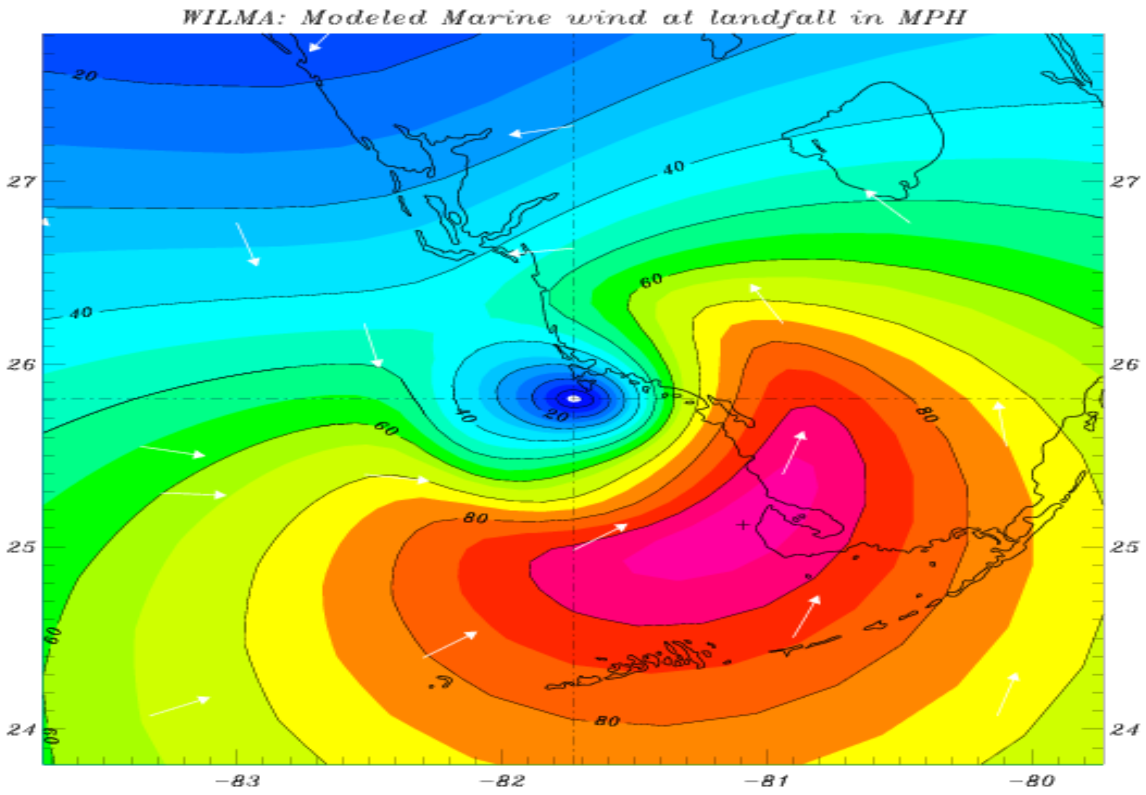
FLORIDA
PUBLIC HURRICANE LOSS MODEL ~~6.0~~6.1

Submitted in compliance with the 2013 Standards of the
Florida Commission on Hurricane Loss Projection Methodology
Submitted on ~~November 1, 2014~~April 8, 2015



Model Identification

Name of Model: Florida Public Hurricane Loss Model ~~6.0~~

Model Version Identification: V6.1

Software Program Version Identification: V6.1

Interim Software Program Version Update Identification:

Software Platform Name and Identifications:

Interim Data Update Designation:

Name of Modeling Organization: Florida International University

Street Address: International Hurricane Research Center, AHC 5

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E-mail Address: hamids@fiu.edu

Date: ~~November 1, 2014~~ April 8, 2015

~~October 29, 2014~~ April 8, 2015

Chair, Florida Commission on Hurricane Loss Projection Methodology
c/o Donna Sirmons
Florida State Board of Administration
1801 Hermitage Boulevard, Suite 100
Tallahassee, FL 32308

Dear Commission Chairman:

I am pleased to inform you that the ~~final~~ revised version of ~~6.0~~6.1 of Florida Public Hurricane Loss Model is ready for review by the Commission. The FPHLM model has been reviewed by professionals having credentials and/or experience in the areas of meteorology, engineering, actuarial science, statistics and computer science; for compliance with the Standards, as documented by the expert certification forms G1-G7.

Enclosed are 7 bound copies of our submission, which includes the summary statement of compliance with the standards, the forms, and the submission checklist.

Please contact me if you have any questions regarding this submission.

Sincerely,



Shahid Hamid, Ph.D., CFA
Professor of Finance, and
Director, Laboratory for Insurance, Economic and Financial Research
International Hurricane Research Center
RB 202B, Department of Finance, College of Business
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Cc: Kevin M. McCarty, Insurance Commissioner

Statement of Compliance and Trade Secret Disclosure Items

The Florida Public Hurricane Loss Model ~~6.0~~6.1 is intended to comply with each Standard of the 2013 Report of Activities released by the Florida Commission on Hurricane Loss Projection Methodology. The required disclosures, forms, and analysis are contained herein.

The source code for the loss model will be available for review by the Professional Team.

Model Submission Checklist

1. Please indicate by checking below that the following has been included in your submission documentation to the Florida Commission on Hurricane Loss Projection Methodology.

Yes	No	Item
X		1. Letter to the Commission
X		a. Refers to the certification forms and states that professionals having credentials and/or experience in the areas of meteorology, statistics, engineering, actuarial science, and computer science have reviewed the model for compliance with the standards
X		b. States model is ready to be reviewed by the Professional Team
X		c. Any caveats to the above statements noted with a complete explanation
X		2. Summary statement of compliance with each individual standard and the data and analyses required in the disclosures and forms
X		3. General description of any trade secret information the modeling organization intends to present to the Professional Team
X		4. Model Identification
X		5. Seven (7) Bound Copies (duplexed)
X		6. Link containing:
X		a. Submission text in PDF format
X		b. PDF file supports highlighting and hyperlinking, and is bookmarked by standard, form, and section
X		c. Data file names include abbreviated name of modeling organization, standards year, and form name (when applicable)
X		d. Form S-6 (Hypothetical Events for Sensitivity and Uncertainty Analysis), if required, in ASCII and PDF format
X		e. Forms M-1 (Annual Occurrence Rates), M-3 (Radius of Maximum Winds and Radii Of Standard Wind Thresholds), V-2 (Mitigation Measures – Range of Changes in Damage), A-1 (Zero Deductible Personal Residential Loss Costs by ZIP Code), A-2 (Base Hurricane Storm Set Statewide Losses), A-3A (2004 Hurricane Season Losses, 2007 FHCF Exposure Data), A-3B (2004 Hurricane Season Losses, 2012 FHCF Exposure Data), A-4A (Output Ranges, 2007 FHCF Exposure Data), A-4B (Output Ranges, 2012 FHCF Exposure Data), A-5 (Percentage Change in Output Ranges, 2007 FHCF Exposure Data), A-7 (Percentage Change in Logical Relationship to Risk), and A-8 (Probable Maximum Loss for Florida) in Excel format
X		7. All hyperlinks to the locations of forms are functional
X		8. Table of Contents
X		9. Materials consecutively numbered from beginning to end starting with the first page (including cover) using a single numbering system, including date and time
X		10. All tables, graphs, and other non-text items consecutively numbered using whole numbers

X		11. All tables, graphs, and other non-text items specifically listed in Table of Contents
X		12. All tables, graphs, and other non-text items clearly labeled with abbreviations defined
X		13. All column headings shown and repeated at the top of every subsequent page for forms and tables
X		14. Standards, disclosures, and forms in *italics*, modeling organization responses in non-italics
X		15. All graphs and maps conform to guidelines in **II. Notification Requirements A.5e.**
X		16. All units of measurement clearly identified with appropriate units used
X		17. All forms included in submission document as Appendices except Forms V-3 (Mitigation Measures – Mean Damage Ratios and Loss Costs, Trade Secret item), A-6 (Logical Relationship to Risk, Trade Secret item), and S-6 (Hypothetical Events for Sensitivity and Uncertainty Analysis)
X		18. Hard copy documentation identical to electronic version
X		19. Signed Expert Certification Forms G-1 to G-7

2. Explanation of "No" responses indicated above. (Attach additional pages if needed.)

Model Name and Identification	Modeler Signature	Date
Florida Public Hurricane Loss Model ~~6.0~~6.1		~~October 29, 2014~~ April 8, 2015

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GENERAL STANDARDS

G-1 Scope of the Computer Model and Its Implementation

A. ***The computer model shall project loss costs and probable maximum loss levels for residential property insured damage from hurricane events.***

The Florida Public Hurricane Loss Model estimates loss costs and probable maximum loss levels from hurricane events for personal lines and commercial lines of residential property. The losses are estimated for building, appurtenant structure, contents, and additional living expense (ALE).

B. ***The modeling organization shall maintain a documented process to assure continual agreement and correct correspondence of databases, data files, and computer source code to slides, technical papers, and modeling organization documents.***

The FPHLM group members follow the process specified in the flowchart of Figure 1 in order to assure continual agreement and correct correspondence of databases, data files, and computer source code to slides, technical papers, and FPHLM documents.

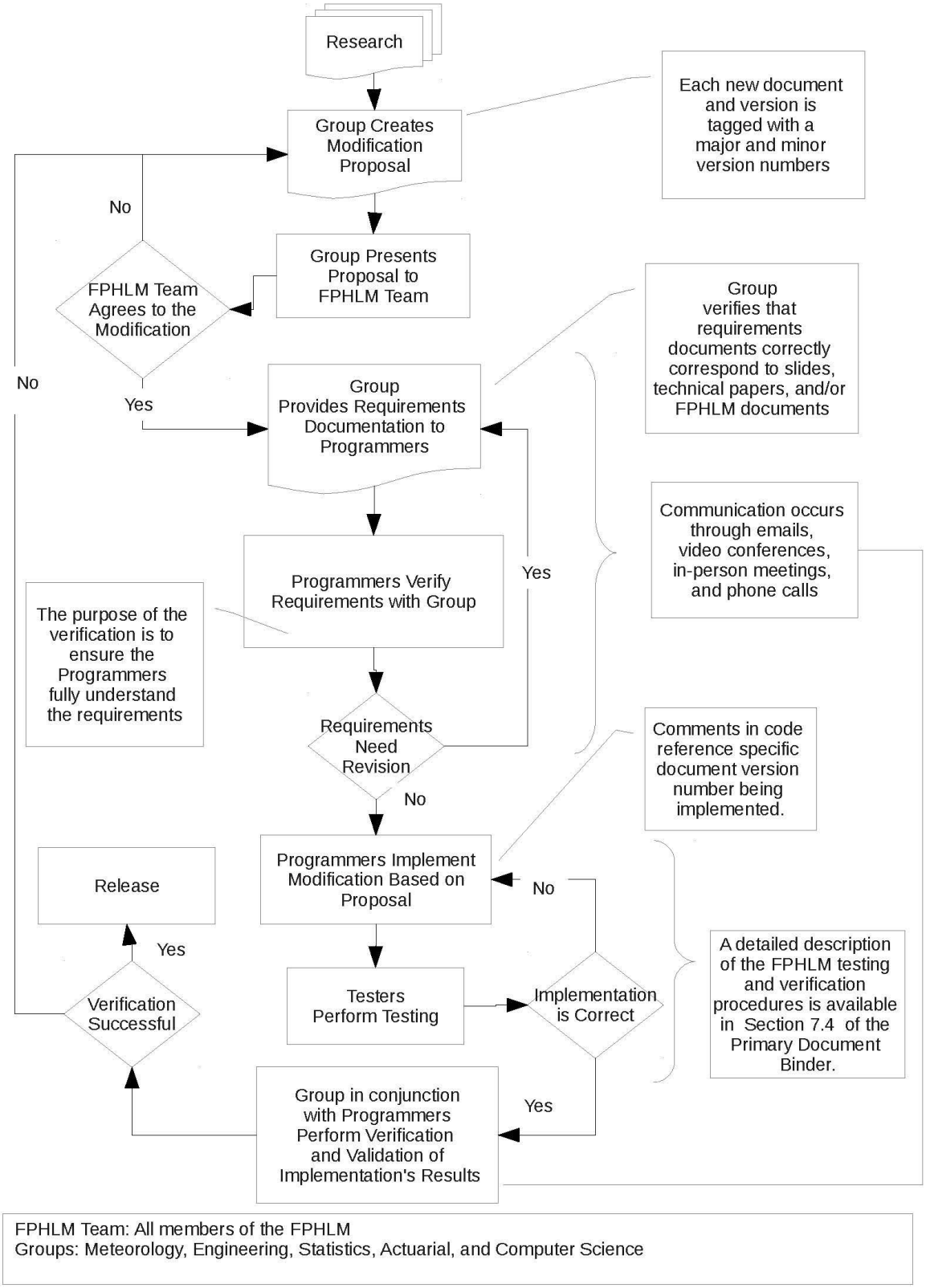


Figure 1. Process to assure continual agreement and correct correspondence.

Disclosures

1. ***Specify the model version identification and software program version identification. If the model submitted for review is implemented on more than one software platform, specify each model software platform. Specify which software platform is the primary software platform and verify how any other software platforms produce the same model output results or are otherwise functionally equivalent as provided for in the "Process for Determining the Acceptability of a Computer Simulation Model" in VI. Review by the Commission, I. Review and Acceptance Criteria for Functionally Equivalent Model Software Platforms.***

The model name is Florida Public Hurricane Loss Model (FPHLM). The current version identification is V6.0.

2. ***Provide a comprehensive summary of the model. This summary shall include a technical description of the model including each major component of the model used to produce residential loss costs and probable maximum loss levels in the State of Florida. Describe the theoretical basis of the model and include a description of the methodology, particularly the wind components, the vulnerability components, and the insured loss components used in the model. The description shall be complete and not reference unpublished work.***

The model is a very complex set of computer programs. The programs simulate probable future hurricane activity, including where and when hurricanes form, their tracks and intensities, their wind fields and sizes; how they decay and how they are affected by the terrain along the tracks after landfall; how the winds interact with different types of residential structures; how much they can damage roofs, windows, doors, interior, and contents, etc.; how much it will cost to rebuild the damaged parts; and how much of the loss will be paid by insurers. The model consists of three major components: wind hazard (meteorology), vulnerability (engineering), and insured loss cost (actuarial). It has over a dozen subcomponents. The major components are developed independently before being integrated. The computer platform is designed to accommodate future subcomponents or enhancements. Following is the description of each of the major components and the computer platform.

METEOROLOGY COMPONENT

Hurricane Track and Intensity

The storm track model generates storm tracks and intensities on the basis of historical storm conditions and motions. The initial seeds for the storms are derived from the HURDAT database. For historical landfalling storms in Florida and neighboring states, the initial positions, intensities, and motions are taken from the track fix 36 hours prior to first landfall. For historical storms that do not make landfall but come within 62 sm (100 km) of the coast, the initial conditions are taken from the track fix 36 hours prior to the point at which the storm first comes within 62 sm of the coast (threat zone) and has a central pressure below 1005 mb. Small, uniform random error terms are added to the initial position, the storm motion change, and the storm intensity change. The initial conditions derived from HURDAT are recycled as necessary to generate thousands of years of stochastic tracks. After the storm is initiated, the subsequent motion and intensity changes are

sampled from empirically derived probability distribution functions over the model domain (Figure 2).

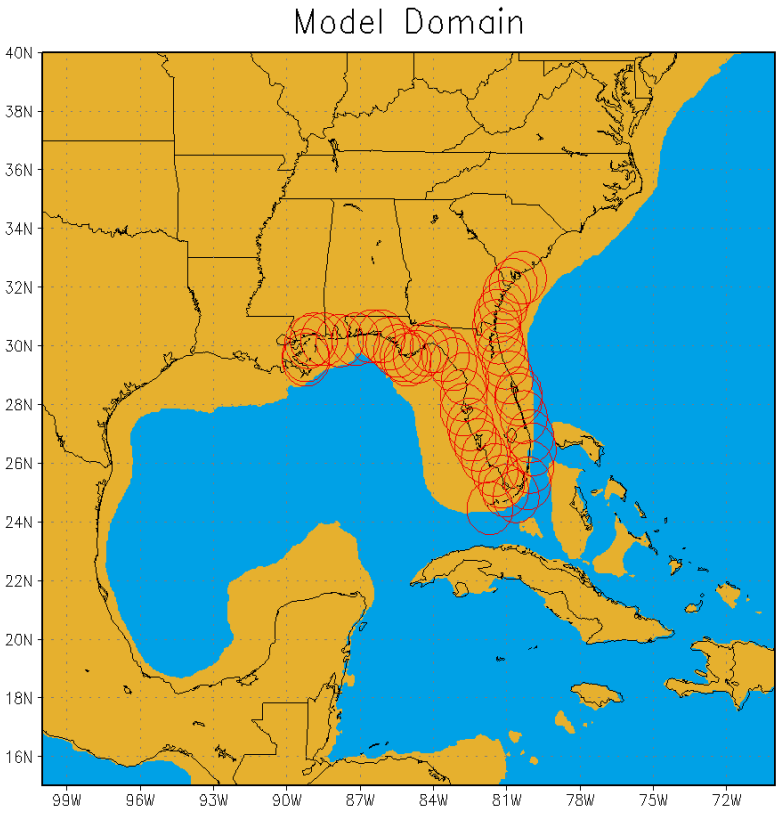


Figure 2. Florida Public Hurricane Loss Model domain. Circles represent the threat zone. Blue color indicates water depth exceeding 656 ft (200 m).

The time evolution of the stochastic storm tracks and intensity are governed by the following equations

$$\Delta x = c\cos(\theta)\Delta t / \cos(y)$$
$$\Delta y = c\sin(\theta)\Delta t$$
$$\Delta p = w\Delta t$$

where (*x,y)* are the longitude and latitude of the storm, (c,θ) are the storm speed and heading (in conventional mathematical sense), *p* is central pressure, *w* is the rate of change in *p*, and Δt is the time step. The time step of the model is currently one hour. The storm speed and direction $(\delta\theta, \delta c)$ are sampled at every 24-hour interval from a probability distribution function (PDF). The intensity change after the initial 24 hours of track evolution is sampled every six hours to capture the more detailed evolution over the continental shelf (shallow water). From the 24-hour change in speed and heading angle, we determine the speed and heading angle at each one-hour time step by assuming the storm undergoes a constant acceleration that gives the 24-hour sampled change in velocity. For changes in pressure, we first sample from a PDF of relative intensity changes, δr, for the six-hour period and then determine the corresponding rate of pressure change, *w*. The

relative intensity is a function of the climatological sea surface temperatures and the upper tropospheric 100 mb temperatures. The PDFs of the changes $(\delta c, \delta\theta, \delta r)$ depend on spatial location, as well as the current storm motion and intensity. These PDFs are of the form

$$PDF(\delta a) = A(\delta a, a, x, y)$$

where a is either c, θ, or r and are implemented as discrete bins that are represented by multi-dimensional matrices (arrays), $A(l,m,i,j)$. The indices (i,j) are the storm location bins. The model domain (100W to 70W, 15N to 40N) is divided into 0.5-degree boxes. The index m represents the bin interval that a falls into. That is, the range of all possible values of a are divided into discrete bins, the number of which depends on the variable, and the index m represents the particular bin a is in at the current time step. As with a, the range of all possible values of the change in a are also discretely binned. Given a set of indices (m,i,j), which represent the current storm location and state, the quantity $A(l,m,i,j)$ represents the probability that the change in $a, \delta a$, will fall into the l'th bin. When A is randomly sampled, one of the bins represented by the l index, e.g. l', is chosen. The change of a is then assigned the midpoint value of the bin associated with l'. A uniform random error term equal to the width of bin l' is added to δa, so that δa may assume any value within the bin l'.

The PDFs described above were generated by parsing the HURDAT database and computing for each track the storm motion and relative intensity changes at every 24- and 6-hour interval, respectively, and then binning them. Once the counts are tallied, they are then normalized to obtain the distribution function. For intensity reports for which pressure is not available, a wind pressure relation developed by Landsea et al. (2004) is used. In cases where there is no pressure report for a track fix in the historical data but there are two pressure reports within a 24-hour period that includes the track fix, the pressures are derived by linear interpolation. Otherwise the pressure is derived by using the wind-pressure relation. Extra-tropical systems, lows, waves, and depressions are excluded. Intensity changes over land are also excluded from the PDFs. To ensure a sufficient density of counts to represent the PDFs for each grid box, counts from nearest neighbor boxes, ranging up to 2 to 5 grid units away (both north-south and east-west direction), are aggregated. Thus, the effective size of the boxes may range from 1.5 to 5.5 degrees but are generally a fixed size for a particular variable. The sizes of the bins were determined by finding a compromise between large bin sizes, which ensure a robust number of counts in each bin to define the PDF, and small bin sizes, which can better represent the detail of the distribution of storm motion characteristics. Detailed examinations of the distributions, as well as sensitivity tests, were done. Bin sizes need not be of equal width, and a nonlinear mapping function is used to provide unequal-sized bins. For example, most storm motion tends to be persistent, with small changes in direction and speed. Thus, to capture this detail, the bins are more fine-grained at lower speed and direction changes.

For intensity change PDFs, boxes which are centered over shallow water (defined to be less than 656 ft deep, see Figure 2) are not aggregated with boxes over deeper waters. Deeper waters may have significantly higher ocean heat content, which can lead to more rapid intensification [see, for example, Shay et al. (2000); DeMaria et al. (2005); Wada and Usui (2007)]. The depth that defines deep and shallow waters is not too critical, as the continental shelf drops rather sharply. The 200 m (656 ft) bathymetric contour line appears to distinguish well estimates of regions with high and low tropical cyclone heat potential (see http://www.aoml.noaa.gov/phod/cyclone/data/).

When gridded long-term analyses of tropical cyclone heat potential, or similar characterization of oceanic heat content, become available, we intend to use that data in lieu of bathymetry.

In Figure 3 we show a sample of tracks generated by the stochastic track and intensity model.

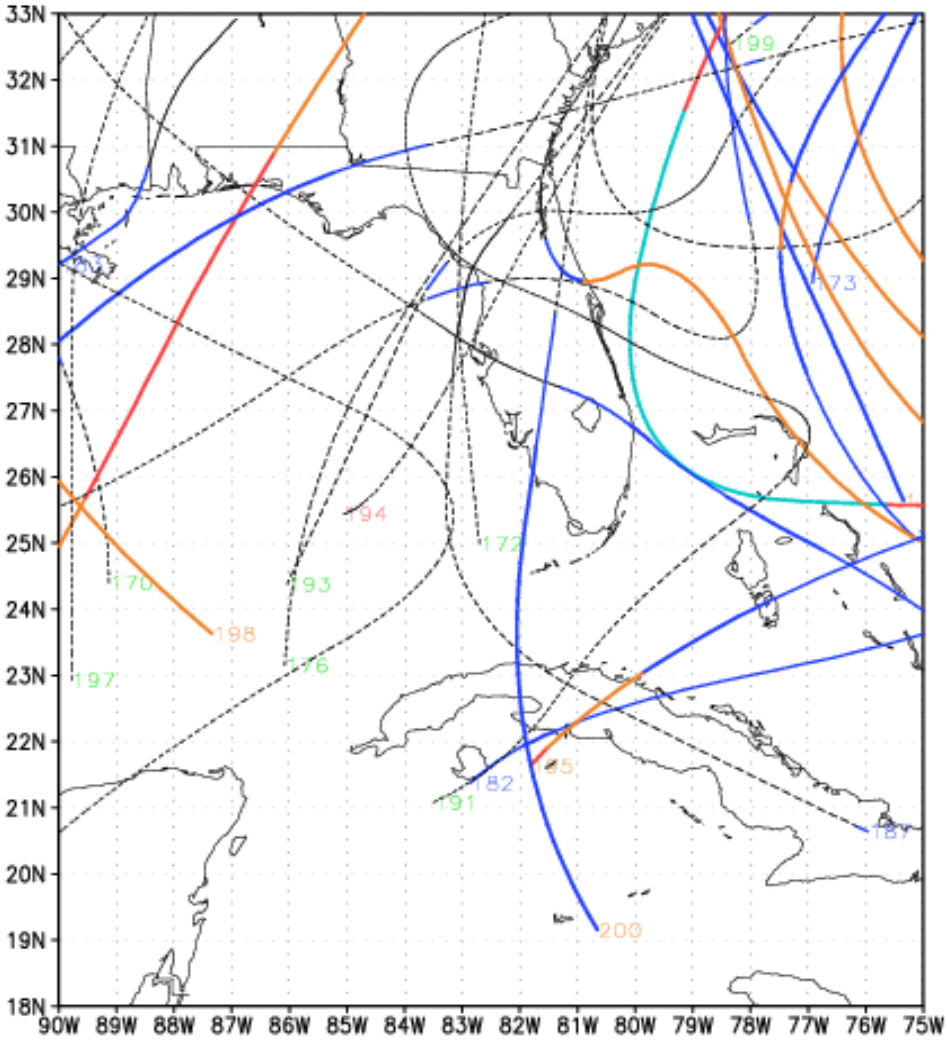


Figure 3. Examples of simulated hurricane tracks. Numbers refer to the stochastic track number, and colors represent storm intensity based on central pressure. Dashed lines represent tropical storm strength winds, and Cat 1-5 winds are represented by black, blue, orange, red, and turquoise, respectively.

When a storm is started, the parameters for radius of maximum winds and *Holland B* are computed and appropriate error terms are added as described below. The *Holland B* term is modeled as follows:

$$B = 1.74425 - 0.007915\,Lat + 0.0000084\,DelP^2 - 0.005024Rmax$$

where *Lat* is the current latitude (degrees) of the storm center, *DelP* is the central pressure difference (mb), and *Rmax* is the radius of maximum winds (km). The random error term for the

Holland B is modeled using a Gaussian distribution with a standard deviation of 0.286. Figure 4 shows a comparison between the Willoughby and Rahn (2004) B dataset (see Standard M-2.1) and the modeled results (scaled to equal the 116 measured occurrences in the observed dataset). The modeled results with the error term have a mean of about 1.38 and are consistent with the observed results. The figure indicates excellent agreement between model and observations.

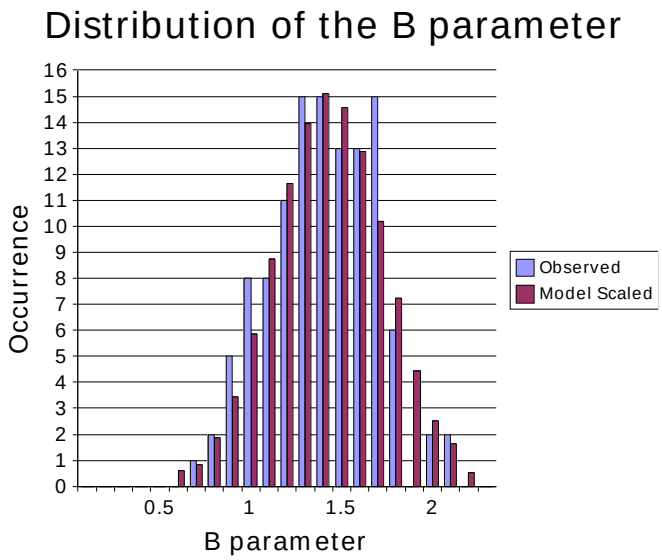


Figure 4. Comparison between the modeled and observed Willoughby and Rahn (2004) *B* dataset.

We developed an *Rmax* model using a landfall *Rmax* database, which includes more than 100 measurements for storms up to 2012. We have opted to model the *Rmax* at landfall rather than the entire basin for a variety of reasons. One is that the distribution of landfall *Rmax* may be different than that over open water. An analysis of the landfall *Rmax* database and the 1988–2007 DeMaria extended best track data shows that there appears to be a difference in the dependence of *Rmax* on central pressure (*Pmin*) between the two datasets (Demuth et al., 2006). The landfall dataset provides a larger set of independent measurements, more than 100 storms compared to about 31 storms affecting the Florida threat area region in the best track data. Since landfall *Rmax* is most relevant for loss cost estimation and has a larger independent sample size, we have chosen to model the landfall dataset. Future studies will examine how the extended best track data can be used to supplement the landfall dataset.

We modeled the distribution of *Rmax* using a gamma distribution. Using the maximum likelihood estimation method, we found the estimated parameters for the gamma distribution, $\hat{k} = 4.76$ and $\hat{\theta} = 5.41$. With these estimated values, we show a plot of the observed and expected distribution in Figure 5. The *Rmax* values are binned in 5 sm intervals, with the *x*-axis showing the end value of the interval.

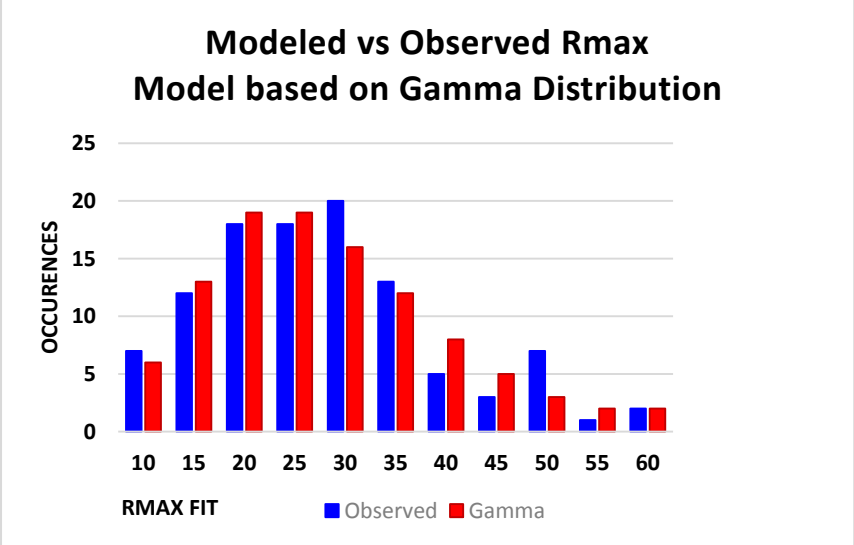


Figure 5. Observed and expected distribution for *Rmax*. The x-axis is the radius in statute miles, and the y-axis is the frequency of occurrence.

An examination of the *Rmax* database shows that intense storms, essentially Category 5 storms, have rather small radii. Thermodynamic considerations (Willoughby, 1998) also suggest that smaller radii are more likely for these storms. Thus, we model Category 5 (*DelP*>90 mb, where *DelP*=1013-*Pmin* and *Pmin* is the central pressure of the storm) storms using a gamma distribution, but with a smaller value of the θ parameter, which yields a smaller mean *Rmax* as well as smaller variance. We have found that for Category 1–4 (*DelP*<80) storms there is essentially no discernable dependence of *Rmax* on central pressure. This is further verified by looking at the mean and variance of *Rmax* in each 10 mb interval. Thus, we model Category 1–4 storms with a single set of parameters. For a gamma distribution, the mean is given by $k\theta$, and variance is $k\theta^2$. For Category 5 storms, we adjust θ such that the mean is equal to the mean of the three Category 5 storms in the database: 1935 No Name, 1969 Camille, and 1992 Andrew. An intermediate zone between *DelP*=80 mb and *DelP*=90 mb is established where the mean of the distribution is linearly interpolated between the Category 1–4 value and the Category 5 value. As the θ value is reduced, the variance is likewise reduced. Since there are insufficient observations to determine what the variance should be for Category 5 storms, we rely on the assumption that variance is appropriately described by the rescaled θ, via $k\theta^2$.

A simple method is used to generate the gamma-distributed values. A uniformly distributed variable, a product of the random number generator that is intrinsic to the FORTRAN compiler, is mapped onto the range of *Rmax* values via the inverse cumulative gamma distribution function. For computational efficiency, a lookup table is used for the inverse cumulative gamma distribution function, with interpolation between table values. Figure 6 shows a test using 100,000 samples of *Rmax* for Category 1–4 storms, binned in 1 sm intervals and compared with the expected values.

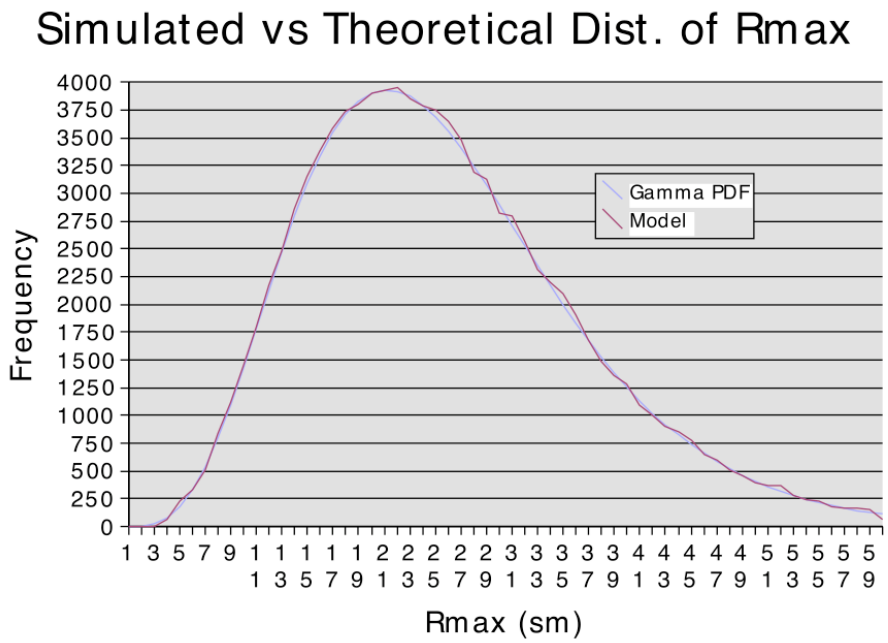


Figure 6. Comparison of 100,000 *Rmax* values sampled from the gamma distribution for Category 1-4 storms to the expected values.

For Category 5 and intermediate Category 4–5 storms, we use the property that the gamma cumulative distribution function is a function of (k,x/θ). Thus, by rescaling θ, we can use the same function (lookup table), but just rescale x (*Rmax*). The rescaled *Rmax* will still have a gamma distribution but with different mean and variance.

The storms in the stochastic model will undergo central pressure changes during the storm life cycle. When a storm is generated, an appropriate *Rmax* is sampled for the storm. To ensure the appropriate mean values of *Rmax* as pressure changes, the *Rmax* is rescaled every time step as necessary. As long as the storm has *DelP* < 80 mb, there is in effect no rescaling. In the stochastic storm generator, we limit the range of *Rmax* from 4 sm to 120 sm.

Storm landfall and decay over land are determined by comparing the storm location *(x,y)* with a 0.6 sm resolution land-sea mask. This land mask is obtained from the U.S. Geological Survey (USGS) land use cover data, and inland bodies of water have been reclassified as land to avoid spurious landfalls. Landfall occurs every time the storm moves from an ocean point to a land point as determined by this land mask. During landfall, the central pressure is modeled by a filling model described in Vickery (2005) and is no longer sampled from the intensity change PDFs. The Vickery (2005) model basically uses an exponentially decaying, in time, function of the central pressure difference with the decay coefficients varying by region on the basis of historical data. The pressure filling model also takes into account the speed and size of the storm. When the storm exits to sea, the land-filling model is turned off and sampling of the intensity change PDFs begins again. A storm is dissipated when its central pressure exceeds 1011 mb.

Wind Field Model

Once a simulated hurricane moves to within a threshold distance of a Florida ZIP Code, the wind field model is turned on. The model is based on the slab boundary layer concept originally conceived by Ooyama (1969) and implemented by Shapiro (1983). Similar models based on this concept have been developed by Thompson and Cardone (1996), Vickery et al. (1995), and Vickery et al. (2000a). The model is initialized by a boundary layer vortex in gradient balance. Gradient balance represents a circular flow caused by balance of forces on the flow whereby the inward directed pressure gradient force is balanced by outward directed Coriolis and centripetal accelerations. The coordinate system translates with the hurricane vortex moving at velocity *c*. The vortex translation is assumed to equal the geostrophic flow associated with the large-scale pressure gradient. In cylindrical coordinates that translate with the moving vortex, equations for a slab hurricane boundary layer under a prescribed pressure gradient are

$$u\frac{\partial u}{\partial r}-\frac{v^2}{r}-fv+\frac{v}{r}\frac{\partial u}{\partial \phi}+\frac{\partial p}{\partial r}-K\left(\nabla^2 u-\frac{u}{r^2}-\frac{2}{r^2}\frac{\partial u}{\partial \phi}\right)+F(c,u)=0=\frac{\partial u}{\partial t}$$

$$u\left(\frac{\partial v}{\partial r}+\frac{v}{r}\right)+fu+\frac{v}{r}\frac{\partial v}{\partial \phi}-\ K\left(\nabla^2 v-\frac{v}{r^2}+\frac{2}{r^2}\frac{\partial u}{\partial \phi}\right)+F(c,v)=0=\frac{\partial v}{\partial t}$$

where *u* and *v* are the respective radial and tangential wind components relative to the moving storm; *p* is the sea level pressure, which varies with radius (*r*); *f* is the Coriolis parameter, which varies with latitude; ϕ is the azimuthal coordinate; *K* is the eddy diffusion coefficient; and *F*(*c*,*u*), *F*(*c*,*v*) are frictional drag terms. All terms are assumed to be representative of means through the boundary layer. The motion of the vortex is determined by the modeled storm track. The symmetric pressure field *p*(*r*) is specified by the Holland (1980) pressure profile with the central pressure specified according to the intensity modeling in concert with the storm track. The model for the *Holland B* pressure profile and the radius of maximum wind are described above. The wind field is solved on a polar grid with a 0.1 *R*/*Rmax* resolution. The input *Rmax* is adjusted to remove a bias caused by a tendency of the wind field solution to place *Rmax* one grid point radially outward from the input value.

The marine surface winds from the slab model are adjusted to land surface winds using a surface friction model. The FPHLM includes the ability to model losses at the "street level." To incorporate this feature, the treatment of land surface friction in the model has been enhanced to provide surface winds at high resolution and to take advantage of recent developments in hurricane boundary layer theory. The 10-minute winds from the slab model are interpolated to a 1 km (0.62 sm) fixed grid covering the entire state of Florida at every time step to obtain a wind swath for each storm. Surface friction is modeled using an effective roughness model (Axe, 2004) based on the Source Area Model of Schmidt and Oke (1990) that takes into account upstream surface roughness elements. The surface roughness elements are derived from the Multi-Resolution Land Characteristics Consortium (MRLC) National Land Classification Database (NLCD) 2011 land cover/land use dataset (Jin et al., 2013) and the Statewide 2004-2011 Florida Water Management District land use classification data (available from the Florida

Department of Environmental Protection). The effective roughness elements are computed for eight incoming wind directions on a grid of approximately 90 m (295 ft) resolution covering the entire state of Florida.

For modeling losses at the ZIP Code level, the effective roughness elements are aggregated over the ZIP Code by a weighted summation of the roughness elements according to population density determined from census block data. The methodology for converting marine winds to actual terrain winds is based on Powell et al. (2003) and Vickery et al. (2009). This method assumes that wind at the top of the marine boundary layer is similar to the wind at the top of the boundary layer over land, and a modified log-wind profile is then used to determine the wind near the land surface. The winds are computed at various height levels that are needed for the vulnerability functions for residential and commercial residential structures.

The effect of the sea-land transition of hurricane winds coming onshore is modeled by modifying the terrain conversion methodology of Vickery et al. (2009). This modification is based on the concept of an internal boundary layer (IBL) (Arya, 1988) that develops as wind transitions from smooth to rough surface conditions. Winds above the IBL are assumed to be in equilibrium with marine roughness. In the equilibrium layer (EL), defined to be one-tenth of the IBL, the winds are assumed to be in equilibrium with the local effective roughness. Between the EL and IBL the winds are assumed to be in equilibrium with vertically varying step-wise changes in roughness associated with upstream surface conditions. This concept of multiple equilibrium layers is similar in philosophy to the method prescribed by the Engineering Sciences Data Unit (ESDU). The coastal transition function produces wind transitions that are very close to the ESDU and modified ESDU values reported in Vickery et al. (2009).

VULNERABILITY COMPONENT: PERSONAL RESIDENTIAL MODEL

The engineering component performs several tasks: (1) it estimates the physical damage to exterior components of typical buildings, including roof cover, roof decking, walls, and openings; (2) it assesses the interior and utilities damage and contents damage due to water penetration through exterior damage and defects to interior walls, ceiling, doors, etc.; (3) it combines the exterior and interior damage to estimate the building and content vulnerabilities; (4) it estimates additional living expenses; and (5) it estimates the appurtenant structure vulnerability (Pinelli et al., 2003a, 2003b, 2004a, 2004b, 2005a, 2005b, 2006, 2007a, 2007b, 2008a, 2008b, 2009a, 2010a, 2011a, 2011b, 2012; Cope, 2004; Cope et al., 2003a, 2003b, 2004b, 2005; Gurley et al., 2003, Torkian at al., 2011, 2014).

Exposure Study

Personal residential single-family home buildings (PRB), either site built (Figure 7) or manufactured (Figure 8), are categorized into typical generic groups with similar structural characteristics, layout, and materials within each group. These buildings can suffer substantial external structural damage (in addition to envelope and interior damage), including collapse under hurricane winds. The approach to assessing damage for each of these building types is to model the building as a whole so that interactions among components can be accounted for. The models are intended to represent the majority of the PRB's in Florida.

An extensive survey of the Florida building stock was carried out to develop a manageable number of building models that represent the majority of the Florida residential building stock. The modelers analyzed several sources of data for building stock information. One source was the Florida Hurricane Catastrophe Fund (FHCF) exposure database. Another source was the Florida counties' property tax appraisers' databases. Although the database contents and format vary county to county, many of these databases contain the structural information needed to define common structural types. The 52 most populous counties were contacted to acquire their tax appraiser database, producing information from 33 counties. These 33 counties account for more than 90% of Florida's population. The residential buildings in each county database were divided into single-family residential buildings and mobile homes.

County property tax appraiser (CPTA) databases contain large quantities of building information, and it was necessary to extract those characteristics related to the vulnerability of buildings to wind. The available building characteristics vary from county to county and include some combination of the following: exterior wall material, interior wall material, roof shape, roof cover, floor covering, foundation, opening protection, year built, number of stories, area per floor, area per unit, and geometry of the building. The parameters important for modeling are roof cover, roof shape, exterior wall material, number of stories, year built, and building area. For each of these categories, the authors extracted statistical information. The dependency between critical building characteristics was also investigated. For example, it was found that roof shape and area of the building are strongly dependent on the year built. The survey statistics were calculated for different eras to account for the correlation between various factors and year built.



Figure 7. Typical single-family homes (Google Earth).

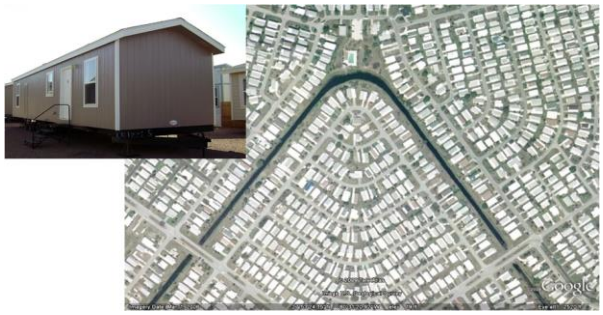


Figure 8. Manufactured homes (Google Earth).

The modelers divided Florida into four regions: North, Central, South, and the Keys. Geography and the statistics from the Florida Hurricane Catastrophe Fund (FHCF) provided guidance for defining regions that would have a similar building mix. For example, North Florida has primarily wood frame houses while South Florida primarily has masonry houses. Figure 9 shows the regions. Each county for which data were available is marked with a star and shaded.

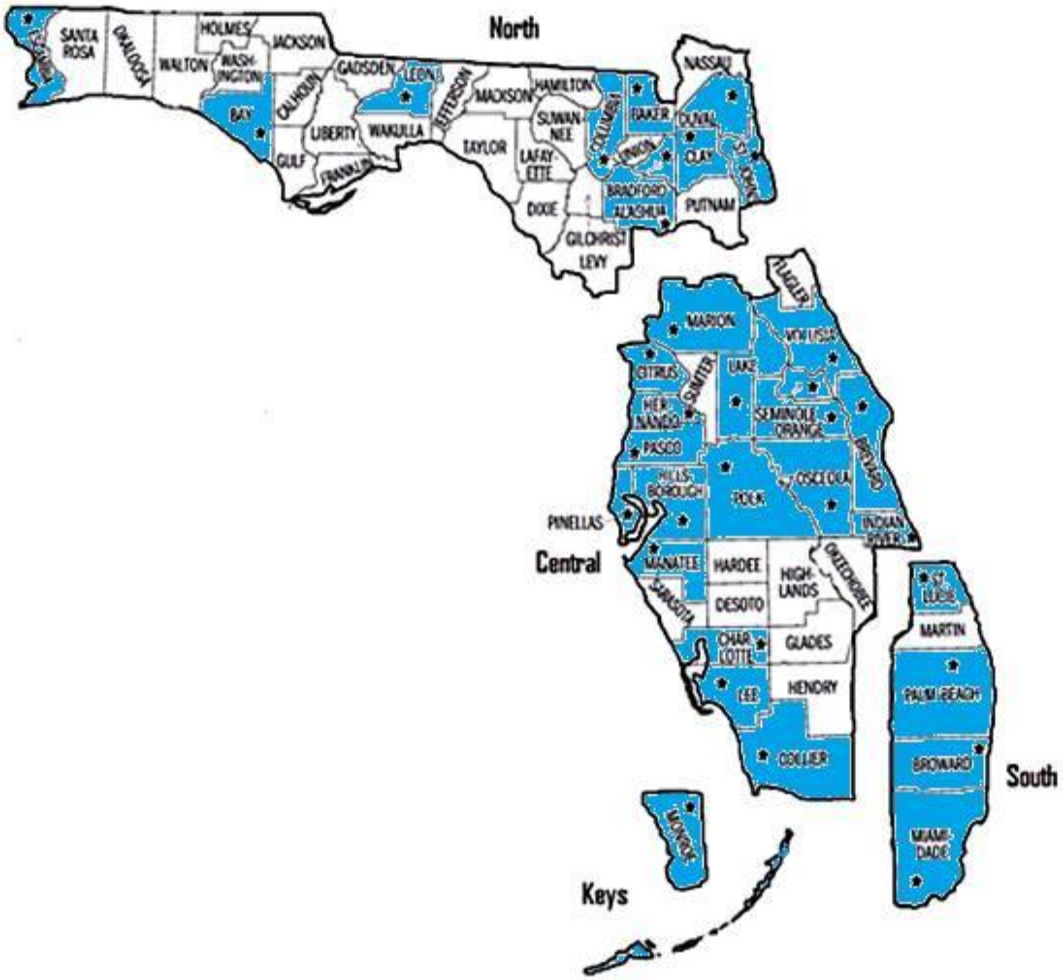


Figure 9. Regional Classification of Florida with the corresponding sample counties (blue and star).

Structural types are delineated by a combination of four characteristics: number of stories (either one or two), roof cover (either shingle, tile, or metal), roof shape (either gable or hip), and exterior wall material (either concrete blocks or timber). Statistics were computed for each structural type in every sampled county. Weighted average techniques were used to extrapolate the results to the remaining counties in each region.

Building Models

Site-Built Home Models

In addition to a classification of building by structural types (wood or masonry walls, hip or gable roof), it was also necessary to classify the buildings by relative strength to reflect changes in construction practice over many years. The vulnerability team has developed strong, medium, and

weak strength models for each site-built structural type to represent relative quality of original construction as well as post-construction mitigation. The weak and medium models have additional variants that reflect historical building practices, roof retrofits, and reroofing of existing structures as mandated by the newer building standards. The strong model has two variants to delineate code requirements that are regionally dependent. One strong variant reflects inland and wind-borne debris region (WBDR) construction, and another (stronger) variant reflects construction in the high velocity hurricane zone (HVHZ).

Both the WBDR and the HVHZ are defined in the Florida Building Code (FBC, 2010):

- WIND-BORNE DEBRIS REGION: Areas within hurricane-prone regions located:
  - 1. Within 1 mile (1.61 km) of the coastal mean high water line where the ultimate design wind speed Vult is 130 mph (58 m/s) or greater; or
  - 2. In areas where the ultimate design wind speed Vult is 140 mph (63 m/s) or greater.
- HIGH VELOCITY HURRICANE ZONE: Broward and Miami-Dade counties

Since the definition of WBDR is linked to the most current wind map in the FBC, its boundaries are not static, and can evolve with changes in the wind speed maps adopted by the FBC. In particular, it was revised in the 2010 edition of the FBC, effective March 2012. The FPHLM has implemented both the pre-2010, and the post 2010 boundaries of the WBDR. Consequently, a building might be assigned to a different WBDR depending on its year built (pre or post 2012).

The three strength categories are based on the same model framework, in which strength is represented by the capacities assigned to the modeled building components. For example, the strong models differ from the weak models by stronger assigned capacities for roof-to-wall (r2w) and stud to sill connections, garage pressure capacity, cracking capacity of masonry walls, gable end walls, decking and shingle capacities. The medium models differ from the weak models by increasing the strength of the roof-to-wall connections (toe nails vs. clips), roof decking capacity (nailing schedule), and masonry wall strength (un-reinforced vs. reinforced).

Any given strong, medium, or weak model may be altered by additional mitigation or retrofit measures individually or in combination. For example, from the base weak model, additional models were derived to represent historical building practices and mitigation techniques. The modified weak W10 model accounts for the use of tongue-and-groove plank decking in pre-1960s buildings. These buildings tend to exhibit higher deck strength capacities than the buildings with the plywood decking implemented in the base weak model, referred to as W00 (Shanmugam et al., 2009).

A modified medium model M10 was adopted that reflects the use of oriented strand board (OSB) decking with staples in the 1980s and pre-Andrew 1990s. This was considered an adequate alternative to nailed plywood at the time. It was, however, weaker in terms of wind resistance and was assigned a weaker deck attachment capacity than the standard medium model.

Additionally, retrofitted weak W01 and medium M01 models were derived from the base weak and medium models. They represent the case in which a structure has been reroofed and the decking re-nailed according to current code requirements. On the basis of the average lifespan of a roof, reroofing would be required periodically throughout the structure's lifetime and would result in an increase in the deck attachment capacity and shingle ratings to meet current building code requirements. The deck attachment capacities of these models were therefore upgraded to produce the retrofitted weak W01 and medium M01 cases. The roof cover was also upgraded to rated shingles (Pinelli et al., 2012).

The base, retrofitted and modified versions of the weak and medium models were developed in order to provide a fine model resolution of quality of construction for homes constructed prior to 1994 and a portion of the homes prior to 2002. Weak and medium models represent approximately 80% of the existing single-family residential inventory in Florida, and are described in Table 1~~Table 1a.~~.

Two basic variations of the strong model represent construction quality for the remaining approximately 20% of the single-family residential inventory. The base strong model, S00, represents modern construction in locations inland, as well as the WBDR that is not overlapping the HVHZ. The difference in strong models between inland, S00, and WBDR, S00-OP, is due to the presence of metal shutters in WBDR. This base strong model incorporates modern requirements for nailing schedules, roof to wall connection products, masonry reinforcing, and roof shingle products and installation methods. The second strong model, S01, has upgrades to the capacity for roof cover, roof decking and roof to wall connections to reflect additional code requirements for HVHZ construction. The strong models are described in Table 2.

All models may be run without opening protection, with plywood opening protection, or with metal panel shutter opening protection installed, with increasing protection respectively.

The distribution of the weak, medium and strong model variations with respect to year built will be presented later in Table 7 and in the discussion of the models' distribution in time.

Table 1~~a.~~. Weak and Medium Models

	Weak			Medium		
	W00 (base)	W01 (retrofitted*)	W10 (modified**)	M00 (base)	M01 (retrofitted*)	M10 (modified***)
Roof to wall	Weak	Weak	Weak	Medium	Medium	Medium
Stud to sill	Weak	Weak	Weak	Medium	Medium	Medium
Roof cover	Weak	Strong	Weak	Weak	Strong	Weak
Roof deck	Weak	Strong	Strong	Medium	Strong	Weak
Wall	Weak	Weak	Weak	Medium	Medium	Medium
Gable end	Weak	Weak	Weak	Weak	Weak	Weak
Garage	Weak	Weak	Weak	Weak	Weak	Weak

*retrofitted refers to re-roof and re-nailed decking, occurring post-1993 for HVHZ and Monroe, and post-2001 for everywhere else. No other retrofits are included.
**modified weak refers to the base weak model with stronger decking to reflect the use of plank decking
***modified medium refers to the base medium model with weak decking to reflect the use of staples and/or OSB

Table ~~1a~~2. Strong Models

	S00 Strong - inland	S00-OP Strong - WBDR	S01 Strong - HVHZ
Roof to wall	Strong	Strong	Upgraded Strong
Stud to sill	Strong	Strong	Strong
Roof cover	Strong	Strong	Upgraded Strong
Roof deck	Strong	Strong	Upgraded Strong
Wall	Strong	Strong	Strong
Gable end	Strong	Strong	Strong
Garage	Strong	Strong	Strong
Shutters	no shutters	metal	metal

Manufactured Homes Model

On the basis of the exposure study, it was decided to model four manufactured home (MH) types: (1) pre-1994—fully tied down, (2) pre-1994—not tied down, (3) post-1994—Housing and Urban Development (HUD) Zone II, and (4) post-1994—HUD Zone III. The partially tied-down homes are assumed to have a vulnerability that is an average of the vulnerabilities of fully tied-down and not tied-down homes. Because little information is available regarding the distribution of manufactured home types by size or geometry, it is assumed that all model types are single-wide manufactured homes. The modeled single-wide manufactured homes are 56 ft x 13 ft, have gable roofs, eight windows, a front entrance door, and a sliding-glass back door.

Damage Matrices

Exterior Damage

The model accounts for a number of construction factors that influence the vulnerability of single-family dwellings, including classification (site-built or manufactured home), size, roof shape, location, age, and a variety of construction details and mitigation measures. The effects of mitigation measures such as code revisions and post-construction upgrades to the wind resistance of homes (e.g., new roof cover on an older home, shutter protection against debris impact, braced garage door, re-nailed roof decking, etc.) are accounted for both individually and in combination by selecting the desired statistical descriptors of the capacities of the various components. Thus the comparative vulnerability of older homes as built, older homes with combinations of mitigation measures, and homes constructed to the new code requirements can be estimated.

The vulnerability model uses a component-based Monte Carlo simulation to determine the external vulnerability at various wind speeds for the different building models. The approach accounts for the resistance capacity of the various building components, the wind-load effects from different directions, and associated uncertainties of capacity and loads to predict exterior damage at various wind speeds. The simulation relates probabilistic strength capacities of building components to a series of three-second peak gust wind speeds through a detailed wind and structural engineering analysis that includes effects of wind-borne debris. Damage to the structure occurs when the loads from wind or flying debris are greater than the components'

capacity to resist them. The vulnerability of a structure at various wind speeds is estimated by quantifying the amount of damage to the modeled components. Damage to a given component may influence the loads on other components, e.g., a change in roof loading from internal pressurization due to a damaged opening. These influences are accounted for through an iterative process of loading, damage assessment, load redistribution, and reloading until convergence is reached. The flow chart in Figure 10 summarizes the Monte Carlo procedure used to predict the external damage. The random variables include wind speed, pressure coefficients, debris impact, and the resistances of the building components (roof cover, roof sheathing, openings, walls, connections).

The damage estimations are affected by uncertainties regarding the behavior and strength of the various components and the load effects produced by hurricane winds. Field and laboratory data that better define these uncertain behaviors can thus be directly included in the model by refining the statistical descriptors of the capacities, load paths, and applied wind loads.

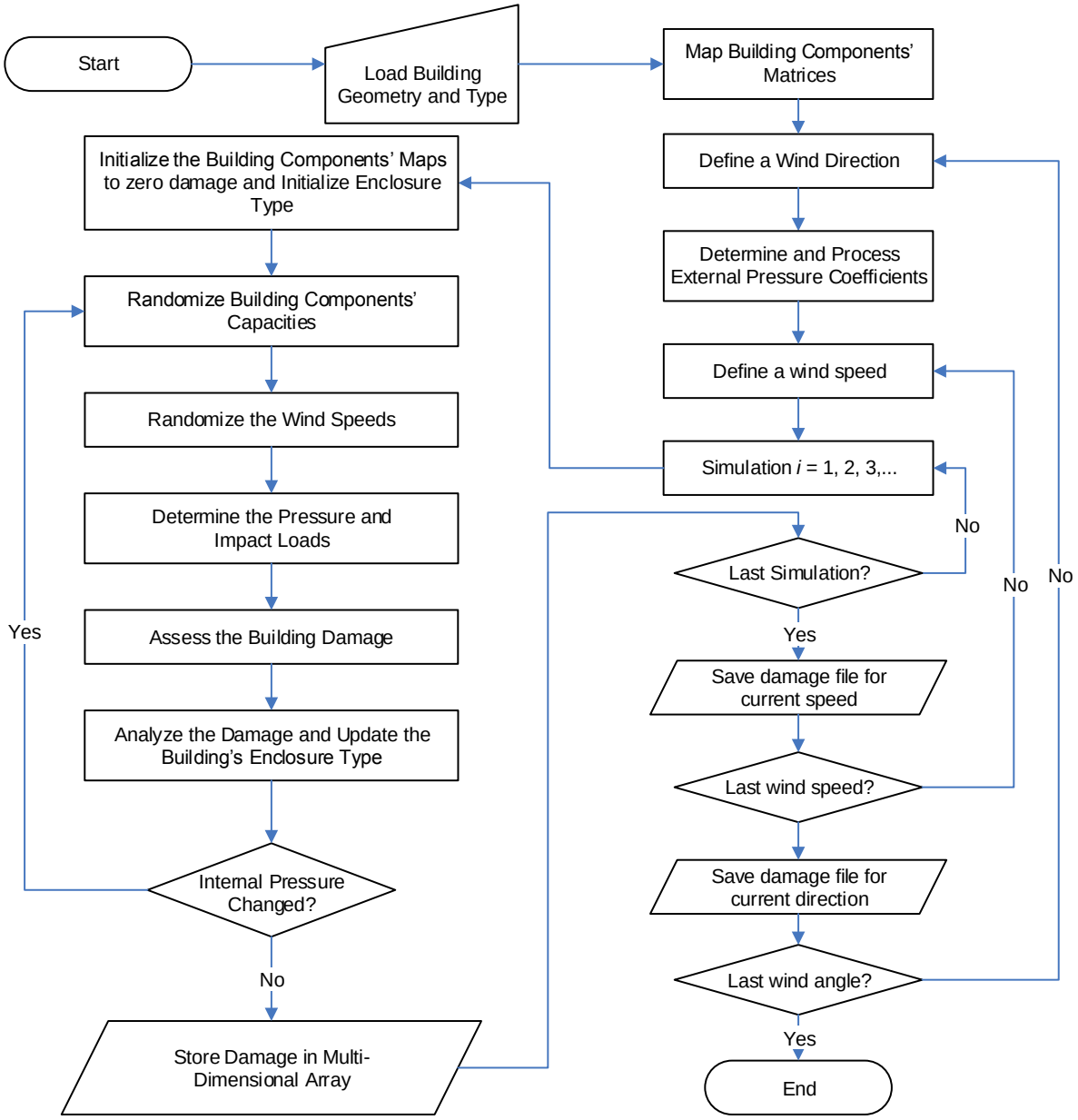


Figure 10. Monte Carlo simulation procedure to predict external damage.

The output of the Monte Carlo simulation model is an estimate of physical damage to structural and exterior components of the modeled home. The results are presented in the form of a damage matrix, where each row presents the output of an individual simulation. The 15 rows of this matrix (Table 3) correspond to damage to 14 components, and the internal pressure of the building upon completion of that simulation (column 11). A separate matrix is created for each peak three-second gust wind speed between 50 and 250 mph in 5 mph increments (50, 55, …, 250 mph) and for each wind angle between 0 and 315 degrees in 45-degree increments. A description of the values in each of the nine columns of the manufactured home damage matrix is given in Table 4. Note that internal pressure is not included as an output from the manufactured home model (Table 4). Changes in internal pressure due to breach are accounted for and utilized to quantify damage, but the final internal pressure value is not needed as an output.

Table 3. Description of values given in the damage matrices for site-built homes.

Col#	Description of Value	Min Value	Max Value
1	% failed roof sheathing	0	100
2	% failed roof cover	0	100
3	% failed roof to wall connections	0	100
4	# of failed walls	0	4
5	# of failed windows	0	15
6	# of failed doors	0	2
7	y or n failed garage	0 = no	1 = yes
8	y or n envelope breached	0 = no	1 = yes
9	# of windows broken by debris impact	0	15
10	% of gable end panels broken	0	100
11	internal pressure	0	Not defined
12	% failed wall panels – front	0	100
13	% failed wall panels – back	0	100
14	% failed wall panels – side	0	100
15	% failed wall panels – side	0	100

Table 4. Description of values given in the damage matrices for manufactured homes.

Col #	Description of Value	Min Value	Max Value
1	# of failed windows (out of 8 for single wide)	0	8
2	# of broken windows that were broken by impact load case	0	8
3	# of failed doors (front and back = 2 total)	0	2
4	% of roof sheathing failed	0	100
5	% of roof cover failed	0	100
6	% of wall sheathing failed	0	100
7	# of failed roof to wall connections (out of 58)	0	58
8	sliding (0 = no sliding, 1 = minor sliding, 2 = major sliding)	0	2
9	overturning (0 = not overturned, 1 = overturned)	0	1

Interior and Utilities Damage

Once the external damage has been calculated for a given Monte Carlo simulation, the internal, utilities, and contents damages to the building are then extrapolated from the external damage. For the interior and utilities of a home, there is no explicit means by which to compute damage. Damage to the interior and utilities occurs when the building envelope is breached, allowing wind and rain to enter. Damage to roof sheathing, roof cover, walls, windows, doors, and gable ends present the greatest opportunities for interior damage. For manufactured homes, sliding and overturning are additional factors.

Interior damage equations were derived as functions of each of the external components. These equations are developed primarily on the basis of experience and engineering judgment. Observations of homes damaged during the 2004 hurricane season helped to validate these predictions. The interior equations are derived by estimating typical percentages of damage to each interior component, given a percentage of damage to an external component. The interior damage as a function of each modeled component is the same for both site-built and manufactured homes.

To model the uncertainties inherent in the determination of interior damage, the output of the equations is multiplied by a random factor with mean unity. The factor is assumed to have a Weibull distribution with tail length parameter 2. For the factor to have mean unity, the scale parameter must be 0.7854, resulting in a variance of 0.2732. This choice of Weibull parameters is assumed to be reasonable, and a sensitivity study was done to confirm that assumption and to show that it has no effect on the mean vulnerability, as expected.

To compute the total interior damage for each model simulation, all values in the damage matrices are converted to percentages of component damage. The interior equations are applied to each component, one at a time. The total interior damage for each simulation is the maximum interior damage value produced by these equations. The maximum value is used instead of a summation to avoid the possibility of counting the same interior damage more than once. That is, once water intrusion from one breach of the envelope has thoroughly damaged any part of the

interior, further water intrusion from other sources will not increase the cost of the damage of that part.

Utilities damage is estimated on the basis of interior damage. A coefficient is defined for each utility (electrical, plumbing, and mechanical), which multiplies the interior equations defined for each component. As in the case of interior damage, the maximum value is retained as the total damage. The utilities coefficients are based on engineering judgment. In both site-built and manufactured homes, it is assumed that electrical damage occurs at half the rate of interior damage (0.5). Plumbing damage is set to 0.35 of interior damage for site-built homes and for manufactured homes. Mechanical damage is set to 0.4 of interior damage for site-built homes and for manufactured homes.

Contents Damage

As with the interior and utilities, the contents of the home are not modeled by Monte Carlo simulations. Contents damage is assumed to be a function of the interior damage caused by each failed component that causes a breach of the building envelope. The functions are based on engineering judgment and are validated using actual claims data.

Additional Living Expenses

Additional Living Expense (ALE) coverage covers only expenses actually paid by the insured. This coverage pays only the increase in living expenses that results directly from the covered damage and having to live away from the insured location. The value of an ALE claim is dependent on the time required to repair a damaged home and the surrounding utilities and infrastructure.

The equations and methods used for manufactured and residential homes are identical. However, it seems logical to reduce the manufactured home ALE predictions because typically a faster repair or replacement time may be expected for these home types. Therefore, an ALE multiplier factor of 0.75 was introduced into the manufactured home model.

Vulnerability Matrices

The estimates of total building damage result in the formulation of vulnerability matrices for each modeled building type. The flowchart in Figure 11 summarizes the procedure used to convert the Monte Carlo simulations of physical external damage into a vulnerability matrix.

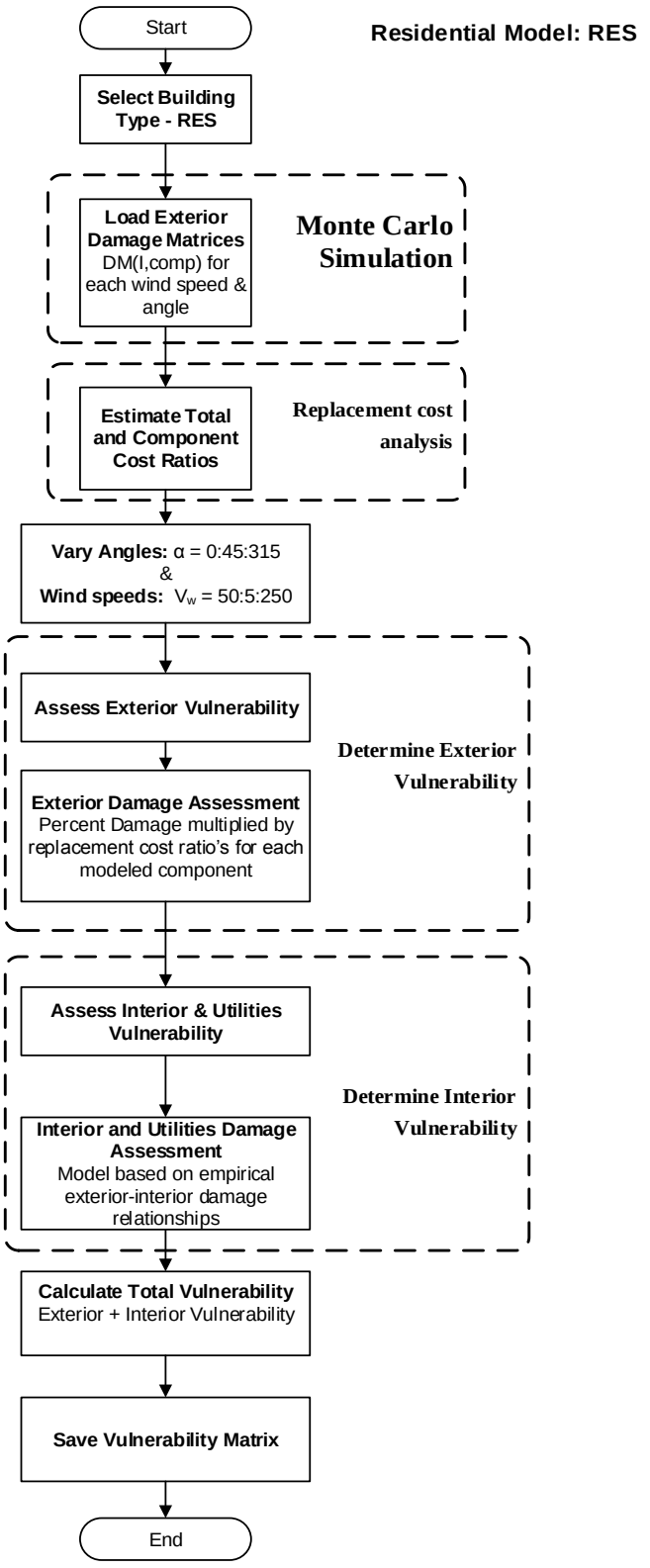


Figure 11. Procedure to create vulnerability matrix.

For each Monte Carlo model, 5000 simulations are performed for each of 8 different wind angles and 41 different wind speeds. This is 5000 x 8 x 41 = 1,640,000 simulations of external damage

per model, which are then expanded to cover interior, utilities, and contents damage, plus ALE, as explained above.

Knowing the components of a home and the typical square footage, the cost of repairing all damaged components is estimated using cost estimation resources [e.g., RSMeans Residential Cost Data (RSMeans, 2008a) and RSMeans Square Foot Costs (RSMeans, 2008b) and Construction Estimating Institute (Langedyk & Ticola, 2002)] and expert advice. These resources provide cost data from actual jobs based on estimates and represent typical conditions. Unmodeled nonstructural interior, plumbing, mechanical, and electrical utilities make up a significant portion of repair costs for a home.

Replacement cost ratios provide a link between modeled physical damage and the corresponding monetary losses. They can be defined as the cost of replacing a damaged component or assembly of a home divided by the cost of constructing a completely new home of the same type. The sum of the replacement cost ratios for all the components of a home is greater than 100% because the replacement costs include the additional costs of removal, repair, and remodeling.

An explicit procedure is used to convert physical damage of the modeled components to monetary damage. Since the replacement ratio of each modeled component is known, the monetary damage resulting from damage to a component expressed as a percentage of the home's value can be obtained by multiplying the damaged percentage of the component by the component's replacement ratio. For example, if 30% of the roof cover is damaged, and for this particular home type the replacement ratio of roof cover is 14%, the value of the home lost as a result of the damaged roof cover would be 0.30 x 0.14 = 4.2%. If the value of this home were $150,000, the cost to replace 30% of the roof would be $150,000 x 0.042 = $6,300. In addition, the costs will be adjusted as necessary because of certain requirements of the Florida building code that might result in an increase of the repair costs (for example, the code might require replacement of the entire roof if 30% or more is damaged).

After the simulation results have been translated into damage ratios, they are then transformed into vulnerability matrices. A total of 4356 matrices for site-built homes is created for different combinations of wall type (frame or masonry), region (North, Central, or South), subregion (high wind velocity zone, wind-borne debris region, or other), roof shape (gable or hip), roof cover (tile or shingle), window protection (shuttered or not shuttered), number of stories (one or two), and strength (base weak W00, modified weak W10, retrofitted weak W01, base medium M00, modified medium M10, retrofitted medium M01, or strong S).

The cells of a vulnerability matrix for a particular structural type represent the probability of a given damage ratio occurring at a given wind speed. The columns of the matrix represent three-second gust wind speeds at 10 m, from 50 mph to 250 mph in 5 mph bands. The rows of the matrix correspond to damage ratios (DR) in 2% increments up to 20%, and then in 4% increments up to 100%. If a damage ratio is DR= 15.3%, it is assigned to the interval 14%<DR<16% with a midpoint DR=15%. After all the simulations have been counted, the total number of instances in each damage interval is divided by the total number of simulations per wind speed to determine the percentage of simulations at any damage state occurring at each speed. These percentages are the conditional probabilities of occurrence of a level of damage, given a certain wind speed. A partial example of a vulnerability matrix is shown in Table 5

Table 5. Partial example of vulnerability matrix.

Damage\Wind Speed (mph)	47.5 to 52.5	52.5 to 57.5	57.5 to 62.5	62.5 to 67.5	67.5 to 72.5
0% to 2%	1	0.99238	0.91788	0.77312	0.61025
2% to 4%	0	0.00725	0.0806	0.21937	0.36138
4% to 6%	0	0.00037	0.001395	0.007135	0.0235
6% to 8%	0	0	0.000125	0.000375	0.0025
8% to 10%	0	0	0	0	0.000375
10% to 12%	0	0	0	0	0.000375
12% to 14%	0	0	0	0	0.000625
14% to 16%	0	0	0	0	0.0005
16% to 18%	0	0	0	0	0.000125
18% to 20%	0	0	0	0	0.00012
20% to 24%	0	0	0	0	0.00025
24% to 28%	0	0	0	0	0

An important plot derived from the vulnerability matrix is the vulnerability curve. The vulnerability curve for any structural type is the plot of the mean damage ratio vs. wind speed. The model can also generate fragility curves (the probability of exceedance of any given damage level as a function of the wind speed) for each vulnerability matrix, although these curves are not used in the model.

Similar vulnerability matrices and vulnerability curves are developed for contents and ALE, one for each structural type. The whole process is also applied to manufactured homes.

Weighted Vulnerability Matrices

Building vulnerability matrices were created for every combination of region (Keys, South, Central, and North), construction type (masonry, wood, or other), roof shape (gable or hip), roof cover (tile or shingle or metal), number of stories (one or two), shutters (with or without), and subregion (inland, wind-borne debris region, or high velocity hurricane zone). However, in general, there is little information available in an insurance portfolio file regarding the structural characteristics and the wind resistance of the insured property. Instead, insurance companies rely on the Insurance Services Office's (ISO) fire resistance classification. Portfolio files have information on ZIP Code and year built. The ISO classification is used to determine if the home is constructed of masonry, timber, or other. The ZIP Code is used to define the region and subregion. The year the home was built is used to assist in defining the strength to be assigned to the home.

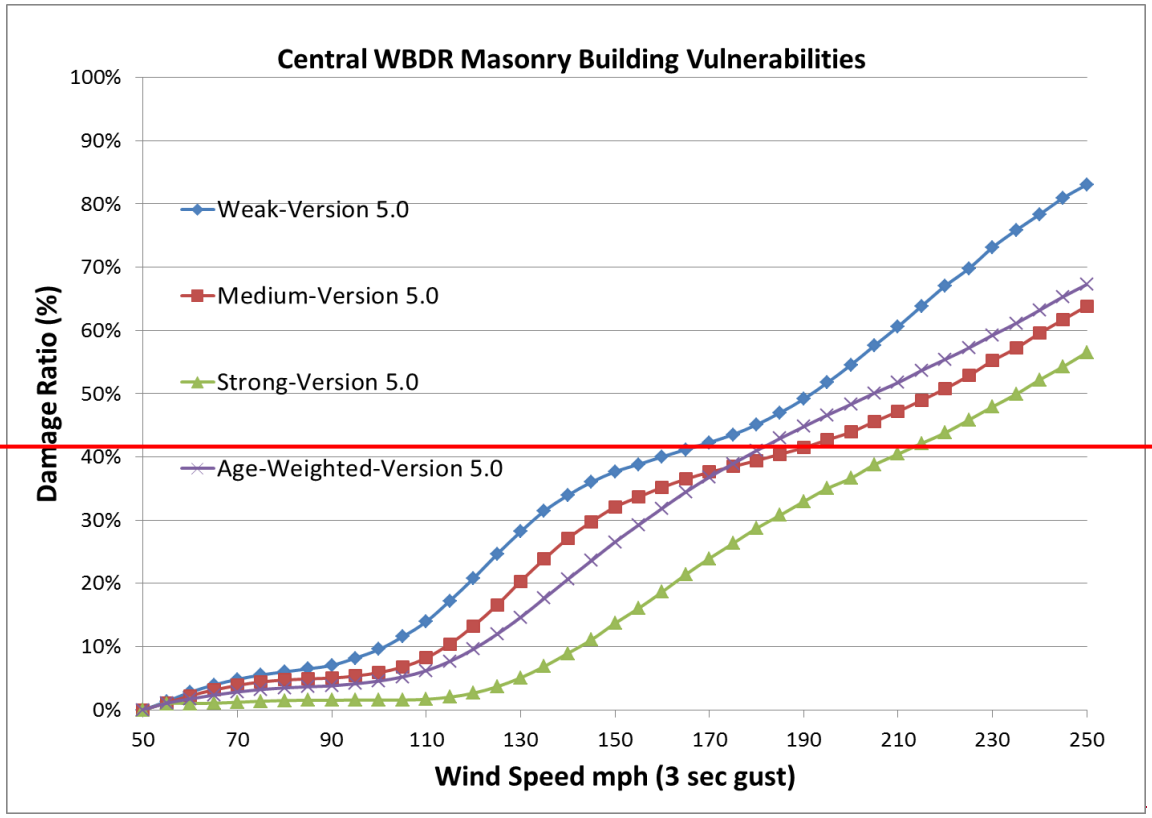
Region, subregion, construction type, and year built are determined from the insurance files. This leaves the roof shape, roof cover, and shutter options undefined. From the exposure study of 33 Florida counties, the distribution of number of stories, roof shapes, and roof cover by age per region can be extrapolated. For each age group, we define a weighted matrix for each construction type in each county belonging to a region and subregion. The weighted matrices are the sum of the corresponding vulnerability model matrices weighted on the basis of their statistical

distribution. For example, consider a masonry home built in the wind-borne debris region of central Florida in 1990. The exposure study indicates that 66% of such homes have gable roofs, 85% have shingle roof cover, and 20% have window shutters. Weight factors can be computed for each model matrix based on these statistics. For example, the Central Florida, gable, tile, no shutters, masonry matrix would have a weight factor of 66% (masonry percent gable) x 15% (percent tile) x 80% (percent without shutters) = 7.9%; this is the percentage of that home type that would be expected in this region, for that year built. Each model matrix is multiplied by its weight factor, and the results are summed. The final result is a weighted matrix that is a combination of all the model matrices and can be applied to an insurance policy if only the ZIP Code, year built, and ISO classification are known. As a result, for each county in each subregion (inland, wind-borne debris region, and high velocity hurricane zone) of each region (Keys, South, Central, and North), there will be sets of weighted matrices (masonry, wood, and others) for weak, medium, and strong structures.

Age-Weighted Matrices

The year built or year of last upgrade of a structure in a portfolio might not be available when performing a portfolio analysis to estimate hurricane losses in a certain region. In that case, it becomes necessary to assume a certain distribution of ages in the region to develop an average vulnerability by combining weak, medium, and strong.

The tax appraisers' databases include effective year of construction and thus provide guidance as to how to weigh the combined weak, medium, and strong model results when year built information is not available in other portfolio files. In each region, the data were analyzed to provide the age statistics. These statistics were used to weigh the average of weak, medium, and strong vulnerabilities in each region. The results are shown in Figure 12 for the wind-borne debris zone in the Central region. The different weighted vulnerability curves are shown for the weak, medium, and strong models, superimposed with the age-weighted vulnerability curve.



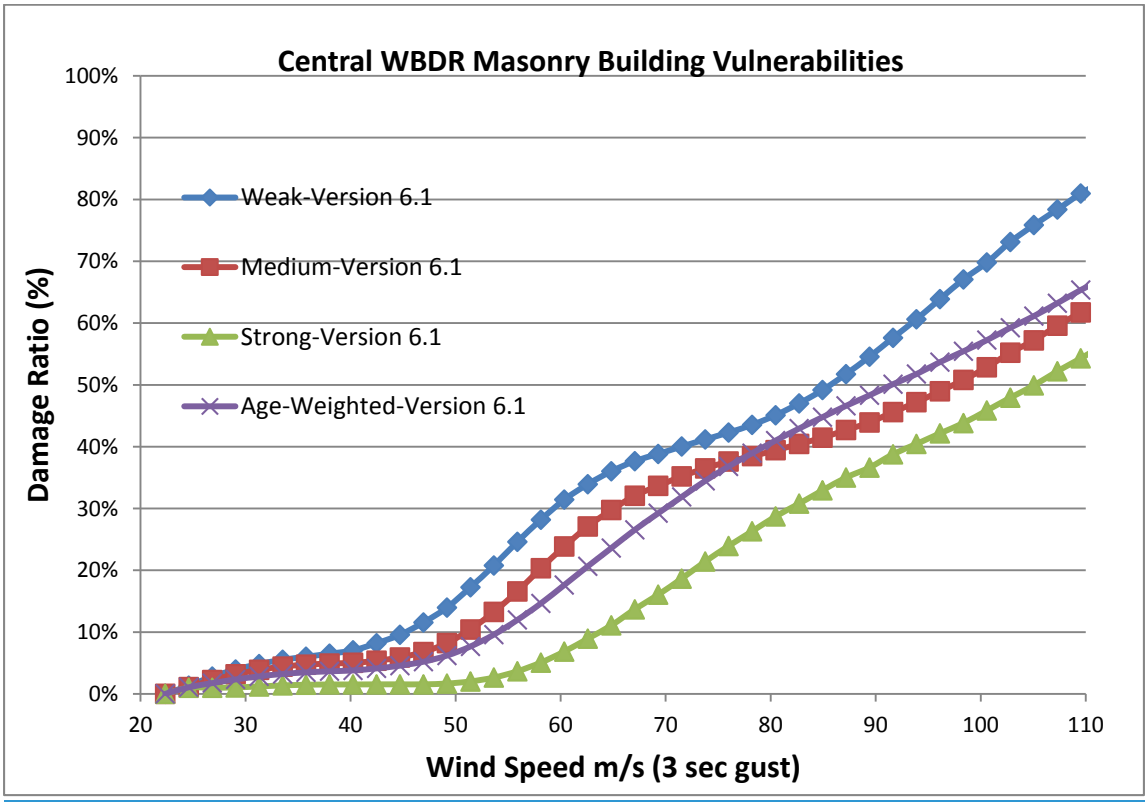


Figure 12. Weighted masonry structure vulnerabilities in the central wind-borne debris region.

Mapping of Insurance Policies to Vulnerability Matrices

The FPHLM processes insurance portfolios from many different insurance companies. Since there is no universal way to classify building characteristics, each company assigns different names or classifications to the building variables. In many cases most of the building structural information in a portfolio is unknown since, in general, detailed records of building characteristics are missing. In a minority of cases, parameters are known, but they do not match any value in the library of the FPHLM. In this case these parameters are classified as "other." For example, the FPHLM models only timber or masonry residential single-family homes. A steel structure would be classified as other.

This makes the mapping of existing portfolio policies to available vulnerability matrices challenging. The engineering team designed a mapping tool to read a policy and assign building characteristics, if unknown or other, on the basis of building population statistics and year built, where the year built serves as a proxy for the strength of the building. The process is summarized in Table 6. Once all the unknown parameters in the policy have been defined, an unweighted vulnerability matrix based on the corresponding combination of parameters can then be assigned. If the number of unknown parameters exceeds a certain threshold defined by ~~the user of the program, he or she always has the choice of using a weighted matrix or age-weighted matrix instead.~~ the actuarial team, a weighted matrix or age-weighted matrix is used instead.

In the few cases in which a policy in a portfolio has a combination of parameters that would result in a vulnerability matrix different than any of the existing matrices in the library of the FPHLM, the program assigns to the policy a so-called "other" weighted matrix (see Table 6 below). The "other" matrices are an average of timber and masonry matrices.

Table 6. Assignment of vulnerability matrix depending on data availability in insurance portfolios.

Data in Insurance Portfolio	Year Built	Exterior Wall	No. of Story	Roof Shape	Roof Cover	Opening Protection	Vulnerability Matrix
Case 1	known	known	known	known	known	known	Use unweighted vulnerability matrix
Case 2	known	known or unknown	Any combination of the four parameters is either unknown or other				use weighted matrix or replace all unknown and others randomly based on stats and use unweighted vulnerability matrix
Case 3	known	other	Any combination of the four parameters is either unknown or other				use the "other" weighted matrix

Data in Insurance Portfolio	Year Built	Exterior Wall	No. of Story	Roof Shape	Roof Cover	Opening Protection	Vulnerability Matrix
Case 4	unknown	known	Any combination of the four parameters is either unknown or other				use age weighted matrix or replace all unknown and others randomly based on stats and use unweighted vulnerability matrix
Case 5	unknown	other	Any combination of the four parameters is either unknown or other				Use age weighted matrices for "other"

Models' Distribution in Time

Over time the codes used for construction in Florida have evolved to reduce wind damage vulnerability. The weak W00, modified weak W10, retrofitted weak W01, medium M00, modified medium M10, retrofitted medium M01, and strong models represent this evolution in time of relative quality of construction in Florida. Each model is representative of the prevalent building type for a certain historical period. However, the assignment of a building strength (its relative vulnerability to wind damage) based on its year of construction is not a straightforward task. The appropriate relationship between age and strength is a function of location within Florida, code in place in that location, and code enforcement policy (also regional). It is therefore important to define the cut-off date between the different periods since the overall aggregate losses in any region are determined as a mixture of homes of various strengths (ages). The cut-off dates are based on both the evolution of the building code and the prevailing local builder/community code enforcement standards in each era.

Given the importance of these issues in the estimation of wind damage vulnerability, a brief history of codes and enforcement is presented next.

Construction practice in South Florida recognized the importance of truss-to-wall connection as early as the 1950s, when it became common to use clips rather than toe nails. The clips were not as strong as modern straps, but they were an improvement over nails. North Florida has fewer historical occurrences of severe hurricane impact, resulting in weaker construction in general than in the south within the same given era. The use of clips became relatively standard statewide by the mid-1980s. The use of improved shingle products and resistant garage doors became more common after Hurricane Andrew.

The issue of code enforcement has also evolved over time. The State of Florida took an active role in uniform enforcement only recently. Prior to Hurricane Andrew, a given county may have built to standards that were worse than or exceeded the code in place at the time. Following consultation with building code development experts, which included the director of the Miami-Dade building department, the president of an engineering consulting firm and consultant to the South Florida Building Code, the consensus was that the issue was not only the contents of the code, but also enforcement of the code.

In an attempt to standardize construction, some cities and counties in Florida adopted building codes, some of the earliest being Clearwater, which adopted a draft of the Standard Building Code (SBC) in 1945 (Cox, 1962); Daytona Beach in 1946 (The Morning Journal, 1946); Bradenton and Manatee counties by 1950; Sarasota County in 1956 (Sarasota Journal, 1956), and Riviera Beach in Palm Beach County in 1957 (The Palm Beach Post, 1957). Miami-Dade and Broward counties adopted the South Florida Building Code (SFBC) in 1957 and 1961, respectively. The SFBC, one of the most stringent codes in the United States, had some wind provisions since its inception. SBC made wind-load provisions mandatory in 1986. Modern wind design started in 1972 and improved considerably for low-rise construction in 1982 (Mehta, 2010). In addition, Florida's construction boom of the 1970s led the state authorities to promote a statewide uniformity of building standards. The first attempt was Chapter 553, "Building Construction Standards," of the Florida Statutes (F.S.), which was enacted in 1974 and required all counties to adopt a code by January 1st, 1975. The statute selected four allowable minimum codes as the pool from which jurisdictions needed to adopt their official building codes, namely: (1) SBC (Southern Building Code Congress International, 1975), (2) the SFBC (South Florida Building Code, 1957), (3) the One and Two Family Dwelling Code, (CABO) (ICC, 1992) and (4) the EPCOT code (enforced in Walt Disney World and based on the SBC, SFBC, and Uniform Building Code) (Reedy Creek Improvement District, 2002). However, the responsibility for the administration and enforcement was left to the discretion of 400 local jurisdictions as diverse as local governments, local school boards, and state agencies (Governor's Report, 1996). The State allowed the jurisdictions to choose any code from the four allowed codes and granted them the authority to amend the code according to their needs, as long as the amendments resulted in more stringent requirements and the power to enforce it.

Problems in the Building Code System

After 1975, there were two main codes in use in Florida before the 1990s: the SFBC in Miami-Dade and Broward counties and the SBC in most of the rest of the state. Although the SFBC was the most stringent code in Florida, this was uncorrelated with compliance and enforcement from many builders, design professionals, and inspectors. To a lesser extent, some of the code stringency was eroded for almost three decades (Getter, 1992; Fronstin & Holtmann, 1994). Some measures that watered down the code included the allowance of power-driven staples instead of nails for roof decking, thinner roofing-felt, 63 mph resisting shingles, and waferboards (pressed wood) as a replacement for plywood for roof decking. A study by Florida A&M University published in 1987 also highlighted deficiencies in code compliance and enforcement in the rest of Florida. Furthermore, the local amendments created a state of confusion, making it difficult for engineers, architects, and contractors to identify the locally administered codes and their jurisdictions (Shingle, 2007; Barnes et al., 1991).

The aftermath of Hurricane Andrew confirmed the concerns reported above. Post-storm damage surveys revealed innumerable violations to the SFBC (the absence of corner columns, vertical reinforcement, and gypsum board used as wall sheathing to name a few) that produced catastrophic failures of buildings (Khan & Suaris, 1993; Siddiq Khan & Associates, 1993). Clearly there were serious shortcomings in the compliance and enforcement process.

For later hurricanes like Opal and Erin in 1995, the rebuild process was also delayed because of the intricacies of the jurisdictional, enforcement, and compliance issues of the codes, exacerbating losses. An expeditious and unambiguous system would have eased proper compliance and enforcement and therefore would have drastically reduced losses (Governor's Report, 1996).

Post-Andrew Building Code Development Enforcement

The South Florida Building Code

Three to four months after Hurricane Andrew, South Florida began to reform the code and the code enforcement system. Engineers became directly involved in the design of residential structures. OSB decking and staples were banned. Wind-rated shingles were required. In 1994 the whole SFBC was reformed and adopted the ASCE 7 wind provisions.

The Florida Building Code

After Hurricane Andrew, local and state agencies were unsure about how to guarantee building safety. Concerns arose that a diminution of insurance availability would occur, which threatened the continuity of economic growth. In response, Governor Lawton Chiles established a Building Codes Study Commission in 1996 to review the current system of codes. The Governor's Commission found that the existing system had led to a "patchwork of technical and administrative processes." Its recommendations led to the formation of the Florida Building Commission in 1998, which was responsible for creating a unified Florida Building Code (Governor's Report, 1996).

For the new unified Florida Building Code (FBC), the Commission selected the SBC, developed in Alabama from 1940 to 1945 (Ratay, 2009), as the base code because 64 out of 67 counties were already using the 1973 and the 1997 versions of the code with amendments (Shingle, 2007). The SFBC was later included as an additional base code in 1999 to meet South Florida's special requirements. The Building Commission worked to reach a consensus among all stakeholders, and the first version of a unified FBC was made effective on March 1, 2002 (Blair, 2009). Studies indicate that the losses due to hurricanes have decreased since the enactment of the FBC (Gurley et al., 2006).

Application of the Building Code History

The history above clearly indicates that a completely accurate accounting of all building practices in every region of Florida going back many decades is not possible, given the limited policy information of age and location. To accommodate the history of residential building construction practice in Florida, buildings were classified into different eras. The classifications shown in Table 7 were adopted for characterizing the regions by age and model. The strength descriptions within Table 7 are provided at the bottom of Table 7 in terms of the nomenclature used in Table 1~~Table 1a~~ and ~~Table 1b.~~Table 2. The specific building eras and classifications per region are based on the evolution of the building codes in Florida and the opinions of the experts consulted.

Table 7. Age classification of the models per region.

	Pre-1960	1960-1970	1971-1980	1981-1993	1994-2001	2002-pres.
HVHZ	⅔ modified Weak, ⅓ Medium	⅔ Weak, ⅓ Medium	½ Weak, ½ modified Medium	⅔ Weak, ⅓ modified Medium	Modified Strong	Modified Strong
Keys	½ modified Weak, ½ Medium	Medium	Medium	Medium	⅓ Medium ⅔ Strong_OP	Strong_OP
WBDR	modified Weak	⅔ Weak, ⅓ Medium	⅓ Weak, ⅔ Medium	⅓ Weak, ⅔ Medium	½ Medium, ½ Strong_OP	Strong_OP
Inland	modified Weak	⅔ Weak, ⅓ Medium	½ Weak, ½ Medium	½ Weak, ½ Medium	½ Medium, ½ Strong	Strong

Table 7 Nomenclature with respect to Table 1~~Table 1a~~ and ~~Table 1b~~Table 2

Strong:	S00
Strong_OP:	S00-OP
Modified Strong:	S01
Medium:	M00
Modified Medium:	M10
Weak:	W00
Modified Weak:	W10

Note: HVHZ means high velocity hurricane zone; WBDR means wind borne debris region. The boundaries of the WBDR vary depending on the year built, and the edition of the FBC which applies, as explained in Standard G-1, in the description of the site-built models.

Appurtenant Structures

Appurtenant structures are not attached to the dwelling or main residence of the home but are located on the insured property. These types of structures could include detached garages, guesthouses, pool houses, sheds, gazebos, patio covers, patio decks, swimming pools, spas, etc. Insurance claims data reveal no obvious relationship between building damage and appurtenant structure claims. The variability of the structures covered by an appurtenant structure policy may be responsible for this result.

Since the appurtenant structures damage is not derived from the building damage, only one vulnerability matrix is developed for appurtenant structures. To model appurtenant structure damage, three equations were developed. Each determines the appurtenant structure insured damage ratio as a function of wind speed. One equation predicts damage for structures highly susceptible to wind damage, the second predicts damage for structures moderately susceptible to wind damage, and the third predicts damage for structures that are affected only slightly by wind.

Because a typical insurance portfolio file gives no indication of the type of appurtenant structure covered under a particular policy, a distribution of the three types (slightly vulnerable, moderately vulnerable, and highly vulnerable) must be assumed and is validated against the claim data.

VULNERABILITY COMPONENT: COMMERCIAL RESIDENTIAL MODEL

Given the hurricane hazard defined by the atmospheric component, the engineering component performs several tasks: (1) it estimates the physical damage to exterior components of typical buildings or apartment units; (2) it assesses the interior and utilities damage and contents damage due to water penetration through exterior damage and defects to interior walls, ceiling, doors, etc.; (3) it combines the exterior and interior damage to estimate the building and content vulnerabilities; (4) it estimates the time related expenses; and (5) it estimates appurtenant structure vulnerability (Pita et al., 2008, 2009a, 2009b, 2009c, 2010, 2011a, 2011b, 2011c, 2012a, 2012b, 2013, 2014; Pinelli et al., 2009b, 2010b, 2012, 2013a, 2013b; Weekes et al., 2009, 2014).

Exposure Study

Most low-rise commercial residential buildings (LB) (Figure 13) can be categorized into a few generic groups having similar structural characteristics, layout, and materials, although they may differ somewhat in dimensions. These buildings can suffer substantial external structural damage, in addition to envelope and interior damage, from hurricane winds. The modeling approach to assessing damage for these building types is the same as that for assessing damage for personal residential buildings, modeling the building as a whole.

However, commercial residential mid- and high-rise buildings (MHB) (Figure 14) are very different from low-rise buildings and single-family homes. The mid-/high-rise buildings are engineered structures, which suffer few structural failures during a windstorm but are subject to water ingress from cladding and opening failures. These buildings, which come in many different types, shapes, height, and geometries, consist of steel, reinforced concrete, timber, masonry, or a combination of different structural materials.

It is not realistic to perform damage simulations on a reduced collection of ‘base’ buildings, as is done for single-family residential and low-rise commercial residential buildings, because that will necessarily leave out a majority of existing mid- and high-rise typologies. For instance, for steel frame structures alone there are a wide variety of possible building shapes and configurations. These different shapes lead to very different wind-loading scenarios and therefore different vulnerabilities. Equally important, the number of MHB is at least an order of magnitude smaller than the number of PRB or LB. It is therefore not feasible to average the losses over a very large number of buildings and compensate small differences between buildings, as in the case of PRB. On the contrary, the analyst is faced with a relatively small number of buildings, each of which is different from the other.

As a result, the FPHLM has adopted a modular approach to model mid- and high-rise buildings. Rather than considering a structure as a whole, the model treats the building as a collection of apartment units. The base modules are typical apartment units, divided as corner and middle units. Thus, buildings with any number of stories and any number of units per floor can be modeled by aggregating the corresponding apartment units vulnerabilities and accounting for correlation of damage among units (e.g., water ingress through an envelope breach in a fifth-floor unit creates problems for lower units with no failures).

To summarize, in the case of LB (low rise buildings), typical models of the whole structure that are representative of the vast majority of this building population in Florida were defined. In the case of MHB (mid-high rise buildings), typical models of individual units that are representative of the vast majority of units in Florida were defined.

An extensive survey of the commercial residential Florida building stock was carried out to generate a manageable number of these building and apartment models to represent the majority of the Florida residential building stock. The modelers analyzed Florida counties' property tax appraisers' (CPTA) databases for building stock information. Although the database contents and format vary from county to county, many of the databases contain the structural information needed to define the most common structural types. Information from 21 counties was collected for commercial residential buildings. The modelers extracted information on several building characteristics for classification, including roof cover, roof shape, exterior wall material, number of stories, year built, building area, foundation type, floor plan, shape, and opening protection.



Figure 13. Typical low-rise buildings (LB).



Figure 14. Examples of mid- and high-rise buildings (MHB).

Commercial Residential Building Survey

In the case of the commercial residential buildings, the CPTAs classify the buildings either as condominiums or as multifamily residential (MFR) based only on the type of ownership. Condo buildings are such that each unit or apartment has a different owner. The condo unit can then be occupied by the owner or by a renter. The CPTAs do not record if the condo unit is rented or owned. Condo owners' expenses include the maintenance and use of the common areas and common facilities because the condo owner actually owns a percentage of the entire facility. The condo buildings relevant to this survey are all classified by the CPTAs as residential. Commercial office condo buildings are out of the scope of the survey.

A MFR building has a single owner who rents the units to tenants. The CPTAs classify MFR buildings with fewer than 10 units (duplex, triplex, and quadruplex) as residential buildings; MFR buildings with 10 units or more are classified as commercial buildings. Both residential and commercial MFR buildings were considered in this survey. MFR buildings are interchangeably referred to as apartment buildings by CPTAs. Residential MFR buildings (fewer than 10 units) account for approximately 70% of the MFR building stock, and the remaining 30% are commercial MFR buildings (10 units or more).

The commercial-residential buildings, regardless of whether they are condos or MFR buildings, were divided in two categories: low-rise (one–three stories) and mid-high rise (four stories and more). Low-rise buildings have three stories or fewer. The survey shows these buildings, which represent the majority of the building stock, have different characteristics than taller buildings. Unanwa (1997) uses a similar definition in his study. The mid- and high-rise buildings tend to be more heterogeneous and necessitate a different treatment in the vulnerability model. Owned as well as rented apartment units are included in this survey; the CPTAs do not distinguish between the two.

Appraisers have confirmed that MFR buildings tend to have fewer stories than condo buildings and the majority of MFR buildings are duplexes, triplexes, and quadruplexes. Also, the proportion of MFR buildings that can be classified as mid-/high-rise is negligible according to available information and consultation with CPTAs.

Building Models

Distinctly different construction characteristics and modes of damage in high winds led to the development of separate models for low-rise commercial residential construction (LB) and mid-/high-rise commercial residential construction (MHR).

Low-Rise Commercial Residential Models

The LB model was developed to represent typical apartment and town-house style structures of three stories or fewer (Figure 13). The model framework is based on the single-family, site-built residential model, which uses a probabilistic description of wind loads and exterior and structural component capacities to project physical damage as a function of wind speed. The components in the LB damage model include roof cover, roof sheathing, roof-to-wall connections, wall type, wall sheathing, windows, entry doors, sliding-glass doors, soffits, and gable end truss integrity.

Given the large array of sizes and geometries for low-rise commercial residential structures, the program is developed to provide flexibility in choosing a building layout and dimensioning details (footprint, overhang length, roof slope, roof shape, etc.). The changes in construction practice over decades in Florida also necessitate flexibility when choosing construction quality with regard to hurricane wind resistance. The model allows the selection of building components with a variety of strength options to represent a range from low to high wind resistance (braced or unbraced gable ends, old or new roof cover, sheathing nailing schedules, etc.).

A standard (default) model was developed based on the building exposure study that quantified average square footage per story, units per story, and other descriptors. Default settings were also developed to represent weak, medium, and strong construction practice. Any given strong, medium, or weak model may be altered by additional mitigation or retrofit measures individually or in combination. For example, reroofing an older apartment can be represented by increasing the probabilistic descriptor of capacity for the roof cover.

Outputs (damage matrices) have been produced for each combination of the following: building height (one, two, or three stories), wall type (timber or masonry), roof shape (hip or gable), strength (weak, medium, or strong), and window protection (no protection or with metal shutters).

Mid-/High-Rise Commercial Residential Models

The mid-/high-rise model uses the Monte Carlo simulation concept, but it differs from the low-rise model in significant ways. There is a high level of variability among mid-/high-rise buildings because of the combination of the number of stories, the number of units per floor, intentionally unique geometries, and the materials used for the exterior. This makes the application of a "standard" or default model unfeasible. Because of the construction methods and materials used in these structures, damage to the superstructure and exterior surfaces of the buildings tends to be relatively minor. The majority of damage accumulation in mid-/high-rise structures is due to water penetration and failure of openings. The model reflects this by focusing on the failure of windows and doors, the ingress of rain water, and the proliferation of water from the source of the ingress to adjacent living units. The structure in whole is not modeled. Rather, individual units are modeled in isolation. That is, the vulnerability of a single unit is explicitly modeled, and damage is assessed to openings as a function of wind speed.

Two different mid-/high-rise classifications are modeled for this study: "closed building" and "open building." Closed buildings are characterized by the location of the unit entry doors at the interior of the building. The sliding-glass doors and windows are all facing the exterior of the building. For the open building model there is exterior corridor access to each unit entry door on one side of the building, and the patio areas are situated on the opposite side of the building (Figure 15). The type of building chosen can increase or decrease the vulnerability of a selected unit because of the exposure of the exterior openings. Middle units in a closed or open building have one or two exterior walls, respectively.

There are three main differences between the low-rise and mid-/high-rise models: (1) the use of a modular (i.e., per unit rather than per building) approach, (2) the exterior components being analyzed for failure, and (3) the use of two basic floor plans. Location of unit within the plan view of the building, unit square footage, and number of available openings are some of the important factors that separate one unit from another.

Corner units are subjected to higher wind pressures that are present along the edges of the building, compared to the middle units, which are located within lower pressure zones at the center of the wall area (Figure 15). Increased square footage typically results in an increase in exterior wall frontage and the number of openings vulnerable to damage.

The MHB model uses the same analysis and output technique as the LB model. The difference is the number of failure types modeled. The MHB model analyzes only the damage to the openings, which include the windows, sliding doors, and entry doors. Each of the components can fail due to pressure or debris impact.

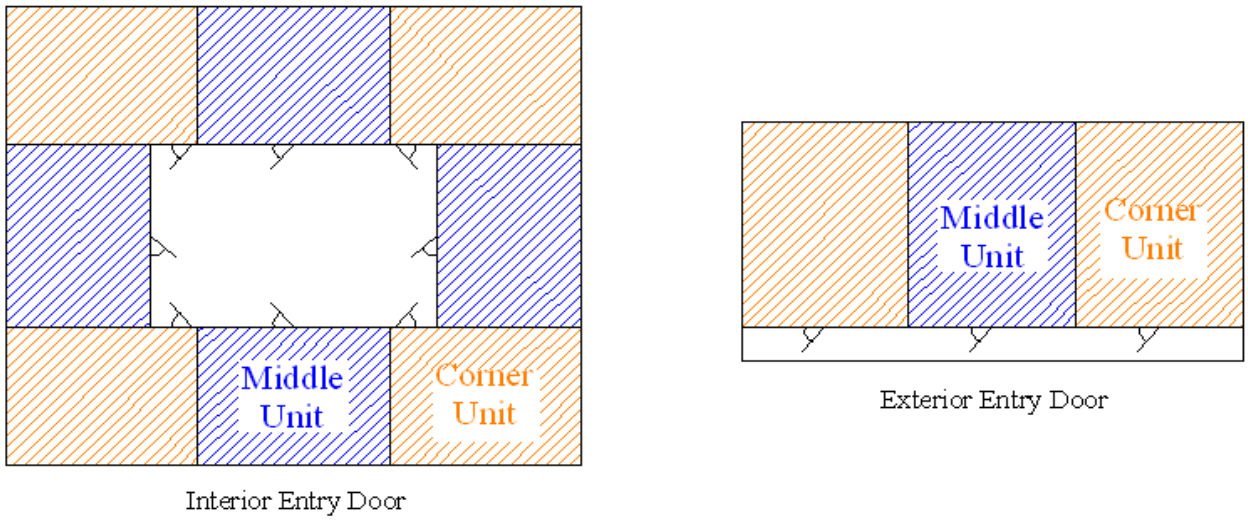


Figure 15. Apartment types according to layout (left: closed building with interior entry door; right: open building with exterior entry door).

Damage Matrices

Exterior Damage

The vulnerability model uses a Monte Carlo simulation based on a component approach to determine the external vulnerability (as shown in Figure 10) at various wind speeds of buildings in the case of LB, or apartment units in the case of MHB. For the case of LB, the procedure is identical to the one described for single-family residential (PRB). In the case of MHB, the simulations address only wind pressure and debris impact on the openings.

The damage assessment is conducted over a range of wind speeds and wind directions, and results are stored in a damage matrix. Probabilistic damage assessment is conducted by first creating an individual building realization by mapping each component according to typical construction practice. Random capacity values are assigned to the various components on the basis of a probability distribution for each component type. This realization is subjected to a peak three-second gust wind speed from a particular direction. Directional loads are calculated using randomized pressure coefficients based on directional modifications to ASCE 7 as well as wind tunnel data (NIST Aerodynamic Database - http://fris2.nist.gov/winddata), and a comparison of resulting surface and internal loads to component capacities is conducted. Damage occurs when the assigned capacity of a component is exceeded by its loading. Once the openings have been checked for failure due to pressure, the damage due to the impact of windborne debris is also evaluated. Damaged components are removed, and a series of checks are performed to determine

if lost components will redistribute loading to adjacent components or change the overall loading. For example, loss of a roof-to-wall connection places additional load on adjacent connections, whereas an envelope breach will potentially alter internal loading—changing the overall loading on most components. Iterative convergence is used to produce the final damage state for that building realization. The results of this single simulation are documented on the basis of the final iteration, another realization of that building is constructed by assigning new random capacities to each component, and the process repeats for the same three-second gust, same wind direction, and newly randomized pressure coefficients based on the number of desired simulations the user would like to run. The process is repeated for eight wind directions and a series of three-second wind speeds between 50 and 250 mph in 5 mph increments.

The output of the Monte Carlo simulation model is an estimate of physical damage to structural and exterior components. The results are in the form of a four-dimensional damage matrix. Each row of the matrix lists the results of one simulation. The amount of damage to each of the modeled components for a simulation is listed in 75 columns. The third dimension represents the peak three-second gust wind speed between 50 and 250 mph in 5 mph increments, and the fourth dimension represents the eight angles between 0 and 315 degrees in 45-degree increments. Table 7 delineates the damage matrix contents for the case of the LB. A description of each of the nine columns of the MHB damage matrix is given in Table 9.

Table 8. Description of damage matrices for LB.

Column #	**Timber Models**	**Masonry Models**
Col 1	Percent roof cover (shingles or tiles) failed	
Col 2	Percent field roof sheathing lost (field roof sheathing is all but overhang)	
Col 3	Percent edge (overhang) roof sheathing failed	
Col 4	Percent roof-to-wall connections failed	
Col 5	Collapse of gable end trusses (0 = no, 1 to 20) starting from side 1	
Col 6	Collapse of gable end trusses (0 = no, 1 to 20) starting from side 2	
Col 7-8	Percent gable end wall covering failed (side 1 and 2, positive for windward, negative for leeward)	
Col 9-10	Percent gable end sheathing failed (side 1 and 2, positive for windward, negative for leeward)	
Col 11- 14	Percent wall covering failed – 1st floor (walls 1-4, positive for windward, negative for Leeward)	Shear Damage Ratio for Masonry Walls- 1st Floor (walls 1-4, positive for windward, negative for leeward)
Col 15-18	Percent wall sheathing failed – 1st floor (walls 1-4, positive for windward, negative for leeward)	Bending Damage Ratio for Masonry Walls- 1st Floor (walls 1-4, positive for windward, negative for leeward)
Col 19-22	Number of windows failed from wind pressure – 1st floor - (walls 1-4, positive for windward, negative for leeward)	
Col 23-26	Number of windows failed from wind Debris– 1st floor - (walls 1-4)	
Col 27	Number of sliding glass doors failed from wind pressure – 1st floor (+ for windward - for leeward)	
Col 28	Number of sliding glass doors failed from debris impact – 1st floor	
Col 29	Number of entry doors failed from wind pressure – 1st floor (+ for windward - for leeward)	
Col 30	Number of entry doors failed from debris impact – 1st floor	
Col 31-50	Repeat Col 11 - Col 30 for 2nd Floor	
Col 51-70	Repeat Col 11 - Col 30 for 3nd Floor	
Col 71	Garage Door Damage (positive for windward, negative for leeward)	
Col 72-75	Percent Soffit Damage (walls 1-4)	

Table 9. Description of the damage matrices for MHB apartments.

Commercial and Single Family Residential	
Column #	**Inner and Outer Stair Models**
Col 1	Number of Windows failed from wind pressure
Col 2	Number of Entry Doors failed from wind pressure
Col 3	Number of Sliding failed from wind pressure
Col 4	Number of Windows failed from debris impact
Col 5	Number of Entry Doors failed from debris impact
Col 6	Number of Sliding failed from debris impact
Col 7	Number of Windows breached from debris impact
Col 8	Number of Entry Doors breach from debris impact
Col 9	Number of Sliding breach from debris impact

Interior and Utilities Damage

The FPHLM introduced a novel approach to assessing the interior damage by considering the physics of the problem. The approach starts from the damage to the building envelope (Weekes et al., 2009), described in the previous section. The model then estimates the amount of wind-driven rain that enters through the breaches and defects in the building envelope and converts it to interior damage. The approach is summarized below. More details are provided in standard V-1 and in (Pita, 2012; Pita et al., 2012a).

The method (Figure 16) combines existing building defects and estimated building envelope damage with the impinging rain to predict the amount of water that will enter a building. This physically based approach models the main contributor to interior damage, addresses the uncertainty in the interior damage source, and documents the individual water ingress contribution of each component to the total water intrusion.

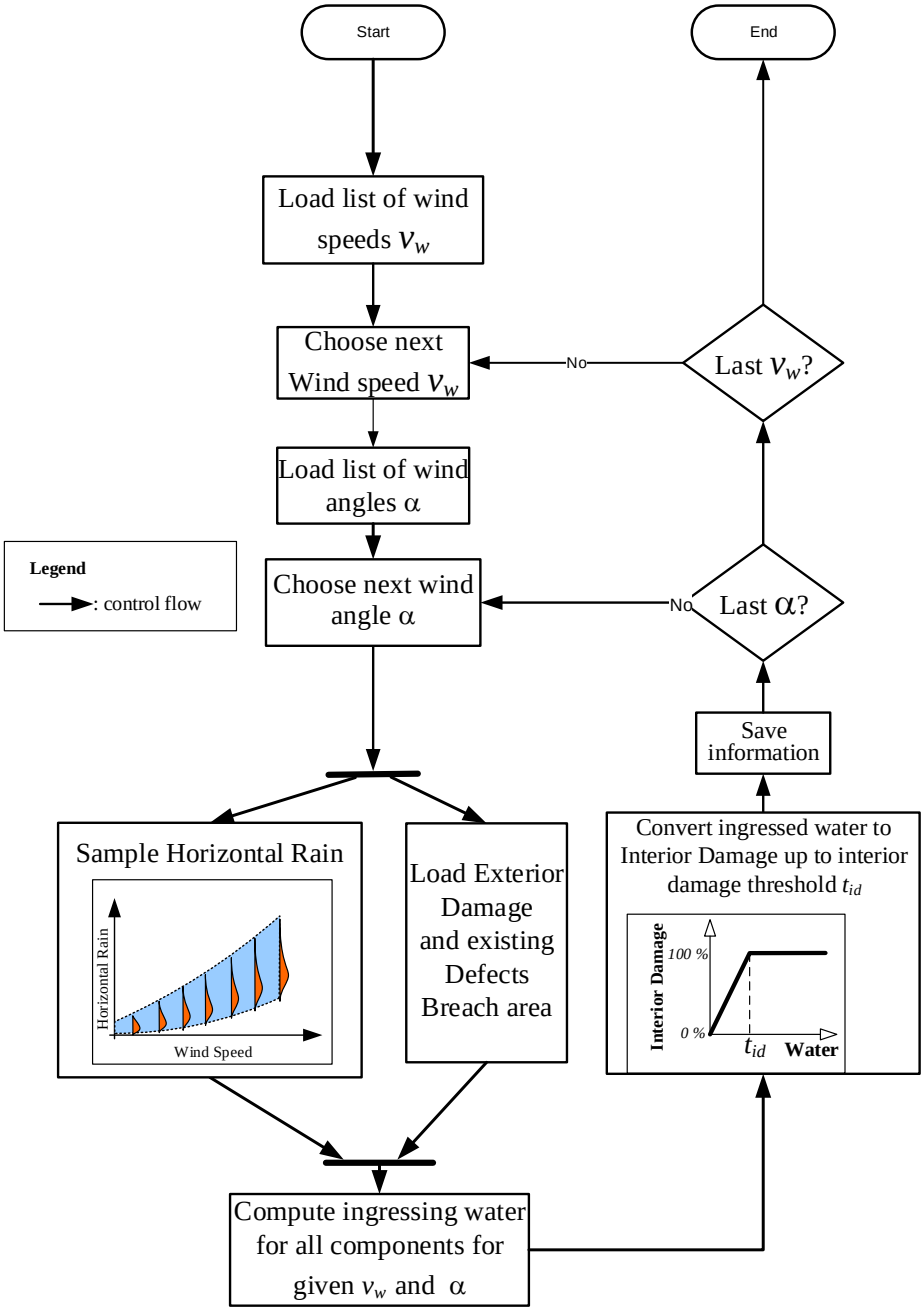


Figure 16. Flowchart of the interior damage model.

The exterior building components that the model considers include roof cover, roof sheathing, wall cover, wall sheathing, gable cover, gable sheathing, windows, doors, and sliding doors. In the case of MHB units, only windows, doors, and sliding doors are considered. For a given wind speed, the model first estimates breach areas of each component from the exterior damage array. The area of existing defects in envelope components is estimated based on surveys (Mullens et al., 2006) and engineering experience.

This approach for both low-rise and mid/high-rise buildings estimates the amount of water that enters through the breaches and defects of each component of the envelope. The total amount of

water is calculated by adding the contribution of all components for a given wind speed, and by estimating the water which percolates from story to story. The final step maps water inside the building to interior damage with a bilinear relationship, where total interior damage is achieved for a certain threshold of height of accumulated water.

Contents Damage

Contents include anything in the building that is not attached to the structure itself. As in the case of interior and utilities damage, the contents damage is assumed to be a function of the amount of water that penetrates the building, and it is therefore proportional to interior damage. The function is based on engineering judgment and is validated using claims data. In the case of a condo building, only the contents of the common areas are covered by the policy. In the case of an apartment building, the personal contents of the renters are not covered by the building policy.

Time Related Expenses

Time Related Expenses refer to loss of rent for owners of apartment buildings, which are mainly low-rise commercial residential buildings. As in the case of interior and utilities damage, the Time Related Expenses are assumed to be a function of the amount of water that penetrates into the building, and they are therefore proportional to interior damage. The function is based on engineering judgment and should be validated using claims data, which is almost non-existent.

Vulnerability Matrices for Low-Rise Buildings

Unweighted Vulnerability Matrices of LB

A description of the process to estimate the total vulnerability of low-rise buildings is displayed in Figure 17. Given a particular building type, the Monte Carlo simulation-generated damage array that expresses the exterior damage in the envelope is loaded. For a particular wind speed and wind direction, each component physical damage is normalized to a percentage value. For instance, the number of damaged doors, windows, and sliding doors is divided by the total number of the corresponding openings; collapsed trusses are divided over the total number of trusses, etc. The cost of the damage is then assessed.

Interior damage is estimated by (1) simulating the amount of wind-driven rain that enters through the breaches and defects in the building envelope, (2) propagating water from floor to floor, and (3) converting to damage to interior and utilities.

Replacement cost ratios provide the link between modeled physical damage and the corresponding monetary losses. They can be defined as the cost of replacing a damaged component or assembly of a building divided by the cost of constructing a completely new building of the same type. An explicit procedure is used to convert physical damage of the modeled components to monetary damage. The procedure is almost identical to the one already described for single-family residential buildings. The damage ratio (DR) as a function of wind speed for the exterior, interior, and utilities is calculated by adding the corresponding costs of damaged exterior plus damaged interior plus damaged utilities divided over the overall building cost that is contingent upon the type and size of the building.

Derivation of the probability distribution functions of damage at each wind speed interval is the final step of the process. For each wind speed interval, the probability of damage given that wind speed interval (i.e., the cells of the vulnerability matrices) is computed as the summation of specific damage ratios for all wind directions divided by the total number of simulations at that particular wind speed interval.

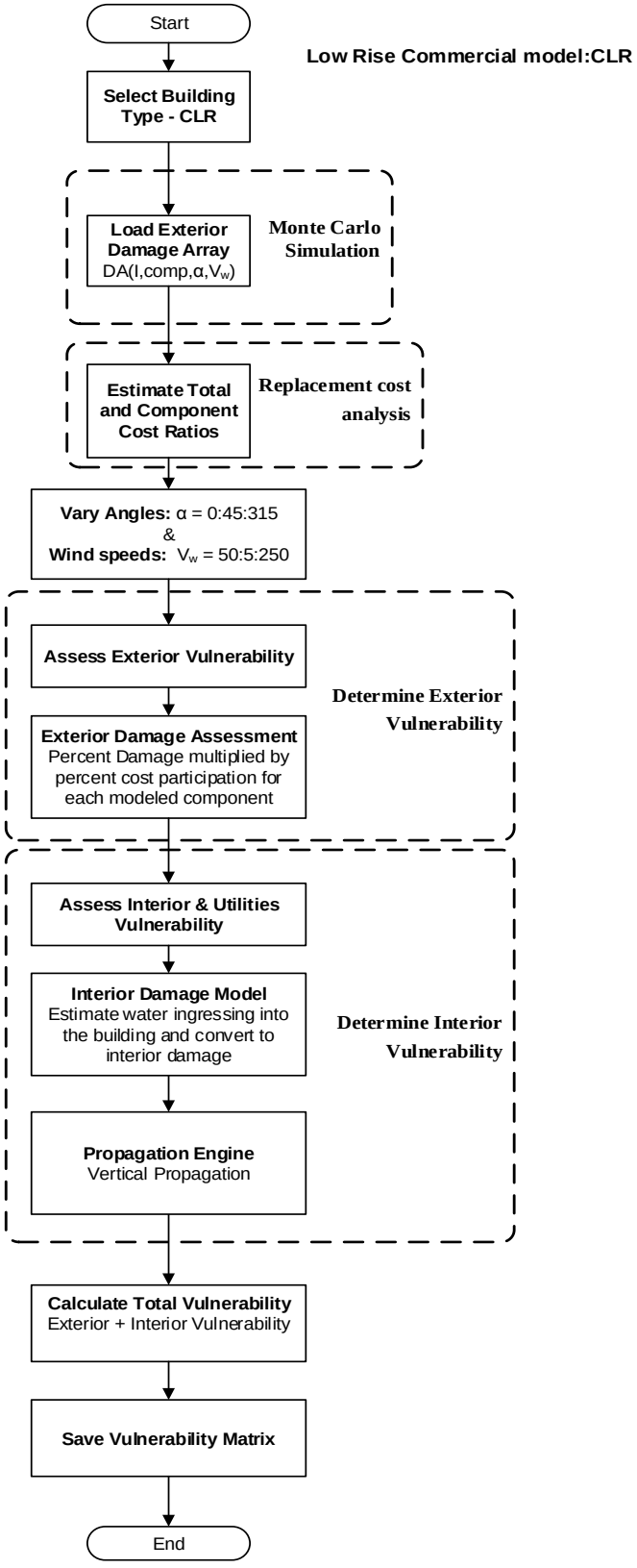


Figure 17. Procedure to create a CR vulnerability matrix.

Weighted Vulnerability Matrices of LB

In the case of LB, vulnerability matrices were created for every combination of construction type (masonry, timber, or other), roof shape (gable or hip), roof cover (tile or shingle or metal), shutters (with or without), number of stories (one, two, or three), and subregion (inland, wind-borne debris region, and high velocity zone). However, in general, there is little information available in an insurance portfolio file regarding the structural characteristics and the wind resistance of the insured property. Instead, insurance companies rely on the ISO fire resistance classification. Portfolio files have information on ZIP Code and year built. The ISO classification is used to determine if the home is constructed of masonry, timber, or other. The ZIP Code is used to define the subregion. The year built is used to assist in defining whether a building should be considered weak, medium, or strong.

From the insurance files, sub-region, construction type, and year built are determined. This leaves the roof shape, roof cover, number of stories, and shutter options undefined. From the exposure study of 21 Florida counties, the distribution of these parameters can be extrapolated. For each age group, we define a weighted matrix for each construction type in each sub-region. The procedure is identical to the one already described for single-family buildings.

Age-Weighted Matrices of LB

The year built or year of last upgrade of a structure in a portfolio may not be available when performing a portfolio analysis to estimate hurricane losses in a certain region. In that case, it becomes necessary to assume a certain distribution of ages in the region to develop an average vulnerability by combining weak, medium, and strong. Here again, the procedure is identical to the one described for single-family residential buildings.

Mapping of Insurance Policies to Vulnerability Matrices for LB

The mapping of the low-rise vulnerability matrices to the insurance policies in any given portfolio is also very similar to the process already reported for single-family buildings.

LB Models' Distribution in Time

The low-rise building models' distribution in time is similar to that of the single-family buildings.

Vulnerability of Mid-/High-Rise Buildings

MHB opening vulnerabilities

In the case of MHB, a process similar to the one described above is followed to derive exterior vulnerability and breach curves for different openings of typical apartment units. These curves are derived for the cases of open and closed buildings, for corner and middle units, with different opening protections (with or without impact-resistant glass; with or without metal shutters). Each vulnerability curve for openings of corner or middle apartment units (window, door, or slider) gives the number or fraction of opening damaged as a function of wind speed. Each breach curve for openings of corner or middle apartment units (window, door, or slider) gives the breach area in ft^2 of opening damaged as a function of wind speed.

MHB building vulnerability

Unlike the single-family home loss model in which interior and exterior damage was aggregated inside the vulnerability module, the aggregation for mid-/high-rise buildings is performed outside that module because of the interior damage propagation. The modular approach produces independent assessments of exterior damage for each unit while also considering the interior water damage that can spread from unit to unit and trigger damage far from its source. Therefore, interior damage is treated in two stages: the first stage occurs as a direct result of the exterior damage, and the second occurs as a consequence of propagation between units. The separate modeling of exterior and interior damage is also well suited to dealing with the insurance issue of different insurance coverage for apartment and condo buildings.

The process for damage estimation for MHB is presented in Figure 18. For each policy in the portfolio, the program reads the information on the building (location and number of stories and units) and assigns a wind speed profile based on its location (i.e., surrounding terrain). The algorithm calculates the number of corner and middle units per floor (a_c and a_M) and loads the corresponding opening vulnerability and breach curves ($V_{C,M}$ and $B_{C,M}$). The vulnerability curves, combined with the wind speed value at every story, W_i, yield the number of openings of each kind damaged at each story, which are then assigned a replacement cost, $C_{W,D,S}$. The result is the cost of damage to the openings at each story (CDO_s), which is then accumulated over all the stories as the total expected cost of damage to the openings (TECDO).

For the interior damage estimation the process is similar. From the wind profile, the corresponding wind speed, W_i, is calculated at each story. For a given story and its corresponding wind speed, the value of the expected breach size for windows, entry door, and sliding door, $B_C^{W.D,S}$ and $B_M^{W.D,S,}$, are retrieved from the corresponding breach curves. The breach size of each component is added to get the total breach size per story. The next step is to estimate the amount of water that will enter a particular story with a given breach size, as described in the section describing the interior damage model. Note that for the sake of simplification, defects are not represented in the flow chart.

Increased water penetration through possible roof cover damage as well as roof defects or ventilation ducts could happen in the upper floors, which would then trickle down to the lower stories. Therefore an additional volume of water penetration is modeled at the upper story.

A scheme for vertical propagation of water between floors was implemented. The water content is then transformed at each story into an interior damage ratio (ID) based on the bilinear relationship described in Standard V-1. The final product of the interior damage assessment is the Expected Interior Damage Ratio (EIDR).

At this point in the process, the algorithm has computed expected damages, both exterior (TECDO) and interior (EIDR), for the particular building of the policy under study. The EIDR is then multiplied by the interior insured value expressed as a percentage of the total insured value BV, thanks to a coefficient k_I which varies for condos and apartment buildings. The final value is the total expected damage value (EDV).

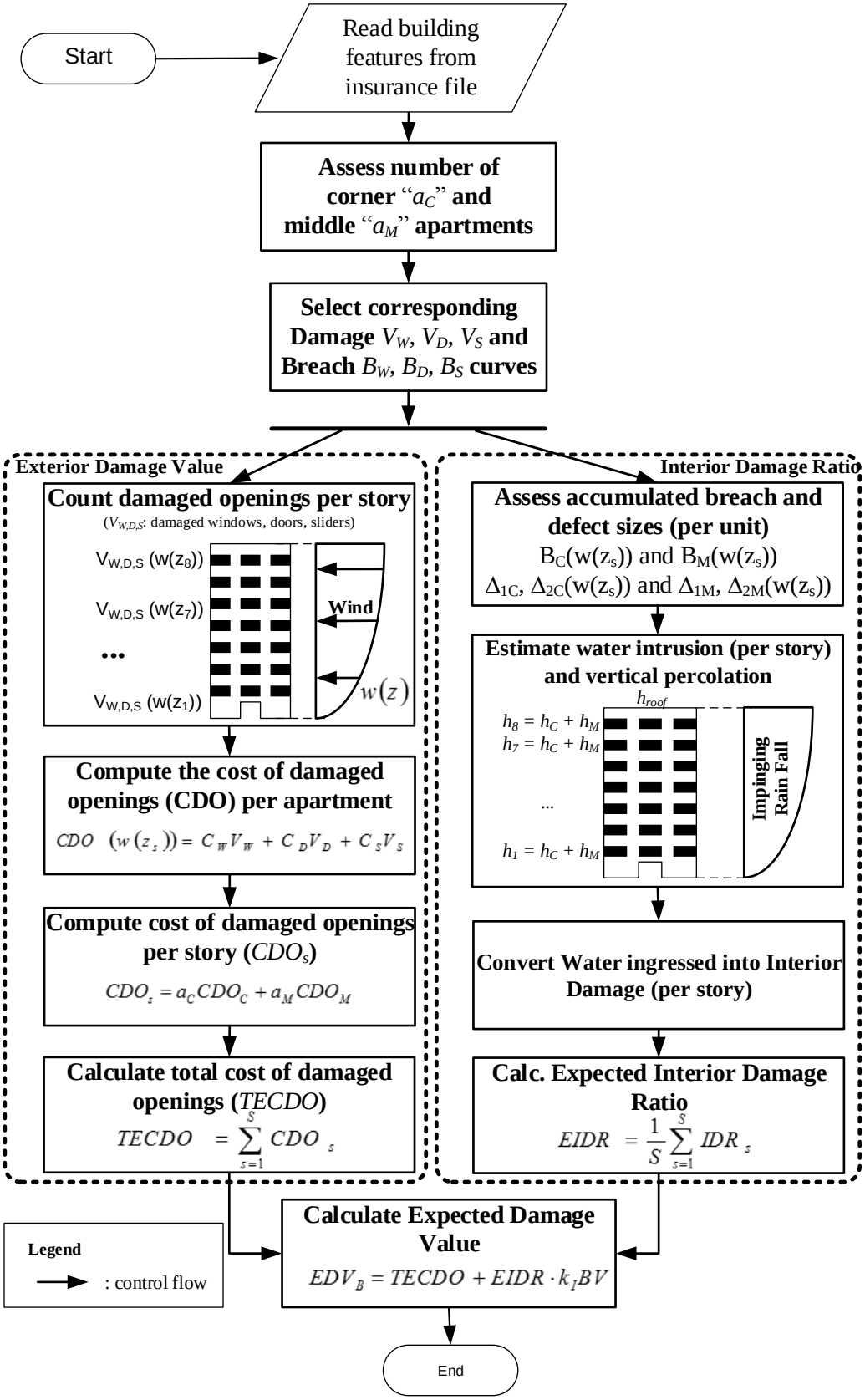


Figure 18. Exterior and interior damage assessment for MHB.

Contents Vulnerability

Contents include anything in the building that is not attached to the structure. In the case of a condo building only the contents of the common areas are covered by the policy. In the case of an apartment building, the personal contents of the renters are not covered by the building policy. In both cases, the contents vulnerability is proportional to the interior vulnerability. The constant of proportionality is based on engineering judgment and is validated using claims data.

Time-Related Expenses

Time-related expenses are coverage for loss of income due to the building damage. The value of a claim is obviously dependent on the time it takes to repair a damaged building as well as the surrounding utilities and infrastructure. This coverage applies only to apartment buildings, where the loss of income is the loss of rent. The time-related expenses are modeled as directly proportional to the interior vulnerability.

Appurtenant Structures

For commercial residential structures, appurtenant structures might include a clubhouse or administration building, which are treated like additional buildings. For other structures such as pools, etc., the appurtenant structures model developed for residential buildings is applicable.

ACTUARIAL COMPONENT

The actuarial component consists of a set of algorithms. The process involves a series of steps: rigorous check of the input data; selection and use of the relevant output produced by the meteorology component; selection and use of the appropriate vulnerability matrices for building structure, contents, appurtenant structure, and additional living expenses; running the actuarial algorithm to produce expected losses; aggregating the losses in a variety of manners to produce a set of expected annual hurricane wind losses; and producing probable maximum losses for various return periods. The expected losses can be reported by construction type (e.g., masonry, frame, manufactured homes), by county or ZIP Code, by policy form (e.g., HO-3, HO-4, etc.), by rating territory, and combinations thereof.

Expected annual losses are estimated for individual policies in the portfolio. They are estimated for building structure, appurtenant structure, contents, and ALE on the basis of their exposures and by using the respective vulnerability matrices or vulnerability curves for the construction types. For each policy, losses are estimated for all the hurricanes in the stochastic set by using appropriate damage matrices and policy exposure data. The losses are then summed over all hurricanes and divided by the number of years in the simulation to get the annual expected loss. These are aggregated at the ZIP Code, county, territory, or portfolio level and then divided by the respective level of aggregated exposure to get the loss costs. This is a computationally demanding method. Each portfolio must be run through the entire stochastic set of hurricanes.

The distribution of losses is driven by both the distribution of damage ratios generated by the engineering component and by the distribution of wind speeds generated by the meteorology

component. The meteorology component provides, for each lat-long grid, the associated probabilities for a common set of wind speeds. Thus, locations are essentially differentiated by their probability distribution of wind speeds. The meteorology component uses up to 56,000 year simulations to generate a stochastic set of storms. The storms are hurricane events at landfall or when bypassing closely. Each simulated storm has a track and a set of modeled windfields at successive time intervals. The windfields generate the one-minute maximum sustained wind speeds for the storm at various locations (lat-long grid) along its track. These one-minute maximum sustained winds are then converted to three-second peak gust winds and corrected for terrain roughness by using the gust wind model and the terrain roughness model.

For each lat-long grid, an accounting is then made of all the simulated storms that pass through it. On the basis of the number of pass-through storms and their peak wind speeds, a distribution of the wind speed is then generated for the grid. On the basis of this distribution, probabilities are generated for each 5-mph interval of wind speeds, starting at 20 mph. These 5-mph bins constitute the column headings of the damage matrices generated by the engineering component.

The engineering group has produced vulnerability matrices for personal residential buildings and vulnerability curves for commercial residential buildings.

Vulnerability matrices are provided for personal residential building structure, contents, appurtenant structures and additional living expenses for a variety of residential construction types and for different policy types. The construction types are masonry, frame, mobile home, and other. The vulnerability matrices are also developed for weak, medium, and strong construction as proxy by year built.

Within each broad construction category, the vulnerability matrices are specific to the roof types and number of stories, etc. Since the policy data do not provide this level of specificity, weighted matrices are used instead, where the weights are the proportion of different roof types in given region as determined by a survey of the building blocks and exposure data. The vulnerability matrices are used as input in the actuarial model.

The starting point for the computations of personal residential losses is the vulnerability matrix with its set of damage intervals and associated probabilities. Appropriate vulnerability matrices are applied separately for building structure, content, appurtenant structure, and ALE. Once the matrix is selected, for a given wind speed, for each of the midpoint of the damage intervals, the ground up loss is computed, the appropriate deductibles and limits are applied, and the loss net of deductible is calculated. More specifically, for each damage outcome the damage ratio is multiplied by insured value to get dollar damages, the deductible is deducted, and net of deductible loss is estimated, subject to the constraints that net loss is ≥ 0 and $\leq$ limit – deductible. Percentage deductibles are converted into dollar amounts. Both the replacement cost and actual cash value are generally assumed to equal the coverage limit. Furthermore, if there are multiple hurricanes in a year in the stochastic set, the wind deductibles are applied to the first hurricane, and any remaining amount is then applied to the second hurricane. If none remains then the general peril deductible can be applied.

The net of deductible loss is multiplied by the probability in the corresponding cell to get the expected loss for the given damage ratio. The results are then averaged across the possible

damages for the given wind speed. Next, the wind probability weighted loss is calculated to produce the expected loss for the property. The expected losses are then adjusted by the appropriate expected demand surge factor.

In the case of low-rise commercial residential structures, the expected damage ratios (EDR) are derived from the vulnerability curves for the maximum wind in the given storms. The EDRs are multiplied by the respective coverage limits to produce the expected ground up building damage value (EDV^B), and expected ground up content damage value (EDV^C) for the storm. The deductible is then applied to these damage values on a pro-rata basis to generate the net of deductible expected losses. The process is repeated across all the storms in the stochastic set to produce the average loss for the policy. The expected losses are then adjusted by the appropriate expected demand surge factor.

In the case of mid-high rise commercial residential buildings, the vulnerability component produces, for a given storm (or given vertical maximum wind profile) and across all the floors in the building, the total expected cost of damage to the openings (TECDO) and the expected interior damage ratio (EIDR). The EIDR is then multiplied by the fraction of the coverage limit corresponding to the value of the interior and added to the TECDO to produce the expected building damage value (EDV^B). The expected content damage value (EDV^C) is produced by multiplying a fraction of the EIDR by the content coverage limit. The deductible is then applied on a pro-rata basis to generate the expected loss for the storms. The process is repeated across all storms to produce the average loss for the policy. The expected losses are then adjusted by the appropriate expected demand surge factor.

For commercial residential policies, if there are multiple risks (multiple structures) within the policy, the default is to apply the deductible at the risk level. The percentage deductible is applied to each risk based on their individual limit. If information is so available, then deductible is applied at the policy level.

The demand surge factors are estimated by a separate model and applied appropriately to each hurricane in the stochastic set. The surge factors for structures are a function of the size of statewide storm losses and are produced separately for the different regions in Florida. The surge factors for content and ALE are functionally related to the surge factor for structure. To estimate the impact of demand surge on the settlement cost of structural claims following a hurricane, data from 1992 to 2007 on a quarterly construction cost index produced by Marshall & Swift/Boeckh are used. The approach to estimating structural demand surge was to examine the index for specific regions impacted by one or more hurricanes since 1992. From the history of the index we projected what the index would have been in the period following the storm had no storm occurred. Any gap between the predicted and actual index was assumed to be due to demand surge. In total ten storm–region combinations are examined. From these ten observations of structural demand surge the functional relationship is generalized.

After the losses are adjusted for demand surge, they are summed across all structures of the type in the grid and also across the grids to get expected aggregate portfolio loss. The model can process any combination of policy type, construction type, deductibles, coverage limits, etc. The model output reports include separate loss estimates for structure, content, appurtenant structure, and ALE. These losses are also reported by construction type (e.g., masonry, frame,

manufactured homes), by county or ZIP Code, by policy form (e.g., HO-3, HO-4, etc.), by rating territory, and combinations thereof.

Another function of the actuarial algorithms is to produce estimates of the probable maximum loss for various return periods. The PML is produced non-parametrically using order statistics of simulated annual losses. Suppose the model produces N years of simulated annual losses. The annual losses L are ordered in increasing order so that $L(1) \leq L(2) \leq \ldots \leq L(N)$. For a return period of Y years, let p = 1-1/Y. The corresponding PML for the return period Y is the pth quantile of the ordered losses. Let k = (N)*p. If k is an integer, then the estimate of the PML is the kth order statistic, L(k), of the simulated losses. If k is not an integer, then let k* = the smallest integer greater than k, and the estimate of the pth quantile is given by L(k*).

COMPUTER SYSTEM ARCHITECTURE

The FPHLM is a large-scale system that is designed to store, retrieve, and process a large amount of historical and simulated hurricane data. In addition, intensive computation is supported for hurricane damage assessment and insured loss projection. To achieve system robustness and flexibility, a three-tier architecture is adopted and deployed in our system. It aims to solve a number of recurring design and development problems and make the application development work easier and more efficient. The computer system architecture consists of three layers: the user interface layer, the application logic layer, and the database layer.
The interface layer offers the user a friendly and convenient user interface to communicate with the system. To offer greater convenience to the users, the system is prototyped on the web so that the users can access the system with existing web-browser software.

The application logic layer activates model logic based on the functionality presented to the user, processes data, and controls the information flow. This is the middle tier in the computer system architecture. It aims to bridge the gap between the user interface and the underlying database and to hide technical details from the users.

The database layer is responsible for data modeling to store, index, manage, and model information for the application. Data needed by the application logic layer are retrieved from the database, and the computational results produced by the application logic layer are stored back to the database.

Software, Hardware, and Program Structure

The system is primarily a web-based application that is hosted on a Tomcat web application server. The backend server environment is Linux and the server side scripts are written in Java Server Pages (JSP) and JavaBeans. Backend probabilistic calculations are coded in C++ using the IMSL library and called through Java Native Interface (JNI). The system uses a PostgreSQL database that runs on a Linux server. Server side software requirements are the IMSL library CNL 5.0, JDBC 3, JNI 1.3.1, and JDK 1.5.

The end-user workstation requirements are minimal. The recommended web browsers are Internet Explorer 8.0 running on Windows XP or Internet Explorer 9.0 running on Windows 7. However,

other modern web browsers such as Mozilla Firefox running on either Windows or Linux should also deliver optimal user experience. Typically, the manufacturer's minimal set of features for a given web browser and operating system combination is sufficient for an optimal operation of the application.

Translation from Model Structure to Program Structure

The FPHLM uses a component-based approach in converting from model to program structure. The model is divided into the following components or modules: Storm Forecast Module, Wind Field Module, Damage Estimation Module, and Loss Estimation Module. Each of these modules fulfills its individual functionality and communicates with other modules via well-defined interfaces. The architecture and program flow of each module are defined in its corresponding use case document following software engineering specifications. Each model element is translated into subroutines, functions, or class methods on a one-to-one basis. Changes to the models are strictly reflected in the software code.

3. ***Provide a flow diagram that illustrates interactions among major model components.***

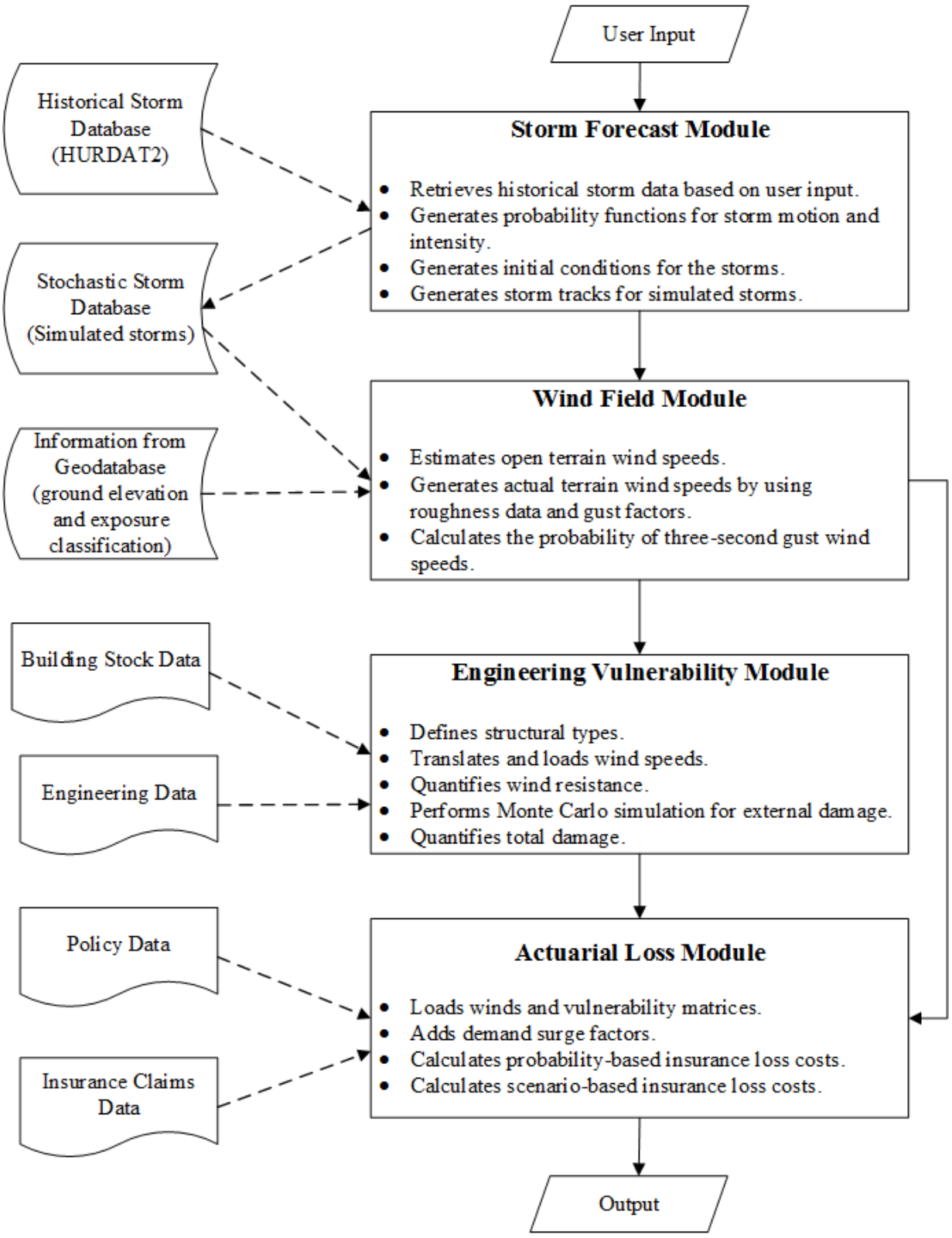


Figure 19. Flow diagram of the computer model.

4. ***Provide a comprehensive list of complete references pertinent to the submission by standard grouping, according to professional citation standards.***

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Iman, R. L., Johnson, M. E., & Schroeder, T. (2000b). Assessing Hurricane Effects. Part 2. Uncertainty Analysis. *Reliability Engineering & System Safety, 78*(2), 147-155.

Lin, L. I. (1989). A concordance correlation coefficient to evaluate reproducibility. *Biometrics, 45*, 255-268.

Reiss, R. D. (1989). *Approximate Distributions of Order Statistics with Applications to Nonparametric Statistics.* New York: Springer Verlag.

Tamhane, A. C., & Dunlop, D. (2000). *Statistics and Data Analysis.* New York: Prentice Hall.

Relevant Web Sites

Applied Insurance Research, Inc. (AIR) page.
http://www.airboston.com_public/html/rmansoft.asp

Applied Research Associates, Inc. (ARA) page.
http://www.ara.com/risk_and_reliability_analysis.htm

ARIS Reference.
http://www.idsscheer.com/international/english/products/aris_design_platform/50324

CIMOSA Reference. http://cimosa.cnt.pl

EQECAT home page. http://www.eqecat.com/

FEMA hurricanes page. http://www.fema.gov/hazards/hurricanes

Florida Water Management District Land Use Data, Statewide 2004-2011, as compiled by the Florida State Department of Environmental Protection:
http://www.dep.state.fl.us/gis/datadir.htm
Actual data is at
http://publicfiles.dep.state.fl.us/otis/gis/data/STATEWIDE_LANDUSE_2004_2011.zip

Global Ecosystems Database (GED). http://www.ngdc.noaa.gov/seg/fliers/se- 2006.shtml

HAZUS Home. http://www.hazus.org/

HAZUS Overview. http://www.nibs.org/hazusweb/verview/overview.php

HAZUS manuals page, http://www.fema.gov/hazus/li_manuals.shtm

HURDAT data. http://www.aoml.noaa.gov/hrd/hurdat/Data Storm.html

IMSL Mathematical & Statistical Libraries. http://www.vni.com/products/imsl

Java Native Interface. http://java.sun.com/docs/books/tutorial/native1.1/

Java Server Pages (TM) Technology. http://java.sun.com/products/jsp/

National Hurricane Center. http://www.nhc.noaa.gov/

NIST Aerodynamic Database - http://fris2.nist.gov/winddata

NOAA Coastal Services Center. http:www.csc.noaa.gov

NOAA EL Nino Page. http://www.elnino.noaa.gov/

NOAA LA Nina Page. http://www.elnino.noaa.gov/lanina.html

PHRLM Manual. http://www.cis.fiu.edu/hurricaneloss

RAMS: Regional Atmospheric Modeling System. http://rams.atmos.colostate.edu/

R.L. Walko, C.J. Tremback, "RAMS: regional atmospheric modeling system, version 4.3/4.4 - Introduction to RAMS 4.3/4.4."
http://www.atmet.com/html/docs/rams/ug44-rams-intro.pdf

RMS home page. http://www.rms.com

The JDBC API Universal Data Access for the Enterprise.
http://java.sun.com/products/jdbc/overview.html

The Interactive Data Language. http://www.rsinc.com/idl/

Track of hurricane Andrew (1992) (Source from NOVA).
http://www.pbs.org/newshour/science/hurricane/facts.html

Tropical cyclone heat potential: http://www.aoml.noaa.gov/phod/cyclone/data/

The Ptolemy Java Applet package.
http://ptolemy.eecs.berkeley.edu/papers/99/HMAD/html/plotb.html

5. ***Provide the following information related to changes in the model from the previously accepted submission to the initial submission this year.***

A. ***Model changes:***

1. A summary description of changes that affect the personal or commercial residential loss costs or probable maximum loss levels,

Meteorological Component

- We updated to a recent version of ~~HURDAT~~HURDAT2 (4/1/2014) which includes storms up through the 2013 season.

- We updated the land use/land cover data set using MRLC NLCD 2011 and the Statewide 2004-2011 Florida Water Management District data set as per Standard M-4.

- We updated the zip code database to the December, 2013 ZIP code boundaries as per Standard G-3.

- The *Rmax* database was revised to include recent storms and revisions to historical storms.

Vulnerability Component

a. The changes in the low-rise CR model include:
   - Projectile count increase in debris impact model
   - Interior pressure sharing between attic and top floor changed
   - Interior pressure calculation in the attic space due to sheathing loss changed
   - Change in the soffit damage computation
   - Reductions in the pressure coefficient (C_p) multiplier
   - Modification of the masonry wall area failure function and its differentiation between unreinforced and reinforced masonry.
   - Changes in the rain admittance factor (RAF) values and incorporation of the new surface run-off coefficient
   - Replacement of the directionality factor (f_{sim}) with a more sophisticated directionality scheme
   - The statistics used to weigh the low-rise CR vulnerability matrices were updated.
   - Modeling of gable end damage for masonry models.
   - Rain penetration model: correction to the dimension of gable end upstream run-off area A_{SR}.
   - Alternative soffit model.

b. The changes in the damage evaluation of Mid/high-rise CR model include
    - An additional volume of water penetration was modeled at the upper story of Mid/high-rise CR model

2. A list of all other changes, and

No other changes are reported.

3. The rationale for each change.

Meteorological Component

- Change made to update to ~~the latest HURDAT~~a recent version of HURDAT2 (4/1/2014) and to take advantage of new observations of *Rmax* that have recently become available for storms that have occurred up to the 2012 hurricane season.

- Updated centroid locations as per Standard G-3 and terrain roughness as per Standard M-4 B.

Vulnerability Component

For LR CR:

- The projectile count in the debris impact model was modified to better reflect P(projectile impact | story height).
- The interior pressure sharing mechanism between attic and top floor was modified to reflect the change in internal pressure sharing due to top floor breaches.
- Interior pressure calculation in the attic space was changed to delineate internal pressure contributions from the flow regimes associated with windward and leeward roof sheathing damage
- ~~The location of the soffit damage routine was modified to better reflect the influence of internal pressure on soffit failure.~~ The loading on the soffits was modified to better reflect current state-of-knowledge.
- The pressure coefficient (C_p) was modified to reduce an overly-conservative extreme load assumption, and to bring the LR CR model into compliance with the personal residential model
- In the case of masonry structures, the damage simulations give the exceedance of the ultimate moment and shear capacities of the masonry wall for both unreinforced (weak models) and reinforced masonry (medium and strong models). Heuristic curves relate the % of masonry capacity exceedance to a % of wall failure, for both shear or out of plane bending. The original heuristics used in v5.0 of the FPHLM did not distinguish between unreinforced and reinforced masonry. A set of heuristics was implemented in the new version of the vulnerability model, which takes into account the differences in behavior

between unreinforced and reinforced masonry. Further modifications of the masonry wall area failure function were implemented to ensure a greater differentiation between unreinforced and reinforced masonry, and to ensure that the strong masonry models were less vulnerable than the strong timber models.

- The rain admittance factor values (RAF) were modified, and the new concept of surface run-off coefficient (SRC) was introduced in the rain intrusion model. This was done to incorporate into the FPHLM the latest test results from the FIU Wall of Wind (Baheru et al., 2014a, 2014b).
- The directionality factor (f_{sim}) in the rain intrusion model was replaced with a directionality scheme, which accounts for the difference in horizontal rain with the storm rotation and its influence on the water intrusion though breaches and defects.
- The statistics used to weigh the vulnerability matrices were updated to take into account the latest information from the tax appraiser databases.
- The modeling of gable end damage for masonry models was modified in v6.1 to correct an omission in v6.0.
- In the rain penetration model, the dimension of the gable end upstream run-off area ASR was corrected for 2- and 3-story models.
- An alternative soffit model was implemented to incorporate the latest data on soffits.

For MH CR

- An additional volume of water penetration is modeled at the upper story, because increased water penetration through possible roof cover damage as well as roof defects or ventilation ducts could happen in the upper floors, which would then trickle down to the lower stories.

B. ***Percentage difference in average annual zero deductible statewide loss costs based on the 2007 Florida Hurricane Catastrophe Fund's aggregate personal and commercial residential exposure data found in the file named "hlpm2007c.exe" for:***

1. All changes combined, and

Overall statewide percentage change in loss cost is ~~5.28~~5.18% decrease.

2. Each individual model component change.

Meteorological Component

The estimated change in statewide loss costs due to the updated *Rmax* and HURDAT is a ~~6.13~~6.70% decrease. The estimated change in statewide loss costs due to updated ZIP code centroids and updated terrain roughness is a ~~0.79~~0.73% increase. The overall change in loss costs resulting from meteorological component is ~~5.32~~5.96% decrease.

Vulnerability Component

The combined statewide percentage change in loss costs due to all the changes in the vulnerability model is an approximate ~~0.34~~0.25% ~~increase~~decrease.

It is not possible to apply each of the individual model component changes reported above by themselves into version 6.0 to isolate any change in losses due to each change alone. The best and most meaningful compromise was to evaluate the influence on the overall losses of all the changes together in version 6.0.

C. ***Color-coded maps by county reflecting the percentage difference in average annual zero deductible statewide loss costs based on the 2007 Florida Hurricane Catastrophe Fund's aggregate personal and commercial residential exposure data found in the file named "hlpm2007c.exe" for each model component change.***

D. ***Color-coded map by county reflecting the percentage difference in average annual zero deductible statewide loss costs based on the 2007 Florida Hurricane Catastrophe Fund's aggregate personal and commercial residential exposure data found in the file named "hlpm2007c.exe" for all model components changed.***

1. Between the previously accepted submission and the revised submission,

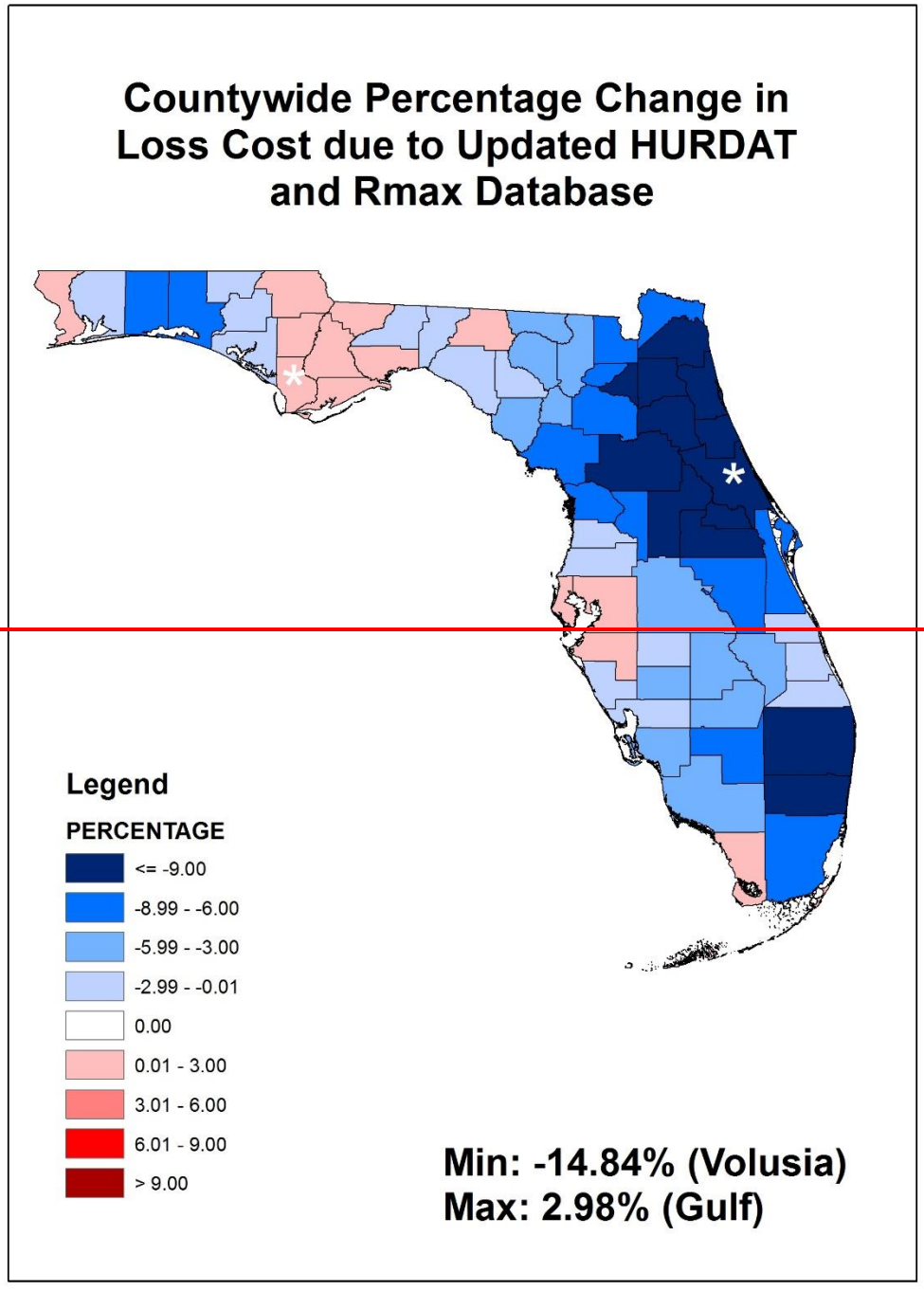
2. Between the initial submission and the revised submission, and

3. Between any intermediate revisions and the revised submission.

6. Provide a list and description of any potential interim updates to underlying data relied upon by the model. State whether the time interval for the update has a possibility of occurring during the period of time the model could be found acceptable by the Commission under the review cycle in this *Report of Activities*.

County tax appraiser databases: some databases might be updated or become available during the period of time the model could be found acceptable by the Commission under the review cycle in this *Report of Activities*.

Cost estimates from different sources like RSMeans or local contractors might be updated during the period of time the model could be found acceptable by the Commission under the review cycle in this Report of Activities.


Countywide Percentage Change in Loss Cost due to Updated HURDAT and Rmax Database
Legend
PERCENTAGE
<= -9.00
-8.99 - -6.00
-5.99 - -3.00
-2.99 - -0.01
0.00
0.01 - 3.00
3.01 - 6.00
6.01 - 9.00
> 9.00
Min: -14.84% (Volusia)
Max: 2.98% (Gulf)

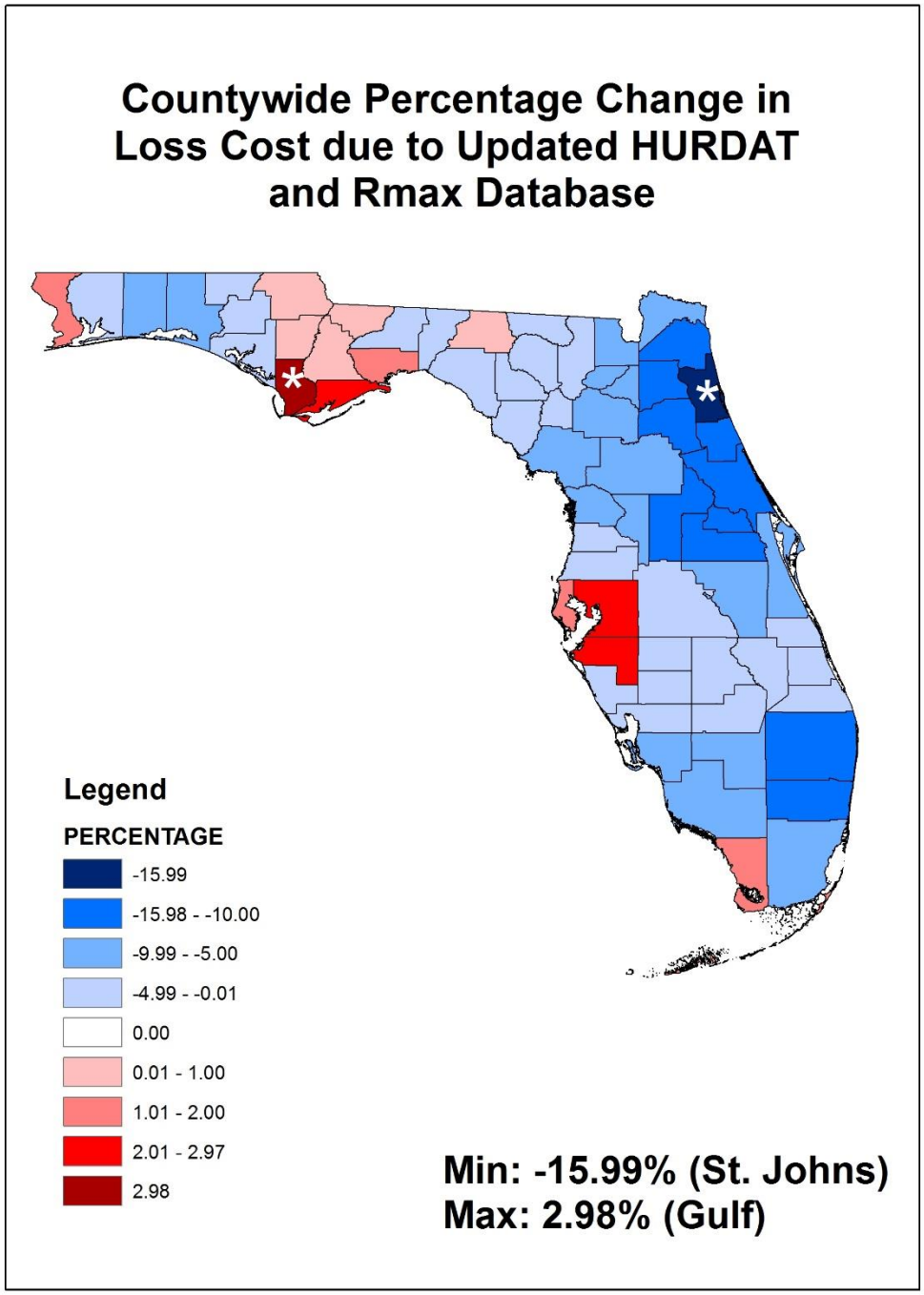
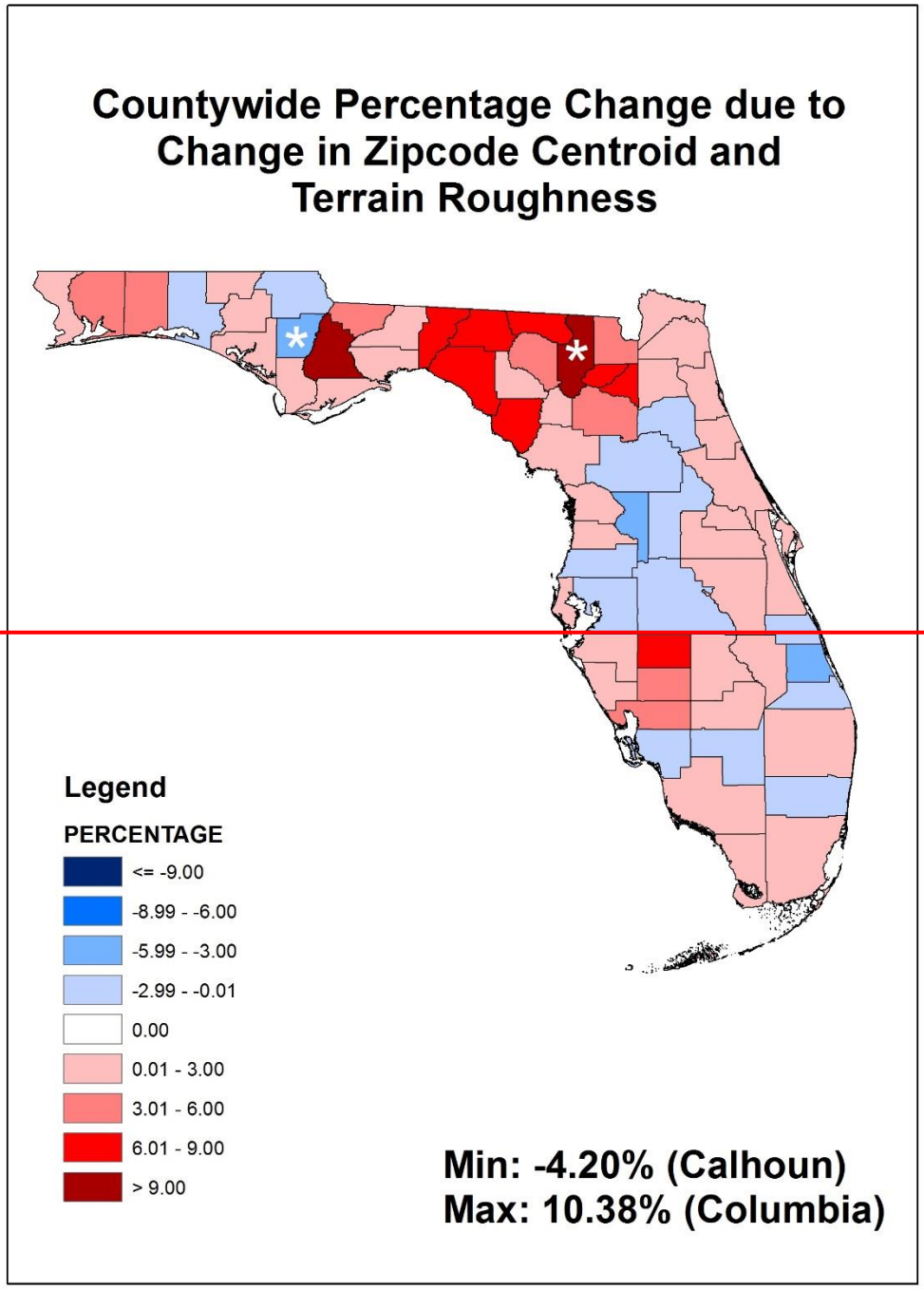


Figure 20. Countywide Percentage Change in Loss Cost due to Updated HURDAT and Rmax Database


Countywide Percentage Change due to Change in Zipcode Centroid and Terrain Roughness
Legend
PERCENTAGE
<= -9.00
-8.99 - -6.00
-5.99 - -3.00
-2.99 - -0.01
0.00
0.01 - 3.00
3.01 - 6.00
6.01 - 9.00
> 9.00
Min: -4.20% (Calhoun)
Max: 10.38% (Columbia)

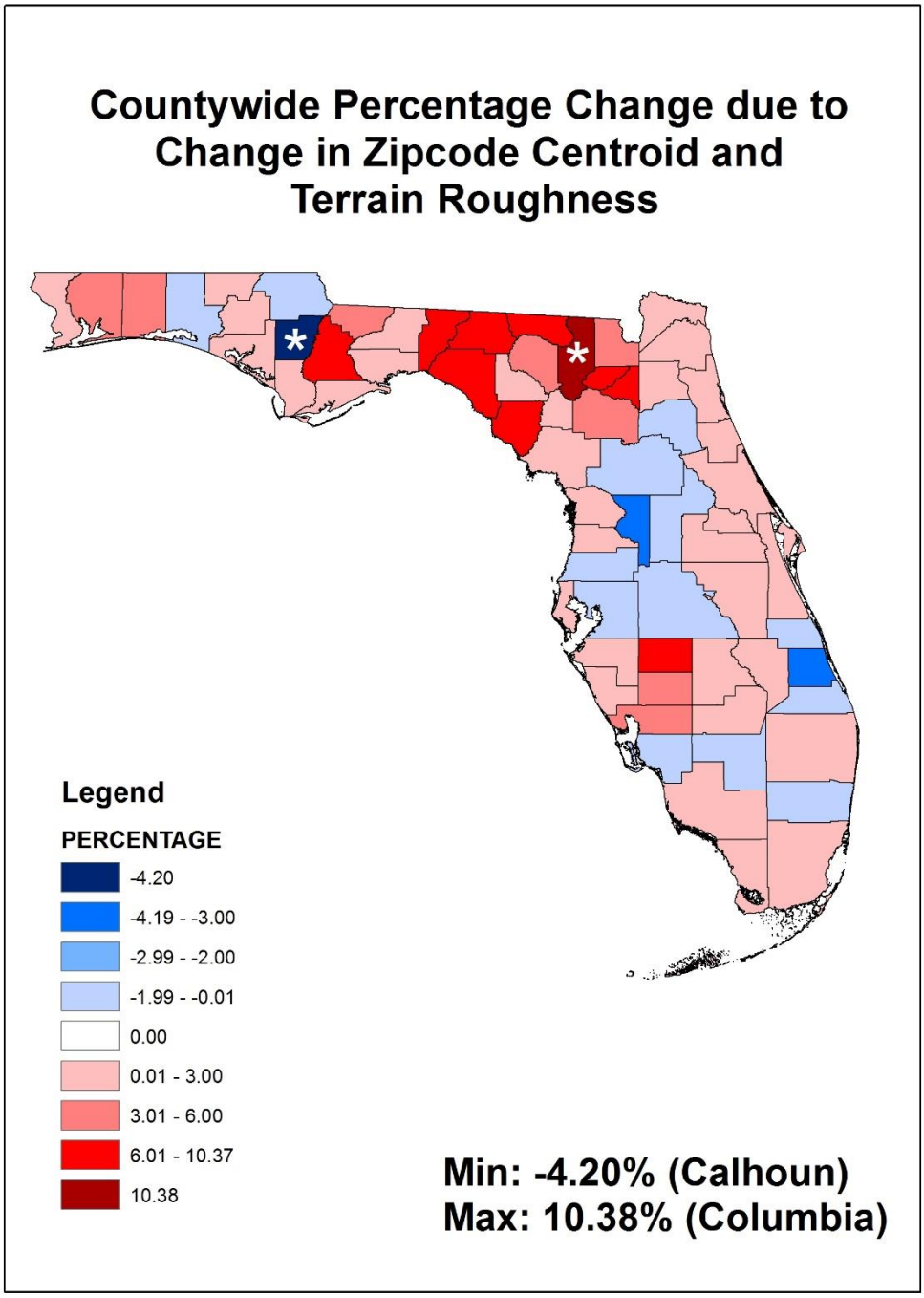
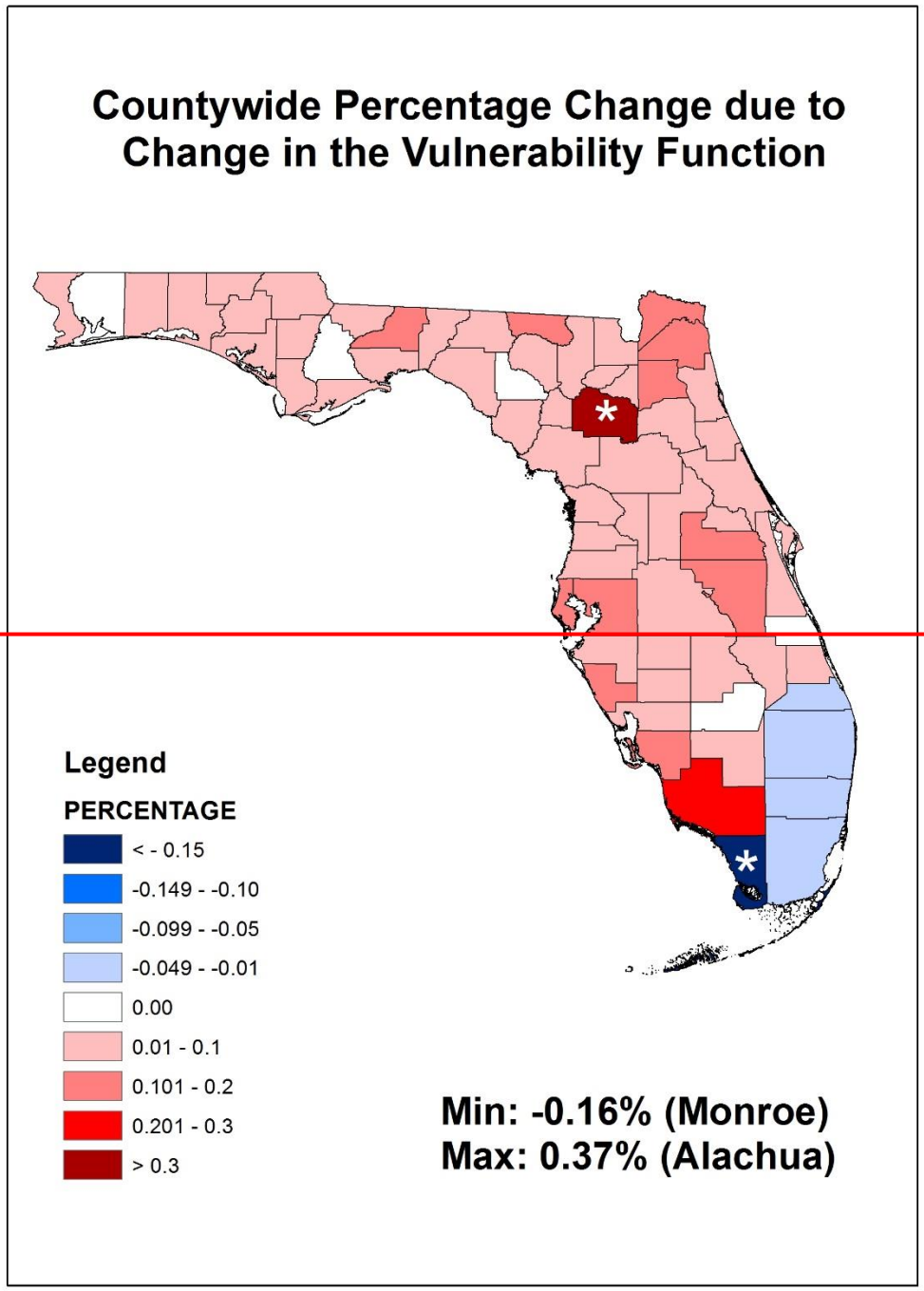


Figure 21. Countywide Percentage Change due to Change in Zipcode Centroid and Terrain Roughnes


Countywide Percentage Change due to Change in the Vulnerability Function
Legend
PERCENTAGE
< - 0.15
-0.149 - -0.10
-0.099 - -0.05
-0.049 - -0.01
0.00
0.01 - 0.1
0.101 - 0.2
0.201 - 0.3
> 0.3
Min: -0.16% (Monroe)
Max: 0.37% (Alachua)

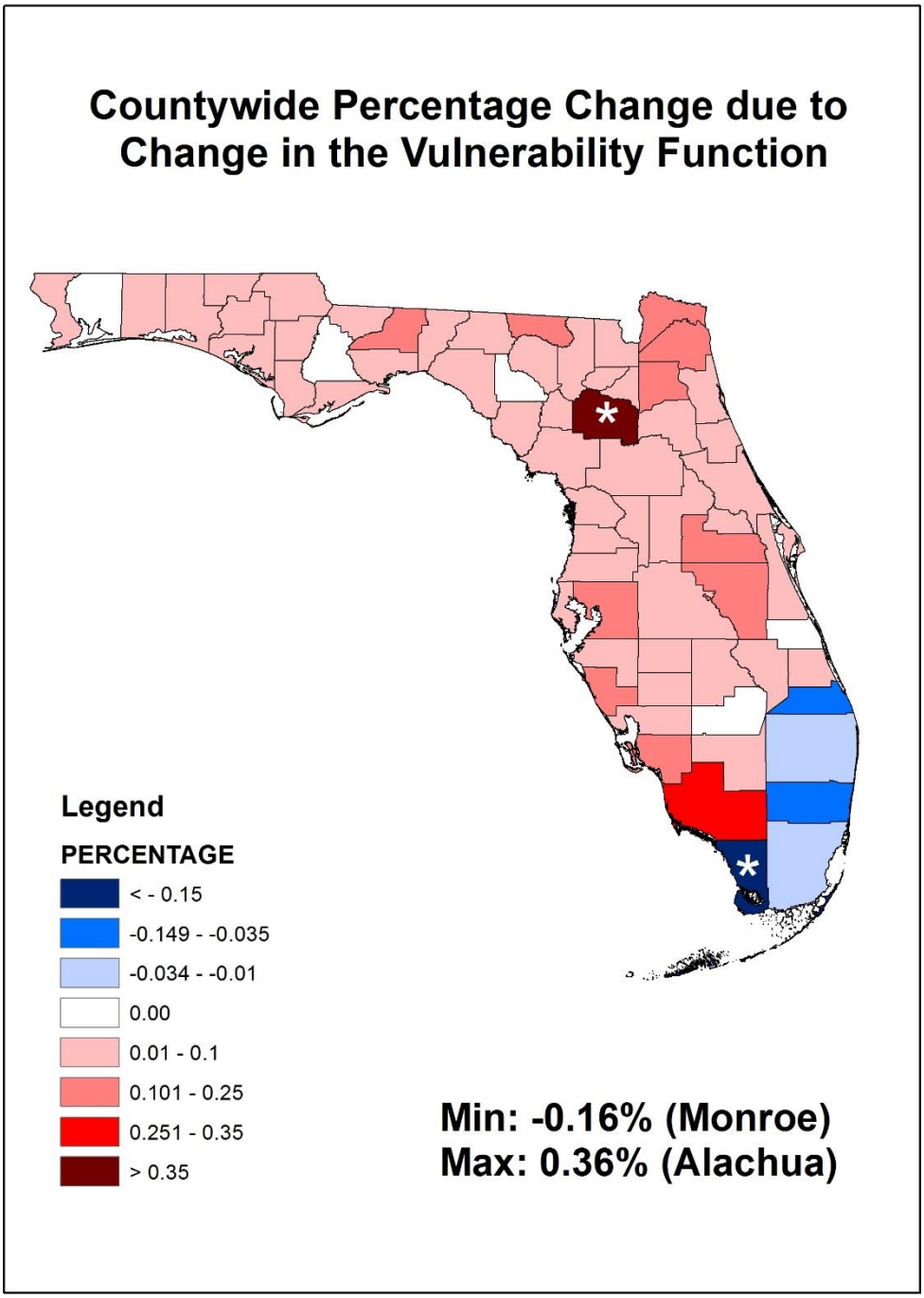


Figure 22. Countywide Percentage Change due to Change in the Vulnerability Function

G-2 Qualifications of Modeling Organization Personnel and Consultants

A. ***Model construction, testing, and evaluation shall be performed by modeling organization personnel or consultants who possess the necessary skills, formal education, and experience to develop the relevant components for hurricane loss projection methodologies.***

The model was developed, tested, and evaluated by a multi-disciplinary team of professors and experts in the fields of meteorology, wind and structural engineering, computer science, statistics, finance, economics, and actuarial science. The experts work primarily at Florida International University, Florida Institute of Technology, Florida State University, University of Florida, Hurricane Research Division of NOAA, and University of Miami.

B. ***The model and model submission documentation shall be reviewed by either modeling organization personnel or consultants in the following professional disciplines: structural/wind engineering (licensed Professional Engineer), statistics (advanced degree), actuarial science (Associate or Fellow of Casualty Actuarial Society), meteorology (advanced degree), and computer/information science (advanced degree). These individuals shall certify Forms G-1 through G-6 as applicable.***

The model has been reviewed by modeler personnel and consultants in the required professional disciplines. These individuals abide by the standards of professional conduct as adopted by their profession.

Disclosures

1. ***Organization Background***

   A. ***Describe the ownership structure of the modeling organization. Describe affiliations with other companies and the nature of the relationship, if any. Indicate if your organization has changed its name and explain the circumstances.***

The model was developed independently by a multi-disciplinary team of professors and experts. The lead university is the Florida International University. The model was commissioned by the Florida Office of Insurance Regulation.

   B. ***If the model is developed by an entity other than a modeling company, describe its organizational structure and indicate how proprietary rights and control over the model and its critical components is exercised. If more than one entity is involved in the development of the model, describe all involved.***

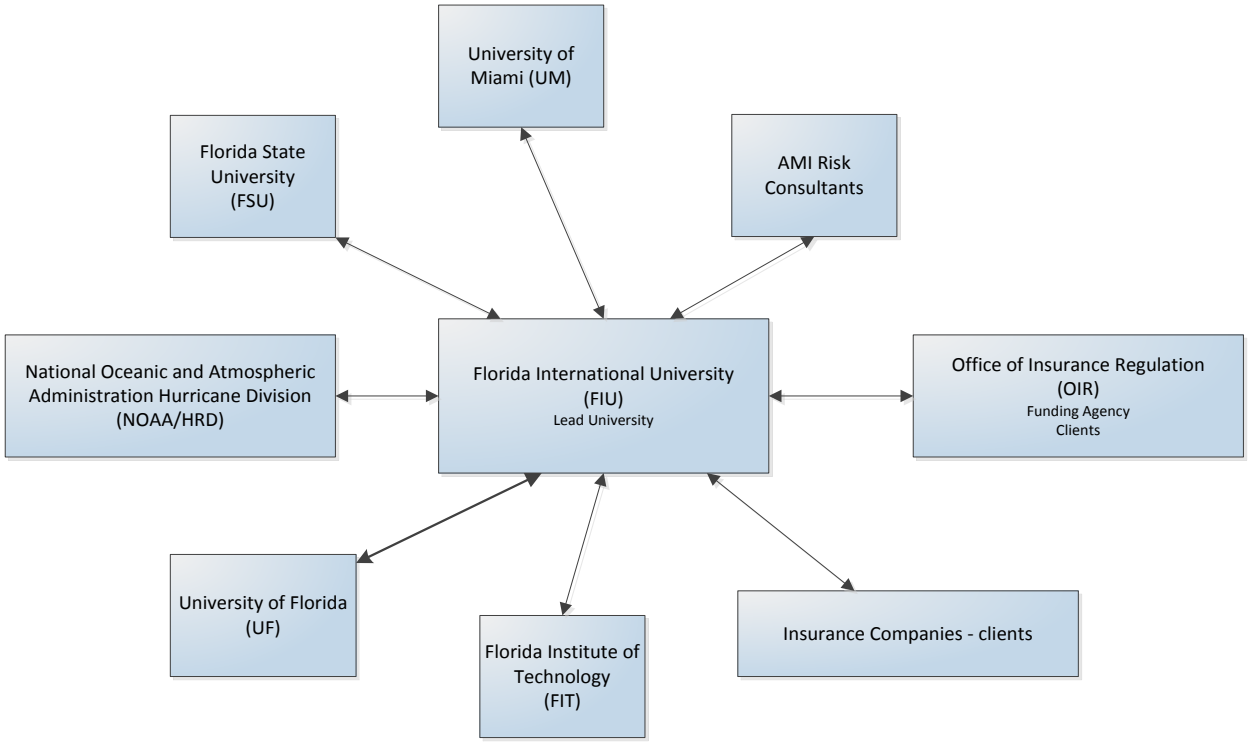


Figure 23. Organizational structure.

The Florida Office of Insurance Regulation (OIR) contracted and funded Florida International University to develop the Florida Public Hurricane Loss Model. The model is based at the Laboratory for Insurance, Financial and Economic Research, which is part of the International Hurricane Research Center at Florida International University. The OIR did not influence the development of the model. The model was developed independently by a team of professors, experts, and graduate students working primarily at Florida International University, Florida Institute of Technology, Florida State University, University of Florida, Hurricane Research Division of NOAA, University of Miami, and AMI Risk Consultants. The copyright for the model belongs to OIR.

C. ***If the model is developed by an entity other than a modeling company, describe the funding source for the model.***

The model was funded by the state legislature at the request of the Florida Office of Insurance Regulation.

D. ***Describe the modeling organization's services.***

Until 2008 the modeler provided services to only one major client, the FL-OIR. Effective January 2009 the modeler is providing services to the firms and organizations in the insurance and reinsurance industries. It has expanded the infrastructure and computational capacity to handle the added load.

The first version of the model was completed in May 2005 and was based on the knowledge and the limited data available prior to the 2004–2005 hurricane seasons. It was not used for purposes

of estimating loss costs for insurance company exposures. Essentially, it was an internal model that was never implemented.

The next version of the model was developed upon the acquisition of a limited amount of meteorological, engineering, and insurance claim data from the 2004–2005 hurricane events and was implemented in March 2006. This version was used to process the insurance company data on behalf of the Florida Office of Insurance Regulation.

In summer 2007 a revised and updated version of the model, 2.6, was accepted by the Florida Commission on Hurricane Loss Projection Methodology and put to immediate use. Another revised and updated version, 3.0, was accepted by the Commission in June 2008. The next updated version of the model was 3.1, which was accepted by the Commission in June 2009. This was followed by version of the model was 4.1, which was accepted by the Commission in August 2011.The latest updated version of the model is 5.0, which was accepted by the Commission in July 2013.

E. ***Indicate if the modeling organization has ever been involved directly in litigation or challenged by a statutory authority where the credibility of one of its U.S. hurricane model versions for projection of loss costs or probable maximum loss levels was disputed. Describe the nature of each case and its conclusion.***

None.

2. ***Professional Credentials***

A. Provide in a chart format (a) the highest degree obtained (discipline and university), (b) employment or consultant status and tenure in years, and (c) relevant experience and responsibilities of individuals currently involved in the acceptability process or in any of the following aspects of the model:

1. Meteorology
2. Statistics
3. Vulnerability
4. Actuarial Science
5. Computer Science

See below.

Table 10. Professional credentials.

Key Personnel	Degree/ Discipline	University	Employment Status	Tenure	Experience
Meteorology:					
Dr. Mark Powell	Ph.D. Meteorology	Florida State University	Senior Atmospheric Scientist HRD/NOAA	35	Meteorology wind field model
Dr. Steve Cocke	Ph.D. Physics	Univ. Texas Austin	Scholar/Scientist FSU, Dept of Meteorology	18	Meteorology track, intensity, roughness models
Bachir Annane	M.S. Meteorology, M.S. Mathematics	Florida State University	Meteorologist, Univ. of Miami	20	Meteorology

Key Personnel	Degree/ Discipline	University	Employment Status	Tenure	Experience
Neal Dorst	B.S. Meteorology	Florida State University	Meteorologist, HRD/NOAA	30	Meteorology
Statistics:					
Dr. S. Gulati	Ph.D. Statistics	University of South Carolina	Professor, Statistics , FIU	23	Statistics tests and nonparametric analysis
Dr. B. M. Golam Kibria	Ph.D. Statistics	University of Western Ontario	Professor of Statistics at FIU	14	Statistics testing and sensitivity analysis
Engineering:					
Dr. Jean-Paul Pinelli	Ph.D. Civil Engineering	Georgia Tech	Professor, CE Florida Institute of Technology	18	Wind engineering, vulnerability functions
Dr. Kurt Gurley	Ph.D. Civil Engineering	University of Notre Dame	Associate Professor, CE University of Florida	15	Wind engineering, simulations
Timothy Johnson	B.S. Civil Engineering	Florida Institute of Technology	M.S. candidate (FIT)	4	Wind engineering, vulnerability functions
Johann Weekes	M.S. Civil Engineering	University of Florida	Ph.D. Candidate (UF Civil)	8	Wind and structural engineering
Actuarial/Finance:					
Dr. Shahid Hamid Project Manager, PI	Ph.D. Economics (Financial), CFA	University of Maryland	Professor of Finance Florida International University	26	Insurance and finance
Gail Flannery	FCAS, Actuary	CAS	VP, AMI Risk Consultants	29	Reviewer, demand surge, actuarial analysis
Aguedo Ingco	FCAS, Actuary	CAS	President, AMI Risk Consultants	39	Reviewer, demand surge
Nino Joseph Paz	FCAS, Actuary	University of Philippines-Diliman	Actuarial supervisor, AMI Risk Consultants	3	Actuarial consulting
Computer Science					
Dr. Shu-Ching Chen	PhD Electrical and Computer Engineering	Purdue University	Professor of Computer Science at FIU	14	Software and database development
Dr. Mei-ling Shyu	PhD Electrical and Computer Engineering	Purdue University	Professor of Electrical and Computer Engineering at UM	14	Software quality assurance
Raul Garcia	BS Computer Science	Florida International University	MS in Computer Science student at FIU	4	Software and database development
Diana Machado	BS Computer Science	Florida International University	MS in Computer Science student at FIU	3	Software and database development
Hsin-Yu Ha	MS Computer Science	Florida International University	Ph.D. candidate at FIU	7	Software and database development
Yimin Yang	MS Computer Science	Florida International University	Ph.D candidate at FIU	5	Software and database development
Fausto Fleites	PhD Computer Science	Florida International University	Consultant	12	Software and database development
Haiman Tian	MS Computer Engineering	Florida International University	Ph.D. in Computer Science student at FIU	1	Software and database development
Samira Pouyanfar	MS Computer Engineering	Sharif University of Technology	Ph.D. in Computer Science student at FIU	1	AI, software and database development
Yilin Yan	MS Computer Science	National Taiwan Ocean University	Ph.D. Student at University of Miami	2	Software design and testing, data processing

Key Personnel	Degree/ Discipline	University	Employment Status	Tenure	Experience
Qinghua Liang	BS Mathematics and Applied Mathematics	Ningbo University	ME Computer Engineering at FIU	1	Software development Image Processing Interaction Design
Junjie Hou	BE Electrical and Information Engineering	China University of Geoscience	ME in Computer Engineering student at FIU	1	Software development and Big Data analysis
Wenbo Wang	BS Computer Science	North China University of Technology	MS in Information Technology student at FIU	1	Software development and project management
Yuexin Liu	BS Mathematics and Applied Mathematics	Dianli University (China)	ME Computer Engineering at FIU	1	Software development
Xiaoyu Dong	BS Mathematics and Applied Mathematics	Hebei University of Science and Technology (China)	MS in Computer Engineering student at FIU	1	Software development and Data processing
Jing Chang	BS Software Engineering	Hebei University of Technology	MS in Information Technology student at FIU	1	Software development and Data processing
Daniel Lopez	Computer Science undergraduate student	Florida International University	Undergraduate CS student at FIU	1	Data processing and software development

B. ***Identify any new employees or consultants (since the previous submission) working on the model or the acceptability process.***

Haiman Tian, Samira Pouyanfar, Yilin Yan, Qinghua Liang, Junjie Hou, Wenbo Wang, Yuexin Liu, Xiaoyu Dong, Jing Chang, Daniel Lopez.

C. ***Provide visual business workflow documentation connecting all personnel related to model design, testing, execution, maintenance, and decision-making.***

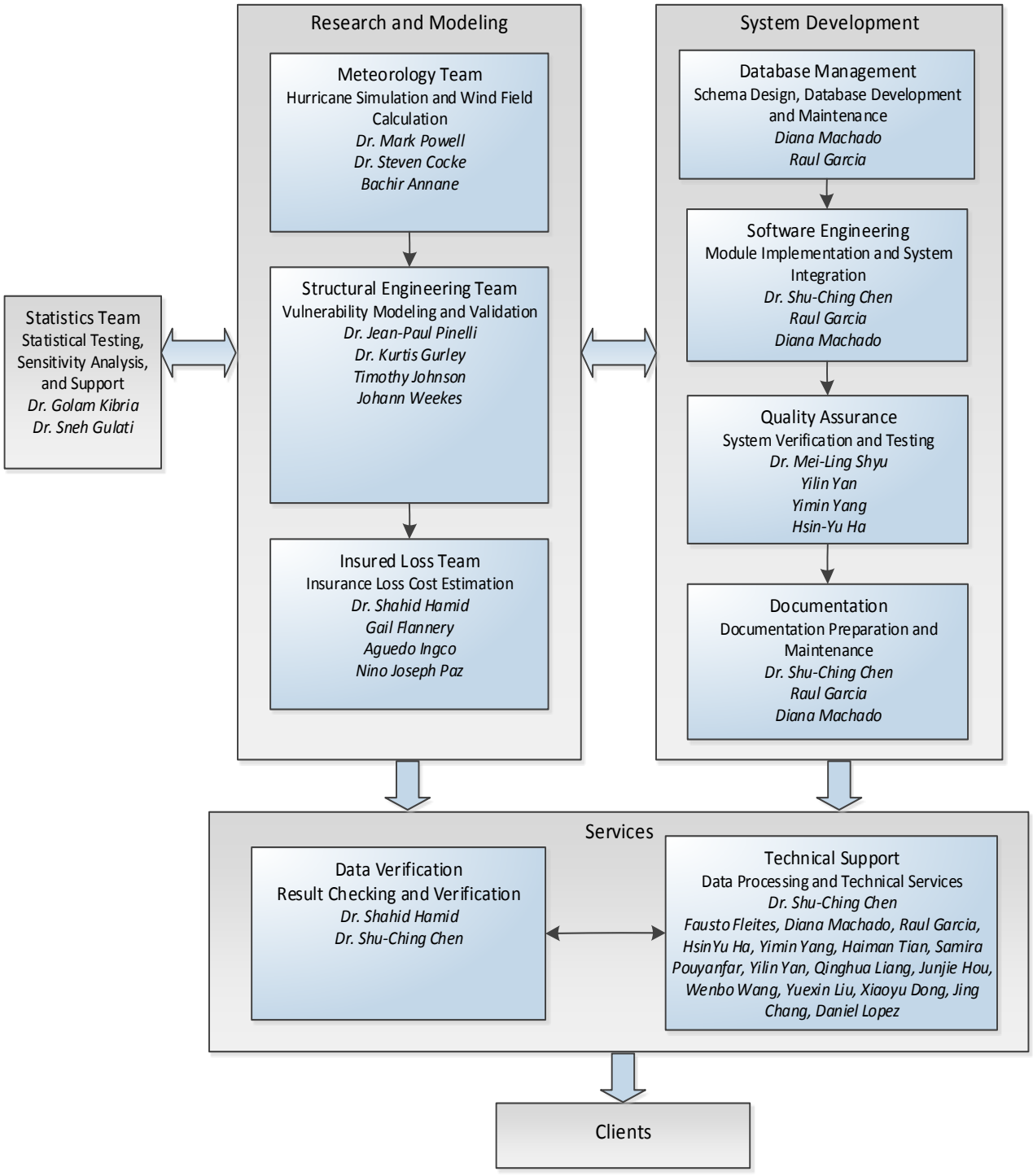


Figure 24. Florida Public Hurricane Loss Model workflow.

D. ***Indicate specifically whether individuals listed in A. and B. are associated with the insurance industry, a consumer advocacy group, or a government entity, as well as their involvement in consulting activities.***

Dr. Mark Powell worked for Hurricane Research Division of NOAA until March 2014 and has since been an independent consultant. Neal Dorst works for the Hurricane Research Division of NOAA.

3. ***Independent Peer Review***

A. Provide reviewer names and dates of external independent peer reviews that have been performed on the following components as currently functioning in the model:

1. Meteorology
2. Statistics
3. Vulnerability
4. Actuarial Science
5. Computer Science

Dr. Gary Barnes, Professor of Meteorology at University of Hawaii, performed the external review of the meteorology component in December 2006. The current version was reviewed by modeler personnel.

Gail Flannery, FCAS, and Aguedo Ingco, FCAS, actuaries and vice president and president, respectively, of AMI Risk Consultants in Miami, performed the external review of the actuarial component and submission. Gail Flannery was also involved in the development of the demand surge model and the commercial residential model.

The vulnerability, statistical, and computer science components were reviewed by modeler personnel.

B. ***Provide documentation of independent peer reviews directly relevant to the modeling organization's responses to the current standards, disclosures, or forms. Identify any unresolved or outstanding issues as a result of these reviews.***

The written independent review of the wind component by Dr. Gary Barnes is presented in Appendix A. No unresolved outstanding issues remain after the review.

Gail Flannery, FCAS, performed the independent review of the actuarial component. She attended many on-site meetings with the model team and helped in the understanding of the requirements of the actuarial standards, disclosures, and forms. She was provided with all relevant forms and supporting documents. She conducted independent analysis of the A forms and asked questions and provided feedback and suggestions; her questions were addressed, and the feedback and suggestions were acted upon so that no unresolved outstanding issues remain. She prepared the submission document for the actuarial standards. A letter from Gail Flannery can be found in Appendix A. See also Form G-5.

C. ***Describe the nature of any on-going or functional relationship the organization has with any of the persons performing the independent peer reviews.***

Dr. Gary Barnes, Professor of Meteorology at University of Hawaii, performed the external review of the version 2.6 meteorology component of the model, particularly the wind field model. He has no on-going or functional relationship to FIU or the modeling organization, other than as an independent reviewer. He did not take part in the development or testing of the model. His role in the model has been confined to being an independent external reviewer.

4. ***Provide a completed Form G-1, General Standards Expert Certification. Provide a link to the location of the form here.***

See Form G-1

5. ***Provide a completed Form G-2, Meteorological Standards Expert Certification. Provide a link to the location of the form here.***

See Form G-2

6. ***Provide a completed Form G-3, Statistical Standards Expert Certification. Provide a link to the location of the form here.***

See Form G-3

7. ***Provide a completed Form G-4, Vulnerability Standards Expert Certification. Provide a link to the location of the form here.***

See Form G-4

8. ***Provide a completed Form G-5, Actuarial Standards Expert Certification. Provide a link to the location of the form here.***

See Form G-5

9. ***Provide a completed Form G-6, Computer Standards Expert Certification. Provide a link to the location of the form here.***

See Form G-6

G-3 Risk Location

A. *ZIP Codes used in the model shall not differ from the United States Postal Service publication date by more than 24 months at the date of submission of the model. ZIP Code information shall originate from the United States Postal Service.*

Our model uses ZIP Code data exclusively from a third-party developer, which bases its information on the ZIP Code definitions issued by the United States Postal Service. The version we used has a USPS vintage of December 2013. The ZIP Code data have been changed in the current release of the model from last year's submission.

B. *ZIP Code centroids, when used in the model, shall be based on population data.*

ZIP Code centroids used in the model are population centroids and are updated at least every 24 months.

C. *ZIP Code information purchased by the modeling organization shall be verified by the modeling organization for accuracy and appropriateness.*

The ZIP Code information is checked for consistency by experts developing our model. Maps showing the ZIP Code boundaries and the associated centroids will be provided to the professional team during the on-site visit.

D. *If any hazard or any model vulnerability components are dependent on ZIP Code databases, the modeling organization shall maintain a logical process for ensuring these components are consistent with the recent ZIP Code database updates.*

All ZIP Code-dependent components are recreated using the latest update of the ZIP code data in the model.

E. *Geocoding methodology shall be consistent and justifiable.*

The FPHLM uses an enterprise class geocoding engine for converting street addresses to latitude-longitude values.

Disclosures

1. ***List the current ZIP Code databases used by the model and the components of the model to which they relate. Provide the effective (official United States Postal Service) date corresponding to the ZIP Code databases.***

The FPHLM uses 5-digit ZIP Codes distributed by Pitney Bowes. The 5-digit ZIP Codes product constitutes a geographic data set that contains the boundaries for each 5-digit ZIP Code in the United States assigned by the U.S. Postal Service.

The ZIP Code data are updated quarterly. The release we used in this submission has a vintage of 2013.12 (December 2013).

The ZIP Code data are used in the Wind Speed Correction and Insured Loss modules of the model.

2. ***Describe in detail how invalid ZIP Codes are handled.***

For historical loss costs where street addresses are not available, we use contemporaneous ZIP Codes and associated population-based centroids to locate the exposure. The Wind Speed Correction module subsequently determines the current (2013) ZIP Code that contains the historical centroid, and the exposure is then modeled on the basis of the 2013 ZIP code centroid location. If a policy has a ZIP Code that cannot be found in the contemporaneous database of ZIP Codes, it is not modeled.

3. ***Describe the data, methods, and process used in the model to convert among street addresses, geocode locations (latitude-longitude), and ZIP Codes.***

The FPHLM uses Street Map Premium for ArcGIS vintage 2014.1 (January 2014) to geocode street addresses.

4. ***List and provide a brief description of each model ZIP Code-based database (e.g., ZIP Code centroids).***

Population-based zip code centroids and roughness. This database provides the zip code centroid location and corresponding population-weighted roughness and distance to coast for each incoming wind direction octant.

Wind-borne Debris Region (WBDR) ZIP Codes. This database provides the lists of Florida ZIP Codes that fall within the WBDR specified by the Florida Building Code.

5. ***Describe the process for updating model ZIP Code-based databases.***

The zip code boundaries received from the vendor are checked and then the boundaries are used in the recalculation of the zip code centroids, roughness and distance to coast.

G-4 Independence of Model Components

The meteorological, vulnerability, and actuarial components of the model shall each be theoretically sound without compensation for potential bias from the other two components.

The meteorology, vulnerability, and actuarial components of the model are theoretically sound and were developed and validated independently before being integrated. The model components were tested individually.

G-5 Editorial Compliance

The submission and any revisions provided to the Commission throughout the review process shall be reviewed and edited by a person or persons with experience in reviewing technical documents who shall certify on Form G-7, Editorial Certification that the submission has been personally reviewed and is editorially correct.

The current submission document has been reviewed and edited by persons who are qualified to perform such tasks. Future revisions and related documentation will likewise be reviewed and edited by the qualified individual listed in Form G-7 and his assistant.

Disclosures

1. ***Describe the process used for document control of the submission. Describe the process used to ensure that the paper and electronic versions of specific files are identical in content.***

All submission document revisions are passed to the Editor and his assistant prior to inclusion in the document. The assistant is responsible for the electronic version of the document and the technical software issues. Several Word tools are utilized to automate the process of formatting and editing the document. For example, we used Source Manager for APA-style bibliographies, consistent formatting via styles for standards, forms and disclosures, cross-references to cite figures and tables, and multi-level lists to ensure consistent numbering. In addition, Word's track changes tool is used to keep track of modifications to the document since the initial submission. An export filter to PDF format is used to export the document directly to PDF format, which subsequently is printed directly to paper via a printer. The PDF and printed document should be identical barring unforeseen bugs in the PDF export plug-in or PDF printing software.

2. ***Describe the process used by the signatories on Forms G-1 through G-6 (Standards Expert Certification forms) to ensure that the information contained under each set of standards is accurate and complete.***

Each signatory was responsible for doing a final review of the standards related to their expertise prior to submission to verify the accuracy and completeness of the information in the submission document. A professional technical editor and an assistant were hired to perform a thorough edit of the document. All signatories were required to proof-read a PDF version of the document to ensure accuracy and completeness. On-site meetings were held to perform a thorough review of the final version of the document.

3. ***Provide a completed Form G-7, Editorial Certification. Provide a link to the location of the form here.***

See Form G-7.

METEOROLOGICAL STANDARDS

M-1 Base Hurricane Storm Set

A. ***Annual frequencies used in both model calibration and model validation shall be based upon the National Hurricane Center HURDAT2 starting at 1900 as of August 15, 2013 (or later). Complete additional season increments based on updates to HURDAT2 approved by the Tropical Prediction Center/National Hurricane Center are acceptable modifications to these storm sets. Peer reviewed atmospheric science literature can be used to justify modifications to the Base Hurricane Storm Set.***

Validation of the FPHLM is based on the 1900–2013 period of historical record as provided in the April, 2014 version of HURDAT released by the National Hurricane Center.

B. ***Any trends, weighting, or partitioning shall be justified and consistent with currently accepted scientific literature and statistical techniques. Calibration and validation shall encompass the complete Base Hurricane Storm Set as well as any partitions.***

Validation and comparison of the FPHLM encompasses the complete Base Hurricane Storm Set provided in HURDAT. We conduct no trending, weighting, or partitioning of the Base Hurricane Set.

Disclosures

1. ***Identify the Base Hurricane Storm Set, the release date, and the time period included to develop and implement landfall and by-passing hurricane frequencies into the model.***

The National Hurricane Center HURDAT file from April, 2014 for the period 1900–2013 is used to establish the official hurricane base set used by our model. All HURDAT storm tracks that have made landfall in Florida or bypassed Florida but passed close enough to produce damaging winds are documented in our archives.

2. ***If the modeling organization has made any modifications to the Base Hurricane Storm Set related to landfall frequency and characteristics, provide justification for such modifications.***

For stochastic hurricane loss modeling, the HURDAT database indicated in Disclosure 1 is used, unmodified, to develop the probability distribution functions for track and intensity changes and to determine storm frequency.

To model historical losses, we developed a Historical Base Set. This base set is based on the latest HURDAT but includes additional data, such as central pressure and *Rmax*, that may not be available in HURDAT but is needed by the wind model.

3. ***Where the model incorporates short-term or long-term modification of the historical data leading to differences between modeled climatology and that in the entire Base Hurricane Storm Set, describe how this is incorporated.***

The FPHLM incorporates no short-term or long-term modifications of the climate record. Storm frequencies are based on historical occurrences derived from HURDAT and thus implicitly contain any long- or short-term variations that are contained in the historical record. No attempt is made to explicitly model long- or short-term variations.

4. ***Provide a completed Form M-1, Annual Occurrence Rates. Provide a link to the location of the form here.***

See Form M-1.

M-2 Hurricane Parameters and Characteristics

Methods for depicting all modeled hurricane parameters and characteristics including but not limited to windspeed, radial distributions of wind and pressure, minimum central pressure, radius of maximum winds, landfall frequency, tracks, spatial and time variant windfields, and conversion factors, shall be based on information documented in currently accepted scientific literature.

All methods used to depict storm characteristics are based on methods described in the peer-reviewed scientific literature. Our scientists developed datasets using data from published reports, the HURDAT database, archives, observations, and analyses from NOAA's Hurricane Research Division, The Florida State University, Florida International University, and the Florida Coastal Monitoring Program.

Disclosures

1. ***Identify the hurricane parameters (e.g., central pressure or radius of maximum winds) that are used in the model.***

Hurricane parameters used in the model include storm track (translation speed and direction of the storm), radius of maximum wind (*Rmax*), Holland surface pressure profile parameter (*B*), the minimum central sea level pressure (*Pmin*), the damage threshold distance, and the pressure decay as a function of time after landfall.

The storm initial position and motion are modeled using the HURDAT database. For pressure decay we use the Vickery (2005) decay model. Vickery developed the model on the basis of pressure observations in HURDAT and NWS-38, together with *Rmax* and storm motion data as described in the publication. The radius of maximum winds at landfall is modeled by fitting a gamma distribution to a comprehensive set of historical data published in NWS-38 by Ho et al. (1987) and supplemented by the extended best track data of DeMaria, NOAA HRD research flight data, and NOAA-AOML-HRD H*Wind analyses (Powell & Houston, 1996; Powell et al., 1996; Powell & Houston, 1998; Powell et al., 1998).

Additional research was used to construct a historical landfall *Rmax-Pmin* database using existing literature (Ho et al., 1987), extended best track data, HRD Hurricane field program data, and the H*Wind wind analysis archive (Demuth et al., 2006). We developed an *Rmax* model using the revised landfall *Rmax* database, which includes more than 100 measurements for hurricanes up to 2012. We have opted to model the *Rmax* at landfall rather than the entire basin for a variety of reasons. One is that the distribution of landfall *Rmax* may be different than that over open water. An analysis of the landfall *Rmax* database and the 1988–2007 extended best track data shows that there appears to be a difference in the dependence of *Rmax* on central pressure (*Pmin*) between the two datasets (Demuth et al., 2006). The landfall dataset provides a larger set of independent measurements (more than 100 storms compared to about 31 storms affecting the Florida threat area region in the best track data). Since landfall *Rmax* is most relevant for loss cost estimation and has a larger independent sample size, we have chosen to model the landfall dataset. Future studies will examine how the extended best track data can be used to supplement the landfall dataset.

Recent research results by Willoughby and Rahn (2004) based on the NOAA-AOML-HRD annual hurricane field program and Air Force reconnaissance flight-level observations are used to create a model for the "*Holland B*" parameter. Ongoing research on the relationship between horizontal surface wind distributions (based on Stepped Frequency Microwave Radiometer observations) to flight level distributions (Powell et al., 2009) is used to correct the flight-level *Rmax* to a surface *Rmax* when developing a relationship for the *Holland B* term. We multiply the flight-level *Rmax* from the Willoughby and Rahn (2004) dataset by 0.815 to estimate the surface *Rmax* (based on SFMR, flight-level maxima pair data). This adjustment keeps the Holland pressure profile parameter consistent with a surface *Rmax* and because of the negative term in the equation produces a larger value of *B* than if a flight-level value of *Rmax* were used. This is consistent with the concept of a stronger radial pressure gradient for the mean boundary layer slab than at flight level (due to the warm core of the storm), which agrees with GPS dropsonde wind profile observations showing boundary layer winds that are stronger than those at the 10,000 ft flight level, which is the level for most of the *B* data in Willoughby and Rahn (2004). The *B* adjustment for a surface *Rmax* produces an overall stronger surface wind field than if *B* were not adjusted. In addition, surface pressures from the "best track" information on HURDAT are used to associate a particular flight-level pressure profile *B* with a surface pressure.

The NOAA-AOML-HRD H*Wind analysis archive was used to develop a relationship between *Rmax* and the extent of damaging winds to make sure that the model would only consider land locations that have potential for damaging winds. HRD wind modeling research initiated by Ooyama (1969) and extended by Shapiro (1983) has been used to develop the HRD wind field model. This model is based on the concept of a slab boundary layer model, a concept pioneered at NOAA-AOML-HRD and now in use by other modelers for risk applications (Thompson & Cardone, 1996; Vickery & Twisdale, 1995; Vickery et al., 2000b). The HURDAT historical database is used to develop the track and intensity model. Historical data used for computing the potential intensity is based on the National Centers for Environmental Prediction (NCEP) sea surface temperature archives and the NCEP reanalysis for determining the upper tropospheric outflow temperatures. Use cases describing the various model functions and their research bases are available with the model documentation.

2. ***Describe the dependencies among variables in the windfield component and how they are represented in the model, including the mathematical dependence of modeled windfield as a function of distance and direction from the center position.***

B depends linearly on latitude and *Rmax*, and quadratically on *DelP*. The gradient wind for the slab boundary layer depends on *Pmin* (through *DelP*) and *B*; the mean slab planetary boundary layer (PBL) wind depends on the gradient wind, the drag coefficient (which depends on wind speed), the air density, the gradients of the tangential and radial components of the wind, and the Coriolis parameter (which also depends on latitude). The wind field model solves the equations of motion on a polar grid with a 0.1 *R/Rmax* radial grid resolution. The input *Rmax* is reduced by 10% to correct a small bias in *Rmax* caused by a tendency of the wind field solution to place *Rmax* radially outward by one grid point. The wind field model terms and dependencies are further described in Powell et al. (2005).

3. ***Identify whether hurricane parameters are modeled as random variables, as functions, or***

as fixed values for the stochastic storm set. Provide rationale for the choice of parameter representations.

Initial storm positions and motion changes derived from HURDAT are modified by the addition of small uniform random error terms. Subsequent storm motion change and intensity are obtained by sampling from empirically derived PDFs as described in Section G-1.2. The random error term for the B parameter is a normal distribution with zero mean and a standard deviation derived from observed reconnaissance aircraft pressure profile fits for B (Willoughby & Rahn, 2004). The radius of maximum winds is sampled from a gamma distribution based on landfall *Rmax* data and is described in more detail below and in Standard G-1.2.

Since *Rmax* is nonnegative and skewed, we model the distribution using a gamma distribution. Using the maximum likelihood estimators, we found the parameters for the gamma distribution to be k=4.76, θ=5.41. A discussion of the goodness of fit for *Rmax* is found in Standard S-1.

An examination of the *Rmax* database shows that intense storms, essentially Category 5 storms, have rather small radii. Thermodynamic considerations (Willoughby, 1998) also suggest that smaller radii are more likely for these storms. Thus, we model Category 5 (*DelP*>90 mb, where *DelP*=1013-*Pmin* and *Pmin* is the central pressure of the storm) storms using a gamma distribution, but with a smaller value of the θ parameter, which yields a smaller mean *Rmax* as well as smaller variance. We have found that for Category 1–4 (*DelP*<80) storms there is essentially no discernable dependence of *Rmax* on central pressure. This is further verified by looking at the mean and variance of *Rmax* in each 10 mb interval. Thus, we model Category 1–4 storms with a single set of parameters. For a gamma distribution, the mean is given by $k\theta$, and variance is $k\theta^2$. For Category 5 storms, we adjust θ such that the mean is equal to the mean of the three Category 5 storms in the database: 1935 No Name, 1969 Camille, and 1992 Andrew. An intermediate zone between *DelP*=80 mb and *DelP*=90 mb is established where the mean of the distribution is linearly interpolated between the Category 1–4 value and the Category 5 value. As the θ value is reduced, the variance is likewise reduced. Since there are insufficient observations to determine what the variance should be for Category 5 storms, we rely on the assumption that variance is appropriately described by the rescaled θ, via $k\theta^2$.

A simple method is used to generate the gamma-distributed values. A uniformly distributed variable is mapped onto the range of *Rmax* values via the inverse cumulative gamma distribution function. For computational efficiency, a lookup table is used for the inverse cumulative gamma distribution function.

For Category 5 and intermediate Category 4–5 storms, we use the property that the gamma cumulative distribution function is a function of (k,x/θ). Thus, by rescaling θ, we can use the same function (lookup table), but just rescale x (*Rmax*). The rescaled *Rmax* will then still have a gamma distribution but with different mean and variance.

The storms in the stochastic model will undergo central pressure changes during the storm life cycle. When a storm is generated, an appropriate *Rmax* is sampled for the storm. To ensure the appropriate mean values of *Rmax* as pressure changes, the *Rmax* is rescaled every time step as necessary. As long as the storm has *DelP* < 80 mb, there is in effect no rescaling. In the stochastic storm generator, we limit the range of *Rmax* from 4 sm to 120 sm. The wind field

solution, after including the translation speed, results in values of *Rmax* that are outside this range less than 2% of the time.

4. ***Describe how any hurricane parameters are treated differently in the historical and stochastic storm sets (e.g., has a fixed value in one set and not the other).***

All historical storm sets consist of input files containing information derived from HURDAT or other observation sources as described in Standard M-1. All stochastic input storm tracks are modeled.

5. ***State whether the model simulates surface winds directly or requires conversion between some other reference level or layer and the surface. Describe the source(s) of conversion factors and the rationale for their use. Describe the process for converting the modeled vortex winds to surface winds including the treatment of the inherent uncertainties in the conversion factor with respect to location of the site compared to the radius of maximum winds over time. Justify the variation in the surface winds conversion factor as a function of hurricane intensity and distance from the hurricane center.***

The mean boundary layer winds computed by the model are adjusted to the surface using results from Powell et al. (2003), which estimated a mean surface wind factor of 77.5% on the basis of over 300 GPS sonde wind profile observations in hurricanes. The surface wind factor is based on the ratio of the surface wind speed at 10 m to the mean wind speed for the 0–500 m layer (mean boundary layer wind speed or MBL) published in Powell et al. (2003). This ratio is far more relevant to a slab boundary layer model than using data based on higher, reconnaissance aircraft flight levels. The depth of the slab boundary layer model is assigned a value of 450 m, which is the level of the maximum mean wind speed from GPS sonde wind profiles published in Powell et al. (2003). The uncertainty of the surface wind factor is ~8%, based on the standard deviation of the measurements, but no attempt is made to model this uncertainty. No radial distance from center or intensity dependent variation of reduction factor is used at this time because of a lack of dependency on these quantities based on examination of GPS dropsonde data (Figure 25).

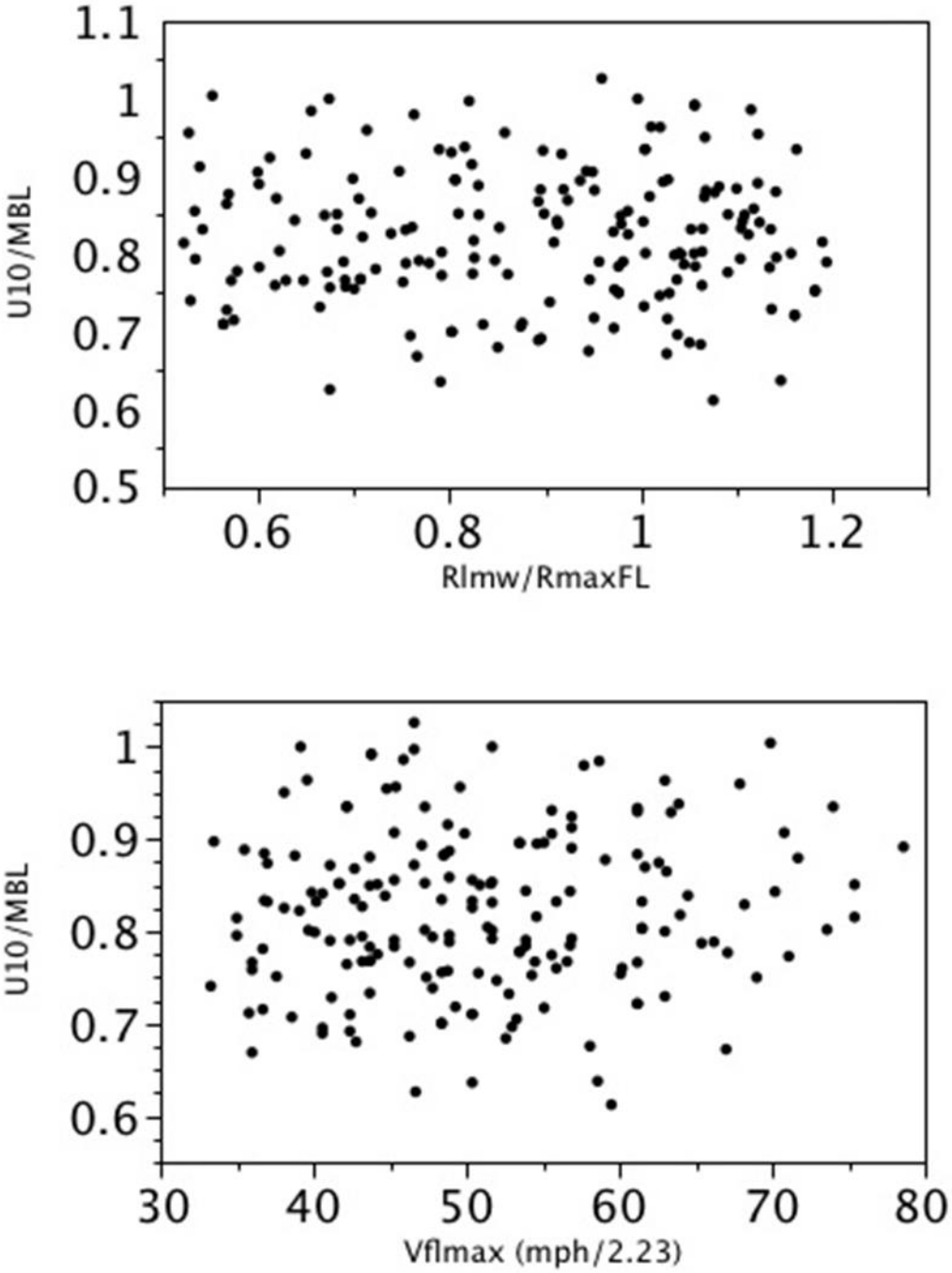


Figure 25. Analysis of 742 GPS dropsonde profiles launched from 2-4 km with flight-level winds at launch greater than hurricane force and with measured surface winds. Upper figure: Dependence of the ratio of 10 m wind speed (U10) to the mean boundary layer wind speed (MBL) on the scaled radius (ratio of radius of last measured wind (Rlmw) to the radius of maximum wind at flight level (RmaxFL). Lower figure: Surface wind factor (U10/MBL) dependence on maximum flight level wind speed (Vflmax, in units of miles per hour / 2.23).

6. ***Describe how the windspeeds generated in the windfield model are converted from sustained***

to gust and identify the averaging time.

Wind speeds from the HRD slab boundary layer wind field model are assumed to represent ten-minute averages. A sustained wind is computed by applying a gust factor to account for the highest one-minute wind speed over the ten-minute period. A peak three-second gust is also computed. Gust factors depend on wind speed and the upstream fetch roughness, which in turn depends on wind direction at a particular location. Gust factor calculations were developed using research in the Engineering Sciences Data Unit (ESDU) series papers as summarized and applied to tropical cyclones by Vickery and Skerlj (2005).

7. ***Describe the historical data used as the basis for the model's hurricane tracks. Discuss the appropriateness of the model stochastic hurricane tracks with reference to the historical hurricane database.***

The hurricane tracks are modeled as a Markov process. Initial storm conditions are derived from HURDAT. Small uniform random perturbations are added to the historical initial conditions, including initial storm location, change in motion, and intensity.

Storm motion is determined by sampling empirical distributions, based on HURDAT, of change in speed and change in direction, as well as change in relative intensity. These functions are also spatially dependent, binned in variable box sizes (typically 2.5 degrees), and enlarged as necessary to ensure sufficient density of storms for the distribution.

The model has been validated by examining key hurricane statistics relative to HURDAT at roughly 30 sm milepost locations along the Gulf and Atlantic coasts. The parameters examined include average central pressure deficit, average heading angle and speed, and total occurrence by Saffir-Simpson category.

8. ***If the historical data are partitioned or modified, describe how the hurricane parameters are affected.***

The FPHLM does not partition or modify the historical data.

9. ***Describe how the coastline is segmented (or partitioned) in determining the parameters for hurricane frequency used in the model. Provide the hurricane frequency distribution by intensity for each segment.***

The model does not use coastline segmentation to determine hurricane frequency.

10. ***Describe any evolution of the functional representation of hurricane parameters during an individual storm life cycle.***

Upon landfall, the evolution of the central pressure changes from sampling a PDF to a decay model described in Vickery (2005). When the storm exits back over water, the pressure is again modeled via the PDF. After landfall, the slab boundary layer, surface drag coefficient changes from a functional marine form to a constant based on a mean aerodynamic roughness length of

0.2 m. The slab boundary layer height increases from 450 m to 1 km after the center makes landfall and decreases back to 450 m if the center exits land to go back to sea.

M-3 Hurricane Probabilities

A. ***Modeled probability distributions of hurricane parameters and characteristics shall be consistent with historical hurricanes in the Atlantic basin.***

Hurricane motion (track) is modeled based on historical geographic probability distributions of hurricane translation velocity and velocity change, initial intensity, intensity change, and potential intensity. Modeled probability distributions for hurricane intensity, forward speed, *Rmax*, and storm heading are consistent with historical hurricanes in the Atlantic basin.

B. ***Modeled hurricane landfall frequency distributions shall reflect the Base Hurricane Storm Set used for category 1 to 5 hurricanes and shall be consistent with those observed for each coastal segment of Florida and neighboring states (Alabama, Georgia, and Mississippi).***

As shown in Form M-1 and the accompanying plots, our model reflects reasonably the 1900–2013 Base Hurricane Set for hurricanes of Saffir-Simpson Categories 1–5 in each coastal region of Florida, as well as in the neighboring states. In addition, a finer scale coastal milepost study of model parameters (occurrence rate, storm translation speed, storm heading, and *Pmin*) was conducted during the development of the model.

C. ***Models shall use maximum one-minute sustained 10-meter windspeed when defining hurricane landfall intensity. This applies both to the Base Hurricane Storm Set used to develop landfall frequency distributions as a function of coastal location and to the modeled winds in each hurricane which causes damage. The associated maximum one-minute sustained 10-meter windspeed shall be within the range of windspeeds (in statute miles per hour) categorized by the Saffir-Simpson Scale.***

Saffir-Simpson Hurricane Scale:

Category	Winds (mph)	Damage
1	74 – 95	Minimal
2	96 – 110	Moderate
3	111 – 129	Extensive
4	130 – 156	Extreme
5	157 or higher	Catastrophic

The HRD wind field model simulates landfall intensity according to the maximum one-minute sustained wind for the 10 m level for both stochastic simulations and the Base Hurricane Set. The Saffir-Simpson damage potential scale is used to further categorize the intensity at landfall, and the range of simulated wind speeds (in miles per hour) is within the range defined in the scale.

Disclosures

1. ***List assumptions used in creating the hurricane characteristic databases.***

The *Holland B* database is based on flight-level pressure profiles corresponding to constant pressure surfaces at 700 mb and below. Because of a lack of surface pressure field data, an assumption is made that the *Holland B* at the surface is equivalent to a *B* determined from information collected at flight level. The surface pressure profile uses *Pmin*, *DelP*, and *Rmax* at the surface. It would be ideal to have a *B* dataset also corresponding to the surface, but such data are not available. The best available data on *B* are flight-level data from Willoughby and Rahn (2004). Willoughby and Rahn (2004) reveal that during major hurricanes most flights flew at 3 km (700 mb). Few lower-level data are available for mature hurricanes, so their plot (Figure 3) of *B* vs. flight level does not provide data about average vertical structure. In lieu of lower-level data, we model *B* using flight data supplied by Willoughby, but with *Rmax* adjusted to a surface *Rmax*, and with surface *DelP* added from NHC best track data for each flight. Since we are modeling hurricane winds during landfall, our *Rmax* model applies only to landfall and is not designed to model the life cycle of *Rmax* as a function of intensity.

2. ***Provide a brief rationale for the probability distributions used for all hurricane parameters and characteristics.***

Form S-3 provides a list of probability distributions used to model hurricane parameters. Further discussion and rationale for these functions are provided in Standard M-2, Disclosure 1 and Standard S-1, Disclosure 1. Some of the details pertaining to data sources used are described below.

Monthly geographic distributions of climatological sea surface temperatures (Reynolds et al., 2002) and upper tropospheric outflow temperatures (Kanamitsu et al., 2002) are used to determine physically realistic potential intensities that help to bound the modeled intensity. Terrain elevation and bathymetry data were obtained from the United States Geological Survey. The radius of maximum wind at landfall is modeled from a comprehensive set of historical data published in NWS-38 by Ho et al. (1987) but supplemented by the extended best track data of DeMaria (Pennington et al., 2000), the HURDAT Reanalysis Project (Landsea et al, 2004), NOAA HRD research flight data, and NOAA-HRD H*Wind analyses (Powell et al., 1996, 1998). The development of the Rmax frequency distribution fit and its comparison to historical hurricane data are discussed in M-2.1, M-2.3 and in Standard S-1. Comparisons of the modeled radius of maximum wind to the observed data are shown in Form M-3.

M-4 Hurricane Windfield Structure

A. *Windfields generated by the model shall be consistent with observed historical storms affecting Florida.*

As described in Statistical Standards S-1, Disclosure 2, comparisons of FPHLM to gridded H*Wind fields indicate that the FPHLM wind fields are consistent with observed historical wind fields from Florida landfalling hurricanes.

B. *The land use and land cover database shall be consistent with National Land Cover Database (NLCD) 2006 or later. Use of alternate data sets shall be justified.*

We use the MRLC NLCD 2011 land use dataset as well as the Statewide 2004-2011 Land Use/Land Cover dataset developed and maintained by the Florida Water Management Districts (WMD) and compiled and distributed by the Florida Department of Environmental Protection. The NLCD dataset became available in Spring 2014 and provides detailed (30 m) land use characteristics circa 2011. The datasets of the individual water management districts were combined in the statewide WMD dataset to form a unified dataset. The WMD data are based on 2004-2011 imagery. Both of these data sets are based on imagery that is generally more current than NLCD 2006.

C. *The translation of land use and land cover or other source information into a surface roughness distribution shall be consistent with current state-of-the-science and shall be implemented with appropriate geographic information system data.*

Land friction is modeled according to the currently accepted, state-of-the-science principles of surface layer similarity theory as described in the disciplines of micrometeorology, atmospheric turbulence, and wind engineering. The geographic distribution of surface roughness is determined by careful studies of aerial photography and satellite remote sensing measurements used to create land use-land cover classification systems. We have developed a roughness dataset at 90 meter resolution covering the state of Florida to enable modeling losses at the "street level." For modeling losses at the ZIP Code level, we use population-weighted roughness.

All street level locations (at 90 m resolution) and population-weighted ZIP Code centroids are assigned roughness values as a function of upstream fetch for each wind direction octant. After landfall, the surface drag coefficient used in the hurricane PBL slab model changes from a marine value to a fixed value associated with a roughness of 0.2 m.

D. *With respect to multi-story buildings, the model windfield shall account for the effects of the vertical variation of winds if not accounted for in the vulnerability functions.*

The modeled wind fields take into account vertical variation through the terrain conversion methodology based on Vickery et al. (2009). The coastal transition function also takes into account variation of wind with height.

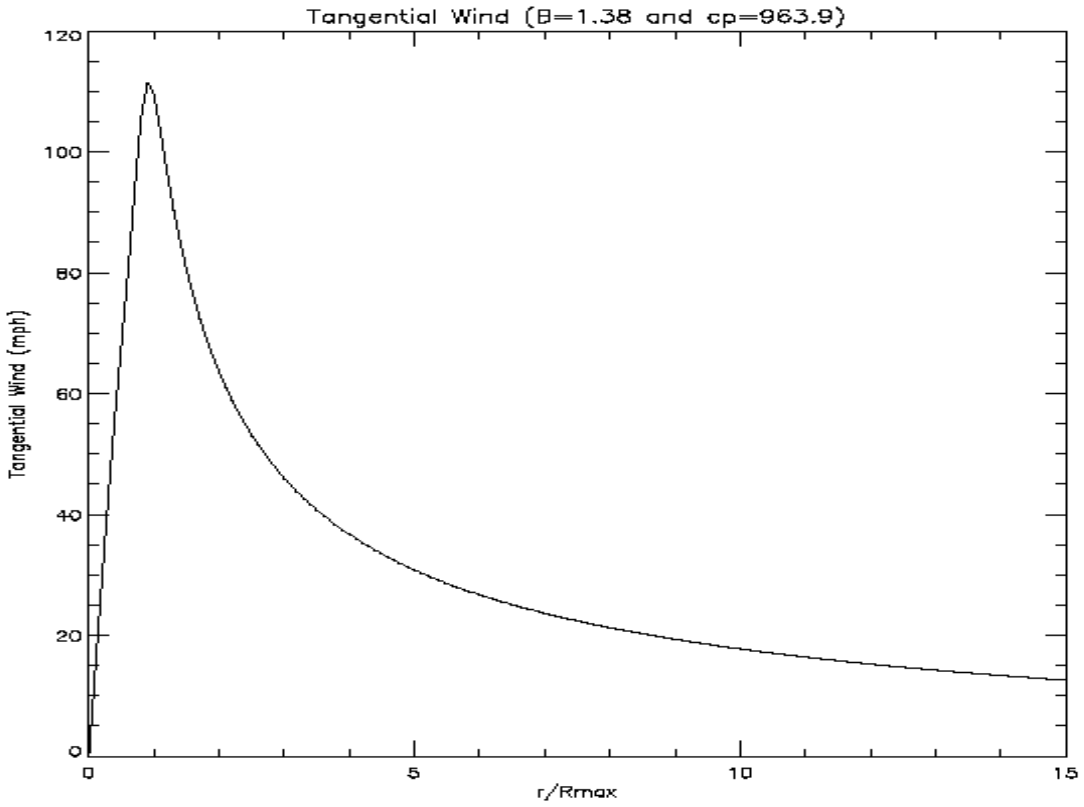


Figure 26. Axisymmetric rotational wind speed (mph) vs. scaled radius for B = 1.38, DelP = 49.1 mb.

Disclosures

1. ***Provide a rotational windspeed (y-axis) versus radius (x-axis) plot of the average or default symmetric wind profile used in the model and justify the choice of this wind profile.***

See Figure 26. The *Holland B* profile has been compared extensively to historical data (Holland, 1980; Willoughby & Rahn, 2004) and found to be a reasonable fit.

2. ***If the model windfield has been modified in any way from the previous submission, provide a rotational windspeed (y-axis) versus radius (x-axis) plot of the average or default symmetric wind profile for both the new and old functions. The choice of average or default shall be consistent for the new and old functions.***

The wind field model has not been modified since the previous submission.

3. ***If the model windfield has been modified in any way from the previous submission, describe variations between the new and old windfield functions with reference to historical storms.***

The wind field model has not been modified since the previous submission.

4. ***Describe how the vertical variation of winds is accounted for in the model where applicable. Document and justify any difference in the methodology for treating historical and stochastic storm sets.***

Vertical variation of wind is accounted for in the terrain conversion methodology described in Vickery et al. (2009). This methodology is a modification of the log wind profile and has been validated against dropsonde data. The coastal transition function, which is based on the above methodology, also incorporates variation with height so that the impact of a larger marine fetch on taller structures in coastal regions can be modeled. The treatment of vertical variation of winds is the same for both historical and stochastic storm sets.

5. ***Describe the relevance of the formulation of gust factor(s) used in the model.***

The gust factors used in the model were developed from hurricane wind speed data and the Engineering Sciences Data Unit methods as described in Vickery and Skerlj (2005).

6. ***Identify all non-meteorological variables that affect windspeed estimation (e.g., surface roughness, topography, etc.).***

Upstream aerodynamic surface roughness within a fixed 45-degree sector extending upstream has an effect on the determination of wind speed for a given street location (latitude and longitude) or ZIP Code centroid and is a significant variable that affects estimation of surface wind speeds. The upstream sectors are defined according to the Tropical Cyclone Winds at Landfall Project (Powell et al., 2004), which characterized upstream wind exposure for each of eight wind direction sectors at over 200 coastal automated weather stations (Figure 27). In additional, a coastal transition function is employed to account for the smooth marine fetch near coastal regions.



Figure 27. Upstream fetch wind exposure photograph for Chatham, MA (left, looking north), and Panama City, FL (right, looking northeast). After Powell et al. (2004).

7. ***Provide the collection and publication dates of the land use and land cover data used in the model and justify their timeliness for Florida.***

We use the 2011 Multi-Resolution Land Characteristics Consortium (MRLC) National Land Cover Database released on March 31, 2014. This is a high-resolution (30 m) land cover dataset that covers not only Florida, but the entire United States, and roughly depicts land characteristics circa 2011 [see Jin et al. (2013) for more details]. We also use the Statewide 2004-2011 Florida Water Management District Land Use/Land Cover dataset based on 2004-2011 imagery. This dataset was published by the Florida Department of Environmental Protection on March 8, 2013.

8. ***Describe the methodology used to convert land use and land cover information into a spatial distribution of roughness coefficients in Florida and adjacent states.***

The land cover classifications provided by the MRLC Land Cover Database and the WMD land use/land cover data are first mapped to roughness values using a lookup table that associates a representative roughness for the land use category on the basis of peer-reviewed literature. An algorithm was developed to merge the datasets based on how well each dataset classified the land surface with respect to surface roughness. An effective roughness model (Axe, 2004) is then used to incorporate upstream roughness elements to provide a more realistic roughness on a 90 m (295 ft) grid covering Florida.

9. ***Demonstrate the consistency of the spatial distribution of model-generated winds with observed windfields for hurricanes affecting Florida. Describe and justify the appropriateness of the databases used in the windfield validations.***

As shown below in Disclosure 10 and in Statistical Standard 1, Disclosure 2, the spatial distribution of model-generated winds is consistent with observed wind fields for hurricanes affecting Florida. The observations are from the the H*Wind surface analyses produced by NOAA's Hurricane Research Division. These analyses are described in detail in Standard S-1, Disclosure 2. The H*wind analyses are highly regarded in the scientific community and have been cited in over 400 peer-reviewed publications.

10. ***Describe how the model's windfield is consistent with the inherent differences in windfields for such diverse hurricanes as Hurricane Charley (2004), Hurricane Jeanne (2004), and Hurricane Wilma (2005).***

The model can represent a wide variety of storms through variation of parameters for radius of maximum winds, central pressure deficit, and *Holland B*. Snapshots of model wind fields at landfall are compared to NOAA-AOML-HRD H*Wind analyses below (for further details see Disclosure 2 for Standard S-1). In these cases, rather than tuning the model to best fit the observations by varying the *Holland B* parameter, we derived the input *B* from the H*Wind analyses. Hurricane Charley, a small, fast moving 2004 hurricane (Figure 28, top), was modeled quite well; the motion asymmetry and extent of strong winds in the core of the storm were captured but the peak wind (near 150 mph) was underestimated by the model. Hurricane Jeanne Figure 28, bottom) struck the central Florida Atlantic coast in 2004. Similar to the observed (H*Wind) field, the modeled wind field maximum is on the right (north) side of the storm, but the model underestimates the peak wind of 105 mph and the area of winds above 70 mph. Wilma made landfall in Florida in 2005 as a very large hurricane (Figure 29). The FPHLM captures the location of maximum winds in the core of the storm and represents the left-right motion asymmetry, but tends to produce too broad of a wind field.

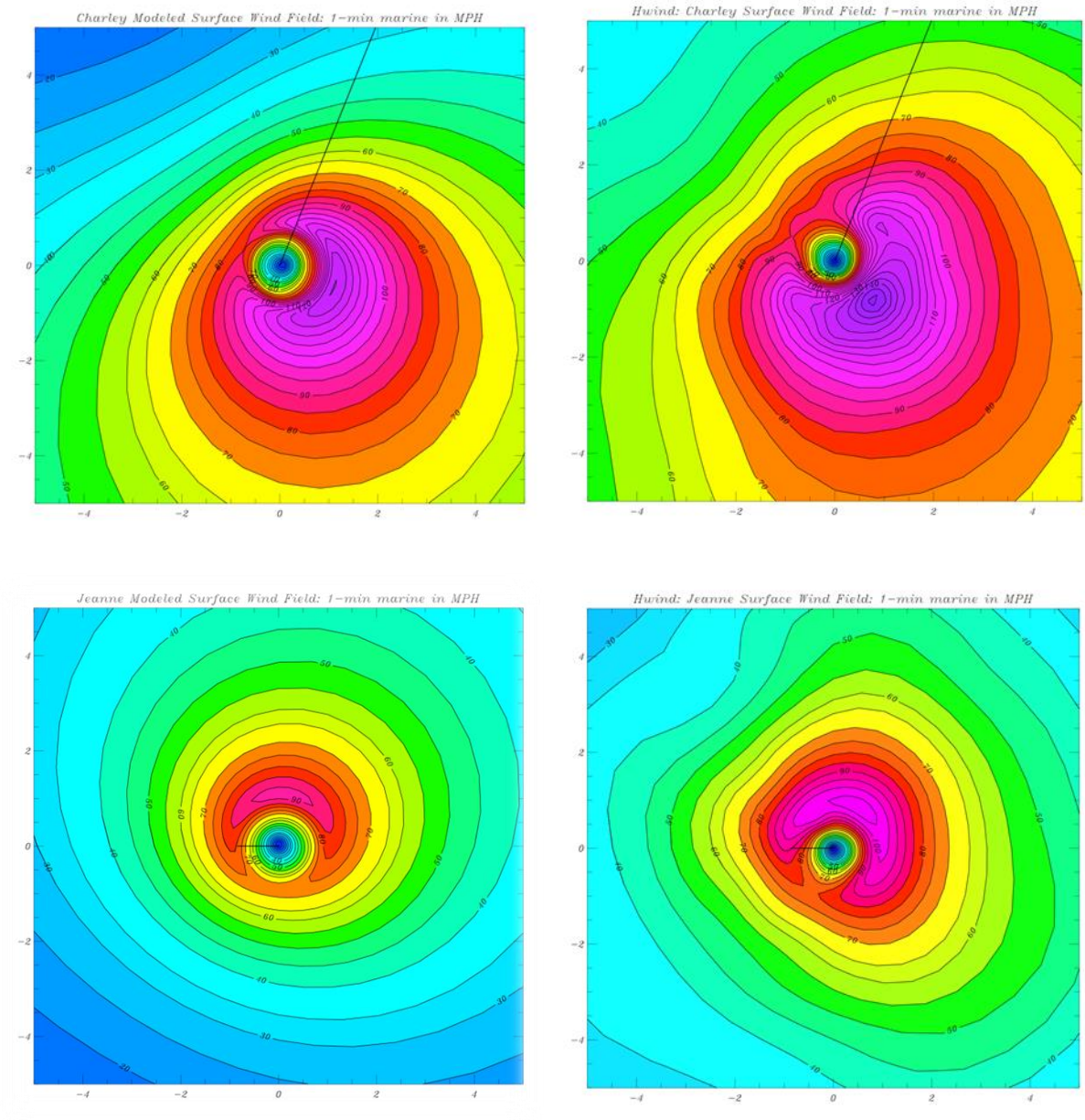


Figure 28. Comparison of modeled (left) and observed (H*Wind, right) landfall wind fields of Hurricane Charley (2004, top) and Hurricane Jeanne (2004, bottom). Line segment indicates storm heading. Horizontal coordinates are in units of *R/Rmax* and winds units of miles per hour. All wind fields are for marine exposure.

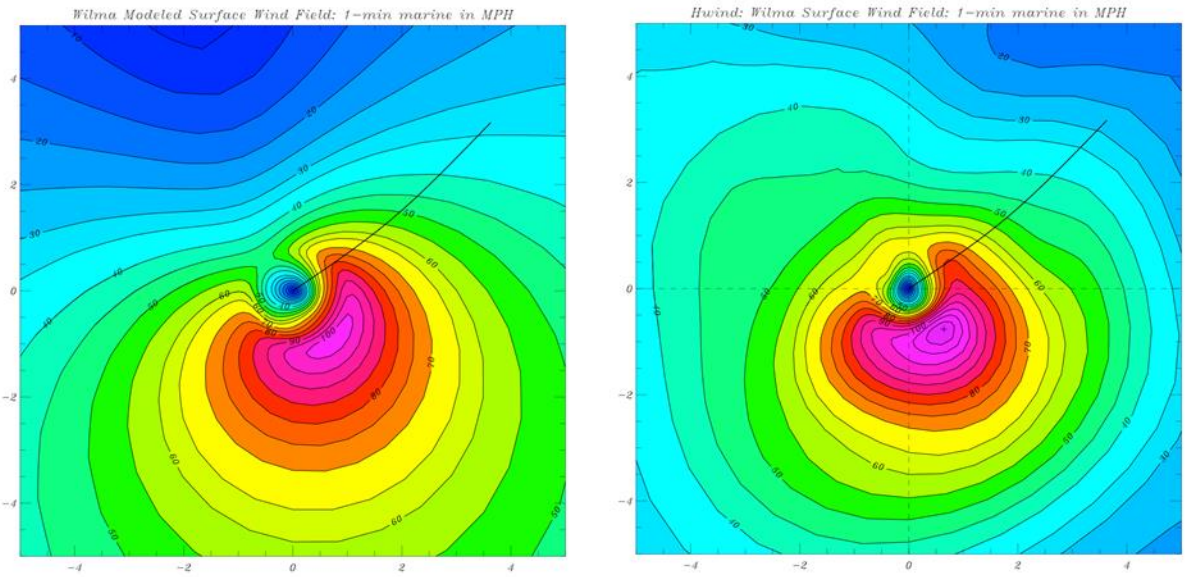


Figure 29. As in Fig. 28 but for Hurricane Wilma of 2005.

11. ***Describe any variations in the treatment of the model windfield for stochastic versus historical storms and justify this variation.***

All historical storm sets consist of input files containing information derived from HURDAT or other observation sources as described in Standard M-1. All stochastic input storm tracks are modeled. The wind field is modeled from the stochastic or historical input files in the same manner.

12. ***Provide a completed Form M-2, Maps of Maximum Winds. Explain the differences between the spatial distributions of maximum winds for open terrain and actual terrain for historical storms. Provide a link to the location of the form here.***

See Form M-2.

The open terrain winds are based on the common assumption that the wind is in equilibrium with open terrain roughness (0.03 m) with infinite fetch. The actual terrain winds are assumed to be in equilibrium with the local (effective) roughness near the surface, but near coastal regions the winds aloft may be more in equilibrium with marine roughness. Thus, it is possible for regions near the coast to have actual terrain winds that are larger than open terrain winds. The spatial distributions of open and actual terrain wind can be quite different because of the coastal transition and the fact that surface roughness in general has a large impact on the wind field. Spatial variations of roughness on the order of a few miles can cause large differences in the wind on that spatial scale.

M-5 Landfall and Over-Land Weakening Methodologies

A. *The hurricane over-land weakening rate methodology used by the model shall be consistent with historical records and with current state-of-the-science.*

Overland weakening rates are based on a pressure decay model developed from historical data as described by a recent paper published in peer-reviewed atmospheric science literature (Vickery, 2005).

B. *The transition of winds from over-water to over-land within the model shall be consistent with current state-of-the-science.*

The transition of winds from over-water to over-land is consistent with the current state of the science through the use of a pressure decay model (Vickery, 2005), a terrain conversion model from marine to actual roughness, and a coastal transition function (Vickery et al., 2009).

Disclosures

1. ***Describe and justify the functional form of hurricane decay rates used by the model.***

The hurricane decay rate function acts to decrease the *DelP* with time after landfall. The functional form is an exponential in time since landfall and is based on historical data (Vickery, 2005).

2. ***Provide a graphical representation of the modeled decay rates for Florida hurricanes over time compared to wind observations.***

The degradation of the wind field of a landfalling hurricane is associated with the filling of the central sea level pressure and the associated weakening of the surface pressure gradient; also the hurricane is over land, where the flow is subject to friction while flowing across obstacles in the form of roughness elements. Maximum wind degradation is shown according to how the maximum sustained surface wind (at the location containing the maximum winds in the storm) changes with time after landfall. At landfall the marine exposure wind is assumed to be representative of the maximum winds occurring onshore. After landfall the open terrain wind is chosen to represent the maximum envelope of sustained winds over land. The NOAA-HRD H*Wind system is used to analyze the maximum winds at a sequence of times following landfalls of Hurricanes Katrina, Charley, Frances, Jeanne, and Wilma. H*Wind uses all available wind observations. The landfall wind field is used as a background field for times after landfall and compared to the available observations at a sequence of times after landfall. An empirical decay is applied to the background field based on the comparisons to the observations. These data are then objectively analyzed to determine the wind field at each time. The model maximum sustained winds are compared to the maximum winds from the H*Wind analyses for the same times and roughness exposures. In general, points after landfall are given for open terrain exposure. At times, even though the storm center is over land, the maximum wind speed may remain over water. For example, in the Hurricane Frances plot (Figure 30), the first three pairs of points represent marine exposure, the next three open terrain, and the final three marine exposure again, while all Hurricane Wilma point pairs (Figure 31) represent marine exposure. The plots

indicate that the public wind field model realistically simulates decay of the maximum wind speed during the landfall process, as well as subsequent strengthening after exit.

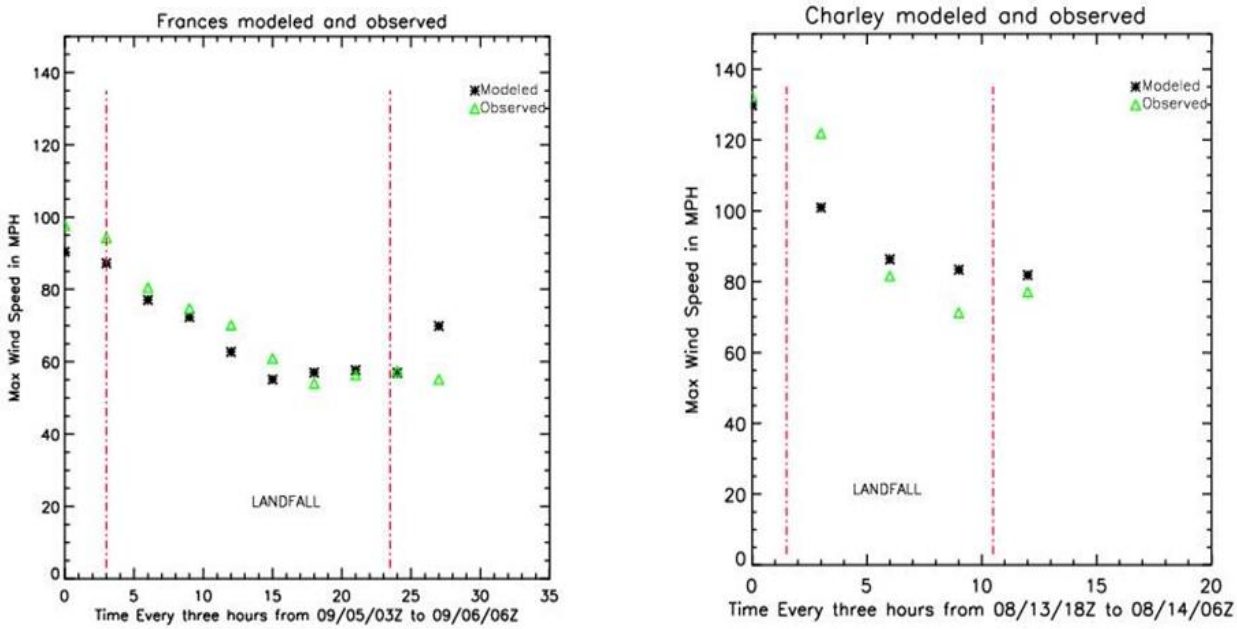


Figure 30. ~~Observed (green) and modeled (black) maximum sustained surface winds as a function of time for 2004 Hurricanes Frances (left) and Charley (right). Landfall is represented by the vertical dash-dot red line at the left and time of exit as the red line on the right.~~ Observed (green) and modeled (black) maximum sustained surface winds as a function of time for 2004 Hurricanes Frances (left) and Charley (right). Landfall is represented by the vertical dash-dot red line at the left and time of exit as the red line on the right. For Hurricane Frances (left) the first three pairs of points represent marine exposure, the next three open terrain, and the final three pairs represent marine exposure. For Hurricane Charley (right) all pairs represent open terrain.

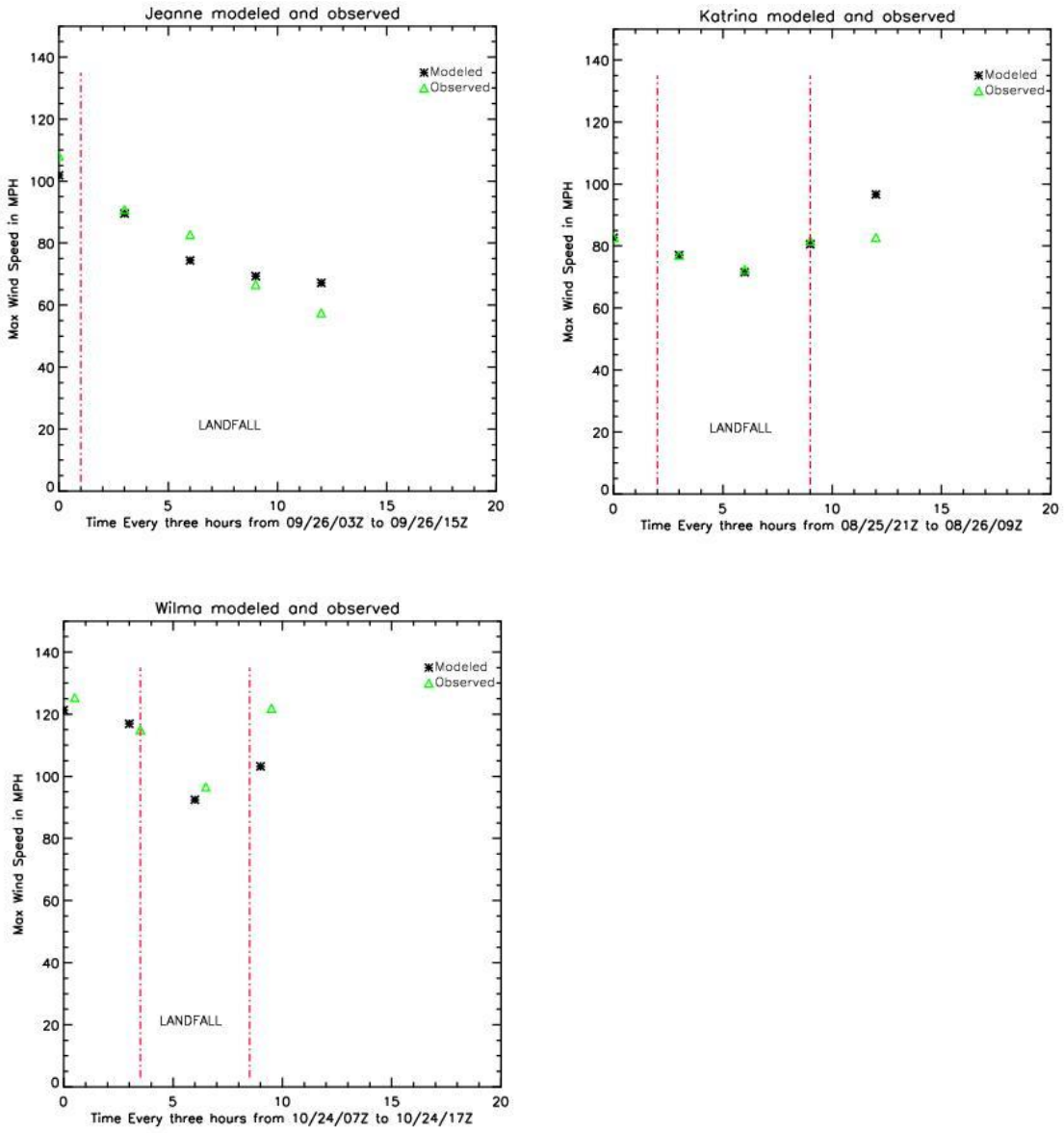


Figure 31. ~~Observed (green) and modeled (black) maximum sustained surface winds as a function of time for Hurricanes Jeanne (2004, top left), Katrina (2005 in South Florida, top right), and Wilma (2005, lower left). Landfall is represented by the vertical dash-dot red line at the left and time of exit as the red line on the right.~~ Observed (green) and modeled (black) maximum sustained surface winds as a function of time for Hurricanes Jeanne (2004, top left, open terrain), Katrina (2005 in South Florida, top right, open terrain), and Wilma (2005, lower left, marine exposure). Landfall is represented by the vertical dash-dot red line at the left and time of exit as the red line on the right.

3. ***Describe the transition from over-water to over-land boundary layer simulated in the model.***

After landfall, the slab boundary layer, surface drag coefficient changes from a functional marine form to a constant based on a mean aerodynamic roughness length of 0.2 m. The slab boundary layer height increases from 450 m to 1 km after the center makes landfall and decreases back to 450 m if the center exits land to go back to sea. To determine surface winds, an effective roughness model is used along with a coastal transition function. The coastal transition function is based on the concept of a growing internal boundary layer (Arya, 1988) for the sea-to-land transition. Within the equilibrium layer, assumed to be one tenth of the internal boundary layer (IBL) height in depth, the wind is assumed to be in equilibrium with the local effective roughness. Above the IBL the wind is assumed to be in equilibrium with marine roughness. Between the equilibrium layer and the IBL we assume that the wind is in equilibrium with vertically varying, stepwise increments of roughness that decay linearly from the local roughness to marine roughness. This is similar in concept to the methodology described in ESDU, and the modeled transition is very close to the ESDU values reported in Vickery et al. (2009).

4. ***Describe any changes in hurricane parameters, other than intensity, resulting from the transition from over-water to over-land.***

See Standard M-2, Disclosure 10. The *Holland B* parameter has a weak dependence on pressure and will undergo slight change. The radius of maximum winds has an implicit dependence on pressure through the scale and shape parameters of the gamma distribution (see M-2, Disclosure 3), and thus strong storms making landfall could undergo some expansion.

5. ***Describe the representation in the model of passage over non-continental U.S. land masses on hurricanes affecting Florida.***

Noncontinental U. S. land masses are identified by a land-ocean mask that keeps track of whether the storm center is over the land or ocean. Storms that pass over noncontinental U.S. land masses (e.g., Cuba) undergo decay, just as storms do crossing continental land masses (e.g., mainland U. S.) using a pressure-filling model (Vickery, 2005).

6. ***Document any differences between the treatment of decay rates in the model for stochastic hurricanes compared to historical hurricanes affecting Florida.***

In the FPHLM model, decay is defined as the change in minimum sea level pressure (*Pmin*) with time after landfall. The input file for the wind field model consists of a hurricane track file that contains storm position, *Pmin*, *Rmax*, and *Holland B* at 1 h frequency. The wind field model is exactly the same for scenario (historical) or stochastic events. When running the model in scenario mode for historical hurricanes affecting Florida, we use a set of historical hurricane tracks as input to the model. When the model is run in stochastic mode, the input hurricane tracks are provided by the track and intensity model. The track and intensity model uses the Vickery (2005) pressure decay after landfall. When a hurricane exits land, the *Pmin* over water is determined on the basis of the Markov process as described in Disclosure G-1.2.

For historical hurricane tracks the landfall pressure is determined from HURDAT or from the Ho et al. (1987) report. If post-landfall pressure data are available in HURDAT, we interpolate pressure values over land. If post-landfall pressure data are not available, we apply the Vickery (2005) pressure decay model to the landfall pressure. After the storm exits land, the pressure is based on HURDAT data. Therefore, decay rates for historical hurricanes are based on HURDAT data if available, or the Vickery decay rate model applied to the HURDAT or Ho et al. (1987) landfall *Pmin*, and decay rates for stochastic hurricanes are based on Vickery (2005).

M-6 Logical Relationships of Hurricane Characteristics

A. *The magnitude of asymmetry shall increase as the translation speed increases, all other factors held constant.*

With all other factors held constant, the wind field asymmetry increases with translation speed. The storm translation speed causes a major right-left (looking in the direction the storm is moving) asymmetry in the wind field, which in turn causes an asymmetry in surface friction since the surface stress is wind-speed dependent. The magnitude of the asymmetry increases as the translation speed increases; there is no asymmetry for a stationary storm except for possible land friction effects if a storm becomes stationary while a large percentage of its circulation is over both land and water.

B. *The mean windspeed shall decrease with increasing surface roughness (friction), all other factors held constant.*

With all other factors held constant, the mean wind speed decreases with increasing surface roughness. However, the gust factor, which is used to estimate the peak one-minute wind and the peak three-second gust over the time period corresponding to the model mean wind increases as a function of turbulence intensity, which increases with surface roughness (Paulsen et al., 2003; Masters, 2004; Powell et al., 2004). For roughness values representative of ZIP Codes in Florida, with residential roughness values on the order of 0.2–0.3 m, the roughness effect on decreasing the mean wind speed overwhelms the enhanced turbulence intensity effect that increases the gust factor.

Disclosures

1. ***Describe how the asymmetric structure of hurricanes is represented in the model.***

The asymmetry of the wind field is determined by the storm translation motion (right-left asymmetry) and the associated asymmetric surface friction. A set of form factors for the wind field also contributes to the asymmetry, and the proximity of the storm to land introduces an additional asymmetry because of the effect of land roughness elements on the flow. Azimuthal variation is introduced through the use of two form factors [see Appendix of Powell et al. (2005) for more detail]. The form factors multiply the radial and tangential profiles and provide a "factorized" ansatz for both the radial and tangential storm–relative wind components. Each form factor contains three constant coefficients that are variationally determined in such a way that the ansatz constructed satisfies (as far as its numerical degrees of freedom permit) the scaled momentum equations for the storm-relative polar wind components.

2. ***Provide a completed Form M-3, Radius of Maximum Winds and Radii of Standard Wind Thresholds. Provide a link to the location of the form here.***

See Form M-3.

3. ***Discuss the radii values for each wind threshold in Form M-3 (Radius of Maximum Winds***

and Radii of Standard Wind Thresholds) with reference to available hurricane observations such as those in HURDAT2. Justify the appropriateness of the databases used in the radii validations.

The Extended Best Track Database (EXBT) [DeMuth et al. (2006), available for download from http://rammb.cira.colostate.edu] contains NHC's estimated *Rmax* and advisory outermost radii of hurricane and tropical storm magnitude winds, which are included in the Supplemental Form M-3 Table 33~~Table 33~~. The EXBT does not contain estimates of the 110 mph wind radius (R110), so we found examples of the R110 from the H*Wind archive. We should mention that NHC considers the outer wind radii quality to be poor because of data sparseness, and therefore NHC does not validate wind radii forecasts. Furthermore, the values in Form M-3 and Form M-3 Supplemental represent relatively small samples at particular pressure values, so the ranges in the radii listed on the forms do not represent the full variability of model outputs or observed radii at a given pressure value. Therefore, comparisons are qualitative.

For *Rmax*, the model minima tend to be smaller than the EXBT for storms with *Pmin* of 930 or less but generally compare well for storms with *Pmin* > 930 mb. Model *Rmax* maxima are greater than the EXBT sample for storms with *Pmin* > of 930–950 mb.

For the outer extent of 110 mph winds (R110), for pressures of 940 mb or less, the model radii minima tend to be smaller than the sampled H*Wind values and either above or below for *Pmin* values at or above 950 mb. The model R110 maxima are all larger than the H*Wind sample.

For the radius of hurricane winds (R74), the model minima tend to be smaller than the EXBT sample; the model R74 maxima also tend to be smaller than the EXBT sample for *Pmin* of 920 mb or less or *Pmin* > 950 mb but are larger than the EXBT sample for storms with *Pmin* of 930–950 mb.

For the outer extent of tropical storm winds (R39), the model minima are generally smaller than the EXBT. The R39 maxima are also typically smaller than the EXBT R39 sample, except for the 930–950 mb range of *Pmin*, in which the model has larger radii.

In general the model maximum radii fall within the range of the EXBT or H*Wind values; however, there are instances where model outer radii exceed the EXBT or H*Wind radii. In such cases it is possible that the EXBT and H*Wind radii are not representative of the full range of tropical cyclone radii.

Form M-1: Annual Occurrence Rates

A. Provide annual occurrence rates for landfall from the dataset defined by marine exposure that the model generates by hurricane category (defined by maximum windspeed at landfall in the Saffir-Simpson scale) for the entire state of Florida and selected regions as defined in Figure 3 [of the 2013 ROA]. List the annual occurrence rate per hurricane category. Annual occurrence rates shall be rounded to two decimal places. The historical frequencies below have been derived from the Base Hurricane Storm Set as defined in Standard M-1 (Base Hurricane Storm Set).

A report detailing the how the counts were determined will be available for review.

Statewide counts are determined using two different methods. Under the heading "Entire State," we provide the counts using the most intense landfall for each storm affecting Florida; that is, there is only one landfall per storm. Under the heading "Entire State Landfalls," we provide the counts of all landfalls for each storm, using only one landfall per region. This table is the sum of the counts for Regions A–D.

See Appendix K.

B. Describe model variations from the historical frequencies.

The modeled frequencies are consistent with the historical record, to the extent that we may consider the historical record reliable. Statewide, the model produces 71.4 Florida landfalls (63.6 storms) in 114 years, compared to 71 landfalls (64 storms) historically. For major (Category 3–5) storms, the model produces 24.2 landfalls, compared to about 26 landfalls historically.

On a regional basis, the model is also consistent with the historical record. In Part C below we show bar charts for each region. The bar charts show reasonable agreement between the modeled and historical frequencies. Goodness of fit tests have been performed and indicate that the model results are consistent with the historical record. These tests will be available for review.

C. Provide vertical bar graphs depicting distributions of hurricane frequencies by category by region of Florida (Figure 3 [of the 2013 ROA]) and for the neighboring states of Alabama/Mississippi and Georgia. For the neighboring states, statistics based on the closest milepost to the state boundaries used in the model are adequate.

Vertical bar charts are shown in the figure below. These charts show the number of hurricanes in a 114-year period. Note that there are two charts for Florida statewide hurricanes. The "FL Landfalls" chart shows the total number of landfalls in the state (basically the sum of Regions A–D), whereas the "FL Hurricanes" chart shows only the number of hurricanes making at least one landfall, and the intensity is the maximum intensity landfall in the case of multiple landfalls.

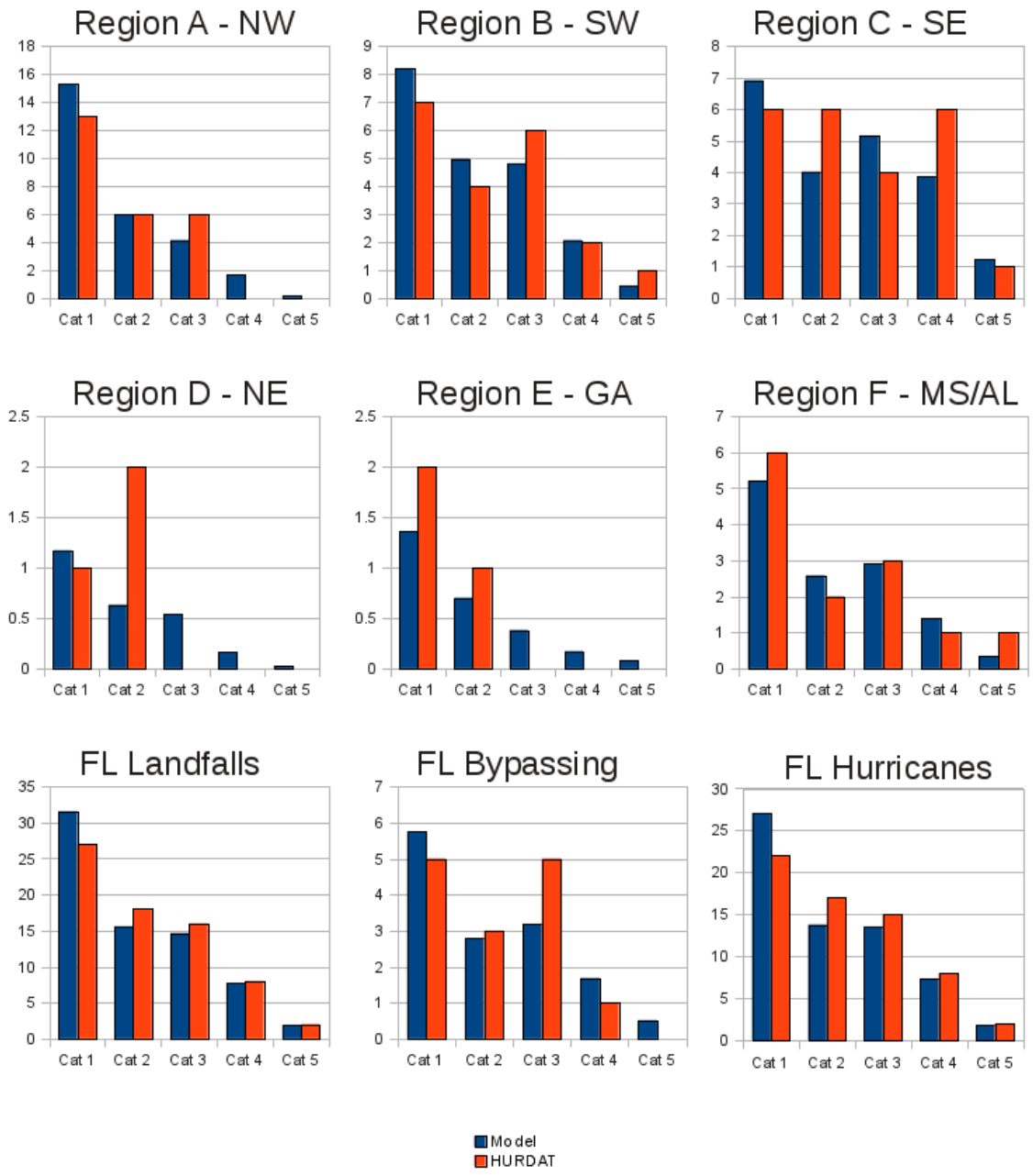


Figure 32. Form M-1 comparison of modeled and historical landfalling hurricane frequency (storms occurring in 114 years) for Regions A–F, FL statewide landfalls (one per FL region), FL bypassing storms, and FL state-wide hurricanes.

D. If the data are partitioned or modified, provide the historical annual occurrence rates for the applicable partition (and its complement) or modification as well as the modeled annual occurrence rates in additional copies of Form M-1 (Annual Occurrence Rates).

Not Applicable.

E. List all hurricanes added, removed, or modified from the previously accepted submission version of the Base Hurricane Storm Set.

Due to recent HURDAT Reanalysis, 6 storms were revised in intensity category or region of impact: 1936 #5, 1941 #5, 1944 #13, 1947 #9, 1948 #9 and King (1950). In addition, two storms were reclassified: the 1926 #10 bypassing storm was upgraded in intensity and Georges (1998) is now considered a landfalling storm rather than bypassing.

F. Provide this form in Excel format. The file name shall include the abbreviated name of the modeling organization, the standards year, and the form name. Form M-1 (Annual Occurrence Rates) shall also be included in a submission appendix.

The form is provided in Excel format. See Appendix J.

Form M-2: Maps of Maximum Winds

A. Provide color maps of the maximum winds for the modeled version of the Base Hurricane Storm Set for land use as set for open terrain and land use as set for actual terrain as defined by the modeling organization.

B. Provide color maps of the maximum winds for a 100-year and a 250-year return period from the stochastic storm set for both open terrain and actual terrain.

C. Plot the position and values of the maximum windspeeds on each contour map.

Actual terrain is the roughness distribution used in the standard version of the model. Open terrain uses the same roughness value of 0.03 meters at all land points.

All maps shall be color coded at the ZIP Code level.

Maximum winds in these maps are defined as the maximum one-minute sustained winds over the terrain as modeled and recorded at each location.

The same color scheme and increments shall be used for all maps.

Use the following eight isotach values and interval color coding:

(1) Minimum damaging	*Blue*
(2) 50 mph	*Medium Blue*
(3) 65 mph	*Light Blue*
(4) 80 mph	*White*
(5) 95 mph	*Light Red*
(6) 110 mph	*Medium Red*
(7) 125 mph	*Red*
(8) 140 mph	*Magenta*

Contouring in addition to these isotach values may be included.

See Appendix L.

Form M-3: Radius of Maximum Winds and Radii of Standard Wind Thresholds

A. *For the central pressures in the table below, provide the minimum and maximum values for 1) the radius of maximum winds (Rmax) used by the model to create the stochastic storm set and minimum and maximum values for the outer radii (R) of 2) Category 3 winds (>110 mph), 3) Category 1 winds (>73 mph), and 4) gale force winds (>40 mph). This information should be readily calculated from the windfield formula input to the model and does not require running the stochastic storm set. Describe the procedure used to complete this Form.*

From the entire set of stochastic track files, 10 sets of track files (totaling 400) were extracted; each set was selected on the basis of the central pressure at landfall as listed in Form M-3. The tracks were processed and the model output from the range of solutions for all times in each track (2600 wind field snapshots) was used to populate the table. Note that the table represents a subset of the possible ranges of *Rmax* because of the selection of landfall tracks close (+/- 0.05 to 0.5 mb) to the pressure values in the table. Note that the *Rmax* values listed also represent model wind field snapshots from when the storms are offshore, while Form M-3 "C" (below) is limited to values at landfall. Input *Rmax* can vary slightly from *Rmax* determined from the gridded wind field because of the effects of translation speed on the wind field and interpolation truncation over the 0.1 *R*/*Rmax* model grid. Observed estimates of *Rmax* and outer wind radii from historical Atlantic basin hurricanes are included for comparison in Table 33~~Table 33~~.

See Appendix M.

B. *Identify the other variables that influence Rmax.*

For our input values of *Rmax* that determine the initial boundary layer mean vortex, we sample *Rmax* from a gamma distribution, which only explicitly depends on central pressure. For *Rmax* determined from the wind field, the translation speed (which is added after the steady state boundary layer model solution is obtained) may also influence *Rmax*.

C. *Provide a box plot and histogram of Central Pressure (x-axis) versus Rmax (y-axis) to demonstrate relative populations and continuity of sampled hurricanes in the stochastic storm set.*

A scatter plot with histograms and box plot is shown below.

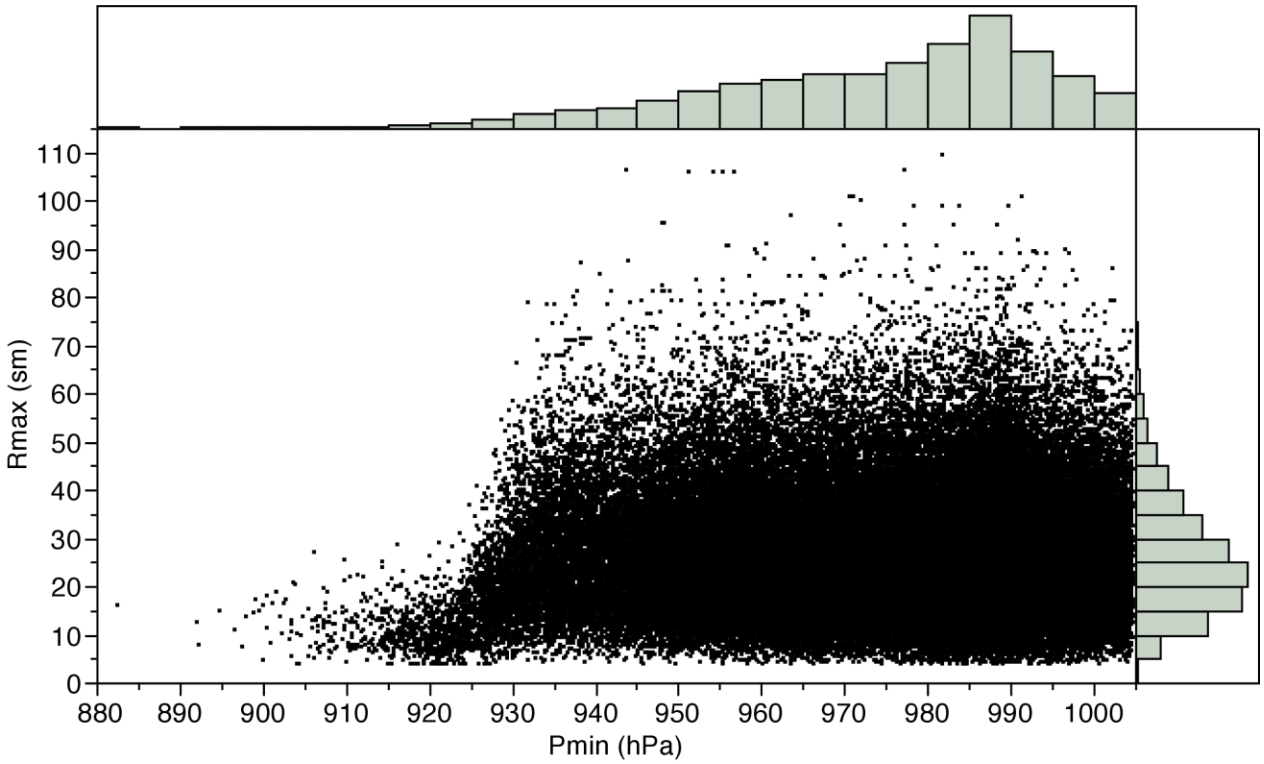


Figure 33. Representative scatter plot of the model input radius of maximum wind (y axis) versus minimum sea-level air pressure at landfall (mb). Relative histograms for each quantity are also shown.

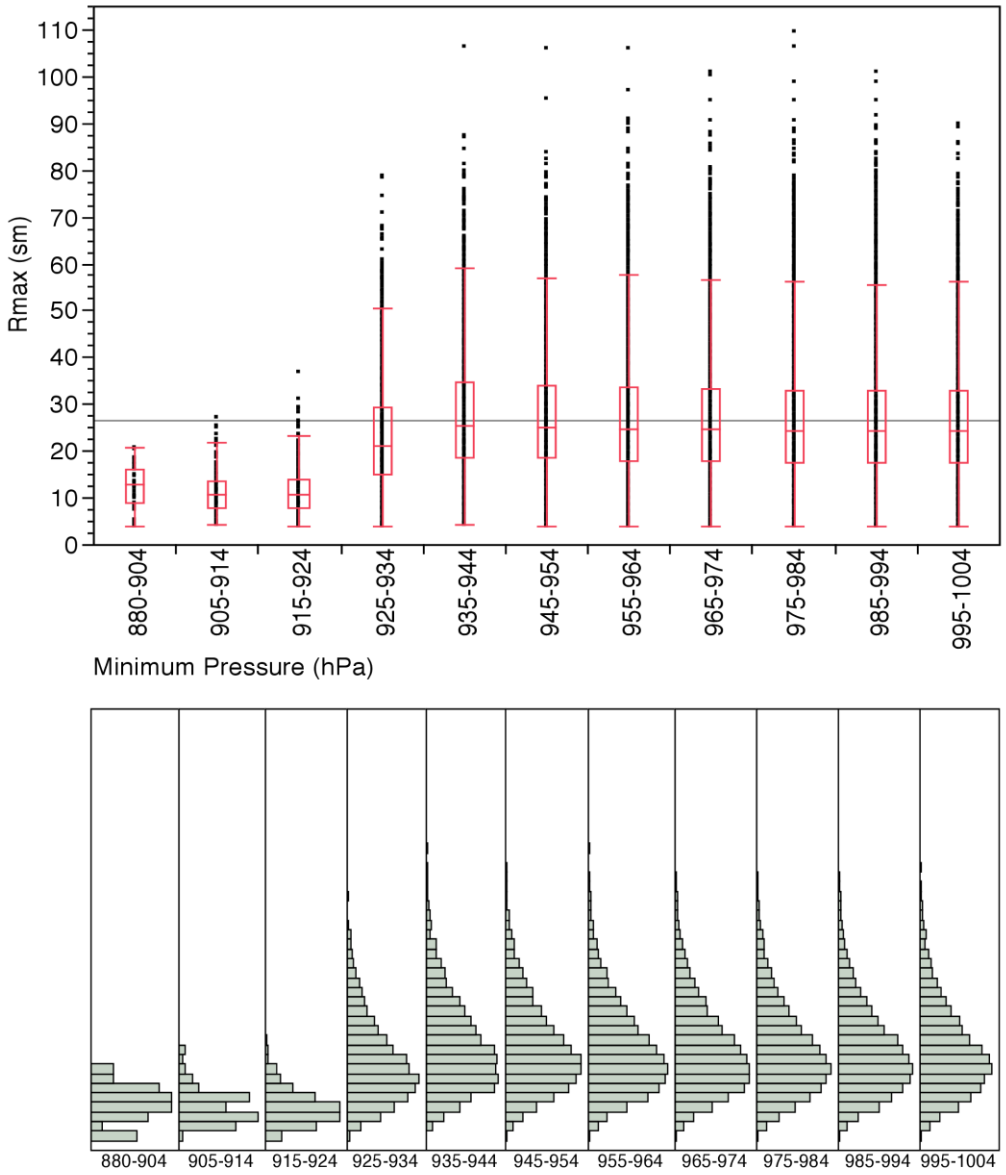


Figure 34. One way box plot (left) of *Rmax* (continuous) response across 10 mb *Pmin* groups. Boxes (and whiskers) are in red; standard deviations are in blue. Histograms (right) for each *Pmin* group.

D. Provide this form in Excel using the format given in the file named "2013FormM3.xlsx." The file name shall include the abbreviated name of the modeling organization, the standards year, and the form name. Form M-3 (Radius of Maximum Winds and Radii of Standard Wind Thresholds) shall also be included in a submission appendix.

The form is provided in Excel format. See Appendix M.

STATISTICAL STANDARDS

S-1 Modeled Results and Goodness-of-Fit

A. ***The use of historical data in developing the model shall be supported by rigorous methods published in currently accepted scientific literature.***

The historical data for the period 1900-2013 were modeled using scientifically accepted methods that have been published in accepted scientific literature.

B. ***Modeled and historical results shall reflect statistical agreement using currently accepted scientific and statistical methods for the academic disciplines appropriate for the various model components or characteristics.***

Modeled and historical results are in agreement as indicated by appropriate statistical and scientific tests. Some of these tests will be discussed below.

Disclosures

1. ***Identify the form of the probability distributions used for each function or variable, if applicable. Identify statistical techniques used for the estimates and the specific goodness-of-fit tests applied. Describe whether the p-values associated with the fitted distributions provide a reasonable agreement with the historical data. Provide a completed Form S-3, Distributions of Stochastic Hurricane Parameters. Provide a link to the location of the form here.***

Form S-3 at the end of this section identifies the form of the probability distribution used for each variable. Some of the methods and distributions are described below.

Historical initial conditions are used to provide the seed for storm genesis in the model. Small uniform random error terms are added to the historical starting positions, intensities and changes in storm motion. Subsequent storm motion and intensity are determined by randomly sampling empirical probability distribution functions derived from the HURDAT historical record.

Figure 35 shows the occurrence rate of both modeled and historical land-falling hurricanes in Florida. The figure shows a high level of agreement between historical and modeled occurrences. A chi-square goodness-of-fit test, for the number of years with 0, 1, and 2 or more hurricanes per year (4 bins each with 5 or more occurrences giving 3 degrees of freedom), gives a *p*-value of approximately 0.483. A comparison of landfalls by region and intensity is given in Form M1. The modeled results are consistent with the historical record, especially given the large uncertainty in the historical observations~~. Goodness of fit tests will be available for review~~.

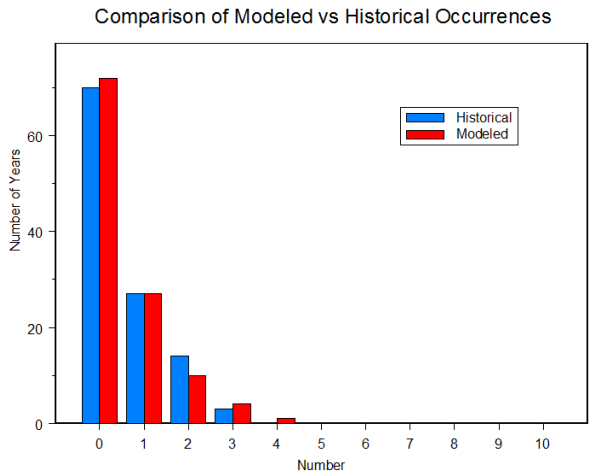


Figure 35. Comparison of modeled vs. historical occurrences.

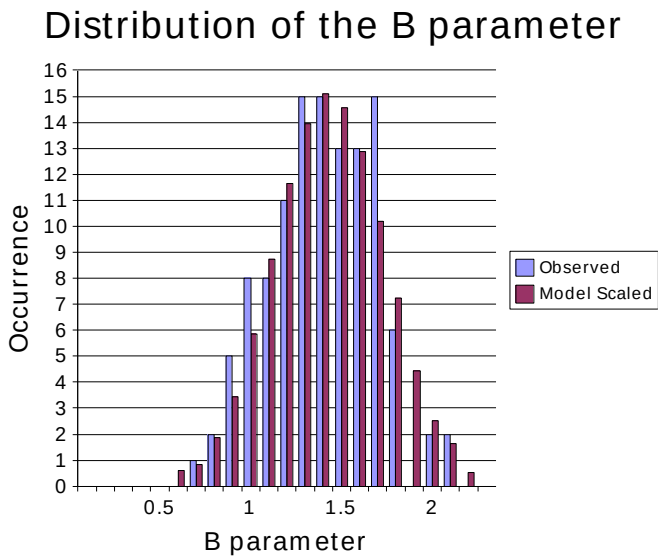


Figure 36. Comparison between the modeled and observed Willoughby and Rahn (2004) *B* data set.

The random error term for the *Holland B* is modeled using a Gaussian distribution with a standard deviation of 0.286. Figure 36 shows a comparison between the Willoughby and Rahn (2004) *B* data set (see Standard M-2.1) and the modeled results (scaled to equal the 116 measured occurrences in the observed data set). The modeled results with the error term have a mean of about 1.38 and are consistent with the observed results. The figure indicates a high level of agreement, and the chi-square goodness-of-fit test gives a *p*-value about 0.57, using 8 degrees of freedom (re-binning to 11 bins and two estimated parameters). A KS goodness-of-fit yields a *p*-value of 0.845 (ks=0.057).

We developed an *Rmax* model using 106 measurements from the revised landfall *Rmax* database which includes observations for storms up to 2012. We have opted to model the *Rmax* at landfall rather than the entire basin for a variety of reasons. One is that the distribution of landfall *Rmax* may be different from the *Rmax* distribution over open water. An analysis of the landfall *Rmax* database and the 1988-2007 DeMaria Extended Best Track data show that there appears to be a difference in the dependence of *Rmax* on central pressure (*Pmin*) between the two data sets. The landfall data set provides a larger set of independent measurements, which is more than 100 storms compared to about 31 storms affecting the Florida threat area region in the Best Track Data. Since landfall *Rmax* is most relevant for loss cost estimation, and has a larger independent sample size, we have chosen to model the landfall data set. Future studies will examine how the Extended Best Track Data can be used to supplement the landfall data set.

Based on the skewness of Rmax and the fact that it is nonnegative, we sought to model the distribution using a gamma distribution. Using the maximum likelihood estimation method, we found the estimated shape and scale parameters for the gamma distribution are 4.76 and 5.41 respectively. Using these estimated values, we plotted the observed and expected distribution in Figure 37. The Rmax values are binned in 5 sm intervals, with the x-axis showing the end value of the interval.

Plot of Observed *Rmax* vs. Gamma Distribution

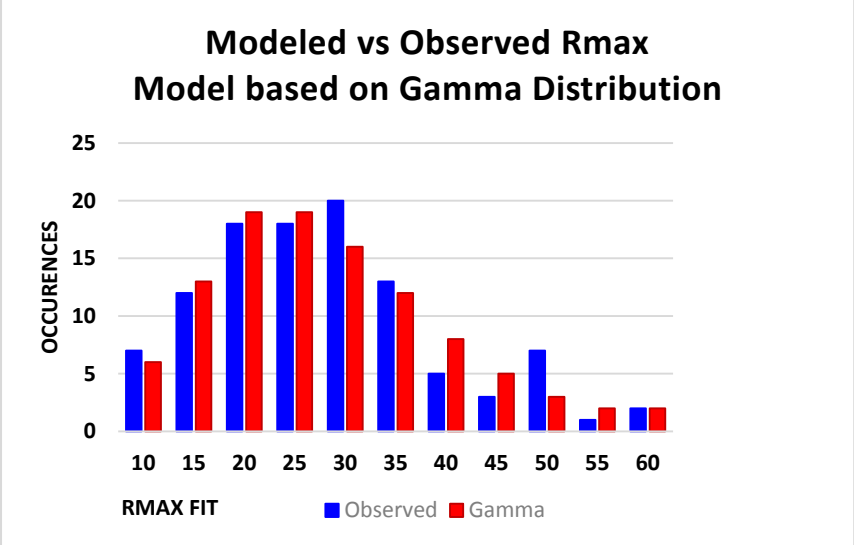


Figure 37. Observed and expected distribution using a gamma distribution.

The gamma distribution showed a reasonable fit. A chi-square goodness of fit test shows A chi-square goodness-of-fit test yields a *p*-value of 0.59 with 6 degrees of freedom (re-binning to 9 bins to ensure more than 5 expected occurrences per bin and 2 estimated parameters.) The KS goodness-of-fit yields a *p*-value of 0.8327 (ks= 0.0605).

2. ***Describe the nature and results of the tests performed to validate the windspeeds generated.***

We compared the cumulative effect of a series of modeled and observed wind fields by comparing the peak winds observed at a particular ZIP Code during the entire storm life-cycle. We also compared our modeled wind fields to those that have been constructed from all available observations which are freely available on the NOAA AOML-HRD web site. A subsequent section describes the process for recording the peak modeled and observed wind speeds (wind swaths) from which the validation statistics are generated. Our validation is based on nine hurricanes that passed by or made landfall in Florida. These hurricanes were well-observed. We will have the ability to add new storms and quickly conduct new validation studies as our validation set grows and we make enhancements to the model. In order to run the Loss Model in "scenario" mode for doing validation studies, we had to construct detailed storm track histories for recent storms affecting Florida using the HURDAT, *Rmax* and *Holland B* databases. The validation suite included 1992 Hurricane Andrew and the following 2004 and 2005 storms: Charley, Frances, Jeanne, Ivan, Dennis, Katrina, Rita, and Wilma. The validations make use of the Hurricane Research Division's Surface Wind Analysis System (H*Wind).

H*WIND

The HRD approach to hurricane wind analysis employed in H*Wind evolved from a series of peer-reviewed, scientific publications analyzing landfalls of major hurricanes including Frederic of 1979, Alicia of 1983, Hugo of 1989, and Andrew of 1992 (Powell et al., 1991; Powell et al., 1996; Powell et al., 1998). In Powell et al. (1991) which described Hurricane Hugo's landfall, a concept was developed for conducting a real-time analysis of hurricane wind fields. The system was first used in real-time during Hurricane Emily in 1993 (Burpee et al., 1994). Since 1994, HRD wind analyses have been conducted on a research basis to create real time hurricane wind field guidance for forecasters at the National Hurricane Center. During hurricane landfall episodes from 1995-2005, HRD scientists have conducted research side by side with hurricane specialists at NHC analyzing wind observations on a regular 3 or 6 hour schedule consistent with NHC's warning and forecast cycle.

An HRD wind analysis requires the input of all available surface weather observations (e.g., ships, buoys, coastal platforms, surface aviation reports, reconnaissance aircraft data adjusted to the surface, etc.). Observational data are downloaded on a regular schedule and then processed to fit the analysis framework. This includes the data sent by NOAA P3 and G4 research aircraft during the HRD hurricane field program, including the Step Frequency Microwave Radiometer measurements of surface winds and U.S. Air Force Reserves (AFRES) C-130 reconnaissance aircraft, remotely sensed winds from the polar orbiting SSM/I and ERS, the QuikScat platform and TRMM microwave imager satellites, and GOES cloud drift winds derived from tracking low level near-infrared cloud imagery from geostationary satellites. These data are composited relative to the storm over a 4-6 hour period. All data are quality controlled and processed to conform to a common framework for height (10 m or 33 feet), exposure (marine or open terrain over land), and averaging period (maximum sustained 1 minute wind speed) using accepted methods from micrometeorology and wind engineering (Powell et al., 1996). This framework is consistent with that used by the National Hurricane Center (NHC) and is readily converted to wind load frameworks used in building codes.

Based on a qualitative examination of various observing platforms and methods used to standardize observations, Powell et al. (2005) suggest that the uncertainty of the maximum wind from a given analysis ranges from 10-20% depending on the observing platform. In general the uncertainty of a given H*Wind analysis is of the order of 10% for analysis of Hurricanes Ivan, Frances, Jeanne, and Katrina, all of which incorporated more accurate surface wind measurements from the Stepped Frequency Microwave Radiometer (SFMR) aboard the NOAA research aircraft. The SFMR data used for those analyses was post-processed during the fall of 2005 using the latest geophysical model function relating wind speed to sea surface foam emissivity. Hurricanes Charley, Dennis, Rita, Wilma, and Andrew did not have the benefit of SFMR measurements but relied on adjusting Air Force reconnaissance observations at the 3 km altitude to the surface with empirical reduction methods. The method used was based on how SFMR measurements compared to flight level winds and depended on storm relative azimuth. Preliminary results suggest that this method has an uncertainty of 15%.

We created wind swaths for both the modeled and observed winds. We also computed the maximum winds at ZIP Codes for both the observed and modeled winds; from that we derived the mean and root-mean-square error (see Table 11 and Table 12).

WIND SWATHS

For each storm in the validation set, the peak sustained surface wind speed is recorded at each ZIP Code in Florida for the duration of the storm event. Observed wind fields from H*Wind and modeled wind fields from the public model are moved along the exact same tracks, which are the observed high-resolution storm tracks assembled from reconnaissance aircraft and radar data. For each storm, the recorded peak of the observed and modeled wind speed is saved at each grid point and each ZIP Code, and the resulting ZIP Code comparison pairs provide the basis for the model validation statistics. The peak grid point values are color contoured and mapped as graphics showing the "swath" of maximum winds swept out by the storm passage. Wind swaths are sometimes confused with wind fields. The winds depicted in a wind swath do not have time continuity, cannot depict a circulation, and therefore cannot be described as a wind field. A wind field represents a vector field that represents a representative instance of the surface wind circulation.

Wind swaths were constructed for both the modeled and observed winds. Maximum marine exposure winds were compared at all ZIP Codes for both the observed and modeled winds (Figure 38) from which we derived the mean and root-mean-square error statistics shown in Table 11 and Table 12. This type of comparison provides an unvarnished assessment of model performance.

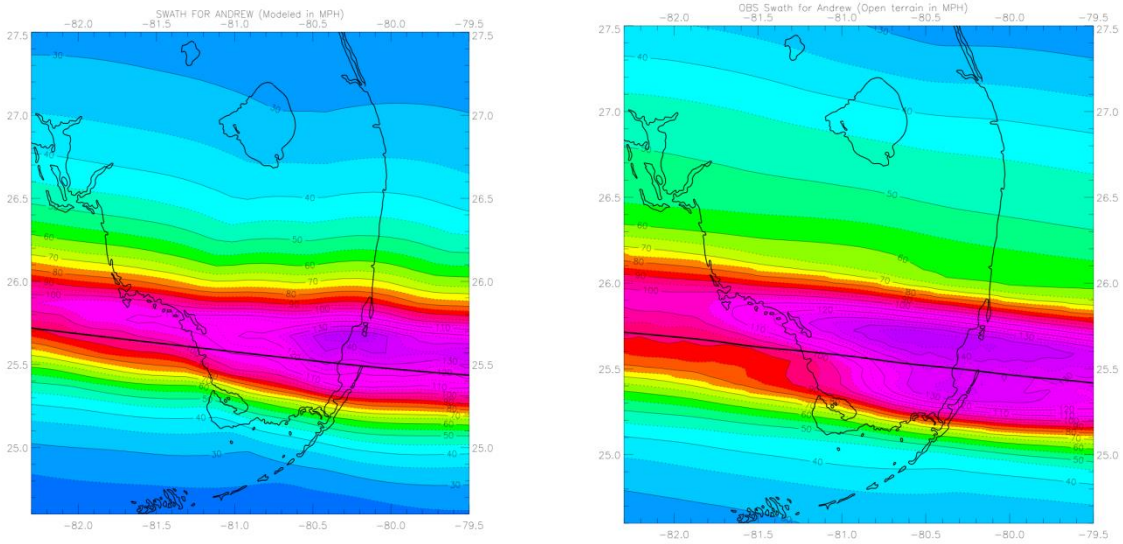


Figure 38. Comparison of modeled (left) and observed (right) swaths of maximum sustained marine surface winds for Hurricane Andrew of 1992 in South Florida. The Hurricane Andrew observed swath is based on adjusting flight-level winds with the SFMR-based wind reduction method.

Table 11. Validation Table based on ZIP Code wind swath comparison of the Public wind field model to H*Wind. Mean errors (bias) of model for the set of validation wind swaths. Errors (upper number in each cell) are computed as Modeled – Observed (Obs) at ZIP Codes were modeled winds were within wind thresholds (model threshold) or where observed winds were within respective wind speed threshold (H*Wind threshold). Number of ZIP Codes for the comparisons is indicated as the lower number in each cell.

Storms	Year	56-74 Model Threshold	75-112 Model Thresh.	>112mph Model Thresh.	>56mph Model Thresh.	56-74 H*Wind Thresh.	75-112 H*Wind Thresh.	>112mph H*Wind Thresh.	>56mph H*Wind Thresh.
Andrew	1992	5.25 92	13.86 107	2.73 100	7.49 299	10.26 139	12.47 54	0.66 88	7.68 281
Charley	2004	12.96 112	21.36 244	-7.36 13	17.80 369	8.58 122	-3.09 63	-8.91 17	3.47 202
Frances	2004	3.99 693	-0.99 96	None	3.38 789	-0.59 372	-4.48 96	None	-1.38 468
Ivan	2004	-6.95 20	-3.35 38	None	-4.59 58	-5.76 22	-3.73 41	None	-4.44 63
Jeanne	2004	6.78 250	3.95 190	None	5.56 440	2.67 225	-3.87 121	None	0.38 346
Dennis	2005	2.45 15	6.98 46	None	5.87 61	5.22 29	7.57 29	-4.37 3	5.87 61
Dennis Keys	2005	None	None	None	None	-12.65 5	None	None	-12.65 5
Katrina	2005	-11.43 77	-2.42 100	None	-6.34 177	-8.93 93	-11.57 149	None	-10.55 242
Rita	2005	6.28 5	14.54 3	None	9.38 8	12.01 5	None	None	12.01 5
Wilma	2005	0.44 133	-9.99 394	None	-7.35 527	6.54 87	-13.35 396	None	-9.77 483

Table 12. Validation Table based on ZIP Code wind swath comparison of the Public wind field model to H*Wind. Root mean square (RMS) wind speed errors (mph) of model for the set of validation wind swaths. Errors are based on Modeled – Observed (Obs) at ZIP Codes where modeled winds were within wind thresholds (model threshold) or where observed winds were within respective wind speed threshold (H*Wind threshold).

Storms	Year	56-74 Model Threshold	75-112 Model Thresh.	>112mph Model Thresh.	>56mph Model Thresh.	56-74 H*Wind Thresh.	75-112 H*Wind Thresh.	>112mph H*Wind Thresh.	>56mph H*Wind Thresh.
Andrew	1992	6.11	15.75	7.024	10.81	12.19	14.26	5.82	11.10
Charley	2004	19.84	26.59	10.08	24.30	16.65	8.60	11.69	14.21
Frances	2004	8.08	11.20	None	8.52	4.99	10.20	None	6.41
Ivan	2004	7.07	5.20	None	5.91	6.11	5.51	None	5.72
Jeanne	2004	10.14	9.65	None	9.93	10.88	6.16	None	9.50
Dennis	2005	3.06	9.19	None	8.12	6.15	9.93	4.59	8.12
Dennis Keys	2005	None	None	None	None	12.67	None	None	12.67
Katrina	2005	14.66	8.25	None	11.49	12.50	17.97	None	16.09
Rita	2005	6.4992	14.54	None	10.28	12.41	None	None	12.41
Wilma	2005	14.73	14.05	None	14.22	12.51	14.83	None	14.44
RMS N	All	10.18 1397	14.87 1218	6.26 113	12.37 2728	9.75 1099	12.79 949	6.71 108	11.19 2156

Comparison of model and H*Wind sustained marine exposure wind speeds at ZIP Codes receiving model wind speeds over the given thresholds (Table 11) indicates a positive bias. For ZIP Codes where model wind speeds exceeded 56 mph, the bias is +3.3 mph ; negative bias was apparent in Hurricanes Ivan, Katrina, and Wilma. At other wind speed thresholds, low bias is evident for winds > 112 mph in Hurricane Charley, and winds of 75-112 mph in Hurricanes Frances, Ivan, Katrina, and Wilma. For winds of 56-74 mph, low bias is noted in Hurricanes Ivan, and Katrina. Errors for Hurricane Andrew are relatively high, but the lack of observations for Hurricane Andrew makes it difficult to determine if it was a Cat 4 or Cat 5 hurricane during its landfall in South Florida. Hurricane Rita in the Keys also shows relatively high bias, but observations indicate that there were fluctuations in intensity over a short period of time during its passage past the Keys. Model errors for Hurricane Charley are also relatively high, likely due to the model producing a wind field that was too broad. When model winds are compared to H*Wind at ZIP Codes exceeding H*Wind and sustained wind speed thresholds of 56 mph are considered, the mean bias is -2.2 mph. However, bias at other wind speed thresholds is larger, primarily caused by large model - H*Wind differences in Hurricanes Andrew, Charley, and Rita.

When swaths are evaluated at ZIP Codes, a positive wind speed bias of ~3 mph is indicated. However, the model can also under-predict swaths for individual cases. While bias correction is an accepted practice for numerical weather prediction, there is no evidence that the model has a consistent bias. The swath bias is probably associated with limitations in specifying the radial pressure profile after landfall. The tendency for the Holland pressure profile parameter to produce too broad an area of strong winds near the eyewall is the most likely cause of bias and is likely a feature found in many of the current risk models. Therefore we have decided to forgo any corrective measures at this point.

Our validation set is unique in that the values of storm position, motion, *Rmax* and *Pmin* are observed, and *B* is determined independently from the H*Wind field. In other words, it is impossible to fine-tune our results. Although additional validation storms are desired, we believe the positive bias for locations with winds > 56 mph is a characteristic of models that use the *Holland B* pressure profile parameter, which tends to produce model fields that are too broad outside the radius of maximum winds. Our validation method provides an objective means of assessing model performance by evaluating the portion of the wind field that contains damaging winds.

The root mean square (RMS) error (Table 12) provides a better estimate of model uncertainty. For ZIP Codes in which model winds were 56-74 mph, the RMS error is +/- 10 mph (~ 15%), for 75-112 mph the error is +/- 15 mph (~16%), and for winds > 112 mph the errors is +/- 6 mph (~ 5%). In general, for winds > 56 mph, the RMS error is +/- 12 mph or ~ 13%. RMS errors are similar for ZIP Codes in which H*Wind wind speeds fell into the respective thresholds.

SUMMARY OF WIND SWATH VALIDATION

Validation of the winds from the wind model against the H*WIND analyses was prepared by considering winds that would be strong enough to be associated with damage. Threshold-based comparisons could miss places where the observed winds were greater than the model and the model was below the threshold. Conversely, observed winds over the same thresholds can be compared to the co-located model grid points but would miss places where the observed winds

were below the threshold. It is important to evaluate the errors both ways to see if a consistent bias is evident. According to our validation statistics, albeit for a relatively small number of cases, wind swath ZIP Code comparisons show evidence of a 3 mph positive bias, but it is not consistent for all storms. The bias is likely related to the limitations of the *Holland B* pressure profile specification. The model uncertainty, as estimated by the RMS error, is on the order of 15%.

3. ***Provide the date of loss of the insurance company data available for validation and verification of the model.***

The following hurricane data from different insurance companies are used to validate the model:

Andrew	1992
Erin	1995
Charley	2004
Frances	2004
Jeanne	2004
Dennis	2005
Wilma	2005
Katrina	2005

4. ***Provide an assessment of uncertainty in loss costs for output ranges using confidence intervals or other accepted scientific characterizations of uncertainty.***

While the model does not automatically produce confidence intervals for the output ranges, the data do allow for the calculation of confidence intervals. We calculated the mean and the standard deviation of the losses for each county, and it was found that the standard errors were within 2.5% of the means for all counties. We also calculated the coefficient of variation (CV) for all counties and drew a histogram which is provided in Figure 39. The range of the CVs was between 2.73 and 5.04. Finally, we computed 95% confidence intervals for the average loss for each county. Some of these intervals are reproduced in Table 13.

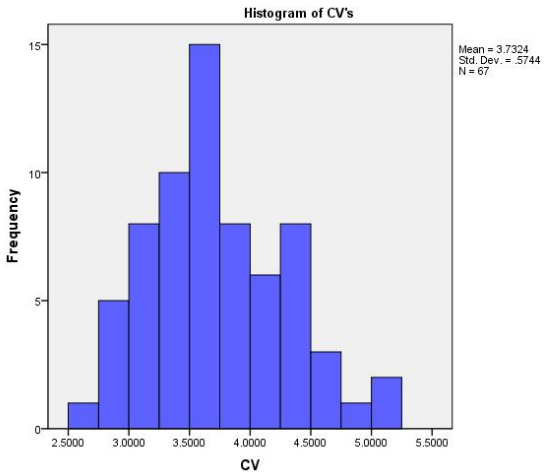


Figure 39. Histogram of CVs for all counties combined.

Table 13. 95% Confidence intervals for mean loss for selected counties (based on 57,000) year simulation.

county	average_loss	stdev_loss	LCL	UCL
Alachua	$11,169,611.60	$44,601,249.27	10803456.01	11535767.19
Brevard	$139,311,789.58	$510,248,609.36	135122884.9	143500694.2
Broward	$426,828,471.38	$1,260,130,308.57	416483385.5	437173557.3
Duval	$38,916,104.37	$171,434,361.95	37508707.7	40323501.04
Escambia	$39,847,556.64	$135,856,703.97	38732236.06	40962877.22
Gulf	$1,724,724.94	$5,957,062.07	1675820.221	1773629.659
Hamilton	$208,068.87	$1,045,196.16	199488.2939	216649.4461
Hillsborough	$191,265,100.55	$622,086,874.69	186158055.6	196372145.5
Jackson	$1,840,534.59	$6,940,982.96	1783552.336	1897516.844
Jefferson	$387,082.01	$1,925,992.02	371270.508	402893.512
Lee	$188,966,509.94	$520,708,638.81	184691733.3	193241286.6
Leon	$10,932,373.90	$48,948,483.98	10530529.53	11334218.27
Madison	$368,149.21	$1,856,878.82	352905.0954	383393.3246
Miami-Dade	$463,773,194.18	$1,401,372,085.16	452268578.8	475277809.6
Monroe	$69,389,783.84	$207,638,466.46	67685168.25	71094399.43
Nassau	$4,886,760.81	$21,816,995.79	4707653.385	5065868.235
Okeechobee	$8,775,340.91	$29,956,287.85	8529413.668	9021268.152
Osceola	$150,388,255.66	$537,527,141.98	145975406.9	154801104.4
Palm Beach	$31,630,062.09	$107,622,237.03	30746533.4	32513590.78
Sarasota	$115,326,276.19	$346,694,082.83	112480078.4	118172474

LCL: 95% Lower Confidence Limit for the Average Loss
UCL: 95% Upper Confidence Limit for the Average Loss

5. ***Justify any differences between the historical and modeled results using current accepted scientific and statistical methods in the appropriate disciplines.***

The various statistical tests as well as other validation tests presented here and elsewhere indicate that any differences between modeled results and historical observations are not statistically significant given the large known uncertainties in the historical record.

6. ***Provide graphical comparisons of modeled and historical data and goodness-of-fit tests. Examples include hurricane frequencies, tracks, intensities, and physical damage.***

For hurricane frequencies as a function of intensity by region, see Form M-1 plots and the goodness-of-fit table. The histogram in Figure 35 compares the modeled and historical annual landfall distribution by number of events per year. The agreement between the two distributions is quite close and the histogram shows a good fit. The chi-square goodness-of-fit test gives a *p*-value of approximately ~~0.749~~ 0.483 as described in S-1.1. Plots and goodness-of-fit tests for the radius of maximum wind and the Holland pressure profile parameter are shown in Disclosure 1 of this standard. Plots and statistical comparisons of historical and modeled losses are shown in Standard S-5, Form S-4 and Form S-5.

7. ***Provide a completed Form S-1, Probability and Frequency of Florida Landfalling Hurricanes per Year. Provide a link to the location of the form here.***

Please see completed Form S-1 at the end of this section.

8. ***Provide a completed Form S-2A, Examples of Loss Exceedance Estimates, using the 2007 Florida Hurricane Catastrophe Fund aggregate personal and commercial residential exposure data. Provide a link to the location of the form here.***

Please see completed Form S-2A at the end of this section.

9. ***Provide a completed Form S-2B, Examples of Loss Exceedance Estimates, using the 2012 Florida Hurricane Catastrophe Fund aggregate personal and commercial residential exposure data. Provide a link to the location of the form here.***

Please see completed Form S-2B at the end of this section.

S-2 Sensitivity Analysis for Model Output

The modeling organization shall have assessed the sensitivity of temporal and spatial outputs with respect to the simultaneous variation of input variables using currently accepted scientific and statistical methods in the appropriate disciplines and have taken appropriate action.

We have performed sensitivity analysis on the temporal and spatial outputs of the model using currently accepted scientific and statistical methods. We examined the effects of five input variables on the expected loss cost. The input variables were as follows:

CP = central pressure (in millibars)
Rmax = radius of maximum winds (in statute miles)
VT = translational velocity (forward speed in miles per hour)
Holland B = pressure profile parameter and
FFP = far field pressure

The effects of the above input variables on the expected loss cost were examined using the methods described by Iman et al. (2000a).

Disclosures

1. ***Identify the most sensitive aspect of the model and the basis for making this determination. Provide a full discussion of the degree to which these sensitivities affect output results and illustrate with an example.***

Figure 40 provides the graph of the standardized regression coefficients of the expected loss cost as a function of the input variables for Category 1, 3 and 5 hurricanes. From the graph, we observe that the sensitivity of expected loss cost depends on the category of the hurricanes. For a Category 1 hurricane, expected loss cost is most sensitive to Holland B parameter followed by *FFP* and then *CP*. For a Category 3 hurricane, expected loss cost is most sensitive to Holland B followed by *FFP* and *Rmax* and finally for a Category 5 hurricane, expected loss cost is most sensitive to *Rmax*, followed by Holland B and then *CP* and *FFP*. The expected loss cost is least sensitive to *Rmax* for Category 1, while the expected loss cost is least sensitive to VT for Category 3 and 5.

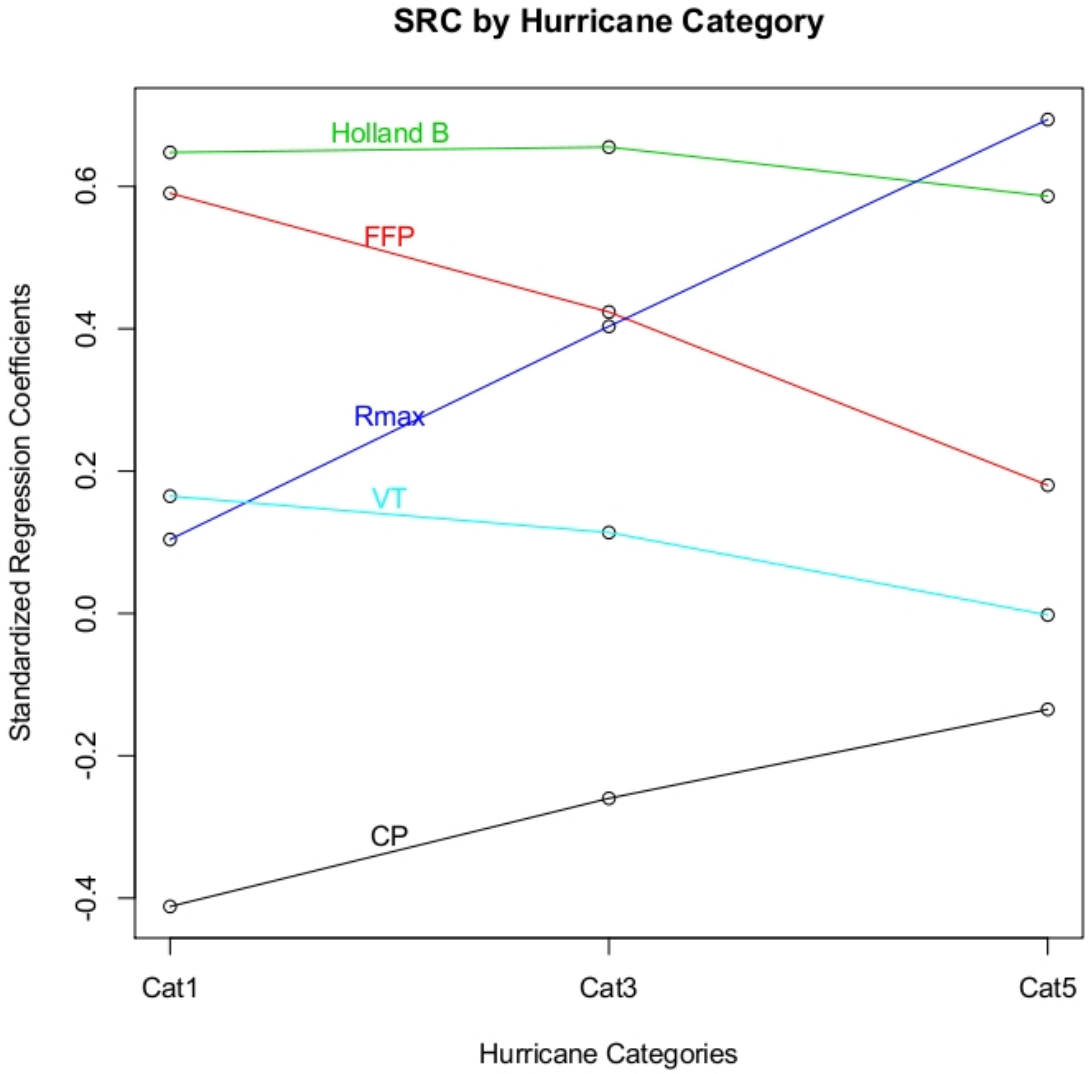


Figure 40. SRCs for Expected Loss Cost for all Input Variables for all Hurricane Categories.

2. ***Describe how other aspects of the model may have a significant impact on the sensitivities in output results and the basis for making this determination.***

Validation studies (described in Standard S-1.2) indicated that air density, boundary layer height, fraction of the boundary layer depth over which the turbulent stresses act, the drag coefficient, the averaging time chosen to represent the boundary layer slab winds, and the conversion of the 0-500 m layer mean wind to 10 m surface wind could all have a significant impact on the output. These quantities were evaluated during the validation process, resulting in the selection of physically consistent values. For example, the values chosen for air density, marine boundary layer height and reduction factor from the mean boundary layer to the surface are representative of near surface GPS dropsonde measurements in hurricanes.

Model wind speeds (and therefore, output results) are very sensitive to surface roughness, which in turn depend on land use/land cover determined from satellite remote sensing. The assignment of roughness to mean land use / land cover classifications as well as the upstream filtering or weighting factor was applied to integrate the upstream roughness elements within a 45 degree sector to windward of the corresponding ZIP Code.

3. ***Describe and justify action or inaction as a result of the sensitivity analyses performed.***

No actions were taken in light of the aforementioned sensitivity experiments.

4. ***Provide a completed Form S-6, Hypothetical Events for Sensitivity and Uncertainty Analysis. (Requirement for models submitted by modeling organizations which have not previously provided the Commission with this analysis. For models previously found acceptable, the Commission will determine, at the meeting to review modeling organization submissions, if an existing modeling organization will be required to provide Form S-6 (Hypothetical Events for Sensitivity and Uncertainty Analysis) prior to the Professional Team on-site review). If applicable, provide a link to the location of the form here.***

Please see the completed Form S-6 at the end of this section.

S-3 Uncertainty Analysis for Model Output

The modeling organization shall have performed an uncertainty analysis on the temporal and spatial outputs of the model using currently accepted scientific and statistical methods in the appropriate disciplines and have taken appropriate action. The analysis shall identify and quantify the extent that input variables impact the uncertainty in model output as the input variables are simultaneously varied.

We have performed uncertainty analysis on the temporal and spatial outputs of the model using currently accepted scientific and statistical methods. We examined the effects of five input variables on the expected loss cost. The input variables were as follows:

CP = central pressure (in millibars)
R*max* = radius of maximum winds (in statute miles)
VT = translational velocity (forward speed in miles per hour)
Holland B = pressure profile parameter and
FFP = far field pressure

The effects of the above input variables on the expected loss cost were examined using the methods described by Iman et al. (2000b).

Disclosures

1. ***Identify the major contributors to the uncertainty in model outputs and the basis for making this determination. Provide a full discussion of the degree to which these uncertainties affect output results and illustrate with an example.***

Figure 41 gives the expected percentage reductions in the variance of expected loss costs for Category 1, 3 and 5 hurricanes as a function of the input variables. As with the sensitivity analysis, the category of the hurricane determines which variables contributes most to the uncertainty of the expected loss costs. For a Category 1 hurricane, the major contributor to the uncertainty in expected loss cost is the Holland B parameter followed by *FFP* and then *CP*. For a Category 3 hurricane, the major contributor to the uncertainty in loss costs is Holland B followed by *Rmax* and then *FFP* and finally for a Category 5 hurricane, the major contributor to the uncertainty of expected loss costs is *Rmax* followed by Holland B and then *FFP* and *CP*. The variable *VT* has negligible effect on the uncertainty in expected loss costs.

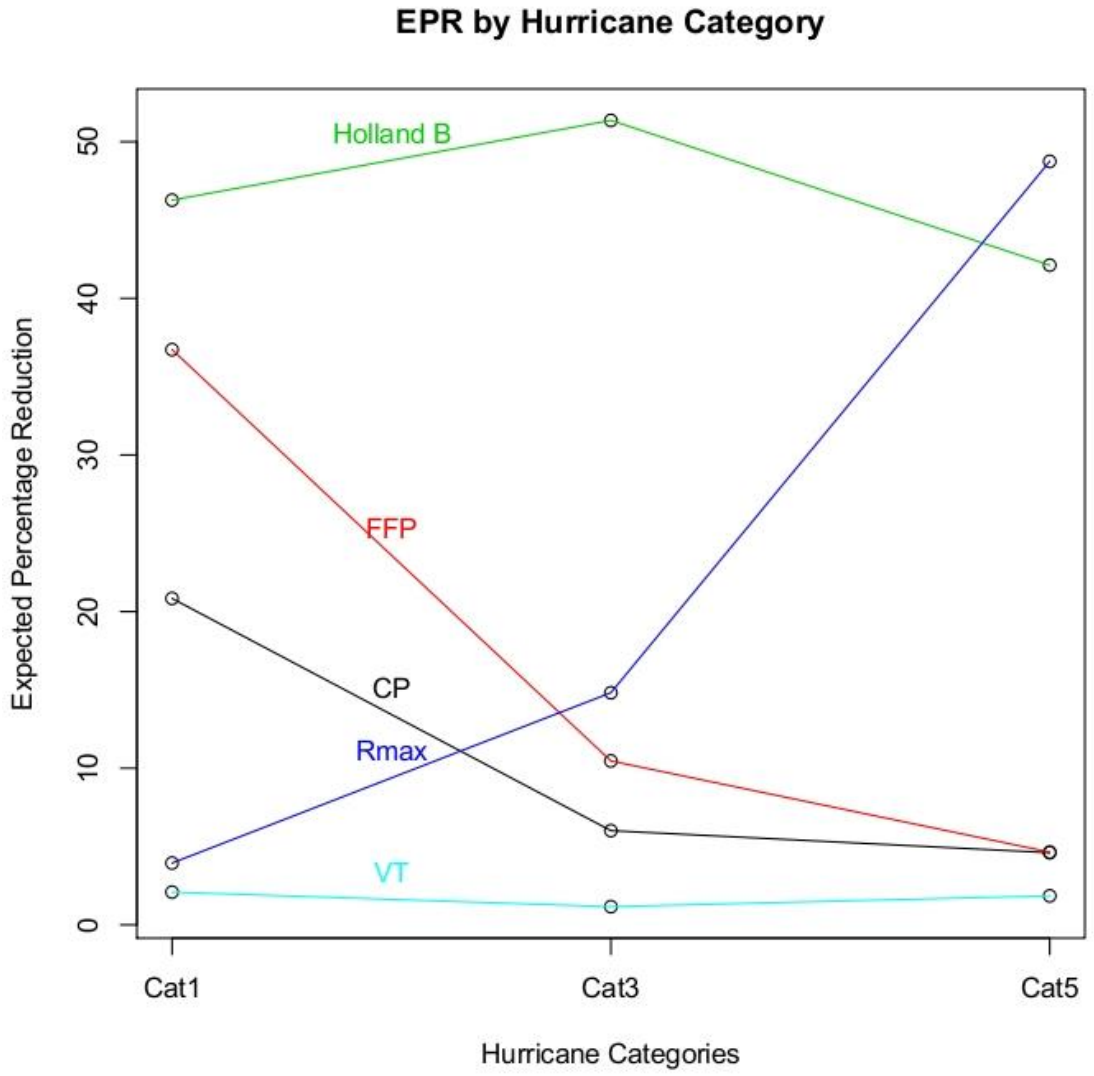


Figure 41. EPRs for Expected Loss Cost for all Input Variables for all Hurricane Categories.

2. ***Describe how other aspects of the model may have a significant impact on the uncertainties in output results and the basis for making this determination.***

Limitations in the HURDAT record contribute to the uncertainty of modeled tracks and pressures. Surface pressure measurements are not always available in HURDAT and estimating surface pressures by pressure-wind relationships is also fraught with uncertainty since well-observed hurricanes can demonstrate a large variation in maximum wind speeds for a given minimum surface pressure. The HURDAT record prior to the advent of satellites in the mid 1960s could have missed or incorrectly classified many hurricanes that affected Florida in the early 20th century. Even today, there is still considerable uncertainty in the assessment of hurricane intensity. Recent research results based on SFMR measurements (Powell et al., 2009) indicate that some Saffir-Simpson 1-3 Category hurricanes may be rated too highly while the Category 4 and 5 storms are probably rated accurately.

Uncertainty in surface roughness has a significant impact on wind uncertainty which in turn leads to a significant impact on losses.

3. ***Describe and justify action or inaction as a result of the uncertainty analyses performed.***

No actions were taken in light of the aforementioned uncertainty analysis.

4. ***Form S-6 (Hypothetical Events for Sensitivity and Uncertainty Analysis), if disclosed under Standard S-2 (Sensitivity Analysis for Model Output), will be used in the verification of Standard S-3 (Uncertainty Analysis for Model Output).***

Please see the completed Form S-6 at the end of this section.

S-4 County Level Aggregation

At the county level of aggregation, the contribution to the error in loss cost estimates attributable to the sampling process shall be negligible.

The error in the county level loss costs induced by the sampling process can be quantified by computing standard errors for the county level loss costs. These loss costs have been computed for all counties in the state of Florida using 57,000 years of simulation. The results indicate that the standard errors are less than 2.5% of the average loss cost estimates for all counties.

Disclosure

1. ***Describe the sampling plan used to obtain the average annual loss costs and output ranges. For a direct Monte Carlo simulation, indicate steps taken to determine sample size. For an importance sampling design, describe the underpinnings of the design.***

The number of simulation years was determined through the following process:

The average loss cost, $\bar{X}_Y$, and standard deviation S_Y, were determined for each county Y using an initial run of an 11,400 year simulation. Then the maximum error of the estimate will be 2.5% of the estimated mean loss cost, if the number of simulation years for county Y is:

$$N_Y = \left(\frac{s_Y}{0.025\bar{\bar{X}}_Y} \right)^2$$

Based on the initial 11,400 year simulation runs, the minimum number of years required is N_Y = 41,667 for Madison County, which had the highest number of years required of all the counties. Therefore, we have decided to use 57,000 (500x114) years of simulation for our final results. For the 57,000 year simulation run we found that the standard errors are less than 2.5% of the average loss costs for each county.

S-5 Replication of Known Hurricane Losses

The model shall estimate incurred losses in an unbiased manner on a sufficient body of past hurricane events from more than one company, including the most current data available to the modeling organization. This standard applies separately to personal residential and, to the extent data are available, to commercial residential. Personal residential experience may be used to replicate structure-only and contents-only losses. The replications shall be produced on an objective body of loss data by county or an appropriate level of geographic detail and shall include loss data from both 2004 and 2005.

Table 14 compares the modeled and actual total losses by hurricane and company for personal residential coverage. Moreover, Figure 42 indicates reasonable agreement between the observed and modeled losses. This was also supported by the various statistical tests described below.

Disclosures

1. ***Describe the nature and results of the analyses performed to validate the loss projections generated by the model for personal and commercial residential separately. Include analyses for the 2004 and 2005 hurricane seasons.***

For model validation purposes, the actual and modeled losses for some selected companies and hurricanes are provided in Table 14.

Table 14. Total Actual vs. Total Modeled Losses.

Company Name	Event	Total Exposure	Total Actual Loss	Total Modeled Loss
A	Charley	14572357458.00	274702333.00	198737119.97
A	Frances	9613407332.00	224656954.00	141606296.14
B	Charley	7155996653.00	110471361.00	124580043.64
B	Frances	1847430290.00	20201407.00	61591326.44
C	Charley	27568302239.00	526544555.00	328479701.97
C	Dennis	8858384208.00	20384468.00	55684738.00
C	Frances	20091682728.00	392510598.00	272506470.42
C	Jeanne	39525022665.00	177552030.00	401863199.62
C	Katrina	6232468582.00	19712702.00	79909488.96
C	Wilma	39461443904.00	340628254.00	543021524.08
D	Charley	1377700566.00	63889029.00	22384465.48
D	Frances	4309535304.00	122776727.00	74034631.42
E	Charley	35580184.00	952353.00	664538.39
E	Frances	316894463.00	10007410.00	4201793.13
E	Charley	2498971217.00	113313510.00	47276226.01
E	Frances	3639401631.00	78377163.00	61057776.86
E	Jeanne	4307858204.00	40245030.00	71539354.47
F	Charley	1386793895.00	32316645.00	20282578.33
G	Charley	587526292.00	3884930.00	6641333.28
G	Frances	189912832.00	2918642.00	3730479.70

G	Katrina	135143330.00	464971.00	858448.83
G	Wilma	767025160.00	6120435.00	9217331.98
H	Charley	844602098.00	78535467.00	51557230.62
H	Dennis	28266337.00	928111.00	2142032.00
H	Frances	665429117.00	59229372.00	23812263.98
H	Jeanne	1854530377.00	74983526.00	54296228.26
H	Katrina	6903619.00	330018.00	234998.48
H	Wilma	727865863.00	47056668.00	18797871.31
I	Charley	2506896464.00	62086256.00	50785421.04
I	Frances	74702419.00	43799401.00	7153089.93
J	Jeanne	6169965775.00	84545829.00	91067794.18
K	Charley	932092266.00	79751698.00	57010970.62
K	Jeanne	2558106618.00	81552694.00	96710294.33
L	Charley	41558803.00	4511656.00	2573848.34
L	Charley	166263166.00	8645559.00	3234270.76
L	Frances	34908100.00	4009884.00	1431110.10
L	Frances	368182344.00	11489176.00	5759558.41
L	Jeanne	78735391.00	3590284.00	3305877.80
L	Jeanne	347104726.00	4812837.00	6090394.27
M	Charley	1517072812.00	15135021.00	22447000.25
M	Frances	804861107.00	9399468.00	16536003.10
M	Jeanne	2272770727.00	9048905.00	27699686.43
N	Charley	9974317521.00	250201871.00	155825974.93
N	Frances	8155340655.00	185676998.00	158122082.28
N	Jeanne	15900477962.00	127752952.00	208018427.62
N	Katrina	482901644.00	1498112.00	4203642.78
N	Wilma	13042930295.00	156638501.00	170034281.48
O	Charley	475100767.00	2015902.00	3091625.89
O	Frances	1086978976.00	2659551.00	4903325.29
O	Jeanne	905676619.00	29144703.00	36604823.82
O	Jeanne	1436506385.00	2059383.00	6232773.82
P	Jeanne	3434049257.00	31066792.00	52236441.42
Q	Andrew	30391564010.00	2984373067.00	2172367393.04
Q	Charley	427213972.00	23395988.00	16336023.08
Q	Charley	51283638860.00	1037108745.00	602805672.89
Q	Dennis	8560926395.00	30098559.00	55014031.00
Q	Erin	3193215496.00	50519119.00	59625732.57
Q	Frances	482335774.00	18467176.00	7903118.39
Q	Frances	36447006477.00	614006549.00	420792734.68
Q	Katrina	19486034141.00	54163254.00	102899060.88
Q	Wilma	80021657140.00	1185407656.00	732908955.18
R	Jeanne	1178562197.00	3125588.00	14861814.76
S	Charley	9721434560.00	111013524.00	216617817.85
S	Frances	12631336130.00	94272660.00	385265228.80
T	Charley	2685932544.00	54207520.00	41706065.94
T	Frances	3554743715.00	121893725.00	52474757.28

Figure 42 provides a comparison of total actual losses vs. total modeled losses for different hurricanes. The comparison indicates a reasonable agreement between the actual and modeled losses. The correlation between actual and modeled losses is found to be 0.970, which shows a strong positive linear relationship between actual and modeled losses. We tested whether the difference in paired mean values equals zero using the paired t test (t = 1.4387, df = 65, p-value = 0.155) and Wilcoxon signed rank test (Z = 0.9454, p-value = 0.3444). Based on these tests, we failed to reject the null hypothesis of equality of paired means and concluded that there is insufficient evidence to suggest a difference between actual and modeled losses. We also observe from Table 14 that about 52% of the actual losses are more than the corresponding modeled losses, and 48% of the modeled losses are more than the corresponding actual losses. This shows that our modeling process is not biased. Following Lin (1989), the bias correction factor (measure of accuracy) is obtained as 0.946, and the sample concordance correlation coefficient is found to be 0.918, which again shows a strong agreement between actual and modeled losses.

Due to the lack of a sufficient body of claims data for commercial losses, extensive statistical tests were not conducted to validate the model losses. ~~However, a tabular comparison of the modeled vs. actual commercial insured loss costs in Table 15 shows a reasonable agreement (Wilcoxon Signed Rank Test Statistic = 26, p-value = 0. 0.3125) between the two. The correlation coefficient between Actual and Modeled losses is 0.952.~~ A tabular comparison of the modeled vs. actual commercial insured loss costs is presented in Table 15 for illustration purposes. A Wilcoxon Signed Rank Test Statistic value of 26 gives a p-value of 0.3125 and the correlation coefficient between Actual and Modeled losses is 0.951.

Table 15. Comparison of Total vs. Actual Losses - Commercial Residential

Company	Event	Total Exposure	Total Actual Loss	Total Modeled Loss
A	Charley	$ 2,265,520,808.00	$ 64,378,393.00	$ 22,652,931.01
A	Jeanne	$ 4,866,082,786.00	$ 34,826,257.00	$ 52,591,725.33
A	Katrina	$ 6,489,785,877.00	$ 11,846,697.00	$ 31,287,853.21
~~A~~	~~Wilma~~	~~$ 20,490,736,703.00~~	~~$ 318,671,056.00~~	~~$ 179,818,326.20~~
C	Frances	$ 863,784,392.00	$ 41,726,240.00	$ 11,337,074.54
C	Jeanne	$ 1,021,543,325.00	$ 8,373,620.00	$ 12,173,750.27
C	Katrina	$ 224,012,300.00	$ 2,175,895.00	$ 8,040,237.83
C	Wilma	$ 2,423,163,266.00	$ 60,926,510.00	$ 18,218,195.64

Company Name	Event	Total Exposure	Total Actual Loss	Total Modeled Loss
D	Charley	$ 2,344,572,547.00	$ 64,378,393.00	$ 22,652,931.01
D	Jeanne	$ 4,866,082,786.00	$ 34,826,257.00	$ 52,591,725.32
D	Katrina	$ 6,489,785,877.00	$ 11,846,697.00	$ 31,287,853.21
D	Wilma	$ 20,489,475,103.00	$ 318,671,056.00	$ 179,818,326.20
Q	Frances	$ 863,784,392.00	$ 42,238,244.00	$ 11,332,725.15
Q	Jeanne	$ 1,021,385,625.00	$ 8,446,718.00	$ 12,173,750.27
Q	Katrina	$ 224,012,300.00	$ 2,178,110.00	$ 8,040,237.83
Q	Wilma	$ 2,423,163,266.00	$ 62,492,371.00	$ 18,218,195.64

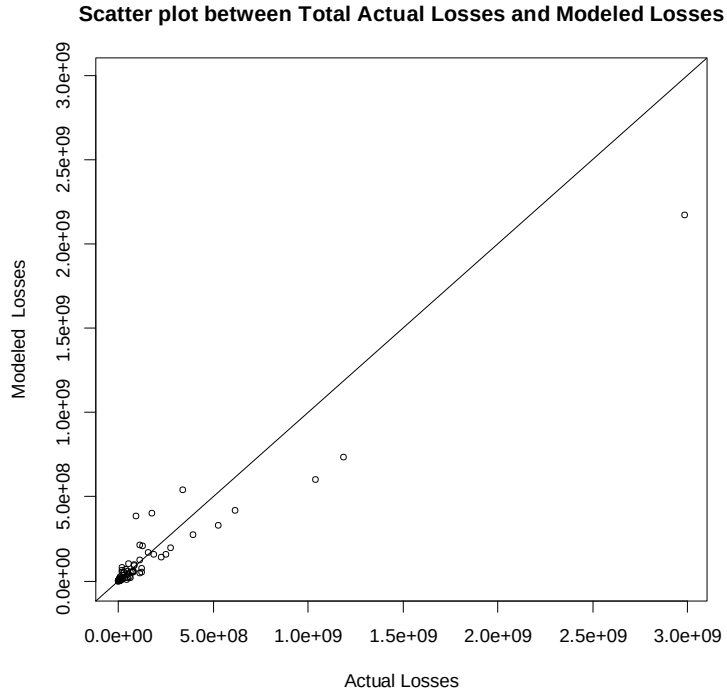
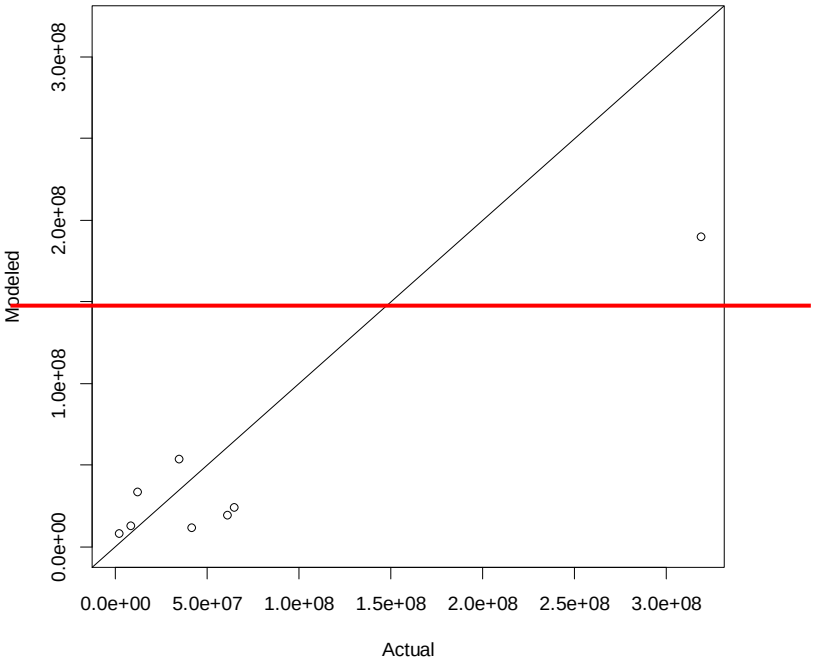


Figure 42. Scatter plot between total actual vs. total modeled losses (Personal Residential).



~~**Figure 47. Scatter plot between actual vs modeled losses (Commercial Residential)**~~

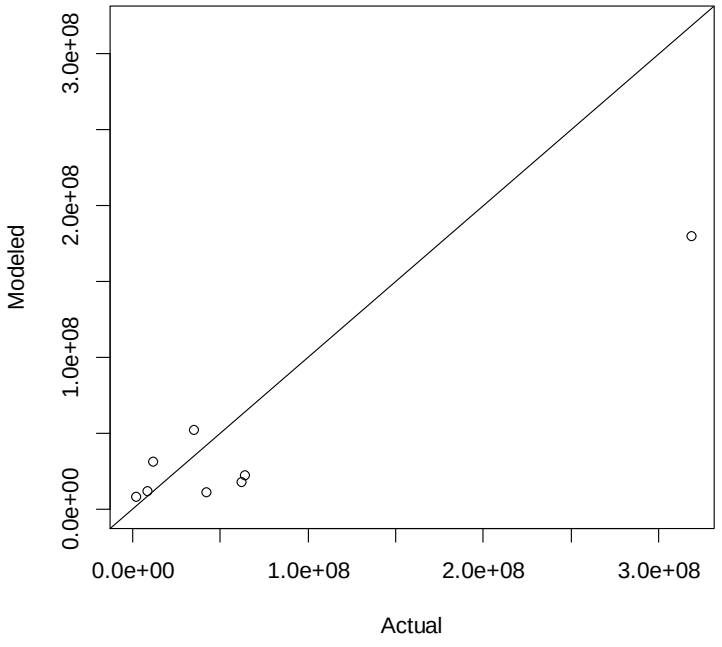


Figure 43. Scatter plot between total losses vs. total modeled losses (Commercial Residential).

2. ***Provide a completed Form S-4, Validation Comparisons. Provide a link to the location of the form here.***

Please see the completed Form S-4 at the end of this section.
Reference:

Lin, L. I. (1989). A concordance correlation coefficient to evaluate reproducibility. *Biometrics*, 45(1), 255-268.

S-6 Comparison of Projected Hurricane Loss Costs

The difference, due to uncertainty, between historical and modeled annual average statewide loss costs shall be reasonable, given the body of data, by established statistical expectations and norms.

The difference, due to uncertainty, between historical and modeled annual average statewide loss costs is reasonable as shown in the following description.

Disclosures

1. ***Describe the nature and results of the tests performed to validate the expected loss projections generated. If a set of simulated hurricanes or simulation trials was used to determine these loss projections, specify the convergence tests that were used and the results. Specify the number of hurricanes or trials that were used.***

Loss costs are generated using a simulated number of hurricanes. The number of years used in the simulations was calculated as described in Standard S-4, and was found to be 57,000. The standard errors are within 2.5% of the means for all counties. From Form S-5 we found that the 95% confidence interval on the difference between the mean of the losses from the historical and modeled contains 0, indicating that there is no statistically significant difference. In addition, as shown in Standard S-5, modeled loss costs have also been validated against insurance company data and are in reasonable agreement with the same.

2. ***Identify and justify differences, if any, in how the model produces loss costs for specific historical events versus loss costs for events in the stochastic hurricane set.***

The historical and stochastic storm loss costs are treated the same.

3. ***Provide a completed Form S-5, Average Annual Zero Deductible Statewide Loss Costs – Historical versus Modeled. Provide a link to the location of the form here.***

Please see the completed Form S-5 at the end of this section.

Form S-1: Probability and Frequency of Florida Landfalling Hurricanes per Year

Complete the table below showing the probability and modeled frequency of landfalling Florida hurricanes per year. Modeled probability shall be rounded to four decimal places. The historical probabilities and frequencies below have been derived from the Base Hurricane Storm Set for the 113 year period 1900-2012 (as given in Form A-2, Base Hurricane Storm Set Statewide Losses). Exclusion of hurricanes that caused zero modeled Florida damage or additional Florida landfalls included in the modeling organization Base Hurricane Storm Set as identified in their response to Standard M-1 (Base Hurricane Storm Set) should be used to adjust the historical probabilities and frequencies provided here.

If the data are partitioned or modified, provide the historical probabilities and frequencies for the applicable partition (and its complement) or modification as well as the modeled probabilities and frequencies in additional copies of Form S-1 (Probability and Frequency of Florida Landfalling Hurricanes per Year). ***Please note that the results here are based on the base set for 1900-2013.***

See Appendix N.

Form S-2A: Examples of Loss Exceedance Estimates (2007 FHCF Exposure Data)

Provide projections of the aggregate personal and commercial insured losses for various probability levels using the notional risk data set specified in Form A-1 (Zero Deductible Personal Residential Loss Costs by ZIP Code) and using the 2007 Florida Hurricane Catastrophe Fund aggregate personal and commercial residential exposure data provided in the file named "hlpm2007c.exe." Provide the total average annual loss for the loss exceedance distribution. If the modeling methodology does not allow the model to produce a viable answer, please state so and why.

See Appendix O.

Form S-2B: Examples of Loss Exceedance Estimates (2012 FHCF Exposure Data)

Provide projections of the aggregate personal and commercial insured losses for various probability levels using the notional risk data set specified in Form A-1 (Zero Deductible Personal Residential Loss Costs by ZIP Code) and using the 2012 Florida Hurricane Catastrophe Fund aggregate personal and commercial residential exposure data provided in the file named "hlpm2012c.exe." Provide the total average annual loss for the loss exceedance distribution. If the modeling methodology does not allow the model to produce a viable answer, please state so and why.

See Appendix P.

Form S-3: Distributions of Stochastic Hurricane Parameters

Provide the probability distribution functional form used for each stochastic hurricane parameter in the model. Provide a summary of the rationale for each functional form selected for each general classification.

See Appendix Q.

Form S-4: Validation Comparisons

A. *Provide five validation comparisons of actual personal residential exposures and loss to modeled exposures and loss. These comparisons must be provided by line of insurance, construction type, policy coverage, county or other level of similar detail in addition to total losses. Include loss as a percent of total exposure. Total exposure represents the total amount of insured values (all coverages combined) in the area affected by the hurricane. This would include exposures for policies that did not have a loss. If this is not available, use exposures for only those policies that had a loss. Specify which was used. Also, specify the name of the hurricane event compared.*

B. *Provide a validation comparison of actual commercial residential exposures and loss to modeled exposures and loss. Use and provide a definition of the model's relevant commercial residential classifications.*

C. *Provide scatter plot(s) of modeled vs. historical losses for each of the required validation comparisons. (Plot the historical losses on the x-axis and the modeled losses on the y-axis.)*

Rather than using directly a specific published hurricane wind field, the winds underlying the modeled loss cost calculations must be produced by the model being evaluated and should be the same hurricane parameters as used in completing Form A-2.

See Appendix R.

Form S-5: Average Annual Zero Deductible Statewide Loss Costs – Historical versus Modeled

Part A

A. Provide the average annual zero deductible statewide personal residential loss costs produced using the list of hurricanes in the Base Hurricane Storm Set as defined in Standard M-1 (Base Hurricane Strom Set) based on the 2007 Florida Hurricane Catastrophe Fund's aggregate personal residential exposure data found in the file named "hlpm2007c.exe."

See Appendix S.

B. Provide a comparison with the statewide personal residential loss costs produced by the model on an average industry basis.

The loss cost produced by the model on an average industry basis is 5.2 billion dollars and the corresponding historical average loss is 6.0 billion dollars.

C. Provide the 95% confidence interval on the differences between the mean of the historical and modeled personal residential loss.

The 95% confidence interval on the difference between the mean of the historical and the mean of the modeled losses is between ~~-1.41~~-1.38 and ~~2.98~~2.96 billion dollars. Since the interval contains 0, we are 95% confident that there is no significant difference between the historical and the modeled losses.

D. If the data are partitioned or modified, provide the average annual zero deductible statewide personal residential loss costs for the applicable partition (and its complement) or modification as well as the modeled average annual zero deductible statewide personal residential loss costs in additional copies of Form S-5.

Not applicable.

Part B

A. Provide the average annual zero deductible statewide personal and commercial residential loss costs produced using the list of hurricanes in the Base Hurricane Storm Set as defined in Standard M-1 (Base Hurricane Storm Set) based on the 2012 Florida Hurricane Catastrophe Fund's aggregate personal and commercial residential exposure data found in the file named "hlpm2012c.exe."

See Appendix S.

B. Provide a comparison with the statewide personal and commercial residential loss costs produced by the model on an average industry basis.

The loss cost produced by the model on an average industry basis is 4.9 billion dollars and the corresponding historical average loss is 5.7 billion dollars.

C. *Provide the 95% confidence interval on the differences between the mean of the historical and modeled personal and commercial residential loss.*

The 95% confidence interval on the difference between the mean of the historical and the mean of the modeled losses is between ~~-1.30~~-1.28 and ~~2.83~~2.80 billion dollars. Since the interval contains 0, we are 95% confident that there is no significant difference between the historical and the modeled losses.

D. *If the data are partitioned or modified, provide the average annual zero deductible statewide personal and commercial residential loss costs for the applicable partition (and its complement) or modification, as well as the modeled average annual zero deductible statewide personal and commercial residential loss costs in additional copies of Form S-5 (Average Annual Zero Deductible Statewide Loss Costs – Historical versus Modeled).*

Not applicable.

Form S-6: Hypothetical Events for Sensitivity and Uncertainty Analysis

We have provided the output in ASCII files based on running a series of hurricanes as provided in the Excel file "FormS5Input09.xls." The output files consist of wind speeds (in miles per hour for one minute sustained 10 meter winds) at hourly intervals over a 21×40 grid for the 500 combinations of initial conditions specified in the Excel file for the following model inputs:

- *CP* = central pressure (in millibars)
- *Rmax* = radius of maximum winds (in statute miles)
- *VT* = translational velocity (forward speed in miles per hour)
- Holland *B* = pressure profile parameter for other input used by the modeler ($0 \le p \le 1$)
- *FFP* = far field pressure (in millibars)

The value of *CP*, *Rmax*, *VT*, *FFP* and Quantile are used as direct inputs. Quantiles from 0 to 1 have been provided in the Excel input file. For the FPHLM (V4.1) model, we used the first quantile input for the Holland *B* parameter.

On a CD, we have provided an ASCII file and a PDF file named FPHLM09Expected Loss Costs. This file gives aggregate and expected loss costs for each input vector for each category of hurricane and contains 3x100=300 rows.

We have also provided, on a CD, the results in an ASCII file and a PDF file named FPHLM09Loss Cost Contour, which contains 3 x 682 = 2,046 rows. This file gives the mean loss cost at each of the 682 land based vertices over all 100 input vectors for each hurricane category.

Distribution of Loss Costs

Figure 44 provides the comparison of CDFs of the Expected Loss Costs for all Hurricane Categories.

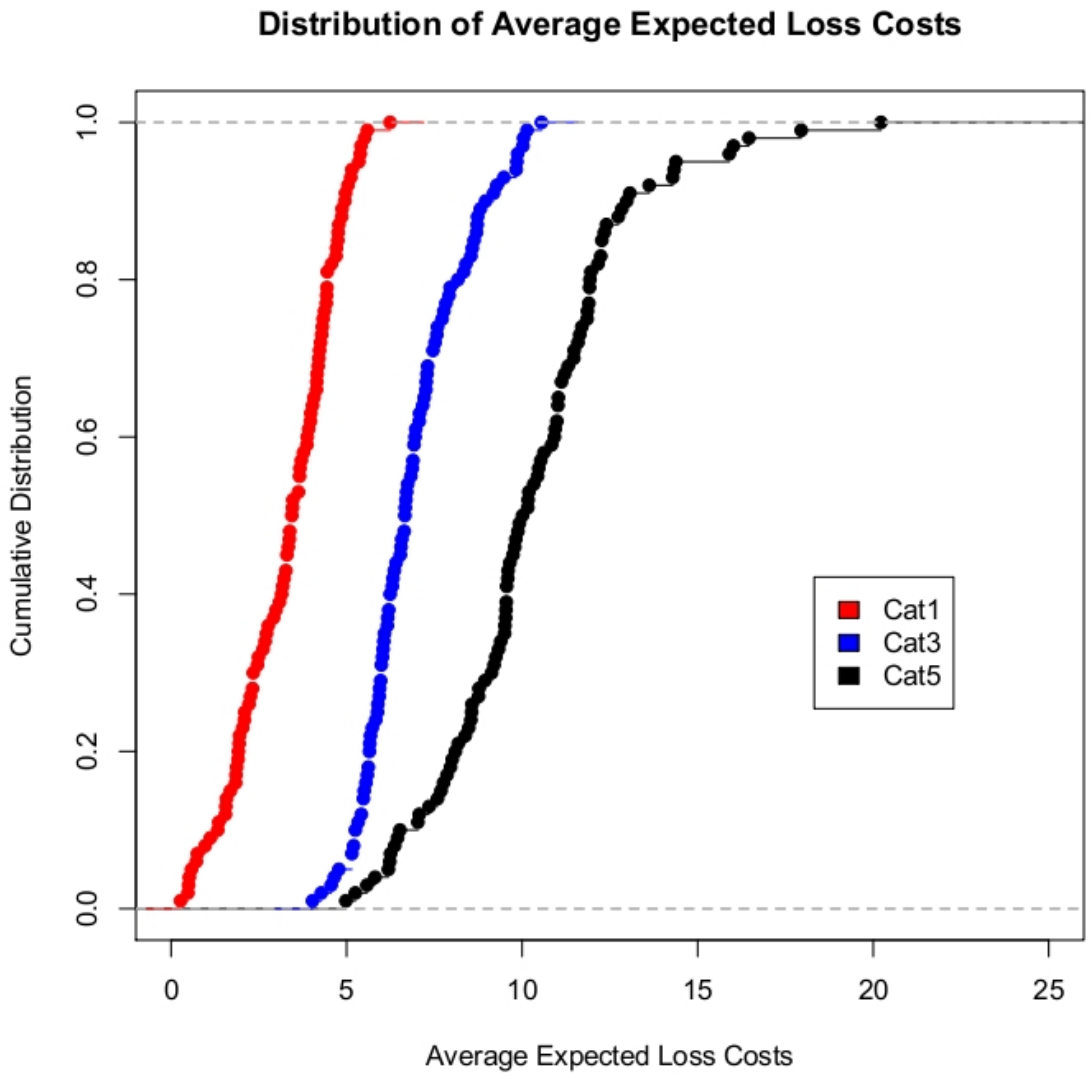


Figure 44. Comparison of CDFs of Loss Costs for all Hurricane Categories.

Figure 45 – Figure 47 show contours of the mean loss cost for Category 1, 3 and 5 hurricane respectively for each land based grid point. The mean percentage loss costs are found to be about between 1.14 %-8.3% for Category 1, between 3.64%-24.6% for Category 3 and between 2.57%-41.84% for Category 5 hurricanes. The largest losses occur shortly after landfall to the right of the hurricane path.

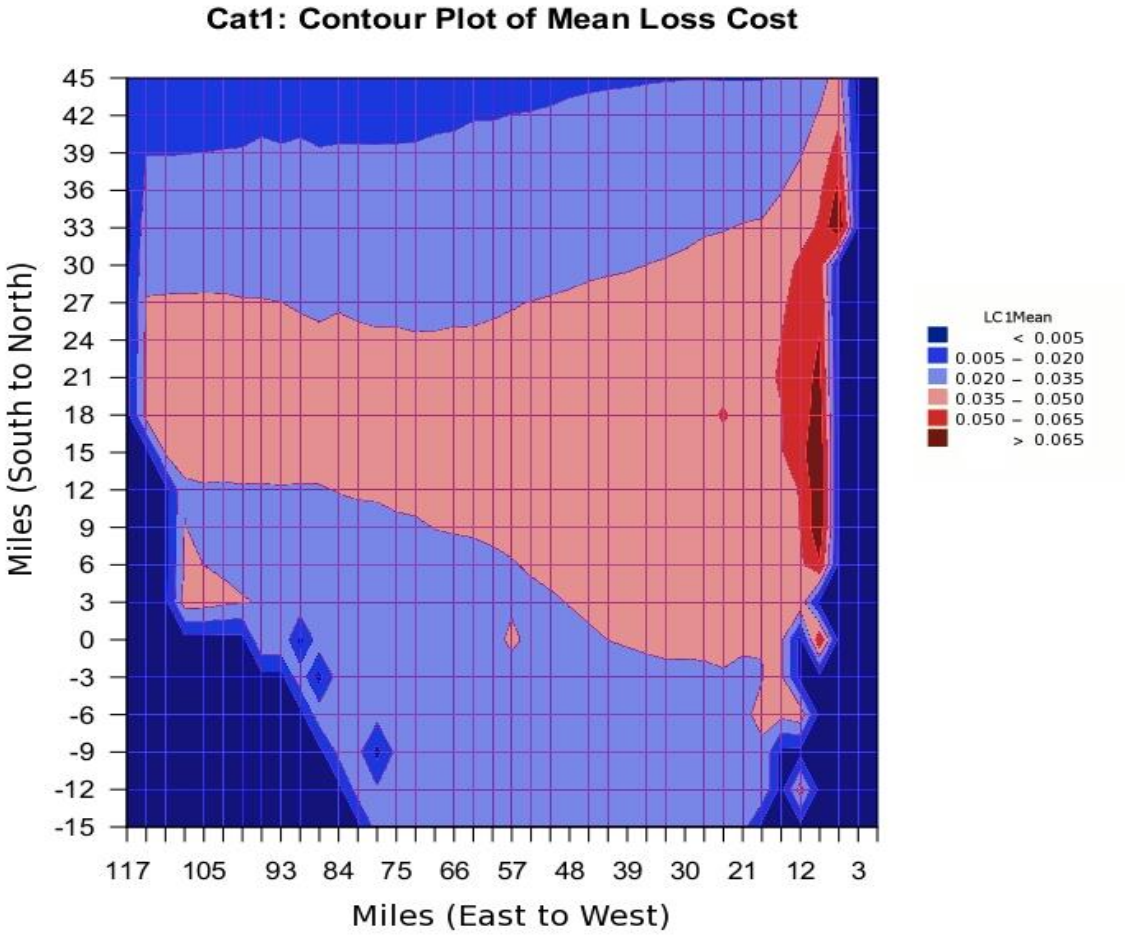


Figure 45. Contour Plot of Loss Cost for a Category 1 Hurricane.

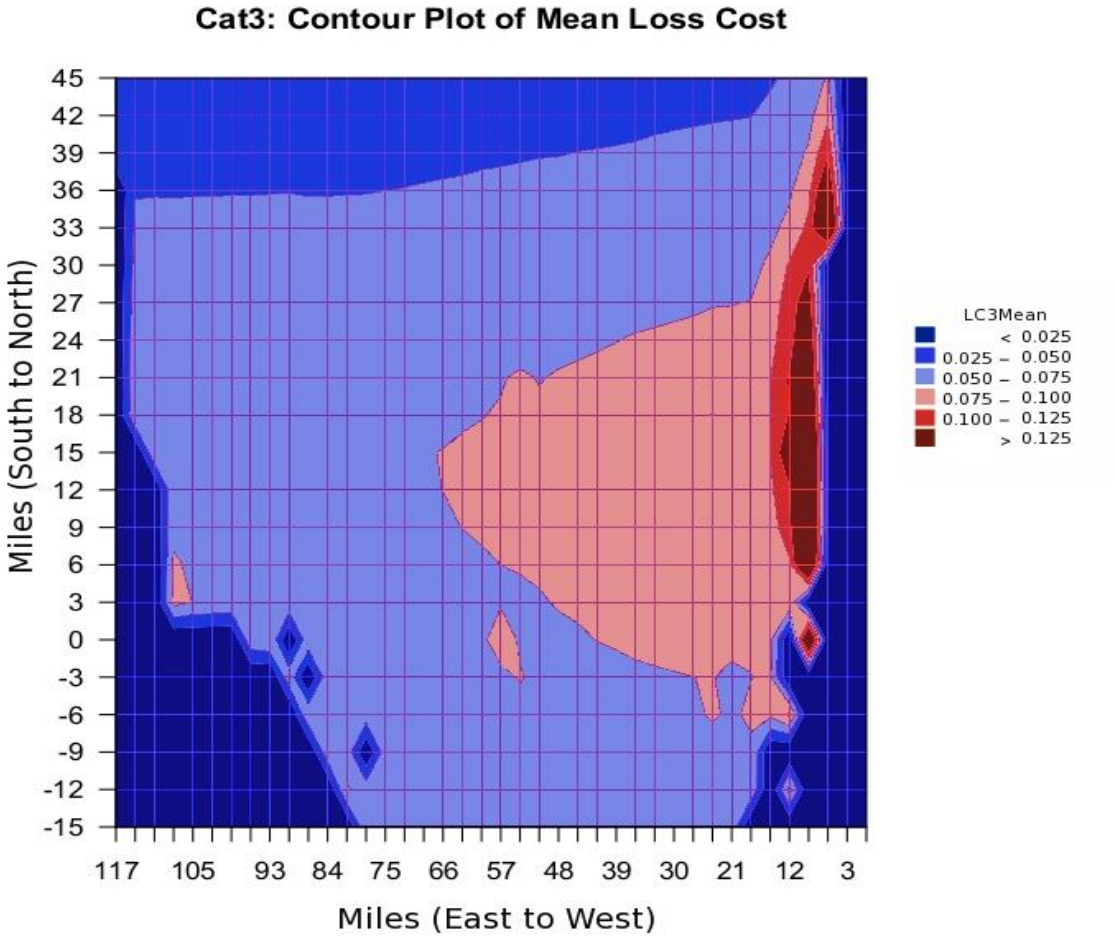


Figure 46. Contour Plot of Loss Cost for a Category 3 Hurricane.

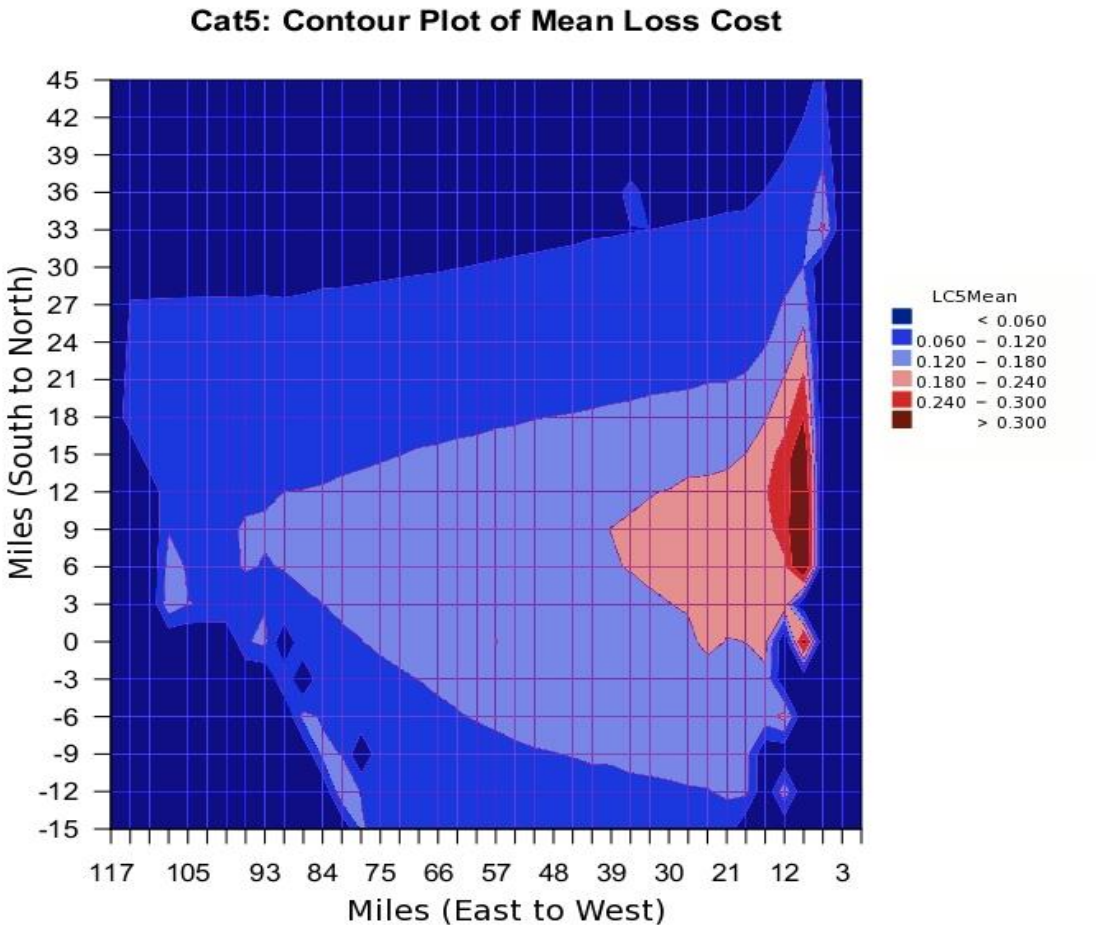


Figure 47. Contour Plot of Loss Cost for a Category 5 Hurricane.

Sensitivity and Uncertainty Analysis for Expected Loss Costs

Sensitivity analysis for the expected loss costs was conducted through the use of the standardized regression coefficients of the expected loss cost as a function of the input variables for Category, 1, 3 and 5 hurricanes. We used the methods described by Iman et al. (2000a, 2000b). The values of standardized regression coefficients are summarized in the table below.

Category	*CP*	*Rmax*	*VT*	*Holland B*	*FFP*
1	-0.4118	0.1039	0.1648	0.6477	0.5905
3	-0.2599	0.4033	0.1137	0.6552	0.4236
5	-0.1349	0.6939	-0.0022	0.5862	0.1801

Figure 48 gives the graph of the standardized regression coefficients for all input variables for Category 1, 3 and 5 hurricanes. From the graph, we observed that the sensitivity of expected loss cost depends on the category of the hurricanes. For a Category 1 hurricane, expected loss cost is

most sensitive to *Holland B* parameter followed by *FFP, CP* and *VT*. For a Category 3 hurricane, expected loss cost is most sensitive to *Holland B* followed by *FFP, Rmax and CP* and finally for a Category 5 hurricane, expected loss cost is most sensitive to *Rmax,* followed by *Holland B, CP* and *FFP*. The expected loss cost is least sensitive to *Rmax* for Category 1 while the expected loss cost is least sensitive to *VT* for Categories 3 and 5.

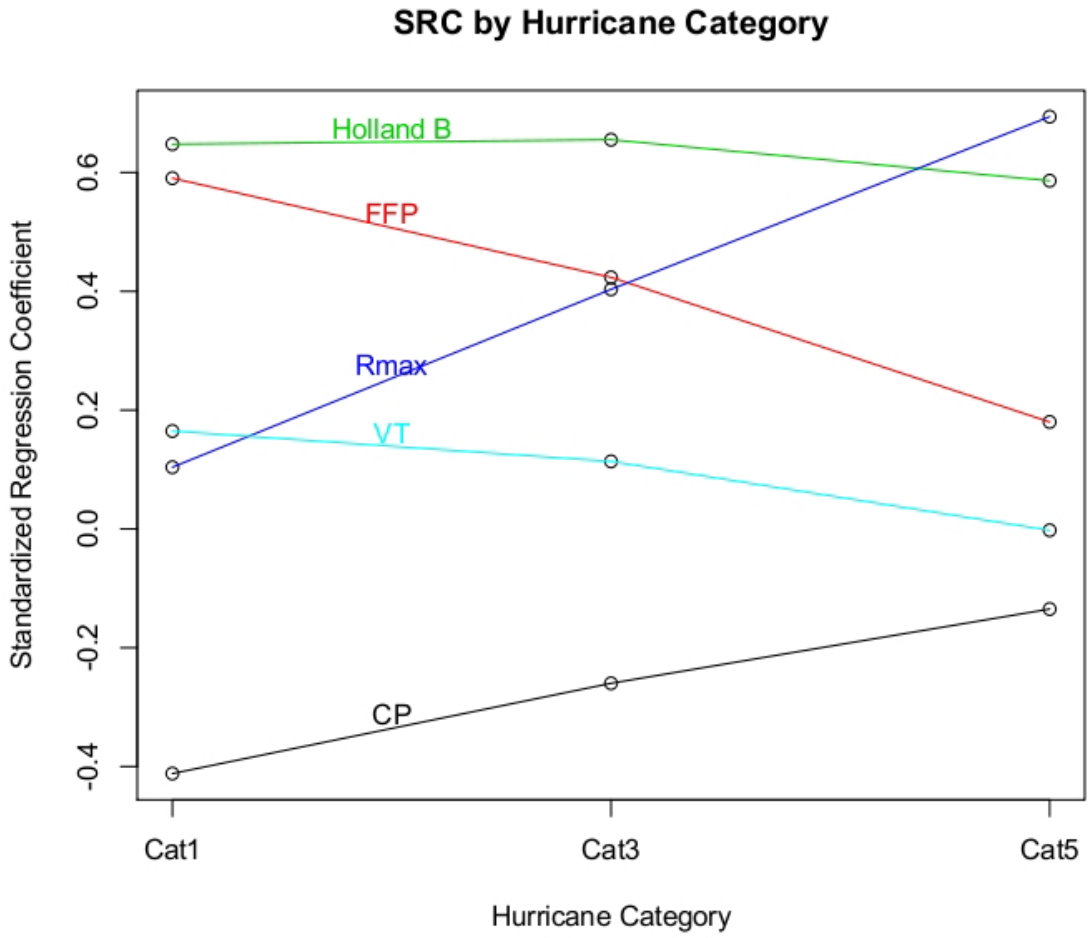


Figure 48. SRCs for expected loss cost for all input variables for all hurricane categories.

Uncertainty analysis for the expected loss costs was conducted through the use of the expected percentage reduction (EPR) in the variance of the expected loss cost as a function of the input variables for Category, 1, 3 and 5 hurricanes. We used the methods described by Iman et al. (2000a, 2000b). The values of EPR's are summarized in the table below.

Category	*CP*	*Rmax*	*VT*	*Holland B*	*FFP*
1	20.8398%	3.9463%	2.0921%	46.2717%	36.7245%
3	6.0155%	14.8201%	1.1625%	51.3594%	10.4668%
5	4.6087%	48.7428%	1.8529%	42.1176%	4.6455%

Figure 49 gives the expected percentage reductions in the variance of expected loss cost for Category 1, 3 and 5 Hurricanes for all input variables. As with the sensitivity analysis, the category of the hurricane determines which variable contributes most to the uncertainty of the expected loss cost. For a Category 1 hurricane, the major contributor to the uncertainty in loss cost is the *Holland B* parameter, followed by *FFP*, then *CP*. For a Category 3 hurricane, the major contributor to the uncertainty in loss cost is *Holland B*, followed by *Rmax*, then *FFP*. For a Category 5 hurricane, the major contributor to the uncertainty of expected loss cost is *Rmax*, followed by *Holland B*, then *FFP*, and finally *CP*. The variable *VT* has negligible effect on the uncertainty in expected loss costs.

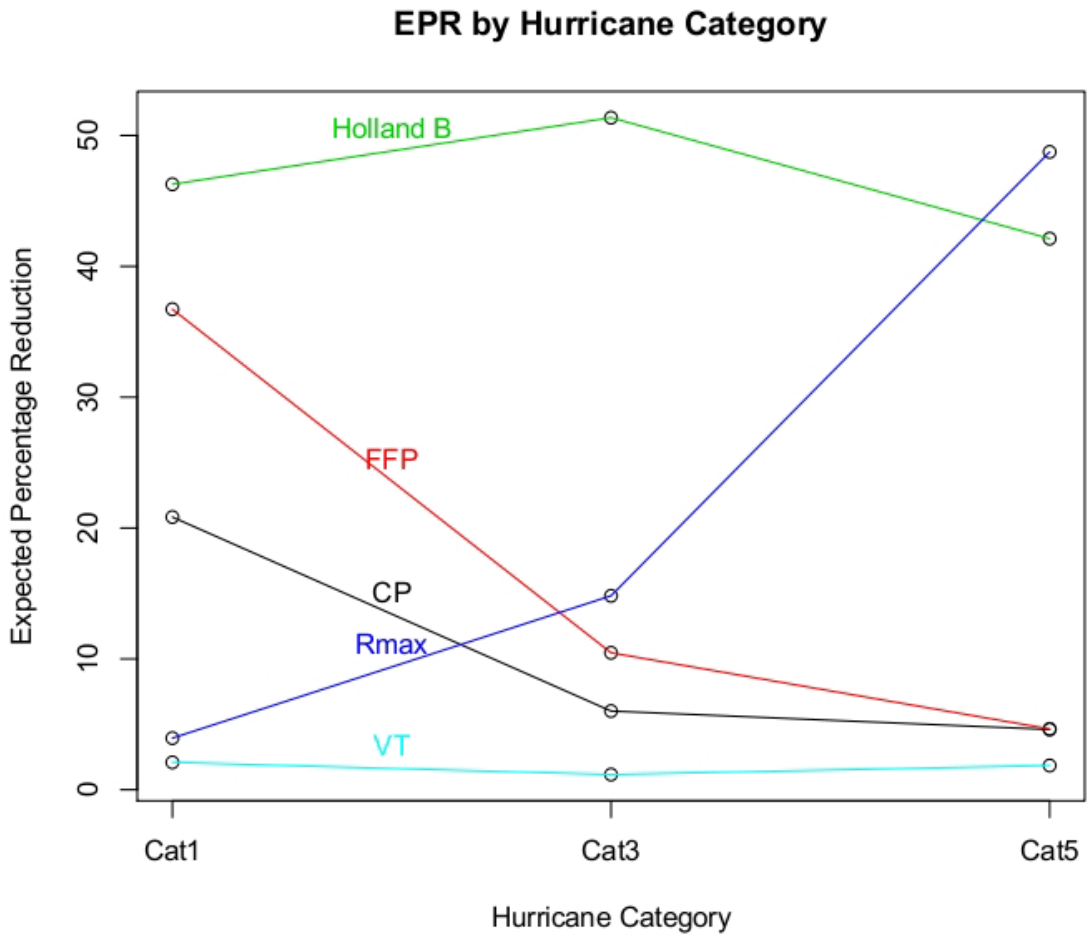


Figure 49. EPRs for Expected Loss Cost for all Input Variables for all Hurricane Categories.

VULNERABILITY STANDARDS

V-1 Derivation of Vulnerability Functions

A. ***Development of the building vulnerability functions shall be based on at least one of the following: (1) historical data, (2) tests, (3) rational structural analysis, and (4) site inspections. Any development of the building vulnerability functions based on rational structural analysis, site inspections, and tests shall be supported by historical data.***

The development of the vulnerabilities is based on a component approach that combines engineering modeling, simulations with engineering judgment, and observed (historical) data. The determination of external damage to buildings is based on structural calculations, tests, and Monte Carlo simulations. The wind loads and strength of the building components in the simulations are based on laboratory and in-situ tests, manufacturer's data, expert opinion based on post-hurricane site inspections of actual damage, and codes and standards. The internal and content damage are extrapolated from the external damage on the basis of expert opinion and site inspections of areas impacted by recent hurricanes and are confirmed using historical claims data.

B. ***The method of derivation of the building vulnerability functions and their associated uncertainties shall be theoretically sound and consistent with fundamental engineering principles.***

The method used in the derivation is based on extrapolating the results of Monte Carlo simulations of physical exterior damage through simple equations based on engineering judgment, expert opinion, and claims data. Uncertainties at each stage are accounted for by distributing the damage according to reasonable probability distributions and are validated with claims data.

The Monte Carlo component models take into account many variations in structural characteristics, and the result clearly filters through the cost estimation model. There are also different and clearly defined costing considerations applied to each structural type. These adjustments come directly from resources developed exclusively for defining repair costs to structures and therefore are theoretically sound.

C. ***Residential building stock classification shall be representative of Florida construction for personal and commercial residential properties.***

A detailed exposure study was carried out to define the most prevalent construction types and characteristics in the Florida residential building stock for different regions. The corresponding engineering models were built for each of the identified common structural types. In the case of the residential model and the low-rise commercial residential model, the models include differing wall types (wood and masonry) of varying strengths (e.g., reinforced or not, various roof to wall connection types), differing roof shapes (hip and gable end), various strengths of roof-to-wall connections (toe nails, clips, straps), varying window types and sizes, opening protection systems, varying garage door pressure capacities, and one and two story houses and one-to-three story commercial residential buildings.

Models of varying combinations of the above characteristics (e.g., wood frame, gable end, no window shutters) were created for four different regions in Florida. In all cases, the probabilistic capacities of the various components were determined by a variety of sources, including testing, test results in the literature, in-field data collection (post-hurricane damage evaluations), manufacturer's specifications and manufacturer's test data, and expert opinion.

In the case of the mid-/high-rise commercial residential model (buildings with more than three stories), the models include different apartment units corresponding to different building layouts (interior or exterior entry door), different locations within the floor plan (corner or middle units), different heights (subject to different probabilities of missile impact and wind speed), and different openings (windows, doors, sliders) with different protection options (none or impact resistant).

D. *Building height/number of stories, primary construction material, year of construction, location, building code, and other construction characteristics, as applicable, shall be used in the derivation and application of building vulnerability functions.*

The structural models include options that allow the representation of building code revisions. Three models were derived for each structural type: weak construction, medium construction, and strong construction. For example, each model for wood frame and gable roof homes has weak, medium, and strong versions. The assignment of a given strength level is based on the assumed age of the home being modeled and the available information on construction practice in that region of the state in that era of construction. Florida Building Code requirements that apply to the repair of existing homes are also taken into consideration when computing the repair costs of a structure. Separate models were also developed for manufactured housing constructed based on pre- and post-1994 HUD regulations and for different wind zones.

In addition to the various models that reflect construction type, region of Florida, and era of construction, each model has numerous additional strength features that can be adjusted before simulations are conducted to represent various combinations of mitigation features. For example, a weak constructed home in central Florida with masonry walls (no reinforcing) may have been recently re-roofed with renailed roof decking and modern code-approved shingles. The simulation model is capable of reflecting this combination of weak original construction and new, strong roof sheathing and roof cover mitigation.

E. *Vulnerability functions shall be separately derived for commercial residential building structures, personal residential building structures, mobile homes, and appurtenant structures.*

The commercial and personal residential building structures, mobile homes, and appurtenant structures are independently derived.

F. *The minimum windspeed that generates damage shall be consistent with fundamental engineering principles.*

The minimum one-minute average sustained wind speed at which some damage is observed is 38 mph (3-second gust 50 mph) for appurtenant structures. Site-built and manufactured homes have a very small probability of some very minor damage at 42 mph (3-second gust 55 mph). This probability becomes more significant at 46 mph (3-second gust 60 mph) and increases with higher

wind speed. Simulations are run for 3-second gusts from 50 mph to 250 mph in 5 mph increments.

G. ***Building vulnerability functions shall include damage as attributable to windspeed and wind pressure, water infiltration, and missile impact associated with hurricanes. Building vulnerability functions shall not include explicit damage to the building due to flood, storm surge, or wave action.***

The vulnerability functions do not explicitly include damage due to flood, storm surge, or wave action. The vulnerability functions for all models (site-built residential, manufactured homes, low-rise commercial residential, and mid-/high-rise commercial residential) include damage due to wind pressure, missile impact and water infiltration.

Disclosures

1. ***Describe any modifications to the building vulnerability component in the model since the previously accepted model.***

   a. The changes in the external damage evaluation of low-rise CR model include:
      - Projectile count increase in debris impact model
      - Interior pressure sharing between attic and top floor changed
      - Interior pressure calculation in the attic space due to sheathing loss changed
      - Change in the soffit damage computation
      - Reduction in the pressure coefficient (C_p) multiplier
      - Modification of the masonry wall area failure function and its differentiation between unreinforced and reinforced masonry.

   b. The changes in the internal damage evaluation of low-rise CR model include:
      - Changes in the rain admittance factor (RAF) values and incorporation of the new surface run-off coefficient
      - Replacement of the directionality factor (f_{sim}) with a more sophisticated directionality scheme

   c. The statistics used to weigh the low-rise CR vulnerability matrices were updated.

   d. The changes in the damage evaluation of Mid/high-rise CR model include an additional volume of water penetration, which was modeled at the upper story of Mid/high-rise CR model

The rationale for these changes is given in Standard G-1

2. ***Provide a flow chart documenting the process by which the building vulnerability functions are derived and implemented.***

The flow chart in Figure 50 summarizes the procedure used in the Monte Carlo simulations to predict the external damage to the different structural types for the case of residential buildings and commercial residential buildings. The random variables include wind speed, pressure coefficients, and the resistances of the various building components (roof cover, roof sheathing, openings, walls, connections).

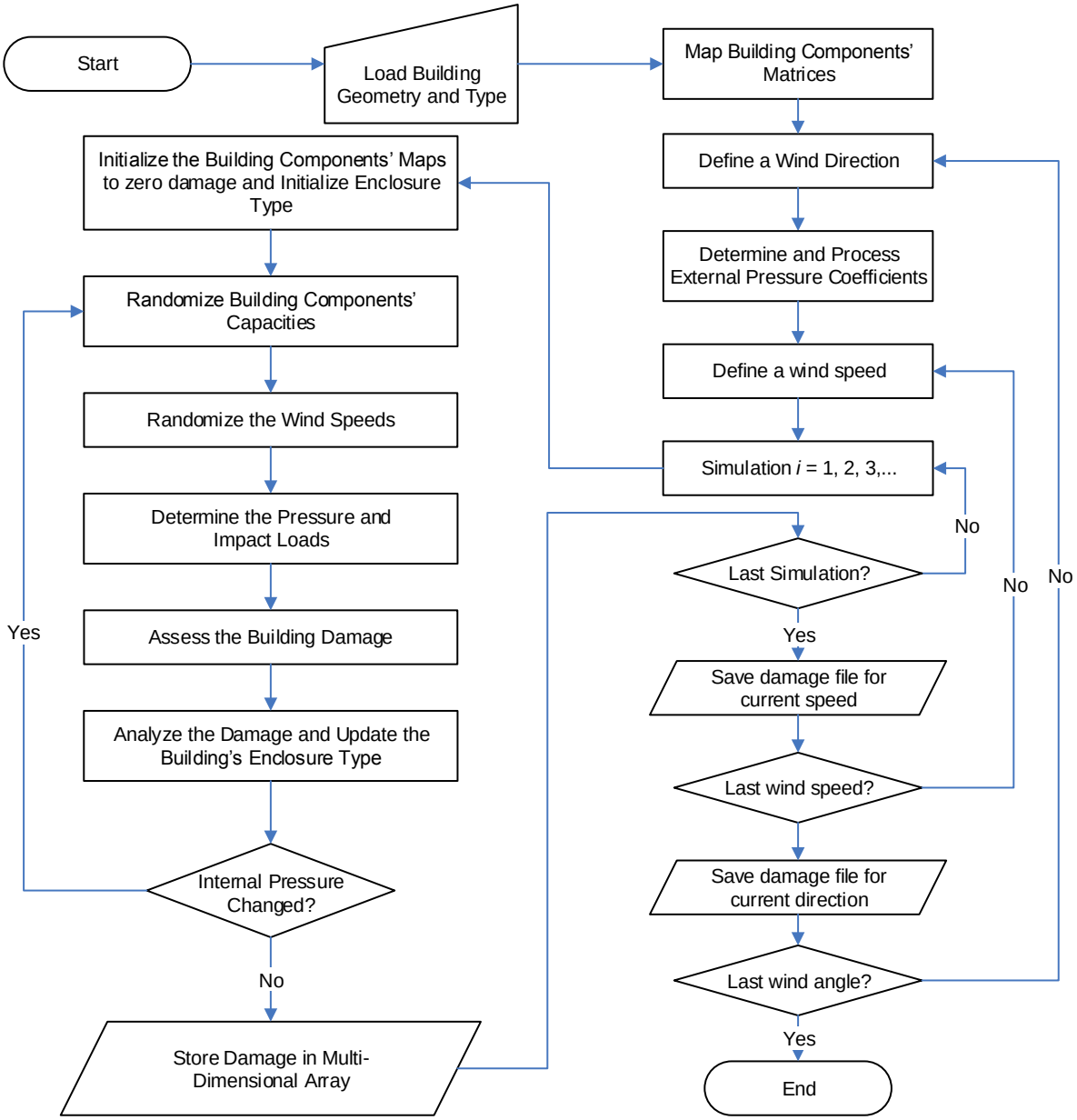


Figure 50. Monte Carlo simulation procedure to predict damage.

The flow charts in Figure 51 summarize the procedure used to convert the results of the Monte Carlo simulations of physical external damage into vulnerability matrices for the cases of the personal residential model (left) and commercial residential model (right).

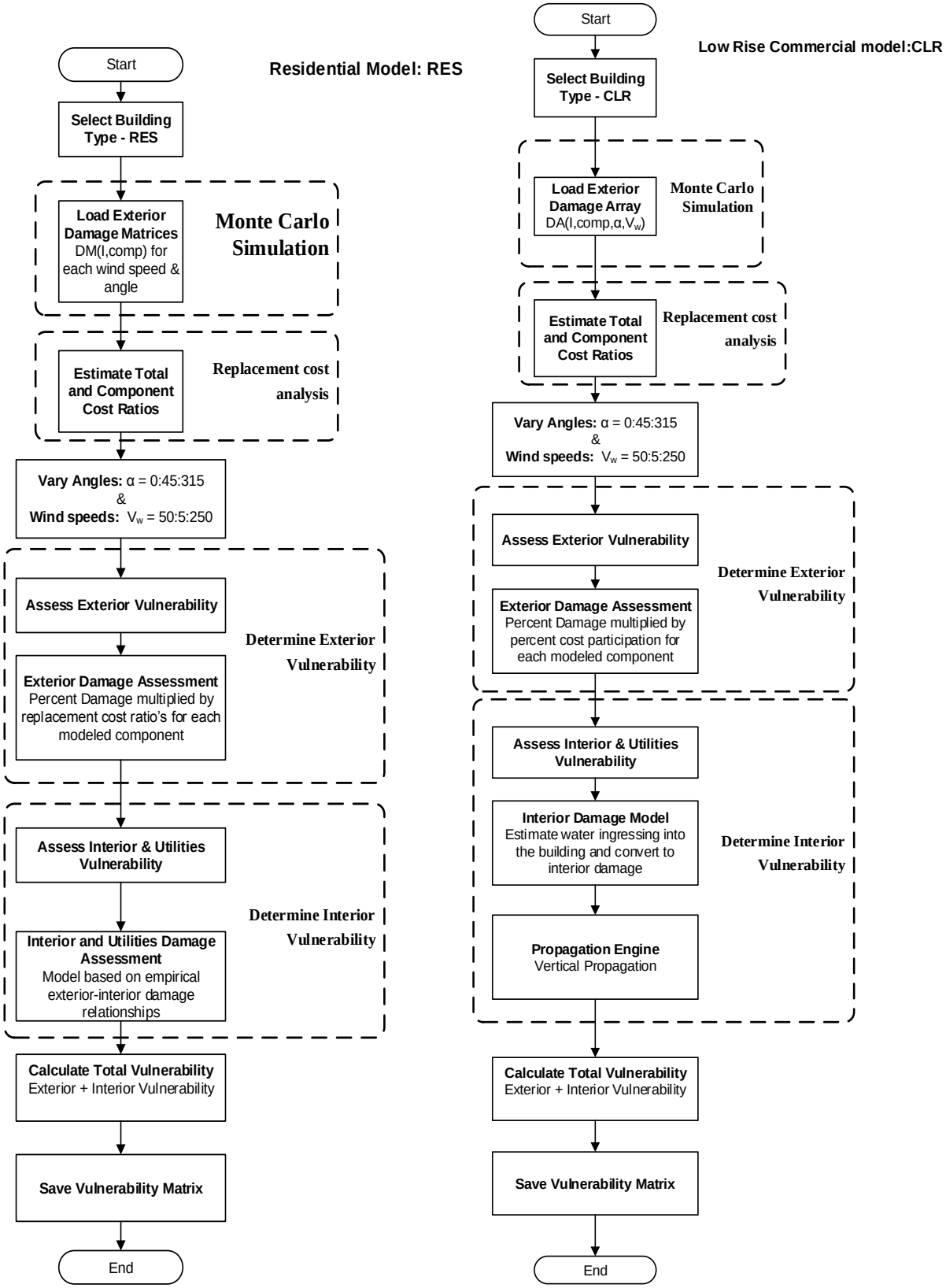


Figure 51. Procedure to create vulnerability matrix.

The flowcharts in Figure 50 and 51 are also partially applicable to the apartment facades of the mid-/high-rise commercial residential model (MHB), in which building components modeled include windows, entry doors, and balcony (sliding-glass) doors. In the case of MHB, a process similar to the one described above is followed to derive exterior vulnerability and breach curves for different openings of typical apartment units. These curves are derived for the cases of open and closed buildings, for corner and middle units, with different opening protections (with or without impact-resistant glass, with or without metal shutters). Each vulnerability curve for openings of corner or middle apartment units (window, door, or slider) gives the number or fraction of openings damaged as a function of wind speed. Each breach curve for openings of corner or middle apartment units (window, door, or slider) gives the breach area in ft^2 of opening damaged as a function of wind speed.

The flow chart in Figure 52 summarizes the procedure used to convert the apartment unit opening vulnerability and breach curves into an overall estimate of building vulnerability. This figure is already presented in Standard G-1, as Figure 18 where the values represented in the flow chart are explained in detail.

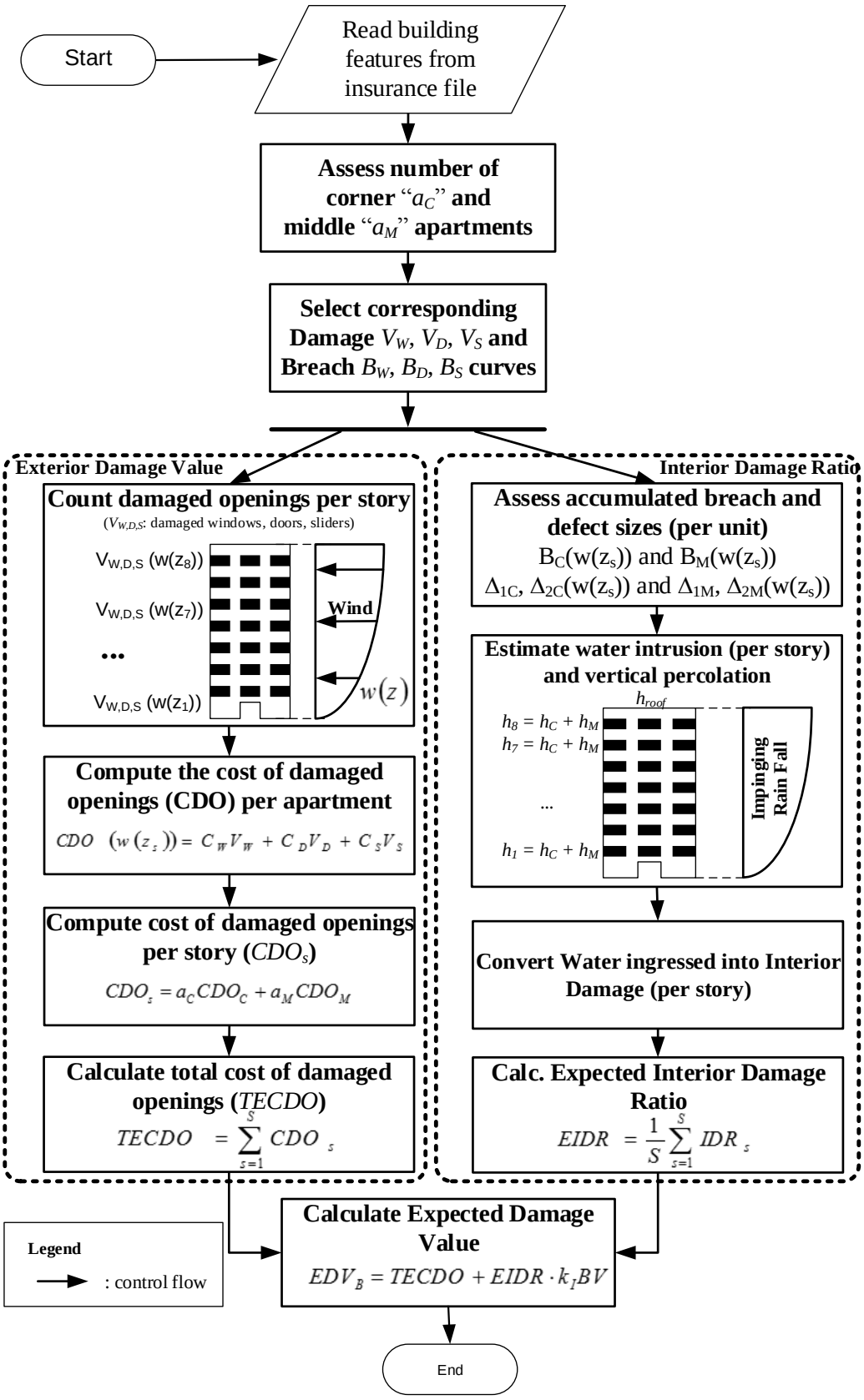


Figure 52. Exterior and interior damage assessment for MHB.

3. ***Describe the nature and extent of actual insurance claims data used to develop the model's building vulnerability functions. Describe in detail what is included, such as, number of policies, number of insurers, date of loss, and number of units of dollar exposure, separated into personal residential, commercial residential, and mobile home.***

Pre-2004 Personal Residential Claims Data

At the request of the Florida Department of Financial Services (FDFS), four insurance companies provided insurance claims data for several hurricanes that impacted Florida prior to 2004, including Andrew. The companies provided the following two types of files:

1. Sample files with 10% of the exposure selected at random, plus the claims on this 10% exposure since 1996
2. Hurricane files with premium files for all hurricane claims since 1996, plus all the corresponding claims data since 1996

Because of a confidentiality agreement, these companies will be referred to as Company A, B, C, or D. These companies represent between 75% and 85% of the insured exposure in the state and approximately 70% of the claims. Most of the data provided come from minor hurricanes and tropical storms that impacted Florida between 1994 and 2002.

Company A provided the only significant data for storms prior to 2004, in particular for Hurricane Andrew, as shown in Table 16. Wind speed estimates are also available, so validation efforts were primarily concentrated on the use of these data. Attempts were made to make use of additional data from Hurricane Opal and other storms. However, the amount of processed data available was too small to be statistically significant for validation.

Table 16. Summary of processed claims data (number of claims provided).

	Hurricane Andrew	Hurricane Georges	Hurricane Opal	Tropical Storm Irene	Tropical Storm Earl	Hurricane Erin
Company A						
Masonry	78636	266	1973	3638	59	11460
Timber	1603	1078	9166	776	89	11878
Manufactured	1775	0	256	184	16	690

Note: Only building, contents, and appurtenant structure claims were provided by Company A (ALE was not provided).

2004 Personal Residential Claims Data

Claims data for the 2004 hurricane season from a series of insurance companies were also used to validate the FPHLM. Although 21 companies submitted data for a total of almost 675,000 claims, only two main companies are detailed here. These two companies (referred to as Company 1 and

Company 2) represent 386,000 claims, mainly for site-built homes. These claims are divided between Hurricanes Charley, Frances, and Jeanne for central Florida, and Hurricane Ivan for the Panhandle. The validation consists of a series of comparisons between the actual claims data and the FPHLM results. The claims files were provided by the insurance companies. Table 17, Table 18, and Table 19 show the number of policies provided by the two companies for the four different hurricanes in 2004. As expected, there are more masonry claims in central Florida and more timber claims in the Panhandle. The claims data for Ivan was not used in the validation process because it was contaminated by storm surge damage.

One additional claims data point became available for the actuarial team prior to this submission, but was not used by the vulnerability team for validation.

Table 17. Company 1: Claim number for each year-build category

Company	Hurricane	Construction	Year Built	Number of Claims
Company 1	Charley	Masonry	yb<1970	5026
Company 1	Charley	Masonry	1970<=yb<1984	8216
Company 1	Charley	Masonry	1984<=yb<1994	11850
Company 1	Charley	Masonry	yb>=1994	8110
Company 1	Charley	Frame	yb<1970	956
Company 1	Charley	Frame	1970<=yb<1984	1232
Company 1	Charley	Frame	1984<=yb<1994	3044
Company 1	Charley	Frame	yb>=1994	677
Company 1	Charley	Manufactured	yb<1994	2966
Company 1	Charley	Manufactured	yb>=1994	212
Company 1	Frances	Masonry	yb<1970	5009
Company 1	Frances	Masonry	1970<=yb<1984	6989
Company 1	Frances	Masonry	1984<=yb<1994	7903
Company 1	Frances	Masonry	yb>=1994	4384
Company 1	Frances	Frame	yb<1970	902
Company 1	Frances	Frame	1970<=yb<1984	2081
Company 1	Frances	Frame	1984<=yb<1994	5648
Company 1	Frances	Frame	yb>=1994	721
Company 1	Frances	Manufactured	yb<1994	3186
Company 1	Frances	Manufactured	yb>=1994	222
Company 1	Ivan	Masonry	yb<1970	2029
Company 1	Ivan	Masonry	1970<=yb<1984	2099
Company 1	Ivan	Masonry	1984<=yb<1994	1719
Company 1	Ivan	Masonry	yb>=1994	1769
Company 1	Ivan	Frame	yb<1970	3048
Company 1	Ivan	Frame	1970<=yb<1984	3956
Company 1	Ivan	Frame	1984<=yb<1994	4829
Company 1	Ivan	Frame	yb>=1994	3890
Company 1	Ivan	Manufactured	yb<1994	634
Company 1	Ivan	Manufactured	yb>=1994	79
Company 1	Jeanne	Masonry	yb<1970	3601
Company 1	Jeanne	Masonry	1970<=yb<1984	5274
Company 1	Jeanne	Masonry	1984<=yb<1994	5698
Company 1	Jeanne	Masonry	yb>=1994	4999
Company 1	Jeanne	Frame	yb<1970	825
Company 1	Jeanne	Frame	1970<=yb<1984	1386
Company 1	Jeanne	Frame	1984<=yb<1994	3430
Company 1	Jeanne	Frame	yb>=1994	674
Company 1	Jeanne	Manufactured	yb<1994	2717
Company 1	Jeanne	Manufactured	yb>=1994	177

Table 18. Company 2: Claim number for each year-built category.

Company	Hurricane	Construction	Year Built	Number of Claims
Company 2	Charley	Masonry	yb<1970	8677
Company 2	Charley	Masonry	1970<=yb<1984	15085
Company 2	Charley	Masonry	1984<=yb<1994	18324
Company 2	Charley	Masonry	yb>=1994	6376
Company 2	Charley	Frame	yb<1970	1920
Company 2	Charley	Frame	1970<=yb<1984	1782
Company 2	Charley	Frame	1984<=yb<1994	3786
Company 2	Charley	Frame	yb>=1994	443
Company 2	Charley	Manufactured	yb<1994	1843
Company 2	Charley	Manufactured	yb>=1994	159
Company 2	Frances	Masonry	yb<1970	8276
Company 2	Frances	Masonry	1970<=yb<1984	11978
Company 2	Frances	Masonry	1984<=yb<1994	11394
Company 2	Frances	Masonry	yb>=1994	3224
Company 2	Frances	Frame	yb<1970	1453
Company 2	Frances	Frame	1970<=yb<1984	3202
Company 2	Frances	Frame	1984<=yb<1994	7731
Company 2	Frances	Frame	yb>=1994	601
Company 2	Frances	Manufactured	yb<1994	1590
Company 2	Frances	Manufactured	yb>=1994	131
Company 2	Ivan	Masonry	yb<1970	1399
Company 2	Ivan	Masonry	1970<=yb<1984	746
Company 2	Ivan	Masonry	1984<=yb<1994	449
Company 2	Ivan	Masonry	yb>=1994	275
Company 2	Ivan	Frame	yb<1970	4004
Company 2	Ivan	Frame	1970<=yb<1984	5546
Company 2	Ivan	Frame	1984<=yb<1994	4637
Company 2	Ivan	Frame	yb>=1994	2229
Company 2	Ivan	Manufactured	yb<1994	171
Company 2	Ivan	Manufactured	yb>=1994	41
Company 2	Jeanne	Masonry	yb<1970	6907
Company 2	Jeanne	Masonry	1970<=yb<1984	10767
Company 2	Jeanne	Masonry	1984<=yb<1994	9629
Company 2	Jeanne	Masonry	yb>=1994	4176
Company 2	Jeanne	Frame	yb<1970	1555
Company 2	Jeanne	Frame	1970<=yb<1984	2087
Company 2	Jeanne	Frame	1984<=yb<1994	4561
Company 2	Jeanne	Frame	yb>=1994	484
Company 2	Jeanne	Manufactured	yb<1994	1401
Company 2	Jeanne	Manufactured	yb>=1994	128

Table 19. Company 1 and Company 2: Claim numbers combined.

Company	Hurricane	Construction	Number of Claims
Company 1	Charley	Masonry	33202
Company 1	Charley	Frame	5909
Company 1	Charley	Manufactured	3178
Company 1	Charley	Other	260
Company 1	Frances	Masonry	24285
Company 1	Frances	Frame	9352
Company 1	Frances	Manufactured	3408
Company 1	Frances	Other	566
Company 1	Ivan	Masonry	7616
Company 1	Ivan	Frame	15723
Company 1	Ivan	Manufactured	713
Company 1	Ivan	Other	100
Company 1	Jeanne	Masonry	19572
Company 1	Jeanne	Frame	6315
Company 1	Jeanne	Manufactured	2894
Company 1	Jeanne	Other	331
Company 2	Charley	Masonry	48462
Company 2	Charley	Frame	7931
Company 2	Charley	Manufactured	2002
Company 2	Charley	Other	582
Company 2	Frances	Masonry	34872
Company 2	Frances	Frame	12987
Company 2	Frances	Manufactured	1721
Company 2	Frances	Other	1134
Company 2	Ivan	Masonry	2869
Company 2	Ivan	Frame	16416
Company 2	Ivan	Manufactured	212
Company 2	Ivan	Other	87
Company 2	Jeanne	Masonry	31479
Company 2	Jeanne	Frame	8687
Company 2	Jeanne	Manufactured	1529
Company 2	Jeanne	Other	1167

The claims are divided by the type of coverage for structure and contents. Company 1 has two types of coverage, replacement cost and actual cash value, but does not specify whether both structure and contents have the same coverage for each claim.

For Company 2, there are six types of coverage, as shown below.

ACV S/ACV C	Structure Actual-Cash-Value, Contents Actual-Cash-Value
ACV S/RC C	Structure Actual-Cash-Value, Contents Replacement-Cost
RC S/ACV C	Structure Replacement-Cost, Contents Actual-Cash-Value
RC S/RC C	Structure Replacement-Cost, Contents Replacement-Cost
SV S/RC C	Structure Stated-Value, Contents Replacement-Cost
SV S/SV C	Structure Stated-Value, Contents Stated-Value

Table 20~~a~~ and ~~Table 21b~~**Error! Reference source not found.**~~Table 21~~ summarize the distribution of claims in both companies.

Table 20~~a.~~ Distribution of coverage for Company 1.

Coverage	Premium Policy Count		Claim Policy Count	
A	44020	1%	2759	2%
R	3706219	99%	163692	98%
Total	3750240		166451	

Table ~~21b~~21. Distribution of coverage for Company 2.

Coverage	Premium Policy Count		Claim Policy Count	
ACV S/ACV C	13173	3%	3496	3%
ACV S/RC C	44805	10%	12150	9%
RC S/ACV C	162122	35%	41484	30%
RC S/RC C	232688	51%	77146	57%
SV S/RC C	235	0%	69	0%
SV S/SV C	6019	1%	1717	1%
Total	459042	100%	136062	100%

There are 29,372 claims with $0 losses (i.e., Loss structure + Loss app + Loss contents + Loss ALE = 0), though they are listed in the claim file of Company 2. They probably correspond to claims whose losses were lower than the deductible.

2004 Personal Residential Claims Data

New claims data for the 2004 hurricane season from a series of insurance companies were also used to validate the FPHLM. Four new insurance companies provided claims data for the 2004 hurricane season. They will be referred to as companies PR2 to 5-2004. Company PR5-2004 has only manufactured homes. See Table PR04a to q. The claims data for Ivan was not used in the validation process because it was contaminated by storm surge damage.

Table 22~~a.~~ 2004 Personal Residential Claims Data

PR04a. Distribution of claims per hurricane for PR-2004 Companies.

	PR2-2004	PR3-2004	PR4-2004	PR5-2004
Charley	12641	34149	289	8030
Frances	12731	27866	200	7,301
Ivan	6202	21424	31	817
Jeanne	11547	19975	248	10,390

PR04b. Distribution of claims per coverage for PR-2004 Companies.

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
A	0	155	0	0
R	43121	103414	768	26,538

PR04c. Distribution of claims per construction type for PR-2004 Companies.

Exterior Wall	PR2-2004	PR3-2004	PR4-2004	PR5-2004
Frame	10760	23471	198	0
Manuf. Homes	0	0	0	26,538
Masonry	31673	79911	569	0
Other	688	32	1	0

PR04d. Distribution of claims per story for PR-2004 Companies.

Stories	PR2-2004	PR3-2004	PR4-2004	PR5-2004
1	0	0	0	26,538
2	0	0	0	0
Unknown	43121	103,414	768	0

PR04e. Distribution of claims per era for PR-2004 Companies.

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
pre1960	1785	7854	125	0
1960-1970	3983	12033	102	0
1971-1980	8312	19,772	145	0
1981-1993	18621	46,525	276	0
1994-2001	5545	14,436	91	0
2002-present	4875	2,785	29	0
MH pre-1994	0	0	0	22172
MH 1994-present	0	0	0	4366

PR04f. Distribution of claims per era for PR-2004 Companies, for hurricane Charley, and construction types Frame and Manufactured Homes.

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
pre1960	119	535	20	0
1960-1970	80	190	2	0
1971-1980	212	471	3	0
1981-1993	956	2752	31	0
1994-2001	128	247	8	0
2002-present	237	29	1	0
MH pre-1994	0	0	0	6665
MH 1994-present	0	0	0	1365

PR04g. Distribution of claims per era for PR-2004 Companies, for hurricane Charley, and construction type Masonry

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
pre1960	409	1870	32	0
1960-1970	972	3051	37	0
1971-1980	1909	5478	46	0
1981-1993	4674	13668	64	0
1994-2001	1580	4877	34	0
2002-present	1271	968	10	0

PR04h. Distribution of claims per era for PR-2004 Companies, for hurricane Charley, and construction type Other

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
pre1960	0	0	0	0
1960-1970	5	0	0	0
1971-1980	35	0	0	0
1981-1993	35	8	0	0
1994-2001	3	1	0	0
2002-present	16	0	0	0

PR04i. Distribution of claims per era for PR-2004 Companies, for hurricane Frances, and construction type Frame and Manufactured Homes

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
pre1960	110	419	7	0
1960-1970	96	218	4	0
1971-1980	555	922	6	0
1981-1993	2845	5689	24	0
1994-2001	265	311	8	0
2002-present-	358	30	3	0
MH pre-1994	0	0	0	6145
MH 1994-present	0	0	0	1156

PR04j. Distribution of claims per era for PR-2004 Companies, for hurricane Frances, and construction type Masonry

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
pre1960	348	1433	15	0
1960-1970	1043	3181	27	0
1971-1980	1906	4770	34	0
1981-1993	3129	8165	56	0
1994-2001	954	2206	15	0
2002-present	864	511	1	0

PR04k. Distribution of claims per era for PR-2004 Companies, for hurricane Frances, and construction type Other

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
pre1960	0	0	0	0
1960-1970	8	0	0	0
1971-1980	50	2	0	0
1981-1993	114	4	0	0
1994-2001	5	3	0	0
2002-present	81	0	0	0

PR04l. Distribution of claims per era for PR-2004 Companies, for hurricane Ivan, and construction type Frame and Manufactured Homes

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
pre1960	140	914	4	0
1960-1970	117	538	2	0
1971-1980	174	759	2	0
1981-1993	626	3292	4	0
1994-2001	302	1636	0	0
2002-present-	273	223	0	0
MH pre-1994	0	0	0	620
MH 1994-present	0	0	0	197

PR04m. Distribution of claims per era for PR-2004 Companies, for hurricane Ivan, and construction type Masonry

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
pre1960	151	1,207	4	0
1960-1970	624	2,557	4	0
1971-1980	1279	3,573	3	0
1981-1993	1320	4,087	6	0
1994-2001	676	2,251	2	0
2002-present	467	378	0	0

PR04n. Distribution of claims per era for PR-2004 Companies, for hurricane Ivan, and construction type Other

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
pre1960	1	0	0	0
1960-1970	0	0	0	0
1971-1980	12	1	0	0
1981-1993	23	2	0	0
1994-2001	3	3	0	0
2002-present	13	1	0	0

PR04o. Distribution of claims per era for PR-2004 Companies, for hurricane Jeanne, and construction type Frame and Manufactured Homes

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
pre1960	137	376	16	0
1960-1970	81	166	2	0
1971-1980	399	493	9	0
1981-1993	1983	2939	30	0
1994-2001	276	296	10	0
2002-present-	290	24	2	0
MH pre-1994	0	0	0	8742
MH 1994-present	0	0	0	1648

PR04p. Distribution of claims per era for PR-2004 Companies, for hurricane Jeanne, and construction type Masonry

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
pre1960	369	1,100	26	0
1960-1970	951	2,132	24	0
1971-1980	1716	3,303	42	0
1981-1993	2795	5,915	61	0
1994-2001	1340	2,604	14	0
2002-present	926	619	12	0

PR04q. Distribution of claims per era for PR-2004 Companies, for hurricane Jeanne, and construction type Other

Year Built	PR2-2004	PR3-2004	PR4-2004	PR5-2004
pre1960	1	0	0	0
1960-1970	5	0	0	0
1971-1980	65	0	0	0
1981-1993	121	4	0	0
1994-2001	13	1	0	0
2002-present	79	2	0	0

2005 Personal Residential Claims Data

Claims data for the 2005 hurricane season from a series of insurance companies were also used to validate the FPHLM. Five insurance companies provided claims data for the 2005 hurricane season. They will be referred to as companies PR1 to 5-2005. Company PR5-2005 has only manufactured homes. See Table PR05a to q. The data for hurricane Rita was not used given the small number of claims.

Table ~~22b~~23. 2005 Personal Residential Claims Data

PR05a. Distribution of claims per hurricane for PR-2005 Companies.

	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
Dennis	3968	1251	3,467	9	232
Katrina	5382	201	2,379	30	78
Rita	56	34	0	1	4
Wilma	62677	9247	21328	264	5,302

PR05b. Distribution of claims per coverage for PR-2005 Companies.

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
A	5990	10733	43	304	0
R	66093	0	27,131	0	5616

PR05c. Distribution of claims per construction type for PR-2005 Companies.

Exterior Wall	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
Frame	6920	1629	2,881	44	0
Manuf. Homes	1402	0	0	0	5616
Masonry	60475	8538	24,292	258	0
Other	3286	566	1	2	0

PR05d. Distribution of claims per story for PR-2005 Companies.

Stories	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
1	664	0	0	0	0
2	146	0	0	0	0
Unknown	71273	10733	27,174	304	0

PR05e. Distribution of claims per era for PR-2005 Companies.

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
pre1960	6204	233	2,526	47	0
1960-1970	10865	770	3,715	58	0
1971-1980	18922	2441	7172	69	0
1981-1993	26412	4498	10202	98	0
1994-2001	7172	1571	2,908	28	0
2002-present	1106	1220	649	4	0
MH pre-1994	1274	0	0	0	4227
MH 1994-present	128	0	0	0	1389

PR05f. Distribution of claims per era for PR-2005 Companies, for hurricane Dennis, and construction type Frame.

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
pre1960	242	26	106	1	0
1960-1970	541	26	73	1	0
1971-1980	815	33	128	2	0
1981-1993	1046	112	452	0	0
1994-2001	573	77	422	0	0
2002-present	66	45	59	0	0
MH pre-1994	36	0	0	0	162
MH 1994-present	18	0	0	0	70

PR05g. Distribution of claims per era for PR-2005 Companies, for hurricane Dennis, and construction type Masonry

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
pre1960	93	21	150	1	0
1960-1970	175	110	324	1	0
1971-1980	140	237	537	2	0
1981-1993	124	255	535	1	0
1994-2001	70	218	562	0	0
2002-present-	12	89	118	0	0

PR05h. Distribution of claims per era for PR-2005 Companies, for hurricane Dennis, and construction type Other

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
pre1960	0	0	0	0	0
1960-1970	0	0	0	0	0
1971-1980	6	0	0	0	0
1981-1993	11	1	0	0	0
1994-2001	0	0	1	0	0
2002-present	0	1	0	0	0

PR05i. Distribution of claims per era for PR-2005 Companies, for hurricane Katrina, and construction type Frame

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
pre1960	60	1	25	0	0
1960-1970	40	1	8	0	0
1971-1980	43	3	10	0	0
1981-1993	91	9	52	0	0
1994-2001	44	3	20	0	0
2002-present	8	4	6	0	0
MH pre-1994	45	0	0	0	68
MH 1994-present	1	0	0	0	10

PR05j. Distribution of claims per era for PR-2005 Companies, for hurricane Katrina, and construction type Masonry

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
pre1960	969	10	410	12	0
1960-1970	1137	26	456	10	0
1971-1980	1428	48	583	4	0
1981-1993	1297	53	727	4	0
1994-2001	133	27	74	0	0
2002-present	23	12	8	0	0

PR05k. Distribution of claims per era for PR-2005 Companies, for hurricane Katrina, and construction type Other

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
pre1960	1	0	0	0	0
1960-1970	14	0	0	0	0
1971-1980	31	1	0	0	0
1981-1993	13	2	0	0	0
1994-2001	4	0	0	0	0
2002-present	0	1	0	0	0

PR05l. Distribution of claims per era for PR-2005 Companies, for hurricane Rita, and construction type Frame

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
pre1960	0	0	0	0	0
1960-1970	1	0	0	0	0
1971-1980	1	2	0	0	0
1981-1993	0	1	0	1	0
1994-2001	0	0	0	0	0
2002-present	0	2	0	0	0
MH pre-1994	1	0	0	0	4
MH 1994-present	0	0	0	0	0

PR05m. Distribution of claims per era for PR-2005 Companies, for hurricane Rita, and construction type Masonry

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
pre1960	6	1	0	0	0
1960-1970	13	2	0	0	0
1971-1980	14	7	0	0	0
1981-1993	17	7	0	0	0
1994-2001	2	10	0	0	0
2002-present	0	1	0	0	0

PR05n. Distribution of claims per era for PR-2005 Companies, for hurricane Rita, and construction type Other

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
pre1960	0	0	0	0	0
1960-1970	0	0	0	0	0
1971-1980	1	0	0	0	0
1981-1993	0	1	0	0	0
1994-2001	0	0	0	0	0
2002-present	0	0	0	0	0

PR05o. Distribution of claims per era for PR-2005 Companies, for hurricane Wilma, and construction type Frame

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
pre1960	323	32	99	2	0
1960-1970	151	51	47	1	0
1971-1980	546	213	212	7	0
1981-1993	2136	786	1084	25	0
1994-2001	164	114	70	4	0
2002-present	29	88	8	0	0
MH pre-1994	1192	0	0	0	3993
MH 1994-present	109	0	0	0	1309

PR05p. Distribution of claims per era for PR-2005 Companies, for hurricane Wilma, and construction type Masonry

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
pre1960	4484	142	1736	31	0
1960-1970	8567	542	2,807	45	0
1971-1980	14288	1721	5702	54	0
1981-1993	20430	3079	7352	65	0
1994-2001	6089	1103	1759	24	0
2002-present-	964	817	450	4	0

PR05q. Distribution of claims per era for PR-2005 Companies, for hurricane Wilma, and construction type Other

Year Built	PR1-2005	PR2-2005	PR3-2005	PR4-2005	PR5-2005
pre1960	26	0	0	0	0
1960-1970	226	12	0	0	0
1971-1980	1609	176	0	0	0
1981-1993	1247	192	0	2	0
1994-2001	93	19	0	0	0
2002-present-	4	160	0	0	0

Commercial Residential Claims Data

Claims data from the 2004 and the 2005 hurricane seasons for commercial residential from four insurance companies (referred to as companies CR1 to 4) were used to validate the commercial residential module of the FPHLM. The details are given below for low rise commercial and for mid/high rise commercial in Tables CR04-LRa to q, CR05-LRa to n, CR04-MRa to q, and CR05-MRa to k. The vast majority of the claims are for low-rise 1 and 2 story buildings.

The policies for company CR2 included commercial line accounts (CLA) for condominium association, apartment building, and homeowners association policies, and the policies for company CR3 included high risk accounts (HRA) in coastal areas.

2004 Low Rise Commercial Residential Claims Data

It is clear from Tables CR04-LRa to q that the vast majority of LR 2004 claims data consists of masonry one and two story tall pre-1994 buildings.

Table ~~22c~~24. 2004 Low Rise Commercial Residential Claims Data

CR04-LRa. Distribution of claims per hurricane for CR LR 2004 companies.

	CR1-LR04	CR2-LR04	CR3-LR04
Charley	575	11	182
Frances	691	78	808
Ivan	166	0	0
Jeanne	285	12	280

CR04-LRb. Distribution of claims per coverage for CR LR 2004 companies.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
A	0	0	0
R	1717	0	0
Not Provided	0	101	1270

CR04-LRc. Distribution of claims per construction type for CR LR 2004 companies.

Exterior Wall	CR1-LR04	CR2-LR04	CR3-LR04
Frame	405	28	240
Masonry	1204	73	1030
Other	108	0	0

CR04-LRd. Distribution of claims per story for CR LR 2004 companies.

Stories	CR1-LR04	CR2-LR04	CR3-LR04
1	806	24	441
2	789	69	677
3	122	8	152

CR04-LRe. Distribution of claims per era for CR LR 2004 companies.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
pre1960	69	1	273
1960-1970	155	28	279
1971-1980	452	31	389
1981-1993	987	41	286
1994-2001	51	0	34
2002-present	3	0	9

CR04-LRf. Distribution of claims per era for CR LR 2004 companies, for hurricane Charley, and construction type Frame.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
pre1960	12	0	20
1960-1970	1	0	11
1971-1980	6	7	19
1981-1993	50	4	20
1994-2001	2	0	2
2002-present	0	0	0

CR04-LRg. Distribution of claims per era for CR LR 2004 companies, for hurricane Charley, and construction type Masonry.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
pre1960	10	0	12
1960-1970	33	0	17
1971-1980	153	0	45
1981-1993	290	0	26
1994-2001	9	0	10
2002-present	0	0	0

CR04-LRh. Distribution of claims per era for CR LR 2004 companies, for hurricane Charley, and construction type Other.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
pre1960	0	0	0
1960-1970	0	0	0
1971-1980	3	0	0
1981-1993	6	0	0
1994-2001	0	0	0
2002-present	0	0	0

CR04-LRi. Distribution of claims per era for CR LR 2004 companies, for hurricane Frances, and construction type Frame.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
pre1960	8	1	58
1960-1970	3	0	11
1971-1980	6	3	22
1981-1993	119	7	33
1994-2001	12	0	3
2002-present	0	0	0

CR04-LRj. Distribution of claims per era for CR LR 2004 companies, for hurricane Frances, and construction type Masonry.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
pre1960	11	0	111
1960-1970	69	25	169
1971-1980	152	17	214
1981-1993	206	25	165
1994-2001	11	0	16
2002-present	2	0	6

CR04-LRk. Distribution of claims per era for CR LR 2004 companies, for hurricane Frances, and construction type Other.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
pre1960	0	0	0
1960-1970	0	0	0
1971-1980	6	0	0
1981-1993	85	0	0
1994-2001	1	0	0
2002-present	0	0	0

CR04-LRl. Distribution of claims per era for CR LR 2004 companies, for hurricane Ivan, and construction type Frame.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
pre1960	5	0	0
1960-1970	11	0	0
1971-1980	49	0	0
1981-1993	66	0	0
1994-2001	6	0	0
2002-present-	0	0	0

CR04-LRm. Distribution of claims per era for CR LR 2004 companies, for hurricane Ivan, and construction type Masonry.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
pre1960	5	0	0
1960-1970	9	0	0
1971-1980	9	0	0
1981-1993	5	0	0
1994-2001	0	0	0
2002-present-	0	0	0

CR04-LRn. Distribution of claims per era for CR LR 2004 companies, for hurricane Ivan, and construction type Other.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
pre1960	0	0	0
1960-1970	0	0	0
1971-1980	0	0	0
1981-1993	1	0	0
1994-2001	0	0	0
2002-present-	0	0	0

CR04-LRo. Distribution of claims per era for CR LR 2004 companies, for hurricane Jeanne, and construction type Frame.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
pre1960	12	0	47
1960-1970	1	0	69
1971-1980	2	1	85
1981-1993	32	5	34
1994-2001	2	0	1
2002-present-	0	0	3

CR04-LRp. Distribution of claims per era for CR LR 2004 companies, for hurricane Jeanne, and construction type Masonry.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
pre1960	6	0	47
1960-1970	28	3	69
1971-1980	64	3	85
1981-1993	124	0	34
1994-2001	7	0	1
2002-present-	1	0	3

CR04-LRq. Distribution of claims per era for CR LR 2004 companies, for hurricane Jeanne, and construction type Other.

Year Built	CR1-LR04	CR2-LR04	CR3-LR04
pre1960	0	0	0
1960-1970	0	0	0
1971-1980	2	0	0
1981-1993	3	0	0
1994-2001	0	0	0
2002-present-	0	0	0

2005 Low Rise Commercial Residential Claims Data

It is clear from Tables CR05-LRa to n that the vast majority of LR 2005 claims data consists of masonry one and two story tall pre-1994 buildings for hurricane Wilma.

Table ~~22d~~25. 2005 Low Rise Commercial Residential Claims Data

CR05-LRa. Distribution of claims per hurricane for CR LR 2005 companies.

	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
Dennis	22	0	0	0
Katrina	68	81	186	0
Wilma	1117	1356	2080	410

CR05-LRb. Distribution of claims per coverage for CR LR 2005 companies.

Year Built	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
A	0	0	0	0
R	1207	0	0	0
Not Provided	0	1437	2266	410

CR05-LRc. Distribution of claims per construction type for CR LR 2005 companies.

Exterior Wall	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
Frame	180	168	102	47
Masonry	933	1269	2164	363
Other	94	0	0	0

CR05-LRd. Distribution of claims per story for CR LR 2005 companies.

Stories	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
1	645	458	955	180
2	498	863	1111	221
3	64	116	200	9

CR05-LRe. Distribution of claims per era for CR LR 2005 companies.

Year Built	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
pre1960	3	112	644	0
1960-1970	98	229	743	0
1971-1980	279	501	559	6
1981-1993	811	578	270	119
1994-2001	16	17	35	196
2002-present	0	0	15	89

CR05-LRf. Distribution of claims per era for CR LR 2005 companies, for hurricane Dennis, and construction type Frame.

Year Built	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
pre1960	0	0	0	0
1960-1970	0	0	0	0
1971-1980	2	0	0	0
1981-1993	12	0	0	0
1994-2001	7	0	0	0
2002-present-	0	0	0	0

CR05-LRg. Distribution of claims per era for CR LR 2005 companies, for hurricane Dennis, and construction type Masonry.

Year Built	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
pre1960	0	0	0	0
1960-1970	0	0	0	0
1971-1980	1	0	0	0
1981-1993	0	0	0	0
1994-2001	0	0	0	0
2002-present-	0	0	0	0

CR05-LRh. Distribution of claims per era for CR LR 2005 companies, for hurricane Dennis, and construction type Other.

Year Built	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
pre1960	0	0	0	0
1960-1970	0	0	0	0
1971-1980	0	0	0	0
1981-1993	0	0	0	0
1994-2001	0	0	0	0
2002-present	0	0	0	0

CR05-LRi. Distribution of claims per era for CR LR 2005 companies, for hurricane Katrina, and construction type Frame.

Year Built	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
pre1960	0	0	2	0
1960-1970	0	0	0	0
1971-1980	1	0	1	0
1981-1993	2	6	1	0
1994-2001	0	0	0	0
2002-present	0	0	0	0

CR05-LRj. Distribution of claims per era for CR LR 2005 companies, for hurricane Katrina, and construction type Masonry.

Year Built	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
pre1960	0	13	62	0
1960-1970	3	9	61	0
1971-1980	4	29	29	0
1981-1993	54	23	23	0
1994-2001	0	1	5	0
2002-present	0	0	2	0

CR05-LRk. Distribution of claims per era for CR LR 2005 companies, for hurricane Katrina, and construction type Other.

Year Built	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
pre1960	0	0	0	0
1960-1970	0	0	0	0
1971-1980	0	0	0	0
1981-1993	4	0	0	0
1994-2001	0	0	0	0
2002-present	0	0	0	0

CR05-LRl. Distribution of claims per era for CR LR 2005 companies, for hurricane Wilma, and construction type Frame.

Year Built	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
pre1960	2	4	46	0
1960-1970	93	0	20	0
1971-1980	248	11	12	0
1981-1993	525	147	19	9
1994-2001	4	0	1	29
2002-present	0	0	0	9

CR05-LRm. Distribution of claims per era for CR LR 2005 companies, for hurricane Wilma, and construction type Masonry.

Year Built	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
pre1960	1	95	534	0
1960-1970	93	220	662	0
1971-1980	248	461	517	6
1981-1993	525	402	227	110
1994-2001	4	16	29	167
2002-present	0	0	13	80

CR05-LRn. Distribution of claims per era for CR LR 2005 companies, for hurricane Wilma, and construction type Other.

Year Built	CR1-LR05	CR2-LR05	CR3-LR05	CR4-LR05
pre1960	0	0	0	0
1960-1970	1	0	0	0
1971-1980	21	0	0	0
1981-1993	64	0	0	0
1994-2001	4	0	0	0
2002-present	0	0	0	0

2004 Mid/High Rise Commercial Residential Claims Data

Table ~~22e~~26. 2004 Mid/High Rise Commercial Residential Claims Data

It is clear from Tables CR04-MRa to n that the number of MHR 2004 claims is very small. It consists mainly of masonry or other four to eleven story tall pre-1994 buildings.

CR04-MRa. Distribution of claims per hurricane for CR MHR 2004 companies.

	CR1-MHR04	CR2-MHR04	CR3-MHR04
Charley	23	4	34
Frances	21	5	56
Jeanne	4	0	15

CR04-MRb. Distribution of claims per coverage for CR MHR 2004 companies.

Year Built	CR1-MHR04	CR2-MHR04	CR3-MHR04
A	0	0	0
R	48	0	0
Not Provided	0	9	105

CR04-MRc. Distribution of claims per construction type for CR MHR 2004 companies.

Exterior Wall	CR1-MHR04	CR2-MHR04	CR3-MHR04
Frame	2	0	2
Masonry	34	9	103
Other	12	0	0

CR04-MRd. Distribution of claims per story for CR MHR 2004 companies.

Stories	CR1-MHR04	CR2-MHR04	CR3-MHR04
4	11	1	23
5	14	7	28
6	5	0	8
7	6	0	15
8	2	1	7
9	2	0	4
10	8	0	2
11	0	0	2
12	0	0	1
13	0	0	1
15	0	0	1
26	0	0	1
36	0	0	1
42	0	0	1

CR04-MRe. Distribution of claims per era for CR MHR 2004 companies.

Year Built	CR1-MHR04	CR2-MHR04	CR3-MHR04
pre1960	1	0	4
1960-1970	1	1	8
1971-1980	21	4	35
1981-1993	25	4	50
1994-2001	0	0	7
2002-present	0	0	1

CR04-MRf. Distribution of claims per era for CR MHR 2004 companies, for hurricane Charley, and construction type Frame.

Year Built	CR1-MHR04	CR2-MHR04	CR3-MHR04
pre1960	0	0	0
1960-1970	0	0	0
1971-1980	0	0	0
1981-1993	0	0	0
1994-2001	0	0	0
2002-present	0	0	0

CR04-MRg. Distribution of claims per era for CR MHR 2004 companies, for hurricane Charley, and construction type Masonry.

Year Built	CR1-MHR04	CR2-MHR04	CR3-MHR04
pre1960	0	0	0
1960-1970	0	0	2
1971-1980	10	4	9
1981-1993	10	0	20
1994-2001	0	0	3
2002-present	0	0	0

CR04-MRh. Distribution of claims per era for CR MHR 2004 companies, for hurricane Charley, and construction type Other.

Year Built	CR1-MHR04	CR2-MHR04	CR3-MHR04
pre1960	0	0	0
1960-1970	0	0	0
1971-1980	1	0	0
1981-1993	2	0	0
1994-2001	0	0	0
2002-present	0	0	0

CR04-MRi. Distribution of claims per era for CR MHR 2004 companies, for hurricane Frances, and construction type Frame.

Year Built	CR1-MHR04	CR2-MHR04	CR3-MHR04
pre1960	0	0	1
1960-1970	0	0	0
1971-1980	0	0	0
1981-1993	2	0	0
1994-2001	0	0	0
2002-present	0	0	0

CR04-MRj. Distribution of claims per era for CR MHR 2004 companies, for hurricane Frances, and construction type Masonry.

Year Built	CR1-MHR04	CR2-MHR04	CR3-MHR04
pre1960	1	0	3
1960-1970	0	1	3
1971-1980	9	0	23
1981-1993	3	4	22
1994-2001	0	0	3
2002-present	0	0	1

CR04-MRk. Distribution of claims per era for CR MHR 2004 companies, for hurricane Frances, and construction type Other.

Year Built	CR1-MHR04	CR2-MHR04	CR3-MHR04
pre1960	0	0	0
1960-1970	0	0	0
1971-1980	1	0	0
1981-1993	5	0	0
1994-2001	0	0	0
2002-present	0	0	0

CR04-MRl. Distribution of claims per era for CR MHR 2004 companies, for hurricane Jeanne, and construction type Frame.

Year Built	CR1-MHR04	CR2-MHR04	CR3-MHR04
pre1960	0	0	0
1960-1970	0	0	0
1971-1980	0	0	0
1981-1993	0	0	1
1994-2001	0	0	0
2002-present	0	0	0

CR04-MRm. Distribution of claims per era for CR MHR 2004 companies, for hurricane Jeanne, and construction type Masonry.

Year Built	CR1-MHR04	CR2-MHR04	CR3-MHR04
pre1960	0	0	0
1960-1970	0	0	3
1971-1980	0	0	3
1981-1993	1	0	7
1994-2001	0	0	1
2002-present	0	0	0

CR04-MRn. Distribution of claims per era for CR MHR 2004 companies, for hurricane Jeanne, and construction type Other.

Year Built	CR1-MHR04	CR2-MHR04	CR3-MHR04
pre1960	0	0	0
1960-1970	1	0	0
1971-1980	0	0	0
1981-1993	2	0	0
1994-2001	0	0	0
2002-present	0	0	0

2005 Mid/High Rise Commercial Residential Claims Data

It is clear from Tables CR05-MRa to k that the number of MHR 2005 claims is very small. It consists mainly of masonry four to ten story tall pre-1994 buildings for hurricane Wilma.

Table ~~22f~~27. 2005 Mid/Hid Rise Commercial Residential Claims Data

CR05-MRa. Distribution of claims per hurricane for CR MHR 2005 companies.

	CR1-MHR05	CR2-MHR05	CR3-MHR05	CR4-MHR05
Katrina	0	0	10	0
Wilma	125	118		42

CR05-MRb. Distribution of claims per coverage for CR MHR 2005 companies.

Year Built	CR1-MHR05	CR2-MHR05	CR3-MHR05	CR4-MHR05
A	0	0	0	0
R	126	0	0	0
Not Provided	0	118	127	42

CR05-MRc. Distribution of claims per construction type for CR MHR 2005 companies.

Exterior Wall	CR1-MHR05	CR2-MHR05	CR3-MHR05	CR4-MHR05
Frame	0	0	1	0
Masonry	107	118	127	42
Other	19	0	0	0

CR05-MRd. Distribution of claims per story for CR MHR 2005 companies.

Stories	CR1-MHR05	CR2-MHR05	CR3-MHR05	CR4-MHR05
4	64	70	54	40
5	17	37	29	0
6	8	3	12	0
7	13	2	6	0
8	9	1	7	0
9	4	4	3	0
10	11	1	3	0
11	0	0	1	0
14	0	0	2	0
15	0	0	2	0
16	0	0	2	0
17	0	0	0	2

18	0	0	1	0
19	0	0	1	0
22	0	0	1	0
23	0	0	1	0
29	0	0	1	0
31	0	0	1	0

CR05-MRe. Distribution of claims per era for CR MHR 2005 companies.

Year Built	CR1-MHR05	CR2-MHR05	CR3-MHR05	CR4-MHR05
pre1960	1	0	8	0
1960-1970	1	6	42	0
1971-1980	52	52	38	0
1981-1993	65	60	34	28
1994-2001	7	0	3	12
2002-present	0	0	2	2

CR05-MRf. Distribution of claims per era for CR MHR 2005 companies, for hurricane Katrina, and construction type Frame.

Year Built	CR1-MHR05	CR2-MHR05	CR3-MHR05	CR4-MHR05
pre1960	0	0	0	0
1960-1970	0	0	0	0
1971-1980	0	0	0	0
1981-1993	0	0	0	0
1994-2001	0	0	0	0
2002-present	0	0	0	0

CR05-MRg. Distribution of claims per era for CR MHR 2005 companies, for hurricane Katrina, and construction type Masonry.

Year Built	CR1-MHR05	CR2-MHR05	CR3-MHR05	CR4-MHR05
pre1960	0	0	1	0
1960-1970	0	0	4	0
1971-1980	0	0	3	0
1981-1993	0	0	1	0
1994-2001	0	0	1	0
2002-present	0	0	0	0

CR05-MRh. Distribution of claims per era for CR MHR 2005 companies, for hurricane Katrina, and construction type Other

Year Built	CR1-MHR05	CR2-MHR05	CR3-MHR05	CR4-MHR05
pre1960	0	0	0	0
1960-1970	0	0	0	0
1971-1980	0	0	0	0
1981-1993	0	0	0	0
1994-2001	0	0	0	0
2002-present	0	0	0	0

CR05-MRi. Distribution of claims per era for CR MHR 2005 companies, for hurricane Wilma, and construction type Frame

Year Built	CR1-MHR05	CR2-MHR05	CR3-MHR05	CR4-MHR05
pre1960	0	0	0	0
1960-1970	0	0	0	0
1971-1980	0	0	0	0
1981-1993	0	0	1	0
1994-2001	0	0	0	0
2002-present	0	0	0	0

CR05-MRj. Distribution of claims per era for CR MHR 2005 companies, for hurricane Wilma, and construction type Masonry

Year Built	CR1-MHR05	CR2-MHR05	CR3-MHR05	CR4-MHR05
pre1960	1	0	7	0
1960-1970	1	6	38	0
1971-1980	40	52	35	0
1981-1993	57	60	32	28
1994-2001	7	0	2	12
2002-present	0	0	2	2

CR05-MRk. Distribution of claims per era for CR MHR 2005 companies, for hurricane Wilma, and construction type Other

Year Built	CR1-MHR05	CR2-MHR05	CR3-MHR05	CR4-MHR05
pre1960	0	0	0	0
1960-1970	0	0	0	0
1971-1980	11	0	0	0
1981-1993	8	0	0	0
1994-2001	0	0	0	0
2002-present	0	0	0	0

4. ***Describe the data, methods, and processes used for the development of the building vulnerability functions.***

A detailed discussion of the data, methods, and processes used for the development of the building vulnerability functions is contained within Standard G.1 and other disclosure items in Standard V.1.

5. ***Summarize site inspections, including the source, and provide a brief description of the resulting use of these data in development, validation, or verification of building vulnerability functions.***

The documentation and statistical analysis of damage caused by landfalling hurricanes has been conducted by a variety of stakeholders, including home builder trade associations (NAHB Research Center, 1993, 1996, 1999; Crandell, 1998), practicing engineers (Keith & Rose, 1994), government agencies (Oliver & Hanson, 1994; FEMA, 1992, 2006), and academic researchers (Kareem, 1985, 1986; Gurley, 2006; Gurley et al., 2006). Some of these studies provide a broad overview of structural performance (FEMA and NAHB reports). Others focus on a particular building component such as roofing (Croft et al., 2006; Meloy et al., 2007) or address a specific building type such as wood frame residential construction (van de Lindt et al., 2007). All such available public access literature regarding the performance of residential infrastructure in hurricane winds was reviewed and used as guidance for the development of the vulnerability model. Those studies that provide statistical assessments of damage to specific building components (Gurley, 2006; Gurley et al., 2006; Gurley and Masters, 2011; Meloy et al., 2007) were used as a means of validating the physical damage estimates of the model. Studies that are more qualitative in nature (e.g., FEMA reports) were used to provide guidance regarding the potential failure modes that were important to replicate in the model. For example, the common observation of gable end failures resulted in a gable end failure component in the model.

Several damage surveys were done in 2004. Damage from Hurricane Charley was reported across the state, and the most severe damage occurred where the eye made landfall near the cities of Punta Gorda and Port Charlotte. A team that consisted of approximately 30 members from UF, FIU, Clemson, and FIT, under the leadership of the Insurance Institute for Business & Home Safety (IBHS), surveyed the extent of the structural damage to homes and manufactured homes in these cities. For several days following the storm the team conducted a detailed statistical survey of damage in the impacted areas. Results of this survey can be found on the IBHS website http://www.ibhs.org/. Other information regarding the damage of Charley and other storms can be found at the Florida Tech Wind and Hurricane Impact Research Laboratory website, http://www.fit.edu/research/whirl/.

Damage from Hurricane Frances was surveyed in areas from Cocoa Beach to Stuart in eastern Florida. Although damage from Hurricane Frances was not as severe as that from Hurricane Charley, the same extensive survey conducted in Punta Gorda and Port Charlotte was also conducted in the impacted areas. Great efforts were made to monitor the strength and resulting damage from the storm as part of the Florida Coastal Monitoring Program. Towers were set up to record wind speeds along the coast in locations where the storm was forecasted to make landfall. Sensors to record the wind-induced pressure were deployed on the roofs of several homes. Following the storm, members of the same team that surveyed damage from Charley photographed and recorded damage throughout the area. Areas of Fort Pierce appeared to be hardest hit and damage was severe to many homes in some areas.

Similar efforts to monitor the winds and survey the damage were made for Hurricane Jeanne. Towers and pressure sensors were again deployed at various locations near where landfall was forecasted. After the storm, members of the team surveyed areas from Stuart to Cocoa Beach. These surveys consisted primarily of cataloging and photographing various observations of damage in the impacted areas, as was done with Hurricane Frances. Damage from Hurricane Jeanne in many locations was very similar to what was seen from Hurricane Frances. In many cases damage to structures that was initially caused by Frances was compounded by Hurricane

Jeanne. Fatigue of structures from the winds of two hurricanes within three weeks most likely played a role in the most severe cases of damage in the areas such as Vero Beach and Fort Pierce. In some areas most of the weak trees and components of homes (shingles, screened porches, fences, etc.) were already damaged by Hurricane Frances, so when Hurricane Jeanne hit little or no further damage was seen. It is very difficult to tell what damage was caused by Hurricane Jeanne and what was caused by Hurricane Frances.

Additionally, engineers working on the physical damage model performed a detailed residential damage study after the 2004 hurricane season to assess the performance of housing built to the Florida Building Code and the Standard Building Code (Gurley, 2006; Gurley et al., 2006; Gurley and Masters, 2011). The data were collected as a part of a study conducted by UF and sponsored by the Florida Building Commission. Site-built single-family homes constructed after Hurricane Andrew-related changes to the standard building code went into effect were targeted for a detailed investigation of damage as a result of the 2004 hurricane season. This study provided a quantitative statistical comparison of the relative performance of homes built between 1994 and 2001 with the performance of those built after the 2001 Florida Building Code replaced the Standard Building Code. This evaluation was accomplished through a systematic survey of homes built from 1994 to 2004 in the areas that experienced the highest wind speeds from the 2004 storms (Charlotte, St. Lucie, Escambia, and Santa Rosa counties). Close to 200 homes were surveyed in these regions to define correlations between damage, age, and construction type. These relationships are referenced to maximum three-second gust wind speed via wind swath maps. An expanded and more detailed version of the conference publication (Gurley, 2006; Gurley et al., 2006) has appeared in the ASCE journal Natural Hazards Review (Gurley and Masters, 2011). The data from this study were used to modify the residential component capacities as this model evolved. Another source of field data is the aerial imagery collected by NOAA after Hurricane Katrina. These images provided a quantification of shingle damage relative to estimated wind speed and were used to validate the roof cover damage output from the physical damage model.

6. ***Describe the research used in the development of the model's building vulnerability functions.***

The engineering team adopted a "component approach" in the development of the vulnerability functions. Although a number of commercial loss projection models have been developed, only a handful of studies are available in the public domain to predict damage for hurricane prone areas. Boswell et al. (1999) attempted to predict the public costs of emergency management and recovery without taking into account losses to individual homeowners. In 1985, Berke et al. presented a computer system simulating economic and social losses caused by hurricane disasters, and a Vulnerability Assessment and Mapping System known as VAMS (Berke et al., 1984) enabled the user to consider various types of hurricanes with varying surges, wind patterns, and points of landfall. This information is of some interest, but it is not directly applicable to residential construction in Florida.

Pita et al. (2013, 2014) give an exhaustive account of the evolution of wind vulnerability models. Most studies for residential losses use post-disaster investigations (FEMA, 1992) or available claims data to fit damage versus wind speed vulnerability curves. For example, a relationship between home damage from insurance data and wind speed was proposed for Typhoons Mireille

and Flo (Mitsuta et al., 1996). Holmes (1996) presented the vulnerability curve for a fully engineered building with strength assumed to have lognormal distribution but indicated the need for more thorough post-disaster investigations to better define damage prediction models. A method for predicting the percentage of damage within an area as a function of wind speed and other parameters was presented by Sill and Kozlowski (1997). The method was intended to move away from curve fitting schemes, but its practical value was hampered by insufficient clarity and transparency. Huang et al. (2001) presented a risk assessment strategy based on an analytical expression for the vulnerability curve. The expression is obtained by regression techniques from insurance claims data for Hurricane Andrew. Khanduri and Morrow (2003) also presented a similar method of assessment of vulnerability and a methodology to translate known vulnerability curves from one region to another region. Although such approaches are simple, they are highly dependent on the construction practices common to the areas represented in the claims data. Recent changes in building codes or construction practices cannot be adequately reflected by regression-derived vulnerability curves. In addition, damage curves obtained by regression from observed data can be misleading because very often, as was the case for Hurricane Andrew, few reliable wind speed data are available. In addition, damage curves regressed from observed data do not adequately represent the influence of primary storm characteristics such as central pressure, forward velocity, radius of maximum wind, the amount of rain, duration, and other secondary parameters such as demand surge and preparedness.

In contrast, a component approach explicitly accounts for both the resistance capacity of the various building components and the load effects produced by wind to predict damage at various wind speeds. In the component approach the resistance capacity of a building can be broken down into the resistance capacity of its components and the connections between them. Damage to the structure occurs when the load effects are greater than the component's capacity to resist them. Once the strength capacities, load demands, and load path(s) are identified and modeled, the vulnerability of a structure at various wind speeds can be estimated. Estimations are affected by uncertainties regarding both the behavior and strength of the various components and the load effects produced by hurricane winds.

7. ***Describe the categories of the different building vulnerability functions. Specifically, include descriptions of the building types and characteristics, building height, number of stories, regions within the state of Florida, year of construction, and occupancy types in which a unique building vulnerability function is used. Provide the total number of building vulnerability functions available for use in the model for personal and commercial residential classifications.***

Vulnerability functions were derived for manufactured and site-built homes, for low-rise commercial residential buildings (one to three stories), and for apartment units of mid-/high-rise commercial residential buildings (four stories and higher).

A total of 4356 un-weighted vulnerability matrices were developed for site-built homes for building. The matrices correspond to different combinations of wall type (frame or masonry), region (north, central, south), subregion (high velocity hurricane zone, wind-borne debris region, inland), roof type (gable or hip), roof cover (metal, tile or shingle), window protection (shuttered or not shuttered), number of stories (one or two), and strength (weak, modified weak, retrofitted

weak; medium, modified medium, retrofitted medium; strong for inland and WBDR, strong for HVHZ—see Table 1~~Table 1a~~ and ~~Table 1b~~Table 2 in the General Standards).

These 4356 building un-weighted matrices were then combined to produce 5226 weighted matrices, and 291 age weighted matrices for site-built homes for building, for each county. Many of the matrices are repeated because many of the counties use the same regional statistics for the weighting.

A total of 648 un-weighted vulnerability matrices were developed for low-rise, commercial residential buildings for building. They correspond to different combinations of wall type (frame or masonry), sub-region (high velocity hurricane zone, wind-borne debris region, inland), roof shape (gable or hip), roof cover (metal, tile or shingle), window protection (shuttered or not shuttered), number of stories (one, two, or three), and strength (weak, medium, or strong).

These 648 matrices were then combined to produce 144 weighted curves for low-rise, commercial residential buildings for building.

180 opening vulnerability curves and 180 associated breach curves were developed for openings of apartment units of mid-/high-rise commercial residential buildings. They correspond to different combinations of building layout (open or closed), unit floor location (corner or middle unit), impact debris zone (high density impact for stories 1 to 3, medium density impact for stories 4 to 7, and low density impact for stories 8 and higher), balconies (with or without sliders) and opening protection (none, impact resistant glass, or shutters).

4 un-weighted vulnerability matrices were developed for manufactured homes for building. They correspond to four manufactured home types: (1) pre-1994—fully tied down, (2) pre-1994—not tied down, (3) post-1994—Housing and Urban Development (HUD) Zone II, and (4) post-1994—HUD Zone III. The partially tied-down homes are assumed to have a vulnerability that is an average of the vulnerabilities of fully tied-down and not tied-down homes. Because little information is available regarding the distribution of manufactured home types by size or geometry, it is assumed that all model types are single-wide manufactured homes. The modeled single-wide manufactured homes are 56 ft x 13 ft, have gable roofs, eight windows, a front entrance door, and a sliding-glass back door. The un-weighted matrices are combined into 6 weighted matrices for building, for pre-1994 (4 regions: North, Central, South, Key) and post-1994 (2 zones: II and III) manufactured homes.

8. ***Describe the process by which local construction practices and building code adoption and enforcement are considered in the model.***

In addition to a classification of building by structural types (wood or masonry walls, hip or gable roof), the buildings are classified by relative strength. Residential construction methods have evolved in Florida as experience with severe winds drives the need to reduce vulnerability.

To address this, the vulnerability team has developed strong, medium, and weak models for each site-built home and low-rise, commercial residential building structural type to represent relative quality of original construction as well as post-construction mitigation. In each region of Florida,

local construction and building code criteria are reflected in the mix of weak, medium, and strong buildings.

In the case of site-built single-family homes, the models are further refined with a modified weak to reflect pre-1960s decking practices, a retrofitted weak to model weak (older) buildings that have been reroofed and decking re-nailed, a modified medium to reflect loss of quality in the construction process in the high velocity hurricane zone before Andrew, a retrofitted medium to model medium buildings that have been reroofed and decking re-nailed, a strong model to reflect modern code requirements for inland structures and those in the WBDR but outside the HVHZ, and a strong model to reflect modern code requirements for structures within the HVHZ . A discussion of these models are provided in the Standard G-1 in the section describing the building models, and Table 1~~Table 1a~~ and ~~Table 1b~~Table 2 (also in G-1) provide an overview of the relative strength among the models stratified by the exterior components included in the models. These additions to the model inventory were prompted by detailed interviews with several experts on the evolution of construction practice (common practice, codes and enforcement) in Florida. Details of this interview process and its outcomes are addressed in the next section, and in the "Models' Distribution in Time" section in Standard G-1.

On the basis of the exposure study, it was also decided to model four manufactured home (MH) types. These types include pre-1994—fully tied down, pre-1994—not tied down, post-1994—HUD Zone II, and post-1994—HUD Zone III, where 1994 delineates older, much weaker styles of manufactured home construction than the post-1994 homes that meet minimum federal construction standards established by HUD.

Models' Distribution in Time

Over time, engineers and builders learned more about the interaction between wind and structures. More stringent building codes were enacted, which, when properly enforced, resulted in stronger structures. The weak, medium, and strong models represent this evolution of relative quality of construction in Florida. Each set of models is representative of the prevalent wind vulnerability of buildings for a certain historical period. It is therefore important to define the cut-off dates between the different periods since the overall aggregate losses in any region are determined as a mixture of homes of various strengths (ages). The cut-off dates depend on the evolution of the building code as well as the prevailing local code enforcement.

This issue of code enforcement has also evolved over time, and the State of Florida took an active role in uniform enforcement relatively recently. Thus, a given county may have built to standards that were worse than or better than the code in place at the time. After consulting with building code development experts, the team concluded that the load provisions have had some wind provisions since at least the 1970s. The classifications shown in Table 28 were adopted for characterizing the regions by age and model. The specific building eras and classifications per region are based on the evolution of the building codes in Florida and the opinions of the experts consulted. The strength descriptions within Table 28 are provided at the bottom of Table 28 in terms of the nomenclature used in Table 1~~Table 1a~~ and ~~Table 1b~~Table 2 (Standard G-1).

Table 28. Age classification of the models per region.

	Pre-1960	1960-1970	1971-1980	1981-1993	1994-2001	2002-pres.
HVHZ	⅔ modified Weak, ⅓ Medium	⅔ Weak, ⅓ Medium	½ Weak, ½ modified Medium	⅔ Weak, ⅓ modified Medium	Modified Strong	Modified Strong
Keys	½ modified Weak, ½ Medium	Medium	Medium	Medium	⅓ Medium ⅔ Strong_OP	Strong_OP
WBDR	modified Weak	⅔ Weak, ⅓ Medium	⅓ Weak, ⅔ Medium	⅓ Weak, ⅔ Medium	½ Medium, ½ Strong_OP	Strong_OP
Inland	modified Weak	⅔ Weak, ⅓ Medium	½ Weak, ½ Medium	½ Weak, ½ Medium	½ Medium, ½ Strong	Strong

Table 28 Nomenclature with respect to Table 1~~Table 1a~~ and ~~Table 1b~~Table 2

Strong:	S00
Strong_OP:	S00-OP
Modified Strong:	S01
Medium:	M00
Modified Medium:	M10
Weak:	W00
Modified Weak:	W10

Note: HVHZ is high velocity hurricane zone; WBDR is wind-borne debris region. The boundaries of the WBDR vary depending on the year built, and the edition of the FBC which applies, as explained in Standard G-1, in the description of the site-built models.

9. ***Describe the development of the vulnerability functions for appurtenant structures.***

Since the appurtenant structures damage is not derived from the building damage, only one vulnerability matrix is developed for appurtenant structures. To model appurtenant structure damage, three equations were developed. Each determines the appurtenant structure insured damage ratio as a function of wind speed. One equation predicts damage for structures highly susceptible to wind damage, the second predicts damage for structures moderately susceptible to wind damage, and the third predicts damage for structures that are affected only slightly by wind. Because a typical insurance portfolio file gives no indication of the type of appurtenant structure covered under a particular policy, a distribution of the three types (slightly vulnerable, moderately vulnerable, and highly vulnerable) must be assumed and is validated against the claims data.

For commercial residential structures, appurtenant structures might include a clubhouse or administration building, which are treated like additional buildings. For other structures such as pools, etc., the appurtenant structures model developed for residential buildings is applicable.

10. ***Describe the relationship between building structure and appurtenant structure vulnerability functions.***

Appurtenant structures are not attached to the dwelling or main residence of the home, but are located on the insured property. These types of structures could include detached garages, guesthouses, pool houses, sheds, gazebos, patio covers, patio decks, swimming pools, spas, etc. Insurance claims data reveal no obvious relationship between building damage and appurtenant structure claims. The variability of the structures covered by an appurtenant structure policy may be responsible for this result. Consequently, building structures and appurtenant structures vulnerability functions were developed independently from each other.

11. ***Describe the assumptions, data, methods, and processes used to develop building vulnerability functions for unknown residential construction types.***

Disclosures 11 and 12 are addressed together below.

12. ***Describe the assumptions, data, methods, and processes used to develop building vulnerability functions when some primary characteristics are unknown.***

The engineering team designed a mapping tool to read a policy and assign building characteristics, if unknown or other, on the basis of building population statistics and year built, where the year built serves as a proxy for the strength of the building. The process is summarized in Table 29. Once all the unknown parameters in the policy have been defined, an unweighted vulnerability matrix based on the corresponding combination of parameters can then be assigned. If the number of unknown parameters exceeds a certain threshold defined by ~~the user, he or she has the choice of using a weighted matrix or age-weighted matrix instead.~~ the actuarial team, a weighted matrix or age-weighted matrix is used instead.

Table 29. Assignment of vulnerability matrix depending on data availability in insurance portfolios.

Data in Insurance Portfolio	Year Built	Exterior Wall	No. of Story	Roof Shape	Roof Cover	Opening Protection	Vulnerability Matrix
Case 1	known	known	known	known	known	known	Use unweighted vulnerability matrix
Case 2	known	known or unknown	Any combination of the four parameters is either unknown or other				use weighted matrix or replace all unknown and others randomly based on stats and use unweighted vulnerability matrix
Case 3	known	other	Any combination of the four parameters is either unknown or other				use the “other” weighted matrix

Data in Insurance Portfolio	Year Built	Exterior Wall	No. of Story	Roof Shape	Roof Cover	Opening Protection	Vulnerability Matrix
Case 4	unknown	known	Any combination of the four parameters is either unknown or other				use age weighted matrix or replace all unknown and others randomly based on stats and use unweighted vulnerability matrix
Case 5	unknown	other	Any combination of the four parameters is either unknown or other				Use age weighted matrices for "other"

13. ***Describe the assumptions, data, methods, and processes used to develop building vulnerability functions for various construction types for renters and condo-unit owners.***

No vulnerability curves were developed specifically for renters and condo-unit owners. In the case of renters, the insurance portfolios never specify if the rental unit is an apartment or a single family home. Therefore, in the absence of that information, the model uses by default personal residential weighted vulnerabilities for contents and ALE. In the case of condo-units, the insurance portfolios never provide information regarding the building within which the unit is located. Therefore, although the FPHLM has a procedure in place to evaluate condo-units losses from within the mid/high-rise commercial residential module, the model uses by default the personal residential weighted vulnerabilities.

14. ***Describe any assumptions, data, methods, and processes used to develop and validate building vulnerability functions concerning insurance company claims.***

The 2004 hurricane season provided a wealth of claims data, used to validate and calibrate the FPHLM (Artiles, 2006; Pinelli et al., 2006). First, the consistency and validity of the data itself was investigated (see standard A-1), and the associated wind speed data was sought from NOAA. The results from the model were then compared to the claims data for hurricanes Charley and Frances. The comparisons were done for the different structural types, for different age categories, and for different insurance companies. They included comparisons of aggregated losses and of vulnerability curves. The comparisons took into account the fact that the actual wind data that caused the damage was not always available, and there were some unknowns regarding the true nature of coverage of many insurance policies. Based on these comparisons, the engineering team recalibrated the engineering model to produce a more accurate and credible predictive capability.

In subsequent years, for every new version of the FPHLM, and as new claims data became available, comparisons of aggregated losses between actual claims data and FPHLM output were performed to validate and calibrate the model. These comparisons are reported in the next disclosure. In all these comparisons, the implicit assumptions are that company claim payment practices including the effects of contractual obligations on the claim payment process are stable over time and do not vary by company.

15. ***Demonstrate that building vulnerability function relationships (building structures and appurtenant structures) are consistent with insurance claims data.***

The building loss consists of external and internal losses. Appurtenant structure losses are derived independently. All the losses are based on a combination of engineering principles, empirical equations, and engineering judgment. They were validated against claims data from several hurricanes. The results are shown in Figure 53 and 54 below. Each dot represents an insurance portfolio.

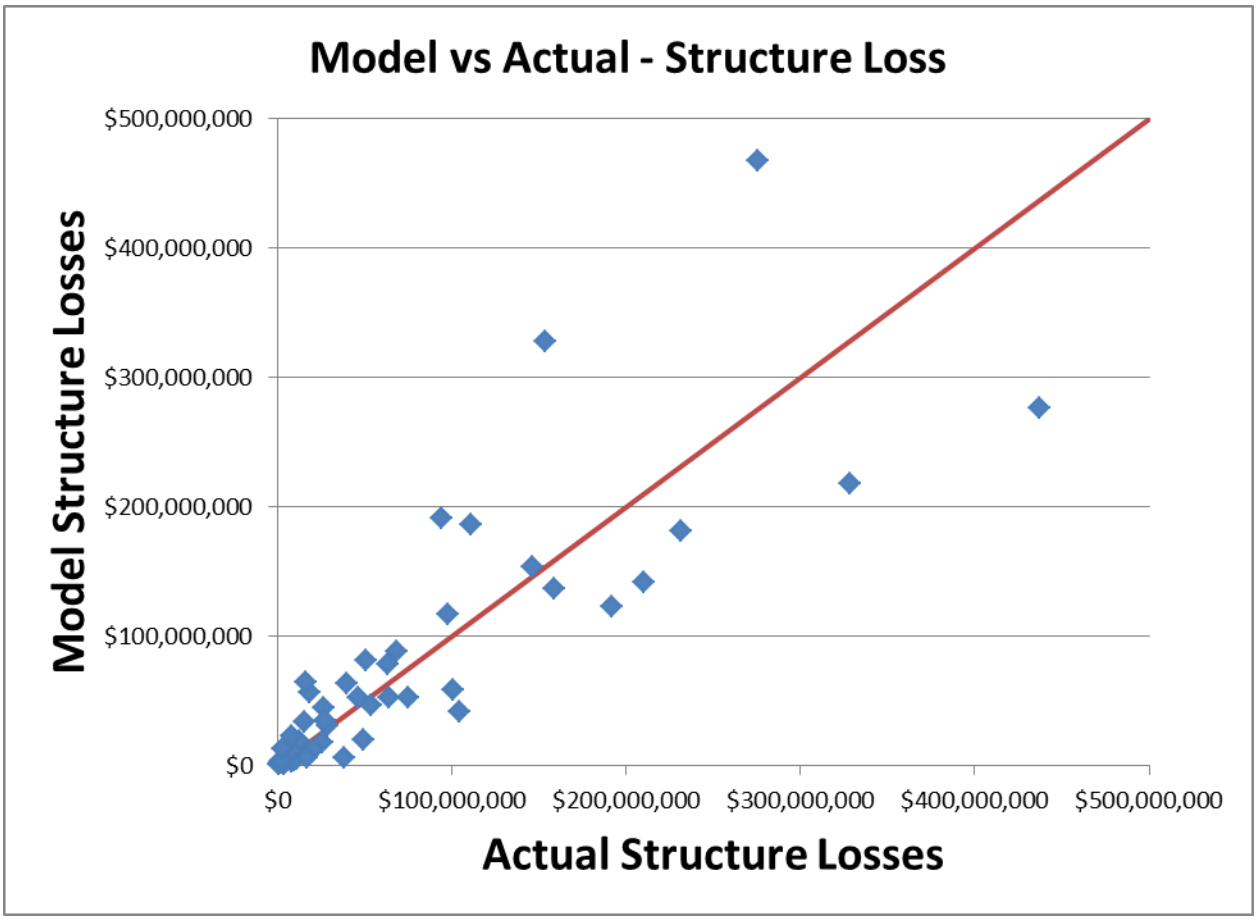


Figure 53. Model vs. Actual-Structural Loss.

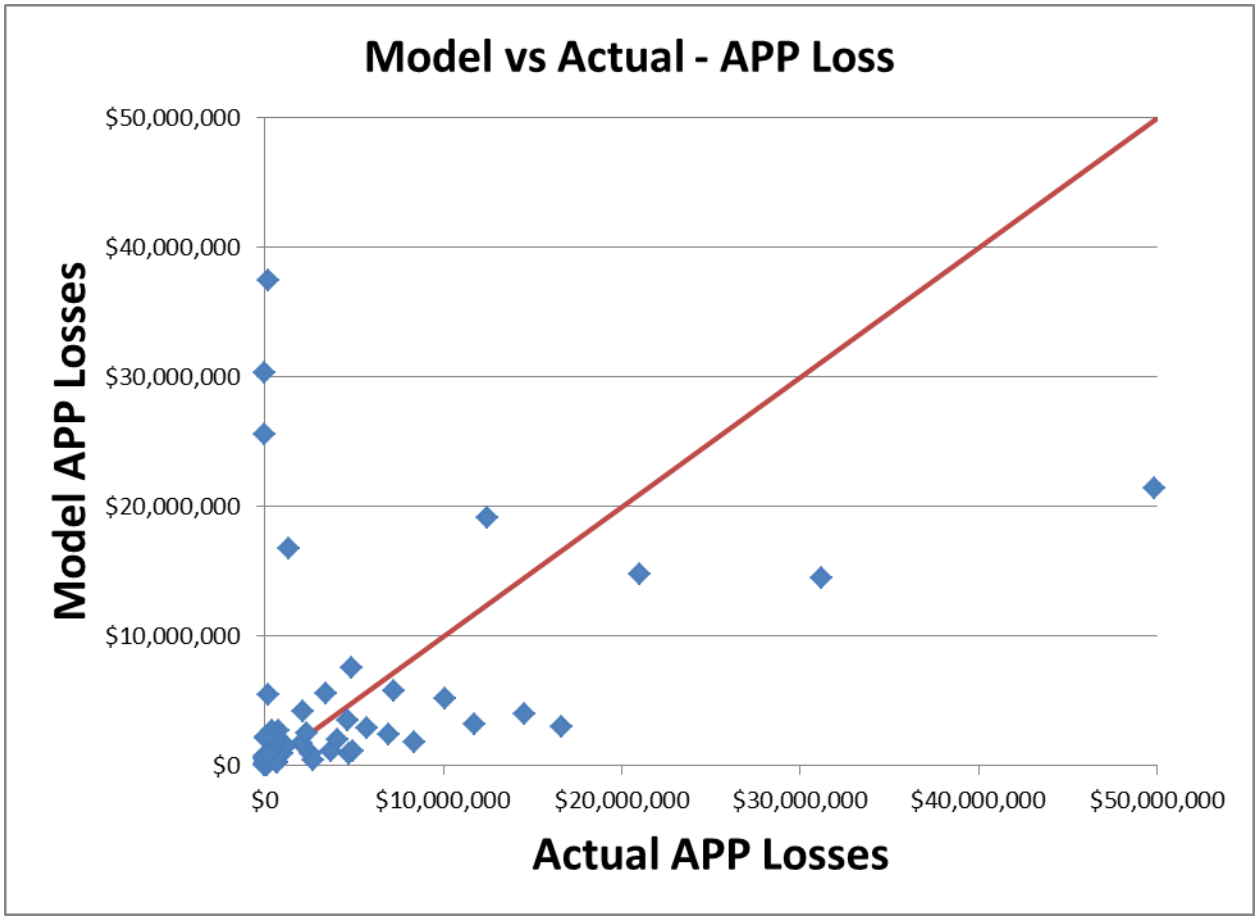


Figure 54. Model vs. Actual-APP Loss.

16. ***Identify the one-minute average sustained windspeed and the windspeed reference height at which the model begins to estimate damage.***

The wind speeds used in the damage model are three-second gusts at 10 m. The lowest three-second gust is 50 mph. The minimum one-minute sustained wind is approximately 40 mph.

17. ***Describe how the duration of windspeeds at a particular location over the life of a hurricane is considered.***

Duration of the storm is not explicitly modeled. The damage accumulation procedures assume sufficient duration of peak loads to account for duration dependent failures.

18. ***Describe how the model addresses wind borne missile impact damage and waterinfiltration.***

Treatment of wind borne missile impact damage

Windborne debris is considered as a source of potential damage to building openings (windows and doors). Based on post-storm damage investigations (e.g. Gurley and Masters, 2011), the model assumes that damaged roof cover from adjacent buildings is the dominant source of windborne debris. The vulnerability of an opening to windborne debris damage is modeled as a

function of the density of the surrounding buildings (e.g. open vs. suburban terrain), wind speed and direction, building age (roof cover strength), height of the opening relative to building height, and opening protection (glass type and / or shutters). If an opening fails as a result of windborne debris impact, the internal pressure and associated building component loads are adjusted and failure checks are repeated. The breached opening is recorded in the damage matrix for use in costing as well as wind driven rain water ingress calculations.

For a given structural type and assigned peak 3-second wind speed (v_{wind}), the probability of damage to an opening ($P_D(v_{wind})$) as:

$$P_D(v_{wind}) = 1 - e^{-N_A * A(v_{wind}) * B(v_{wind}) * C * D(v_{wind})}$$

Where:

- N_A is the total number of available missile objects in the area upwind of the structure being analyzed. For example, the total number of shingles on the neighboring upwind house.
- $A(v_{wind})$ is the fraction of potential missile objects that are in the air at a given 3-second gust wind speed (v_{wind}). For example, the percentage of the shingles on the upwind neighboring roof that were damaged and available for flight.
- $B(v_{wind})$ is probability of the missile hitting the structure. A free shingle upwind of the structure may or may not strike the subject building. A trajectory model is used to determine this parameter.
- C is the fraction of the total area of a particular opening (window, entry door or sliding door) to area of the impact wall in which it exists. If a shingle does strike the building, C is the probability that it struck the subject opening.
- $D(v_{wind})$ is the probability that the impacting missile has enough momentum to damage the component impacted.

Each of the above parameters is considered in more detail below.

N_A is the total number of potential missiles that are upwind of the target structure. It is assumed that surrounding buildings are similar to that of the target building and therefore have approximately the same roof cover. The total number of potential missiles is dependent on the exposure category of the area and the wind direction. The particular exposure category chosen by the user determines the location of the surrounding buildings. There are eight building surrounding the structure in “Urban” and “Suburban” exposures while there are only four buildings cornering the target building in “Open” exposures. Distances from the surrounding buildings to the subject building also changes from urban to suburban to open. N_A is evaluated for each of 8 directions (Figure 55). For wind directions that are perpendicular or parallel to ridgeline of the buildings, it is assumed that N_A is equal to the number of shingles from the adjacent building. For wind directions diagonal to the ridgeline of the building it is assumed that there is full contributions from the building diagonal to ridgeline and a partial contribution from the adjacent structures (25% contribution).

$A(v_{wind})$ is the percentage of the number of potential missiles (N_A) that are assumed to become airborne and become actual missiles in the wind field upwind of the subject building. Roof cover is assumed to become airborne if it is damaged in the wind field. Thus $A(v_{wind})$ is determined by

assuming the neighboring structures are of the same age as the subject with respect to the capacity of the roof cover. The vulnerability of the roof cover at the speed v_{wind} being evaluated is used to populate $A(v_{wind})$. A matrix of mean percent roof cover damage for various roof cover strengths was created and used as the input for the $A(v_{wind})$ variable. The appropriate $A(v_{wind})$ for a given simulation is selected via table lookup and randomized for implementation. In this manner, homes with older and weaker roof cover are assumed to be subjected to a higher $A(v_{wind})$ value than homes with newer and stronger roof cover. This is consistent with post-storm investigation studies that have identified a correlation between roof cover age and vulnerability (e.g. Gurley and Masters, 2011; Liu and Pogorzelski et al., 2010).

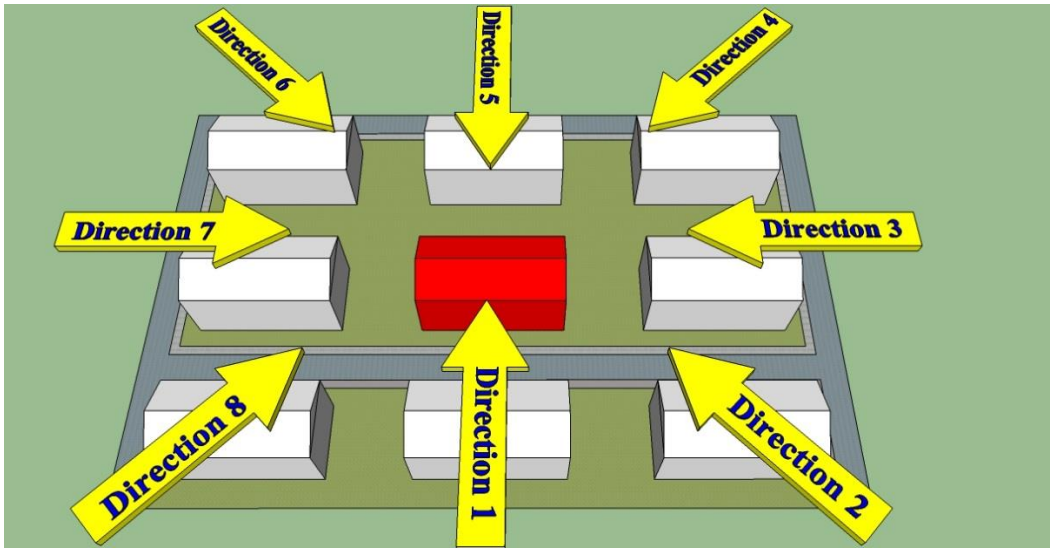


Figure 55. Evaluating NA for eight approach directions.

$B(v_{wind})$ is probability of a airborne missile hitting the subject building. Referring to Figure 55, for a given direction, any airborne shingles that approach the subject building may fall short of, fly over, or strike the building. This is a function of the missile object, distance (sparse or dense neighborhoods), and wind speed and turbulence. A stochastic flight trajectory model (Laboy et al., 2013) is employed in a Monte Carlo framework (100,000 simulations). Inputs to this model include the flight object parameters (e.g. shingles), distance from source to target (dense or sparse neighborhoods), local wind turbulence (suburban or open terrain), and wind speed. A series of curves were developed to determine the mean probability of available debris striking the subject building (stratified by floor) as a function of the above mentioned variables, and are stored in a library to access for a given vulnerability simulation.

C is the fraction of the total area of a particular opening category (window, entry door or sliding door) to area of the impacted wall in which it exists. Now that the probability of a floor being hit has been determined ($B(v_{wind})$), the probability of the debris hitting the opening of interest is assessed. This is the area of the opening divided by the total wall area of the floor. The C value for a 4ft by 4ft window on a wall with dimensions 10ft by 40ft is equal to .04. Based on this value, if a projectile was to strike this wall, there is 4% chance of it hitting the window being evaluated.

D(v_{wind}) is the probability that a window impacted by debris will be damaged. It is a function of the missile object, impact velocity, angle of incidence, and material being impacted. The missile object is roof cover (shingles). The impact velocity and angle of incidence is captured by the flight trajectory model used to determine parameter B. The material being impacted is either standard annealed or impact resistant glass. A recent experimental study evaluated the momentum threshold required for shingles to break unprotected residential window glass. The study concluded that the wind speed necessary to remove and transport shingles a sufficient distance to the target convey sufficient momentum to break annealed glass (Masters et al., 2010). This is incorporated in the current model by assigning a value of 1.0 (100%) to the D parameter. That is, shingles will break standard glass if impact occurs.

Mitigation of damage from debris impact can be achieved via impact resistant glazing products (i.e. impact resistant glass) and / or exterior impact protection (plywood or metal shutters). This is implemented by reducing the probability of missile impact rather than adjusting the impact damage capacity (B is adjusted rather than D). The effect is combinatorial, such that impact resistant glass with shutters is less vulnerable than standard glass with shutters.

The implementation of the above components results in a probability of debris damage value as a function of wind speed, direction, building density / terrain, height of the opening on the building face, and window protection. A random number draw from a uniform distribution then determines the occurrence of damage for each opening on the subject building.

Treatment of water infiltration in the commercial residential model

The modelers developed a novel approach to assess interior damage. The method complements the component approach described above to compute the damage to the building envelope (Weekes et al., 2009). The method is summarized in Figure 56. The model estimates the amount of wind-driven rain that enters through the breaches and defects (also referred to as pre-existing deficiencies) in the building envelope and converts it to interior damage. The approach is described below.

The building components that the model considers for low rise buildings are roof cover, roof sheathing, wall cover, wall sheathing, gable cover, gable sheathing, windows, entry doors and sliding doors. For an initial wind speed, the model starts loading the exterior damage array, expressed as breach areas of each component for thousands of simulation runs. It has been demonstrated that in buildings subjected to hurricane winds, the interior damage may start well before there are any breaches in the envelope (Mullens et al., 2006). The interior damage at this early stage is non-negligible and is caused by the building's existing defects that may be hidden or not, such as cracks, poorly caulked electrical outlets and ventilation ducts, inadequately sealed windows and doors, soffits, baseboards, door thresholds, etc. (Lstiburek, 2005). An estimated area of existing defects or deficiencies in envelope components is accounted for.

The quantification of existing defects is based on the surveys published in Mullens et al. (2006) and the American Society of Heating, Refrigerating and Air-Conditioning (ASHRAE) Handbook (2001) for estimating the infiltration area. To capture the quality of the construction, the model

applies defect densities depending on the building's strength, which is related to the year built. Thus, strong buildings will have fewer defects than medium and weak buildings.

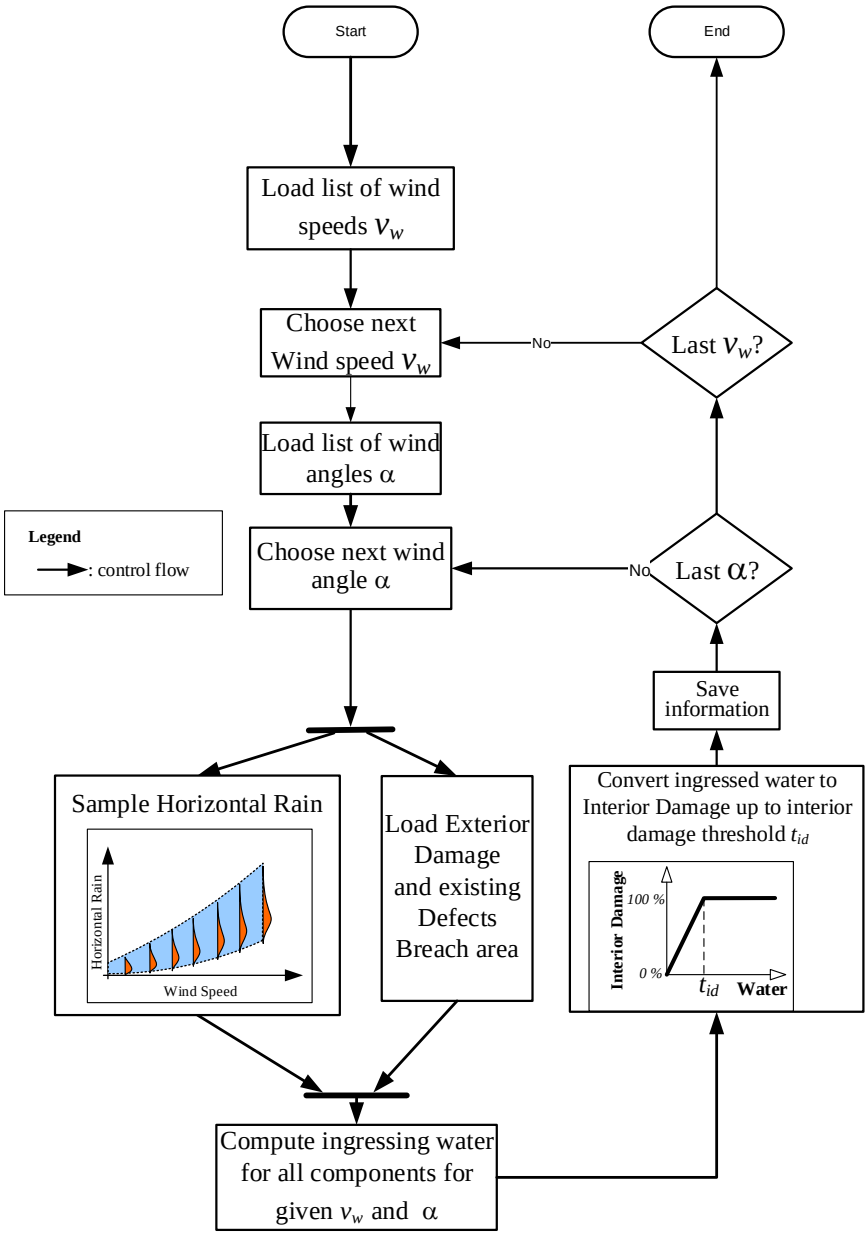


Figure 56. Flowchart of the interior damage model.

Recent studies have shown that water ingress via wind driven rain cannot be attributed exclusively to envelope breach, installation, or product defects. Properly manufactured, installed, and caulked fenestration may nonetheless offer leakage paths in extreme wind conditions, the severity of which is highly dependent on the specific product (Salzano et al., 2010). As this line of research matures, its findings will be incorporated within the above framework.

In order to estimate water intrusion into the buildings, a study was performed to estimate the likely accumulated horizontally impinging rain on a structure during a hurricane event. This study used a simulation model that is composed of a simplified wind model and the R-CLIPER rain rate model developed at NOAA HRD (Lonfat et al., 2007) and is used operationally at NHC. The simplified wind model is based on Holland (1980) and includes parameters for the pressure profile ("B"), Rmax, translation speed and central pressure. Additionally, the Vickery (2005) pressure filling model was used to decay the storms. Storm parameters are sampled from distributions relevant to Florida. The R-CLIPER model determines the vertically free-falling rain rates at each time step of the simulation. The R-CLIPER rain rate is essentially an azimuthally averaged rain rate that varies as a function of radius and maximum intensity of the storm. A detailed presentation of this study is given in Pita et al. (2012a) and Pita (2012).

The study simulates the duration of the event from the time a location enters the storm affected area (within 450 km of the storm center) until exit. The number of storm simulations was 100,000 and for each simulation, 91 locations were selected to record the accumulated horizontal rain ("*HR*") and maximum three-second wind gust at 10 m. Each location was specified to be a multiple of 10 km away from the storm closest approach to center (from 450 km to the left of the storm to 450 km to the right of the storm, in steps of 10 km. A direct hit is at 0 km). The time step of the model was 0.1 hr. In addition to the total horizontal rain during the event, separate accumulations were recorded starting at the time that a location experiences the peak wind of the storm event ("HR_2"). The horizontal rain accumulated prior to the maximum peak gust ("HR_1") is computed as the difference: $HR_1=HR-HR_2$. The resulting accumulations are then distributions of horizontal rain as a function of the peak three-second wind gust for 10 meter height.

Since *HR1* and *HR2* are not uniformly distributed through time (with higher concentration around the max wind speed), not all surfaces of a building will be subject to equal shares of horizontal rain as the storm rotates around the building. To account for this, we developed a directionality scheme where, during the rain simulation process, we record and calculate the *HR1* and *HR2* values while the wind direction falls into successive 45° octants.

The distribution of the horizontal rain at a particular location as a function of time is illustrated in Figure 57. α_m is the fraction of *HR1* (i.e. the fraction of the area under the curve) while the wind direction is in a particular octant "m" (where m = 1, 2 … i represents the possible total number of changes in the wind direction prior to the occurrence of max wind speed). Similarly, β_n represents the fraction of *HR2* while the wind direction is in a particular octant "n" (where n= 1,2,3…*j* represents the possible total number of changes in the wind direction after the occurrence of max wind speed). The vulnerability model assumes the peak wind to occur at the center angle of the sector or octant (at time t_{wmax} in Figure 57). For the sake of consistency with the damage model, in the rain study, the sectors are defined so that the peak wind occurs at the center of the sector which contains the max wind.

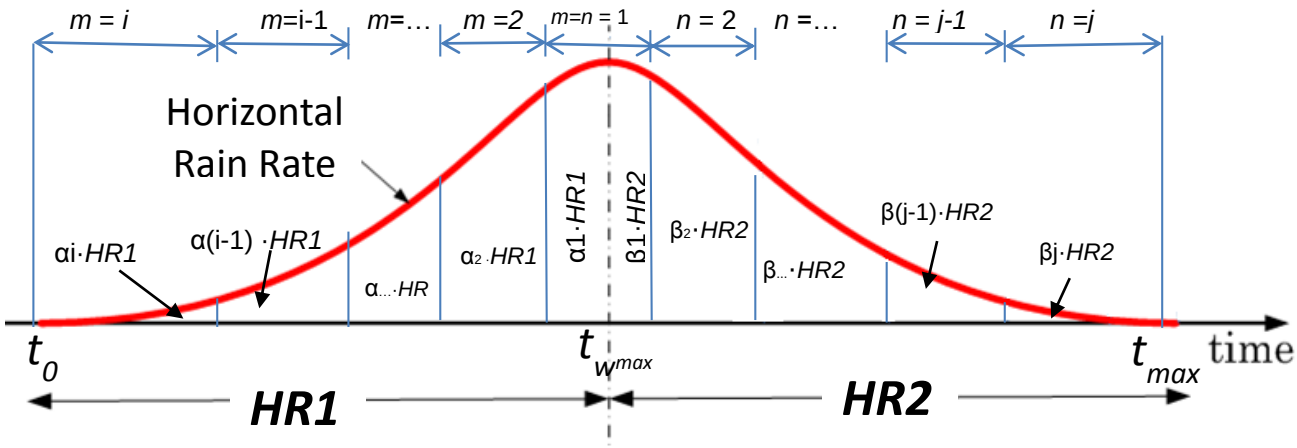


Figure 57. Horizontal rain rate as a function of storm duration.

The overall volume of free stream horizontal rain (*HR*) expected at a particular location can be reduced to the following equation:

$$HR = \sum_{m=1}^{i} \alpha_m * HR_1 + \sum_{n=1}^{j} \beta_n * HR_2$$

Where α_m is the fraction of *HR1* for a given wind direction octant and *i* is the total number of wind direction changes between the initial start of the storm (t_0) and the time of max wind speed (t_{wmax}). Consequently, $\sum_{m=1}^{i} \alpha_m = 1$ and m = 1 represents the wind direction octant at t_{wmax}, and m=i represents the wind direction at the beginning of the storm, t_0. If i=1 it means that the wind has blown in the same octant from t_0 to t_{wmax} .

Similarly, β_n is the fraction of *HR2* for a given wind direction octant and *j* is the total number of wind direction changes from the time of max wind speed to the end of the storm. Consequently, $\sum_{n=1}^{j} \beta_n = 1$ and n = 1 represents the wind direction at the time of maximum wind velocity (t_{wmax}), while n= j represents the wind direction at the end of the storm t_{max}.

Water intrusion model for low-rise CR buildings.

The FPHLM interior damage model performs Monte Carlo simulations to estimate the total volume of water that penetrates through a building envelope on a component by component basis, through either defects in the component or breaches. Each simulation corresponds to a given wind direction octant (from 0° to 315° in 45° increments) and a given maximum wind speed (from 50 to 250 mph, in 5 mph increments). Each component is evaluated for both the directly impinging and the surface runoff rain. The total volume of water $V_{to\ \ ci}$ for each component C_i can therefore be expressed by the general equation.

$$V_{tot_{Ci}} = V_{IR_{Ci}} + V_{SR_{Ci}} = RAF \cdot HR \cdot A_{o_{Ci}} + SRC \cdot HR \cdot A_{SR_i}$$

Where:

- $V_{IR_{Ci}}$ is the volume of horizontal impinging water penetrating through the component C_i
- $V_{SR_{Ci}}$ is the volume of surface run-off water penetrating through the component C_i
- RAF is the rain admittance factor, which transforms the horizontal rain in impinging rain
- SRC is the surface runoff coefficient, which transforms the horizontal rain in surface run-off
- A_{oCi} is the open area of the component C_i, either through defect and/or breach
- A_{SRCi} is the reference surface runoff area or upstream area of the defect or breach collecting water, for component C_i, which is a function of the wind direction;
- *HR* is the horizontal rain, either *HR1* or *HR2* (before or after the occurrence of the maximum wind speed), sampled for each maximum wind speed from the full distribution of horizontal rain from the simulation.

The rain admittance factor (*RAF*) is the fraction of the approaching horizontal rain that strikes the building. It accounts for the effect of a large portion of the rain moving around the structure with the wind rather than striking the building surface and is dependent on the building shape. Both RAF and SRC are independent of the wind speed, but both are a function of the wind direction with respect to the building. The values of RAF and SRC are the result of an extensive testing program carried on at the Wall of Wind at FIU (Baheru et al., 2014a, 2014b).

For any given simulation, the link between the rain study and the vulnerability model is the maximum wind speed w_{max}. As the storm rotates before and after the occurrence of the maximum wind speed, it subjects any given defect or breach on a particular surface to all the fractions of impinging rain corresponding to the different wind directions (or octants) from the storm rotation.

Consequently, before t_{wmax} (i.e. before the occurrence of w_{max} and the occurrence of any breach in the model for that simulation), the total value of impinging rain penetrating through a component defect area A_{d_Ci} is the sum of the corresponding fractions of impinging rain over the wind direction octants θ_m, as the storm rotates from its start to t_{wmax}.

$$V_{IR1_{Ci}} = \left[\sum_{m=1}^{4} RAF_{\theta m} * \overline{a_m}(w_{max})\right] * HR_1 * A_{d_Ci}$$

Where:
$\overline{a_m}(w_{max})$ is the mean fraction of HR_1 for the the wind direction octants θ_m. It is a function of w_{max}.
$RAF_{\theta m}$ is the rain admittance factor for the the wind direction octant θ_m, which transforms the free field horizonal rain into impinging rain.

Simlilarly, the total value of surface run-off water penetrating through a defect is the sum of the corresponding fractions of surface run-off water over the wind direction octants θ_m, as the storm rotates from its start to t_{wmax}. The total quantity *HR1* can be factored out of the sumation, since it is independent of the angle.

$$V_{SR1_{Ci}} = \left[\sum_{m=1}^{4} SRC_{\theta m} * \overline{a_m}(V_{max}) * A_{SR_{Ci_{\theta m}}}\right] * HR_1$$

Where:
$SRC_{\theta m}$ is the surface run-off coefficient for a wind direction octant θ_m, which transforms the free field horizonal rain into run-off water.

For each damage simulation, θ_1 is the wind direction or octant at t_{wmax}, θ_2 is the previous octant in the rotation (45 degrees), and so on.

After t_{wmax} (i.e. after the occurrence of w_{max} and the occurrence of some breaches in the model for that simulation), the total amount of impinging rain penetrating through the breach and the remaining defects of componnet C_i is the sum of the corresponding fractions of impinging rain over the wind direction octants θ_n, as the storm rotates from t_{wmax} to its end.

$$V_{IR2_{Ci}} = \left[\sum_{n=1}^{5} RAF_{\theta n} * \overline{\beta_n}(w_{max})\right] * HR_2 * A_{oCi}$$

Where:
$\overline{\beta_n}(w_{max})$ is the mean fraction of *HR2* for the the wind direction octants θ_n. It is a function of w_{max}. $RAF_{\theta n}$ is the RAF value for a wind direction octant θ_n.

Similarly, the total value of surface run-off penetrating through a component breach and its remaining defects is the sum of the corresponding fractions of surface run-off water over the wind direction octants θ_n, as the storm rotates from t_{wmax} to its end. The total quantity *HR2* can be factored out of the sumation, since it is independent of the angle.

$$V_{SR2_{Ci}} = \left[\sum_{n=1}^{5} SRC_{\theta n} * \overline{\beta_n(w_{max})} * A_{SR_{Ci_{\theta n}}}\right] * HR_2$$

Where $SRC_{\theta n}$ is the SRC value for a wind direction octant θ_n. For each damage simulation, θ_1 is the wind direction or octant at t_{wmax}, θ_2 is the next octant in the rotation (45 degrees), and so on.

Over the entire duration of the storm, the total amount of water penetrating through a component will be:

$$V_{tot_{Ci}} = V_{IR_{Ci}} + V_{SR_{Ci}} = V_{IR1_{Ci}} + V_{SR1_{Ci}} + V_{IR2_{Ci}} + V_{SR2_{Ci}}$$

The volume of water in the equation above can be transformed in heights of water at each story by dividing by the floor area of the story A_b.

$$h_{tot_{Ci}} = \frac{V_{tot_{Ci}}}{A_b}$$

Water intrusion model for mid/high-rise CR buildings.

There is no data available on RAF and SRC for mid/high-rise buildings at this point. Therefore the water intrusion model has not changed and is the same as the previous version 5.0 of the FPHLM. The product of the areas of the breaches and defects by the impinging rain conveys the amount of water that enters the building. The water penetration at each story is computed as follows.

Water penetration through components defects or pre-existing deficiencies:

$$h_{C_i}^{d} = \frac{f_{sim} \cdot f_{Run} \cdot RAF\left[HR_1 \underbrace{\left(d_{C_i} A_{C_i}\right)}_{\text{Total Defects Area}} + HR_2 \underbrace{\left(d_{C_i} A_{C_i} S_{C_i}\right)}_{\text{Post-breach Defects Area}} \right]}{A_b}$$

Water penetration through breaches:

$$h_{C_i}^{b} = \frac{f_{sim} \cdot f_{Run} \cdot RAF\left[HR_2 \cdot A_{C_i}^{B}\right]}{A_b}$$

Where:

- h^d_{Ci}: height of water that accumulates due to defects in component *i*, in inches
- h^b_{Ci}: height of water that accumulates due to envelope breaches in component *i*, in inches
- f_{sim}: adjustment factor which takes into account that defects and breaches will progressively change from windward to leeward or vice-versa as the storm rotates
- f_{Run}: adjustment factor for the water that runs-off the external surfaces of the building and ingress through the defects and breaches and into the building
- RAF: rain admittance factor
- d_{Ci}: defects percentage
- A_{Ci}: area of component *i*
- A^B_{Ci}: breach area of component *i*
- A_b: floor area
- HR_1 : mean value of the accumulated horizontal rain prior to maximum wind speed
- HR_2 : mean value of the accumulated horizontal rain after the occurrence of maximum wind speed
- S_{Ci} : survival factor for component $i = 1 - A^B_{Ci} / A_{Ci}$

Rain admittance factor, RAF.

Straube and Burnett (2000) and Blocken and Carmeliet (2010) suggest values for RAF between 0.5 and 1.0 for mid-/high-rise buildings. Accordingly, the FPHLM adopted a value of 0.6 for mid/high-rise buildings, except for the last story where a value of 1.0 was adopted.

Water percolation for both LR and MHR CR

In multi-story low-rise buildings, a portion of the ingressed water percolates downward from story to story. The interior damage model assumes the percolation ρ to be 12% of the ingressed water at each story for low rise building (plywood floors) and 10 % for mid/high rise building (concrete slabs). These values of percolation are based on engineering judgment, supported by calibration of the model with the insurance claims data, and thus can be updated when new research becomes available.

Figure 58 illustrates the percolation mechanism for water ingressing at a given story from pre-existing deficiencies and breaches in any component C_i. Upper story "*j*" gets rain from the pre-existing deficiencies and the breached openings, which is converted into the heights of ingressed water, $h^d_{C_j}$ and $h^b_{C_j}$, respectively. A fraction of these water heights percolates down as $\rho h^d_{C_j}$ and $\rho h^b_{C_j}$. Rain also enters in the second story "*k*" through pre-existing deficiencies and the openings as $h^d_{C_k}$ and $h^b_{C_k}$, respectively.

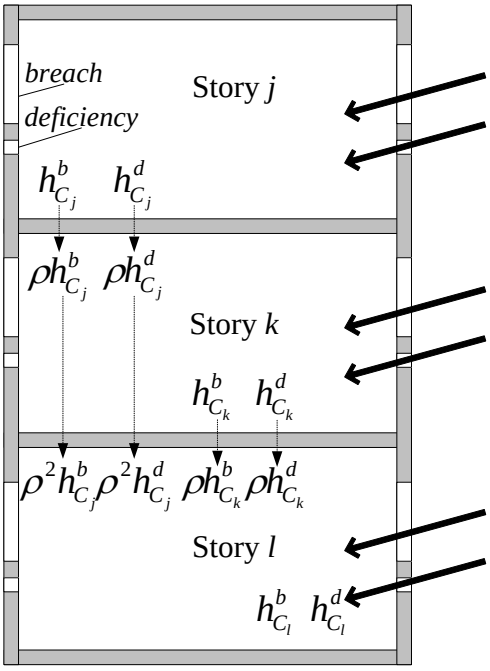


Figure 58. Diagram of water intrusion through breaches, deficiencies and percolation in a 3-story building.

The total amount of water in story *k* of Figure 58 is:

$$h_k = \sum_C \left[\rho \left(h^b_{C_j} + h^d_{C_j} \right) + \left(h^b_{C_k} + h^d_{C_k} \right) \right]$$

Likewise, the total water height at the first story "*l*" of a 3-story building is:

$$h_l = \sum_C \left[\rho^2 \left(h^b_{C_j} + h^d_{C_j} \right) + \rho \left(h^b_{C_k} + h^d_{C_k} \right) + \left(h^b_{C_l} + h^d_{C_l} \right) \right]$$

Thus in 2-story and 3-story buildings, the first story gets the percolated water from the second story by adding a $\rho h^d_{C_2}$ or $\rho h^b_{C_2}$ to the water coming from deficiencies and breaches respectively. The amount of water percolating downward is not subtracted from the total amount of water at the

story where it originated. It is assumed that even if water percolates downward, it still has the potential to produce damage before leaking downward.

In conclusion, these approaches for LR and MHR CR estimate the amount of water that enters through each component of the envelope. The total amount of water is calculated by adding the contributions of all components for a given wind speed, including percolation. The final step maps water inside the building to interior damage with a bilinear relationship, where total interior damage is achieved for a certain threshold of height of accumulated water (currently set at 1 inch).

Treatment of water infiltration in the personal residential model

The overall building damage is the sum of external damage plus interior damage plus utilities damage. In the PR model, the interior damage is extrapolated from the external damage, and the utilities damage is proportional to the interior damage, based on heuristics derived from engineering judgment validated with claims data. This model implicitly includes water infiltration at moderate to high wind speeds.

In damage surveys of past hurricanes (Gurley, 2006), it was observed that a number of houses that were not damaged on the outside did experience losses from water penetration. The heuristic interior damage model was adjusted to address these observations. In order to model rain induced damage, even in the absence of external damage at low wind speeds, a leak internal damage model was developed, which is independent of external damage at low wind speeds, while at higher wind speeds, the relationship between internal and external damage was maintained.

The leak model creates a smooth transition between interior damage at low wind speed (governed by leaks) and interior damage at high wind speed (governed by water penetration through breaches) by means of a polynomial equation coupled with an exponential decay function. The shape of the polynomial model was defined based on engineering judgment and calibrated and validated based on damage observed during the 2004 hurricane season, and the corresponding claims data (Artiles, 2006; Johnson, 2011). The model was first implemented in V3.1 of the FPHLM.

19. ***Provide a completed Form V-1, One Hypothetical Event. Provide a link to the location of the form here.***

See Form V-1.

The model computes the damage based on actual terrain three-second gust winds at 10 m, that are obtained from the given open terrain one-minute sustained winds, and the losses are aggregated twice: once among the ZIP Codes with the same actual terrain three-second gust wind and once among the ZIP Codes with the same open terrain one-minute sustained wind. Because all the ZIP Codes do not have the same roughness, identical open terrain one-minute sustained winds result in different actual terrain three-second gust winds. Occasional bumps in the one-minute sustained winds plot are due to this process of conversion and re-aggregation. The modelers do confirm that the structures used in completing the form are identical to those in the table provided in the Standard.

The insured value for the condo association of the 20 story concrete structure with 8 apartments per story was changed from $100,000 to $15,000,000 since this is a more realistic insured value for a condo association for a building of these characteristics. The change was necessary since the value of the external damage in the model is computed on the basis of the actual replacement value of the damage openings. The actual value of these repairs can be disproportionally high if compared to an arbitrarily low and unrealistic insured value. The adjustment in the insured value of the 20 story concrete structure then provides more realistic damage ratios. The resulting large discrepancies in damage ratios vs. wind speed between the personal residential reference structures in Form V-1 (i.e. timber, masonry, and manufactured home) and the engineered commercial residential reference structure are due to the fact that they correspond to widely different types of structures. Therefore, it is informative to report them separately, which is done in the last two tables of Part A of the form.

Finally, it must be noted that the values of aggregated damage over exposure in Part B are more than three times larger than in the previous submission. This is due to the fact that in this new submission, the number of ZIP codes in the input file was reduced drastically (from 335 to 96). Since the wind speeds of the eliminated ZIP codes were less than 50 mph, their loss was always zero, but their exposure influenced the overall loss cost ratio. Not having those policies included for this submission drastically reduces the overall exposure and therefore raises the overall loss cost. In other words, we are now dividing a similar aggregated damage by a much smaller exposure.

V-2 Derivation of Contents and Time Element Vulnerability Functions

A. ***Development of the contents and time element vulnerability functions shall be based on at least one of the following: (1) historical data, (2) tests, (3) rational structural analysis, and (4) site inspections. Any development of the contents and time element vulnerability functions based on rational structural analysis, site inspections, and tests shall be supported by historical data.***

The development of the vulnerabilities is based on a component approach that combines engineering modeling, simulations with engineering judgment, and observed (historical) data. The content and time element vulnerabilities are extrapolated from the building damage on the basis of expert opinion and site inspections of areas impacted by recent hurricanes and are confirmed using historical claims data.

B. ***The relationship between the modeled building and contents vulnerability functions and historical building and contents losses shall be reasonable.***

The relationship between the modeled structure and the contents vulnerability functions is reasonable, on the basis of the relationship between historical structure and contents losses.

C. ***Time element vulnerability function derivations shall consider the estimated time required to repair or replace the property.***

Time element vulnerability function derivations consider the estimated time required to repair or replace the property.

D. ***The relationship between the modeled building and time element vulnerability functions and historical building and time element losses shall be reasonable***.

For Personal Residential risks the model uses time element vulnerability functions derived from the relationship between building damage and additional living expense. The vulnerability functions have been calibrated using historical claims data on building and additional living expense.

For Commercial Residential risks the relationship between modeled structure and time element loss costs is reasonable. Since no historical loss data were available for calibration, the relationship combines engineering and actuarial judgment.

E. ***Time element vulnerability functions used by the model shall include time element coverage claims associated with wind, flood, and storm surge damage to the infrastructure caused by a hurricane.***

The time element vulnerability functions produced by the model consider time element claims arising from wind, flood, and storm surge damage to the infrastructure. The model does not distinguish explicitly between direct and indirect loss. For Personal Residential risks the time element vulnerability functions were calibrated against claims data that include both types of

losses. For Commercial Residential risks the recognition of claims due to indirect loss is based on judgment since no historical loss data were available for calibration.

Disclosures

1. ***Describe any modifications to the contents and time element vulnerability component in the model since the previously accepted model.***

- No change to report for Personal Residential home owners.
- For Commercial Residential, the contents damage is now assumed to be equal to 100% of the interior damage, instead of 50%.
- For condo unit owners and apartment unit renters, the contents damage is now assumed to be equal to 100% of the interior damage, instead of 50%; and, the time element vulnerability functions are assumed to be the same than for low-rise commercial residential, instead of 50% of the interior damage.

2. ***Describe Provide a flow chart documenting the process by which the contents vulnerability functions are derived and implemented.***

Personal Residential model

Contents include anything in the home that is not attached to the structure itself. Like the interior and utilities, the contents of the home are not modeled in the exterior damage Monte Carlo simulations. Contents damage is modeled as a function of the interior damage caused by each exterior component failure that causes a breach of the building envelope. The function is based on engineering judgment and validated using claims data. The resulting computation of contents vulnerability functions is a 3 stage process as described in Figure 59, and discussed in disclosure 3 below.

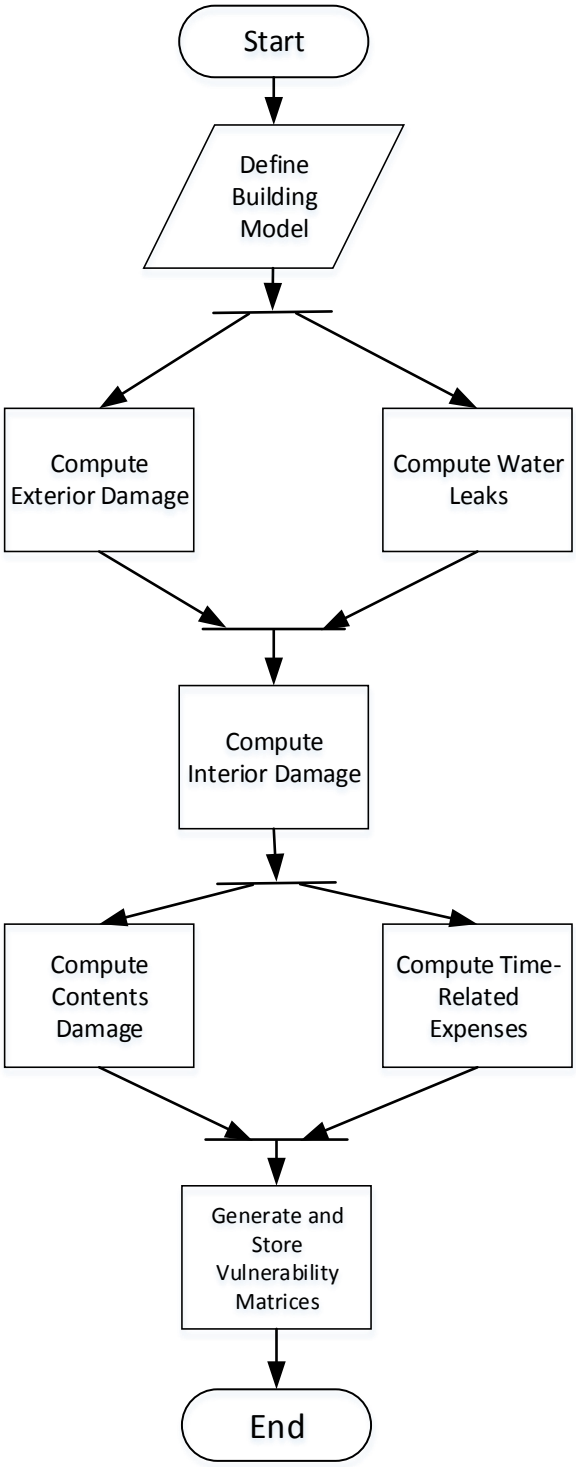


Figure 59~~a.~~. Derivation of contents and additional living expenses vulnerabilities for PR.

Commercial Residential model

The contents vulnerability functions for commercial residential structures are derived from the interior vulnerabilities (which are described in disclosure 18 of standard V-1), and it is

represented by ~~Figure 69b~~**Error! Reference source not found.**~~Figure 60~~ below. In other words, the contents vulnerability functions are set to be proportional to the interior vulnerabilities.

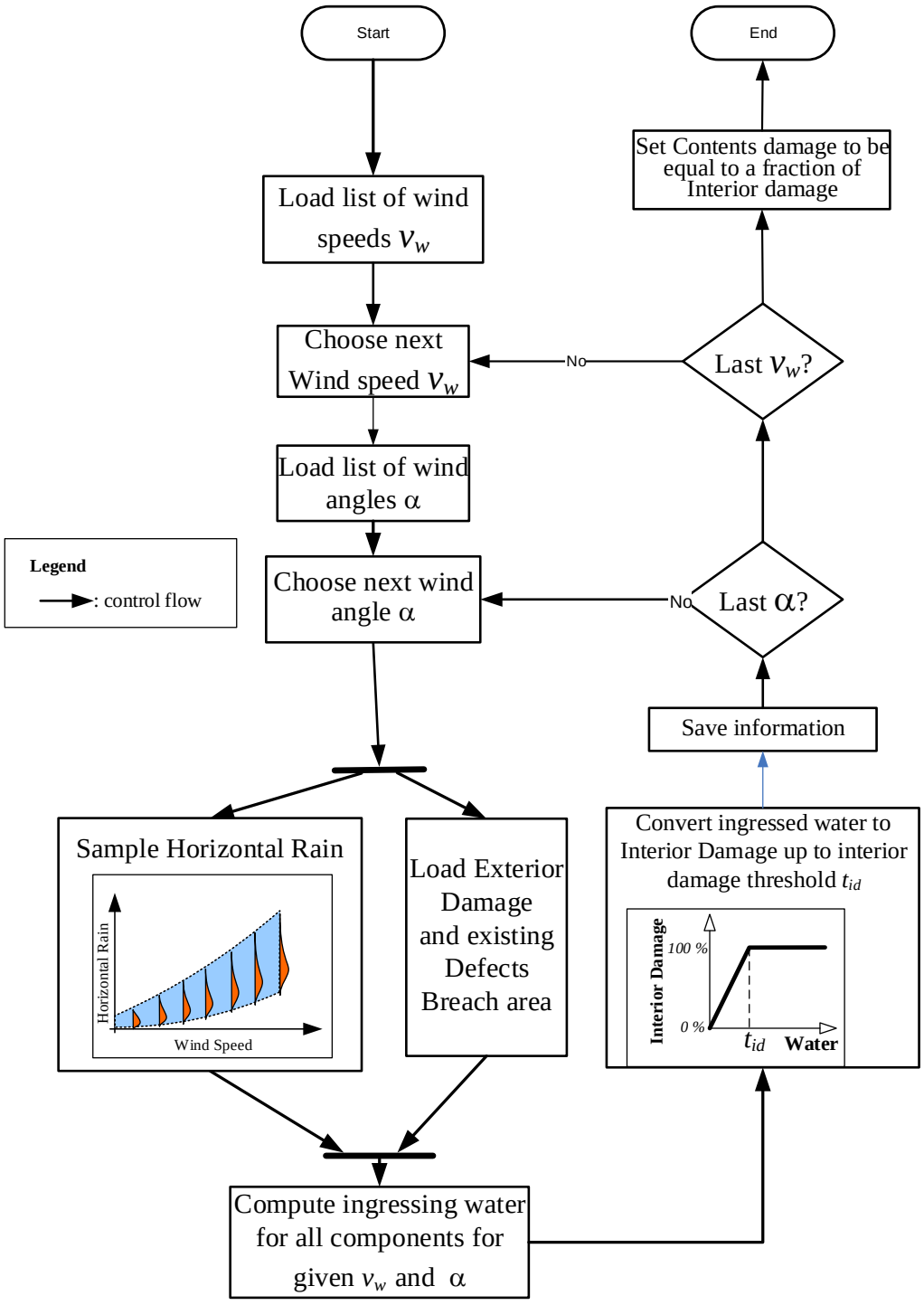


Figure ~~69b~~60. Derivation of contents vulnerabilities for CR.

3. ***Describe the data and methods used to develop vulnerability functions for contents coverage associated with personal and commercial residential buildings.***

Personal Residential model

For each building model, the first stage in the development of contents vulnerability functions corresponds to the external damage assessment through Monte Carlo simulations as discussed in standards G-1 and V-1. In the personal residential model, this is complemented by an empirical estimate of water penetration from wind driven rain due to exterior breaches or leakage paths in undamaged structures (see disclosure 18 of standard V-1). The second stage corresponds to the computation of internal damage. Damage to the interior occurs when the building envelope is breached, allowing wind and rain to ingress. Damage to roof sheathing, roof cover, walls, windows, doors, and gable ends present the possible sources of water ingress. Interior damage equations are derived as heuristic functions of each of these components failure. These relationships are developed primarily on the basis of experience and engineering judgment. Observations of homes damaged during the 2004 hurricane season (Gurley, 2006) helped to validate the predictions. The third stage in the damage estimation (Figure 59) extrapolates the damage to contents from the interior damage, based on a heuristic function. This empirical function is based on engineering judgment and was validated against claims data for Hurricanes Andrew, Charley, and Frances, among others.

Commercial Residential model

Contents damage is assumed to be proportional to interior damage. Therefore, the methods used to develop vulnerability functions for contents coverage associated with commercial residential structures are the same as the methods used for interior damage vulnerability functions. The contents damage is determined by vulnerability functions which correspond to different combinations of wall type (frame or masonry), sub-region (high velocity hurricane zone, wind-borne debris region, inland), roof shape (gable or hip), roof cover (metal, tile or shingle), window protection (shuttered or not shuttered), number of stories (one, two, or three), and strength (weak, medium, or strong).

Based on engineering judgment, contents damage ratio in mid/high-rise buildings (more than three stories) is also estimated to be proportional to the total estimated interior damage ratio for the building.

4. ***Describe the number of contents vulnerability functions and whether different contents vulnerability relationships are used for personal residential, commercial residential, mobile home, condo unit owners, apartment renter unit location, and other similar building classes for wind related damage.***

Contents vulnerability functions were derived for manufactured and site-built homes, and for low-rise commercial residential buildings (one to three stories).

A total of 4356 un-weighted contents vulnerability matrices were developed for site-built homes. The matrices correspond to different combinations of wall type (frame or masonry), region (north, central, south), subregion (high velocity hurricane zone, wind-borne debris region, inland), roof type (gable or hip), roof cover (metal, tile or shingle), window protection (shuttered or not shuttered), number of stories (one or two), and strength (weak, modified weak, retrofitted weak;

medium, modified medium, retrofitted medium; strong for inland and WBDR, strong for HVHZ—see Table 1~~Table 1a~~ and ~~Table 1b~~Table 2 in the General Standards).

These 4356 contents un-weighted matrices were then combined to produce 5226 contents weighted matrices, and 291 contents age weighted matrices for site-built homes for building, for each county. Many of the matrices are repeated because many of the counties use the same regional statistics for the weighting.

A total of 648 un-weighted contents vulnerability matrices were developed for low-rise, commercial residential buildings for building. They correspond to different combinations of wall type (frame or masonry), sub-region (high velocity hurricane zone, wind-borne debris region, inland), roof shape (gable or hip), roof cover (metal, tile or shingle), window protection (shuttered or not shuttered), number of stories (one, two, or three), and strength (weak, medium, or strong).

These 648 matrices were then combined to produce 144 contents weighted curves for low-rise, commercial residential buildings for building.

4 un-weighted contents vulnerability matrices were developed for manufactured homes for building. They correspond to four manufactured home types: (1) pre-1994—fully tied down, (2) pre-1994—not tied down, (3) post-1994—Housing and Urban Development (HUD) Zone II, and (4) post-1994—HUD Zone III. The partially tied-down homes are assumed to have a vulnerability that is an average of the vulnerabilities of fully tied-down and not tied-down homes. The un-weighted matrices are combined into 6 weighted matrices for building, for pre-1994 (4 regions: North, Central, South, Key) and post-1994 (2 zones: II and III) manufactured homes.

The contents vulnerability functions used for condo unit owners and apartment unit renters are the contents vulnerability functions for personal residential buildings, as explained in disclosure 13 of standard V-1.

5. ***Provide a flow chart documenting the process by which the time element vulnerability functions are derived and implemented.***

Personal residential model

Additional living expenses are assumed to be a function of the interior damage caused by each exterior component failure that causes a breach of the building envelope. The function is based on engineering judgment and validated using claims data. The resulting computation of additional living expenses vulnerability functions is a 3 stage process as described in Figure 59 of disclosure 1, and discussed in disclosure 4 below.

Commercial Residential

The process by which the time element expenses vulnerability functions are derived and implemented for commercial residential structures is similar to the process for interior damage already described in disclosure 18 of standard V-1, and is represented in Figure 61.

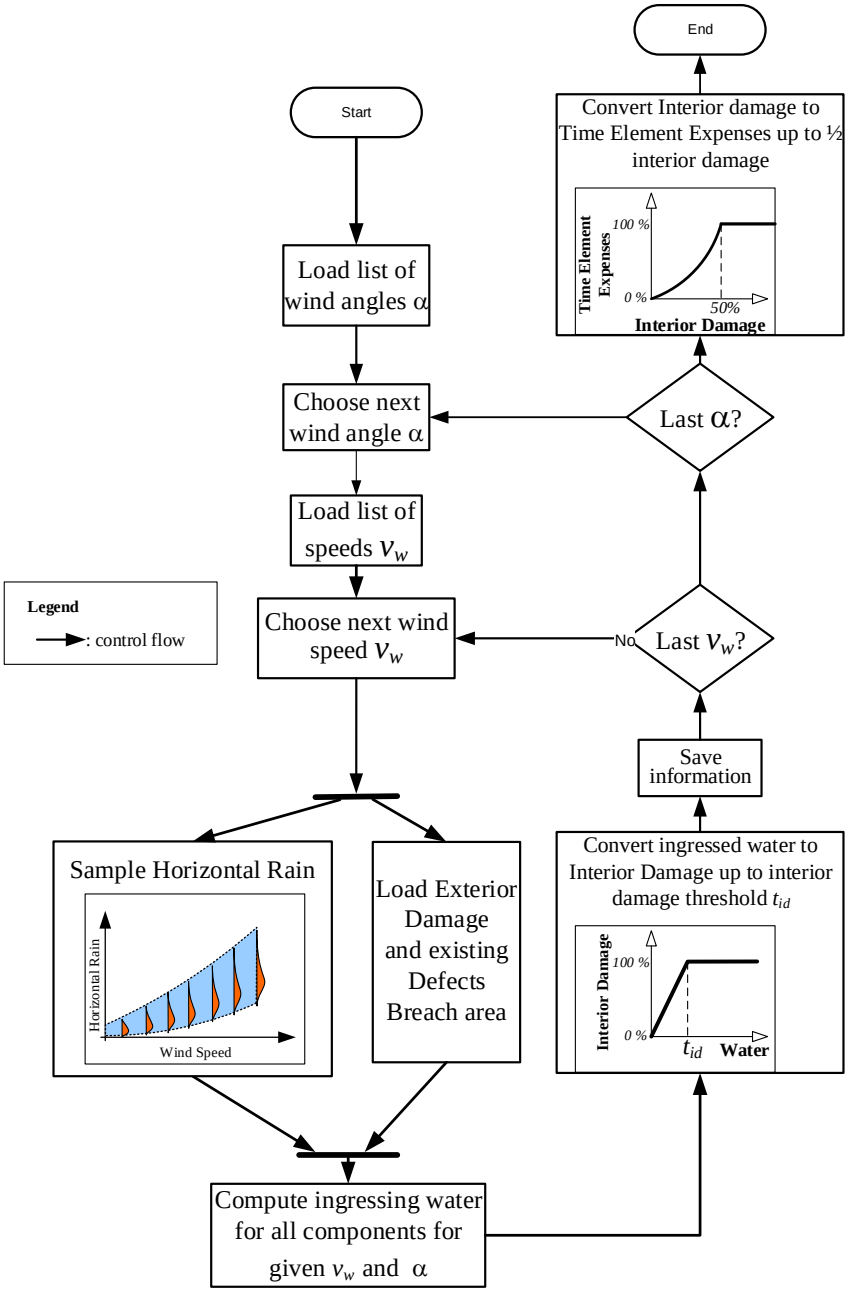


Figure 61. Derivation of time related expenses vulnerabilities for CR.

6. ***Describe the data and methods used to develop vulnerability functions for time element coverage associated with personal and commercial residential buildings. State whether the model considers both direct and indirect loss to the insured property. For example, direct loss could be for expenses paid to house policyholders in an apartment while their home is being repaired. Indirect loss could be for expenses incurred for loss of power (e.g., food spoilage).***

Personal Residential

Additional Living Expense (ALE) is coverage for expenses that arise when an individual must live away from the damaged home. ALE coverage comprises expenses actually paid by the insured. This coverage does not pay all living expenses, only the increase in living expense that results from the covered damage. The value of an ALE claim is dependent on the time needed to repair a damaged home as well as the utilities and infrastructure. Time element or Additional Living Expenses (ALE) are modeled as a function of interior damage. The equations and methods used for manufactured and residential homes are identical. However, it seems logical to reduce the manufactured home ALE predictions because typically a faster repair or replacement time may be expected for these home types. Therefore, an ALE multiplier factor of 0.75 was introduced into the manufactured home model.

Commercial Residential

Owners of apartment buildings may purchase Time Element coverage in addition to wind coverage on the structure and contents. For commercial properties Time Element is an optional coverage and is therefore not purchased by all insured. It is generally a relatively expensive coverage. Some insurance carriers may not even offer Time Element coverage on commercial properties. The coverage will reimburse the owner of the building for business income lost or extra expenses incurred after a hurricane. Both "business income" and "extra expense" are subject to specific definitions and limitations within the coverage form.

We estimate Time Element (TE) losses as a heuristic function of interior damage (ID) as follows:

$$TE = 2ID^2 + ID$$

We do not allocate any portion of the structure deductible to the Time Element loss. We are assuming that Time Element Limits will be exhausted once interior damage reaches approximately 50%. From an underwriting perspective, it is necessary to restrict Time Element coverage limits in order to avoid any disincentive to rapid repairs.

In the case of mid/high rise condominium association policies no time element coverage is assumed, so it is not modeled.

Direct vs. Indirect Loss

The time element losses are based on empirical functions relating those losses to the interior damage to the structure. The model does not distinguish explicitly between direct and indirect losses to the structure. However, the functions are calibrated against claims data that include both types of losses.

7. ***State the minimum threshold at which time element loss is calculated (e.g., loss is estimated for building damage greater than 20% or only for category 3, 4, 5 events). Provide documentation of validation test results to verify the approach used.***

Time element losses for Personal Residential and low-rise Commercial Residential buildings are calculated as a function of interior damage. There is no minimum threshold at which time element loss is calculated since it is believed that even with minimum interior damage, some time element losses might exist, e.g., when residents are subject to a mandatory evacuation. The approach was validated against claims data as reported in disclosure 12.

8. ***Describe how modeled time element loss costs take into consideration the damage (including damage due to storm surge, flood, and wind) to local and regional infrastructure.***

Time element losses for Personal Residential and low-rise Commercial Residential buildings are calculated as a function of interior damage to the structure. They do not explicitly consider the degree of flood or storm surge damage to the infrastructure. For Personal Residential losses there is potentially some influence of such damage injected through the validation process. For low-rise Commercial Residential losses, however, there were no historical time element losses available for validation.

9. ***Describe the relationship between building structure and contents vulnerability functions.***

The contents vulnerability is a function of the interior damage, which is a main contributor to the building vulnerability. Consequently, the relationship between contents vulnerability and structure vulnerability follows the relationship between overall building structure vulnerability and interior vulnerability.

10. ***Describe the relationship between building structure and time element vulnerability functions.***

The time element vulnerability is a function of the interior damage, which is a main contributor to the building vulnerability. Consequently, the relationship between time element vulnerability and structure vulnerability follows the relationship between overall building structure vulnerability and interior vulnerability.

11. ***Describe the assumptions, data, methods, and processes used to develop contents and time element vulnerability functions for unknown residential construction types.***

Disclosures 10 and 11 are addressed together below.

12. ***Describe the assumptions, data, methods, and processes used to develop contents and time element vulnerability functions when some of the primary characteristics are unknown.***

The development of contents and time element vulnerability functions for unknown residential construction types, or when some of the primary characteristics are unknown, follows the process described in disclosures 11 and 12 of standard V-1.

13. ***Describe any assumptions, data, methods, and processes used to develop and validate contents and time element vulnerability functions concerning insurance company claims.***

The 2004 hurricane insurance provided a wealth of claim data, used to validate and calibrate the FPHLPM (Artiles, 2006; Pinelli et al., 2006). First, the consistency and validity of the data itself was investigated (see standard A-1), and the associated wind speed data was sought from NOAA. The results from the model were then compared to the claim data for hurricanes Charley and Frances. The comparisons were done for the different structural types, for different age categories, and for different insurance companies. They included comparisons of aggregated losses and of vulnerability curves. The comparisons took into account the fact that the actual wind data that caused the damage was not always available, and there was some unknowns regarding the true nature of coverage of many insurance policies. Based on these comparisons, the engineering team recalibrated the engineering model to produce a more accurate and credible predictive capability.

In subsequent years, for every new version of the FPHLM, and as new claim data became available, comparisons of aggregated losses between actual claim data and FPHLM output were performed to validate and calibrate the model. These comparisons are reported in the next disclosure.

The assumption is that company claim payment practices including the effects of contractual obligations on the claim payment process are stable over time and do not vary by company.

14. ***Demonstrate that contents and time element vulnerability function relationships are consistent with insurance claims data.***

The building loss consists of external and internal losses. Contents and time element losses are a function of the interior structure loss. All the losses are based on a combination of engineering principles, empirical equations, and engineering judgment. They were validated against claim data from several hurricanes as described above. The results are shown in Figure 62 and 63 below. Each dot represents an insurance portfolio.

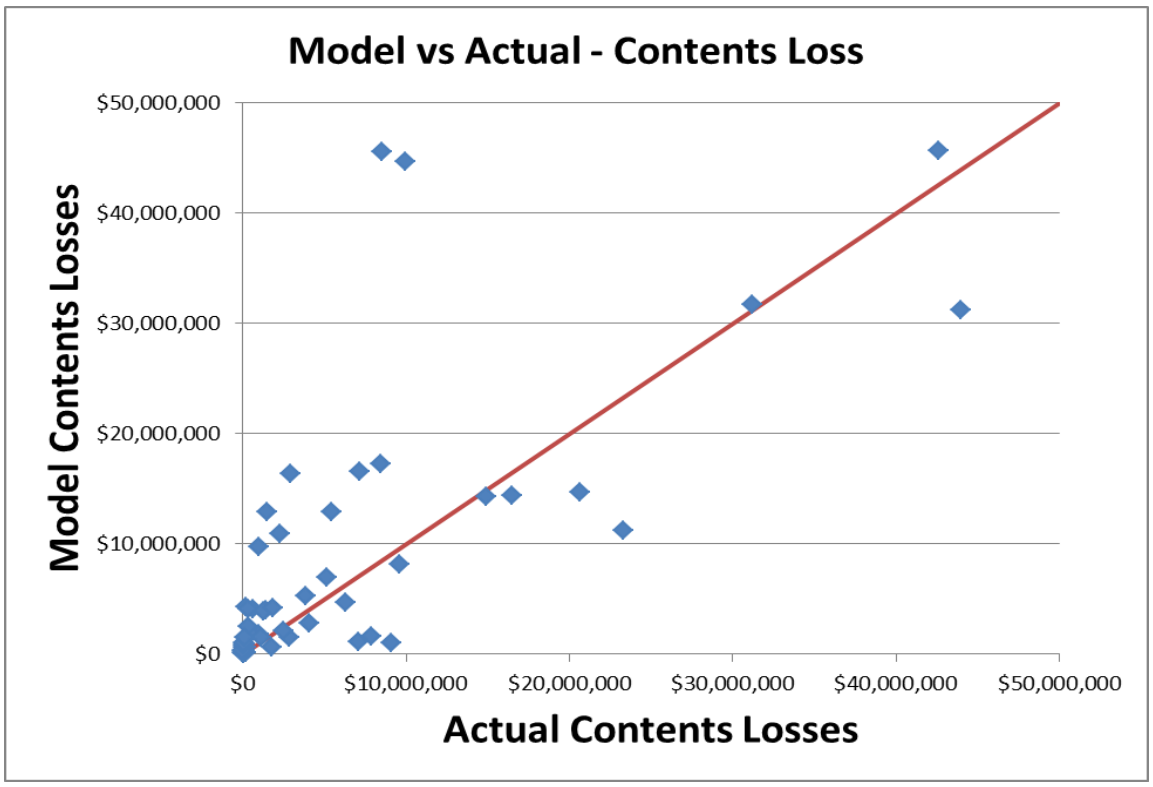


Figure 62. Model vs. Actual-Contents Loss.

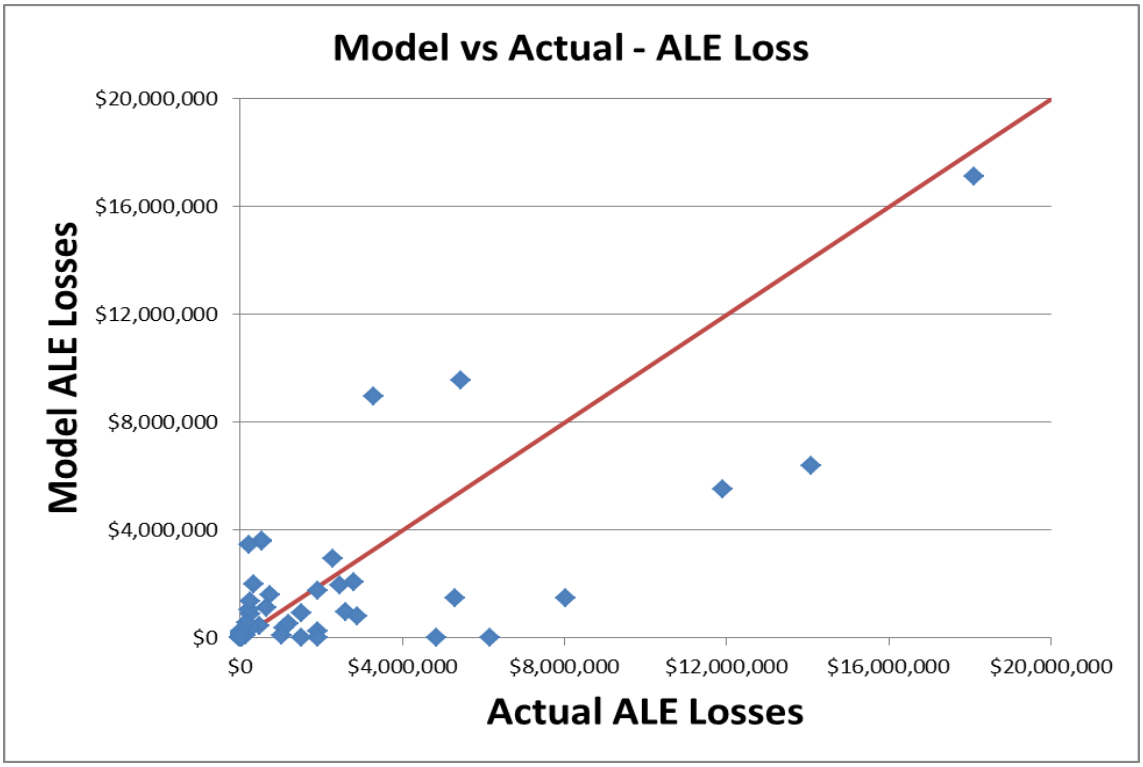


Figure 63. Model vs. Actual-ALE Loss.

V-3 Mitigation Measures

A. Modeling of mitigation measures to improve a building's wind resistance and the corresponding effects on vulnerability shall be theoretically sound and consistent with fundamental engineering principles. These measures shall include fixtures or construction techniques that enhance the performance of the building and its contents and shall consider:

- ***Roof strength***
- ***Roof covering performance***
- ***Roof-to-wall strength***
- ***Wall-to-floor-to-foundation strength***
- ***Opening protection***
- ***Window, door, and skylight strength.***

Modeling of mitigation measures to improve a structure's wind resistance is theoretically sound and includes the fixtures mentioned above. The following structures were modeled:

Base case as defined by Commission
Mitigated case as defined by Commission
Base plus one mitigation at a time

The mitigations include gable bracing, rated shingles, metal roof, stronger sheathing capacity, stronger roof-to-wall connections, stronger wall-to-sill connections, masonry reinforced walls, multiple opening protection options, and wind/missile resistant glass.

B. Application of mitigation measures that enhance the performance of the building and its contents shall be justified as to the impact on reducing damage whether done individually or in combination.

The base cases are very weak cases, where the interior damage is governed by the sheathing loss at low to moderate wind speeds. The application of mitigation measures is justified and the results show the following (Figures 64 through 67).

Bracing the gable end, using rated shingles, using a membrane, or using a metal roof alone does not provide any benefit when all other components remain weak, as required by Form V-2. For example, regardless of the type of roof cover used, if the home loses its weak sheathing panels, there will be little benefit in mitigating the roof cover or gable end alone. Combining mitigation measures, however, does indeed reduce the vulnerability of the home, as demonstrated in the bottom section of Form V-2. The observed negative values in Form V-2 corresponding to the braced gable end mitigation are from round off of smaller values within the uncertainty scatter of the model and indicate zero change.

The hip roof has a greater impact in reducing the losses, especially in the case of frame structures. Because the base frame structure is inherently weaker, there is comparatively a higher gain with the hip timber structure than with the hip masonry structure. For example, a weak home with a hip roof is not vulnerable to gable end collapse.

Improving the roof sheathing capacity (8d nails) alone reduces the damage at wind speeds up to 100 mph and 120 mph sustained winds for wood and masonry structures, respectively, but at higher wind speeds the mitigation becomes counter-effective (Figure 64 and 66). The behavior of the damage curve with mitigated sheathing after 100 (wood) and 120 (masonry) mph sustained winds is due to the still very weak roof-to-wall connections. Loss of sheathing reduces the uplift on the roof-to-wall connections. Thus, the stronger deck results in higher loads on the connections, which the connections are not prepared to absorb. This effect was recently experimentally identified through destructive testing of real structures with toe-nail connections and strong decking attachment (Shanmugam et al., 2009).

Clips and straps are very effective for frame structures, less so for masonry structures. The model emphasizes interior damage due to loss of sheathing, roof cover, or gable end, which are all independent of the roof-to-wall connection strength. If the strength of the plywood deck and roof cover is not increased, increasing the roof-to-wall connections alone will do little good at low to moderate wind speeds. At higher wind speeds, the integrity of the box system in the frame structure is improved by the stronger roof-to-wall connection, hence the more pronounced benefit for the frame structure than for masonry.

Clips and straps for wall-to-sill plate connections are very effective at high wind speeds for frame structures because they improve the integrity of the box system. Similarly, the reinforcing of the walls for masonry structures is more effective at high wind speeds when unreinforced walls become vulnerable.

Opening protections are effective, and more so at higher wind speeds. This follows logically, as the internal pressurization caused by an opening breach is critical to the failure of other components only at higher wind speeds.

A mitigated structure with a combination of individual mitigations (as per standards definition) shows improved performance over the base structure and each of the individual mitigations.

The nonzero damage between 40 and 60 mph sustained winds, the convergence of the base, and all mitigation cases in this wind speed range reflect the incorporation of non-exterior damage-related losses in the model. Water penetration through windows and doors is possible even without window or door breach (Salzano et al., 2010). This portion of the model is not dependent upon mitigations, thus the convergence of curves in Figures 64 through 67 in that wind speed range.

Disclosures

1. ***Describe any modifications to mitigation measures in the model since the previously accepted model.***

None to be reported.

2. ***Provide a completed Form V-2, Mitigation Measures – Range of Changes in Damage. Provide a link to the location of the form here.***

See Form V-2. Notice that there are no entries for the Wall-Foundation Strength rows for timber structures because the model does not have the capability to model wall-to-foundation anchors or straps for timber structures. The model does account for wall-to-sill plate connections, but not the sill plate-to-foundation connections. There are no field data to indicate that this is a significant failure mode. The connection to the foundation can be weak and is reflected in the wall-to-sill capacity (toe-nails, clips, straps).

3. ***Provide a description of the mitigation measures used by the model that are not listed in Form V-2, Mitigation Measures – Range of Changes in Damage.***

None to be reported.

4. ***Describe how mitigation is implemented in the model. Identify any assumptions.***

The various mitigation options delineated in Forms V-2 and V-3 are implemented in the model by varying the capacity model parameters (mean and coefficient of variation) to reflect the strength of a given component. For example, weak (base model) roof covering is represented by a random value for each shingle, with the specific capacity values for a given Monte Carlo simulation randomly assigned on the basis of a specified probability density function, mean, and coefficient of variation assigned to weak shingles. If the strong roof cover mitigation option is chosen, a different mean and coefficient of variation, reflecting higher capacity and less variability, are used to randomly assign capacities to the shingles. This same approach is used for every component for which a mitigation option is modeled. One or any combination of mitigation measures may be selected prior to running the Monte Carlo simulation. The stronger resistances of the mitigated components are directly reflected in the randomly assigned capacities of those components. In the case of membrane, the mitigation is modeled through a reduction of the interior damage due to loss of roof cover and subsequent water penetration.

In the case of membrane, the mitigation is modeled through a reduction of the interior damage due to loss of roof cover and subsequent water penetration.

5. ***Describe the process used to ensure that multiple mitigation factors are correctly combined in the model.***

Each mitigation option (e.g., sheathing, roof cover, membrane, roof-to-wall connections) is modeled and accounted for independently, allowing any combination to be chosen. As reflected in the results in Figures 64 through 67, it is assumed that the effect of mitigating one component can change the vulnerability but not the capacity of other components via the influence that mitigation has on loading or load sharing. It is also assumed that any given mitigation does not necessarily produce improved overall performance for all wind speeds. An example already discussed is the influence of the roof sheathing strength on the vulnerability of roof-to-wall connections, caused by the influence of intact strong roof sheathing on the uplift acting on weak roof-to-wall connections. Another example is the influence of opening vulnerability on the performance of other components (walls, sheathing, and roof-to-wall connections), as the change in internal pressure resulting from opening failure changes the loading on these other components.

In summary, mitigation options may be selected individually or in combination, but the effects of a given mitigation on other components and on overall building vulnerability, should not be and are not isolated in the model.

Form V-1: One Hypothetical Event

A. *Windspeeds for 96 ZIP Codes and sample personal and commercial residential exposure data are provided in the file named "FormV1Input13.xls." The windspeeds and ZIP Codes represent a hypothetical hurricane track. Model the sample personal and commercial residential exposure data provided in the file against these windspeeds at the specified ZIP Codes and provide the damage ratios summarized by windspeed (mph) and construction type.*

The wind speeds provided are one-minute sustained 10-meter wind speeds. The sample personal and commercial residential exposure data provided consist of four structures (one of each construction type: wood frame, masonry, mobile home, and concrete) individually placed at the population centroid of each of the ZIP Codes provided. Each ZIP Code is subjected to a specific wind speed. For completing Part A, Estimated Damage for each individual wind speed range is the sum of ground up loss to all structures in the ZIP Codes subjected to that individual wind speed range, excluding demand surge and storm surge. Subject Exposure is all exposures in the ZIP Codes subjected to that individual wind speed range. For completing Part B, Estimated Damage is the sum of the ground up loss to all structures of a specific type (wood frame, masonry, mobile home, or concrete) in all of the wind speed ranges, excluding demand surge and storm surge. Subject Exposure is all exposures of that specific type in all of the ZIP Codes.

One reference structure for each of the construction types shall be placed at the population center of the ZIP Codes. Do not include contents, appurtenant structures, or time element coverages.

Reference Frame Structure:	***Reference Masonry Structure:***
One story *Unbraced gable end roof* *Normal shingles (55mph)* *½" plywood deck* *6d nails, deck to roof members* *Toe nail truss to wall anchor* *Wood framed exterior walls* *5/8" diameter anchors at 48" centers for wall/floor/foundation connections* *No shutters* *Standard glass windows* *No door covers* *No skylight covers* *Constructed in 1980*	*One story* *Unbraced gable end roof* *Normal shingles (55mph)* *½" plywood deck* *6d nails, deck to roof members* *Toe nail truss to wall anchor* *Masonry exterior walls* *No vertical wall reinforcing* *No shutters* *Standard glass windows* *No door covers* *No skylight covers* *Constructed in 1980*
Reference Mobile Home Structure:	***Reference Concrete Structure:***
Tie downs *Single unit* *Manufactured in 1980*	*Twenty story* *Eight apartment units per story* *No shutters* *Standard glass windows* *Constructed in 1980*

B. Confirm that the structures used in completing the form are identical to those in the above table for the reference structures. If additional assumptions are necessary to complete this form (for example, regarding structural characteristics, duration or surface roughness), provide the reasons why the assumptions were necessary as well as a detailed description of how they were included.

The modelers do confirm that the structures used in completing the form are identical to those in the table provided in the standard.

C. Provide a plot of the Form V-1 (One Hypothetical Event), Part A data.

See Appendix T.

Form V-2: Mitigation Measures – Range of Changes in Damage

A. Provide the change in the zero deductible personal residential reference building damage rate (not loss cost) for each individual mitigation measure listed in Form V-2 (Mitigation Measures – Range of Changes in Damage) as well as for the combination of the four mitigation measures provided for the Mitigated Frame Building and the Mitigated Masonry Building below.

See Appendix U.

B. If additional assumptions are necessary to complete this form (for example, regarding duration or surface roughness), provide the rationale for the assumptions as well as a detailed description of how they are included.

C. Provide this Form in Excel format without truncation. The file name shall include the abbreviated name of the modeling organization, the standards year, and the form name. Form V-2 (Mitigation Measures – Range of Changes in Damage) shall be included in a submission appendix.

Reference Frame Building:	**Reference Masonry Building:**
One story *Unbraced gable end roof* *Normal shingles (55mph)* *½" plywood deck* *6d nails, deck to roof members* *Toe nail truss to wall anchor* *Wood framed exterior walls* *5/8" diameter anchors at 48" centers for wall/floor/foundation connections* *No shutters* *Standard glass windows* *No door covers* *No skylight covers* *Constructed in 1980*	*One story* *Unbraced gable end roof* *Normal shingles (55mph)* *½" plywood deck* *6d nails, deck to roof members* *Toe nail truss to wall anchor* *Masonry exterior walls* *No vertical wall reinforcing* *No shutters* *Standard glass windows* *No door covers* *No skylight covers* *Constructed in 1980*
Mitigated Frame Building: *Rated shingles (110mph)* *8d nails, deck to roof members* *Truss straps at roof* *Plywood Shutters*	**Mitigated Masonry Building:** *Rated shingles (110mph)* *8d nails, deck to roof members* *Truss straps at roof* *Plywood Shutters*

Reference and mitigated buildings are fully insured buildings with a zero deductible building only policy.

Place the reference building at the population centroid for ZIP Code 33921.

Wind speeds used in the form are one-minute sustained 10-meter wind speeds.

See Appendix U.

Form V-3: Mitigation Measures – Mean Damage Ratios and Loss Costs (Trade Secret Item)

A. *Provide the mean damage ratio (prior to any insurance considerations) to the reference building for each individual mitigation measure listed in Form V-3 (Mitigation Measures – Mean Damage Ratios and Loss Costs) as well as the percent damage for the combination of the four mitigation measures provided for the Mitigated Frame building and the Mitigated Masonry building below.*

See Form V-3 below. Notice that for the 60 mph column all the vulnerabilities coincide at 6%. This is because at these low wind speeds, no significant damage is activated to trigger any significant difference between the different cases.

B. *Provide the loss cost rounded to three decimal places, for the reference building and for each individual mitigation measure listed in Form V-3 (Mitigation Measurers – Mean Damage Ratios and Loss Costs, Trade Secret item) as well as the loss cost for the combination of the four mitigation measures provided for the Mitigated Frame Building and the Mitigated Masonry Building below.*

See Form V-3 below.

C. *If additional assumptions are necessary to complete this Form (for example, regarding duration or surface roughness), provide the rationale for the assumptions as well as a detailed description of how they are included.*

D. *Provide a graphical representation of the vulnerability curves for the reference and the fully mitigated buildings.*

See Figures 64 through 67. Because there are too many vulnerability curves to plot in one figure, for the sake of clarity, the mitigations were divided in four sets for both masonry and frame structures. In each figure, there are two horizontal axes: the upper axis represents the actual terrain three-second gusty winds; the lower axis represents the actual terrain one-minute sustained winds. The conversion between three-second gust and one-minute sustained winds depends on the roughness of the terrain. Therefore, on each plot, the value of the roughness parameter for Lee County is indicated. Finally, please note that, as explained in the previous section, mitigating the roof shingles alone, or the metal roof alone, or the membrane alone without mitigating the roof deck (upgrading nail size and or spacing) or the roof-to-wall connections does not improve the overall vulnerability of the structure. Consequently, in Figures 64 through 67, the curves for the base case and the rated shingle, metal roof, and membrane cases are superimposed on each other. This result is dependent on the base case weak sheathing connection and should not be interpreted to imply that reroofing is not an effective mitigation. Reroofing is only ineffective for the case of a very weak roof deck. The combination of re-nailing the decking and reroofing (now required practice) is an effective mitigation.

Reference Frame Structure: *One story* *Unbraced gable end roof* *Normal shingles (55mph)* *½" plywood deck* *6d nails, deck to roof members* *Toe nail truss to wall anchor* *Wood framed exterior walls* *5/8" diameter anchors at 48" centers for wall/floor/foundation connections* *No shutters* *Standard glass windows* *No door covers* *No skylight covers* *Constructed in 1980*	***Reference Masonry Structure:*** *One story* *Unbraced gable end roof* *Normal shingles (55mph)* *½" plywood deck* *6d nails, deck to roof members* *Toe nail truss to wall anchor* *Masonry exterior walls* *No vertical wall reinforcing* *No shutters* *Standard glass windows* *No door covers* *No skylight covers* *Constructed in 1980*
Mitigated Frame Structure: *Rated shingles (110mph)* *8d nails, deck to roof members* *Truss straps at roof* *Plywood Shutters*	***Mitigated Masonry Structure:*** *Rated shingles (110mph)* *8d nails, deck to roof members* *Truss straps at roof* *Plywood Shutters*

Reference and mitigated buildings *are fully insured building structures with a zero deductible building only policy.*

Place the reference building *at the population centroid for ZIP Code.*

Wind speeds used in the form are one-minute sustained 10-meter wind speeds.

Form V-3: Mitigation Measures – Mean Damage Ratio (1 min)

INDIVIDUAL MITIGATION MEASURES			MEAN DAMAGE RATIO										LOSS COSTS	
			FRAME BUILDING					MASONRY BUILDING					FRAME BUILDING	MASONRY BUILDING
			WIND SPEED (MPH)					WIND SPEED (MPH)					ACROSS ALL WINDSPEEDS	
			60	85	110	135	160	60	85	110	135	160		
	REFERENCE BUILDING		6%	14%	40%	56%	67%	6%	14%	36%	47%	62%	$5.544	$5.414
ROOF STRENGTH														
	BRACED GABLE ENDS		6%	14%	40%	55%	66%	6%	14%	36%	47%	61%	$5.544	$5.414
	HIP ROOF		6%	13%	38%	50%	64%	6%	13%	35%	43%	59%	$5.415	$5.287
ROOF COVERING														
	METAL		6%	14%	40%	56%	67%	6%	14%	36%	47%	62%	$5.538	$5.409
	RATED SHINGLES (110 MPH)		6%	14%	40%	56%	67%	6%	14%	36%	47%	62%	$5.538	$5.409
	MEMBRANE		6%	14%	40%	56%	67%	6%	14%	36%	47%	62%	$5.544	$5.414
	NAILING OF DECK 8d		6%	9%	40%	60%	67%	6%	8%	31%	48%	63%	$5.115	$4.985
ROOF-WALL STRENGTH														
	CLIPS		6%	14%	38%	48%	59%	6%	14%	35%	43%	54%	$5.541	$5.412
	STRAPS		6%	14%	38%	46%	51%	6%	14%	35%	43%	53%	$5.541	$5.412
WALL-FLOOR STRENGTH														
	TIES OR CLIPS		6%	14%	38%	52%	64%	-	-	-	-	-	$5.541	-
	STRAPS		6%	14%	38%	52%	64%	-	-	-	-	-	$5.541	-
WALL FOUNDATION STRENGTH														
	LARGER ANCHORS OR CLOSER SPACING		-	-	-	-	-	-	-	-	-	-	-	-
	STRAPS		-	-	-	-	-	-	-	-	-	-	-	-
	VERTICAL REINFORCING		-	-	-	-	-	6%	14%	35%	42%	48%	-	$5.409
OPENING PROTECTION														
	WINDOW SHUTTERS	PLYWOOD	6%	14%	37%	54%	67%	6%	13%	33%	45%	62%	$5.533	$5.399
		STEEL	6%	14%	36%	53%	66%	6%	13%	31%	44%	61%	$5.527	$5.399
		ENGINEERED	6%	13%	34%	51%	65%	6%	13%	30%	42%	60%	$5.523	$5.396
	DOOR AND SKYLIGHT COVERS		6%	14%	39%	55%	66%	6%	14%	35%	46%	61%	$5.541	$5.412
WINDOW DOOR, SKYLIGHT STRENGTH		LAMINATED	6%	14%	35%	52%	65%	6%	13%	31%	43%	60%	$5.528	$5.400
		IMPACT GLASS	6%	14%	34%	50%	63%	6%	13%	30%	41%	58%	$5.526	$5.399
	ENTRY DOORS	HIGH STRENGTH	6%	14%	39%	55%	66%	6%	14%	35%	46%	61%	$5.543	$5.414
	GARAGE DOORS	HIGH STRENGTH	6%	12%	38%	55%	67%	6%	11%	34%	46%	62%	$5.409	$5.287
	SLIDING GLASS DOORS	HIGH STRENGTH	6%	14%	39%	55%	66%	6%	14%	35%	46%	61%	$5.540	$5.412

MITIGATION MEASURES IN COMBINATION	MEAN DAMAGE RATIO										LOSS COSTS	
	FRAME BUILDING					MASONRY BUILDING					FRAME BUILDING	MASONRY BUILDING
	WIND SPEED (MPH)					WIND SPEED (MPH)					ACROSS ALL WINDSPEEDS	
	60	85	110	135	160	60	85	110	135	160		
MITIGATED BUILDING	6%	8%	29%	41%	50%	6%	8%	27%	39%	52%	$5.099	$4.973

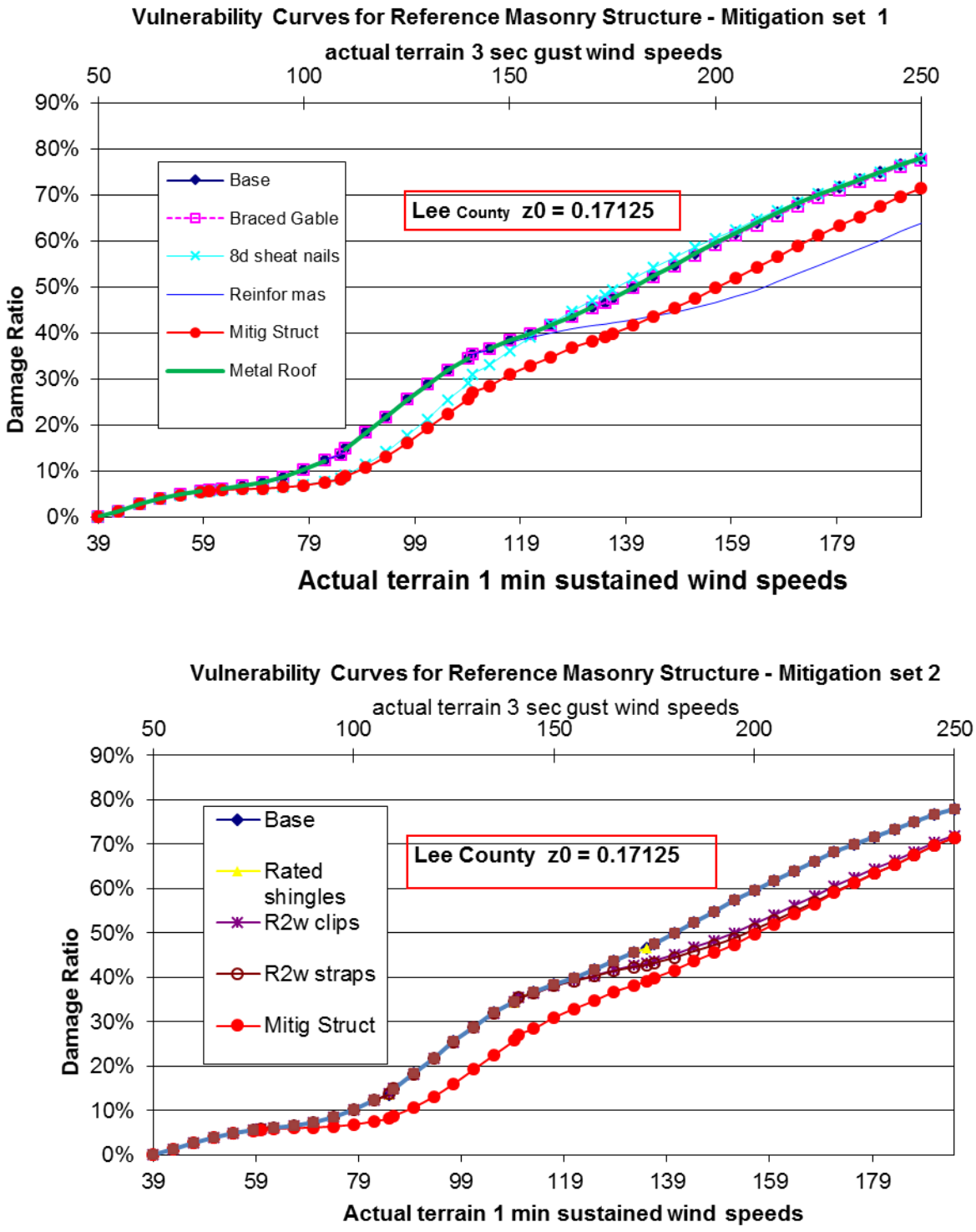
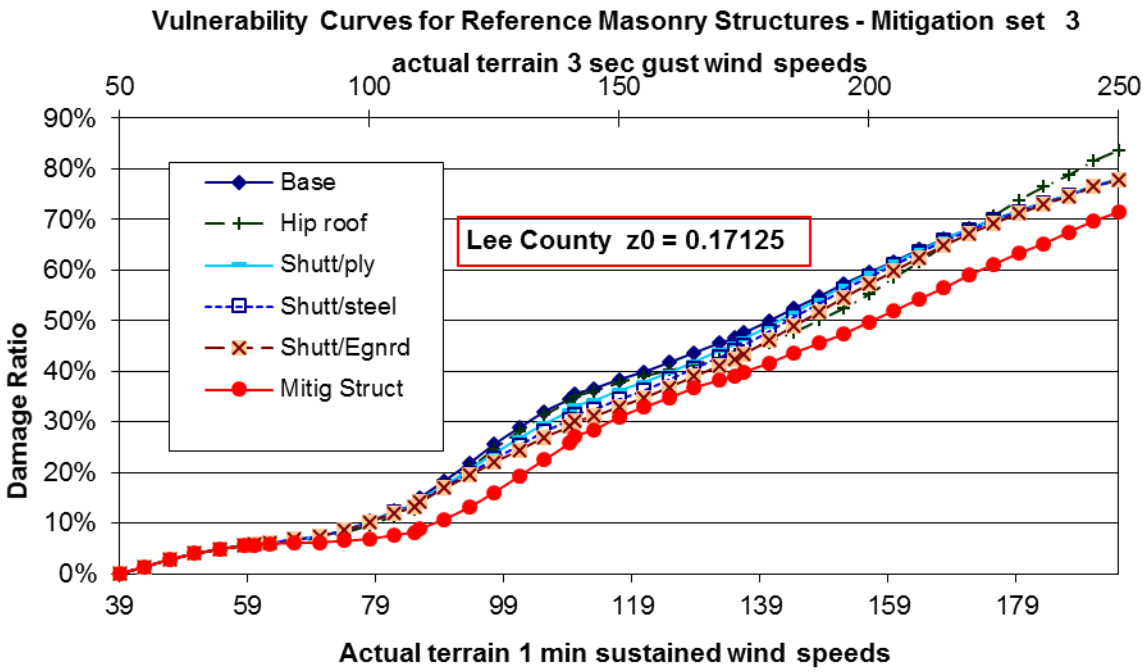


Figure 64. Mitigation measures for masonry homes.



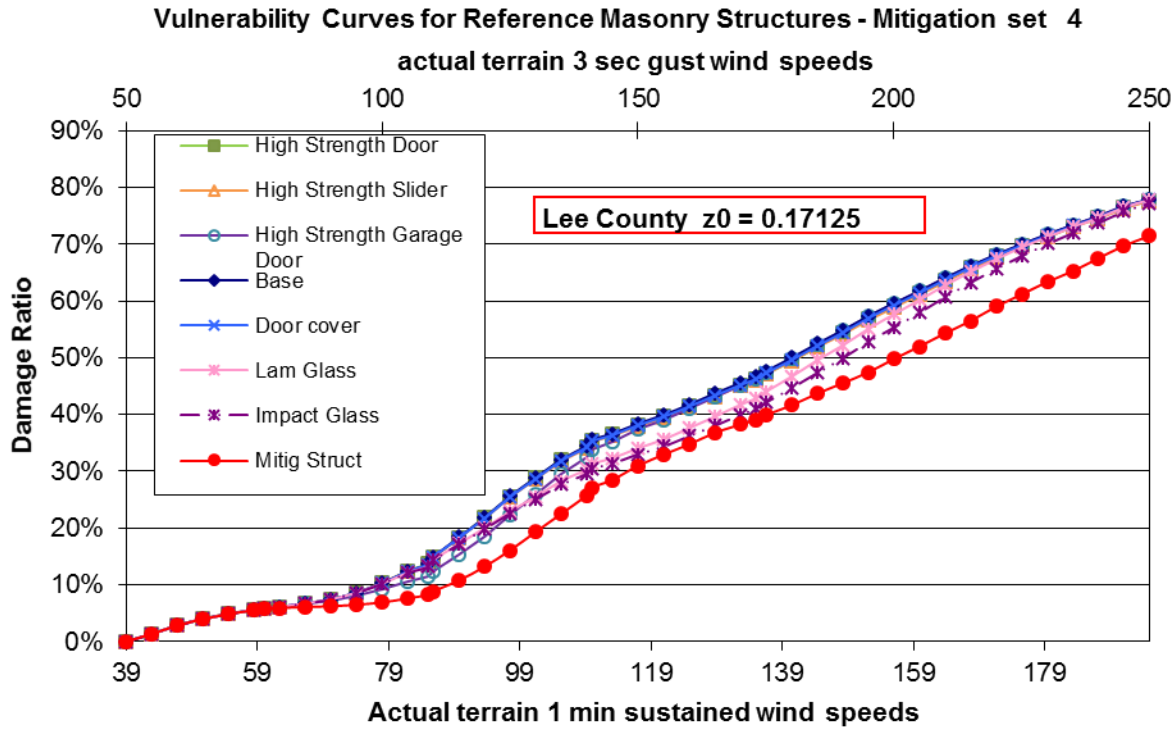
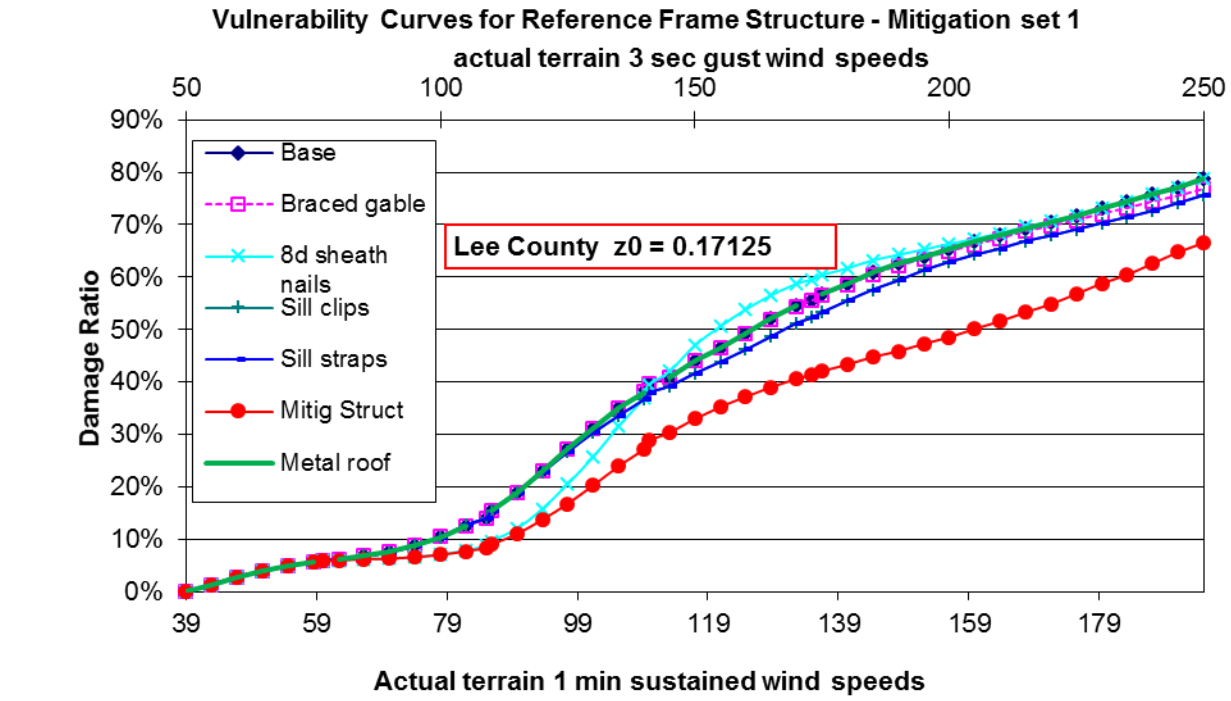


Figure 65. Mitigation measures for masonry homes.



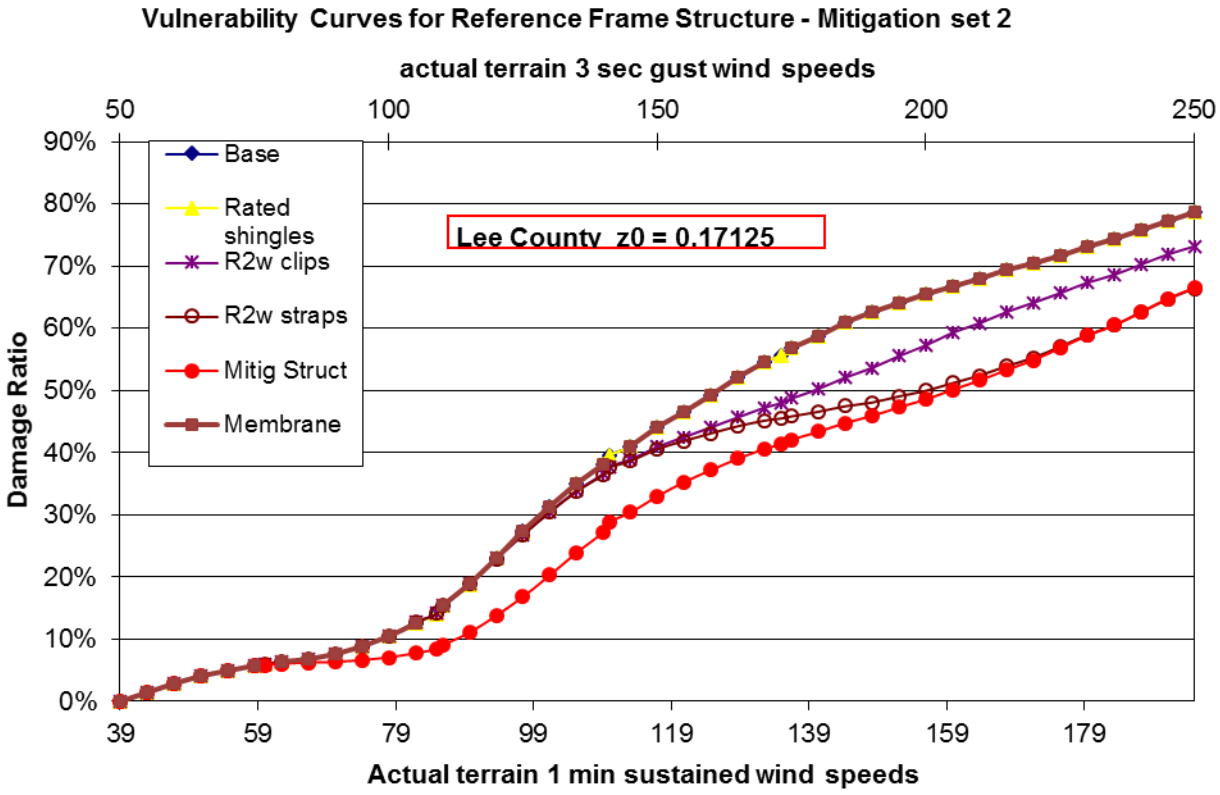
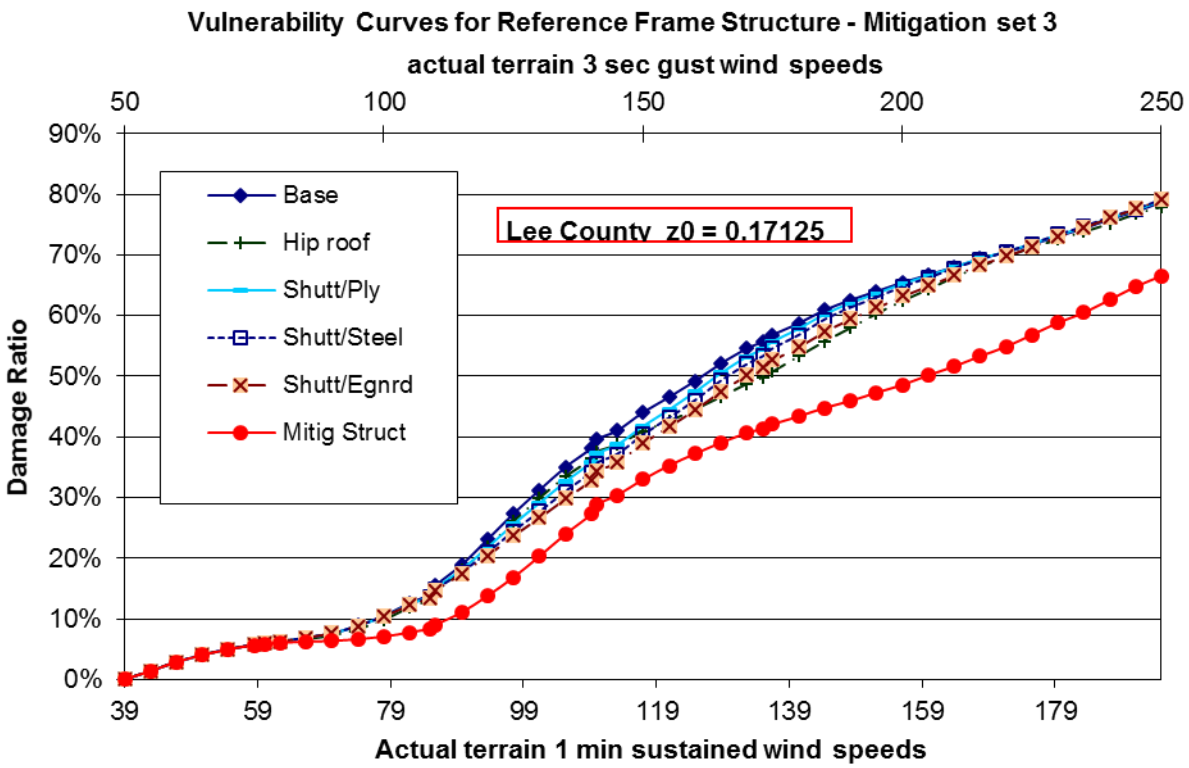


Figure 66. Mitigation measures for frame homes.



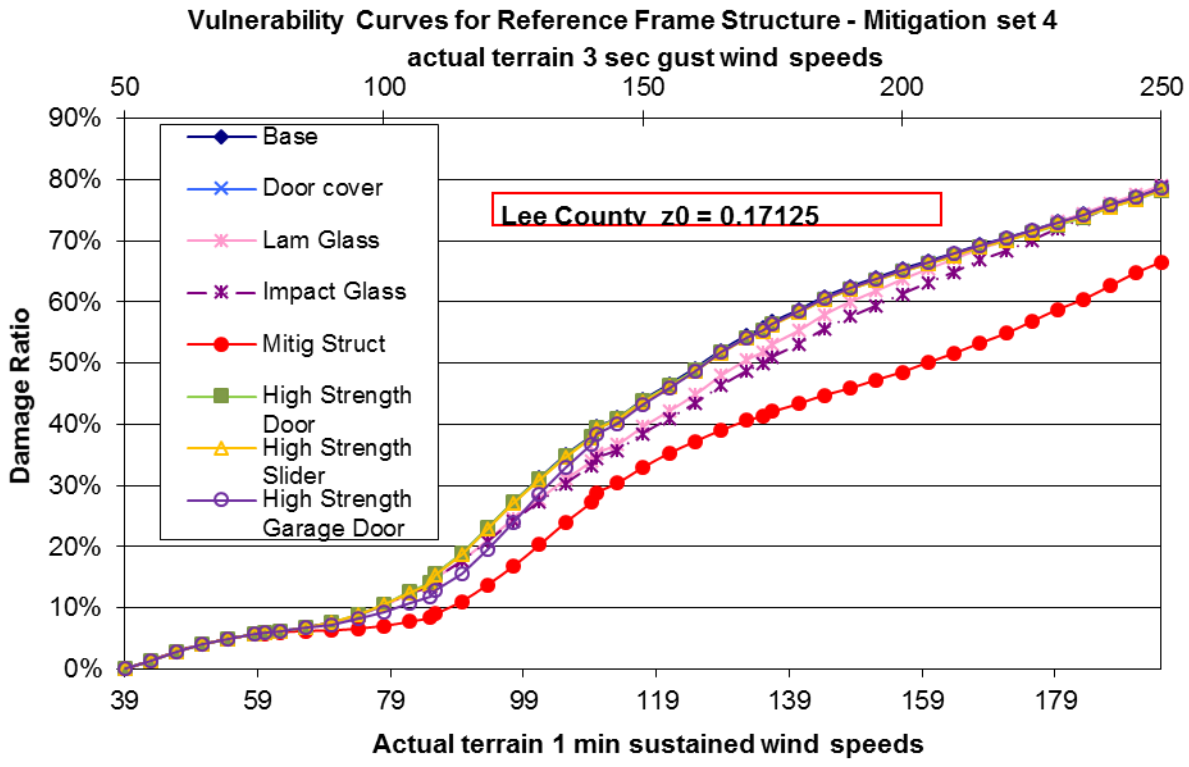


Figure 67. Mitigation measures for frame homes.

ACTUARIAL STANDARDS

A-1 Modeling Input Data

A. ***When used in the modeling process or for verification purposes, adjustments, edits, inclusions, or deletions to insurance company input data used by the modeling organization shall be based upon accepted actuarial, underwriting, and statistical procedures.***

B. ***All modifications, adjustments, assumptions, inputs and input file identification, and defaults necessary to use the model shall be actuarially sound and shall be included with the model output report. Treatment of missing values for user inputs required to run the model shall be actuarially sound and described with the model output report.***

Disclosures

1. ***Identify depreciation assumptions and describe the methods and assumptions used to reduce insured losses on account of depreciation. Provide a sample calculation for determining the amount of depreciation and the actual cash value (ACV) losses.***

For both replacement cost and ACV policies, the value of structures and contents are generally assumed to equal the insured limit. In the rare case where data on property value are available from the insurance company and that value exceeds the limit, the value provided is used to estimate the ground-up damages.

Depreciation is considered in the model, but not explicitly. The damage ratios were calibrated to insured losses that contained a mix of replacement cost and ACV policies but primarily replacement cost. Consequently there is an implicit allowance for depreciation (of an unknown degree) built into the modeled losses.

2. ***Identify insurance-to-value assumptions and describe the methods and assumptions used to determine the true property value and associated losses. Provide a sample calculation for determining the property value and guaranteed replacement cost losses.***

The model assumes that the insured value is the true value of the property except in rare cases when the insurance company provides a separate property value that is higher than the insured value.

3. ***Describe the methods used to distinguish among policy form types (e.g., homeowners, dwelling property, mobile home, tenants, condo unit owners).***

The input record provided by the company includes a "policy form" code. If there is any ambiguity, the company is contacted for clarification.

4. ***Disclose, in a model output report, the specific type of input that is required to use the model***

or model output in a residential property insurance rate filing. Such input includes, but is not limited to, optional features of the model, type of data to be supplied by the model user and needed to derive loss projections from the model, and any variables that a model user is authorized to set in using the model. Include the model name and version identification on the model output report. All items included in the output form submitted to the Commission shall be clearly labeled and defined.

A model output report follows.

Table 30. Output report for OIR data processing.

Output Report for OIR Data Processing

Florida Public Hurricane Loss Model: Release 6.0

OIR Data Processing Results: <Company Name: OIR Filing Number>

Report Content:
- Original Number of the policies in data set
- Process steps to formalize the data set
- Numbers of policies which are excluded due to certain reason, e.g. invalid ZIP Codes, invalid format, etc.
- Numbers of: Construction Types, Territory Codes, Policy Forms, Program Codes, etc.

- Coverage limits for building, appurtenant structure, content, additional living expense
- Distribution of deductibles
- Number of records that change values for different types of roof shape, roof cover, roof membrane, roof to wall connection, nailing of deck, garage door, opening protection, due to missing or illogical values
-Number of records for a county whose name is changed due to inconsistencies with the zip codes

- Number of policies to generate the estimated losses
- Number of files in the report

The results are aggregated by different combinations of counties, ZIP Codes, policy forms, program codes, and territory codes as applicable.

In case if there are:
- more than 1 construction type
- more than 1 policy form
- more than 1 program code
- more than 1 territory code

There will be 40 files in the report for personal residential policies with names as below:

<CompanyName>_PERSONAL_Loss_ConstType.xls
<CompanyName>_PERSONAL_Loss_County.xls
<CompanyName>_PERSONAL_Loss_PolicyForm.xls
<CompanyName>_PERSONAL_Loss_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_TerritoryCode.xls
<CompanyName>_PERSONAL_Loss_Zipcode.xls
<CompanyName>_PERSONAL_Loss_ConstType_PolicyForm.xls
<CompanyName>_PERSONAL_Loss_ConstType_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_ConstType_TerritoryCode.xls
<CompanyName>_PERSONAL_Loss_County_ConstType.xls
<CompanyName>_PERSONAL_Loss_County_PolicyForm.xls
<CompanyName>_PERSONAL_Loss_Zipcode_ConstType.xls

<CompanyName>_PERSONAL_Loss_County_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_County_TerritoryCode.xls
<CompanyName>_PERSONAL_Loss_Zipcode_PolicyForm.xls
<CompanyName>_PERSONAL_Loss_PolicyForm_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_PolicyForm_TerritoryCode.xls
<CompanyName>_PERSONAL_Loss_TerritoryCode_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_Zipcode_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_Zipcode_TerritoryCode.xls
<CompanyName>_PERSONAL_Loss_ConstType_PolicyForm_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_ConstType_PolicyForm_TerritoryCode.xls
<CompanyName>_PERSONAL_Loss_ConstType_TerritoryCode_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_County_ConstType_PolicyForm.xls
<CompanyName>_PERSONAL_Loss_County_ConstType_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_County_ConstType_TerritoryCode.xls
<CompanyName>_PERSONAL_Loss_County_PolicyForm_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_County_PolicyForm_TerritoryCode.xls
<CompanyName>_PERSONAL_Loss_County_TerritoryCode_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_Zipcode_ConstType_PolicyForm.xls
<CompanyName>_PERSONAL_Loss_Zipcode_ConstType_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_Zipcode_PolicyForm_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_ConstType_PolicyForm_TerritoryCode_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_County_ConstType_PolicyForm_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_County_ConstType_PolicyForm_TerritoryCode.xls
<CompanyName>_PERSONAL_Loss_County_ConstType_TerritoryCode_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_County_PolicyForm_TerritoryCode_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_Zipcode_ConstType_PolicyForm_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_PolicyForm_TerritoryCode_ProgramCode.xls
<CompanyName>_PERSONAL_Loss_County_ConstType_PolicyForm_TerritoryCode_ProgramCode.xls

There will be 9 files in the report for commercial residential policies with names as below:

< CompanyName>_COMMERCIAL_Loss_ConstType.xls
<CompanyName>_COMMERCIAL_Loss_County.xls
<CompanyName>_COMMERCIAL_Loss_TerritoryCode.xls
<CompanyName>_COMMERCIAL_Loss_Zipcode.xls
<CompanyName>_COMMERCIAL_Loss_ConstType_TerritoryCode.xls
<CompanyName>_COMMERCIAL_Loss_County_ConstType.xls
<CompanyName>_COMMERCIAL_Loss_Zipcode_ConstType.xls
<CompanyName>_COMMERCIAL_Loss_County_TerritoryCode.xls
<CompanyName>_COMMERCIAL_Loss_County_ConstType_TerritoryCode.xls

There will be 9 files in the report for combined personal and commercial residential policies with names as below:

< CompanyName>_Loss_ConstType.xls
<CompanyName>_Loss_County.xls
<CompanyName>_Loss_TerritoryCode.xls
<CompanyName>_Loss_Zipcode.xls
<CompanyName>_Loss_ConstType_TerritoryCode.xls
<CompanyName>_Loss_County_ConstType.xls
<CompanyName>_Loss_ZIPcode_ConstType.xls
<CompanyName>_Loss_County_TerritoryCode.xls
<CompanyName>_Loss_County_ConstType_TerritoryCode.xls

5. ***Provide a copy of the input form(s) used in the model with options chosen to reflect the Florida hurricane model under review. Describe the process followed by the user to generate the model output produced from the input form. Include the model name and version***

identification on the input form. All items included in the input form submitted to the Commission shall be clearly labeled and defined.

Florida Public Hurricane Loss Model: Version 6.0
Inputs for Personal Residential Policies

Provide input data only for policies that include wind coverage. The policy records should be saved in .txt files with the following format:

PolicyID,Zipcode,YearBuilt,ConstructionType,PropertyValue,StructureCoverage,AppCoverage,ContentCoverage, ALECoverage,Deductible,HurricaneDeductible,NatureOfCoverage,County,Address,City,Form,Programcode,Territ oryCode, Year retrofitted,NumberOfStories,LocationOfUnit,NumberOfUnits,Areaof building,Roof shape,Roof cover,Roof membrane,Roof to wall connection, DeckAttachment,Garage door,Opening protection.

Input files containing personal residential policies to be processed through version 6.0 of the Florida Public Hurricane Loss Model should adhere to the format specifications contained in this document.

Please observe the following when preparing the input file:

(a) Provide one policy per line in a comma-separated values file (.csv).
(b) Do not use comma within the fields' values (e.g., as thousand separators or within addresses).
(c) Include the name of each column in the first line of the file.
(d) For fields that require a code, enter the code that more closely represents the data value.
(e) Only include policies with wind coverage.

Each policy should contain a total of 29 attributes. Attributes 1-17 are the minimum required attributes. Attributes 18-29 are required secondary modifiers. Please always provide all 29 attributes.

1. Policy Id — A unique identifier for this policy in the data file. An alphanumeric text.

2. ZIP Code — The ZIP Code where this building is located. A 5-digit number.

3. Year Built — The year in which the property was built. A 4-digit number or UNKNOWN.

4. Construction Type — The construction type of the building. Please encode the data to one of the following:

Value	Code
Frame	1
Masonry	2
Manufactured	3
Other	4
Unknown	5

5. Structure Coverage — The structure coverage amount in dollars. Enter 0 if none.

6. App. Coverage — The appurtenant structure coverage amount in dollars. Enter 0 if none.

7. Content Coverage The content coverage amount in dollars. Enter 0 if none.

8. ALE Coverage The additional living expense coverage amount in dollars. Enter 0 if none.

9. Deductible The deductible amount for perils other than hurricane. Dollar amount (convert percentages to dollar amounts).

10. Hurricane Deductible The hurricane deductible amount in dollars (convert percentages to dollar amounts)

11. Nature of Coverage The settlement option on the structure. Please encode the data to one of the following:

Value	Code
Replacement Cost	R
Actual Cash Value	A

12. County The name of the county where the building is located.

13. Address The street address or geographic coordinates of the building. If providing coordinates, please enter as longitude; latitude.

14. City The name of the city where the building is located.

15. Form Policy Form (HO-1, HO-2, HO-3, HO-5, HO-8, HO-4, HO-6, DP-1, DP-2, DP-3, etc.)

16. Program Code Use one uppercase letter to represent each company program.

17. Territory Code Use the territory codes reflected in your rate manual.

18. Year Retrofitted The 4-digit year when the property was retrofitted (brought up to code).
If only the year of roof replacement is known, enter the 4-digit year when the roof was replaced followed by R (i.e. if the roof was replaced in 1999, enter 1999R).
If not retrofitted enter NA. If not known enter UNKNOWN.

19. Number of Stories Number of stories in the building (e.g., 1, 2, 3, etc.) or UNKNOWN.

20. Location of Unit The story in which the unit is located (e.g., 1, 2, 3, etc.) or UNKNOWN.
Only applicable to HO-4 and HO-6 policies. Enter "NA" for all other policy types.

21. Sliders Indicates whether the unit has sliders. Please encode the data to one of the following:

Value	Code
No Sliders	0
Sliders	1
Unknown	2
Not HO-4 / HO-6	NA

22. Area of Property The total number of square feet for all floors of the insured property or UNKNOWN.

23. Roof Shape Please encode the data to one of the following:

Value	Code
Unbraced Gable	1
Braced Gable	2
Gable (Unknown bracing)	3
Hip	4
Other	5
Unknown	6

24. Roof Cover Please encode the data to one of the following:

Value	Code
Unrated Shingles	1
Rated Shingles (Current FBC)	2
Shingles (Unknown rating)	3
Tiles	4
Metal	5
Other FBC Compliant	6
Other Non-FBC Compliant	7
Unknown	8

25. Roof Membrane Please encode the data to one of the following:

Value	Code
Regular Underlayment	1
Secondary Water Resistance	2
Other	3
Unknown	4

26. Roof-to-Wall Connection Please encode the data to one of the following:

Value	Code
Toe Nails	1
Clips	2
Straps	3
Other	4
Unknown	5

27. Deck Attachment Please encode the data to one of the following:

Value	Code
Planks	1
Sheathing with 6d@6/12”	2
Sheathing with 8d@6/12”	3
Sheathing with 8d@6/6”	4
Other	5
Unknown	6

28. Garage Door Please encode the data to one of the following:

Value	Code
No garage door	0
Unbraced	1
Braced	2
Unknown	3

29. Opening Protection If at least one glazed opening is not protected, enter as no protection. If there is more than one type of opening protection, use the most predominant type code.
If the only known information is that the policy qualifies for a Basic or Hurricane windstorm loss reduction credit, use code 2.

Value	Code
No Protection	0
Plywood	1
Metal	2
Impact Resistant Glass	3
Other	4
Unknown	5

Example data file with two policies:

PolicyID,ZIPCode,YearBuilt,ConstructionType,StructureCoverage,AppCoverage,ContentCoverage,ALECoverage,Deductible,HurricaneDeductible,NatureOfCoverage,County,Address,City,Form,ProgramCode,TerritoryCode,YearRetrofitted,NumberOfStories,LocationOfUnit,Sliders,AreaOfBuilding,RoofShape,RoofCover,RoofMembrane,RoofToWallConnection,DeckAttachment,GarageDoor,OpeningProtection
ABC100,33143,1981,2,50000,0,20000,8000,1000,1000,R,Miami-Dade,123 Main Street,Miami,HO-6,A,35,NA,1,UNKNOWN,2,1245,6,7,3,5,5,3,5
ABC210,34109,1995,2,115000,0,20000,10000,2500,2500,R,Collier,-81.345593;26.017147,Naples,HO-6,A,35,NA,1,UNKNOWN,2,UNKNOWN,6,7,3,5,5,3,5

Note: the attributes should be separated by comma only.

Florida Public Hurricane Loss Model: Version 6.0
Inputs for Commercial Residential Policies

Provide input data for the Florida Public Hurricane Loss Model that meets the following specifications:
The policy records should be saved in .txt files with the following format:

PolicyID,Location ID,Building ID,Zipcode,YearBuilt,ConstructionType,Number of Stories,Number of Units,Property Value,StructureCoverage,AppCoverage,ContentCoverage,TimeElementCoverage,Deductible, HurricaneDeductible,Coinsurance,NatureOfCoverage,County,Address,City,Form,ProgramCode,TerritoryCode, Year retrofitted,Roof shape,Roof cover,Roof membrane,Roof to wall connection, DeckAttachment,Appurtenant structure,Opening protection,Building layout, AreaofBuilding, Residential Type.

Input files containing personal residential policies to be processed through version 6.0 of the Florida Public Hurricane Loss Model should adhere to the format specifications contained in this document.

Please observe the following when preparing the input file:

(a) Provide one policy per line in a comma-separated values file (.csv).
(b) Do not use comma within the fields' values (e.g., as thousand separators or within addresses).
(c) Include the name of each column in the first line of the file.
(d) For fields that require a code, enter the code that more closely represents the data value.
(e) Only include policies with wind coverage.

Each policy should contain a total of 35 attributes.

1. Policy Id A unique identifier for this policy in the data file. An alphanumeric text.

2. Location Id A unique identifier for the location of the covered building. An alphanumeric text.

3. Building Id A unique identifier for the building. An alphanumeric text.

4. Residency Type Please encode the data to one of the following:

Value	Code
Apartment Building	1
Condominium	2
Unknown	3

5. ZIP Code The ZIP Code where this building is located. A 5-digit number.

6. Year Built The year in which the property was built. A 4-digit number or UNKNOWN.

7. Construction Type The construction type of the building. Please encode the data to one of the following:

Value	Code
Frame	1
Masonry	2
Manufactured	3
Other	4
Unknown	5

8. Structure Coverage The structure coverage amount in dollars. Enter 0 if none.

9. App. Coverage The appurtenant structure coverage amount in dollars. Enter 0 if none.

10. Content Coverage The content coverage amount in dollars. Enter 0 if none.

11. Time Element Coverage The time element coverage amount in dollars. Enter 0 if none.

12. Deductible The deductible amount for perils other than hurricane. Dollar amount (convert percentages to dollar amounts).

13. Hurricane Deductible The hurricane deductible amount in dollars (convert percentages to dollar amounts)

14. Deductible Type The type of hurricane deductible. Please encode the data to one of the following:

Value	Code
Per calendar year	1
Per occurrence	2

15. Nature of Coverage The settlement option on the structure. Please encode the data to one of the following:

Value	Code
Replacement Cost	R
Actual Cash Value	A

16. County The name of the county where the building is located.

17. Address The street address or geographic coordinates of the building. If providing coordinates, please enter as longitude; latitude.

18. City The name of the city where the building is located.

19. Form Policy Form. If company offers different base forms of coverage, enter company code; otherwise, enter 0.

20. Program Code Use one uppercase letter to represent each company program.

21. Territory Code Use the territory codes reflected in your rate manual.

22. Year Retrofitted The 4-digit year when the property was retrofitted (brought up to code).
If only the year of roof replacement is known, enter the 4-digit year when the roof was replaced followed by R (i.e. if the roof was replaced in 1999, enter 1999R).
If not retrofitted enter NA. If not known enter UNKNOWN.

23. Number of Stories Number of stories in the building (e.g., 1, 2, 3, etc.) or UNKNOWN.

24. Total Units The number of units in the building (e.g., 1, 2, 3, etc.) or UNKNOWN.

25. Units per Story The number of units per story (e.g., 1, 2, 3, etc.) or UNKNOWN.

26. Sliders Indicates whether the unit has sliders. Please encode the data to one of the following:

Value	Code
No Sliders	0
Sliders	1
Unknown	2

27. Area of Property The total number of square feet for all floors of the insured property or UNKNOWN.

28. Roof Shape Please encode the data to one of the following:

Value	Code
Unbraced Gable	1
Braced Gable	2
Gable (Unknown bracing)	3
Hip	4
Other	5
Unknown	6

29. Roof Cover Please encode the data to one of the following:

Value	Code
Unrated Shingles	1
Rated Shingles (Current FBC)	2
Shingles (Unknown rating)	3
Tiles	4
Metal	5
Other FBC Compliant	6
Other Non-FBC Compliant	7
Unknown	8

30. Roof Membrane Please encode the data to one of the following:

Value	Code
Regular Underlayment	1
Secondary Water Resistance	2
Other	3
Unknown	4

31. Soffit Please encode the data to one of the following:

Value	Code
None	0
Vinyl	1
Aluminum	2
Plywood	3
Other	4
Unknown	5

32. Roof-to-Wall Connection Please encode the data to one of the following:

Value	Code
Toe Nails	1
Clips	2
Straps	3
Other	4
Unknown	5

33. Deck Attachment Please encode the data to one of the following:

Value	Code
Planks	1
Sheathing with 6d@6/12”	2
Sheathing with 8d@6/12”	3
Sheathing with 8d@6/6”	4
Other	5
Unknown	6

34. Opening Protection If at least one glazed opening is not protected, enter as no protection.
If there is more than one type of opening protection, use the most predominant type code.
If the only known information is that the policy qualifies for a Basic or Hurricane windstorm loss reduction credit, use code 2.

Value	Code
No Protection	0
Plywood	1
Metal	2
Impact Resistant Glass	3
Other	4
Unknown	5

35. Building Layout Please encode the data to one of the following:

Value	Code
Open (Access to units through external balcony)	1
Closed (Access to units through the interior)	2
Unknown	3

Example data file with two policies:

PolicyId,LocationId,BuildingId,ResidencyType,ZIPCode,YearBuilt,ConstructionType,StructureCoverage,AppCoverage,ContentCoverage,TimeElementCoverage,Deductible,HurricaneDeductible,DeductibleType,NatureOfCoverage,County,Address,City,Form,ProgramCode,TerritoryCode,YearRetrofitted,NumberOfStories,TotalUnits,UnitsPerStory,Sliders,AreaOfProperty,RoofShape,RoofCover,RoofMembrane,Soffit,RoofToWallConnection,DeckAttachment,OpeningProtection,BuildingLayout
ABC100,1,1,1,33143,1981,2,10000000,250000,20000,0,500000,500000,2,R,Miami-Dade,123 Main Street,Miami,0,A,35, NA,8,40,5,1,21346,5,6,3,4,4,5,3,1
ABC100,2,1,1,34109,1981,2,8000000,250000,20000,0,450000,450000,2,R,Collier,-81.345593;26.017147,Naples,0,A,42,
NA,6,30,5,1,19464,5,6,3,4,4,5,3,1

Note the attributes should be separated by comma only.

6. ***Describe actions performed to ensure the validity of insurer data used for model inputs or validation/verification.***

A series of functions is executed to check and validate the data and to prepare it for processing. The checklist below outlines the initial tests that are performed. In addition the mitigation attributes are checked for valid, numeric entries, and are mapped to the code description. Pre-processing produces a summary report that identifies any major issues that require contacting the company.

Following pre-processing, a preliminary model run is performed in order to identify any inconsistencies between attributes, e.g. zip code and county. Any inconsistencies are resolved before the model is run and output produced.

Table 31. Checklist for the Pre-processing.

Note: LMs is coverage limit for building structure, LMapp is coverage limit for appurtenant structure, LMc is coverage limit for contents, and LMale is coverage limit for time element.

PolicyID	* There are no null values. * All duplicates (if any) have valid policy information.	
Zipcode	* There are no null values. * All values belong to the set of 5-digit zipcodes in Florida.	
YearBuilt	* There are no null values (Note: policies with no YearBuilt should have for value 0). * All values are 4-digit numbers. * There are no values exceeding the current year. * There are no non-zero values less than 1700.	
ConstType	* There are no null values. * All values are either *masonry*, *frame*, *manufactured*, or *other*.	
PropValue	* There are no null values. * There are no negative values. * If all values are equal to 0, then they are updated to equal LMs. * The actural Property Values will be updated to the larger numeric value between Property Value and Structure Limit	
LMs	* There are no null or non-numeric values. * There are no negative values.	
LMapp	* There are no null or non-numeric values. * There are no negative values. * If all values are equal to 0 (because it's missing but the company covers Lmapp), then they are updated to 10% of LMs. (double check with Dr. Hamid)	
LMc	* There are no null or non-numeric values. * There are no negative values.	
LMale	* There are no null or non-numeric values. * There are no negative values. * If all values are equal to 0, then the ALE limits will be updated in the program as follows: (1) 20% of LMs or (2) 40% of LMc if LMs is zero but LMc > 0 or (3) 40% of LMapp if both LMs and LMc are zero	
Deduc	* There are no null or non-numeric values. * There are no negative values. * All percentages are converted to numeric values. (Sometimes the percentages are represented as 2, 5, 10, 02, 05, 000002, 000005, 000010 instead of 2%, 5%, 10%)	
HurrDeduc	* There are no null or non-numeric values. * There are no negative values. * All percentages are converted to numeric values. (Sometimes the percentages are represented as 2, 5, 10, 02, 05, 000002, 000005, 000010 instead of 2%, 5%, 10%) * Normally Hurricane Deductible should be no less than 500.	
Coverage	* There are no null values. * The format is correct (i.e. value is equal to *A* or *R*).	
County	* There are no null values. * All county names are spelled only one way (i.e. all caps & no spelling errors, etc.). * All names are counties in Florida. * For counties as Miami-Dade (Miami Dade, Dade), St. Johns (Saint Johns, St Johns), St. Lucie (Saint Lucie, St Lucie), make sure only one type of spelling is used.	
PolicyForm	* If the field is present, values cannot be null. * The format is correct (i.e. value is equal to *DP-3*, *HO-6*, etc.).	
ProgramCode	* If the field is present, values cannot be null. * The format is correct (i.e. value is equal to *A*, *B*, etc.).	
TerritoryCode	* If the field is present, values cannot be null. * The format is correct (i.e. value is equal to *36*, *11*, etc.).	

Missing attribute values	If attributes of roof shape, roof cover, opening protection, roof to wall connection are all unknown use weighted matrix. If one or more of the attributes are known, use the values and replace the unknowns by randomly assigning values based on survey statistics and then use unweighted matrices.

A-2 Event Definition

A. ***Modeled loss costs and probable maximum loss levels shall reflect all insured wind related damages from storms that reach hurricane strength and produce minimum damaging windspeeds or greater on land in Florida.***

B. ***Time element loss costs shall reflect losses due to infrastructure damage caused by a hurricane.***

Disclosures

1. ***Describe how damage from model generated storms (landfalling and by-passing) is excluded or included in the calculation of loss costs and probable maximum loss levels for the state of Florida.***

Damages are computed for all Florida land-falling and certain by-passing storms in the stochastic set that attain hurricane level wind speeds. The following by-passing hurricanes are included:

-Non-land-falling hurricanes with point of closest approach in region A, B, C, D, E or F and open terrain winds greater than 30 mph in at least one Florida ZIP Code.

-Land-falling hurricanes in regions E or F with open terrain winds greater than 30 mph in at least one Florida ZIP Code.

2. ***Describe how damage resulting from concurrent or preceding flood or hurricane storm surge is treated in the calculation of loss costs and probable maximum loss levels for the state of Florida.***

Damage from concurrent or preceding flood or storm surge is not considered in the calculation of loss costs and probable maximum loss. The model assumes that wind is the only cause of loss from each hurricane.

A-3 Coverages

A. ***The methods used in the development of building loss costs shall be actuarially sound.***

B. ***The methods used in the development of appurtenant structure loss costs shall be actuarially sound.***

C. ***The methods used in the development of contents loss costs shall be actuarially sound.***

D. ***The methods used in the development of time element coverage loss costs shall be actuarially sound.***

Disclosures

1. ***Describe the methods used in the model to calculate loss costs for building coverage associated with personal and commercial residential properties.***

Personal Residential Buildings

The model includes a set of vulnerability matrices for personal residential buildings. The matrices specify the probability of damage of a given magnitude at various wind speeds. For each building in the policy portfolio the applicable matrix for that building is used to determine the expected percent damage at a given wind speed. This determination is made storm by storm for every storm in the stochastic set. The resulting damages, adjusted for policy limits, deductibles and demand surge, are aggregated across all storms to calculate the loss cost per $1,000 of exposure.

Commercial Residential Buildings

For low-rise commercial residential buildings (three stories or fewer) the model includes a set of vulnerability curves. The curves specify the expected damage rate by wind speed.

For mid-/high-rise commercial residential buildings (over three stories), the model estimates exterior damage to the building by aggregating expected damage per story and interior damage as a function of the volume of water intrusion resulting from breached openings on each story.

Similar to the approach applied to personal residential buildings, expected damages for commercial residential buildings are determined for each storm, adjusted for policy provisions and demand surge, and aggregated to calculate the loss cost per $1,000 of exposure.

2. ***Describe the methods used in the model to calculate loss costs for appurtenant structure coverage associated with personal and commercial residential properties.***

Expected damages for both personal residential and commercial residential appurtenant structures are determined by policy for each storm in the stochastic set, adjusted for policy provisions and demand surge, and aggregated across all storms to calculate the loss cost per $1,000 of exposure. Expected damages are determined as follows:

Personal Residential Appurtenant Structures

Since the appurtenant structures damage is not derived from the building damage, only one vulnerability matrix is applied for appurtenant structures. The typical insurance portfolio gives no indication of the type of appurtenant structure covered under a particular policy. Therefore, a distribution of the three types (slightly vulnerable, moderately vulnerable, and highly vulnerable) was assumed in developing this matrix, and the result was then validated against claim data.

Commercial Residential Appurtenant Structures

For commercial residential exposures, appurtenant structures might include a clubhouse or administration building. These are modeled like additional buildings. For other structures such as pools, the appurtenant structures vulnerability matrix developed for residential buildings is applied.

3. ***Describe the methods used in the model to calculate loss costs for contents coverage associated with personal and commercial residential properties.***

Expected damages for both personal residential and commercial residential contents coverage are determined for each storm in the stochastic set, adjusted for policy provisions and demand surge, and aggregated across all storms to calculate the loss cost per $1,000 of exposure. Expected damages are determined as follows:

Personal Residential Contents

Contents losses are a function of the internal damage. The model applies empirical functions that are based on engineering judgment and were validated against claim data for Hurricanes Andrew, Charley, and Frances. Figure 68 shows masonry claims data from Hurricane Andrew, the cubic polynomial trend fit, and the model curve for the High Velocity Hurricane Zone (HVHZ), which consists of Miami-Dade and Broward counties. Notice that in this case the fit between model and data is reasonable where the density of data is higher. A resulting set of vulnerability matrices are applied to determine expected percent contents damage for a given wind speed.

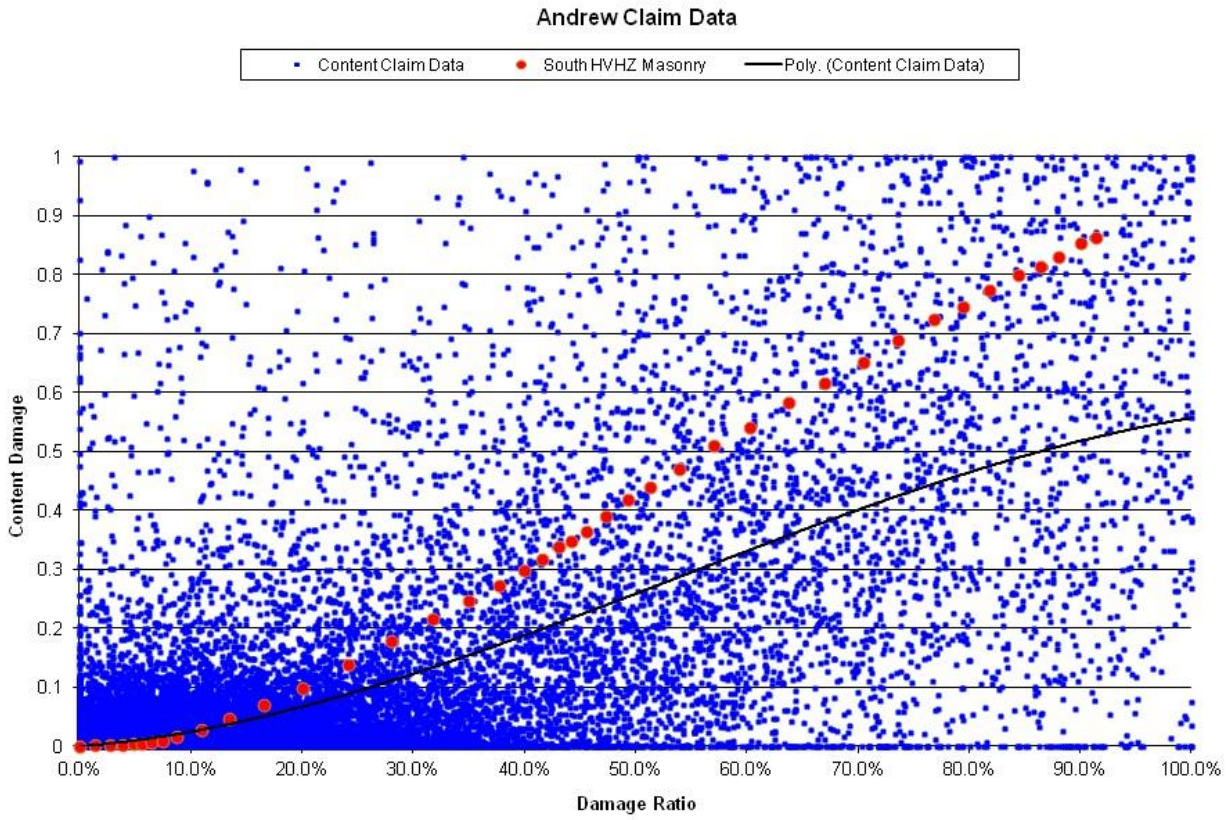


Figure 68. Modeled vs. actual relationship between structure and content damage ratios for Hurricane Andrew.

Commercial Residential Contents

Contents damage in low-rise buildings (three stories or fewer) is modeled as a proportion of interior damage. The resulting set of vulnerability curves vary by subregion and number of stories and specify expected percent damage by wind speed.

Contents damage in mid-/high-rise buildings (over three stories) is also determined as a proportion of total estimated interior damage to the building. The interior damage is estimated by determining the expected number of openings (windows, doors, sliding-glass doors) per story to be breached, and the resulting volume of water intrusion in each story.

The assumptions underlying contents damage development are based on engineering judgment.

4. ***Describe the methods used in the model to calculate loss costs for time element coverage associated with personal and commercial residential properties.***

Expected damages for both personal residential and commercial residential time element coverage are determined for each storm in the stochastic set, adjusted for policy provisions and demand surge, and aggregated across all storms to calculate the loss cost per $1,000 of exposure. Expected damages are determined as follows:

Personal Residential Time Element

Personal residential time element damages are based on an empirical function relating those damages to the interior damage to the structure. The model does not distinguish explicitly between direct and indirect loss to the structure, but the function is calibrated against claim data that include both types of losses. Vulnerability matrices specify the expected percent loss for a given wind speed.

Commercial Residential Time Element

The time element damages associated with low-rise buildings (three stories or fewer) are modeled using functions that relate those damages to interior damage to the building. The resulting set of vulnerability curves specify expected percent damage by wind speed.

Time element damages in mid-/high-rise buildings (over three stories) are not modeled.

A-4 Modeled Loss Cost and Probable Maximum Loss Considerations

A. ***Loss cost projections and probable maximum loss levels shall not include expenses, risk load, investment income, premium reserves, taxes, assessments, or profit margin.***

B. ***Loss cost projections and probable maximum loss levels shall not make a prospective provision for economic inflation.***

C. ***Loss cost projections and probable maximum loss levels shall not include any explicit provision for direct hurricane storm surge losses.***

D. ***Loss cost projections and probable maximum loss levels shall be capable of being calculated from exposures at a geocode (latitude-longitude) level of resolution.***

E. ***Demand surge shall be included in the model's calculation of loss costs and probably maximum loss levels using relevant data.***

F. ***The methods, data, and assumptions used in the estimation of demand surge shall be actuarially sound.***

Disclosures

1. ***Describe the method or methods used to estimate annual loss costs and probable maximum loss level. Identify any source documents used and research performed.***

To estimate annual loss costs and probable maximum loss levels, losses are estimated for individual policies in the portfolio for each hurricane in a stochastic set of storms. Losses are estimated separately for structure, appurtenant structure, contents, and time element coverage.

The meteorological component of the model generates the stochastic set of hurricanes and derives an expected three-second gust wind speed, by latitude and longitude, for each hurricane in that set of storms.

The engineering component of the model consists of a set of vulnerability matrices for personal residential exposures and a set of vulnerability curves for low-rise commercial residential exposures. The matrices specify the probability of damage of a given magnitude at various wind speeds. The curves specify the expected damage rate by wind speed. For mid-rise and high-rise commercial residential exposures, the model estimates exterior damage by aggregating expected damage per story and interior damage as a function of the volume of water intrusion resulting from breached openings on each story.

The estimated damages are reduced by applicable deductibles and increased to allow for the impact of demand surge on claim costs.

The modeled insured losses can then be summed across all properties in a ZIP Code or across all ZIP Codes in a county to obtain expected aggregate loss. The losses can also be aggregated by policy form, construction type, rating territories, etc.

Finally, modeled losses are divided by the number of years in the simulation and by the total amount of insurance to estimate annual loss costs.

Modeled losses for storms occurring in the same year of the simulation are summed to produce annual storm losses. Probable maximum loss levels are calculated from the ordered set of annual losses as described in Standard A-6, Disclosure # 11.

The following sources were used in the research:

Hogg, R. V., & Klugman, S. (1984). *Loss Distributions.* New York: Wiley.

Klugman, S., Panjer, H., & Willmot, G. (1998). *Loss Models: From Data to Decisions.* New York: Wiley.

Wilkinson, M. E. (1982). Estimating Probable Maximum Loss with Order Statistics. *Casualty Actuarial Society, LXIX*, pp. 195-209.

2. ***Identify the highest level of resolution for which loss costs and probable maximum loss levels can be provided. Identify all possible resolutions available for the reported output ranges.***

Losses are calculated at the policy/coverage level for each storm in the stochastic set. When the street address of the exposures is available for input, each policy is associated with a latitude and longitude. In that case loss costs and probable maximum loss levels can be provided by latitude and longitude.

Losses can be summarized across any policy characteristic provided in the exposures. Therefore, loss costs and probable maximum loss levels can be aggregated by characteristics such as policy form, coverage, construction, deductible, latitude-longitude, ZIP Code, county, rating territory, roof shape, or whatever is provided for input.

For the reported output ranges, the resolutions available are defined by the policy characteristics provided in the exposures, namely, policy form, ZIP Code, construction and deductible. ZIP Codes can be aggregated to the county, region, or statewide level.

3. ***Describe how the model incorporates demand surge in the calculation of loss costs and probable maximum loss levels.***

Demand surge factors by coverage are calculated for each storm in the stochastic set and are applied to the estimated losses for that storm. For each storm, demand surge is assumed to be a function of coverage, region, and the storm's estimated statewide losses before consideration of demand surge.

General Form of the Demand Surge Functions

The functions applied to determine the demand surge for each storm are of the form

Structure: Surge Factor = c + p1 x ln (statewide storm losses) + p2,

where c is a constant,
p1 is a constant for all regions except Monroe County,
p2 varies by region, and
"statewide storm losses" are the estimated losses, before demand surge, for the storm under consideration.

Appurtenant Structures: Surge Factor = Structure Factor.

Contents: Surge Factor = [(Structure Factor – 1) x 30%] + 1.

Additional Living Expenses: Surge Factor = 1.5 x Structure Factor - .5.

Development of the Demand Surge Function for Structure

To estimate the impact of demand surge on the settlement cost of structural claims following a hurricane we used a quarterly construction cost index produced by Marshall & Swift/Boeckh. We considered the history of the index from first quarter 1992 through second quarter 2007. There is an index for each of 52 ZIP Codes in Florida representing 42 counties. We grouped the indices to produce a set of regional indices, weighting each ZIP Code index with population.

The approach to estimating structural demand surge was to examine the index for specific regions impacted by one or more hurricanes since 1992. From the history of the index, we projected what the index would have been in the period following the storm had no storm occurred. Any gap between the predicted and actual index was assumed to be due to demand surge. In total we examined ten storm–region combinations. From these ten observations of structural demand surge, we generalized to the functional relationship shown above.

Monroe County was treated as an exception. There were no storms of any severity striking Monroe during the period of our observations. We believe, though, that the location of and limited access to the Keys will result in an unusually high surge in reconstruction costs after a storm, particularly since the Overseas Highway could be damaged by storm surge or seriously blocked by debris. We have therefore judgmentally selected demand surge parameters for Monroe in excess of those indicated for the remainder of South Florida.

Development of the Contents Demand Surge Function

The approach to determining the contents demand surge function was to relate any surge in consumer prices in Southeast Florida following hurricanes Katrina and Wilma to the estimated structure demand surge following those storms. We used a sub-index of the Miami-Ft. Lauderdale Consumer Price Index for this purpose and compared the projected and actual indices after the storms. Since the surge in consumer prices was roughly 30% of the surge in construction costs, we selected that percentage as the relationship between structure and contents demand surge.

Development of Time Element (TE) Demand Surge Function

To estimate TE demand surge we first examined the relationship between structure losses and TE losses in the validation dataset. This dataset includes losses from three storms (Andrew, Charley, and Frances) and eleven insurance companies. We then compared the predicted increase in TE losses associated with various increases in structure losses. That generalized relationship is the TE demand surge function shown above.

TE demand surge is related to structure demand surge in the following sense: structure surge is caused by an inability of the local construction industry to meet the sudden demand for materials and labor following a storm. A high surge in construction costs suggests a more serious mismatch between the demand for repairs and the supply of materials and labor. This mismatch translates into longer delays in the completion of repairs and rebuilding, which in turn implies a higher surge in TE costs.

Because the model's TE surge is determined as a function of structure surge, Monroe County TE surge factors are higher than those for the remainder of South Florida. We believe this is reasonable because of the unusual delays in repair and rebuilding that are likely to occur following a major storm in the Keys, especially if there is damage to US 1 or to bridges connecting the islands.

Treatment of Demand Surge for Storms Impacting both the Florida Panhandle and Alabama

The Northwest region is segregated from the remainder of the North to allow for demand surge that is a function of combined Florida–Alabama losses from storms impacting both states. The Northwest region consists of all Panhandle counties west of Leon and Wakulla. The definition of this region was selected by considering which counties experienced losses from Hurricanes Ivan, Frederic, and Elena, i.e., from storms that impacted both states. Not all counties in the Northwest region experienced losses from these three specific storms, but losses in neighboring counties suggest that that they are nevertheless at risk for inclusion in a combined Florida–Alabama event.

Demand surge factors for the Northwest region are determined as an upward adjustment to the factors for the Northeast–North Central region. The purpose of this adjustment is to correct for an understatement of the model's demand surge that occurs when only the Florida losses from a combined Florida–Alabama event are used to determine the level of demand surge from a storm.

4. ***Provide citations to published papers, if any, that were used to develop how the model estimates demand surge.***

No published papers were used in the demand surge development.

A-5 Policy Conditions

A. ***The methods used in the development of mathematical distributions to reflect the effects of deductibles and policy limits shall be actuarially sounds.***

B. ***The relationship among the modeled deductible loss costs shall be reasonable.***

C. ***Deductible loss costs shall be calculated in accordance with s.627.701(5)(a), F.S.***

Disclosures

1. ***Describe the methods used in the model to treat deductibles (both flat and percentage), policy limits, replacement costs, and insurance-to-value when projecting loss costs.***

In practice the insurance companies often allocate deductibles to structure, content, AP, and ALE on a pro-rata loss basis. Thus, if for example, structure and content damages before deductible are $20,000 and $6,000 respectively, and the deductible is $3,000, then (20,000/26,000)(3,000) = $2,308 is allocated to structure and (6,000/26,000)(3,000) = $692 is allocated to contents. This means that the various damages have to be considered and deductibles applied simultaneously. The deductibles must be allocated among the different losses and the truncation applied to each loss separately on a pro-rata basis.

For the pro-rata deductible method to work optimally, the functional relationships between structure damage and others should be estimated, and for each interval or class of structural damage, the corresponding mean and variance of the C, AP, and ALE damages should be specified. The conditional probabilities for C, AP, and ALE will then be the same as those for structural damage. An independent content matrix is somewhat problematic and may create biases in estimates of net of deductible losses. For structures we are likely to have damage ratio ranges or intervals of 0 to 2%, 2% to 4%, 4% to 6%, etc. For each interval (and its midpoint), ideally we may want to use the mean and variance of the corresponding damage ratios for contents, AP, and ALE. In practice, since the damage matrix for different types of losses are not directly related, we need to use the mean of the content, or AP, or ALE damage vector conditional on windspeeds since the windspeed is the only common frame of reference to the various types of damages.

$$\text{Expected Structure Loss} = E(L_S) = \sum_{D_S}^{L+D_S} (DM_i - D_S)\, p_S(x_i w) + \sum LM_S\, p_S(x_i w)$$

$$\text{Expected Content Loss} = E(L_C) = \sum_{C_S}^{L+C_S} (f(X_i) - D_C)\, p_C(x_i w) + \sum LM_C\, p_C(x_i w)$$

$$\text{Expected Appurtenant Loss} = E(L_{AP}) = \sum (g(X_i) - D_{AP})\, p_S(x_i w) + \sum LM_{AP}\, p_S(x_i w)$$

$$\text{Expected ALE Loss} = E(L_{ALE}) = \sum (h(X_i) - D_{ALE})\, p_S(x_i w) + \sum LM_{ALE}\, p_S(x_i w)$$

$$\text{Expected Loss} = E(L) = E(L_S) + E(L_C) + E(L_{AP}) + E(L_{ALE})$$

where each of the losses net of deductible is ≥ 0 and where the deductibles DS, DC, DAP, DALE are applied on a pro-rata basis to the respective damages as follows:

$D_S = [DM_S /(DM_S + C + AP + ALE)] * D$
$D_C = [C /(DM_S + C + AP + ALE)] * D$
$D_{AP} = [AP /(DM_S + C + AP + ALE)] * D$
$D_{ALE} = [ALE /(DM_S + C + AP + ALE)] * D$

For this method to work, ideally, the joint probabilities of the losses must be estimated and used. In practice such joint probabilities are hard to estimate and validate. Thus, the engineering component should ideally provide for each structural damage interval, and given a wind speed, the mean and variance of damage ratio for content, AP, and ALE. The model uses the mean C, AP, and ALE for the given wind speed to determine the allocation of deductible to the various coverages.

Personal Residential

In the damage matrices, each wind speed interval is associated with a distribution of possible damage ratios. Each damage ratio is multiplied by insured value to determine dollar damages, the deductible is deducted, and net of deductible loss is estimated, subject to the constraints that net loss is ≥ 0 and ≤ limit – deductible.

Commercial Residential

The deductible is deducted from expected loss for each building.

Personal and Commercial Residential

The deductible is allocated to coverage by first calculating expected losses for each coverage, assuming zero deductible, and then allocating the deductible to coverage based on those losses.

Percentage deductibles are converted into dollar amounts.

Both the replacement cost and property value are assumed to equal the coverage limit unless the property value is provided as an input.

2. ***Provide an example of how insurer loss (loss net of deductibles) is calculated. Discuss data or documentation used to confirm or validate the method used by the model.***

Personal Residential

For each damage ratio:
Loss net of deductible = (Damage Ratio x Bldg Value) – Deductible, but not less than zero or greater than limit – deductible.

Example
Bldg value = $200,000. Limit = $180,000. Deductible = $3,000. J^{th} Damage ratio = 5%.

Loss net of deductible = .05 x 200,000 - 3,000 = $7,000. If the J^{th} Damage ratio = 1%, then loss net of deductible = 0. If the damage ratio is 95%, then the loss net of deductible = $180,000 - $3,000 = $177,000.

The deductible method used by model is based on Hogg and Klugman (1984). Modeled losses net of deductible were validated against insurance company losses for Hurricanes Andrew, Charley, and Frances.

Commercial Residential

The deductible is deducted from the expected damage for each building.

Example
Building Limit = $1,000,000. Deductible = 3 % or $30,000. Expected Damage Ratio = 10%.
Expected Damage = $1,000,000 x 10% = $100,000.
Loss net of deductible = $100,000 - $30,000 = $70,000.

3. ***Describe how the model calculates annual deductibles.***

If there are multiple Hurricanes in a year in the stochastic set, the wind deductibles are applied to the first hurricane, and any remaining amount is then applied to the second hurricane. If none of the wind deductible remains, then the general peril deductible is applied. This is the case for both personal and commercial residential policies.

A-6 Loss Output

A. ***The methods, data, and assumptions used in the estimation of probable maximum loss levels shall be actuarially sound.***

B. ***Loss costs shall not exhibit an illogical relation to risk, nor shall loss costs exhibit a significant change when the underlying risk does not change significantly.***

C. ***Loss costs produced by the model shall be positive and non-zero for all valid Florida ZIP Codes.***

D. ***Loss costs cannot increase as the quality of construction type, materials and workmanship increases, all other factors held constant.***

E. ***Loss costs cannot increase as the presence of fixtures or construction techniques designed for hazard mitigation increases, all other factors held constant.***

F. ***Loss costs cannot increase as the quality of building codes and enforcement increases, all other factors held constant.***

G. ***Loss costs shall decrease as deductibles increase, all other factors held constant.***

H. ***The relationship of loss costs for individual coverages, (e.g., buildings and appurtenant structures, contents, and time element shall be consistent with the coverages provided.***

I. ***Output ranges shall be logical for the type of risk being modeled and deviations supported.***

J. ***All other factors held constant, output ranges produced by the model shall in general reflect lower loss costs for:***

   A. masonry construction versus frame construction,

   B. personal residential risk exposure versus mobile home risk exposure,

   C. inland counties versus coastal counties, and

   D. northern counties versus southern counties.

K. ***For loss cost and probable maximum loss level estimates derived from or validated with historical insured hurricane losses, the assumptions in the derivations concerning (1) construction characteristics, (2) policy provisions, (3) coinsurance, (4) contractual provisions, and (5) relevant underwriting***

practices underlying those losses, as well as any actuarial modifications, shall be appropriate based on the type of risk being modeled.

Disclosures

1. ***Provide a completed Form A-1, Zero Deductible Personal Residential Loss Costs by ZIP Code. Provide a link to the location of the form here.***

See Form A-1

2. ***Provide a completed Form A-2, Base Hurricane Storm Set Statewide Losses . Provide a link to the location of the form here.***

See Form A-2

3. ***Provide a completed Form A-3A, 2004 Hurricane Season Losses, using the 2007 Florida Hurricane Catastrophe Fund aggregate personal and commercial residential exposure data. Provide a link to the location of the form here.***

See Form A-3A

4. ***Provide a completed Form A-3B, 2004 Hurricane Season Losses, using the 2012 Florida Hurricane Catastrophe Fund aggregate personal and commercial residential exposure data. Provide a link to the location of the form here.***

See Form A-3B

5. ***Provide a completed Form A-4A, Output Ranges, using the 2007 Florida Hurricane Catastrophe Fund aggregate personal and commercial residential exposure data. Provide a link to the location of the form here.***

See Form A-4A

6. ***Provide a completed Form A-4B, Output Ranges, using the 2012 Florida Hurricane Catastrophe Fund aggregate personal and commercial residential exposure data. Provide a link to the location of the form here.***

See Form A-4B

7. ***Provide a completed Form A-5, Percentage Change in Output Ranges, using the 2007 Florida Hurricane Catastrophe Fund aggregate personal and commercial residential exposure data. Provide a link to the location of the form here.***

See Form A-5

8. ***A completed Form A-6, Logical Relationship to Risk (Trade Secret item) shall be provided during the closed meeting portion of the Commission meeting to review the model for***

acceptability.

See Form A-6

9. ***Provide a completed Form A-7, Percentage Change in Logical Relationship to Risk. Provide a link to the location of the form here.***

See Form A-7

10. ***Provide a completed Form A-8, Probable Maximum Loss for Florida. Provide a link to the location of the form here.***

See Form A-8

11. ***Describe how the model produces probable maximum loss levels.***

Probable maximum loss is produced nonparametrically using order statistics of simulated annual losses.

The model produces N simulated annual losses, represented by $X_1, X_2, \ldots, X_N$. The data are ordered so that $X_{(1)} \leq X_{(2)} \leq \ldots \leq X_{(N)}$.

For a return period of Y years, let p = 1-1/Y. The corresponding PML for the return period Y is the pth quantile of the ordered losses.

Let k = (N)*p. If k is an integer, then the estimate of the PML is the kth order statistic, $X_{(k)}$, of the simulated losses. If k is not an integer, then let k* = the smallest integer greater than k, and the estimate of the pth quantile is given by $X_{(k^*)}$.

12. ***Provide citations to published papers, if any, that were used to estimate probable maximum loss levels.***

Wilkinson, M. E. (1982). Estimating Probable Maximum Loss with Order Statistics. *Casualty Actuarial Society, LXIX*, pp. 195-209.

13. ***Describe how the probable maximum loss levels produced by the model include the effects of personal and commercial residential insurance coverage.***

The model can produce probable maximum loss levels separately for personal and commercial residential exposures or on a combined basis. To produce the probable maximum loss on a combined basis, modeled losses for both personal and commercial exposures are aggregated for each year in the simulation before the years are ordered. Because modeled losses are used as the basis for the probable maximum loss level, the effects of policy limits, deductibles, etc. are reflected in the probable maximum loss estimates.

14. ***Explain any difference between the values provided on Form A-8 (Probable Maximum Loss for Florida) and those provided on Form S-2B (Examples of Loss Exceedance Estimates,***

2012 FHCF Exposure Data).

The values on Form A-8 and Form S-2 are the same.

15. ***Provide an explanation for all anomalies in the loss costs that are not consistent with the requirements of this standard.***

Form A-4: In Form A-4 the county weighted average loss cost for masonry sometimes exceeds frame because the masonry weights are greater in ZIP Codes with higher loss costs.

Form A-6: There are anomalies in the Building Code and Building Strength tests in Form A-6. The anomalies are the result of the following model assumptions:

- The model assumes no difference in structure strength between the 1998 and 2004 Building Codes in the HVHZ.

- The model assumes no difference in structure strength between 1974 and 1992 Mobile Homes and does vary damages based on tie-downs.

- The model assumes no difference in structure strength between the 1980 and 1998 Building Codes as they apply to Commercial Residential construction, except in the HVHZ where metal shutters were required after 1994.

16. ***Provide an explanation of the differences in output ranges between the previously accepted submission and the current submission.***

The meteorology component of the model changed as described in Standard G-1, affecting both personal and commercial residential loss costs. In addition there were changes to the vulnerability component for commercial residential. The overall impact on personal residential loss costs was a reduction of roughly -5%, with changes at the county level ranging between -14% and +10%. The overall impact on commercial residential loss costs was a reduction of -8%, but the vulnerability changes resulted in significant loss cost increases, well over 100%, for a number of the smaller counties.

17. ***Identify the assumptions used to account for the effects of coinsurance on commercial residential loss costs.***

The model assumes properties are insured to value and makes no adjustment to losses for coinsurance penalties.

18. ***Describe how loss adjustment expenses are considered within the loss cost and probable maximum loss level estimates.***

No provision for loss adjustment expense is included in the loss cost or probable maximum loss level estimates.

Form A-1: Zero Deductible Personal Residential Loss Costs by ZIP Code

A. *Provide three maps, color-coded by ZIP Code (with a minimum of 6 value ranges), displaying zero deductible personal residential loss costs per $1,000 of exposure for frame, masonry, and mobile home.*

B. *Create exposure sets for these exhibits by modeling all of the buildings from Notional Set 3 described in the file "NotionalInput13.xlsx" geocoded to each ZIP Code centroid in the state, as provided in the model. Provide the predominant County name and the Federal Information Processing Standards (FIPS) Code associated with each ZIP code centroid. Refer to the Notional Policy Specification below for additional modeling information. Explain any assumptions, deviations, and differences from the prescribed exposure information.*

C. *Provide, in the format given in the file named "2013FormA1.xlsx," the underlying loss cost data rounded to 3 decimal places used for A. above in both Excel and PDF format. The file name shall include the abbreviated name of the modeling organization, the standards year, and the form name.*

See Appendix B.

Form A-2: Base Hurricane Storm Set Statewide Losses

A. Provide the total insured loss and the dollar contribution to the average annual loss assuming zero deductible policies for individual historical hurricanes using the Florida Hurricane Catastrophe Fund's aggregate personal and commercial residential exposure data found in the files named "hlpm2007c.exe" and "hlpm2012c.exe." The list of hurricanes in this form should include all Florida and by-passing hurricanes in the modeling organization Base Hurricane Storm Set, as defined in Standard M-1 (Base Hurricane Storm Set).

The table below contains the minimum number of hurricanes from HURDAT2 to be included in the Base Hurricane Storm Set, based on the 113-year period 1900-2012. Each hurricane has been assigned an ID number. As defined in Standard M-1 (Base Hurricane Storm Set) the Base Hurricane Storm Set for the modeling organization may exclude hurricanes that had zero modeled impact, or it may include additional hurricanes when there is clear justification for the changes. For hurricanes in the table below resulting in zero loss, the table entry should be left blank. Additional hurricanes included in the model's Base Hurricane Storm Set shall be added to the table below in order of year and assigned an intermediate ID number as the hurricane falls within the bounding ID numbers.

B. Provide this form in Excel format. The file name shall include the abbreviated name of the modeling organization, the standards year, and the form name. Form A-2 Base Hurricane Storm Set Statewide Losses) shall also be included in a submission appendix.

See Appendix C.

Form A-3A: 2004 Hurricane Season Losses (2007 FHCF Exposure Data)

A. Provide the percentage of residential zero deductible losses, rounded to four decimal places, and the monetary contribution from Hurricane Charley (2004), Hurricane Frances (2004), Hurricane Ivan (2004), and Hurricane Jeanne (2004) for each affected ZIP Code, individually and in total. Include all ZIP Codes where losses are equal to or greater than $500,000.

Use the 2007 Florida Hurricane Catastrophe Fund's aggregate personal and commercial residential exposure data found in the file named "hlpm2007c.exe."

Rather than using directly a specified published windfield, the winds underlying the loss cost calculations must be produced by the model being evaluated and should be the same hurricane parameters as used in completing Form A-2 (Base Hurricane Storm Set Statewide Losses).

B. Provide maps color-coded by ZIP Code depicting the percentage of total residential losses from each hurricane, Hurricane Charley (2004), Hurricane Frances (2004), Hurricane Ivan (2004), and Hurricane Jeanne (2004) and for the cumulative losses using the following interval coding:

Red	Over 5%
Light Red	2% to 5%
Pink	1% to 2%
Light Pink	0.5% to 1%
Light Blue	0.2% to 0.5%
Medium Blue	0.1% to 0.2%
Blue	Below 0.1%

The relevant storm track should be plotted on each map.

C. Provide this form in Excel format. The file name shall include the abbreviated name of the modeling organization, the standards year, and the form name. Form A-3A (2004 Hurricane Season Losses, 2007 FHCF Exposure Data) shall also be included in a submission appendix.

See Appendix D.

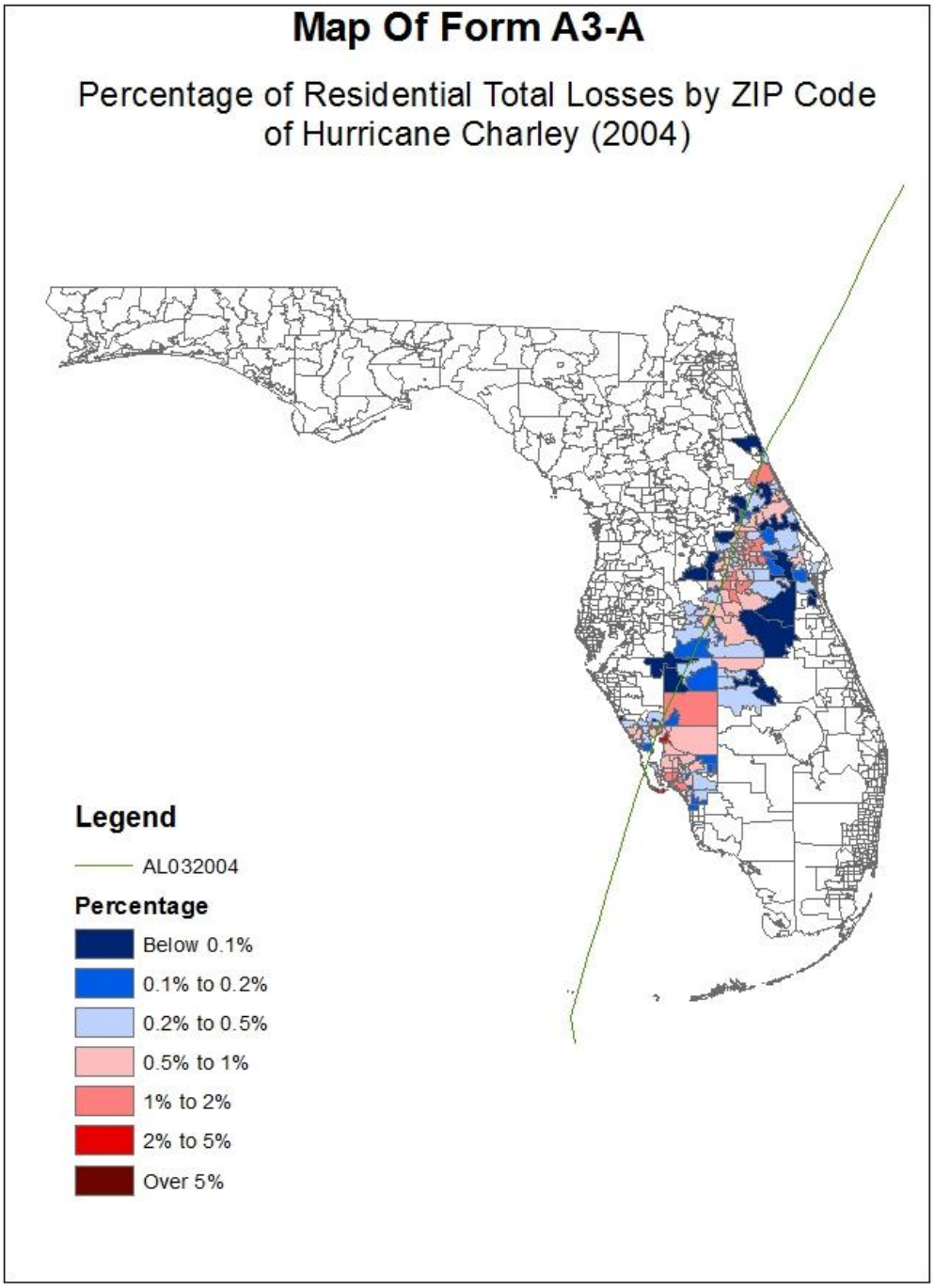


Figure 69. Percentage of residential total losses by ZIP code of Hurricane Charley (2004).

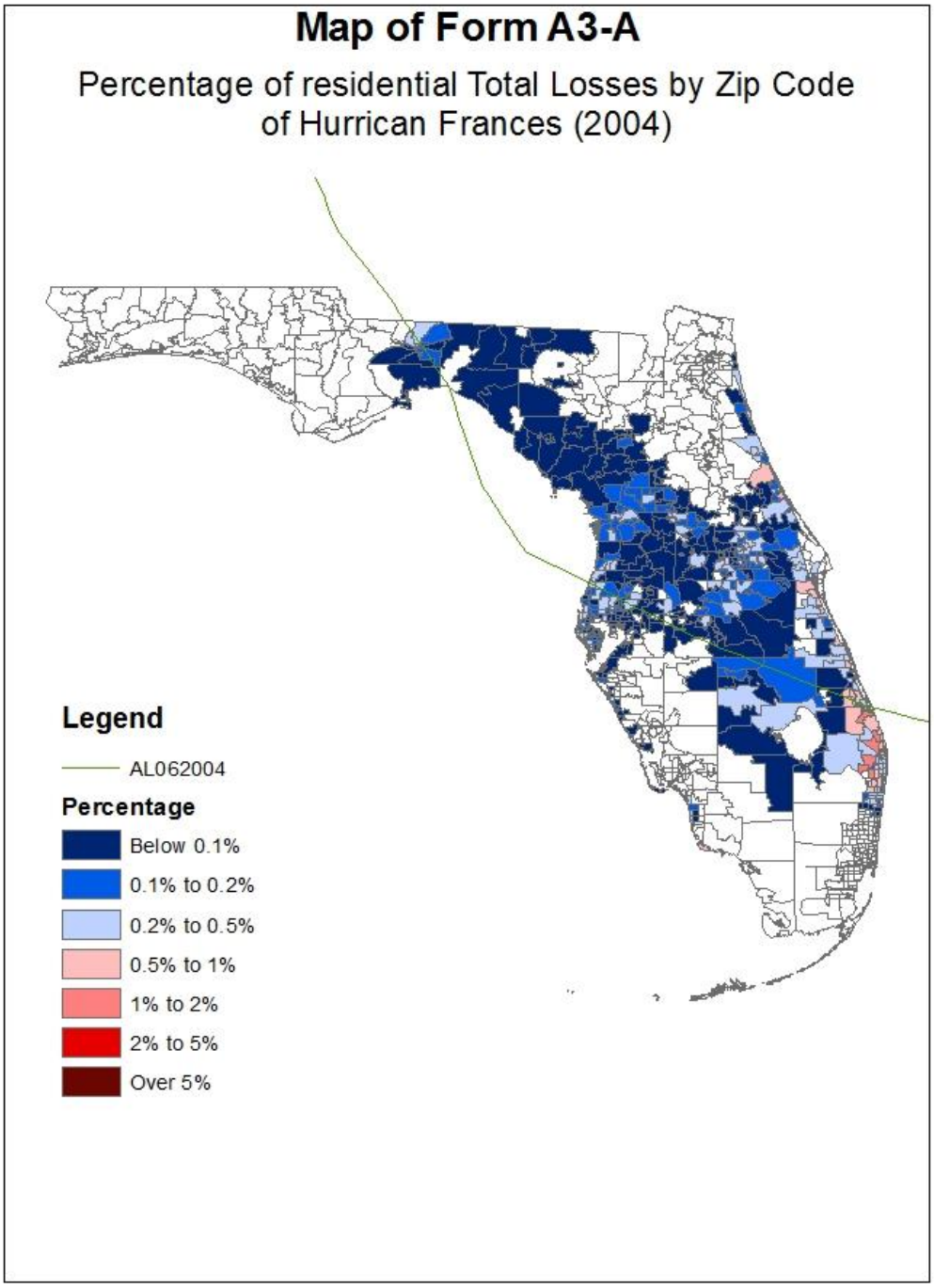


Figure 70. Percentage of residential total losses by ZIP code of Hurricane Frances (2004).

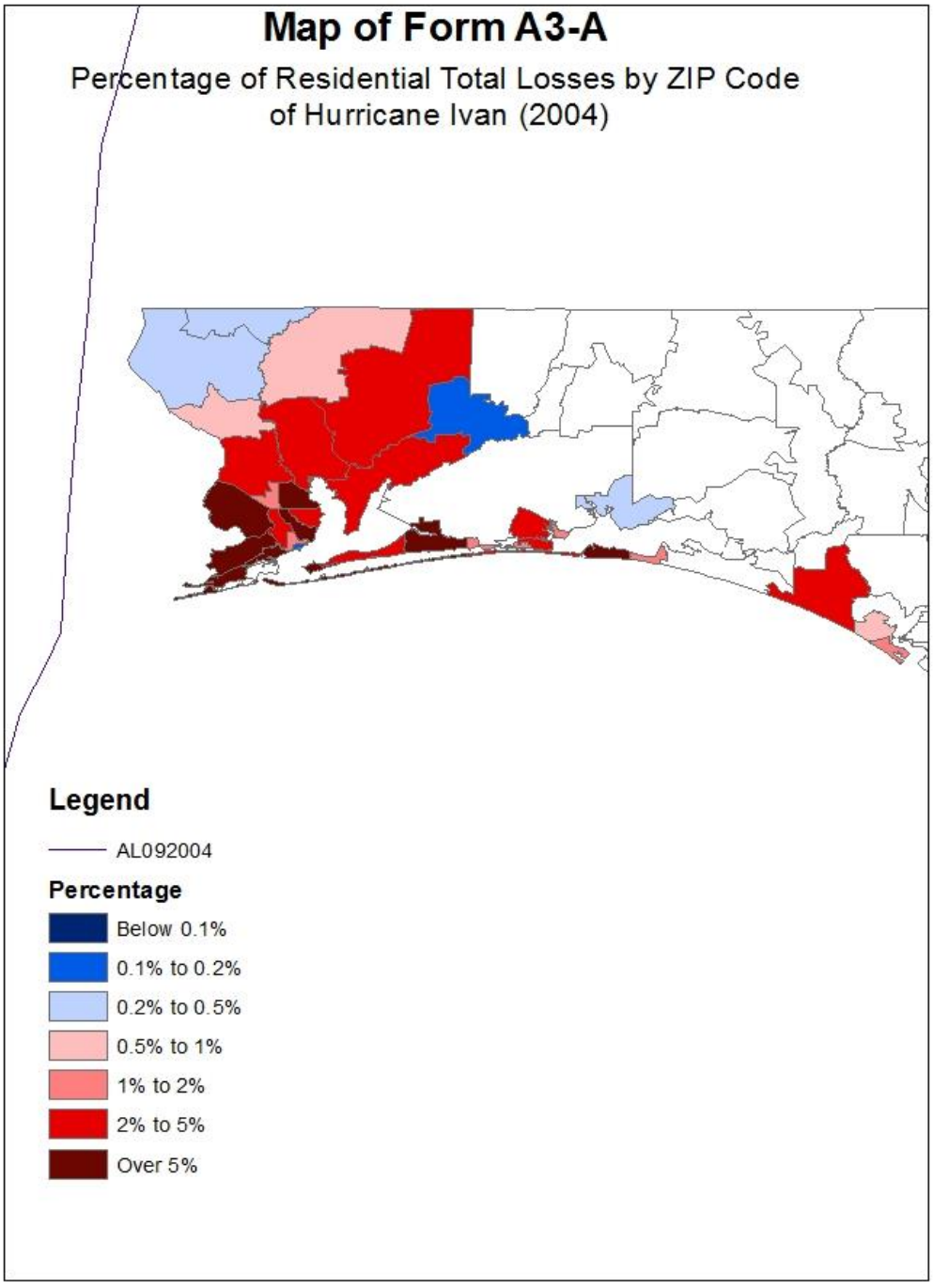


Figure 71. Percentage of residential total losses by ZIP code of Hurricane Ivan (2004).

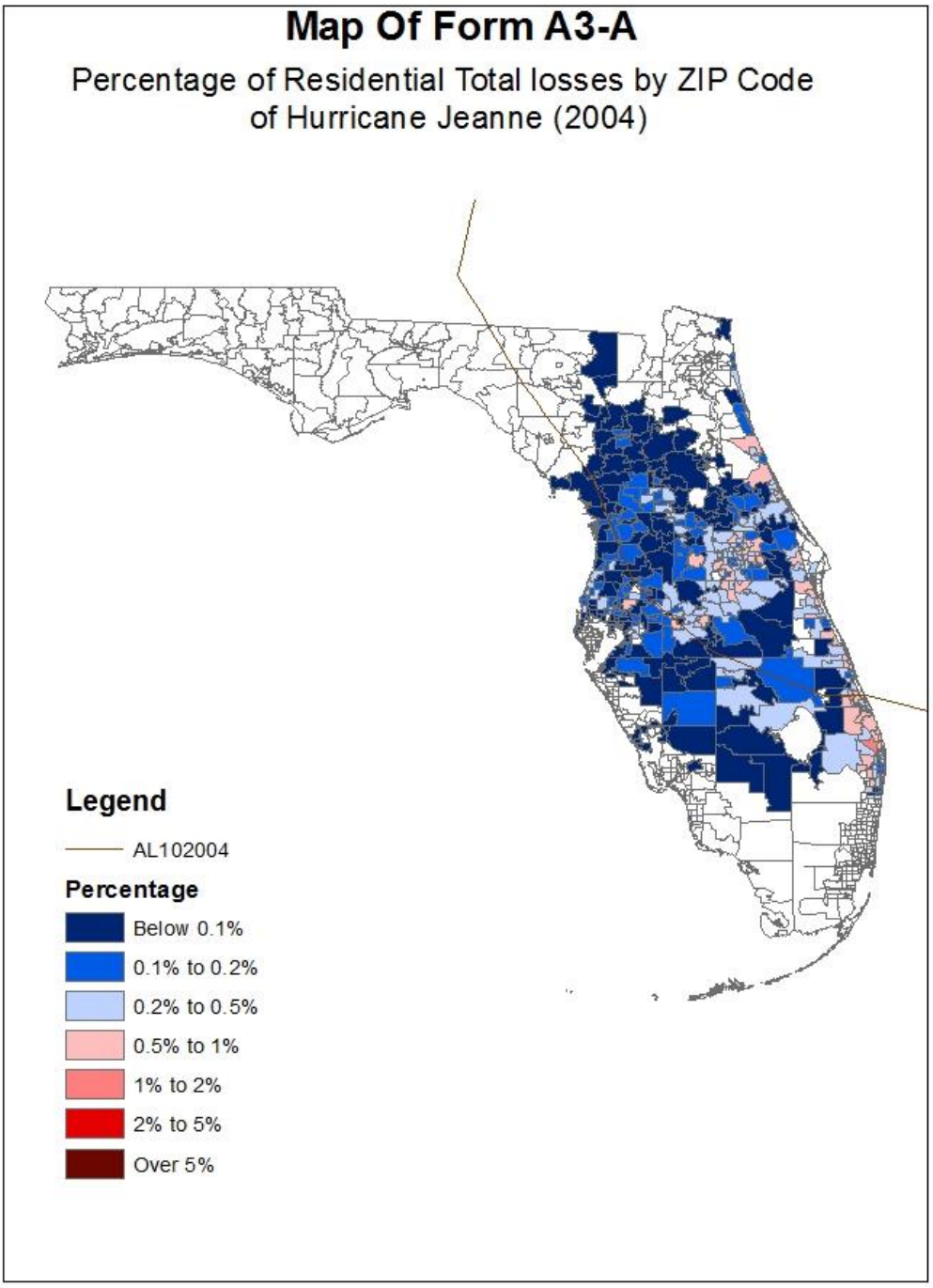


Figure 72. Percentage of residential total losses by ZIP code of Hurricane Jeanne (2004).

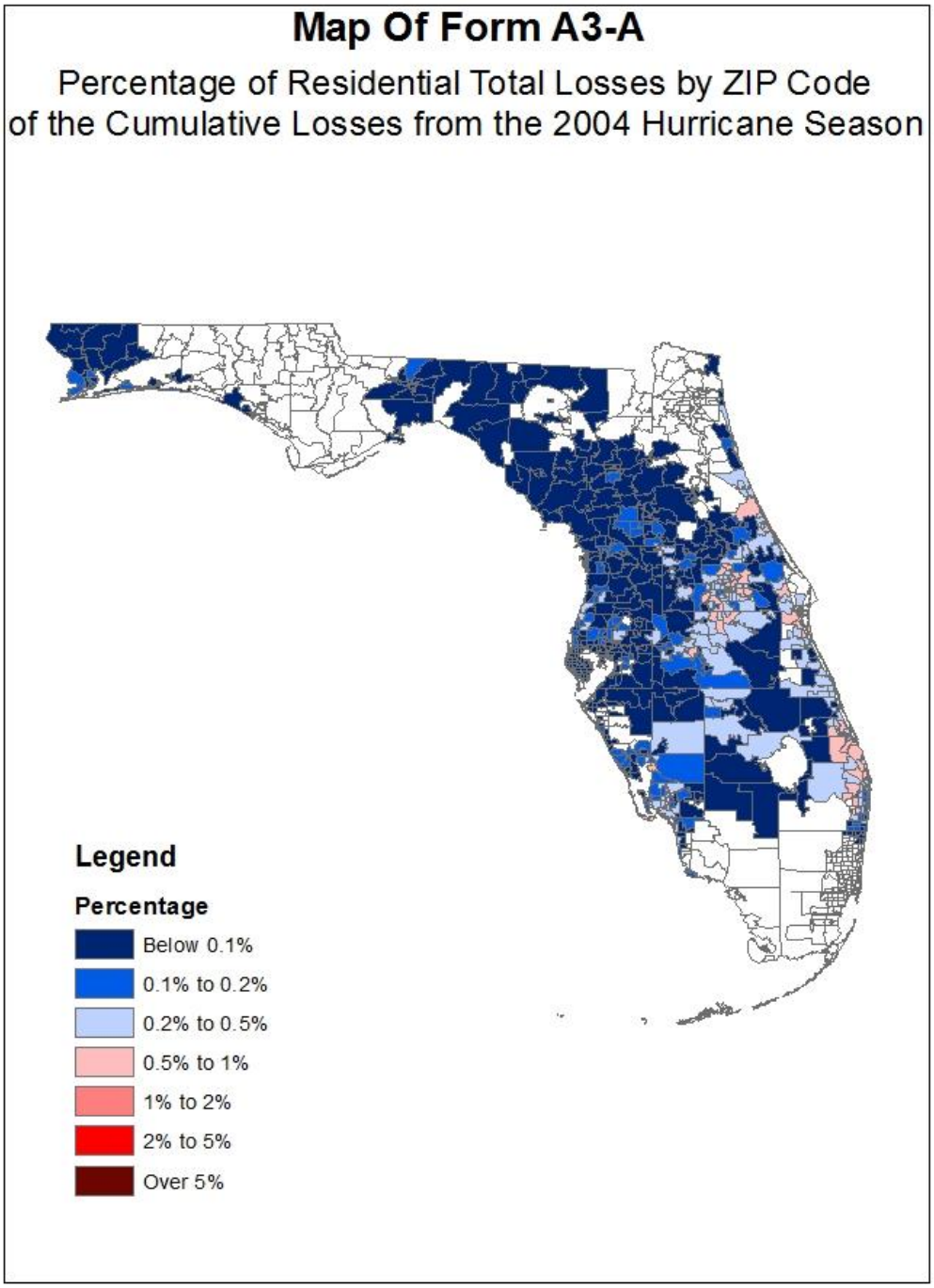


Figure 73. Percentage of residential total losses by ZIP code of the cumulative losses from the 2004 Hurricane Season.

Form A-3B: 2004 Hurricane Season Losses (2012 FHCF Exposure Data)

A. *Provide the percentage of residential zero deductible losses, rounded to four decimal places, and the monetary contribution from Hurricane Charley (2004), Hurricane Frances (2004), Hurricane Ivan (2004), and Hurricane Jeanne (2004) for each affected ZIP Code, individually and in total. Include all ZIP Codes where losses are equal to or greater than $500,000.*

   Use the 2012 Florida Hurricane Catastrophe Fund's aggregate personal and commercial residential exposure data found in the file named "hlpm2012c.exe."

   Rather than using directly a specified published windfield, the winds underlying the loss cost calculations must be produced by the model being evaluated and should be the same hurricane parameters as used in completing Form A-2 (Base Hurricane Storm Set Statewide Losses).

B. *Provide maps color-coded by ZIP Code depicting the percentage of total residential losses from each hurricane, Hurricane Charley (2004), Hurricane Frances (2004), Hurricane Ivan (2004), and Hurricane Jeanne (2004) and for the cumulative losses using the following interval coding:*

Red	*Over 5%*
Light Red	*2% to 5%*
Pink	*1% to 2%*
Light Pink	*0.5% to 1%*
Light Blue	*0.2% to 0.5%*
Medium Blue	*0.1% to 0.2%*
Blue	*Below 0.1%*

   The relevant storm track should be plotted on each map.

C. *Provide this form in Excel format. The file name shall include the abbreviated name of the modeling organization, the standards year, and the form name. Form A-3B (2004 Hurricane Season Losses, 2012 FHCF Exposure Data) shall also be included in a submission appendix.*

See Appendix D.

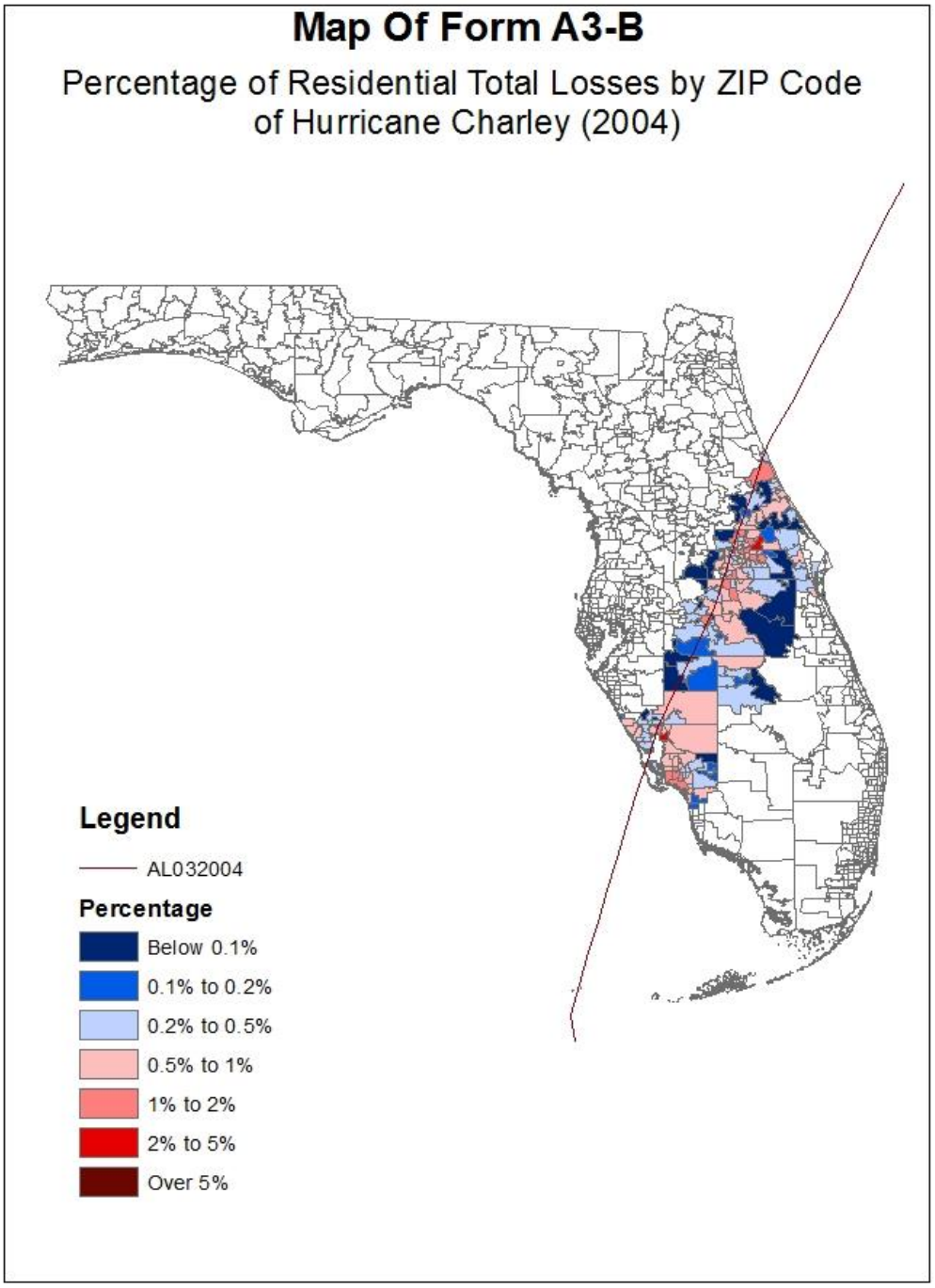


Figure 74. Percentage of residential total losses by ZIP code of Hurricane Charley (2004).

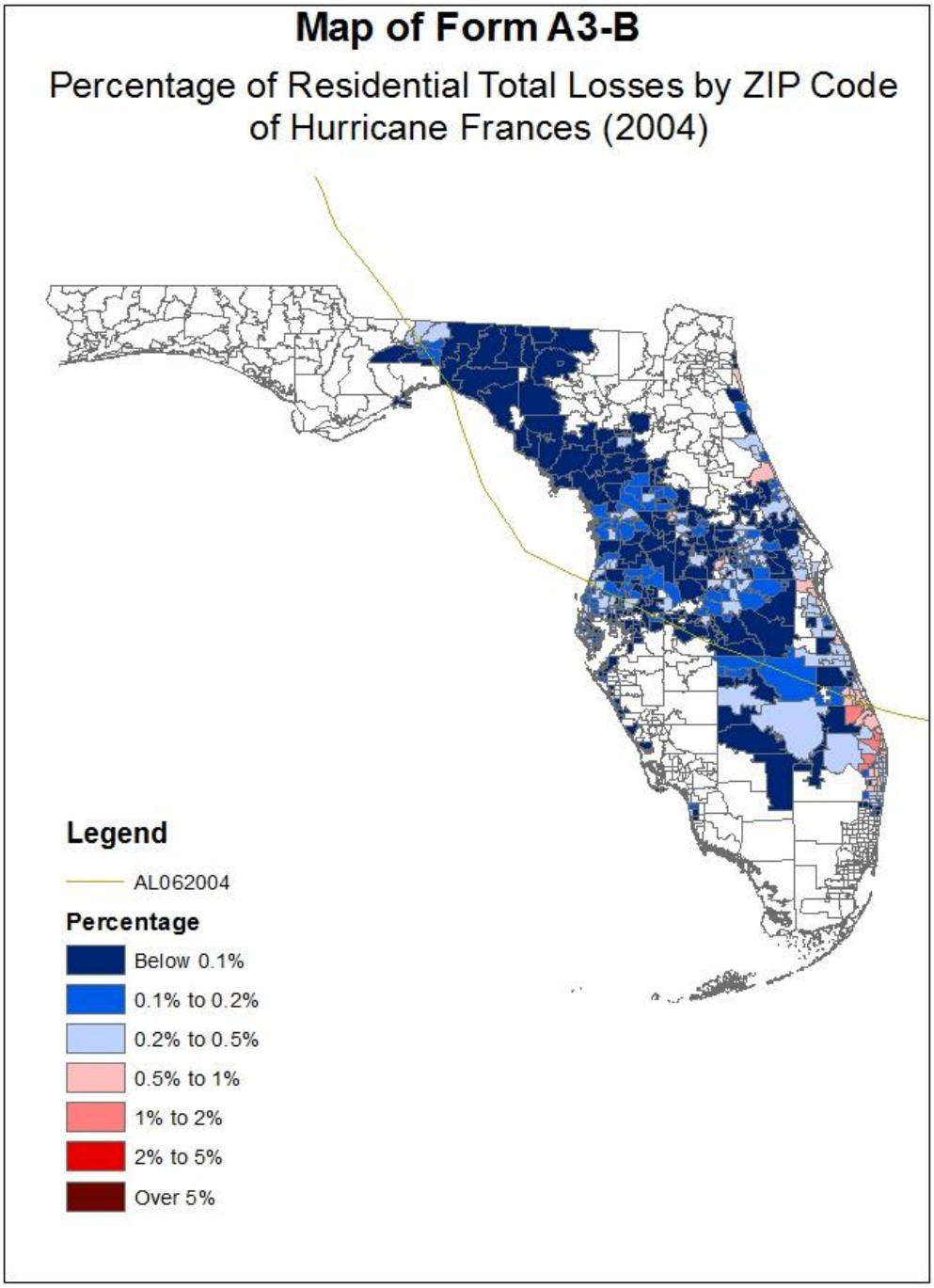


Figure 75. Percentage of residential total losses by ZIP code of Hurricane Frances (2004).

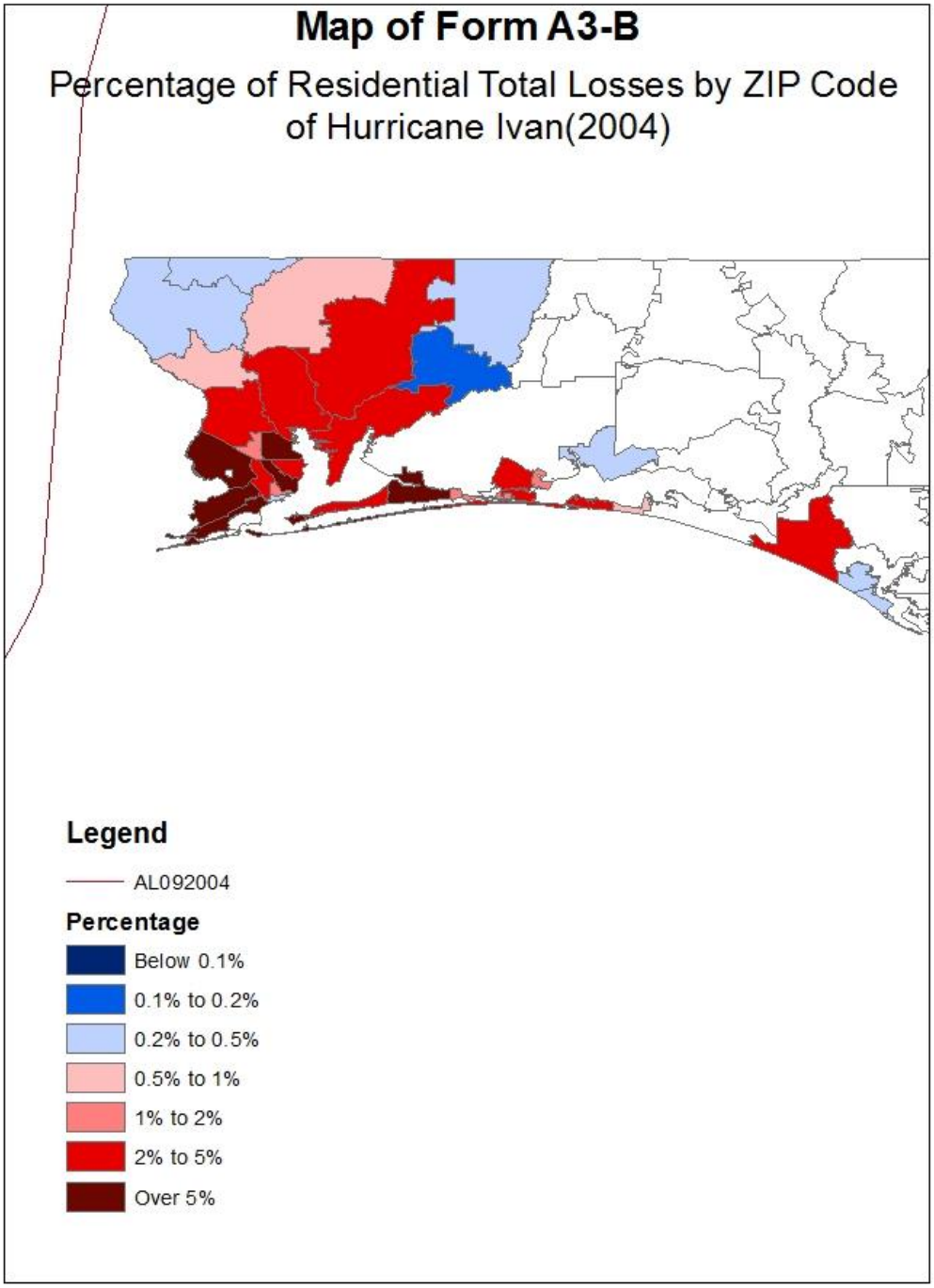


Figure 76. Percentage of residential total losses by ZIP code of Hurricane Ivan (2004).

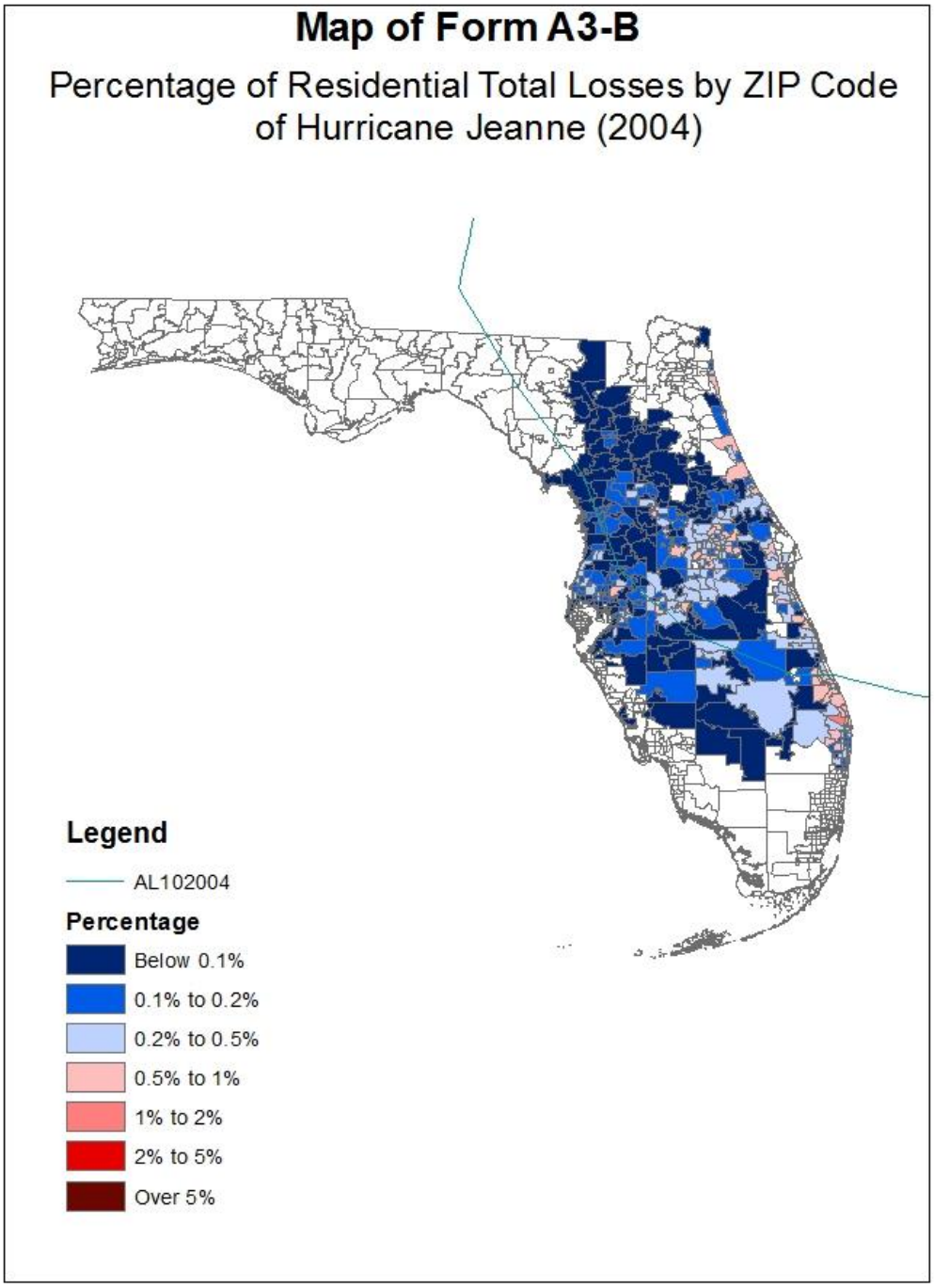


Figure 77. Percentage of residential total losses by ZIP code of Hurricane Jeanne (2004).

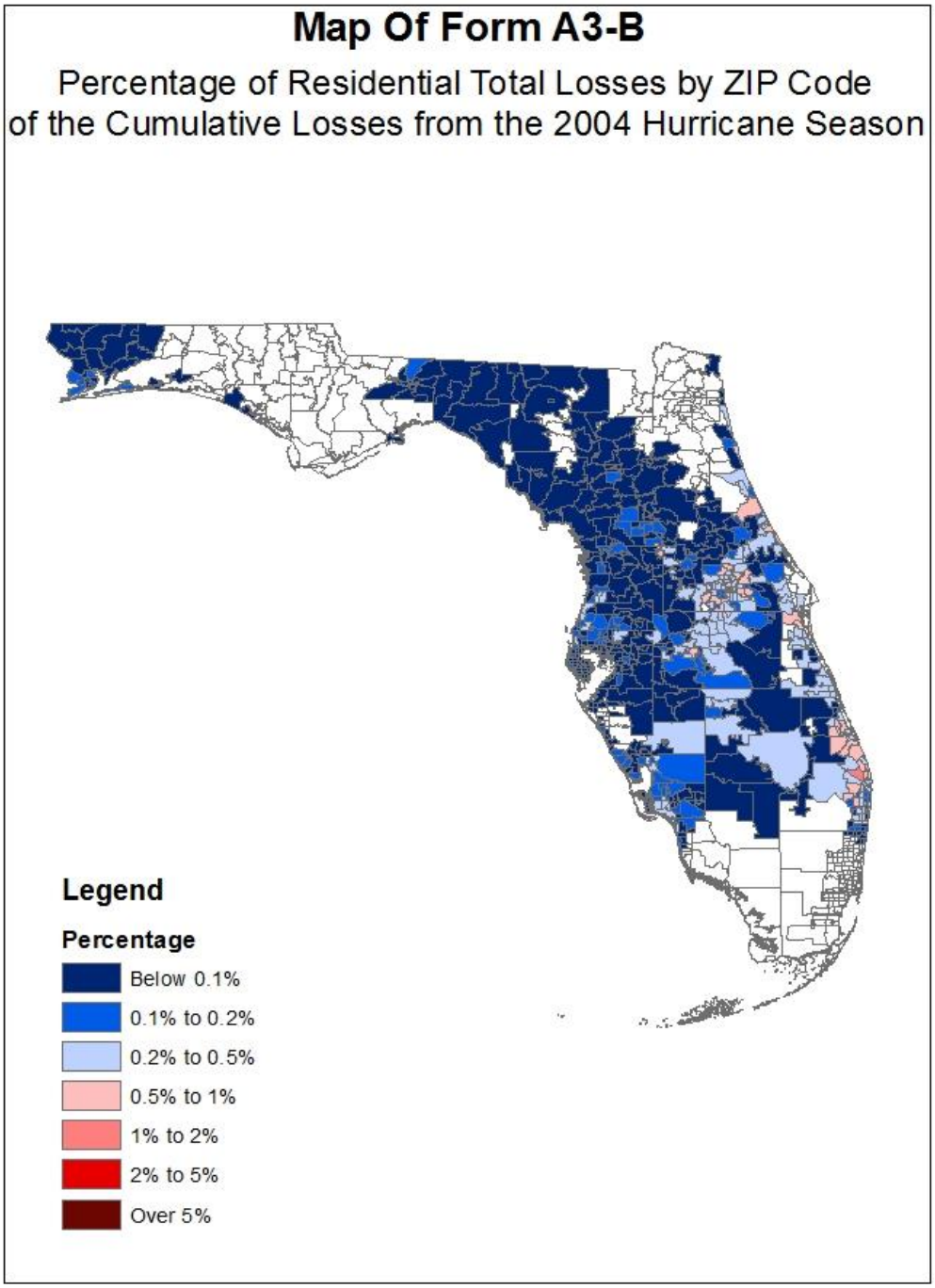


Figure 78. Percentage of residential total losses by ZIP code of the cumulative losses from the 2004 Hurricane Season.

Form A-4A: Output Ranges (2007 FHCF Exposure Data)

A. Provide personal and commercial residential output ranges in the format shown in the file named "2013FormA4A.xlsx" by using an automated program or script. Provide this form in Excel format. The file name shall include the abbreviated name of the modeling organization, the standards year, and the form name. Form A-4A (Output Ranges, 2007 FHCF Exposure Data) shall also be included in a submission appendix.

B. Provide loss costs rounded to three (3) decimal places by county. Within each county, loss costs shall be shown separately per $1,000 of exposure for frame owners, masonry owners, frame renters, masonry renters, frame condo unit owners, masonry condo unit owners, mobile home, and commercial residential. For each of these categories using ZIP Code centroids, the output range shall show the highest loss cost, the lowest loss cost, and the weighted average loss cost. The aggregate residential exposure data for this form shall be developed from the information in the file named "hlpm2007c.exe," except for insured value and deductibles information. Insured values shall be based on the output range specifications below. Deductible amounts of 0% and as specified in the output range specifications will be assumed to be uniformly applied to all risks. When calculating the weighted average loss costs, weight the loss costs by the total insured value calculated above. Include the statewide range of loss costs (i.e., low, high, and weighted average).

C. If a modeling organization has loss costs for a ZIP Code for which there is no exposure, give the loss costs zero weight (i.e., assume the exposure in that ZIP Code is zero). Provide a list in the submission document of those ZIP Codes where this occurs.

D. If a modeling organization does not have loss costs for a ZIP Code for which there is some exposure, do not assume such loss costs are zero, but use only the exposures for which there are loss costs in calculating the weighted average loss costs. Provide a list in the submission document of the ZIP Codes where this occurs.

E. All anomalies in loss costs that are not consistent with the requirements of Standard A-6 (Loss Output) and have been explained in Disclosure A-6.15 shall be shaded.

Indicate if per diem is used in producing loss costs for Coverage D (ALE) in the personal residential output ranges. If a per diem rate is used in the submission, a rate of $150.00 per day per policy shall be used.

See Appendix E.

Form A-4B: Output Ranges (2012 FHCF Exposure Data)

A. *Provide personal and commercial residential output ranges in the format shown in the file named "2013FormA4B.xlsx" by using an automated program or script. Provide this form in Excel format. The file name shall include the abbreviated name of the modeling organization, the standards year, and the form name. Form A-4B (Output Ranges, 2012 FHCF Exposure Data) shall also be included in a submission appendix.*

B. *Provide loss costs rounded to three (3) decimal places by county. Within each county, loss costs shall be shown separately per $1,000 of exposure for frame owners, masonry owners, frame renters, masonry renters, frame condo unit owners, masonry condo unit owners, mobile home, and commercial residential. For each of these categories using ZIP Code centroids, the output range shall show the highest loss cost, the lowest loss cost, and the weighted average loss cost. The aggregate residential exposure data for this form shall be developed from the information in the file named "hlpm2012c.exe," except for insured value and deductibles information. Insured values shall be based on the output range specifications below. Deductible amounts of 0% and as specified in the output range specifications will be assumed to be uniformly applied to all risks. When calculating the weighted average loss costs, weight the loss costs by the total insured value calculated above. Include the statewide range of loss costs (i.e., low, high, and weighted average).*

C. *If a modeling organization has loss costs for a ZIP Code for which there is no exposure, give the loss costs zero weight (i.e., assume the exposure in that ZIP Code is zero). Provide a list in the submission document of those ZIP Codes where this occurs.*

D. *If a modeling organization does not have loss costs for a ZIP Code for which there is some exposure, do not assume such loss costs are zero, but use only the exposures for which there are loss costs in calculating the weighted average loss costs. Provide a list in the submission document of the ZIP Codes where this occurs.*

E. *All anomalies in loss costs that are not consistent with the requirements of Standard A-6 (Loss Output) and have been explained in Disclosure A-6.15 shall be shaded.*

   Indicate if per diem is used in producing loss costs for Coverage D (ALE) in the personal residential output ranges. If a per diem rate is used in the submission, a rate of $150.00 per day per policy shall be used.

See Appendix E.

Form A-5: Percentage Change in Output Ranges (2007 FHCF Exposure Data)

A. *Provide summaries of the percentage change in average loss cost output range data compiled in Form A-4A (Output Ranges, 2007 FHCF Exposure Data) relative to the equivalent data compiled from the previously accepted model in the format shown in the file named "2013FormA5.xlsx."*

For the change in output range exhibit, provide the summary by:

- *Statewide (overall percentage change),*
- *By region, as defined in Figure 4 – North, Central and South,*
- *By county, as defined in Figure 5 – Coastal and Inland.*

B. *Provide this form in Excel format. The file name shall include the abbreviated name of the modeling organization, the standards year, and the form name. All tables in Form A-5 (Percentage Change in Output Ranges, 2007 FHCF Exposure Data) shall also be included in a submission appendix.*

C. *Provide color-coded maps by county reflecting the percentage changes in the average loss costs with specified deductibles for frame owners, masonry owners, frame renters, masonry renters, frame condo unit owners, masonry condo unit owners, mobile home, and commercial residential from the output ranges from the previously accepted model.*

Counties with a negative percentage change (reduction in loss costs) shall be indicated with shades of blue; counties with a positive percentage change (increase in loss costs) shall be indicated with shades of red; and counties with no percentage change shall be white. The larger the percentage change in the county, the more intense the color-shade.

See Appendix F.

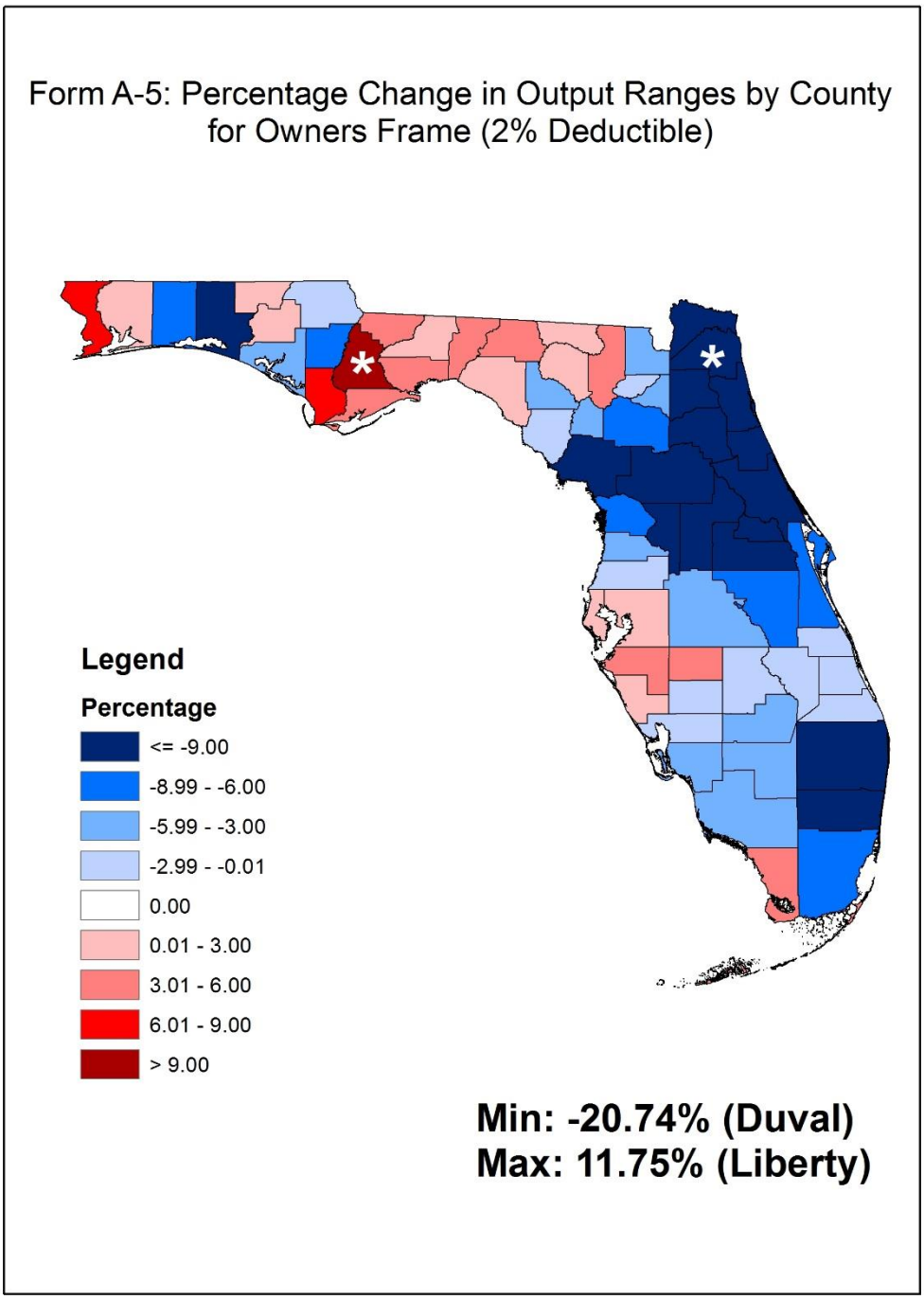


Figure 79. Percentage change in output ranges by county for owners frame (2% deductible).

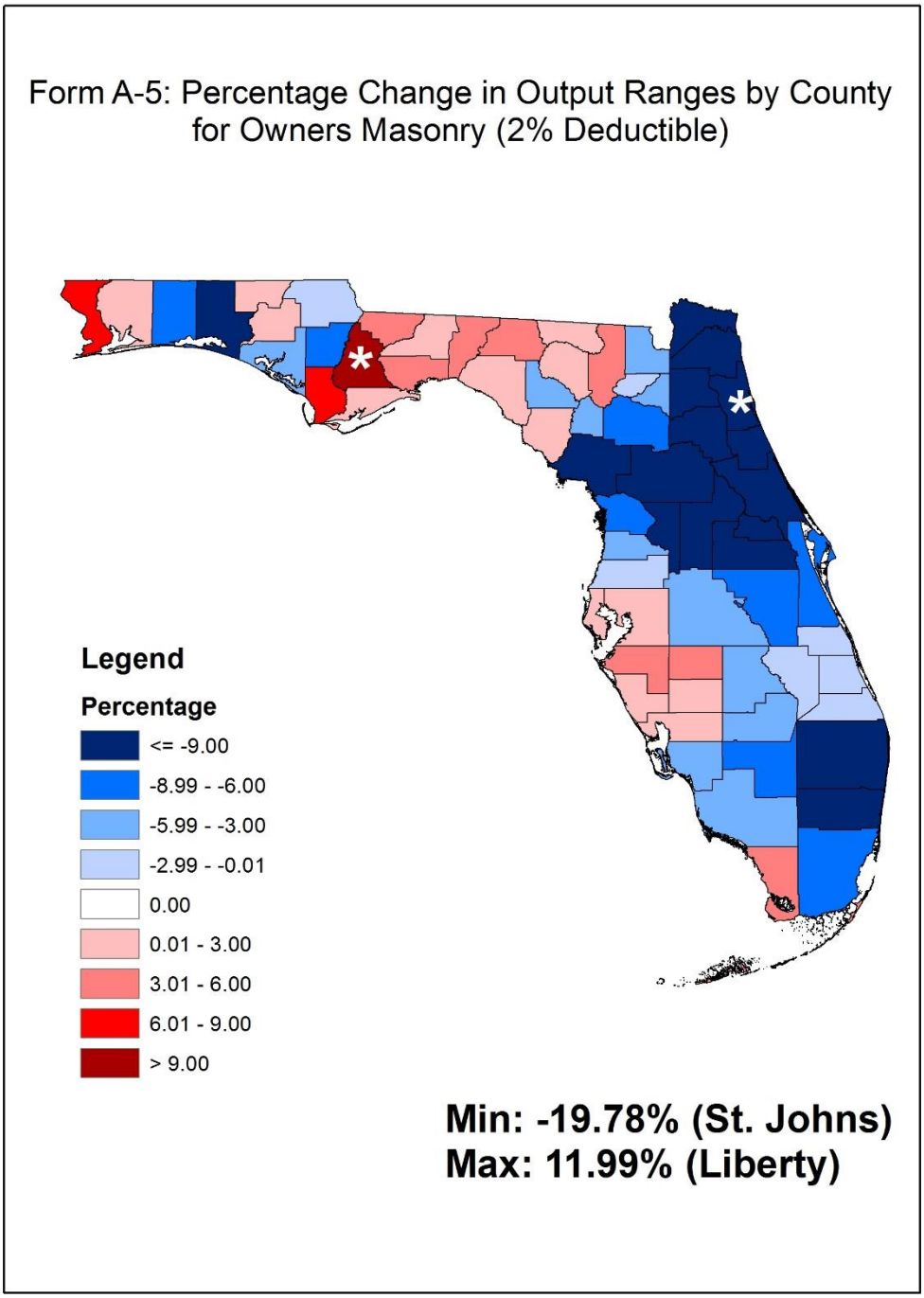


Figure 80. Percentage change in output ranges by county for owners masonry (2% deductible).

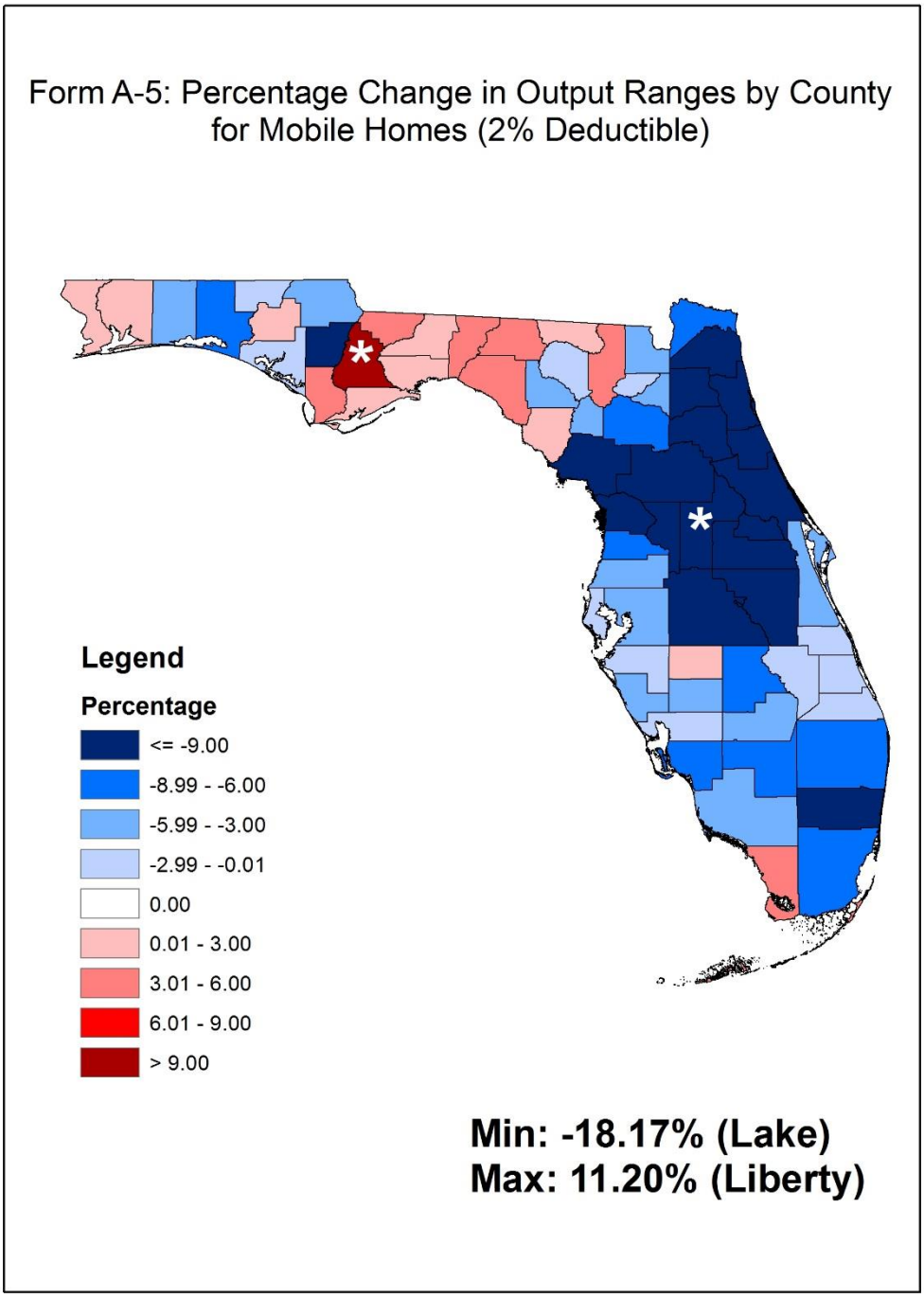


Figure 81. Percentage change in output ranges by county for mobile homes (2% deductible).

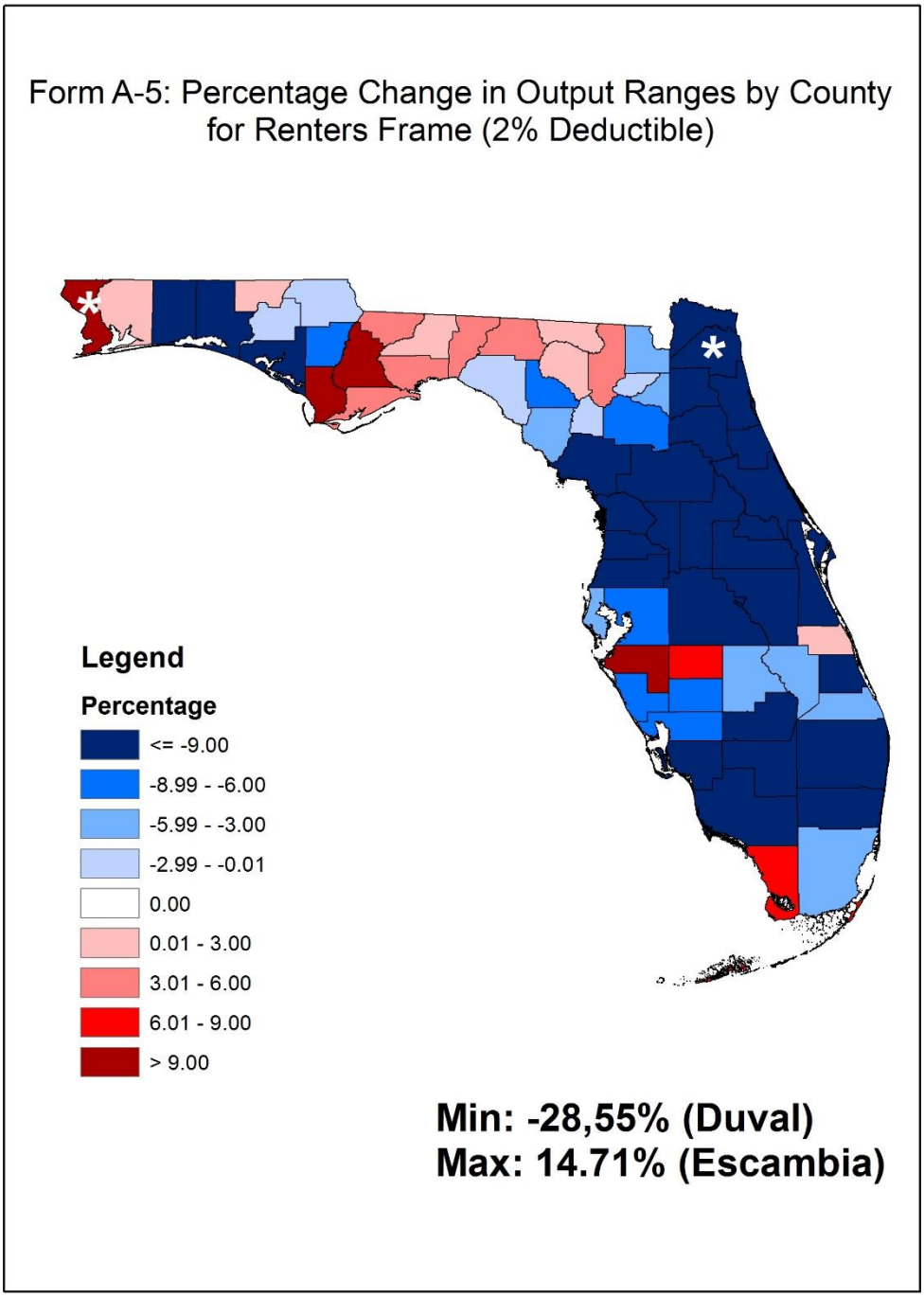


Figure 82. Percentage change in output ranges by county for renters frame (2% deductible).

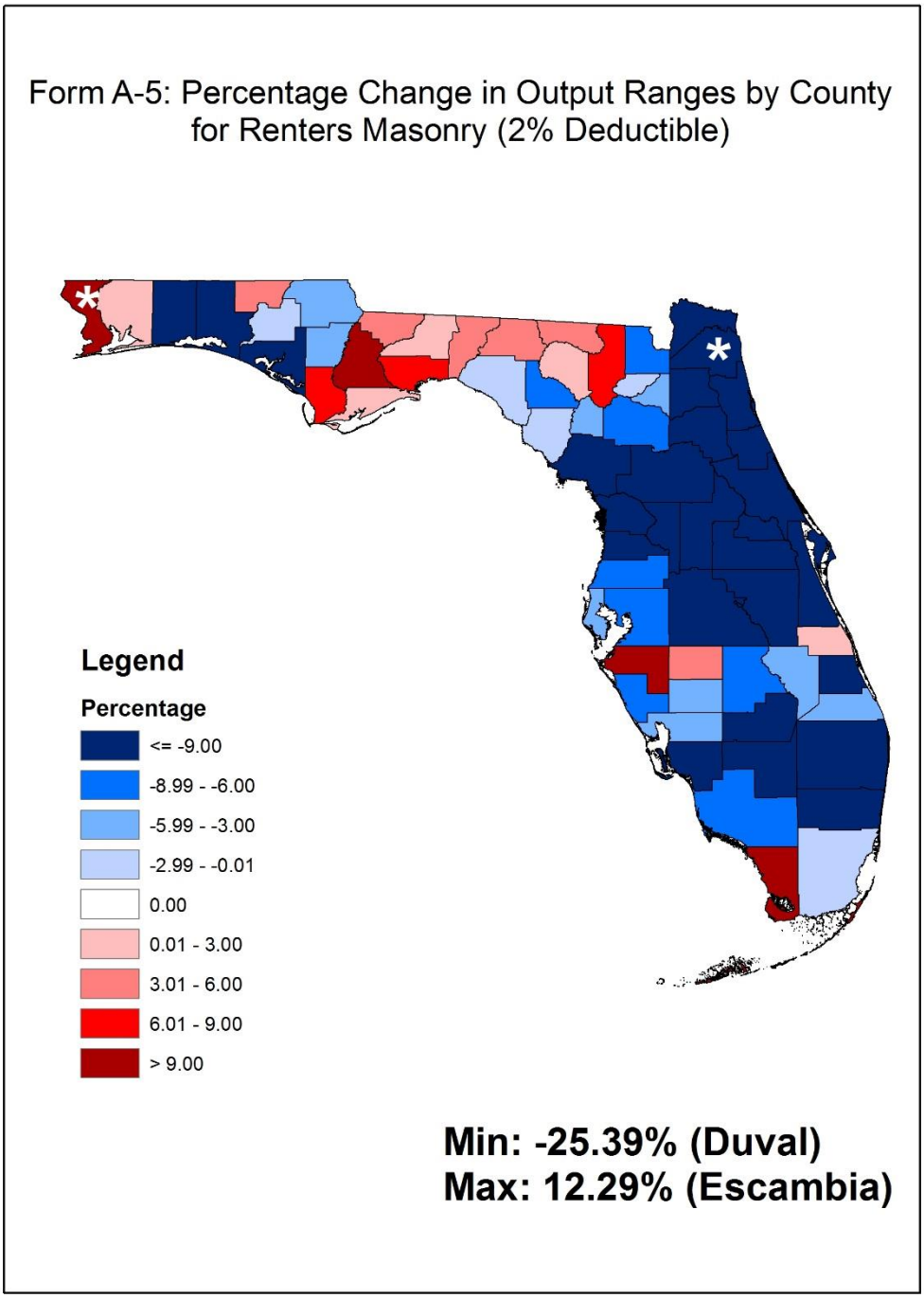


Figure 83. Percentage change in output ranges by county for renters masonry (2% deductible).

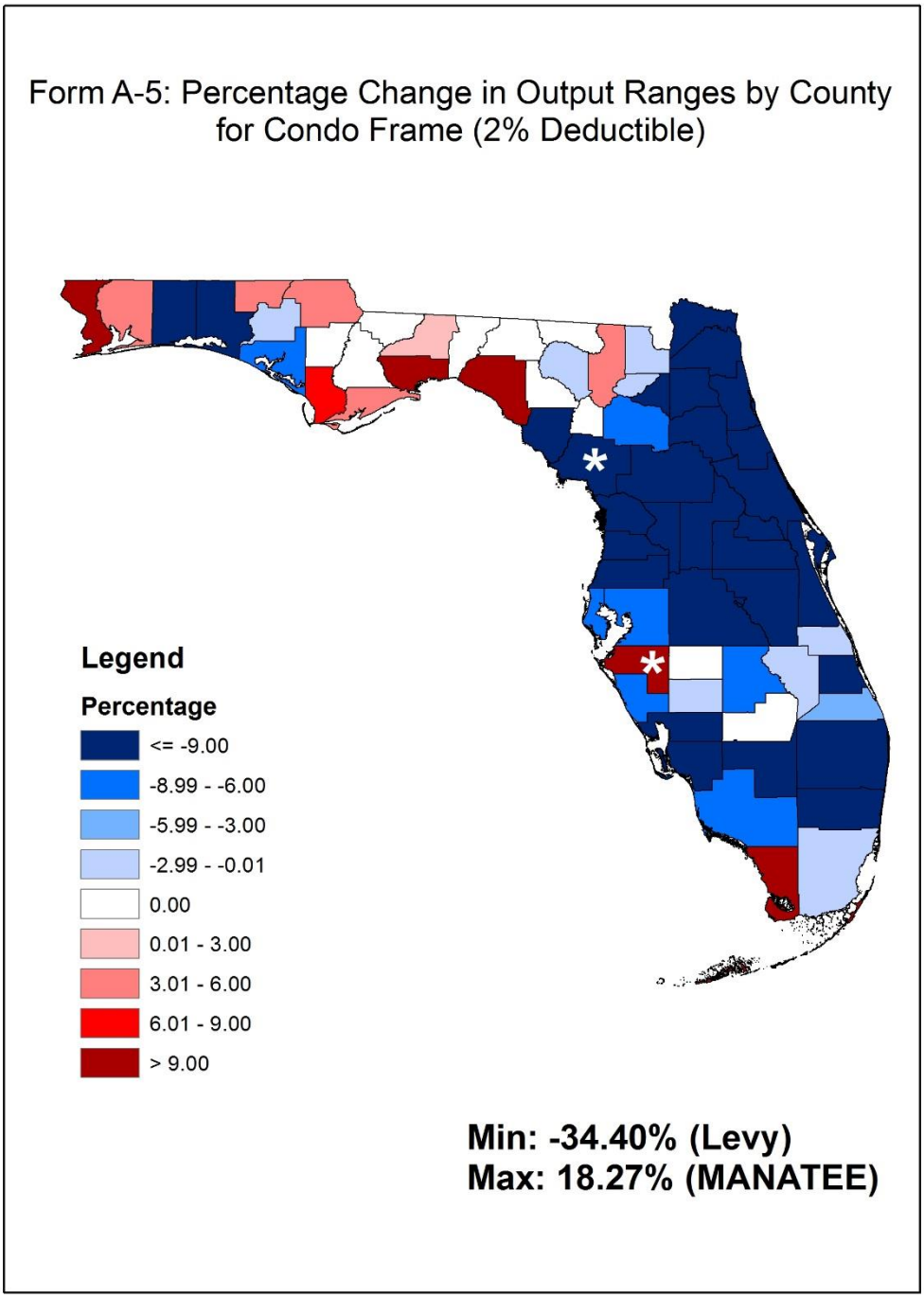


Figure 84. Percentage change in output ranges by county for condo frame (2% deductible).

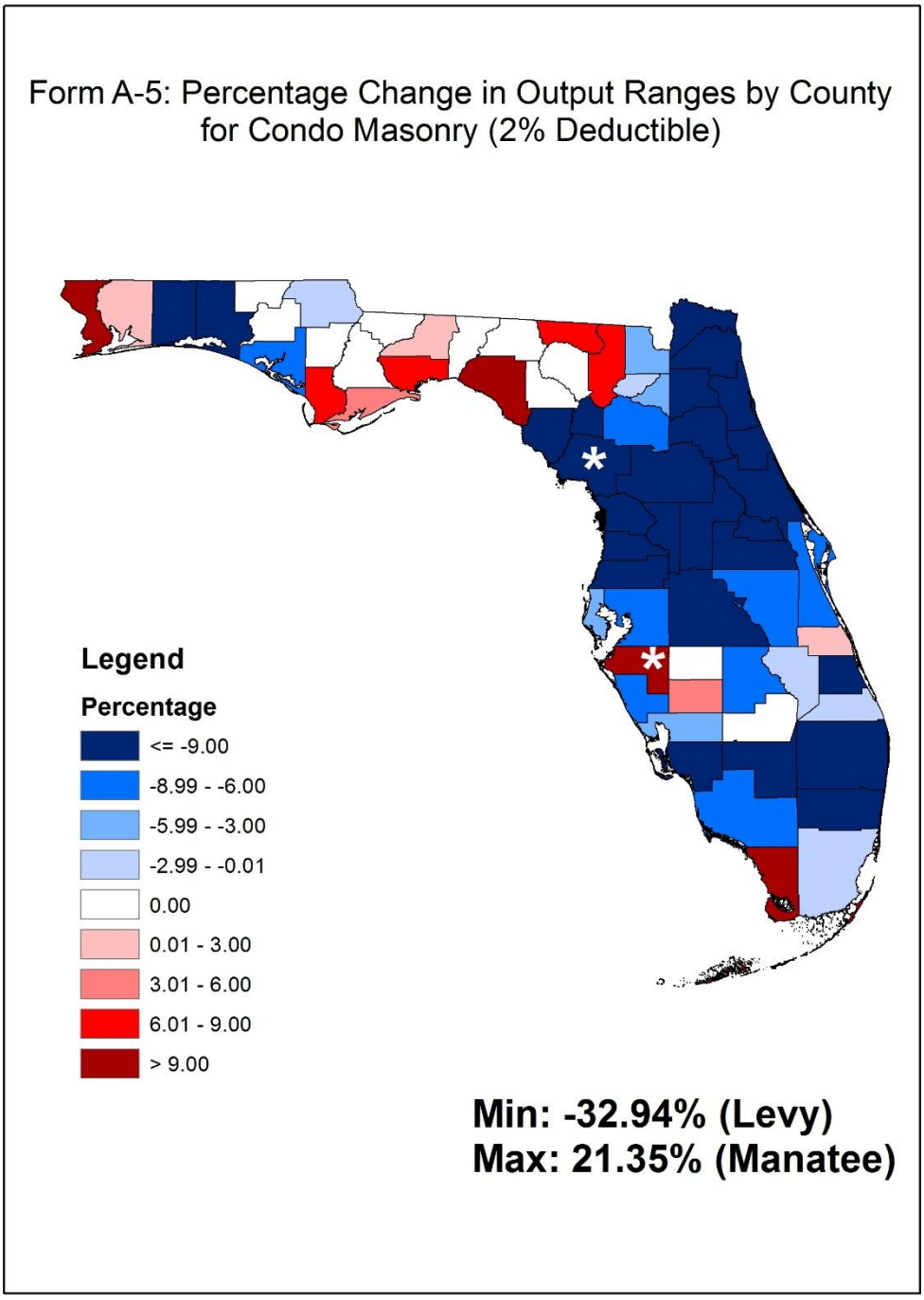


Figure 85. Percentage change in output ranges by county for condo masonry (2% deductible).

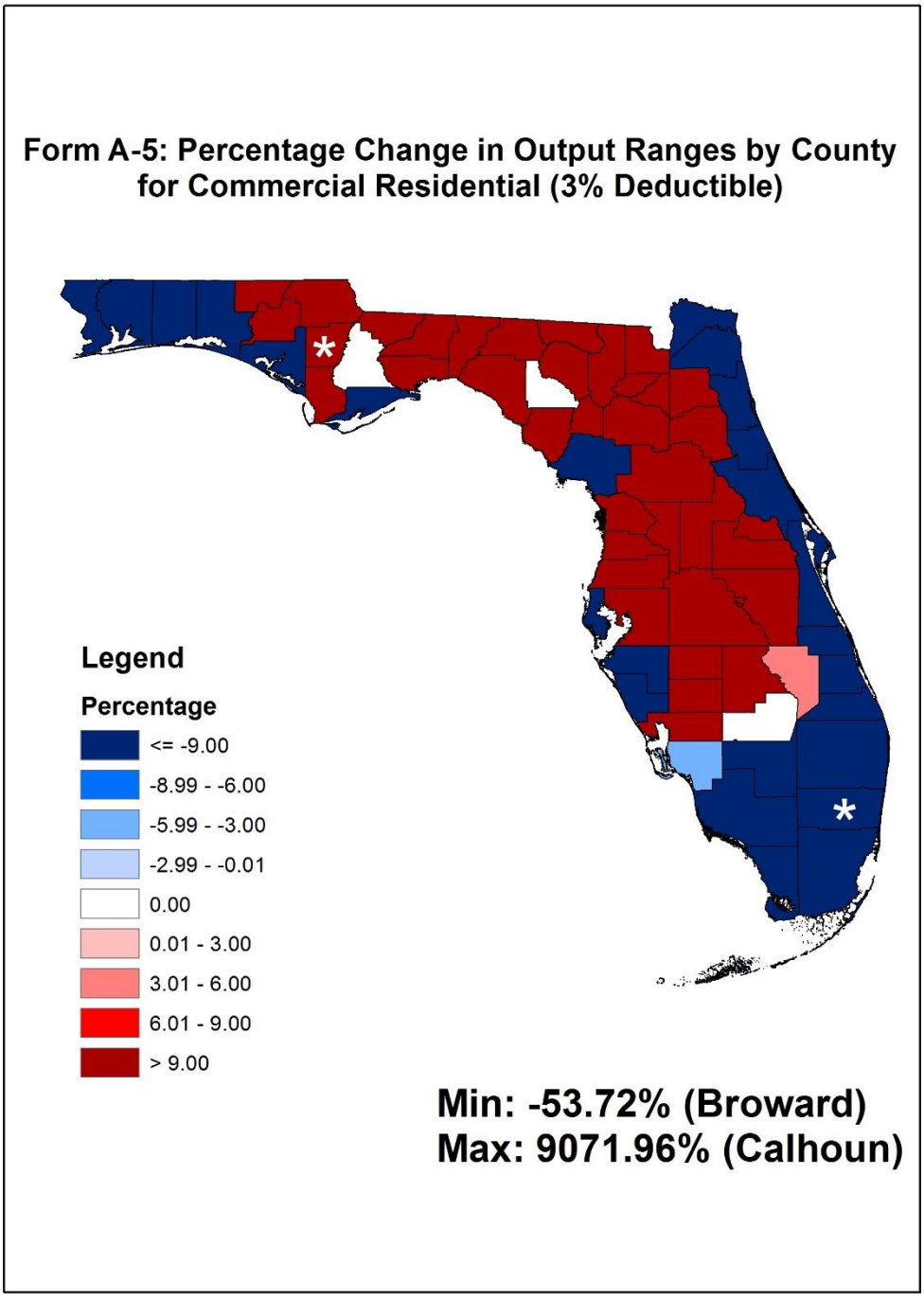


Figure 86. Percentage change in output ranges by county for commercial residential (3% deductible).

Form A-6: Personal Residential Output Ranges

A. *Provide the logical relationship to risk exhibits in the format shown in the file named "2013FormA6.xlsx."*

B. *Create exposure sets for each exhibit by modeling all of the coverages from the appropriate Notional Set listed below at each of the locations in "Location Grid A" as described in the file "NotionalInput13.xlsx." Refer to the Notional Policy Specifications below for additional modeling information. Explain any assumptions, deviations, and differences from the prescribed exposure information.*

Exhibit	Notional Set
Deductible Sensitivity	Set 1
Construction Sensitivity	Set 2
Policy Form Sensitivity	Set 3
Coverage Sensitivity	Set 4
Building Code/Enforcement (Year Built) Sensitivity	Set 5
Building Strength Sensitivity	Set 6
Condo Unit Floor Sensitivity	Set 7
Number of Stories Sensitivity	Set 8

Models shall treat points in Location Grid A as coordinates that would result from a geocoding process. Models shall treat points by simulating loss at exact location or by using the nearest modeled parcel/street/cell in the model.

Report results for each of the points in "Location Grid A" individually, unless specified. Loss cost per $1,000 of exposure shall be rounded to 3 decimal places.

C. *All anomalies in loss costs that are not consistent with the requirements of Standard A-6 (Loss Output) and have been explained in Disclosure A-6.15 shall be shaded.*

D. *Create an exposure set and report loss costs results for the strong owners frame buildings (Notional Set 6) for each of the points in "Location Grid B" as described in the file "NotionalInput13.xlsx." Provide a color-coded contour map of the loss costs. Provide a scatter plot of the loss costs (y-axis) against distance to the coast (x-axis).*

See Appendix G.

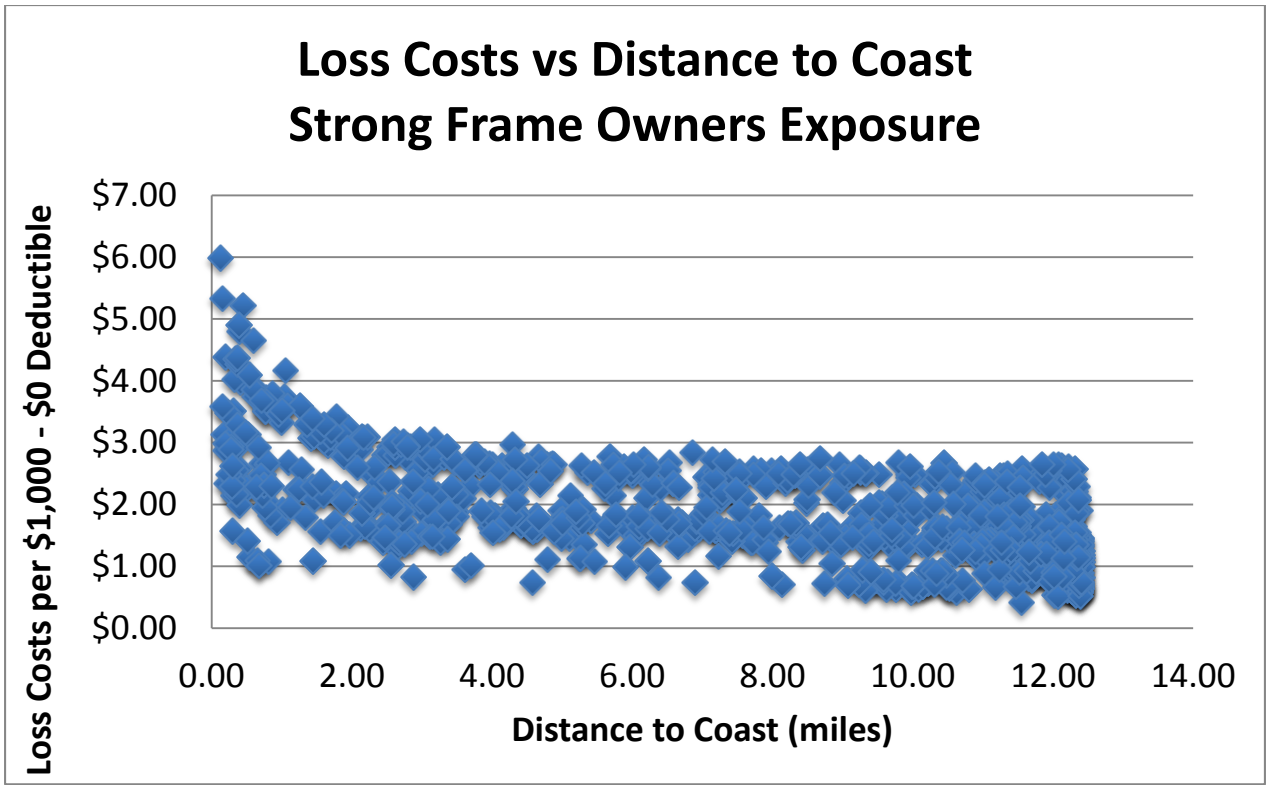


Figure 87. Loss Costs vs. Distance to the Coast Strong Owners Frame Exposures

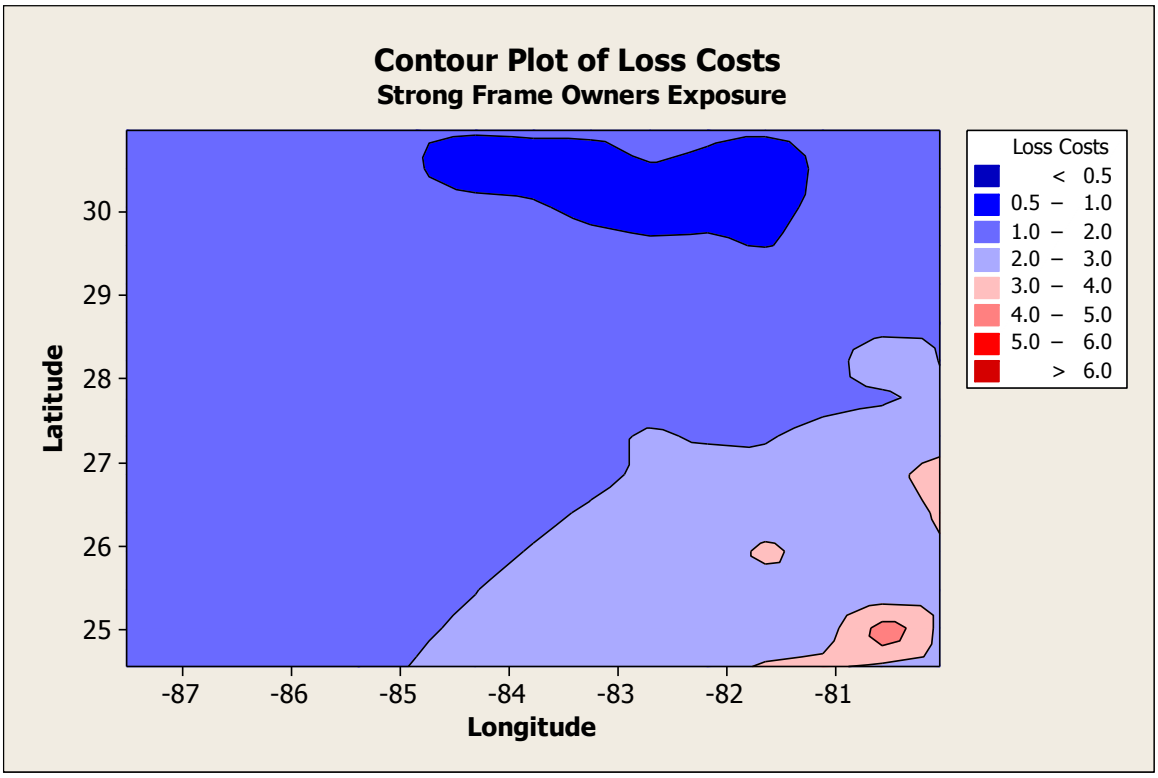


Figure 88. Contour Plot of Loss Costs for Strong Frame Owners Exposure

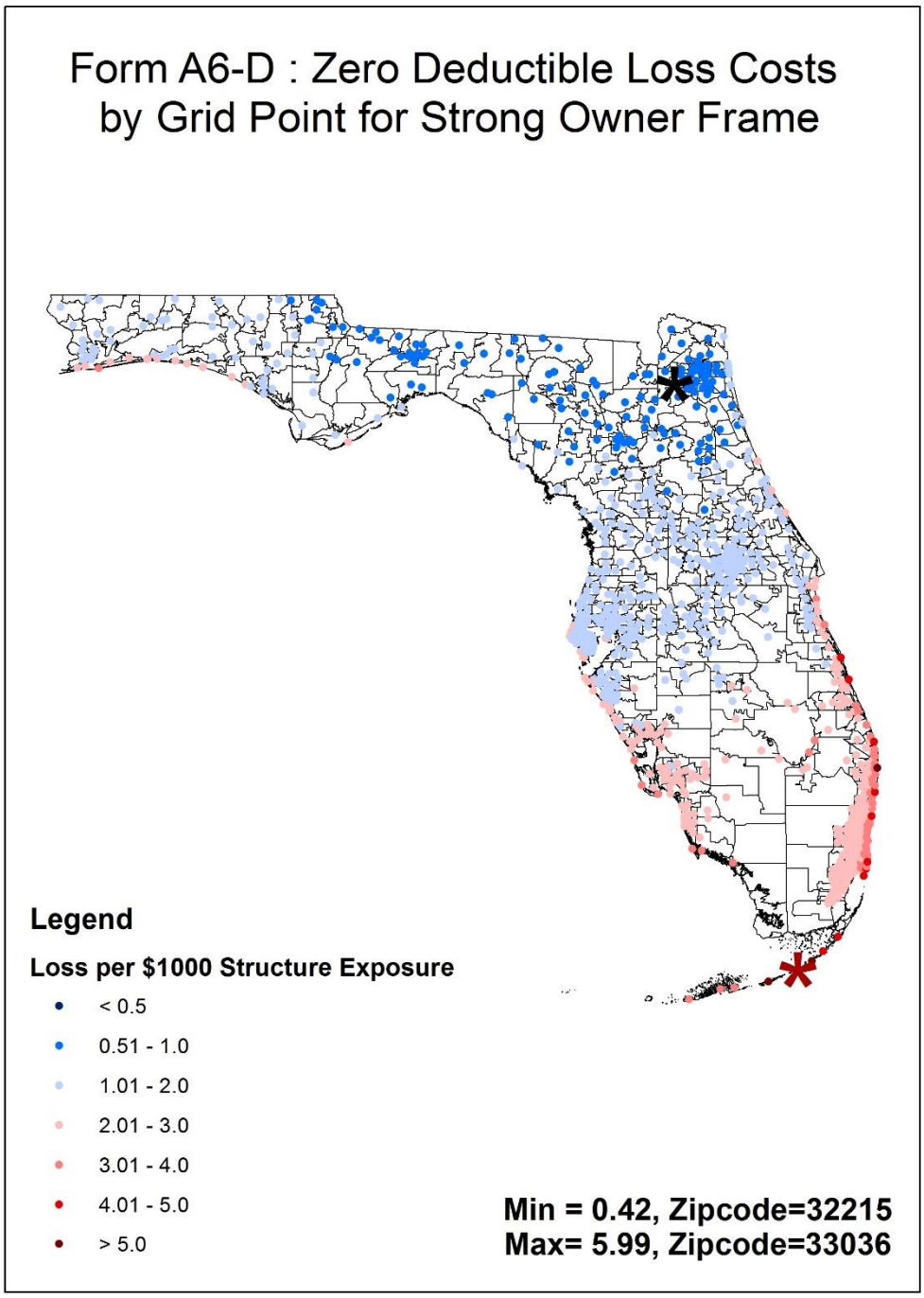


Figure 89. Zero Deductible Loss Costs by Grid Point for Strong Owner Frame.

Form A-7: Percentage Change in Logical Relationship to Risk

A. Provide summaries of the percentage change in logical relationship to risk exhibits from the previously accepted model in the format shown in the file named *"2013FormA7.xlsx."*

B. Create exposure sets for each exhibit by modeling all of the coverages from the appropriate Notional Set listed below at each of the locations in "Location Grid B" as described in the file "NotionalInput13.xlsx." Refer to the Notional Policy Specifications provided in Form A-6 (Logical Relationship to Risk, Trade Secret item) for additional modeling information. Explain any assumptions, deviations, and differences from the prescribed exposure information.

Exhibit	Notional Set
Deductible Sensitivity	Set 1
Construction Sensitivity	Set 2
Policy Form Sensitivity	Set 3
Coverage Sensitivity	Set 4
Building Code/Enforcement (Year Built) Sensitivity	Set 5
Building Strength Sensitivity	Set 6
Condo Unit Floor Sensitivity	Set 7
Number of Stories Sensitivity	Set 8

Models shall treat points in Location Grid B as coordinates that would result from a geocoding process. Models shall treat points by simulating loss at exact location or by using the nearest modeled parcel/street/cell in the model.

Provide the results statewide (overall percentage change) and by the regions defined in Form A-5 (Percentage Change in Output Ranges, 2007 FHCF Exposure Data).

C. Provide this form in Excel format. The file name shall include the abbreviated name of the modeling organization, the standards year, and the form name. All tables in Form A-7 (Percentage Change in Logical Relationship to Risk) shall also be included in a submission appendix.

See Appendix H.

Form A-8: Probable Maximum Loss for Florida

A. *Provide a detailed explanation of how the Expected Annual Hurricane Losses and Return Periods are calculated.*

For each range of losses:

Expected Annual Hurricane Losses = Total Loss / Number of years in the simulation,

Where:

Total Loss = Sum of losses for all simulated years with aggregate storm losses in the range.

Return Period = 1 / Probability of exceeding the average loss in the range,

Where:

Average Loss = Total Loss / Number of years with aggregate storm losses in the range,

And

Probability of exceeding the average loss in the range =

(Number of years with aggregate storm losses > Average Loss) / Number of years in the simulation.

B. *Complete Part A showing the personal and commercial residential probable maximum loss for Florida. For the Expected Annual Hurricane Losses column, provide personal and commercial residential, zero deductible statewide loss costs based on the 2012 Florida Hurricane Catastrophe Fund's aggregate personal and commercial residential exposure data found in the file named "hlpm2012c.exe."*

In the column, Return Period (Years), provide the return period associated with the average loss within the ranges indicated on a cumulative basis.

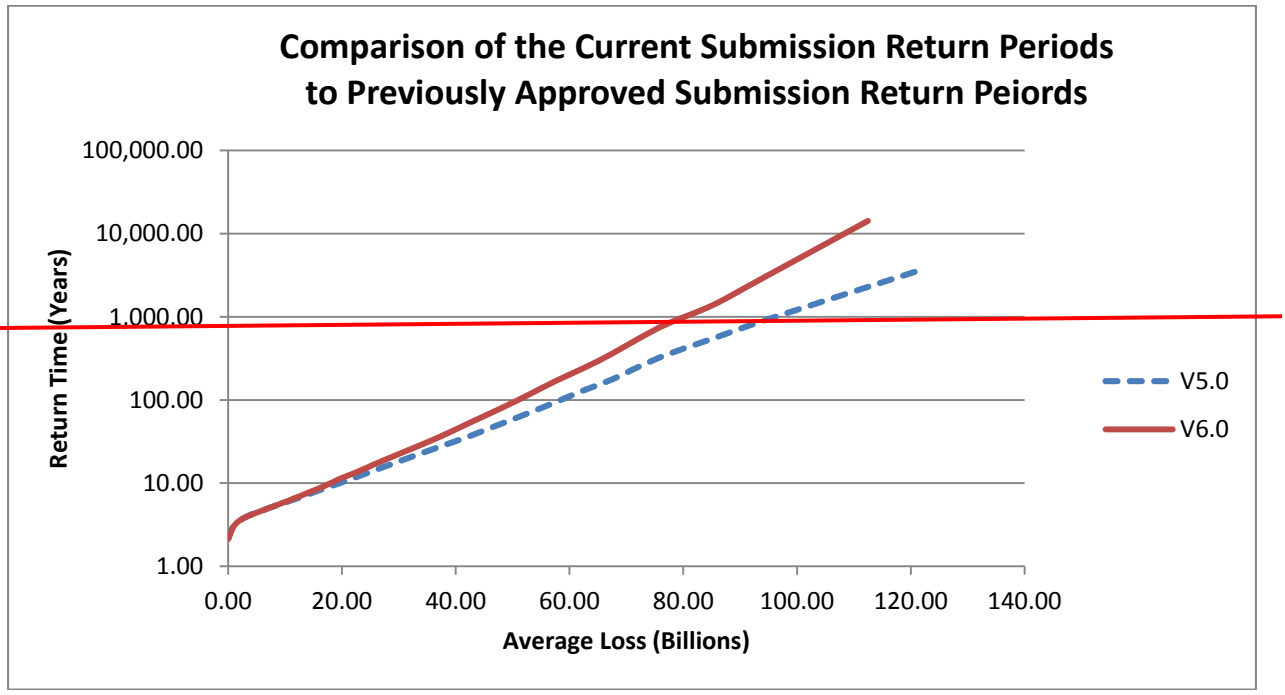
For example, if the average loss is $4,705 million for the range $4,501 million to $5,000 million, provide the return period associated with a loss that is $4,705 million or greater.

For each loss range in millions ($1,001-$1,500, $1,501-$2,000, $2,001-$2,500) the average loss within that range should be identified and then the return period associated with that loss calculated. The return period is then the reciprocal of the probability of the loss equaling or exceeding this average loss size.

The probability of equaling or exceeding the average of each range should be smaller as the ranges increase (and the average losses within the ranges increase). Therefore, the return period associated with each range and average loss within that range should be larger as the ranges increase. Return periods shall be based on cumulative probabilities.

A return period for an average loss of $4,705 million within the $4,501-$5,000 million range should be lower than the return period for an average loss of $5,455 million associated with a $5,001- $6,000 million range.

C. *Provide a graphical comparison of the current submission Residential Return Periods loss curve to the previously accepted submission Residential Return Periods loss curve. Residential Return Period (Years) shall be shown on the y-axis on a log 10 scale with Losses in Billions shown on the x-axis. The legend shall indicate the corresponding submission with a solid line representing the current year and a dotted line representing the previously accepted submission.*



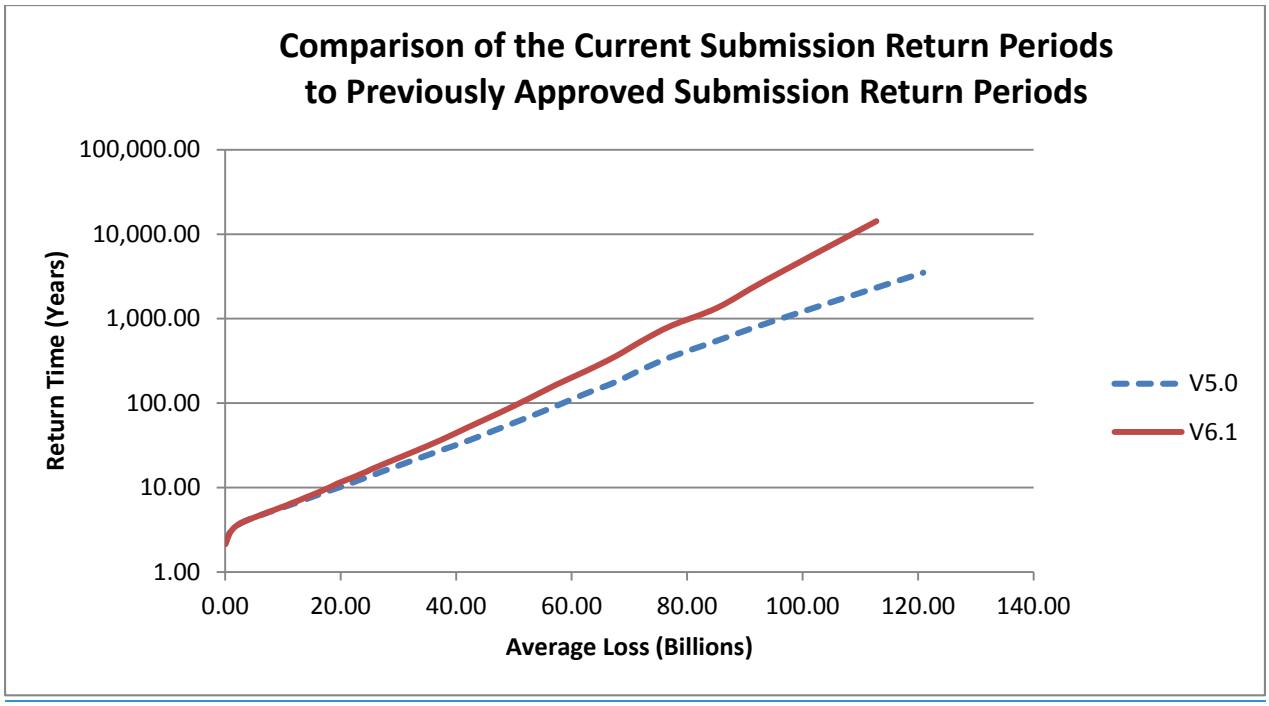


Figure 90. Comparison of return periods.

D. Provide the estimated loss and uncertainty interval for each of the Personal and Commercial Residential Return Periods given in Part B. Describe how the uncertainty intervals are derived.

The uncertainty intervals (except for the top event) are approximate 95% confidence intervals.

Let $X_1, X_2, \ldots, X_N$ be the ordered set of annual losses produced by the simulation with $X_{(1)} \leq X_{(2)} \leq \ldots \leq X_{(N)}$.

Since the sample is large enough to assume a normal approximation for the pth quantile of the ordered set, an approximate 95% confidence interval for the PML is given by $(X_{(r)}, X_{(s)})$, where

$$r = Np - 1.96\sqrt{Np(1-p)}$$

$$s = Np + 1.96\sqrt{Np(1-p)}$$

and N and p are defined as in Standard A-11, i.e.

N = number of years in the simulation

and

p = 1 – 1 / return period.

If r and/or s are not integers, let r* be the smallest integer greater than r and let s* be the smallest integer greater than or equal to s. The 95% approximate confidence interval is given by $(X_{(r^*)}, X_{(s^*)})$

The top event itself is estimated by the highest order statistic, $X_{(N)}$. Although it is not possible to compute a confidence interval for the top event using the above methods, an upper bound can be placed on the ***expected*** top event, $E(X_{(N)})$.

As per Wilkinson (1982), $E(X_{(N)}) \leq \mu + \frac{(N-1)\sigma}{\sqrt{2N-1}}$

where μ and σ are the mean and the standard deviation of the losses, respectively.

Thus an upper bound for the top even is computed as :

$$\bar{X} + \frac{(N-1)s}{\sqrt{2N-1}}$$

where $\bar{X}$ is the sample mean of the simulated annual losses and s is the sample standard deviation.

E. Provide this form in Excel format. The file name shall include the abbreviated name of the modeling organization, the standards year, and the form name. Form A-8 (Probable Maximum Loss for Florida) shall also be included in a submission appendix.

See Appendix I.

COMPUTER STANDARDS

C-1 Documentation

A. *Model functionality and technical descriptions shall be documented formally in an archival format separate from the use of letters, slides, and unformatted text files.*

The Florida Public Hurricane Loss (FPHLM) model formally documents the model functionality and technical descriptions in the primary document, an archival format separate from the use of letters, slides, and unformatted text files. The primary document uses standard software practices to formally describe the model's requirements and complete software design and implementation specifications. All documentation related to the model is maintained in the project's primary document, a central location that is easily accessible.

B. *The modeling organization shall maintain a primary document repository, containing or referencing a complete set of documentation specifying the model structure, detailed software description, and functionality. Development of the documentation shall be indicative of accepted software engineering practices.*

The FPHLM maintains a primary document repository to satisfy the aforementioned requirements. In addition, the FPHLM maintains a user manual, designed for the end user, which provides a high-level introduction and a step-by-step guide to the whole system. All the documents are available for inspection on the project's primary document repository. Accepted software engineering practices are used to render all the documents more readable, self-contained, consistent, and easy to understand. Every component of the system is documented with standard use case, class, data flow, sequence diagrams, etc. The diagrams describe in detail the structure, logic flow, information exchange among submodules, etc. of each component and increase the visibility of the system. The diagrams describing the component functionality and structure also make each component of the system reusable and easily maintainable.

C. *All computer software (i.e., user interface, scientific, engineering, actuarial, data preparation, and validation) relevant to the submission shall be consistently documented and dated.*

The primary document repository contains all of the required documentation organized in chapters and sections linked to one another on the basis of their mutual relationships. Thus, the entire document can be viewed as a hierarchical referencing scheme in which each module is linked to its sub-module, which ultimately refers to the corresponding codes.

D. *The modeling organization shall maintain (1) a table of all changes in the model from the previously accepted submission to the initial submission this year and (2) a table of all substantive changes since this year's initial submission.*

These tables are maintained and documented and will be available for review.

E. *Documentation shall be created separately from the source code.*

Databases and formats of all the input/output data files are comprehensively documented. All source code is properly documented in terms of both in-line detailed comments and external higher-level documentation, and they are maintained under version control systems. Source-code documentation has been created separately from the source code.

C-2 Requirements

The modeling organization shall maintain a complete set of requirements for each software component as well as for each database or data file accessed by a component. Requirements shall be updated whenever changes are made to the model.

The FPHLM is divided into several major modules, each of them providing one or more inputs to other modules. Requirements of each of the modules, including input/output formats, are precisely documented. In addition to maintaining a detailed documentation of each module of the system using standard software practices, several other documents are maintained as part of a large-scale project management requirement, including a quality assurance document, a system hardware and software specification document, a training document, a model maintenance document, a testing document, a user manual, etc. Moreover, detailed documentation has been developed for the database consisting of the schema and information about each table. Additionally, information about the format for each data file (in the form of an Excel or text file) accessed by different programs is documented. Whenever changes are made to a model, the corresponding requirements documentation is updated to reflect such changes.

Disclosure

1. ***Provide a description of the documentation for interface, human factors, functionality, documentation, data, human and material resources, security, and quality assurance.***

The user interface, functionality requirements, and material resources of each of the modules are described in the relevant module documentation. Database schema and table formats are separately documented for the whole system and attached to the primary document. A separate software testing and quality assurance document describes the system quality, performance, and stability concerns. Additionally, a user manual and a human resource management document are maintained. Additionally, security, software and hardware specifications for the system as well as training plans are documented.

C-3 Model Architecture and Component Design

The modeling organization shall maintain and document (1) detailed control and data flow diagrams and interface specifications for each software component, (2) schema definitions for each database and data file, and (3) diagrams illustrating model-related flow of information and its processing by modeling organization personnel or team. Documentation shall be to the level of components that make significant contributions to the model output.

Interface specifications for each of the modules are included in the module documentation. In addition, the user manual provides further information about the user interface specification. Control and data flow diagrams are presented at various levels of the model documentation. High-level flow diagrams are used to illustrate the flow of the whole system and the interactions among modules. More technical and detailed diagrams are used in module-level descriptions.

The database schema are documented and attached as part of the primary document. A detailed schema representation of the active database is documented with additional information such as database maintenance, tuning, data loading methodologies, etc. to provide a complete picture of the database maintained for the project.

Additionally, UML information flow diagrams illustrating the flow of model-related information and its processing by modeling organization personnel or team are included in the project's primary document.

These documents will be made available to the professional team during the on-site visit.

C-4 Implementation

A. *The modeling organization shall maintain a complete procedure of coding guidelines consistent with accepted software engineering practices.*

The FPHLM has developed and followed a set of coding guidelines that is consistent with accepted software practices. This set includes guidelines for version control, code revision history maintenance, etc. All the developers involved in the system development adhere to the instructions in these documents.

B. *The modeling organization shall maintain a complete procedure used in creating, deriving, or procuring and verifying databases or data files accessed by components.*

The FPHLM uses a PostgreSQL database to store the related data necessary for the model. The database documentation includes the procedures for creating and deriving the database. Data files are generated by different modules and used as interfaces between modules. Several data verification techniques are undertaken to ensure the correctness. Details about these are included in the module documentation.

C. *All components shall be traceable, through explicit component identification in the flow diagrams, down to the code level.*

Traceability, from requirements to the code level and vice versa, is maintained throughout the system documentation.

D. *The modeling organization shall maintain a table of all software components affecting loss costs, with the following table columns: (1) Component name, (2) Number of lines of code, minus blank and comment lines; and (3) Number of explanatory comment lines.*

The FPHLM primary document includes a table that gives the above-requested information. The table is available for review by the professional team.

E. *Each component shall be sufficiently and consistently commented so that a software engineer unfamiliar with the code shall be able to comprehend the component logic at a reasonable level of abstraction.*

All the software codes are properly provided with code-level comments, and a consistent format is maintained throughout the software modules. These code-level comments include a summary of important changes, names of developers involved in each modification, function headers, and in-line comments to explain potentially ambiguous software code.

F. *The modeling organization shall maintain the following documentation for all components or data modified by items identified in Standard G-1 (Scope of the Computer Model and Its Implementation), Disclosure 5:*

1. ***A list of all equations and formulas used in documentation of the model with definitions of all terms and variables.***

2. ***A cross-referenced list of implementation source code terms and variable names to items within F.1.***

Tables that map the equations and formulas used in documentation of the model implementation source code terms and variable names were added as glossaries to the model's documentation, thus combining F.1 and F.2 into the same table. These tables enhance the model's documentation and include the equations and formulas for each module (not just the modified ones from the prior year's submission).

Disclosure

1. ***Specify the hardware, operating system, other software, and all computer languages required to use the model.***

The system is primarily a web-based application that is hosted on a Tomcat web application server. The backend server environment is Linux and the server side scripts are written in Java Server Pages (JSP) and JavaBeans. Backend probabilistic calculations are coded in C++ using the IMSL library and called through Java Native Interface (JNI). The system uses a PostgreSQL database that runs on a Linux server. Server side software requirements are the IMSL library CNL 5.0, JDBC 3, JNI 1.3.1, and JDK 1.5.

The end-user workstation requirements are minimal. The recommended web browsers are Internet Explorer 8.0 running on Windows XP or Internet Explorer 9.0 running on Windows 7. However, other modern web browsers such as Mozilla Firefox running on either Windows or Linux should also deliver optimal user experience. Typically, the manufacturer's minimal set of features for a given web browser and operating system combination is sufficient for an optimal operation of the application.

C-5 Verification

A. *General*

For each component, the modeling organization shall maintain procedures for verification, such as code inspections, reviews, calculation crosschecks, and walkthroughs, sufficient to demonstrate code correctness. Verification procedures shall include tests performed by modeling organization personnel other than the original component developers.

The FPHLM software verification is done in three stages:

1. Code inspection and verification by the code developer.
2. Inspection of the input and validation of the output by the system modeler.
3. Review and extensive testing of the code by modeler personnel who are not part of the original component development.

The first level of verification includes code-level debugging, walking through the code to ensure a proper flow, inspection of internal variables through intermediate output printing and error logging, use of exception handling mechanisms, calculation crosschecks, and verification of the output against sample calculations provided by the system modeler.

In the second level of the verification, the modeler is provided with sample inputs and corresponding outputs. The modeler then conducts black-box testing to verify the results against his or her model. Finally, each component is rigorously tested by modeler personnel not responsible for original component development.

B. *Component Testing*

1. The modeling organization shall use testing software to assist in documenting and analyzing all components.

Component testing (C-5.B) and data testing (C-5.C) are done in the third level of verification. The system is rigorously checked for the correctness, precision, robustness, and stability of the whole system. Calculations are performed outside the system and compared against the system-generated results to ensure the system correctness. Extreme and unexpected inputs are given to the system to check the robustness. Wide series of test cases are developed to check the stability and the consistency of the system.

2. Unit tests shall be performed and documented for each component.

Unit testing is done at the first and third levels of verification. The developer tests all the units as the unit is developed and modified. Then all the units are tested again by the external testing team. Both black-box and white-box tests are performed and documented in a separate testing document.

3. ***Regression tests shall be performed and documented on incremental builds.***

Regression testing is performed for each module. In this kind of testing methodology, the modules that have undergone some changes and revisions are retested to ensure that the changes have not affected the entire system in any undesired manner.

4. ***Aggregation tests shall be performed and documented to ensure the correctness of all model components. Sufficient testing shall be performed to ensure that all components have been executed at least once.***

Aggregation testing is performed at all three levels of verification. Aggregation testing is performed by running each major module as a complete package. It is ensured that all components have been executed at least once during the testing procedure. All the test cases executed are described in the software testing and verification documentation.

C. Data Testing

1. ***The modeling organization shall use testing software to assist in documenting and analyzing all databases and data files accessed by components.***

The FPHLM uses a PostgreSQL database to store the required data. Data integrity and consistency are maintained by the database itself. Moreover, different queries are issued and PL/SQL is implemented to check the database. PostgreSQL has a very robust loader, which is used to load the data into the database. The loader maintains a log that depicts if the loading procedure has taken place properly and completely without any discrepancy. Data files are manually tested using commercial data manipulation software such as Microsoft Excel and Microsoft Access.

2. ***The modeling organization shall perform and document integrity, consistency, and correctness checks on all databases and data files accessed by the components.***

All the tests are well documented in a separate testing document.

Disclosures

1. ***State whether any two executions of the model with no changes in input data, parameters, code, and seeds of random number generators produce the same loss costs and probable maximum loss levels.***

The model produces the same loss costs and probable maximum loss levels if it is executed more than once with no changes in input data, parameters, code, and seeds of random number generators.

2. ***Provide an overview of the component testing procedures.***

The FPHLM software testing and verification is done in three stages.

[A] Code inspection and the verification by the code developer

The code developer performs a sufficient amount of testing on the code and does not deliver the code until he or she is convinced of the proper functionality and robustness of the code.

The first level of verification includes code-level debugging, walking through the code to ensure proper flow, inspection of internal variables through intermediate output printing and error logging, use of exception handling mechanisms, calculation crosschecks, and verification of the output against sample calculations provided by the system modeler.

[B] Verification of results by the person who developed the system model

Once the first level of testing is done, the developer sends the sample inputs and the generated results back to the modeler. Then the system modeler double-checks the results against his or her model. The code is not used in the production environment unless approved by the modeler.

[C] Review and extensive testing of the code by modeler personnel other than the original component developers.

The system is rigorously checked by modeler personnel (testers) other than the original component developers for the correctness, precision, robustness, and stability of the whole system. Calculations are performed outside the system and compared against the system generated results to ensure the system correctness. Extreme and unexpected inputs are given to the system to check the robustness. Wide series of test cases are developed to check the stability and the consistency of the system.

Unit testing, regression testing, and aggregation testing (both white-box and black-box) are performed and documented.

Any flaw in the code is reported to the developer, and the bug-corrected code is again sent to the tester. The tester then performs unit testing again on the modified units. Additionally, regression testing is performed to determine if the modification affects any other parts of the code.

Different testing tools and software packages are used to test different components of the system. The detailed list of the various testing tools and/or techniques used for different components of the system is provided in the main document and will be available for audit.

3. ***Provide a description of verification approaches used for externally acquired data, software, and models.***

The verification approaches used for externally acquired data, software, and models are documented in the primary document. They will be available for review by the professional team

C-6 Model Maintenance and Revision

A. *The modeling organization shall maintain a clearly written policy for model revision, including verification and validation of revised components, databases, and data files.*

The FPHLM is periodically enhanced to reflect new knowledge acquired about hurricanes and Florida ZIP Code information. A clearly written policy for model revision is maintained in the primary document.

B. *A revision to any portion of the model that results in a change in any Florida residential hurricane loss cost or probable maximum loss level shall result in a new model version identification.*

Whenever a revision results in a change in any Florida residential hurricane loss cost or probable maximum loss level, a new model version identification will be assigned to the revision. Verification and validation of the revised units are repeated according to the above-mentioned "software verification procedures" document.

C. *The modeling organization shall use tracking software to identify and describe all errors, as well as modifications to code, data, and documentation.*

The FPHLM uses Subversion to identify and describe all errors, as well as modifications to code, data, and documentation. Subversion is a revision control system widely used in recent years by important projects and has been termed the successor of CVS (Concurrent Versions System). We can record the history of source files and documents by using Subversion.

D. *The modeling organization shall maintain a list of all model versions since the initial submission for this year. Each model description shall have unique version identification, and a list of additions, deletions, and changes that define that version.*

A list of all model versions since the initial submission will be maintained. Each model revision will have a unique model version number (i.e., unique version identification) and a list of additions, deletions, and changes that define that version. The unique model version will consist of the scheme "V[major].[minor]." The terms "[major]" and "[minor]" are positive integers that correspond to substantial and minor changes in the model, respectively. A minor change in the model would cause the minor number to be incremented by one, and similarly, a major change in the model would cause the major number to be incremented by one with the minor reset to zero. The rules that prompt changes in the major and minor numbers are described in Disclosure 2.

Disclosures

1. Identify procedures used to maintain code, data, and documentation.

The FPHLM's software development team employs source revision and control software for all software development. In particular, the FPHLM employs Subversion, an accepted and effective system for managing simultaneous development of files. Recently, it has been used in large programming projects both in the open-source community and in the corporate world to track modifications to source code and documentation files. Subversion maintains a record of the changes to each file and allows the user to revert to a previous version, merge versions, and track changes. This software is able to record the information for each file, the date of each change, the author of each change, the file version, and the comparison of the file before and after the changes. The detailed information will be made available to the professional team during the on-site visit.

2. Describe the rules underlying the model and code revision identification systems.

The model identification system consists of the scheme "V[major].[minor]." The terms "[major]" and "[minor]" are positive integers that correspond to major and minor changes in the model, respectively. A minor change causes the minor number to be incremented by one, and similarly, a major change causes the major number to be incremented by one with the minor number reset to zero. The rules that prompt major or minor changes in the model are the following:

Rules that trigger a change in the major number:

- Updates in any of the main modules of the FPHLM: any change resulting in the partial or total modification of the algorithm/model of the Storm Generation, Wind Field, Damage Estimation, and/or Insurance Loss models.

Rules that trigger a change in the minor number:

- Slight changes to the Storm Generation, Wind Field, and/or Damage Estimation modules: small updates such as a change in the Holland B parameter or any change to correct deficiencies that do not result in a new algorithm for the component.

- Updates to correct errors in the computer code: modifications in the code to correct deficiencies or errors such as a code bug in the computer program.

- Changes in the probability distribution functions using updated or corrected historical data, such as the updates of the HURDAT2 database: each year the model updates its HURDAT database with the latest HURDAT2 data released by the National Hurricane Center, which is used as the input in the Storm Generation Model.

- Updates of the ZIP Code list: every two years the ZIP Codes used in the model must be updated according to information originating from the United States Postal Service.

- Updates in the validation of the vulnerability matrices: the incorporation of new data, such as updated winds and insurance data, may trigger a tune-up of the vulnerability matrices used in the Insurance Loss Model.

If any change results in a change in loss costs estimates or probable maximum loss level, there will be at least a change in the minor revision number.

Consequently, for the submission of November 1, 2014, the Florida Public Hurricane Loss Model changed its version identification from V5.0 to V6.0 because of the incorporation of the most recent HURDAT2 database, the updated ZIP Code list, and the changes in the meteorological and vulnerability models. For a detailed description of the aforementioned changes, please refer to Standard G-1, Disclosure 5.

C-7 Security

The modeling organization shall have implemented and fully documented security procedures for: (1) secure access to individual computers where the software components or data can be created or modified, (2) secure operation of the model by clients, if relevant, to ensure that the correct software operation cannot be compromised, (3) anti-virus software installation for all machines where all components and data are being accessed, and (4) secure access to documentation, software, and data in the event of a catastrophe.

The FPHLM maintains a set of security procedures to protect data and documents from deliberate and inadvertent changes. These procedures include both physical and electronic measures. A set of policies identifies different security issues and addresses each of them. All the security measures are properly documented and attached to the primary document.

Disclosure

1. ***Describe methods used to ensure the security and integrity of the code, data, and documentation.***

Electronic measures include the use of different authorization levels, special network security enforcements, and regular backups. Each developer is given a separate username and password and assigned a level of authorization so that even a developer cannot change another developer's code. The users of the system are given usernames and passwords so that unauthorized users cannot use the system. External users are not allowed direct access to any of the data sources of the system. The network is extensively monitored for any unauthorized actions using standard industry practices. Since the system runs on a Linux sever environment, minimal virus attacks are expected.

Any sensitive or confidential data (insurance data, for example) are kept on an unshared disk on a system that has user access control and requires a login. Screen locks are enforced whenever the machine is left unattended. In addition, for system security and reliability purposes, we also deploy a development environment besides the production environment. Modifications to the code and data are done in the development environment and tested by in-house developers. The final production code and data can only be checked into the production environment by the authorized personnel. The models resulting from the FPHLM project can only be used by the authorized users. Authorized user accounts are created by the project manager. Regular backups of the server are taken and stored in two ways: physically and electronically. Backups are performed daily and are kept for six weeks. Nightly backups of all UNIX data disks and selected Windows data disks (at user requests) are performed over the network onto LT02 and LT03 tapes. The tape drives have built-in diagnostics and verification to ensure that the data is written correctly to the tapes. This ensures that if the tape is written successfully, it will be readable, provided no physical damage occurred to the tape. A copy of each backup is placed in a secure and hurricane-protected building. Additionally, the application server and the database server are physically secured in a secure server room with alarm systems. In case of disasters, we have implemented a set of preparation procedures and recovery plans as outlined in "FIU SCIS Hurricane Preparation Procedures.

Appendix A - Expert Review Letters

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
November 1, 2014

Assessment of the meteorological portion of the State of Florida Public Hurricane Model

February 15, 2007

Gary M. Barnes
Professor, Department of Meteorology
School of Ocean and Earth Science and Technology
University of Hawaii at Manoa

Introduction

My review of the State of Florida Public Hurricane Model is based on a three day visit to Florida International University in December, and an examination of the submission draft provided to me in February. I have had full access to the meteorological portion of the model, access to the draft for the Florida commission, and access to prior submittals to the commission from several other groups in order to establish a sense of what is desired by the commission. I am pleased to report that the issues that I have raised have received their attention and I believe that the model meets all the standards set forth by the commission. Ultimately this model, when linked to engineering and actuarial components, will provide objective guidance for the estimation of wind losses from hurricanes for the state of Florida. It does not address losses from other aspects of a tropical cyclone such as storm surge, or fresh water flooding. I now offer specific comments on each of the six meteorological standards established by the commission to ascertain this model's suitability.

M-1 Official Hurricane Set

The consortium of scientists working on the Public model have adopted HURDAT (1900- 2006) to determine landfall frequency and intensity at landfall. The NWS report by Ho et al. (1987), DeMaria's extension of the best track, H*Wind analyses (Powell & Houston, 1996, 1998; Powell et al. 1996, 1998) and NOAA Hurricane Research Division aircraft data are used to estimate the radius of maximum winds (RMW) at landfall. The strength of HURDAT is that it is the most complete and accessible historical record for hurricanes making landfall or passing closely by Florida. HURDAT weaknesses include the abbreviated record and questionable intensity estimates for those hurricanes early in the record, especially those that remain offshore. Evidence for the shortness of record is the impact of the last few hurricane seasons on landfall return frequency. The meteorological team has scrutinized the base set developed by the commission and made a number of adjustments to the dataset based on refereed literature and the HURDAT record. I have looked at several of these adjustments in detail and find the corrections to be an improvement over the initial base set.

M-2 Hurricane Characteristics

The model has two main components. The track portion of the model produces a storm with either an initial location or genesis point and an intensity that is derived from an empirical distribution derived from HURDAT (2006). Storm motion and intensity is then initialized by using a Monte Carlo approach, drawing from probability density functions (PDFs) based on the historical

dataset to create a life for a bogus hurricane. Examination of the PDFs reveals that they are faithful to the observed patterns for storms nearing Florida, and the evolution of any particular hurricane appears realistic.

The second component of the meteorological model is the wind field generated for a given hurricane, which only comes into play when the hurricane comes close enough to place high winds over any given ZIP Code of Florida. To generate a wind field the minimum sea-level pressure (MSLP) found in the eye, the RMW at landfall, and a distant environmental pressure (1013 mb) are entered into the Holland (1980) B model for the axisymmetric pressure distribution around the hurricane. The behavior of the RMW is based on a variety of sources that include Ho et al. (1987), DeMaria's extension of the best track data, H*wind analyses, and aircraft reconnaissance radial wind profiles. The B coefficient is based on the extensive aircraft dataset acquired in reconnaissance and research flights over the last few decades. RMW and B use a random or error term to introduce variety into the model. The Holland pressure field is used to produce a gradient wind at the top of the boundary layer. The winds in the boundary layer are estimated following the work proposed by Ooyama (1969) and later utilized by Shapiro (1983) which includes friction and advection effects. These boundary layer winds are reduced to surface winds (10 m) using reduction factors based on the work of Powell et al. (2003). Maximum sustained winds and 3 second gusts are estimated using the guidance of Vickery and Skerlj (2005). Once the hurricane winds come ashore there are further adjustments to the wind to account for local roughness as well as the roughness of the terrain found upstream of the location under scrutiny. The pressure decay of the hurricane is modeled to fit the observations presented by Vickery (2005).

Gradient balance has been demonstrated to be an accurate representation for vortex scale winds above the boundary layer by Willoughby (1990) and is a fine initial condition. The slab boundary layer concept of Ooyama and Shapiro has been shown to produce wind fields much like observed once storm translation and surface friction come into play. The reduction to 10 m altitude is based on Powell et al. (2003); they use the state of the art Global Positioning System sondes to compare surface and boundary layer winds.

Perhaps the most questionable part of the wind portion of the model is the reliance on the estimates of the RMW at landfall. The scatter in RMW for a given MSLP is large; larger RMWs coupled with the B parameter control the size of the annulus of the damaging winds. The typical length of an aircraft leg from the eye is about 150 km so the choice of the B parameter is based on a small radial distance in the majority of hurricanes. The collection of quality wind observations over land in hurricanes remains a daunting task; therefore the actual response of the hurricane winds to variations in roughness is less certain. Applying roughness as a function of ZIP Code is a coarse approximation to reality. However, this is the approach chosen by the commission, and given the data limitations, a reasonable course to take.

M-3 Landfall Intensity

The model uses one minute winds at 10 m elevation to determine intensity at landfall and categorizes each hurricane according to the Saffir-Simpson classification. The model considers any hurricane that makes landfall or comes close enough to place high winds over Florida. Multiple landfalls are accounted for, and decay over land between these landfalls is also estimated. Maximum wind speeds for each category of the Saffir-Simpson scheme are reasonable as is the

worst possible hurricane the model generates. Simulations are conducted for a hypothetical 60,000 years. Any real climate change would alter results, but maybe not as much as have an actual record of order of 1,000 years to base the PDFs on.

M-4 Hurricane Probabilities

Form M-1 demonstrates that the model is simulating the landfalls very well for the entire state, region A (NW Florida) and region B (SW Florida). There are subsections of the state where the historical and the simulated landfalls have a discrepancy. In region C (SE Florida) the observations show an unrealistic bias toward Category 3 storms. This is likely due to an overestimate of intensity for the hurricanes prior to the advent of aircraft sampling or advanced satellite techniques. The historical distribution for region C also does not fit any accepted distributions that we typically see for atmospheric phenomena. This discrepancy is probably due to the shortness of the historical record. I note that other models also have difficulty with this portion of the coast. I believe the modeled distribution, based on tens of thousands of years, is more defensible than the purported standard. Regions D (NE Florida) and E (Georgia) have virtually no distribution to simulate, again pointing to a very short historical record. There is no documented physical reason why these two regions have escaped landfall events. Perhaps a preferred shape of the Bermuda High may bias the situation, but this remains speculative.

M-5 Land Friction and Weakening

Land use and land cover are based on high resolution satellite imagery. Roughness for a particular location is then based on HAZUS tables that assign a roughness to a particular land use. There are newer assessments from other groups but the techniques were not consistently applied throughout the state, nor are the updated HAZUS maps for 2000 available yet. Winds at a particular location are a function of the roughness at that point and conditions upwind. A pressure decay model based on the work of Vickery (2005) produces weakening winds that are reasonable approximations of the observed decay rates of several hurricanes that made landfall in Florida in 2004 and 2005.

The maps (Form M-2) of the 100 year return period maximum sustained winds shows the following trends: (1) a reduction in the sustained winds from south to north, (2) a reduction of winds from coastal to inland ZIP Codes, and (3) the highest winds in the Keys and along the SE and SW coasts. The plotting thresholds requested by the commission partially obfuscate the gradients in wind speed, but Form M-2 produced with finer contours highlights the above trends clearly. The open terrain maps look logical; the actual terrain maps are perhaps overly sensitive to the local roughness. Convective scale motions, which cannot be resolved in this type of model, would probably be responsible for making the winds closer to the open terrain results.

M-6 Logical Relationships of Hurricane Characteristics

The RMW is a crucial but poorly measured variable. Making RMW a function of intensity and latitude explains only a small portion of the variance (~20%). Examination of aircraft reconnaissance radial profiles shows that RMW is highly variable. Currently there are no other schemes available to explain more of the variance. Form M-3 reflects the large range of RMW.

Note that only the more intense hurricanes (MSLP < 940 mb) show a trend, and only with the upper part of the range. Even open ocean studies of the RMW show such large scatter.

Tests done during my visits show that wind speed decreases as a function of roughness, all other variables being held constant. The evolution of the wind field as a hurricane comes ashore is logical.

Summary

The consortium that has assembled the meteorological portion of the Public Model for Hurricane Wind Losses for the State of Florida is using the HURDAT with corrections based on other refereed literature. These data yield a series of probability density functions that describe frequency, location, and intensity at landfall. Once a hurricane reaches close enough to the coast the gradient winds are estimated using the equations by Holland (1980), then a sophisticated wind model (Ooyama 1969, Shapiro 1983) is applied to calculate the boundary layer winds. Reduction of this wind to a surface value is based on recent boundary layer theory and observations. Here the consortium has exploited other sources of data (e.g., NOAA/AOML/HRD aircraft wind profiles and GPS sondes) to produce a surface wind field. As the wind field transitions from marine to land exposure changes in roughness are taken into account. Form M-1 (frequency and category at landfall as a function of coastal segment) and Form M-2 (100 year return maximum sustained winds for Florida) highlight the good performance of the model.

I suspect that the differences between the historical record and the simulation are largely due to the shortness and uncertainty of the record. If the consortium had the luxury of 1000 years of observations agreement between the record and the simulation would be improved. I believe that the meteorological portion of the model is meeting all the standards established by the commission. Tests of the model against H*Wind analyses and the production of wind speed swaths go beyond the typical quality controls of prior models and demonstrate that this model is worthy of consideration by the commission.

AMI Risk Consultants, Inc.
Actuarial & Risk Management Consulting Services
1336 SW 146th Ct, Miami, Florida 33184, USA Tel No: (305)273-1589 Fax No:(305)330-5427 www.amirisk.com

October 28, 2014

Dr. Shahid Hamid
Professor of
Finance,
Department of
Finance, CBA
and International Hurricane
Research Center Florida
International University, RB 202
B Miami, FL 33199

Re: Florida Public Hurricane Loss Model
Version 6.0
Independent Actuarial Review

Dear Dr. Hamid:

AMI Risk Consultants, Inc. was engaged by the International Hurricane Research Center ("IHRC") at Florida International University ("FIU") to review the actuarial components of its hurricane model, ***Florida Public Hurricane Loss Model, Version 6.0***. I am a Fellow of the Casualty Actuarial Society, a Member of the American Academy of Actuaries, and have more than twenty-five years of actuarial experience in the property/casualty insurance industry. I am an employee of the actuarial consulting firm AMI Risk Consultants, Inc.

It is my understanding that between Versions 5.0 and 6.0 there were significant changes to the Florida Public Hurricane Loss Model ("FPHLM"). Those changes included:

- Updates to HURDAT, ZIP codes and land use/land cover data.
- Revisions to the Rmax database.
- Minor revisions to the Mid/high-rise Commercial Residential vulnerabilities to include water penetration at the top story.
- Numerous additions, refinements and modifications to the Commercial Residential vulnerabilities, including updated statistics used to weight the vulnerability matrices.

My review is based the IHRC's November 2014 model submission to the Commission. I revisited each of the Actuarial Standards, and have the following comments:

Standard A-1: I reviewed the revised data input record formats for Personal and Commercial Residential policies.

Standard A-2: Although Version 6.0 incorporates a new set of stochastic storms, the criteria for inclusion/exclusion have not changed, and the computer code categorizing each storm is also unchanged.

Standard A-3: There were major refinements to the vulnerability component of the Commercial Residential model for low-rise exposures, and minor revisions for mid/high- rise exposures. The basic approach to estimating loss costs by coverage, however, has not changed in this version of the model for either Personal or Commercial Residential.

Standard A-4: The treatment of the items detailed in this standard, such as expenses, inflation, storm surge, geocoding, and demand surge has not changed with this version of the model.

Standard A-5: The methods used by the model to reflect the impact of deductibles and policy limits on losses have not changed since the prior submission.

Standard A-6: I tested the loss costs for compliance with this standard. I examined Forms A-1, A-2, A-3, A-4 and A-8 for reasonability, and compared the results to the prior submission. I examined loss cost changes by county, separating the impacts of meteorology and engineering revisions. Larger positive and negative changes were examined at the zip code level. There were significant loss cost changes for a number of counties, in particular for low-rise commercial residential due to the vulnerability modifications.

I identified the anomalies in Form A-6, and determined the reason for each.

I reviewed Form A-8 for overall reasonability and verified the return time and uncertainty intervals with manual calculations for a sample of entries.

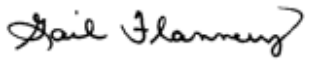
I tested loss costs at the zip code level in instances where compliance could not be verified from the weighted averages in Form A-4.

Conclusion:

My conclusion is that the Florida Public Hurricane Model v6.0 reflects reasonable actuarial assumptions, and meets the Commission's Standards A-1 through A-6.

If you have any questions about my review, I would be happy to discuss them.

Sincerely,



Gail Flannery, FCAS, MAAA
Consulting Actuary

Appendix B – Form A-1: Zero Deductible Personal Residential Loss Costs by ZIP Code

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
November 1, 2014

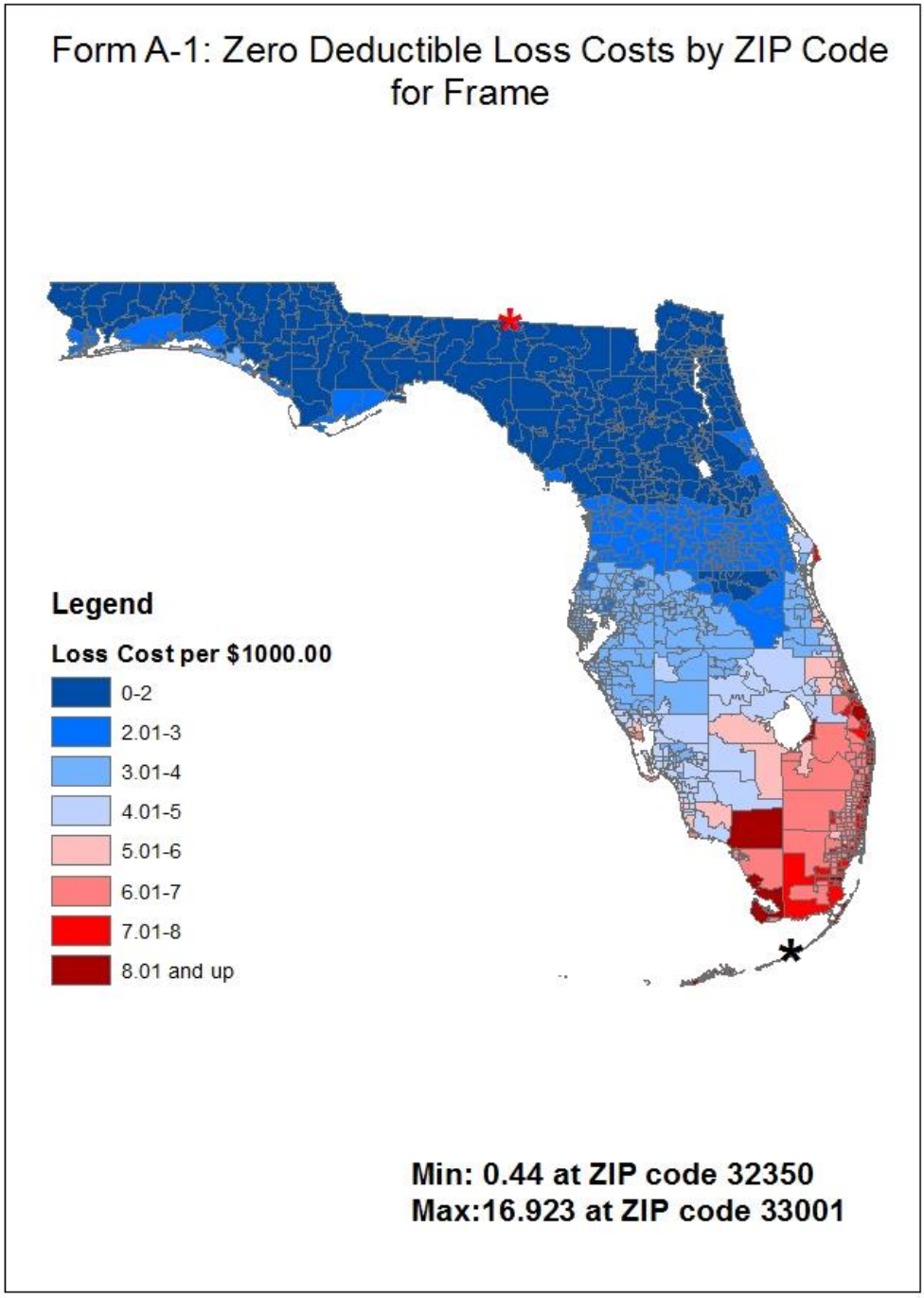


Figure 91. Zero deductible loss costs by ZIP code for frame.

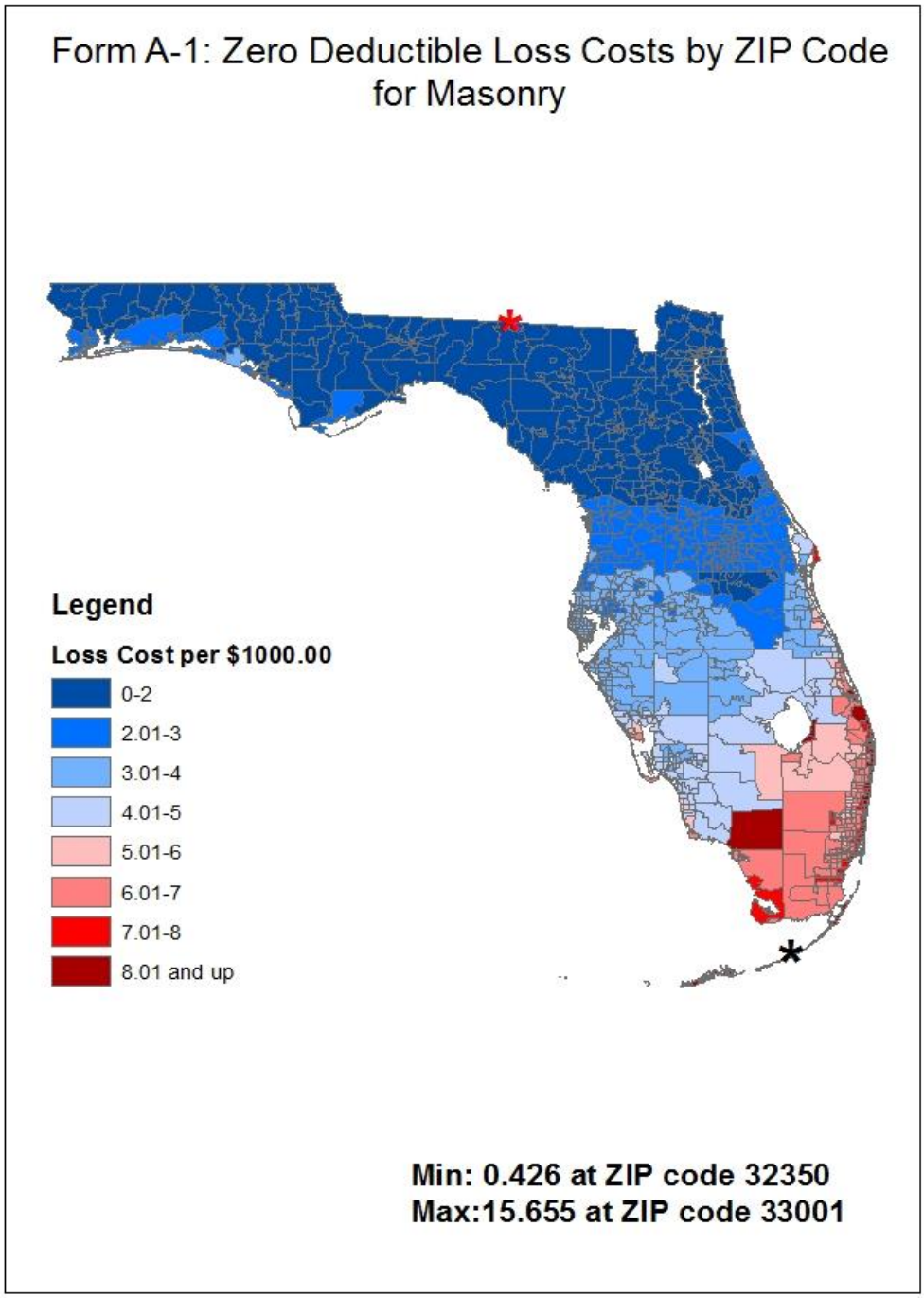


Figure 92. Zero deductible loss costs by ZIP code for masonry.

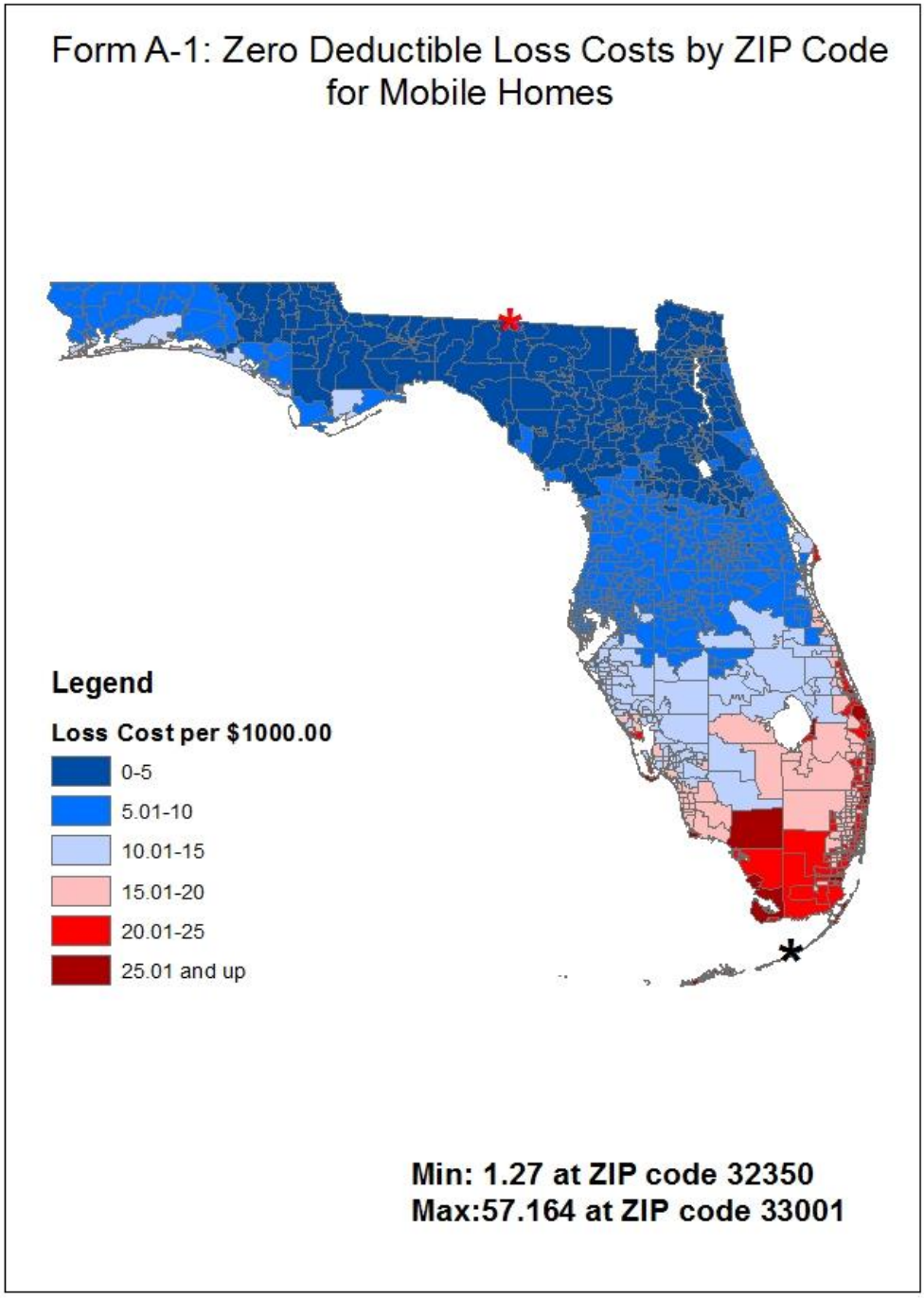


Figure 93. Zero deductible loss costs by ZIP code for mobile homes.

Appendix C – Form A-2: Base Hurricane Storm Set Statewide Loss Costs

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
~~November 1, 2014~~March 15, 2015

Form A-2: Base Hurricane Storm Set Statewide Losses

Florida International University
Florida Public Hurricane Loss Model 6.0
November 1, 2014

				2007 FHCF Exposure Data		2012 FHCF Exposure Data	
ID	Landfall/Closest Approach Date	Year	Name	Personal and Commercial Residential Insured Losses ($)	Dollar Contribution	Personal and Commercial Residential Insured Losses ($)	Dollar Contribution
005	8/10/1901	1901	NoName04-1901	334,142,523	2,931,075	336,515,263	2,951,888
010	9/11/1903	1903	NoName03-1903	10,478,690,354	91,918,336	10,546,830,510	92,516,057
015	10/16/1904	1904	NoName04-1904	3,741,037,925	32,816,122	3,767,409,130	33,047,449
020	6/16/1906	1906	NoName02-1906	3,765,234,507	33,028,373	3,868,016,719	33,929,971
025	9/25/1906	1906	NoName06-1906	882,266,546	7,739,180	822,038,687	7,210,866
030	10/18/1906	1906	NoName08-1906	19,697,999,372	172,789,468	19,503,548,973	171,083,763
035	10/11/1909	1909	NoName11-1909	1,005,041,352	8,816,152	1,045,615,473	9,172,066
040	10/17/1910	1910	NoName05-1910	32,605,248,200	286,010,949	30,294,689,315	265,742,889
045	8/8/1911	1911	NoName02-1911	336,620,205	2,952,809	369,542,989	3,241,605
050	9/10/1912	1912	NoName04-1912	53,591,521	470,101	49,788,424	436,741
055	7/31/1915	1915	NoName01-1915	972,833,486	8,533,627	860,573,538	7,548,891
060	9/3/1915	1915	NoName04-1915	439,047,793	3,851,296	416,172,869	3,650,639
065	7/4/1916	1916	NoName02-1916	540,317,188	4,739,624	530,438,106	4,652,966
070	10/17/1916	1916	NoName14-1916	1,214,633,892	10,654,683	1,102,985,978	9,675,316
075	9/26/1917	1917	NoName04-1917	1,957,489,564	17,170,961	1,737,395,516	15,240,312
080	9/9/1919	1919	NoName02-1919	248,763,639	2,182,137	220,043,173	1,930,203
085	10/24/1921	1921	TampaBay06-1921	22,308,585,607	195,689,347	20,250,489,162	177,635,870
090	9/13/1924	1924	NoName05-1924	37,587,644	329,716	32,563,981	285,649
095	10/20/1924	1924	NoName10-1924	8,259,612,152	72,452,738	7,996,276,840	70,142,779
100	7/27/1926	1926	NoName01-1926	4,549,024,075	39,903,720	3,887,479,897	34,100,701
105	9/18/1926	1926	GreatMiami07-1926	46,328,879,414	406,393,679	43,412,183,415	380,808,626
110	10/20/1926	1926	NoName10-1926	3,641,378,114	31,941,913	3,491,560,445	30,627,723
115	8/7/1928	1928	NoName01-1928	4,546,850,817	39,884,656	4,368,948,257	38,324,108
120	9/16/1928	1928	LakeOkeechobee04-1928	50,116,920,589	439,622,110	46,299,369,108	406,134,817
125	9/27/1929	1929	NoName02-1929	14,514,720,221	127,322,107	13,909,345,611	122,011,804
130	8/29/1932	1932	NoName03-1932	2,101,266,085	18,432,159	2,229,846,757	19,560,059
135	7/30/1933	1933	NoName05-1933	1,095,721,945	9,611,596	1,193,759,140	10,471,571
140	9/3/1933	1933	NoName11-1933	13,519,017,558	118,587,873	12,880,382,372	112,985,810
145	9/2/1935	1935	LaborDay03-1935	22,750,969,717	199,569,910	19,938,791,775	174,901,682
150	11/4/1935	1935	NoName07-1935	7,921,622,010	69,487,912	7,683,250,877	67,396,938
155	7/28/1936	1936	NoName05-1936	2,196,734,231	19,269,599	2,272,901,368	19,937,731
160	8/11/1939	1939	NoName02-1939	3,486,626,025	30,584,439	3,338,694,579	29,286,795
165	10/6/1941	1941	NoName05-1941	9,733,253,873	85,379,420	9,361,960,260	82,122,458

Form A-2: Base Hurricane Storm Set Statewide Losses

Florida International University
Florida Public Hurricane Loss Model 6.0
November 1, 2014

				2007 FHCF Exposure Data		2012 FHCF Exposure Data	
ID	Landfall/Closest Approach Date	Year	Name	Personal and Commercial Residential Insured Losses ($)	Dollar Contribution	Personal and Commercial Residential Insured Losses ($)	Dollar Contribution
170	10/18/1944	1944	NoName13-1944	28,809,014,952	252,710,657	26,607,730,318	233,401,143
175	6/22/1945	1945	NoName01-1945	6,361,935,035	55,806,448	6,511,707,502	57,120,241
180	9/15/1945	1945	NoName09-1945	18,845,663,248	165,312,836	17,857,775,934	156,647,157
185	10/7/1946	1946	NoName06-1946	14,522,047,133	127,386,378	13,737,640,709	120,505,620
190	9/17/1947	1947	NoName04-1947	28,142,722,248	246,865,985	26,634,196,286	233,633,301
195	10/11/1947	1947	NoName09-1947	9,126,005,122	80,052,677	8,831,064,446	77,465,478
200	9/21/1948	1948	NoName08-1948	14,371,116,036	126,062,421	14,088,054,453	123,579,425
205	10/5/1948	1948	NoName09-1948	8,898,839,378	78,059,995	8,899,201,863	78,063,174
210	8/26/1949	1949	NoName02-1949	33,651,912,419	295,192,214	31,711,206,335	278,168,477
215	8/29/1950	1950	Baker-1950	606,901,754	5,323,700	579,044,443	5,079,337
220	9/3/1950	1950	Easy-1950	10,176,454,948	89,267,149	9,772,277,066	85,721,729
225	10/17/1950	1950	King-1950	21,211,622,940	186,066,868	20,533,530,222	180,118,686
230	9/25/1953	1953	Florence-1953	381,383,117	3,345,466	313,931,073	2,753,781
235	9/24/1956	1956	Flossy-1956	810,935,727	7,113,471	719,783,654	6,313,892
240	9/9/1960	1960	Donna-1960	23,614,172,363	207,141,863	20,999,063,239	184,202,309
241	9/14/1960	1960	Ethel-1960	301	3	233	2
245	8/26/1964	1964	Cleo-1964	16,255,229,204	142,589,730	15,794,383,362	138,547,222
250	9/9/1964	1964	Dora-1964	4,199,644,880	36,838,990	4,016,541,448	35,232,820
255	10/14/1964	1964	Isbell-1964	9,848,953,117	86,394,326	10,038,113,741	88,053,629
260	9/7/1965	1965	Betsy-1965	9,587,626,030	84,101,983	9,505,638,528	83,382,794
265	6/8/1966	1966	Alma-1966	14,174,701,192	124,339,484	13,517,964,712	118,578,638
270	10/2/1966	1966	Inez-1966	350,731,992	3,076,596	335,736,138	2,945,054
275	10/16/1968	1968	Gladys-1968	5,448,467,752	47,793,577	5,165,474,205	45,311,177
280	6/18/1972	1972	Agnes-1972	117,435,997	1,030,140	100,160,757	878,603
285	9/22/1975	1975	Eloise-1975	1,493,385,837	13,099,876	1,131,789,873	9,927,981
290	9/3/1979	1979	David-1979	9,991,920,956	87,648,429	9,600,986,838	84,219,183
295	9/11/1979	1979	Frederic-1979	1,207,700,463	10,593,864	1,085,315,440	9,520,311
300	8/29/1985	1985	Elena-1985	192,665,827	1,690,051	193,943,587	1,701,260
305	11/20/1985	1985	Kate-1985	451,555,233	3,961,011	434,141,678	3,808,260
310	10/12/1987	1987	Floyd-1987	283,780,602	2,489,304	292,996,476	2,570,145
315	8/24/1992	1992	Andrew-1992	20,617,950,634	180,859,216	19,960,320,629	175,090,532
320	8/1/1995	1995	Erin-1995	5,164,875,698	45,305,927	4,983,555,875	43,715,402
325	10/3/1995	1995	Opal-1995	3,451,965,023	30,280,395	2,926,798,993	25,673,675

Form A-2: Base Hurricane Storm Set Statewide Losses

Florida International University
Florida Public Hurricane Loss Model 6.0
November 1, 2014

				2007 FHCF Exposure Data		2012 FHCF Exposure Data	
ID	Landfall/Closest Approach Date	Year	Name	Personal and Commercial Residential Insured Losses ($)	Dollar Contribution	Personal and Commercial Residential Insured Losses ($)	Dollar Contribution
330	7/17/1997	1997	Danny-1997	84,759,542	743,505	70,927,574	622,172
335	9/1/1998	1998	Earl-1998	19,082,228	167,388	9,505,772	83,384
340	9/25/1998	1998	Georges-1998	949,465,831	8,328,648	987,965,794	8,666,367
345	10/15/1999	1999	Irene-1999	5,718,374,290	50,161,178	5,927,896,358	51,999,091
350	8/13/2004	2004	Charley-2004	8,245,308,579	72,327,268	7,150,288,824	62,721,832
355	9/4/2004	2004	Frances-2004	12,247,091,671	107,430,629	12,190,191,364	106,931,503
360	9/14/2004	2004	Ivan-2004	714,395,662	6,266,629	670,485,308	5,881,450
365	9/25/2004	2004	Jeanne-2004	14,153,386,059	124,152,509	13,167,376,296	115,503,301
370	7/7/2005	2005	Dennis-2005	952,126,920	8,351,991	900,579,385	7,899,819
375	8/24/2005	2005	Katrina-2005	4,683,073,208	41,079,590	4,633,735,654	40,646,804
380	10/20/2005	2005	Wilma-2005	18,993,719,933	166,611,578	18,267,879,540	160,244,557
385	9/10/2008	2008	Ike-2008	41,274	362	83,260	730
	Total			684,511,870,498	6,004,490,092	648,876,393,691	5,691,898,190

Form A-2: Base Hurricane Storm Set Statewide Losses

Florida International University
Florida Public Hurricane Loss Model 6.1
March 15, 2015

				2007 FHCF Exposure Data		2012 FHCF Exposure Data	
ID	Landfall/Closest Approach Date	Year	Name	Personal and Commercial Residential Insured Losses ($)	Dollar Contribution	Personal and Commercial Residential Insured Losses ($)	Dollar Contribution
005	8/10/1901	1901	NoName04-1901	333,435,307	2,924,871	336,136,230	2,948,563
010	9/11/1903	1903	NoName03-1903	10,439,468,271	91,574,283	10,499,955,766	92,104,875
015	10/16/1904	1904	NoName04-1904	3,709,792,683	32,542,041	3,730,155,872	32,720,666
020	6/16/1906	1906	NoName02-1906	3,742,888,559	32,832,356	3,839,569,825	33,680,437
025	9/25/1906	1906	NoName06-1906	879,375,759	7,713,822	820,553,845	7,197,841
030	10/18/1906	1906	NoName08-1906	19,751,809,782	173,261,489	19,546,527,668	171,460,769
035	10/11/1909	1909	NoName11-1909	1,003,952,819	8,806,604	1,043,258,469	9,151,390
040	10/17/1910	1910	NoName05-1910	32,522,154,496	285,282,057	30,199,758,393	264,910,161
045	8/8/1911	1911	NoName02-1911	336,180,781	2,948,954	369,225,333	3,238,819
050	9/10/1912	1912	NoName04-1912	53,404,286	468,459	49,655,256	435,572
055	7/31/1915	1915	NoName01-1915	970,585,281	8,513,906	858,991,225	7,535,011
060	9/3/1915	1915	NoName04-1915	438,663,284	3,847,924	416,035,695	3,649,436
065	7/4/1916	1916	NoName02-1916	538,782,707	4,726,164	529,334,225	4,643,283
070	10/17/1916	1916	NoName14-1916	1,210,045,906	10,614,438	1,100,363,284	9,652,310
075	9/26/1917	1917	NoName04-1917	1,949,995,489	17,105,224	1,732,549,669	15,197,804
080	9/9/1919	1919	NoName02-1919	248,740,938	2,181,938	219,271,068	1,923,430
085	10/24/1921	1921	TampaBay06-1921	22,303,558,825	195,645,253	20,248,475,470	177,618,206
090	9/13/1924	1924	NoName05-1924	37,566,345	329,529	32,553,820	285,560
095	10/20/1924	1924	NoName10-1924	8,207,697,754	71,997,349	7,933,612,307	69,593,090
100	7/27/1926	1926	NoName01-1926	4,549,106,341	39,904,442	3,884,966,382	34,078,652
105	9/18/1926	1926	GreatMiami07-1926	46,513,902,476	408,016,688	43,527,399,131	381,819,291
110	10/20/1926	1926	NoName10-1926	3,625,570,686	31,803,252	3,470,283,771	30,441,086
115	8/7/1928	1928	NoName01-1928	4,539,651,185	39,821,502	4,360,557,129	38,250,501
120	9/16/1928	1928	LakeOkeechobee04-1928	50,100,463,199	439,477,747	46,281,441,115	405,977,554
125	9/27/1929	1929	NoName02-1929	14,440,975,483	126,675,224	13,828,676,352	121,304,179
130	8/29/1932	1932	NoName03-1932	2,080,330,393	18,248,512	2,203,956,005	19,332,947
135	7/30/1933	1933	NoName05-1933	1,092,389,212	9,582,362	1,190,065,769	10,439,173
140	9/3/1933	1933	NoName11-1933	13,536,879,759	118,744,559	12,904,672,302	113,198,880
145	9/2/1935	1935	LaborDay03-1935	22,799,729,514	199,997,627	19,950,713,496	175,006,259
150	11/4/1935	1935	NoName07-1935	7,860,845,215	68,954,783	7,610,824,601	66,761,619
155	7/28/1936	1936	NoName05-1936	2,184,468,503	19,162,004	2,260,383,359	19,827,924
160	8/11/1939	1939	NoName02-1939	3,481,165,735	30,536,542	3,333,500,389	29,241,231
165	10/6/1941	1941	NoName05-1941	9,688,396,550	84,985,935	9,315,753,944	81,717,140

Form A-2: Base Hurricane Storm Set Statewide Losses

Florida International University
Florida Public Hurricane Loss Model 6.1
March 15, 2015

				2007 FHCF Exposure Data		2012 FHCF Exposure Data	
ID	Landfall/Closest Approach Date	Year	Name	Personal and Commercial Residential Insured Losses ($)	Dollar Contribution	Personal and Commercial Residential Insured Losses ($)	Dollar Contribution
170	10/18/1944	1944	NoName13-1944	28,763,980,864	252,315,622	26,564,630,986	233,023,079
175	6/22/1945	1945	NoName01-1945	6,355,765,144	55,752,326	6,505,062,313	57,061,950
180	9/15/1945	1945	NoName09-1945	18,800,524,325	164,916,880	17,813,540,663	156,259,129
185	10/7/1946	1946	NoName06-1946	14,503,512,475	127,223,794	13,720,943,673	120,359,155
190	9/17/1947	1947	NoName04-1947	28,092,662,143	246,426,861	26,584,059,545	233,193,505
195	10/11/1947	1947	NoName09-1947	9,064,078,596	79,509,461	8,755,844,027	76,805,649
200	9/21/1948	1948	NoName08-1948	14,317,775,581	125,594,523	14,026,392,230	123,038,528
205	10/5/1948	1948	NoName09-1948	8,881,823,457	77,910,732	8,879,507,504	77,890,417
210	8/26/1949	1949	NoName02-1949	33,633,104,921	295,027,236	31,690,377,474	277,985,767
215	8/29/1950	1950	Baker-1950	605,253,684	5,309,243	577,981,386	5,070,012
220	9/3/1950	1950	Easy-1950	10,166,399,053	89,178,939	9,762,896,794	85,639,446
225	10/17/1950	1950	King-1950	21,307,417,871	186,907,174	20,603,128,522	180,729,198
230	9/25/1953	1953	Florence-1953	379,591,244	3,329,748	313,198,155	2,747,352
235	9/24/1956	1956	Flossy-1956	806,862,937	7,077,745	717,757,062	6,296,115
240	9/9/1960	1960	Donna-1960	23,580,620,883	206,847,552	20,951,588,528	183,785,864
241	9/14/1960	1960	Ethel-1960	301	3	233	2
245	8/26/1964	1964	Cleo-1964	16,278,609,897	142,794,824	15,801,247,907	138,607,438
250	9/9/1964	1964	Dora-1964	4,194,026,534	36,789,706	4,012,707,444	35,199,188
255	10/14/1964	1964	Isbell-1964	9,807,107,483	86,027,259	9,994,229,786	87,668,682
260	9/7/1965	1965	Betsy-1965	9,532,670,535	83,619,917	9,444,922,717	82,850,199
265	6/8/1966	1966	Alma-1966	14,128,560,190	123,934,739	13,468,879,226	118,148,063
270	10/2/1966	1966	Inez-1966	349,985,199	3,070,046	334,797,063	2,936,816
275	10/16/1968	1968	Gladys-1968	5,445,611,429	47,768,521	5,159,701,709	45,260,541
280	6/18/1972	1972	Agnes-1972	117,401,074	1,029,834	100,147,182	878,484
285	9/22/1975	1975	Eloise-1975	1,490,626,106	13,075,668	1,130,303,373	9,914,942
290	9/3/1979	1979	David-1979	9,967,202,314	87,431,599	9,568,883,914	83,937,578
295	9/11/1979	1979	Frederic-1979	1,202,148,130	10,545,159	1,082,086,904	9,491,990
300	8/29/1985	1985	Elena-1985	192,363,662	1,687,401	193,729,939	1,699,385
305	11/20/1985	1985	Kate-1985	451,179,586	3,957,716	433,986,075	3,806,895
310	10/12/1987	1987	Floyd-1987	283,314,206	2,485,212	292,373,062	2,564,676
315	8/24/1992	1992	Andrew-1992	20,663,642,338	181,260,021	20,011,470,463	175,539,215
320	8/1/1995	1995	Erin-1995	5,157,161,135	45,238,256	4,977,551,280	43,662,731
325	10/3/1995	1995	Opal-1995	3,449,216,551	30,256,286	2,921,257,845	25,625,069

				2007 FHCF Exposure Data		2012 FHCF Exposure Data	
ID	Landfall/Closest Approach Date	Year	Name	Personal and Commercial Residential Insured Losses ($)	Dollar Contribution	Personal and Commercial Residential Insured Losses ($)	Dollar Contribution
330	7/17/1997	1997	Danny-1997	84,523,709	741,436	70,784,237	620,914
335	9/1/1998	1998	Earl-1998	19,076,413	167,337	9,506,715	83,392
340	9/25/1998	1998	Georges-1998	941,366,739	8,257,603	977,879,820	8,577,893
345	10/15/1999	1999	Irene-1999	5,683,253,353	49,853,100	5,884,180,999	51,615,623
350	8/13/2004	2004	Charley-2004	8,235,249,333	72,239,029	7,141,910,923	62,648,341
355	9/4/2004	2004	Frances-2004	12,221,956,719	107,210,147	12,162,419,477	106,687,890
360	9/14/2004	2004	Ivan-2004	712,018,497	6,245,776	668,927,083	5,867,781
365	9/25/2004	2004	Jeanne-2004	14,142,247,465	124,054,802	13,153,086,414	115,377,951
370	7/7/2005	2005	Dennis-2005	949,105,991	8,325,491	898,741,811	7,883,700
375	8/24/2005	2005	Katrina-2005	4,648,092,442	40,772,741	4,592,644,961	40,286,359
380	10/20/2005	2005	Wilma-2005	18,895,819,645	165,752,804	18,154,872,511	159,253,268
385	9/10/2008	2008	Ike-2008	41,274	362	83,260	730
	Total			683,674,924,952	5,997,148,464	647,739,400,362	5,681,924,565

Appendix D – Form A-3: Cumulative Losses from the 2004 Hurricane Season

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
~~November 1, 2014~~ March 15, 2015

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34950	0	0.00%	19,103,050	0.16%	0	0.00%	22,571,623	0.16%	41,674,727	0.12%
34667	0	0.00%	39,684,929	0.33%	0	0.00%	54,161,842	0.38%	93,846,771	0.27%
32686	0	0.00%	4,385,437	0.04%	0	0.00%	4,402,327	0.03%	8,787,764	0.03%
33960	0	0.00%	681,113	0.01%	0	0.00%	1,124,336	0.01%	2,218,684	0.01%
32828	106,047,246	1.29%	45,802,511	0.38%	0	0.00%	78,848,532	0.56%	230,698,290	0.66%
34102	0	0.00%	41,512,673	0.34%	0	0.00%	0	0.00%	41,512,861	0.12%
34951	0	0.00%	42,992,561	0.35%	0	0.00%	58,942,254	0.42%	101,934,825	0.29%
32970	0	0.00%	0	0.00%	0	0.00%	0	0.00%	708,620	0.00%
34668	0	0.00%	26,471,792	0.22%	0	0.00%	46,083,441	0.33%	72,555,233	0.21%
32829	34,025,151	0.41%	14,414,038	0.12%	0	0.00%	23,449,896	0.17%	71,889,085	0.20%
34103	0	0.00%	46,481,045	0.38%	0	0.00%	24,813,288	0.18%	71,295,424	0.20%
34952	0	0.00%	95,119,918	0.78%	0	0.00%	107,392,671	0.76%	202,512,730	0.58%
33820	608,226	0.01%	0	0.00%	0	0.00%	0	0.00%	1,153,882	0.00%
34669	0	0.00%	10,805,821	0.09%	0	0.00%	6,877,626	0.05%	17,683,448	0.05%
32547	0	0.00%	0	0.00%	20,487,938	2.87%	0	0.00%	20,487,938	0.06%
32830	0	0.00%	0	0.00%	0	0.00%	0	0.00%	1,079,658	0.00%
34953	0	0.00%	106,041, 759	0.87%	0	0.00%	114,996,976	0.82%	221,038,736	0.63%
33538	0	0.00%	5,885,524	0.05%	0	0.00%	5,908,922	0.04%	11,794,446	0.03%
32548	0	0.00%	0	0.00%	26,211,974	3.67%	0	0.00%	26,211,974	0.07%
32407	0	0.00%	0	0.00%	6,841,582	0.96%	0	0.00%	6,841,628	0.02%
34105	0	0.00%	7,117,844	0.06%	0	0.00%	0	0.00%	7,118,375	0.02%
32124	5,159,109	0.06%	3,547,410	0.03%	0	0.00%	5,209,862	0.04%	13,916,381	0.04%
32266	0	0.00%	7,136,273	0.06%	0	0.00%	7,162,452	0.05%	14,298,724	0.04%
32832	29,405,905	0.36%	17,333,137	0.14%	0	0.00%	24,042,511	0.17%	70,781,553	0.20%
32408	0	0.00%	0	0.00%	8,876,249	1.24%	0	0.00%	8,876,488	0.03%
33823	16,447,490	0.20%	0	0.00%	0	0.00%	37,139,918	0.26%	53,996,966	0.15%
33540	0	0.00%	4,500,085	0.04%	0	0.00%	7,623,029	0.05%	12,123,114	0.03%
32550	0	0.00%	0	0.00%	13,217,829	1.85%	0	0.00%	13,217,829	0.04%
32833	16,279,725	0.20%	7,355,022	0.06%	0	0.00%	13,694,223	0.10%	37,328,970	0.11%
34956	0	0.00%	6,829,207	0.06%	0	0.00%	6,855,395	0.05%	13,684,603	0.04%
32692	0	0.00%	781,217	0.01%	0	0.00%	0	0.00%	781,217	0.00%
33541	0	0.00%	10,284,593	0.08%	0	0.00%	16,801,253	0.12%	27,085,848	0.08%
33966	10,696,686	0.13%	0	0.00%	0	0.00%	0	0.00%	10,696,704	0.03%
34108	21,934,667	0.27%	58,231,937	0.48%	0	0.00%	0	0.00%	80,167,375	0.23%
34957	0	0.00%	59,036,288	0.48%	0	0.00%	59,205,209	0.42%	118,241,572	0.34%
32127	71,743,996	0.87%	72,155,513	0.59%	0	0.00%	87,922,006	0.62%	231,821,515	0.66%
32976	0	0.00%	30,697,591	0.25%	0	0.00%	95,378,176	0.68%	126,075,776	0.36%

ZIP Code	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33825	44,180,190	0.54%	13,244,113	0.11%	0	0.00%	38,304,860	0.27%	95,729,163	0.27%
32693	0	0.00%	3,414,376	0.03%	0	0.00%	0	0.00%	3,474,895	0.01%
33542	0	0.00%	9,021,880	0.07%	0	0.00%	15,802,067	0.11%	24,823,964	0.07%
33401	0	0.00%	55,999,693	0.46%	0	0.00%	41,934,519	0.30%	97,934,823	0.28%
33967	13,290,394	0.16%	0	0.00%	0	0.00%	0	0.00%	13,290,394	0.04%
32835	66,709,719	0.81%	31,571,535	0.26%	0	0.00%	67,267,447	0.48%	165,548,702	0.47%
34109	0	0.00%	8,999,771	0.07%	0	0.00%	0	0.00%	9,000,078	0.03%
32128	43,864,705	0.53%	18,845,403	0.15%	0	0.00%	36,911,762	0.26%	99,621,870	0.28%
32694	0	0.00%	0	0.00%	0	0.00%	702,284	0.01%	702,284	0.00%
33543	0	0.00%	26,343,040	0.22%	0	0.00%	35,769,456	0.25%	62,112,503	0.18%
34251	0	0.00%	0	0.00%	0	0.00%	4,954,577	0.04%	4,954,577	0.01%
32836	48,971,137	0.60%	24,296,704	0.20%	0	0.00%	59,628,318	0.42%	132,896,159	0.38%
34110	13,844,283	0.17%	13,923,648	0.11%	0	0.00%	0	0.00%	27,768,349	0.08%
32129	35,551,190	0.43%	21,457,674	0.18%	0	0.00%	29,366,636	0.21%	86,375,500	0.25%
33827	9,508,783	0.12%	3,940,728	0.03%	0	0.00%	6,568,919	0.05%	20,018,430	0.06%
33544	0	0.00%	18,205,793	0.15%	0	0.00%	24,402,034	0.17%	42,607,827	0.12%
33403	0	0.00%	18,417,024	0.15%	0	0.00%	13,416,711	0.10%	31,833,791	0.09%
32837	109,585,372	1.33%	35,635,982	0.29%	0	0.00%	80,016,714	0.57%	225,238,068	0.64%
32413	0	0.00%	0	0.00%	29,901,353	4.19%	0	0.00%	29,901,883	0.08%
32130	0	0.00%	0	0.00%	0	0.00%	3,197,792	0.02%	3,253,570	0.01%
34677	0	0.00%	21,461,602	0.18%	0	0.00%	3,202,207	0.02%	24,663,809	0.07%
32696	0	0.00%	5,333,083	0.04%	0	0.00%	4,577,873	0.03%	9,910,956	0.03%
33404	0	0.00%	64,401,303	0.53%	0	0.00%	42,671,506	0.30%	107,073,648	0.30%
33971	32,326,221	0.39%	0	0.00%	0	0.00%	0	0.00%	32,326,221	0.09%
32839	32,477,540	0.40%	11,709,354	0.10%	0	0.00%	26,992,807	0.19%	71,179,702	0.20%
33405	0	0.00%	40,968,979	0.34%	0	0.00%	21,270,258	0.15%	62,239,516	0.18%
32132	14,213,189	0.17%	14,296,466	0.12%	0	0.00%	17,523,893	0.12%	46,033,548	0.13%
33830	21,783,831	0.27%	515,066	0.00%	0	0.00%	30,117,719	0.21%	52,416,616	0.15%
33547	0	0.00%	0	0.00%	0	0.00%	17,207,117	0.12%	17,498,074	0.05%
33972	11,064,691	0.13%	0	0.00%	0	0.00%	0	0.00%	11,402,061	0.03%
33406	0	0.00%	39,596,640	0.33%	0	0.00%	23,518,387	0.17%	63,115,065	0.18%
33548	0	0.00%	6,151,334	0.05%	0	0.00%	6,175,953	0.04%	12,327,287	0.04%
33407	0	0.00%	40,995,273	0.34%	0	0.00%	29,929,038	0.21%	70,924,442	0.20%
32134	0	0.00%	0	0.00%	0	0.00%	3,448,813	0.02%	3,513,462	0.01%
34681	0	0.00%	2,213,618	0.02%	0	0.00%	2,222,471	0.02%	4,436,089	0.01%
33549	0	0.00%	17,883,587	0.15%	0	0.00%	17,953,026	0.13%	35,836,613	0.10%
33408	0	0.00%	96,476,036	0.79%	0	0.00%	77,066,547	0.55%	173,543,240	0.49%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32135	1,051,272	0.01%	894,527	0.01%	0	0.00%	1,055,354	0.01%	3,001,153	0.01%
32701	34,758,551	0.42%	15,404,100	0.13%	0	0.00%	25,805,295	0.18%	75,967,946	0.22%
33409	0	0.00%	40,871,999	0.34%	0	0.00%	23,170,620	0.16%	64,042,771	0.18%
32136	24,837,256	0.30%	20,663,230	0.17%	0	0.00%	25,059,945	0.18%	70,560,431	0.20%
33834	3,648,905	0.04%	0	0.00%	0	0.00%	3,684,995	0.03%	7,367,393	0.02%
34683	0	0.00%	34,427,754	0.28%	0	0.00%	34,530,487	0.25%	68,958,241	0.20%
32702	0	0.00%	0	0.00%	0	0.00%	1,270,125	0.01%	1,270,125	0.00%
32561	0	0.00%	0	0.00%	48,484,546	6.79%	0	0.00%	48,484,546	0.14%
33410	0	0.00%	108,972,402	0.89%	0	0.00%	95,878,564	0.68%	204,851,220	0.58%
32137	7,128,904	0.09%	53,754,148	0.44%	0	0.00%	76,984,559	0.55%	137,867,611	0.39%
34684	0	0.00%	22,862,208	0.19%	0	0.00%	4,384,609	0.03%	27,246,818	0.08%
32703	26,052,853	0.32%	801,083	0.01%	0	0.00%	45,043,975	0.32%	71,897,911	0.20%
33411	0	0.00%	136,604,090	1.12%	0	0.00%	89,148,854	0.63%	225,753,140	0.64%
33836	0	0.00%	0	0.00%	0	0.00%	0	0.00%	589,901	0.00%
34685	0	0.00%	20,785,168	0.17%	0	0.00%	1,605,540	0.01%	22,390,708	0.06%
32563	0	0.00%	0	0.00%	31,299,846	4.38%	0	0.00%	31,299,846	0.09%
33412	0	0.00%	53,696,662	0.44%	0	0.00%	47,532,957	0.34%	101,229,619	0.29%
33837	38,356,961	0.47%	17,540,865	0.14%	0	0.00%	32,138,459	0.23%	88,036,286	0.25%
32564	0	0.00%	0	0.00%	1,044,739	0.15%	0	0.00%	1,044,739	0.00%
33413	0	0.00%	22,942,007	0.19%	0	0.00%	14,802,084	0.11%	37,744,098	0.11%
33838	7,001,746	0.09%	1,834,049	0.02%	0	0.00%	4,806,175	0.03%	13,641,969	0.04%
33980	47,532,264	0.58%	0	0.00%	0	0.00%	1,399,929	0.01%	48,932,229	0.14%
32565	0	0.00%	0	0.00%	3,626,455	0.51%	0	0.00%	3,626,455	0.01%
33414	0	0.00%	132,998,967	1.09%	0	0.00%	101,348,460	0.72%	234,347,502	0.67%
32141	34,058,766	0.41%	34,258,295	0.28%	0	0.00%	42,037,047	0.30%	110,354,109	0.31%
33839	3,514,603	0.04%	1,577,236	0.01%	0	0.00%	3,549,283	0.03%	8,641,122	0.02%
34688	0	0.00%	11,373,025	0.09%	0	0.00%	11,404,139	0.08%	22,777,165	0.06%
32707	72,929,432	0.89%	25,248,338	0.21%	0	0.00%	60,392,537	0.43%	158,570,307	0.45%
33556	0	0.00%	29,082,626	0.24%	0	0.00%	29,193,699	0.21%	58,276,324	0.17%
33981	28,485,716	0.35%	0	0.00%	0	0.00%	0	0.00%	28,485,716	0.08%
32566	0	0.00%	0	0.00%	38,030,237	5.32%	0	0.00%	38,030,237	0.11%
33415	0	0.00%	47,010,500	0.39%	0	0.00%	35,554,771	0.25%	82,565,458	0.23%
34972	0	0.00%	14,119,382	0.12%	0	0.00%	20,184,229	0.14%	34,303,611	0.10%
34689	0	0.00%	23,643,987	0.19%	0	0.00%	23,705,068	0.17%	47,349,055	0.13%
32708	119,572,105	1.45%	41,381,624	0.34%	0	0.00%	99,078,408	0.70%	260,032,137	0.74%
33982	52,430,262	0.64%	0	0.00%	0	0.00%	7,258,567	0.05%	59,688,828	0.17%
33416	0	0.00%	0	0.00%	0	0.00%	0	0.00%	687,625	0.00%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34973	0	0.00%	0	0.00%	0	0.00%	0	0.00%	543,135	0.00%
33841	12,137,887	0.15%	0	0.00%	0	0.00%	7,685,105	0.05%	19,886,354	0.06%
34690	0	0.00%	7,910,841	0.06%	0	0.00%	7,942,259	0.06%	15,853,099	0.05%
32709	3,536,896	0.04%	1,268,504	0.01%	0	0.00%	2,915,515	0.02%	7,720,915	0.02%
33558	0	0.00%	30,013,855	0.25%	0	0.00%	19,979,648	0.14%	49,993,502	0.14%
33983	69,742,447	0.85%	0	0.00%	0	0.00%	1,930,605	0.01%	71,673,115	0.20%
32568	0	0.00%	0	0.00%	2,421,316	0.34%	0	0.00%	2,421,316	0.01%
33417	0	0.00%	50,996,571	0.42%	0	0.00%	33,040,268	0.23%	84,037,121	0.24%
34266	84,734,223	1.03%	0	0.00%	0	0.00%	20,016,397	0.14%	104,750,620	0.30%
34974	0	0.00%	39,603,558	0.33%	0	0.00%	47,252,973	0.34%	87,235,851	0.25%
34691	0	0.00%	11,270,200	0.09%	0	0.00%	18,553,915	0.13%	29,824,115	0.08%
33559	0	0.00%	9,064,616	0.07%	0	0.00%	9,096,087	0.06%	18,160,703	0.05%
33701	0	0.00%	12,180,937	0.10%	0	0.00%	0	0.00%	12,181,152	0.03%
32569	0	0.00%	0	0.00%	10,659,348	1.49%	0	0.00%	10,659,348	0.03%
33418	0	0.00%	183,855,098	1.51%	0	0.00%	149,446,737	1.06%	333,301,998	0.95%
33843	24,623,351	0.30%	4,574,347	0.04%	0	0.00%	13,254,137	0.09%	42,451,835	0.12%
33702	0	0.00%	19,201,493	0.16%	0	0.00%	0	0.00%	19,201,721	0.05%
32570	0	0.00%	0	0.00%	18,641,328	2.61%	0	0.00%	18,641,328	0.05%
34268	0	0.00%	0	0.00%	0	0.00%	0	0.00%	552,063	0.00%
33844	50,140,311	0.61%	19,083,213	0.16%	0	0.00%	50,627,517	0.36%	119,851,041	0.34%
32712	1,414,362	0.02%	1,422,924	0.01%	0	0.00%	58,507,325	0.42%	61,344,612	0.17%
33703	0	0.00%	22,560,352	0.19%	0	0.00%	0	0.00%	22,560,479	0.06%
32571	0	0.00%	0	0.00%	25,788,897	3.61%	0	0.00%	25,788,897	0.07%
34269	15,518,000	0.19%	0	0.00%	0	0.00%	3,614,646	0.03%	19,132,685	0.05%
32713	33,434,392	0.41%	0	0.00%	0	0.00%	33,763,482	0.24%	67,528,030	0.19%
33704	0	0.00%	17,478,335	0.14%	0	0.00%	0	0.00%	17,478,359	0.05%
32148	0	0.00%	0	0.00%	0	0.00%	3,134,945	0.02%	3,134,945	0.01%
33846	0	0.00%	0	0.00%	0	0.00%	509,000	0.00%	516,068	0.00%
34695	0	0.00%	17,311,272	0.14%	0	0.00%	0	0.00%	17,311,370	0.05%
32714	37,854,791	0.46%	3,173,316	0.03%	0	0.00%	38,192,560	0.27%	79,220,667	0.23%
33563	0	0.00%	0	0.00%	0	0.00%	14,107,722	0.10%	14,502,766	0.04%
32856	0	0.00%	0	0.00%	0	0.00%	0	0.00%	631,346	0.00%
33705	0	0.00%	14,893,004	0.12%	0	0.00%	0	0.00%	14,893,509	0.04%
33847	0	0.00%	0	0.00%	0	0.00%	0	0.00%	781,233	0.00%
33706	0	0.00%	60,877,198	0.50%	0	0.00%	0	0.00%	60,879,503	0.17%
33848	1,064,498	0.01%	0	0.00%	0	0.00%	754,557	0.01%	2,112,538	0.01%
33565	0	0.00%	9,207,839	0.08%	0	0.00%	21,452,103	0.15%	30,659,942	0.09%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33990	89,919,691	1.09%	0	0.00%	0	0.00%	0	0.00%	89,919,702	0.26%
33707	0	0.00%	42,326,409	0.35%	0	0.00%	0	0.00%	42,327,181	0.12%
33424	0	0.00%	1,037,383	0.01%	0	0.00%	702,682	0.01%	1,740,082	0.00%
33849	0	0.00%	0	0.00%	0	0.00%	724,368	0.01%	1,047,686	0.00%
34698	0	0.00%	34,189,575	0.28%	0	0.00%	7,303,838	0.05%	41,493,413	0.12%
33566	0	0.00%	12,181,141	0.10%	0	0.00%	21,290,888	0.15%	33,472,029	0.10%
34981	0	0.00%	7,997,794	0.07%	0	0.00%	8,028,339	0.06%	16,026,133	0.05%
33991	50,743,628	0.62%	0	0.00%	0	0.00%	0	0.00%	50,743,727	0.14%
33708	0	0.00%	58,413,708	0.48%	0	0.00%	0	0.00%	58,415,990	0.17%
33425	0	0.00%	0	0.00%	0	0.00%	0	0.00%	722,549	0.00%
33850	6,023,507	0.07%	0	0.00%	0	0.00%	8,313,356	0.06%	14,425,858	0.04%
32718	0	0.00%	0	0.00%	0	0.00%	0	0.00%	702,242	0.00%
33567	0	0.00%	6,351,821	0.05%	0	0.00%	10,312,431	0.07%	16,664,252	0.05%
34982	0	0.00%	56,482,817	0.46%	0	0.00%	52,342,424	0.37%	108,825,353	0.31%
33709	0	0.00%	14,080,138	0.12%	0	0.00%	0	0.00%	14,080,217	0.04%
32577	0	0.00%	0	0.00%	5,276,885	0.74%	0	0.00%	5,276,885	0.02%
33426	0	0.00%	33,485,213	0.27%	0	0.00%	15,710,415	0.11%	49,195,688	0.14%
34275	0	0.00%	1,930,083	0.02%	0	0.00%	0	0.00%	1,930,426	0.01%
33851	2,114,855	0.03%	646,710	0.01%	0	0.00%	1,869,260	0.01%	4,630,824	0.01%
34134	49,853,099	0.61%	50,142,182	0.41%	0	0.00%	0	0.00%	99,995,452	0.28%
34983	0	0.00%	84,123,739	0.69%	0	0.00%	91,339,285	0.65%	175,463,023	0.50%
33993	45,145,960	0.55%	0	0.00%	0	0.00%	0	0.00%	45,145,960	0.13%
33710	0	0.00%	25,266,525	0.21%	0	0.00%	0	0.00%	25,266,669	0.07%
32578	0	0.00%	0	0.00%	1,585,109	0.22%	0	0.00%	1,585,109	0.00%
33852	32,207,724	0.39%	32,395,573	0.27%	0	0.00%	53,964,519	0.38%	118,567,816	0.34%
32720	3,525,381	0.04%	3,544,719	0.03%	0	0.00%	24,794,042	0.18%	31,864,142	0.09%
33569	0	0.00%	3,560,505	0.03%	0	0.00%	54,085,227	0.38%	57,645,800	0.16%
34135	12,565,371	0.15%	0	0.00%	0	0.00%	0	0.00%	12,566,174	0.04%
34984	0	0.00%	34,096,128	0.28%	0	0.00%	36,994,903	0.26%	71,091,031	0.20%
33711	0	0.00%	13,145,840	0.11%	0	0.00%	0	0.00%	13,146,355	0.04%
32579	0	0.00%	0	0.00%	9,640,244	1.35%	0	0.00%	9,640,244	0.03%
33428	0	0.00%	36,472,412	0.30%	0	0.00%	0	0.00%	36,472,717	0.10%
33853	36,316,306	0.44%	8,080,147	0.07%	0	0.00%	23,649,552	0.17%	68,046,005	0.19%
33570	0	0.00%	544,226	0.00%	0	0.00%	9,443,474	0.07%	9,987,726	0.03%
33712	0	0.00%	11,954,259	0.10%	0	0.00%	0	0.00%	11,954,328	0.03%
32580	0	0.00%	0	0.00%	2,671,515	0.37%	0	0.00%	2,671,515	0.01%
34420	0	0.00%	8,824,372	0.07%	0	0.00%	14,961,292	0.11%	23,785,664	0.07%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34986	0	0.00%	61,514,840	0.50%	0	0.00%	66,775,948	0.47%	128,291,009	0.36%
33713	0	0.00%	17,628,746	0.14%	0	0.00%	0	0.00%	17,628,810	0.05%
33430	0	0.00%	6,398,536	0.05%	0	0.00%	6,419,329	0.05%	12,817,877	0.04%
33855	3,033,637	0.04%	1,829,821	0.02%	0	0.00%	3,063,661	0.02%	7,927,119	0.02%
33572	0	0.00%	517,389	0.00%	0	0.00%	18,371,630	0.13%	18,889,019	0.05%
34987	0	0.00%	10,387,309	0.09%	0	0.00%	11,284,463	0.08%	21,671,805	0.06%
33714	0	0.00%	8,512,014	0.07%	0	0.00%	0	0.00%	8,512,432	0.02%
33431	0	0.00%	26,571,625	0.22%	0	0.00%	0	0.00%	26,572,360	0.08%
34705	0	0.00%	1,395,903	0.01%	0	0.00%	2,233,210	0.02%	3,629,113	0.01%
32724	24,645,008	0.30%	2,202,465	0.02%	0	0.00%	24,869,922	0.18%	51,717,396	0.15%
33573	0	0.00%	7,448,885	0.06%	0	0.00%	25,706,634	0.18%	33,155,690	0.09%
33715	0	0.00%	27,822,161	0.23%	0	0.00%	10,068,439	0.07%	37,890,600	0.11%
32583	0	0.00%	0	0.00%	14,320,741	2.00%	0	0.00%	14,320,741	0.04%
33432	0	0.00%	41,896,540	0.34%	0	0.00%	0	0.00%	41,900,338	0.12%
32159	0	0.00%	31,278,637	0.26%	0	0.00%	52,495,241	0.37%	83,773,878	0.24%
33857	1,029,609	0.01%	2,495,404	0.02%	0	0.00%	3,637,060	0.03%	7,162,073	0.02%
32725	65,668,633	0.80%	880,240	0.01%	0	0.00%	66,313,505	0.47%	132,862,378	0.38%
33716	0	0.00%	5,910,281	0.05%	0	0.00%	0	0.00%	5,910,827	0.02%
33433	0	0.00%	50,189,719	0.41%	0	0.00%	0	0.00%	50,191,221	0.14%
32301	0	0.00%	13,881,726	0.11%	0	0.00%	0	0.00%	13,881,726	0.04%
33858	0	0.00%	0	0.00%	0	0.00%	0	0.00%	814,760	0.00%
32726	0	0.00%	15,128,297	0.12%	0	0.00%	25,873,316	0.18%	41,001,630	0.12%
34990	0	0.00%	120,792,975	0.99%	0	0.00%	121,200,562	0.86%	241,993,653	0.69%
33434	0	0.00%	36,207,792	0.30%	0	0.00%	0	0.00%	36,210,231	0.10%
32302	0	0.00%	2,188,167	0.02%	0	0.00%	0	0.00%	2,188,167	0.01%
33859	21,858,561	0.27%	5,701,694	0.05%	0	0.00%	13,479,751	0.10%	41,040,006	0.12%
33576	0	0.00%	5,344,808	0.04%	0	0.00%	5,366,118	0.04%	10,710,926	0.03%
34991	0	0.00%	1,048,295	0.01%	0	0.00%	1,042,563	0.01%	2,090,867	0.01%
33435	0	0.00%	51,070,789	0.42%	0	0.00%	26,878,405	0.19%	77,949,580	0.22%
32303	0	0.00%	27,227,230	0.22%	0	0.00%	0	0.00%	27,227,230	0.08%
33860	0	0.00%	567,687	0.00%	0	0.00%	16,172,670	0.11%	16,740,399	0.05%
34992	0	0.00%	0	0.00%	0	0.00%	0	0.00%	793,375	0.00%
32162	0	0.00%	58,350,102	0.48%	0	0.00%	58,578,391	0.42%	116,928,493	0.33%
33436	0	0.00%	61,566,701	0.51%	0	0.00%	39,565,928	0.28%	101,132,828	0.29%
34285	6,991,606	0.09%	7,031,721	0.06%	0	0.00%	0	0.00%	14,024,196	0.04%
32304	0	0.00%	9,660,004	0.08%	0	0.00%	0	0.00%	9,660,004	0.03%
33437	0	0.00%	109,252,677	0.90%	0	0.00%	75,296,801	0.53%	184,549,753	0.52%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34286	38,960,439	0.47%	0	0.00%	0	0.00%	0	0.00%	39,278,367	0.11%
32305	0	0.00%	4,495,668	0.04%	0	0.00%	0	0.00%	4,495,668	0.01%
34711	0	0.00%	55,698,848	0.46%	0	0.00%	99,964,159	0.71%	155,663,077	0.44%
32730	8,111,636	0.10%	0	0.00%	0	0.00%	5,847,936	0.04%	14,118,948	0.04%
34428	0	0.00%	13,149,717	0.11%	0	0.00%	5,785,735	0.04%	18,935,452	0.05%
34145	0	0.00%	62,023,994	0.51%	0	0.00%	0	0.00%	62,026,360	0.18%
34994	0	0.00%	42,753,724	0.35%	0	0.00%	42,770,214	0.30%	85,524,064	0.24%
32164	0	0.00%	36,512,724	0.30%	0	0.00%	36,657,583	0.26%	73,170,321	0.21%
33438	0	0.00%	1,145,835	0.01%	0	0.00%	985,358	0.01%	2,131,193	0.01%
34287	36,195,760	0.44%	0	0.00%	0	0.00%	0	0.00%	36,195,768	0.10%
34429	0	0.00%	14,490,477	0.12%	0	0.00%	14,529,190	0.10%	29,019,668	0.08%
34288	19,891,063	0.24%	0	0.00%	0	0.00%	670,180	0.00%	20,561,324	0.06%
32732	15,194,637	0.18%	4,799,077	0.04%	0	0.00%	10,944,613	0.08%	30,938,327	0.09%
34430	0	0.00%	0	0.00%	0	0.00%	0	0.00%	619,885	0.00%
34996	0	0.00%	70,841,613	0.58%	0	0.00%	65,795,869	0.47%	136,637,813	0.39%
32025	0	0.00%	0	0.00%	0	0.00%	7,605,466	0.05%	7,605,521	0.02%
33440	0	0.00%	10,527,355	0.09%	0	0.00%	10,565,392	0.08%	21,092,748	0.06%
34289	2,875,330	0.04%	0	0.00%	0	0.00%	0	0.00%	3,219,594	0.01%
32308	0	0.00%	19,731,273	0.16%	0	0.00%	0	0.00%	19,731,276	0.06%
34714	1,749,905	0.02%	1,759,968	0.01%	0	0.00%	15,880,026	0.11%	19,389,899	0.06%
34431	0	0.00%	15,412,870	0.13%	0	0.00%	11,406,026	0.08%	26,818,896	0.08%
34997	0	0.00%	124,317,151	1.02%	0	0.00%	113,141,544	0.80%	237,458,955	0.67%
33865	1,307,080	0.02%	0	0.00%	0	0.00%	741,713	0.01%	2,055,399	0.01%
33441	0	0.00%	17,591,313	0.14%	0	0.00%	0	0.00%	17,593,087	0.05%
32309	0	0.00%	22,908,546	0.19%	0	0.00%	0	0.00%	22,908,553	0.07%
34432	0	0.00%	16,313,866	0.13%	0	0.00%	16,378,762	0.12%	32,692,628	0.09%
32168	58,737,363	0.71%	36,143,013	0.30%	0	0.00%	59,307,477	0.42%	154,187,852	0.44%
34715	0	0.00%	7,890,835	0.06%	0	0.00%	14,514,061	0.10%	22,404,897	0.06%
33442	0	0.00%	29,636,174	0.24%	0	0.00%	0	0.00%	29,637,328	0.08%
32310	0	0.00%	5,537,440	0.05%	0	0.00%	0	0.00%	5,537,440	0.02%
32735	0	0.00%	3,182,364	0.03%	0	0.00%	4,882,008	0.03%	8,064,372	0.02%
33584	0	0.00%	0	0.00%	0	0.00%	15,666,136	0.11%	15,927,737	0.05%
34433	0	0.00%	8,427,419	0.07%	0	0.00%	6,407,983	0.05%	14,835,402	0.04%
32169	59,905,448	0.73%	68,580,312	0.56%	0	0.00%	75,705,378	0.54%	204,191,138	0.58%
33867	720,744	0.01%	0	0.00%	0	0.00%	0	0.00%	1,034,432	0.00%
32311	0	0.00%	13,207,897	0.11%	0	0.00%	0	0.00%	13,208,056	0.04%
32736	0	0.00%	0	0.00%	0	0.00%	11,882,303	0.08%	12,029,477	0.03%

ZIP Code	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33585	0	0.00%	1,011,208	0.01%	0	0.00%	1,336,379	0.01%	2,347,588	0.01%
34434	0	0.00%	13,302,516	0.11%	0	0.00%	10,238,071	0.07%	23,540,586	0.07%
33868	0	0.00%	4,749,300	0.04%	0	0.00%	10,524,929	0.07%	15,274,229	0.04%
33444	0	0.00%	20,710,764	0.17%	0	0.00%	3,779,827	0.03%	24,490,752	0.07%
34293	39,501,072	0.48%	0	0.00%	0	0.00%	0	0.00%	39,502,485	0.11%
32312	0	0.00%	35,365,637	0.29%	0	0.00%	0	0.00%	35,365,640	0.10%
33445	0	0.00%	53,933,484	0.44%	0	0.00%	12,295,233	0.09%	66,229,071	0.19%
32738	75,722,371	0.92%	0	0.00%	0	0.00%	56,821,770	0.40%	133,021,321	0.38%
34436	0	0.00%	8,306,996	0.07%	0	0.00%	8,340,154	0.06%	16,647,150	0.05%
33870	29,103,822	0.35%	12,755,847	0.10%	0	0.00%	35,827,252	0.25%	77,686,922	0.22%
33446	0	0.00%	66,205,879	0.54%	0	0.00%	48,833,278	0.35%	115,039,501	0.33%
33447	0	0.00%	749,992	0.01%	0	0.00%	0	0.00%	1,241,853	0.00%
32174	83,548,339	1.02%	84,034,685	0.69%	0	0.00%	113,444,435	0.81%	281,027,459	0.80%
33872	40,481,469	0.49%	15,215,174	0.12%	0	0.00%	40,870,416	0.29%	96,567,058	0.27%
33873	21,441,094	0.26%	0	0.00%	0	0.00%	12,473,182	0.09%	34,111,109	0.10%
32317	0	0.00%	10,498,163	0.09%	0	0.00%	0	0.00%	10,498,167	0.03%
32034	0	0.00%	0	0.00%	0	0.00%	6,489,328	0.05%	6,489,571	0.02%
32176	78,341,998	0.95%	63,116,149	0.52%	0	0.00%	72,186,284	0.51%	213,644,431	0.61%
32601	0	0.00%	1,840,440	0.02%	0	0.00%	7,062,660	0.05%	8,903,100	0.03%
33592	0	0.00%	4,512,962	0.04%	0	0.00%	7,468,378	0.05%	11,981,342	0.03%
33875	17,948,714	0.22%	7,915,951	0.07%	0	0.00%	22,156,391	0.16%	48,021,056	0.14%
32744	4,622,412	0.06%	0	0.00%	0	0.00%	4,668,232	0.03%	9,335,443	0.03%
34442	0	0.00%	29,056,449	0.24%	0	0.00%	22,523,375	0.16%	51,579,824	0.15%
33876	8,096,746	0.10%	6,029,512	0.05%	0	0.00%	11,495,619	0.08%	25,621,877	0.07%
32603	0	0.00%	0	0.00%	0	0.00%	1,272,826	0.01%	1,380,756	0.00%
33594	0	0.00%	1,583,398	0.01%	0	0.00%	57,646,291	0.41%	59,229,742	0.17%
32179	0	0.00%	4,119,439	0.03%	0	0.00%	7,144,529	0.05%	11,263,968	0.03%
33877	660,671	0.01%	0	0.00%	0	0.00%	531,659	0.00%	1,363,650	0.00%
32038	0	0.00%	0	0.00%	0	0.00%	3,189,674	0.02%	3,189,674	0.01%
32746	93,180,894	1.13%	42,769,706	0.35%	0	0.00%	70,190,008	0.50%	206,140,608	0.59%
32180	0	0.00%	0	0.00%	0	0.00%	1,577,500	0.01%	1,577,500	0.00%
32605	0	0.00%	1,217,579	0.01%	0	0.00%	18,501,938	0.13%	19,719,517	0.06%
32606	0	0.00%	3,271,675	0.03%	0	0.00%	17,587,096	0.12%	20,858,771	0.06%
33455	0	0.00%	106,129,787	0.87%	0	0.00%	96,503,512	0.68%	202,633,396	0.58%
33597	0	0.00%	4,875,865	0.04%	0	0.00%	6,571,690	0.05%	11,447,555	0.03%
34446	0	0.00%	25,685,189	0.21%	0	0.00%	15,988,488	0.11%	41,673,677	0.12%
33880	31,957,269	0.39%	1,251,105	0.01%	0	0.00%	43,992,363	0.31%	77,200,737	0.22%

ZIP Code	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32607	0	0.00%	4,035,014	0.03%	0	0.00%	17,209,102	0.12%	21,244,117	0.06%
33598	0	0.00%	0	0.00%	0	0.00%	3,545,669	0.03%	3,602,258	0.01%
34447	0	0.00%	0	0.00%	0	0.00%	0	0.00%	546,509	0.00%
33881	40,428,439	0.49%	17,736,002	0.15%	0	0.00%	49,971,855	0.35%	108,136,296	0.31%
32608	0	0.00%	22,598,771	0.19%	0	0.00%	22,674,040	0.16%	45,272,811	0.13%
34448	0	0.00%	15,899,130	0.13%	0	0.00%	11,655,164	0.08%	27,554,295	0.08%
34731	0	0.00%	13,528,955	0.11%	0	0.00%	18,623,442	0.13%	32,152,400	0.09%
32750	38,636,685	0.47%	21,406,919	0.18%	0	0.00%	39,015,991	0.28%	99,059,594	0.28%
32609	0	0.00%	1,101,247	0.01%	0	0.00%	6,183,954	0.04%	7,285,202	0.02%
33458	0	0.00%	150,580,216	1.24%	0	0.00%	122,147,284	0.87%	272,727,746	0.77%
34449	0	0.00%	1,978,479	0.02%	0	0.00%	1,371,651	0.01%	3,350,129	0.01%
32751	59,301,691	0.72%	25,470,217	0.21%	0	0.00%	43,797,633	0.31%	128,569,541	0.37%
32327	0	0.00%	11,104,721	0.09%	0	0.00%	0	0.00%	11,104,721	0.03%
32893	0	0.00%	0	0.00%	0	0.00%	0	0.00%	581,992	0.00%
34450	0	0.00%	16,257,612	0.13%	0	0.00%	16,318,135	0.12%	32,575,748	0.09%
33884	81,052,004	0.99%	27,382,930	0.22%	0	0.00%	71,138,440	0.50%	179,573,374	0.51%
33601	0	0.00%	0	0.00%	0	0.00%	0	0.00%	541,928	0.00%
33460	0	0.00%	39,834,562	0.33%	0	0.00%	24,318,057	0.17%	64,152,737	0.18%
34734	3,914,000	0.05%	0	0.00%	0	0.00%	8,589,672	0.06%	12,567,440	0.04%
33602	0	0.00%	11,671,417	0.10%	0	0.00%	5,255,156	0.04%	16,926,573	0.05%
33461	0	0.00%	38,763,446	0.32%	0	0.00%	18,561,027	0.13%	57,324,610	0.16%
34452	0	0.00%	14,829,731	0.12%	0	0.00%	14,887,896	0.11%	29,717,627	0.08%
32754	17,000,319	0.21%	12,658,772	0.10%	0	0.00%	20,919,577	0.15%	50,578,668	0.14%
33603	0	0.00%	689,943	0.01%	0	0.00%	687,674	0.00%	1,377,617	0.00%
33462	0	0.00%	58,848,207	0.48%	0	0.00%	28,248,668	0.20%	87,097,050	0.25%
34736	0	0.00%	9,664,500	0.08%	0	0.00%	17,342,404	0.12%	27,006,905	0.08%
33604	0	0.00%	1,039,766	0.01%	0	0.00%	1,037,089	0.01%	2,076,855	0.01%
34453	0	0.00%	15,352,325	0.13%	0	0.00%	11,632,631	0.08%	26,984,956	0.08%
33463	0	0.00%	70,666,223	0.58%	0	0.00%	35,598,778	0.25%	106,265,142	0.30%
32331	0	0.00%	1,553,606	0.01%	0	0.00%	0	0.00%	1,553,606	0.00%
34737	0	0.00%	4,819,313	0.04%	0	0.00%	7,513,199	0.05%	12,332,512	0.04%
33605	0	0.00%	767,596	0.01%	0	0.00%	764,222	0.01%	1,531,819	0.00%
32615	0	0.00%	0	0.00%	0	0.00%	9,037,515	0.06%	9,433,076	0.03%
33464	0	0.00%	0	0.00%	0	0.00%	0	0.00%	506,372	0.00%
32757	0	0.00%	1,229,718	0.01%	0	0.00%	31,314,733	0.22%	32,544,481	0.09%
33606	0	0.00%	17,960,443	0.15%	0	0.00%	1,890,717	0.01%	19,851,160	0.06%
32192	0	0.00%	0	0.00%	0	0.00%	528,681	0.00%	825,866	0.00%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33890	8,796,300	0.11%	1,737,328	0.01%	0	0.00%	4,000,482	0.03%	14,534,110	0.04%
34739	951,192	0.01%	1,278,685	0.01%	0	0.00%	2,233,530	0.02%	4,463,407	0.01%
33607	0	0.00%	781,348	0.01%	0	0.00%	778,259	0.01%	1,559,606	0.00%
32617	0	0.00%	1,994,143	0.02%	0	0.00%	3,383,141	0.02%	5,377,284	0.02%
32759	3,317,558	0.04%	3,336,956	0.03%	0	0.00%	4,786,909	0.03%	11,441,422	0.03%
32618	0	0.00%	2,949,983	0.02%	0	0.00%	2,961,791	0.02%	5,911,774	0.02%
33467	0	0.00%	129,822,882	1.07%	0	0.00%	101,287,687	0.72%	231,110,728	0.66%
32052	0	0.00%	1,723,206	0.01%	0	0.00%	0	0.00%	1,806,621	0.01%
32901	0	0.00%	36,958,752	0.30%	0	0.00%	47,059,449	0.33%	84,018,663	0.24%
34741	50,717,812	0.62%	19,041,295	0.16%	0	0.00%	37,218,835	0.26%	106,977,942	0.30%
33609	0	0.00%	1,976,401	0.02%	0	0.00%	1,968,564	0.01%	3,944,965	0.01%
33468	0	0.00%	0	0.00%	0	0.00%	0	0.00%	581,160	0.00%
32053	0	0.00%	981,074	0.01%	0	0.00%	0	0.00%	1,019,757	0.00%
32619	0	0.00%	1,545,547	0.01%	0	0.00%	0	0.00%	1,598,765	0.00%
32195	0	0.00%	2,171,090	0.02%	0	0.00%	3,563,907	0.03%	5,734,997	0.02%
33610	0	0.00%	1,016,583	0.01%	0	0.00%	13,377,004	0.09%	14,393,587	0.04%
33469	0	0.00%	79,740,355	0.65%	0	0.00%	73,221,992	0.52%	152,962,473	0.43%
32903	0	0.00%	66,641,577	0.55%	0	0.00%	81,094,415	0.58%	147,736,656	0.42%
34601	0	0.00%	11,973,578	0.10%	0	0.00%	21,202,237	0.15%	33,175,815	0.09%
34743	78,076,522	0.95%	22,075,641	0.18%	0	0.00%	63,392,606	0.45%	163,544,770	0.46%
33611	0	0.00%	27,252,998	0.22%	0	0.00%	4,445,714	0.03%	31,698,712	0.09%
32055	0	0.00%	0	0.00%	0	0.00%	6,325,750	0.04%	6,325,884	0.02%
32904	0	0.00%	45,570,240	0.37%	0	0.00%	63,576,609	0.45%	109,146,980	0.31%
34602	0	0.00%	8,511,276	0.07%	0	0.00%	8,545,154	0.06%	17,056,430	0.05%
32621	0	0.00%	1,715,443	0.01%	0	0.00%	1,338,594	0.01%	3,054,037	0.01%
33470	0	0.00%	55,071,431	0.45%	0	0.00%	46,073,546	0.33%	101,144,977	0.29%
34744	101,323,079	1.23%	32,467,763	0.27%	0	0.00%	81,696,917	0.58%	215,487,759	0.61%
32763	12,912,704	0.16%	611,247	0.01%	0	0.00%	13,035,805	0.09%	26,559,756	0.08%
33612	0	0.00%	2,718,323	0.02%	0	0.00%	18,240,358	0.13%	20,958,681	0.06%
34461	0	0.00%	14,922,288	0.12%	0	0.00%	14,981,061	0.11%	29,903,349	0.08%
32905	0	0.00%	37,173,528	0.31%	0	0.00%	63,249,042	0.45%	100,422,832	0.29%
32622	0	0.00%	0	0.00%	0	0.00%	640,660	0.00%	640,660	0.00%
33471	0	0.00%	4,999,824	0.04%	0	0.00%	5,019,959	0.04%	10,019,783	0.03%
33896	10,798,912	0.13%	1,294,197	0.01%	0	0.00%	13,303,612	0.09%	25,396,720	0.07%
32764	5,945,978	0.07%	2,162,637	0.02%	0	0.00%	3,672,774	0.03%	11,781,389	0.03%
33613	0	0.00%	18,160,870	0.15%	0	0.00%	18,196,480	0.13%	36,357,350	0.10%
32340	0	0.00%	4,067,874	0.03%	0	0.00%	0	0.00%	4,067,917	0.01%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33755	0	0.00%	14,314,804	0.12%	0	0.00%	1,457,462	0.01%	15,772,266	0.04%
34604	0	0.00%	7,945,604	0.07%	0	0.00%	5,662,961	0.04%	13,608,565	0.04%
33897	0	0.00%	0	0.00%	0	0.00%	29,542,000	0.21%	29,818,055	0.08%
34746	96,226,031	1.17%	35,880,210	0.29%	0	0.00%	76,344,853	0.54%	208,451,094	0.59%
32765	150,925,297	1.84%	47,421,353	0.39%	0	0.00%	109,526,423	0.78%	307,873,073	0.87%
33614	0	0.00%	4,480,822	0.04%	0	0.00%	4,458,324	0.03%	8,939,146	0.03%
32907	0	0.00%	54,047,156	0.44%	0	0.00%	102,437,647	0.73%	156,484,803	0.44%
33756	0	0.00%	26,625,868	0.22%	0	0.00%	5,448,241	0.04%	32,074,109	0.09%
33898	44,698,635	0.54%	8,705,911	0.07%	0	0.00%	25,070,728	0.18%	78,475,274	0.22%
34747	60,514,586	0.74%	38,039,374	0.31%	0	0.00%	60,997,869	0.43%	159,551,830	0.45%
32766	37,585,155	0.46%	15,624,989	0.13%	0	0.00%	27,842,910	0.20%	81,053,054	0.23%
33615	0	0.00%	25,595,571	0.21%	0	0.00%	2,430,922	0.02%	28,026,493	0.08%
32059	0	0.00%	889,044	0.01%	0	0.00%	0	0.00%	889,044	0.00%
32908	0	0.00%	11,175,160	0.09%	0	0.00%	19,431,706	0.14%	30,606,866	0.09%
34606	0	0.00%	40,535,415	0.33%	0	0.00%	22,959,344	0.16%	63,494,759	0.18%
32625	0	0.00%	2,717,012	0.02%	0	0.00%	1,574,161	0.01%	4,291,173	0.01%
34748	0	0.00%	31,057,437	0.25%	0	0.00%	48,707,245	0.35%	79,764,714	0.23%
32767	0	0.00%	0	0.00%	0	0.00%	1,101,141	0.01%	1,117,938	0.00%
33616	0	0.00%	6,640,603	0.05%	0	0.00%	1,027,043	0.01%	7,667,646	0.02%
34465	0	0.00%	27,855,550	0.23%	0	0.00%	16,827,250	0.12%	44,682,800	0.13%
32909	0	0.00%	45,012,330	0.37%	0	0.00%	67,152,927	0.48%	112,165,285	0.32%
34607	0	0.00%	14,696,444	0.12%	0	0.00%	8,476,412	0.06%	23,172,855	0.07%
32626	0	0.00%	3,864,005	0.03%	0	0.00%	0	0.00%	3,928,296	0.01%
33475	0	0.00%	1,009,356	0.01%	0	0.00%	920,543	0.01%	1,929,902	0.01%
33617	0	0.00%	6,218,690	0.05%	0	0.00%	26,725,911	0.19%	32,944,601	0.09%
32344	0	0.00%	6,195,842	0.05%	0	0.00%	0	0.00%	6,195,855	0.02%
33759	0	0.00%	11,304,026	0.09%	0	0.00%	0	0.00%	11,304,419	0.03%
34608	0	0.00%	44,552,731	0.37%	0	0.00%	25,868,002	0.18%	70,420,733	0.20%
33476	0	0.00%	4,952,701	0.04%	0	0.00%	3,979,467	0.03%	8,932,169	0.03%
33901	40,856,080	0.50%	0	0.00%	0	0.00%	0	0.00%	40,856,153	0.12%
33618	0	0.00%	23,708,653	0.19%	0	0.00%	23,777,576	0.17%	47,486,230	0.13%
33477	0	0.00%	92,562,703	0.76%	0	0.00%	84,016,146	0.60%	176,579,222	0.50%
32628	0	0.00%	1,373,601	0.01%	0	0.00%	0	0.00%	1,373,601	0.00%
34609	0	0.00%	51,763,425	0.42%	0	0.00%	33,266,895	0.24%	85,030,319	0.24%
33760	0	0.00%	8,691,317	0.07%	0	0.00%	0	0.00%	8,691,730	0.02%
32062	0	0.00%	870,770	0.01%	0	0.00%	0	0.00%	890,769	0.00%
33619	0	0.00%	738,582	0.01%	0	0.00%	11,742,426	0.08%	12,481,008	0.04%

ZIP Code	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33478	0	0.00%	47,557,765	0.39%	0	0.00%	44,101,040	0.31%	91,658,805	0.26%
34610	0	0.00%	9,851,476	0.08%	0	0.00%	9,890,959	0.07%	19,742,435	0.06%
33761	0	0.00%	19,603,414	0.16%	0	0.00%	5,401,107	0.04%	25,004,521	0.07%
32346	0	0.00%	2,565,881	0.02%	0	0.00%	0	0.00%	2,565,881	0.01%
32771	68,389,978	0.83%	35,830,785	0.29%	0	0.00%	53,391,833	0.38%	157,612,595	0.45%
33903	79,568,421	0.97%	0	0.00%	0	0.00%	0	0.00%	79,568,568	0.23%
33762	0	0.00%	6,391,167	0.05%	0	0.00%	0	0.00%	6,391,329	0.02%
32064	0	0.00%	1,531,241	0.01%	0	0.00%	0	0.00%	1,559,517	0.00%
32347	0	0.00%	4,395,301	0.04%	0	0.00%	0	0.00%	4,395,301	0.01%
34470	0	0.00%	12,042,447	0.10%	0	0.00%	12,078,216	0.09%	24,120,664	0.07%
34753	0	0.00%	2,536,628	0.02%	0	0.00%	4,494,381	0.03%	7,031,009	0.02%
33904	144,356,540	1.76%	0	0.00%	0	0.00%	0	0.00%	144,356,979	0.41%
33480	0	0.00%	240,629,170	1.98%	0	0.00%	161,248,657	1.14%	401,879,388	1.14%
32631	0	0.00%	0	0.00%	0	0.00%	546,966	0.00%	577,179	0.00%
33763	0	0.00%	1,361,974	0.01%	0	0.00%	1,354,932	0.01%	2,716,906	0.01%
32348	0	0.00%	3,628,430	0.03%	0	0.00%	0	0.00%	3,628,430	0.01%
34471	0	0.00%	21,723,439	0.18%	0	0.00%	38,565,395	0.27%	60,288,834	0.17%
32773	31,923,648	0.39%	14,240,038	0.12%	0	0.00%	23,804,412	0.17%	69,968,098	0.20%
33905	42,616,752	0.52%	0	0.00%	0	0.00%	1,574,844	0.01%	44,191,644	0.13%
34613	0	0.00%	23,868,847	0.20%	0	0.00%	14,670,125	0.10%	38,538,972	0.11%
33764	0	0.00%	20,124,562	0.17%	0	0.00%	0	0.00%	20,125,236	0.06%
32066	0	0.00%	1,476,080	0.01%	0	0.00%	0	0.00%	1,476,080	0.00%
34755	0	0.00%	0	0.00%	0	0.00%	906,430	0.01%	1,301,934	0.00%
34472	0	0.00%	18,334,372	0.15%	0	0.00%	29,565,654	0.21%	47,900,025	0.14%
33765	0	0.00%	8,058,230	0.07%	0	0.00%	0	0.00%	8,058,505	0.02%
34614	0	0.00%	5,244,388	0.04%	0	0.00%	4,020,558	0.03%	9,264,946	0.03%
33907	35,523,591	0.43%	0	0.00%	0	0.00%	0	0.00%	35,523,754	0.10%
34756	0	0.00%	3,225,675	0.03%	0	0.00%	6,488,156	0.05%	9,713,831	0.03%
32775	613,516	0.01%	0	0.00%	0	0.00%	755,678	0.01%	1,817,516	0.01%
33624	0	0.00%	31,363,403	0.26%	0	0.00%	31,479,911	0.22%	62,843,315	0.18%
34473	0	0.00%	17,320,262	0.14%	0	0.00%	17,389,342	0.12%	34,709,603	0.10%
32634	0	0.00%	0	0.00%	0	0.00%	0	0.00%	593,966	0.00%
33483	0	0.00%	43,228,083	0.35%	0	0.00%	13,176,719	0.09%	56,405,586	0.16%
34474	0	0.00%	19,508,288	0.16%	0	0.00%	19,562,932	0.14%	39,071,220	0.11%
33625	0	0.00%	16,165,465	0.13%	0	0.00%	16,229,282	0.12%	32,394,747	0.09%
33908	127,635,205	1.55%	0	0.00%	0	0.00%	0	0.00%	127,635,702	0.36%
32776	0	0.00%	8,850,497	0.07%	0	0.00%	13,122,325	0.09%	22,127,614	0.06%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33484	0	0.00%	45,938,345	0.38%	0	0.00%	10,951,295	0.08%	56,889,833	0.16%
33767	0	0.00%	46,471,107	0.38%	0	0.00%	27,648,207	0.20%	74,119,314	0.21%
33909	46,994,640	0.57%	0	0.00%	0	0.00%	0	0.00%	46,994,640	0.13%
34475	0	0.00%	3,880,823	0.03%	0	0.00%	3,894,944	0.03%	7,775,767	0.02%
33626	0	0.00%	27,677,234	0.23%	0	0.00%	1,135,260	0.01%	28,812,494	0.08%
34758	60,471,992	0.74%	22,457,209	0.18%	0	0.00%	48,705,658	0.35%	131,634,859	0.37%
33060	0	0.00%	6,202,860	0.05%	0	0.00%	0	0.00%	6,203,415	0.02%
32777	0	0.00%	0	0.00%	0	0.00%	679,706	0.00%	686,455	0.00%
34476	0	0.00%	30,908,533	0.25%	0	0.00%	31,030,943	0.22%	61,939,475	0.18%
32778	0	0.00%	15,317,663	0.13%	0	0.00%	24,559,299	0.17%	39,877,000	0.11%
34759	59,015,178	0.72%	24,929,175	0.20%	0	0.00%	46,140,649	0.33%	130,085,002	0.37%
33486	0	0.00%	22,194,641	0.18%	0	0.00%	0	0.00%	22,195,001	0.06%
32920	18,303,868	0.22%	31,138,464	0.26%	0	0.00%	38,274,267	0.27%	87,716,600	0.25%
32779	79,477,788	0.97%	46,050,155	0.38%	0	0.00%	80,223,777	0.57%	205,751,720	0.58%
33062	0	0.00%	33,256,485	0.27%	0	0.00%	0	0.00%	33,259,279	0.09%
34760	0	0.00%	0	0.00%	0	0.00%	1,678,844	0.01%	1,692,963	0.00%
33487	0	0.00%	56,136,626	0.46%	0	0.00%	17,100,838	0.12%	73,237,924	0.21%
33770	0	0.00%	16,273,525	0.13%	0	0.00%	1,103,610	0.01%	17,377,134	0.05%
34478	0	0.00%	0	0.00%	0	0.00%	0	0.00%	507,506	0.00%
33629	0	0.00%	4,155,267	0.03%	0	0.00%	4,137,838	0.03%	8,293,105	0.02%
32780	48,905,987	0.60%	49,189,116	0.40%	0	0.00%	80,993,257	0.57%	179,088,360	0.51%
34761	27,847,275	0.34%	1,254,560	0.01%	0	0.00%	56,977,976	0.40%	86,079,812	0.24%
33912	80,823,809	0.98%	0	0.00%	0	0.00%	0	0.00%	80,824,217	0.23%
33771	0	0.00%	14,771,297	0.12%	0	0.00%	0	0.00%	14,771,815	0.04%
32922	6,911,559	0.08%	16,586,589	0.14%	0	0.00%	23,131,425	0.16%	46,629,573	0.13%
34479	0	0.00%	8,102,942	0.07%	0	0.00%	14,586,524	0.10%	22,689,466	0.06%
33913	32,443,154	0.39%	0	0.00%	0	0.00%	0	0.00%	32,443,374	0.09%
33064	0	0.00%	8,070,829	0.07%	0	0.00%	0	0.00%	8,071,444	0.02%
34762	0	0.00%	0	0.00%	0	0.00%	721,430	0.01%	1,131,086	0.00%
32640	0	0.00%	0	0.00%	0	0.00%	4,298,184	0.03%	4,380,283	0.01%
33772	0	0.00%	32,214,088	0.26%	0	0.00%	0	0.00%	32,214,353	0.09%
34480	0	0.00%	11,855,832	0.10%	0	0.00%	18,736,756	0.13%	30,592,588	0.09%
33914	144,795,628	1.76%	0	0.00%	0	0.00%	0	0.00%	144,795,773	0.41%
32641	0	0.00%	0	0.00%	0	0.00%	3,767,535	0.03%	4,063,365	0.01%
33773	0	0.00%	11,803,201	0.10%	0	0.00%	0	0.00%	11,803,364	0.03%
32924	0	0.00%	0	0.00%	0	0.00%	574,015	0.00%	1,309,152	0.00%
34481	0	0.00%	19,347,047	0.16%	0	0.00%	19,423,979	0.14%	38,771,025	0.11%

ZIP Code	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32359	0	0.00%	2,012,696	0.02%	0	0.00%	0	0.00%	2,012,696	0.01%
33774	0	0.00%	27,257,776	0.22%	0	0.00%	3,537,531	0.03%	30,795,307	0.09%
33916	19,151,183	0.23%	0	0.00%	0	0.00%	0	0.00%	19,151,183	0.05%
32501	0	0.00%	0	0.00%	12,695,131	1.78%	0	0.00%	12,695,131	0.04%
34482	0	0.00%	21,097,030	0.17%	0	0.00%	21,181,350	0.15%	42,278,379	0.12%
32784	0	0.00%	6,381,857	0.05%	0	0.00%	10,906,111	0.08%	17,287,969	0.05%
33067	0	0.00%	2,545,365	0.02%	0	0.00%	0	0.00%	2,545,452	0.01%
32643	0	0.00%	0	0.00%	0	0.00%	5,579,278	0.04%	5,723,750	0.02%
32926	14,837,460	0.18%	26,557,090	0.22%	0	0.00%	44,871,424	0.32%	86,265,973	0.25%
33634	0	0.00%	10,722,777	0.09%	0	0.00%	0	0.00%	11,158,324	0.03%
33917	72,131,631	0.88%	0	0.00%	0	0.00%	0	0.00%	72,131,666	0.20%
32502	0	0.00%	0	0.00%	1,096,428	0.15%	0	0.00%	1,096,428	0.00%
33776	0	0.00%	33,092,170	0.27%	0	0.00%	637,173	0.00%	33,729,343	0.10%
32927	32,345,871	0.39%	32,534,717	0.27%	0	0.00%	53,927,214	0.38%	118,807,803	0.34%
33493	0	0.00%	1,687,812	0.01%	0	0.00%	1,215,745	0.01%	2,903,557	0.01%
32503	0	0.00%	0	0.00%	41,925,104	5.87%	0	0.00%	41,925,104	0.12%
34484	0	0.00%	3,246,485	0.03%	0	0.00%	3,254,812	0.02%	6,501,298	0.02%
33635	0	0.00%	9,193,279	0.08%	0	0.00%	814,418	0.01%	10,007,697	0.03%
33069	0	0.00%	16,261,412	0.13%	0	0.00%	0	0.00%	16,262,246	0.05%
33777	0	0.00%	26,563,837	0.22%	0	0.00%	0	0.00%	26,565,005	0.08%
34202	0	0.00%	0	0.00%	0	0.00%	3,531,047	0.03%	3,531,154	0.01%
32504	0	0.00%	0	0.00%	31,500,500	4.41%	0	0.00%	31,500,500	0.09%
33919	100,919,784	1.23%	0	0.00%	0	0.00%	0	0.00%	100,920,966	0.29%
33778	0	0.00%	16,684,026	0.14%	0	0.00%	0	0.00%	16,684,418	0.05%
32080	0	0.00%	33,653,272	0.28%	0	0.00%	38,478,469	0.27%	72,131,984	0.20%
33920	8,349,100	0.10%	0	0.00%	0	0.00%	0	0.00%	8,530,882	0.02%
32505	0	0.00%	0	0.00%	18,040,108	2.53%	0	0.00%	18,040,108	0.05%
33637	0	0.00%	1,181,593	0.01%	0	0.00%	6,898,722	0.05%	8,080,315	0.02%
34769	53,071,654	0.65%	14,914,666	0.12%	0	0.00%	42,254,623	0.30%	110,240,943	0.31%
33496	0	0.00%	51,606,566	0.42%	0	0.00%	6,684,181	0.05%	58,290,747	0.17%
32506	0	0.00%	0	0.00%	38,471,168	5.39%	0	0.00%	38,471,168	0.11%
32789	111,852,576	1.36%	40,733,111	0.33%	0	0.00%	92,956,176	0.66%	245,541,863	0.70%
33921	54,480,026	0.66%	18,794,760	0.15%	0	0.00%	18,850,658	0.13%	92,125,444	0.26%
32648	0	0.00%	709,896	0.01%	0	0.00%	0	0.00%	709,896	0.00%
32931	23,943,986	0.29%	76,975,184	0.63%	0	0.00%	94,554,400	0.67%	195,473,570	0.56%
32082	0	0.00%	59,262,678	0.49%	0	0.00%	59,474,350	0.42%	118,737,028	0.34%
33922	30,338,437	0.37%	0	0.00%	0	0.00%	0	0.00%	30,338,437	0.09%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34205	0	0.00%	3,563,283	0.03%	0	0.00%	0	0.00%	3,563,635	0.01%
32507	0	0.00%	0	0.00%	81,168,508	11.36%	0	0.00%	81,168,508	0.23%
34488	0	0.00%	0	0.00%	0	0.00%	4,530,405	0.03%	4,796,637	0.01%
33073	0	0.00%	2,120,695	0.02%	0	0.00%	0	0.00%	2,120,884	0.01%
34771	34,435,116	0.42%	19,087,523	0.16%	0	0.00%	30,169,554	0.21%	83,692,194	0.24%
33498	0	0.00%	21,572,384	0.18%	0	0.00%	0	0.00%	21,941,724	0.06%
33781	0	0.00%	11,947,965	0.10%	0	0.00%	0	0.00%	11,948,076	0.03%
32932	0	0.00%	0	0.00%	0	0.00%	601,446	0.00%	1,226,228	0.00%
34772	43,969,032	0.53%	15,698,242	0.13%	0	0.00%	35,133,828	0.25%	94,801,102	0.27%
33782	0	0.00%	14,907,345	0.12%	0	0.00%	0	0.00%	14,907,856	0.04%
32084	0	0.00%	19,729,233	0.16%	0	0.00%	19,797,341	0.14%	39,526,574	0.11%
34207	0	0.00%	16,192,982	0.13%	0	0.00%	0	0.00%	16,193,888	0.05%
32792	89,737,500	1.09%	30,769,861	0.25%	0	0.00%	74,383,831	0.53%	194,891,193	0.55%
34773	2,122,737	0.03%	2,547,414	0.02%	0	0.00%	4,025,454	0.03%	8,695,605	0.02%
33924	85,068,331	1.03%	13,049,415	0.11%	0	0.00%	9,794,621	0.07%	107,912,366	0.31%
32934	511,452	0.01%	39,312,160	0.32%	0	0.00%	55,065,346	0.39%	94,888,958	0.27%
34208	0	0.00%	0	0.00%	0	0.00%	1,732,150	0.01%	1,732,777	0.00%
34491	0	0.00%	23,392,665	0.19%	0	0.00%	34,776,525	0.25%	58,169,190	0.17%
32793	0	0.00%	0	0.00%	0	0.00%	0	0.00%	893,437	0.00%
33076	0	0.00%	1,001,041	0.01%	0	0.00%	0	0.00%	1,001,082	0.00%
32935	0	0.00%	81,578,571	0.67%	0	0.00%	102,143,363	0.72%	183,722,387	0.52%
32086	0	0.00%	2,320,279	0.02%	0	0.00%	20,495,679	0.15%	22,815,958	0.06%
34209	0	0.00%	7,089,196	0.06%	0	0.00%	0	0.00%	7,090,080	0.02%
32794	0	0.00%	0	0.00%	0	0.00%	0	0.00%	586,481	0.00%
32653	0	0.00%	0	0.00%	0	0.00%	10,521,819	0.07%	10,993,285	0.03%
33785	0	0.00%	28,018,483	0.23%	0	0.00%	18,684,460	0.13%	46,702,943	0.13%
34210	0	0.00%	17,891,968	0.15%	0	0.00%	0	0.00%	17,893,174	0.05%
33786	0	0.00%	8,332,055	0.07%	0	0.00%	6,174,495	0.04%	14,506,551	0.04%
32937	5,983,753	0.07%	108,064,925	0.89%	0	0.00%	116,923,261	0.83%	230,971,939	0.66%
32796	26,649,606	0.32%	26,804,518	0.22%	0	0.00%	36,980,927	0.26%	90,435,051	0.26%
33928	37,016,953	0.45%	0	0.00%	0	0.00%	0	0.00%	37,017,210	0.11%
34212	0	0.00%	0	0.00%	0	0.00%	1,468,163	0.01%	1,468,220	0.00%
32514	0	0.00%	0	0.00%	36,921,511	5.17%	0	0.00%	36,921,511	0.10%
34637	0	0.00%	5,274,021	0.04%	0	0.00%	5,294,823	0.04%	10,568,844	0.03%
32656	0	0.00%	0	0.00%	0	0.00%	6,871,076	0.05%	6,871,076	0.02%
33647	0	0.00%	50,964,923	0.42%	0	0.00%	74,043,537	0.53%	125,008,459	0.36%
32798	0	0.00%	0	0.00%	0	0.00%	4,201,160	0.03%	4,233,664	0.01%

ZIP Code	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34638	0	0.00%	13,316,278	0.11%	0	0.00%	13,369,552	0.09%	26,685,830	0.08%
32940	3,859,787	0.05%	92,708,149	0.76%	0	0.00%	122,885,892	0.87%	219,453,827	0.62%
33931	56,349,006	0.69%	12,798,374	0.11%	0	0.00%	0	0.00%	69,147,658	0.20%
32233	0	0.00%	3,124,916	0.03%	0	0.00%	18,065,301	0.13%	21,190,217	0.06%
34639	0	0.00%	23,741,356	0.19%	0	0.00%	23,834,261	0.17%	47,575,617	0.14%
33932	0	0.00%	0	0.00%	0	0.00%	0	0.00%	589,914	0.00%
34215	0	0.00%	1,625,666	0.01%	0	0.00%	1,109,602	0.01%	2,735,268	0.01%
34498	0	0.00%	726,095	0.01%	0	0.00%	0	0.00%	738,009	0.00%
34216	0	0.00%	9,400,037	0.08%	0	0.00%	0	0.00%	9,585,345	0.03%
32801	18,704,835	0.23%	8,111,818	0.07%	0	0.00%	15,776,291	0.11%	42,592,945	0.12%
34217	0	0.00%	33,130,555	0.27%	0	0.00%	4,427,005	0.03%	37,557,699	0.11%
32802	0	0.00%	0	0.00%	0	0.00%	0	0.00%	952,706	0.00%
32095	0	0.00%	1,425,800	0.01%	0	0.00%	9,449,611	0.07%	10,875,411	0.03%
33510	0	0.00%	593,806	0.00%	0	0.00%	17,551,609	0.12%	18,145,424	0.05%
33935	0	0.00%	0	0.00%	0	0.00%	7,498,249	0.05%	7,659,698	0.02%
34218	0	0.00%	729,723	0.01%	0	0.00%	0	0.00%	950,788	0.00%
32803	58,720,538	0.71%	20,146,083	0.17%	0	0.00%	48,653,740	0.35%	127,520,362	0.36%
33511	0	0.00%	1,998,097	0.02%	0	0.00%	34,075,933	0.24%	36,074,067	0.10%
34785	0	0.00%	9,449,923	0.08%	0	0.00%	12,856,380	0.09%	22,306,304	0.06%
33936	22,563,750	0.27%	0	0.00%	0	0.00%	0	0.00%	22,563,783	0.06%
32804	51,281,934	0.62%	21,058,914	0.17%	0	0.00%	51,768,029	0.37%	124,108,877	0.35%
34219	0	0.00%	0	0.00%	0	0.00%	17,119,956	0.12%	17,119,956	0.05%
32805	19,110,754	0.23%	7,545,202	0.06%	0	0.00%	19,297,828	0.14%	45,953,784	0.13%
34786	76,316,479	0.93%	55,055,916	0.45%	0	0.00%	96,939,439	0.69%	228,311,834	0.65%
33513	0	0.00%	7,938,180	0.07%	0	0.00%	7,969,937	0.06%	15,908,117	0.05%
32664	0	0.00%	589,316	0.00%	0	0.00%	591,684	0.00%	1,181,000	0.00%
33938	0	0.00%	0	0.00%	0	0.00%	0	0.00%	596,465	0.00%
34787	1,915,407	0.02%	1,926,552	0.02%	0	0.00%	63,451,569	0.45%	67,293,528	0.19%
34221	0	0.00%	2,836,180	0.02%	0	0.00%	2,824,239	0.02%	5,660,491	0.02%
32806	76,067,328	0.93%	25,330,041	0.21%	0	0.00%	62,928,422	0.45%	164,325,792	0.47%
33514	0	0.00%	1,032,344	0.01%	0	0.00%	1,397,796	0.01%	2,430,140	0.01%
32948	0	0.00%	4,005,239	0.03%	0	0.00%	6,224,414	0.04%	10,229,653	0.03%
34788	0	0.00%	12,276,024	0.10%	0	0.00%	21,463,919	0.15%	33,739,947	0.10%
32807	54,359,111	0.66%	16,065,324	0.13%	0	0.00%	39,415,979	0.28%	109,840,415	0.31%
32949	0	0.00%	7,791,485	0.06%	0	0.00%	9,800,988	0.07%	17,592,472	0.05%
34223	52,604,301	0.64%	3,858,173	0.03%	0	0.00%	0	0.00%	56,462,609	0.16%
32808	36,582,682	0.45%	1,691,708	0.01%	0	0.00%	51,420,431	0.36%	89,694,821	0.25%

ZIP Code	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32667	0	0.00%	2,338,766	0.02%	0	0.00%	2,348,182	0.02%	4,686,948	0.01%
32950	0	0.00%	13,834,881	0.11%	0	0.00%	17,501,278	0.12%	31,336,159	0.09%
34224	46,069,843	0.56%	0	0.00%	0	0.00%	2,364,921	0.02%	48,434,764	0.14%
32526	0	0.00%	0	0.00%	42,321,431	5.92%	0	0.00%	42,321,431	0.12%
32809	48,213,201	0.59%	14,710,325	0.12%	0	0.00%	35,025,340	0.25%	97,948,866	0.28%
32951	0	0.00%	63,797,440	0.52%	0	0.00%	124,409,778	0.88%	188,207,650	0.53%
32668	0	0.00%	3,409,333	0.03%	0	0.00%	2,864,289	0.02%	6,273,622	0.02%
32102	0	0.00%	0	0.00%	0	0.00%	1,797,953	0.01%	1,797,953	0.01%
32810	40,508,547	0.49%	1,187,534	0.01%	0	0.00%	40,895,291	0.29%	82,591,372	0.23%
33801	0	0.00%	3,232,071	0.03%	0	0.00%	38,942,708	0.28%	42,175,114	0.12%
32952	22,299,828	0.27%	74,609,333	0.61%	0	0.00%	87,357,603	0.62%	184,266,765	0.52%
32669	0	0.00%	7,491,189	0.06%	0	0.00%	7,519,597	0.05%	15,010,786	0.04%
32811	28,158,383	0.34%	11,537,380	0.09%	0	0.00%	23,643,630	0.17%	63,339,393	0.18%
33802	0	0.00%	647,742	0.01%	0	0.00%	1,115,296	0.01%	1,763,077	0.01%
32953	20,748,846	0.25%	48,369,723	0.40%	0	0.00%	67,471,947	0.48%	136,590,516	0.39%
32812	78,311,422	0.95%	22,935,293	0.19%	0	0.00%	56,744,404	0.40%	157,991,119	0.45%
34652	0	0.00%	18,141,462	0.15%	0	0.00%	30,111,509	0.21%	48,252,970	0.14%
33803	0	0.00%	2,085,149	0.02%	0	0.00%	36,733,893	0.26%	38,819,164	0.11%
33945	904,454	0.01%	0	0.00%	0	0.00%	0	0.00%	904,454	0.00%
34228	0	0.00%	83,308,583	0.68%	0	0.00%	23,506,672	0.17%	106,816,224	0.30%
34653	0	0.00%	18,581,327	0.15%	0	0.00%	18,640,235	0.13%	37,221,562	0.11%
32955	32,768,703	0.40%	67,167,013	0.55%	0	0.00%	92,747,959	0.66%	192,683,675	0.55%
33521	0	0.00%	562,404	0.00%	0	0.00%	784,114	0.01%	1,346,518	0.00%
33946	13,665,186	0.17%	6,754,808	0.06%	0	0.00%	6,768,791	0.05%	27,188,785	0.08%
34229	0	0.00%	15,747,822	0.13%	0	0.00%	0	0.00%	15,747,908	0.04%
32814	8,875,498	0.11%	5,619,379	0.05%	0	0.00%	6,836,793	0.05%	21,331,670	0.06%
34654	0	0.00%	24,730,869	0.20%	0	0.00%	15,738,663	0.11%	40,469,533	0.11%
33805	0	0.00%	862,575	0.01%	0	0.00%	19,129,493	0.14%	19,992,148	0.06%
33947	29,866,866	0.36%	0	0.00%	0	0.00%	955,166	0.01%	30,822,033	0.09%
34655	0	0.00%	39,951,402	0.33%	0	0.00%	40,081,434	0.28%	80,032,836	0.23%
33523	0	0.00%	7,798,919	0.06%	0	0.00%	13,100,621	0.09%	20,899,540	0.06%
32957	0	0.00%	560,966	0.00%	0	0.00%	843,245	0.01%	1,404,211	0.00%
33948	62,472,130	0.76%	0	0.00%	0	0.00%	0	0.00%	62,472,321	0.18%
32250	0	0.00%	7,263,202	0.06%	0	0.00%	27,622,623	0.20%	34,885,825	0.10%
34231	0	0.00%	37,937,238	0.31%	0	0.00%	0	0.00%	37,938,826	0.11%
32533	0	0.00%	0	0.00%	30,557,000	4.28%	0	0.00%	30,557,000	0.09%
34797	0	0.00%	1,724,759	0.01%	0	0.00%	4,589,598	0.03%	6,314,357	0.02%

ZIP Code	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32958	0	0.00%	73,089,219	0.60%	0	0.00%	96,730,209	0.69%	169,819,462	0.48%
32817	73,195,254	0.89%	21,620,066	0.18%	0	0.00%	53,054,216	0.38%	147,869,536	0.42%
32534	0	0.00%	0	0.00%	11,555,457	1.62%	0	0.00%	11,555,457	0.03%
33525	0	0.00%	11,723,230	0.10%	0	0.00%	18,253,158	0.13%	29,976,394	0.09%
32818	43,086,809	0.52%	1,614,198	0.01%	0	0.00%	59,610,301	0.42%	104,311,308	0.30%
33950	207,982,428	2.53%	0	0.00%	0	0.00%	7,221,597	0.05%	215,204,345	0.61%
32535	0	0.00%	0	0.00%	2,437,717	0.34%	0	0.00%	2,437,717	0.01%
33809	0	0.00%	19,199,639	0.16%	0	0.00%	44,793,943	0.32%	63,993,586	0.18%
32960	0	0.00%	58,677,426	0.48%	0	0.00%	71,378,532	0.51%	130,056,128	0.37%
32819	79,768,684	0.97%	34,912,380	0.29%	0	0.00%	80,494,628	0.57%	195,175,692	0.55%
34234	0	0.00%	1,413,736	0.01%	0	0.00%	0	0.00%	1,413,978	0.00%
33810	0	0.00%	25,224,256	0.21%	0	0.00%	52,298,119	0.37%	77,522,375	0.22%
32961	0	0.00%	0	0.00%	0	0.00%	574,037	0.00%	1,056,133	0.00%
32112	0	0.00%	0	0.00%	0	0.00%	2,384,647	0.02%	2,464,525	0.01%
33527	0	0.00%	7,162,299	0.06%	0	0.00%	11,463,620	0.08%	18,625,919	0.05%
33952	117,404,608	1.43%	0	0.00%	0	0.00%	0	0.00%	117,404,608	0.33%
32820	12,234,218	0.15%	6,067,853	0.05%	0	0.00%	9,091,864	0.06%	27,393,935	0.08%
33811	0	0.00%	0	0.00%	0	0.00%	22,521,569	0.16%	22,781,010	0.06%
32962	0	0.00%	62,985,231	0.52%	0	0.00%	75,054,765	0.53%	138,040,125	0.39%
32113	0	0.00%	2,771,197	0.02%	0	0.00%	4,483,533	0.03%	7,254,730	0.02%
33953	16,846,754	0.21%	0	0.00%	0	0.00%	818,830	0.01%	17,665,584	0.05%
32821	32,674,747	0.40%	1,288,414	0.01%	0	0.00%	23,749,844	0.17%	57,713,005	0.16%
34236	0	0.00%	30,023,171	0.25%	0	0.00%	0	0.00%	30,024,248	0.09%
32963	0	0.00%	212,914,044	1.75%	0	0.00%	508,993,266	3.61%	721,908,548	2.05%
32114	36,287,489	0.44%	22,264,468	0.18%	0	0.00%	36,564,680	0.26%	95,116,637	0.27%
32680	0	0.00%	2,876,231	0.02%	0	0.00%	0	0.00%	2,919,053	0.01%
33812	0	0.00%	0	0.00%	0	0.00%	9,353,719	0.07%	9,437,843	0.03%
33954	38,976,639	0.47%	0	0.00%	0	0.00%	0	0.00%	39,126,334	0.11%
32822	69,929,848	0.85%	22,591,138	0.19%	0	0.00%	50,450,039	0.36%	142,971,025	0.41%
33813	0	0.00%	1,246,332	0.01%	0	0.00%	86,084,675	0.61%	87,331,065	0.25%
32964	0	0.00%	553,470	0.00%	0	0.00%	665,863	0.00%	1,219,335	0.00%
32681	0	0.00%	0	0.00%	0	0.00%	0	0.00%	549,045	0.00%
34945	0	0.00%	8,828,136	0.07%	0	0.00%	11,670,401	0.08%	20,498,536	0.06%
33955	63,445,661	0.77%	0	0.00%	0	0.00%	0	0.00%	63,445,717	0.18%
34238	0	0.00%	5,840,287	0.05%	0	0.00%	0	0.00%	5,841,021	0.02%
32116	592,597	0.01%	757,562	0.01%	0	0.00%	835,679	0.01%	2,185,838	0.01%
34946	0	0.00%	11,143,245	0.09%	0	0.00%	14,444,953	0.10%	25,588,198	0.07%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32824	68,613,999	0.83%	23,553,748	0.19%	0	0.00%	55,963,710	0.40%	148,131,457	0.42%
33956	44,875,000	0.55%	0	0.00%	0	0.00%	0	0.00%	45,006,545	0.13%
34239	0	0.00%	1,404,126	0.01%	0	0.00%	0	0.00%	1,404,193	0.00%
32541	0	0.00%	0	0.00%	44,517,471	6.23%	0	0.00%	44,517,471	0.13%
33815	0	0.00%	0	0.00%	0	0.00%	6,860,552	0.05%	7,206,464	0.02%
32966	0	0.00%	38,804,341	0.32%	0	0.00%	61,279,111	0.43%	100,083,548	0.28%
32117	37,950,618	0.46%	22,743,157	0.19%	0	0.00%	31,413,761	0.22%	92,107,536	0.26%
34947	0	0.00%	31,611,688	0.26%	0	0.00%	39,592,696	0.28%	71,204,732	0.20%
32825	111,966,431	1.36%	36,894,342	0.30%	0	0.00%	81,696,404	0.58%	230,557,178	0.65%
33957	170,646,455	2.08%	9,576, 743	0.08%	0	0.00%	0	0.00%	180,223,332	0.51%
32967	0	0.00%	43,867,252	0.36%	0	0.00%	53,711,256	0.38%	97,578,797	0.28%
32118	106,516,515	1.30%	80,657,243	0.66%	0	0.00%	89,275,645	0.63%	276,449,403	0.79%
32826	39,506,765	0.48%	13,300,206	0.11%	0	0.00%	28,377,053	0.20%	81,184,024	0.23%
32119	62,086,048	0.76%	44,487,680	0.37%	0	0.00%	54,445,236	0.39%	161,018,964	0.46%
34949	0	0.00%	67,753,527	0.56%	0	0.00%	93,624,980	0.66%	161,379,395	0.46%
33534	0	0.00%	0	0.00%	0	0.00%	5,498,509	0.04%	5,586,959	0.02%
32968	0	0.00%	34,999,044	0.29%	0	0.00%	38,829,022	0.28%	73,828,066	0.21%
34242	0	0.00%	50,101,465	0.41%	0	0.00%	0	0.00%	50,104,356	0.14%
32827	17,379,709	0.21%	6,926,442	0.06%	0	0.00%	14,216,305	0.10%	38,522,455	0.11%
Total	$8,205,958,838		$12,151,952,390		$712,235,666		$14,058,937,899		$35,162,284,715	

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34950	0	0.00%	16,571,114	0.14%	0	0.00%	19,448,971	0.15%	36,020,150	0.11%
34667	0	0.00%	37,392,077	0.31%	0	0.00%	48,788,907	0.37%	86,180,984	0.26%
32686	0	0.00%	3,216,363	0.03%	0	0.00%	4,226,896	0.03%	7,443,259	0.02%
33960	0	0.00%	560,548	0.00%	0	0.00%	716,087	0.01%	1,665,206	0.01%
32828	106,095,760	1.49%	49,575,285	0.41%	0	0.00%	79,452,861	0.61%	235,123,906	0.71%
34102	0	0.00%	39,694,689	0.33%	0	0.00%	0	0.00%	39,698,314	0.12%
34951	0	0.00%	36,832,536	0.30%	0	0.00%	47,048,957	0.36%	83,881,504	0.25%
34668	0	0.00%	27,992,712	0.23%	0	0.00%	43,058,749	0.33%	71,051,461	0.22%
32829	31,241,285	0.44%	14,756,440	0.12%	0	0.00%	21,746,214	0.17%	67,743,940	0.21%
34103	0	0.00%	41,496,417	0.34%	0	0.00%	18,818,153	0.14%	60,316,590	0.18%
34952	0	0.00%	86,424,093	0.71%	0	0.00%	96,617,404	0.74%	183,041,748	0.55%
33820	0	0.00%	0	0.00%	0	0.00%	0	0.00%	739,164	0.00%
34669	0	0.00%	11,272,188	0.09%	0	0.00%	8,041,685	0.06%	19,313,873	0.06%
32547	0	0.00%	0	0.00%	21,291,621	3.18%	0	0.00%	21,291,621	0.06%
34953	0	0.00%	105,079,371	0.87%	0	0.00%	113,645,474	0.87%	218,724,846	0.66%
33538	0	0.00%	5,128,206	0.04%	0	0.00%	5,142,240	0.04%	10,270,446	0.03%
32548	0	0.00%	0	0.00%	22,555,484	3.36%	0	0.00%	22,555,484	0.07%
32407	0	0.00%	0	0.00%	2,357,764	0.35%	0	0.00%	2,357,852	0.01%
34105	0	0.00%	7,549,782	0.06%	0	0.00%	0	0.00%	7,553,105	0.02%
32124	4,470,008	0.06%	3,761,227	0.03%	0	0.00%	4,519,995	0.03%	12,751,231	0.04%
32266	0	0.00%	7,200,237	0.06%	0	0.00%	7,220,309	0.06%	14,420,545	0.04%
32832	31,688,660	0.44%	19,289,017	0.16%	0	0.00%	25,901,483	0.20%	76,879,161	0.23%
32408	0	0.00%	0	0.00%	2,732,881	0.41%	0	0.00%	2,733,160	0.01%
33823	17,201,210	0.24%	0	0.00%	0	0.00%	33,133,054	0.25%	50,675,821	0.15%
33540	0	0.00%	4,057,600	0.03%	0	0.00%	6,415,377	0.05%	10,472,977	0.03%
32550	0	0.00%	0	0.00%	5,364,741	0.80%	0	0.00%	5,364,741	0.02%
32833	17,248,986	0.24%	7,999,107	0.07%	0	0.00%	12,851,899	0.10%	38,099,992	0.12%
34956	0	0.00%	5,538,635	0.05%	0	0.00%	5,553,749	0.04%	11,092,384	0.03%
32692	0	0.00%	584,949	0.00%	0	0.00%	0	0.00%	584,949	0.00%
33541	0	0.00%	10,267,657	0.08%	0	0.00%	15,374,622	0.12%	25,642,292	0.08%
33966	17,386,873	0.24%	0	0.00%	0	0.00%	0	0.00%	17,387,018	0.05%
34108	0	0.00%	56,660,946	0.47%	0	0.00%	0	0.00%	56,670,883	0.17%
34957	0	0.00%	74,028,585	0.61%	0	0.00%	73,970,818	0.56%	147,999,515	0.45%
32127	56,305,624	0.79%	56,757,606	0.47%	0	0.00%	68,064,485	0.52%	181,127,715	0.55%
32976	0	0.00%	26,268,363	0.22%	0	0.00%	77,749,953	0.59%	104,018,318	0.31%
33825	43,531,939	0.61%	13,255,166	0.11%	0	0.00%	33,424,775	0.26%	90,211,881	0.27%
32693	0	0.00%	4,692,981	0.04%	0	0.00%	0	0.00%	4,763,158	0.01%
33542	0	0.00%	9,803,111	0.08%	0	0.00%	15,724,383	0.12%	25,527,494	0.08%
33401	0	0.00%	69,748,920	0.58%	0	0.00%	52,391,032	0.40%	122,140,188	0.37%
33967	26,074,392	0.37%	0	0.00%	0	0.00%	0	0.00%	26,074,392	0.08%
32835	52,913,684	0.74%	27,523,455	0.23%	0	0.00%	53,443,897	0.41%	133,881,037	0.41%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34109	0	0.00%	7,659,610	0.06%	0	0.00%	0	0.00%	7,664,868	0.02%
32128	43,683,158	0.61%	21,859,040	0.18%	0	0.00%	37,169,411	0.28%	102,711,609	0.31%
32694	0	0.00%	0	0.00%	0	0.00%	747,391	0.01%	747,391	0.00%
33543	0	0.00%	22,204,068	0.18%	0	0.00%	29,618,222	0.23%	51,822,290	0.16%
34251	0	0.00%	0	0.00%	0	0.00%	5,120,763	0.04%	5,120,763	0.02%
32836	50,826,039	0.71%	28,541,510	0.24%	0	0.00%	61,395,425	0.47%	140,762,974	0.43%
34110	12,840,311	0.18%	12,928,769	0.11%	0	0.00%	0	0.00%	25,770,423	0.08%
32129	32,603,392	0.46%	21,150,081	0.17%	0	0.00%	27,316,036	0.21%	81,069,509	0.25%
33827	8,678,410	0.12%	3,865,752	0.03%	0	0.00%	6,133,883	0.05%	18,678,044	0.06%
33544	0	0.00%	20,062,935	0.17%	0	0.00%	25,397,642	0.19%	45,460,578	0.14%
33403	0	0.00%	19,664,620	0.16%	0	0.00%	14,418,182	0.11%	34,082,829	0.10%
32837	97,760,548	1.37%	36,347,606	0.30%	0	0.00%	72,116,706	0.55%	206,224,861	0.62%
32413	0	0.00%	0	0.00%	26,224,912	3.91%	0	0.00%	26,225,109	0.08%
32130	0	0.00%	0	0.00%	0	0.00%	3,158,840	0.02%	3,217,579	0.01%
34677	0	0.00%	22,474,982	0.19%	0	0.00%	3,565,143	0.03%	26,040,125	0.08%
32696	0	0.00%	5,786,548	0.05%	0	0.00%	5,154,613	0.04%	10,941,161	0.03%
33545	0	0.00%	9,875,041	0.08%	0	0.00%	11,825,933	0.09%	21,700,975	0.07%
33404	0	0.00%	68,655,912	0.57%	0	0.00%	51,099,409	0.39%	119,755,802	0.36%
33971	18,223,736	0.26%	0	0.00%	0	0.00%	0	0.00%	18,223,795	0.06%
32839	27,674,170	0.39%	12,126,474	0.10%	0	0.00%	23,289,167	0.18%	63,089,811	0.19%
33405	0	0.00%	37,140,194	0.31%	0	0.00%	19,785,204	0.15%	56,925,405	0.17%
32132	10,838,307	0.15%	10,928,989	0.09%	0	0.00%	13,182,199	0.10%	34,949,495	0.11%
33830	21,359,234	0.30%	581,655	0.00%	0	0.00%	28,303,521	0.22%	50,244,410	0.15%
33547	0	0.00%	0	0.00%	0	0.00%	21,251,727	0.16%	21,570,818	0.07%
33972	8,810,291	0.12%	0	0.00%	0	0.00%	0	0.00%	9,166,203	0.03%
33406	0	0.00%	36,963,320	0.30%	0	0.00%	22,943,719	0.18%	59,907,078	0.18%
33548	0	0.00%	7,428,803	0.06%	0	0.00%	7,449,645	0.06%	14,878,447	0.05%
33973	4,128,441	0.06%	0	0.00%	0	0.00%	0	0.00%	4,128,441	0.01%
33407	0	0.00%	41,520,263	0.34%	0	0.00%	30,421,396	0.23%	71,942,285	0.22%
32134	0	0.00%	0	0.00%	0	0.00%	3,397,847	0.03%	3,465,196	0.01%
34681	0	0.00%	2,013,369	0.02%	0	0.00%	2,019,024	0.02%	4,032,393	0.01%
33549	0	0.00%	16,494,620	0.14%	0	0.00%	16,537,879	0.13%	33,032,500	0.10%
33974	7,680,365	0.11%	0	0.00%	0	0.00%	0	0.00%	7,680,365	0.02%
33408	0	0.00%	100,196,536	0.83%	0	0.00%	80,233,146	0.61%	180,430,207	0.55%
32701	29,617,617	0.42%	14,148,100	0.12%	0	0.00%	22,481,408	0.17%	66,247,126	0.20%
33409	0	0.00%	37,132,148	0.31%	0	0.00%	24,830,982	0.19%	61,963,292	0.19%
32136	22,340,473	0.31%	18,814,906	0.16%	0	0.00%	22,578,749	0.17%	63,734,127	0.19%
33834	4,638,533	0.07%	0	0.00%	0	0.00%	2,980,677	0.02%	7,646,143	0.02%
34683	0	0.00%	36,652,630	0.30%	0	0.00%	36,718,485	0.28%	73,371,115	0.22%
32702	0	0.00%	0	0.00%	0	0.00%	1,233,093	0.01%	1,233,093	0.00%
33976	7,270,041	0.10%	0	0.00%	0	0.00%	0	0.00%	7,270,041	0.02%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32561	0	0.00%	0	0.00%	38,780,840	5.78%	0	0.00%	38,780,840	0.12%
33410	0	0.00%	114,352,787	0.94%	0	0.00%	100,728,051	0.77%	215,080,970	0.65%
32137	0	0.00%	60,611,487	0.50%	0	0.00%	80,284,999	0.61%	140,897,195	0.43%
34684	0	0.00%	24,933,603	0.21%	0	0.00%	5,291,175	0.04%	30,224,778	0.09%
32703	26,131,341	0.37%	640,118	0.01%	0	0.00%	39,958,787	0.30%	66,730,246	0.20%
33411	0	0.00%	138,821,191	1.14%	0	0.00%	93,164,421	0.71%	231,985,855	0.70%
34685	0	0.00%	22,712,376	0.19%	0	0.00%	22,753,869	0.17%	45,466,245	0.14%
32563	0	0.00%	0	0.00%	33,064,291	4.93%	0	0.00%	33,064,291	0.10%
33412	0	0.00%	53,891,418	0.44%	0	0.00%	47,712,656	0.36%	101,604,096	0.31%
33837	34,449,692	0.48%	17,743,483	0.15%	0	0.00%	29,128,785	0.22%	81,321,959	0.25%
32564	0	0.00%	0	0.00%	1,059,033	0.16%	0	0.00%	1,059,033	0.00%
33413	0	0.00%	22,107,224	0.18%	0	0.00%	14,705,207	0.11%	36,812,447	0.11%
33838	5,345,871	0.08%	1,826,829	0.02%	0	0.00%	4,158,033	0.03%	11,330,733	0.03%
33980	43,464,745	0.61%	0	0.00%	0	0.00%	0	0.00%	43,465,205	0.13%
32565	0	0.00%	0	0.00%	4,034,848	0.60%	0	0.00%	4,034,848	0.01%
33414	0	0.00%	134,649,849	1.11%	0	0.00%	104,822,816	0.80%	239,472,772	0.73%
32141	27,362,908	0.38%	27,593,532	0.23%	0	0.00%	33,361,253	0.25%	88,317,693	0.27%
33839	3,361,629	0.05%	1,814,167	0.02%	0	0.00%	3,399,408	0.03%	8,575,205	0.03%
34688	0	0.00%	13,092,622	0.11%	0	0.00%	13,115,095	0.10%	26,207,717	0.08%
32707	66,656,116	0.94%	25,905,144	0.21%	0	0.00%	55,570,190	0.42%	148,131,449	0.45%
33556	0	0.00%	32,239,739	0.27%	0	0.00%	32,326,239	0.25%	64,565,978	0.20%
33981	24,853,784	0.35%	0	0.00%	0	0.00%	0	0.00%	24,853,814	0.08%
32566	0	0.00%	0	0.00%	42,505,783	6.34%	0	0.00%	42,505,783	0.13%
33415	0	0.00%	43,989,366	0.36%	0	0.00%	33,955,239	0.26%	77,944,808	0.24%
34972	0	0.00%	12,652,101	0.10%	0	0.00%	17,784,989	0.14%	30,437,091	0.09%
34689	0	0.00%	24,058,686	0.20%	0	0.00%	24,083,174	0.18%	48,141,860	0.15%
32708	109,107,124	1.53%	41,537,469	0.34%	0	0.00%	90,944,234	0.69%	241,588,826	0.73%
33982	40,327,114	0.57%	0	0.00%	0	0.00%	7,171,179	0.05%	47,498,293	0.14%
33841	10,713,199	0.15%	0	0.00%	0	0.00%	6,886,446	0.05%	17,658,012	0.05%
34690	0	0.00%	7,491,906	0.06%	0	0.00%	7,513,155	0.06%	15,005,061	0.05%
32709	3,025,121	0.04%	1,273,827	0.01%	0	0.00%	2,520,086	0.02%	6,819,033	0.02%
33558	0	0.00%	31,651,358	0.26%	0	0.00%	21,882,754	0.17%	53,534,112	0.16%
33983	68,819,763	0.97%	0	0.00%	0	0.00%	1,772,828	0.01%	70,592,814	0.21%
32568	0	0.00%	0	0.00%	2,586,376	0.39%	0	0.00%	2,586,376	0.01%
33417	0	0.00%	50,166,301	0.41%	0	0.00%	33,182,545	0.25%	83,349,062	0.25%
34266	60,586,545	0.85%	0	0.00%	0	0.00%	16,089,831	0.12%	76,676,376	0.23%
34974	0	0.00%	34,420,980	0.28%	0	0.00%	40,950,771	0.31%	75,767,701	0.23%
34691	0	0.00%	12,904,884	0.11%	0	0.00%	18,920,801	0.14%	31,825,684	0.10%
33559	0	0.00%	10,391,772	0.09%	0	0.00%	10,417,180	0.08%	20,808,952	0.06%
33701	0	0.00%	13,612,843	0.11%	0	0.00%	0	0.00%	13,613,595	0.04%
32569	0	0.00%	0	0.00%	10,657,923	1.59%	0	0.00%	10,657,923	0.03%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33418	0	0.00%	193,338,981	1.59%	0	0.00%	157,152,487	1.20%	350,491,631	1.06%
33843	20,644,380	0.29%	4,457,517	0.04%	0	0.00%	11,196,608	0.09%	36,298,506	0.11%
33702	0	0.00%	22,189,043	0.18%	0	0.00%	0	0.00%	22,190,638	0.07%
32570	0	0.00%	0	0.00%	20,537,266	3.06%	0	0.00%	20,537,266	0.06%
33844	43,021,980	0.60%	18,384,919	0.15%	0	0.00%	43,501,970	0.33%	104,908,868	0.32%
32712	960,223	0.01%	967,767	0.01%	0	0.00%	56,672,691	0.43%	58,600,682	0.18%
33703	0	0.00%	23,450,688	0.19%	0	0.00%	0	0.00%	23,450,958	0.07%
32571	0	0.00%	0	0.00%	28,939,256	4.32%	0	0.00%	28,939,256	0.09%
34269	20,313,093	0.28%	0	0.00%	0	0.00%	4,112,363	0.03%	24,425,456	0.07%
32713	28,543,385	0.40%	0	0.00%	0	0.00%	28,862,632	0.22%	57,725,560	0.17%
33704	0	0.00%	18,672,678	0.15%	0	0.00%	0	0.00%	18,673,778	0.06%
32148	0	0.00%	0	0.00%	0	0.00%	3,244,820	0.02%	3,244,820	0.01%
34695	0	0.00%	18,677,785	0.15%	0	0.00%	0	0.00%	18,679,182	0.06%
32714	36,307,350	0.51%	3,526,374	0.03%	0	0.00%	36,669,359	0.28%	76,503,083	0.23%
33563	0	0.00%	0	0.00%	0	0.00%	15,906,041	0.12%	16,327,942	0.05%
33705	0	0.00%	15,475,524	0.13%	0	0.00%	0	0.00%	15,475,758	0.05%
33847	0	0.00%	0	0.00%	0	0.00%	0	0.00%	587,565	0.00%
33706	0	0.00%	65,201,337	0.54%	0	0.00%	0	0.00%	65,202,930	0.20%
33848	0	0.00%	0	0.00%	0	0.00%	0	0.00%	1,013,489	0.00%
33565	0	0.00%	9,265,688	0.08%	0	0.00%	19,093,862	0.15%	28,359,550	0.09%
33990	71,612,473	1.00%	0	0.00%	0	0.00%	0	0.00%	71,612,476	0.22%
33707	0	0.00%	45,292,710	0.37%	0	0.00%	0	0.00%	45,294,253	0.14%
33849	0	0.00%	0	0.00%	0	0.00%	638,998	0.00%	974,266	0.00%
34698	0	0.00%	36,984,091	0.31%	0	0.00%	8,457,422	0.06%	45,441,512	0.14%
33566	0	0.00%	12,576,913	0.10%	0	0.00%	20,115,652	0.15%	32,692,571	0.10%
34981	0	0.00%	7,236,977	0.06%	0	0.00%	7,256,851	0.06%	14,493,828	0.04%
33991	47,431,826	0.67%	0	0.00%	0	0.00%	0	0.00%	47,431,826	0.14%
33708	0	0.00%	65,153,016	0.54%	0	0.00%	0	0.00%	65,155,148	0.20%
33850	6,019,992	0.08%	0	0.00%	0	0.00%	7,985,730	0.06%	14,076,560	0.04%
33567	0	0.00%	5,513,329	0.05%	0	0.00%	7,991,269	0.06%	13,504,599	0.04%
34982	0	0.00%	51,398,224	0.42%	0	0.00%	47,536,078	0.36%	98,934,527	0.30%
33709	0	0.00%	16,806,222	0.14%	0	0.00%	0	0.00%	16,806,676	0.05%
32577	0	0.00%	0	0.00%	5,352,596	0.80%	0	0.00%	5,352,596	0.02%
33426	0	0.00%	30,903,340	0.25%	0	0.00%	15,647,973	0.12%	46,551,372	0.14%
34275	0	0.00%	2,036,200	0.02%	0	0.00%	0	0.00%	2,036,591	0.01%
33851	1,505,449	0.02%	563,918	0.00%	0	0.00%	1,333,385	0.01%	3,402,752	0.01%
34134	56,473,429	0.79%	56,912,881	0.47%	0	0.00%	0	0.00%	113,386,414	0.34%
34983	0	0.00%	81,241,180	0.67%	0	0.00%	87,987,331	0.67%	169,228,514	0.51%
33993	42,709,266	0.60%	0	0.00%	0	0.00%	0	0.00%	42,709,266	0.13%
33710	0	0.00%	25,045,747	0.21%	0	0.00%	0	0.00%	25,045,797	0.08%
32578	0	0.00%	0	0.00%	1,597,850	0.24%	0	0.00%	1,597,850	0.00%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33852	29,890,272	0.42%	30,141,089	0.25%	0	0.00%	47,532,754	0.36%	107,564,116	0.33%
32720	927,004	0.01%	933,904	0.01%	0	0.00%	21,414,179	0.16%	23,275,087	0.07%
33569	0	0.00%	566,674	0.00%	0	0.00%	26,522,659	0.20%	27,089,340	0.08%
34135	10,440,625	0.15%	0	0.00%	0	0.00%	0	0.00%	10,441,161	0.03%
34984	0	0.00%	35,097,839	0.29%	0	0.00%	37,991,224	0.29%	73,089,063	0.22%
33711	0	0.00%	13,187,267	0.11%	0	0.00%	0	0.00%	13,187,630	0.04%
32579	0	0.00%	0	0.00%	10,690,025	1.59%	0	0.00%	10,690,025	0.03%
33428	0	0.00%	38,170,403	0.31%	0	0.00%	0	0.00%	38,170,960	0.12%
33853	23,723,694	0.33%	6,504,235	0.05%	0	0.00%	16,261,287	0.12%	46,489,216	0.14%
33570	0	0.00%	0	0.00%	0	0.00%	11,736,320	0.09%	12,009,128	0.04%
33712	0	0.00%	11,505,858	0.09%	0	0.00%	0	0.00%	11,506,020	0.03%
32580	0	0.00%	0	0.00%	2,801,942	0.42%	0	0.00%	2,801,942	0.01%
33854	948,556	0.01%	543,060	0.00%	0	0.00%	752,969	0.01%	2,244,585	0.01%
34420	0	0.00%	8,708,161	0.07%	0	0.00%	11,344,965	0.09%	20,053,126	0.06%
34986	0	0.00%	64,659,022	0.53%	0	0.00%	70,199,697	0.54%	134,858,974	0.41%
33713	0	0.00%	17,204,759	0.14%	0	0.00%	0	0.00%	17,204,820	0.05%
33430	0	0.00%	5,605,325	0.05%	0	0.00%	5,616,016	0.04%	11,221,351	0.03%
33855	2,637,471	0.04%	1,676,840	0.01%	0	0.00%	2,667,182	0.02%	6,981,493	0.02%
33572	0	0.00%	943,205	0.01%	0	0.00%	21,061,772	0.16%	22,004,977	0.07%
34987	0	0.00%	12,963,901	0.11%	0	0.00%	14,037,026	0.11%	27,000,940	0.08%
33714	0	0.00%	8,103,916	0.07%	0	0.00%	0	0.00%	8,104,171	0.02%
33431	0	0.00%	27,860,381	0.23%	0	0.00%	0	0.00%	27,862,184	0.08%
34705	0	0.00%	1,398,148	0.01%	0	0.00%	2,054,641	0.02%	3,452,789	0.01%
32724	24,205,233	0.34%	1,002,875	0.01%	0	0.00%	24,474,337	0.19%	49,682,446	0.15%
33573	0	0.00%	9,116,832	0.08%	0	0.00%	29,743,190	0.23%	38,860,084	0.12%
33715	0	0.00%	37,343,327	0.31%	0	0.00%	15,743,937	0.12%	53,087,263	0.16%
32583	0	0.00%	0	0.00%	15,561,560	2.32%	0	0.00%	15,561,560	0.05%
33432	0	0.00%	47,732,890	0.39%	0	0.00%	0	0.00%	47,736,815	0.14%
32159	0	0.00%	31,699,201	0.26%	0	0.00%	50,371,685	0.38%	82,070,885	0.25%
33857	988,227	0.01%	2,071,755	0.02%	0	0.00%	2,967,190	0.02%	6,027,171	0.02%
32725	52,342,793	0.73%	789,267	0.01%	0	0.00%	52,926,282	0.40%	106,058,342	0.32%
33716	0	0.00%	5,029,045	0.04%	0	0.00%	0	0.00%	5,029,621	0.02%
33433	0	0.00%	55,281,734	0.46%	0	0.00%	0	0.00%	55,283,653	0.17%
32301	0	0.00%	14,271,522	0.12%	0	0.00%	0	0.00%	14,271,522	0.04%
32726	0	0.00%	14,213,222	0.12%	0	0.00%	21,488,858	0.16%	35,702,080	0.11%
34990	0	0.00%	123,420,411	1.02%	0	0.00%	123,666,577	0.94%	247,087,091	0.75%
33434	0	0.00%	40,259,136	0.33%	0	0.00%	0	0.00%	40,261,309	0.12%
33859	23,825,095	0.33%	6,251,473	0.05%	0	0.00%	12,897,481	0.10%	42,974,049	0.13%
33576	0	0.00%	5,861,857	0.05%	0	0.00%	5,878,142	0.04%	11,739,999	0.04%
33435	0	0.00%	54,388,394	0.45%	0	0.00%	30,247,867	0.23%	84,636,642	0.26%
32303	0	0.00%	28,633,557	0.24%	0	0.00%	0	0.00%	28,633,557	0.09%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33860	0	0.00%	0	0.00%	0	0.00%	16,231,480	0.12%	16,428,762	0.05%
32162	0	0.00%	91,609,395	0.76%	0	0.00%	91,854,467	0.70%	183,463,862	0.56%
33436	0	0.00%	87,312,516	0.72%	0	0.00%	46,328,552	0.35%	133,641,364	0.40%
34285	7,285,251	0.10%	7,334,619	0.06%	0	0.00%	0	0.00%	14,620,496	0.04%
32304	0	0.00%	8,865,877	0.07%	0	0.00%	0	0.00%	8,865,877	0.03%
33578	0	0.00%	4,712,781	0.04%	0	0.00%	22,749,756	0.17%	27,462,564	0.08%
32163	0	0.00%	2,222,008	0.02%	0	0.00%	2,228,119	0.02%	4,450,127	0.01%
33437	0	0.00%	84,133,803	0.69%	0	0.00%	60,682,065	0.46%	144,816,197	0.44%
34286	30,809,054	0.43%	0	0.00%	0	0.00%	0	0.00%	31,014,756	0.09%
32305	0	0.00%	6,361,162	0.05%	0	0.00%	0	0.00%	6,361,162	0.02%
34711	0	0.00%	54,990,658	0.45%	0	0.00%	89,987,312	0.69%	144,977,984	0.44%
32730	7,444,977	0.10%	0	0.00%	0	0.00%	5,549,831	0.04%	13,292,652	0.04%
33579	0	0.00%	649,653	0.01%	0	0.00%	16,227,982	0.12%	16,877,641	0.05%
34428	0	0.00%	12,725,915	0.11%	0	0.00%	6,478,380	0.05%	19,204,295	0.06%
34145	0	0.00%	60,970,485	0.50%	0	0.00%	0	0.00%	60,982,251	0.18%
34994	0	0.00%	40,847,909	0.34%	0	0.00%	40,799,378	0.31%	81,647,350	0.25%
32164	0	0.00%	41,098,197	0.34%	0	0.00%	41,213,389	0.31%	82,311,586	0.25%
33438	0	0.00%	1, 129,444	0.01%	0	0.00%	971,030	0.01%	2,100,473	0.01%
34287	30,461,820	0.43%	0	0.00%	0	0.00%	0	0.00%	30,462,539	0.09%
34429	0	0.00%	13,568,618	0.11%	0	0.00%	13,598,294	0.10%	27,166,913	0.08%
32024	0	0.00%	0	0.00%	0	0.00%	9,272,363	0.07%	9,272,363	0.03%
34288	21,244,565	0.30%	0	0.00%	0	0.00%	0	0.00%	21,581,651	0.07%
32732	14,105,884	0.20%	5,108,316	0.04%	0	0.00%	10,268,629	0.08%	29,482,829	0.09%
34996	0	0.00%	65,456,278	0.54%	0	0.00%	60,637,861	0.46%	126,094,264	0.38%
32025	0	0.00%	0	0.00%	0	0.00%	8,758,073	0.07%	8,758,073	0.03%
33440	0	0.00%	9,051,962	0.07%	0	0.00%	9,076,526	0.07%	18,128,489	0.05%
34289	3,174,813	0.04%	0	0.00%	0	0.00%	0	0.00%	3,477,856	0.01%
32308	0	0.00%	18,407,334	0.15%	0	0.00%	0	0.00%	18,407,376	0.06%
34714	1,613,370	0.02%	1,624,584	0.01%	0	0.00%	19,113,099	0.15%	22,351,054	0.07%
34431	0	0.00%	10,455,443	0.09%	0	0.00%	8,669,145	0.07%	19,124,588	0.06%
34997	0	0.00%	118,432,247	0.98%	0	0.00%	107,931,681	0.82%	226,364,019	0.69%
33865	1,182,374	0.02%	0	0.00%	0	0.00%	705,163	0.01%	1,894,731	0.01%
33441	0	0.00%	19,451,518	0.16%	0	0.00%	0	0.00%	19,454,049	0.06%
32309	0	0.00%	29,732,686	0.25%	0	0.00%	0	0.00%	29,732,686	0.09%
34432	0	0.00%	14,097,289	0.12%	0	0.00%	14,135,930	0.11%	28,233,219	0.09%
32168	47,377,596	0.66%	31,952,183	0.26%	0	0.00%	47,895,666	0.37%	127,225,446	0.39%
34715	0	0.00%	11,230,763	0.09%	0	0.00%	18,576,908	0.14%	29,807,671	0.09%
33442	0	0.00%	27,950,864	0.23%	0	0.00%	0	0.00%	27,953,403	0.08%
34291	6,058,001	0.09%	0	0.00%	0	0.00%	0	0.00%	6,058,001	0.02%
32310	0	0.00%	5,173,740	0.04%	0	0.00%	0	0.00%	5,173,740	0.02%
32735	0	0.00%	3,449,822	0.03%	0	0.00%	4,877,147	0.04%	8,326,970	0.03%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33584	0	0.00%	0	0.00%	0	0.00%	16,663,695	0.13%	16,899,705	0.05%
34433	0	0.00%	8,735,334	0.07%	0	0.00%	6,948,080	0.05%	15,683,414	0.05%
32169	50,219,306	0.70%	57,508,533	0.47%	0	0.00%	63,315,287	0.48%	171,043,125	0.52%
33867	0	0.00%	0	0.00%	0	0.00%	0	0.00%	575,302	0.00%
32311	0	0.00%	16,049,274	0.13%	0	0.00%	0	0.00%	16,049,300	0.05%
32736	0	0.00%	0	0.00%	0	0.00%	12,405,438	0.09%	12,562,591	0.04%
33585	0	0.00%	1,086,488	0.01%	0	0.00%	1,382,065	0.01%	2,468,553	0.01%
34434	0	0.00%	12,916,043	0.11%	0	0.00%	10,364,080	0.08%	23,280,124	0.07%
33868	0	0.00%	4,729,329	0.04%	0	0.00%	9,167,376	0.07%	13,896,704	0.04%
33444	0	0.00%	22,435,429	0.19%	0	0.00%	4,568,326	0.03%	27,003,938	0.08%
34293	39,655,684	0.56%	0	0.00%	0	0.00%	0	0.00%	39,658,351	0.12%
32312	0	0.00%	40,000,881	0.33%	0	0.00%	0	0.00%	40,000,903	0.12%
33445	0	0.00%	53,023,666	0.44%	0	0.00%	13,516,203	0.10%	66,540,068	0.20%
32738	58,322,976	0.82%	0	0.00%	0	0.00%	46,473,811	0.35%	105,267,468	0.32%
34436	0	0.00%	7,864,201	0.06%	0	0.00%	7,885,971	0.06%	15,750,172	0.05%
33870	31,099,533	0.44%	12,706,909	0.10%	0	0.00%	31,429,044	0.24%	75,235,487	0.23%
33446	0	0.00%	69, 035,647	0.57%	0	0.00%	51,971,285	0.40%	121,007,301	0.37%
32174	92,278,979	1.29%	73,495,476	0.61%	0	0.00%	93,293,594	0.71%	259,068,048	0.78%
33872	29,507,327	0.41%	13,042,590	0.11%	0	0.00%	29,829,465	0.23%	72,379,382	0.22%
33873	18,289,803	0.26%	0	0.00%	0	0.00%	11,016,389	0.08%	29,405,543	0.09%
33449	0	0.00%	21,922,071	0.18%	0	0.00%	17,936,024	0.14%	39,858,095	0.12%
32317	0	0.00%	13,573,734	0.11%	0	0.00%	0	0.00%	13,573,734	0.04%
32034	0	0.00%	0	0.00%	0	0.00%	5,852,096	0.04%	5,852,098	0.02%
32176	57,192,097	0.80%	46,377,903	0.38%	0	0.00%	52,926,157	0.40%	156,496,157	0.47%
32601	0	0.00%	2,811,800	0.02%	0	0.00%	8,134,412	0.06%	10,946,212	0.03%
33592	0	0.00%	5,525,846	0.05%	0	0.00%	5,540,693	0.04%	11,066,539	0.03%
33875	18,563,917	0.26%	9,634,057	0.08%	0	0.00%	22,670,811	0.17%	50,868,784	0.15%
32744	3,616,104	0.05%	0	0.00%	0	0.00%	3,656,751	0.03%	7,318,922	0.02%
34442	0	0.00%	28,427,650	0.23%	0	0.00%	22,646,688	0.17%	51,074,338	0.15%
33876	9,053,161	0.13%	6,976,590	0.06%	0	0.00%	12,849,963	0.10%	28,879,714	0.09%
32603	0	0.00%	0	0.00%	0	0.00%	1,521,835	0.01%	1,675,364	0.01%
33594	0	0.00%	1,068,288	0.01%	0	0.00%	31,597,152	0.24%	32,665,439	0.10%
32179	0	0.00%	4,170,515	0.03%	0	0.00%	5,677,886	0.04%	9,848,401	0.03%
33877	530,622	0.01%	0	0.00%	0	0.00%	0	0.00%	1,134,850	0.00%
32038	0	0.00%	0	0.00%	0	0.00%	3,692,667	0.03%	3,692,667	0.01%
32746	91,487,751	1.28%	46,218,420	0.38%	0	0.00%	70,842,007	0.54%	208,548,178	0.63%
32180	0	0.00%	0	0.00%	0	0.00%	1,697,115	0.01%	1,697,115	0.01%
32605	0	0.00%	1,383,294	0.01%	0	0.00%	20,279,654	0.15%	21,662,948	0.07%
33596	0	0.00%	0	0.00%	0	0.00%	28,000,675	0.21%	28,402,664	0.09%
32606	0	0.00%	2,619,927	0.02%	0	0.00%	17,986,157	0.14%	20,606,084	0.06%
33455	0	0.00%	93,000,649	0.77%	0	0.00%	84,694,704	0.65%	177,695,418	0.54%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33597	0	0.00%	4,696,308	0.04%	0	0.00%	4,709,435	0.04%	9,405,743	0.03%
34446	0	0.00%	25,407,397	0.21%	0	0.00%	17,312,862	0.13%	42,720,259	0.13%
33880	28,970,822	0.41%	1,039,844	0.01%	0	0.00%	38,269,586	0.29%	68,280,252	0.21%
32607	0	0.00%	3,242,173	0.03%	0	0.00%	16,565,258	0.13%	19,807,431	0.06%
33598	0	0.00%	0	0.00%	0	0.00%	5,917,888	0.05%	6,195,174	0.02%
33881	35,100,019	0.49%	17,758,042	0.15%	0	0.00%	43,097,845	0.33%	95,955,906	0.29%
32608	0	0.00%	25,594,744	0.21%	0	0.00%	25,658,658	0.20%	51,253,403	0.16%
34448	0	0.00%	14,670,301	0.12%	0	0.00%	11,099,089	0.08%	25,769,391	0.08%
34731	0	0.00%	12,318,525	0.10%	0	0.00%	16,148,984	0.12%	28,467,509	0.09%
32750	35,595,604	0.50%	21,436,806	0.18%	0	0.00%	35,991,612	0.27%	93,024,022	0.28%
32609	0	0.00%	714,270	0.01%	0	0.00%	6,025,202	0.05%	6,739,472	0.02%
33458	0	0.00%	148,143,861	1.22%	0	0.00%	120,408,732	0.92%	268,552,718	0.81%
34449	0	0.00%	1,956,207	0.02%	0	0.00%	1,470,348	0.01%	3,426,555	0.01%
32751	50,172,173	0.70%	24,998,544	0.21%	0	0.00%	38,641,073	0.29%	113,811,791	0.34%
34450	0	0.00%	14,636,149	0.12%	0	0.00%	14,671,915	0.11%	29,308,063	0.09%
33884	75,622,218	1.06%	29,585,420	0.24%	0	0.00%	66,557,836	0.51%	171,765,475	0.52%
33601	0	0.00%	0	0.00%	0	0.00%	0	0.00%	562,309	0.00%
33460	0	0.00%	35,144,447	0.29%	0	0.00%	21,743,922	0.17%	56,888,444	0.17%
34734	4,113,783	0.06%	0	0.00%	0	0.00%	8,103,998	0.06%	12,280,419	0.04%
33602	0	0.00%	16,030,098	0.13%	0	0.00%	8,112,461	0.06%	24,142,558	0.07%
33461	0	0.00%	37,375,560	0.31%	0	0.00%	19,640,135	0.15%	57,015,904	0.17%
34452	0	0.00%	14,756,689	0.12%	0	0.00%	14,796,552	0.11%	29,553,241	0.09%
32754	15,298,001	0.21%	11,805,965	0.10%	0	0.00%	18,684,482	0.14%	45,788,447	0.14%
33462	0	0.00%	53,778,742	0.44%	0	0.00%	28,032,579	0.21%	81,811,505	0.25%
34736	0	0.00%	10,761,927	0.09%	0	0.00%	17,006,111	0.13%	27,768,038	0.08%
33604	0	0.00%	771,977	0.01%	0	0.00%	768,870	0.01%	1,540,847	0.00%
34453	0	0.00%	15,091,247	0.12%	0	0.00%	11,790,844	0.09%	26,882,091	0.08%
33463	0	0.00%	67,330,761	0.56%	0	0.00%	36,199,023	0.28%	103,529,944	0.31%
32331	0	0.00%	1,736,529	0.01%	0	0.00%	0	0.00%	1,736,529	0.01%
32190	0	0.00%	0	0.00%	0	0.00%	521,804	0.00%	521,863	0.00%
34737	0	0.00%	4,815,614	0.04%	0	0.00%	7,120,487	0.05%	11,936,102	0.04%
33605	0	0.00%	926,268	0.01%	0	0.00%	920,680	0.01%	1,846,948	0.01%
32615	0	0.00%	0	0.00%	0	0.00%	10,349,291	0.08%	10,615,524	0.03%
32757	0	0.00%	1,205,825	0.01%	0	0.00%	31,452,512	0.24%	32,658,355	0.10%
33606	0	0.00%	19,340,417	0.16%	0	0.00%	1,922,007	0.01%	21,262,424	0.06%
33890	10,933,850	0.15%	0	0.00%	0	0.00%	3,723,250	0.03%	14,692,584	0.04%
34739	525,026	0.01%	930,078	0.01%	0	0.00%	1,615,631	0.01%	3,070,735	0.01%
33607	0	0.00%	943,416	0.01%	0	0.00%	938,075	0.01%	1,881,491	0.01%
32617	0	0.00%	2,165,386	0.02%	0	0.00%	2,852,196	0.02%	5,017,582	0.02%
32759	2,700,232	0.04%	2,722,935	0.02%	0	0.00%	3,833,338	0.03%	9,256,505	0.03%
32618	0	0.00%	3,302,200	0.03%	0	0.00%	3,311,535	0.03%	6,613,734	0.02%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33467	0	0.00%	113,655,805	0.94%	0	0.00%	89,899,499	0.69%	203,555,617	0.62%
32052	0	0.00%	1,881,420	0.02%	0	0.00%	0	0.00%	1,913,626	0.01%
32901	0	0.00%	32,262,932	0.27%	0	0.00%	40,648,927	0.31%	72,912,090	0.22%
34741	33,965,278	0.48%	16,811,452	0.14%	0	0.00%	25,336,454	0.19%	76,113,183	0.23%
33609	0	0.00%	3,034,629	0.03%	0	0.00%	3,017,329	0.02%	6,051,959	0.02%
33468	0	0.00%	953,497	0.01%	0	0.00%	751,951	0.01%	1,705,448	0.01%
32336	0	0.00%	589,646	0.00%	0	0.00%	0	0.00%	589,646	0.00%
32053	0	0.00%	1,114,044	0.01%	0	0.00%	0	0.00%	1,133,340	0.00%
32195	0	0.00%	2,304,560	0.02%	0	0.00%	3,080,578	0.02%	5,385,138	0.02%
33610	0	0.00%	0	0.00%	0	0.00%	12,146,373	0.09%	12,563,838	0.04%
33469	0	0.00%	77,389,810	0.64%	0	0.00%	70,921,599	0.54%	148,311,491	0.45%
32054	0	0.00%	0	0.00%	0	0.00%	3,597,870	0.03%	3,597,870	0.01%
32903	0	0.00%	61,585,529	0.51%	0	0.00%	73,120,219	0.56%	134,706,299	0.41%
34601	0	0.00%	12,112,335	0.10%	0	0.00%	12,145,404	0.09%	24,257,739	0.07%
34743	47,183,780	0.66%	20,389,006	0.17%	0	0.00%	38,705,744	0.30%	106,278,531	0.32%
33611	0	0.00%	26,985,137	0.22%	0	0.00%	3,339,204	0.03%	30,324,342	0.09%
32055	0	0.00%	0	0.00%	0	0.00%	6,741,010	0.05%	6,741,010	0.02%
32904	0	0.00%	43,092,338	0.36%	0	0.00%	59,222,727	0.45%	102,315,144	0.31%
34602	0	0.00%	8, 154,153	0.07%	0	0.00%	8,176,941	0.06%	16,331,094	0.05%
32621	0	0.00%	1,642,996	0.01%	0	0.00%	1,406,229	0.01%	3,049,225	0.01%
33470	0	0.00%	52,911,788	0.44%	0	0.00%	44,490,142	0.34%	97,401,930	0.29%
34744	67,848,308	0.95%	31,064,172	0.26%	0	0.00%	55,102,167	0.42%	154,014,648	0.47%
32763	12,731,972	0.18%	739,229	0.01%	0	0.00%	12,868,940	0.10%	26,340,140	0.08%
33612	0	0.00%	1,652,084	0.01%	0	0.00%	16,058,429	0.12%	17,710,513	0.05%
34461	0	0.00%	15,093,216	0.12%	0	0.00%	15,135,301	0.12%	30,228,518	0.09%
32905	0	0.00%	30,365,844	0.25%	0	0.00%	50,564,672	0.39%	80,930,927	0.25%
32622	0	0.00%	0	0.00%	0	0.00%	710,724	0.01%	722,571	0.00%
33471	0	0.00%	4,241,188	0.04%	0	0.00%	4,253,077	0.03%	8,494,265	0.03%
33896	15,206,730	0.21%	2,391,291	0.02%	0	0.00%	15,339,197	0.12%	32,937,217	0.10%
32764	4,912,926	0.07%	2,222,233	0.02%	0	0.00%	3,246,318	0.02%	10,381,477	0.03%
33613	0	0.00%	17,425,720	0.14%	0	0.00%	17,447,864	0.13%	34,873,584	0.11%
32340	0	0.00%	4,290,269	0.04%	0	0.00%	0	0.00%	4,290,269	0.01%
32906	0	0.00%	0	0.00%	0	0.00%	0	0.00%	754,981	0.00%
33755	0	0.00%	15,759,516	0.13%	0	0.00%	2,487,020	0.02%	18,246,536	0.06%
34604	0	0.00%	9,004,603	0.07%	0	0.00%	7,019,590	0.05%	16,024,193	0.05%
33472	0	0.00%	37,083,457	0.31%	0	0.00%	19,288,500	0.15%	56,371,958	0.17%
33897	807,541	0.01%	813,407	0.01%	0	0.00%	25,145,586	0.19%	26,766,534	0.08%
34746	72,224,098	1.01%	33,825,944	0.28%	0	0.00%	57,562,956	0.44%	163,612,998	0.50%
32765	145,206,990	2.04%	51,692,010	0.43%	0	0.00%	106,345,699	0.81%	303,244,699	0.92%
33614	0	0.00%	2,793,231	0.02%	0	0.00%	2,776,187	0.02%	5,569,419	0.02%
32907	0	0.00%	51,035,503	0.42%	0	0.00%	92,853,188	0.71%	143,888,691	0.44%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33756	0	0.00%	28,183,185	0.23%	0	0.00%	6,573,370	0.05%	34,756,555	0.11%
33473	0	0.00%	10,154,574	0.08%	0	0.00%	8,637,657	0.07%	18,792,231	0.06%
33898	43,984,467	0.62%	10,274,520	0.08%	0	0.00%	25,711,536	0.20%	79,970,523	0.24%
34747	47,683,730	0.67%	38,817,180	0.32%	0	0.00%	57,084,989	0.44%	143,585,899	0.43%
32766	38,082,194	0.53%	17,467,318	0.14%	0	0.00%	28,453,413	0.22%	84,002,925	0.25%
33615	0	0.00%	25,715,646	0.21%	0	0.00%	3,234,724	0.02%	28,950,370	0.09%
32059	0	0.00%	928,217	0.01%	0	0.00%	0	0.00%	928,217	0.00%
32908	0	0.00%	11,815,691	0.10%	0	0.00%	19,397,762	0.15%	31,213,452	0.09%
34606	0	0.00%	37,367,171	0.31%	0	0.00%	23,598,857	0.18%	60,966,028	0.18%
32625	0	0.00%	2,895,969	0.02%	0	0.00%	1,507,705	0.01%	4,403,674	0.01%
34748	0	0.00%	31,731,505	0.26%	0	0.00%	45,240,167	0.35%	76,971,671	0.23%
32767	0	0.00%	0	0.00%	0	0.00%	1,007,015	0.01%	1,022,371	0.00%
33616	0	0.00%	7,913,316	0.07%	0	0.00%	1,258,344	0.01%	9,171,660	0.03%
34465	0	0.00%	27,038,252	0.22%	0	0.00%	18,078,160	0.14%	45,116,412	0.14%
32060	0	0.00%	7,781,057	0.06%	0	0.00%	0	0.00%	7,930,725	0.02%
32909	0	0.00%	43,233,755	0.36%	0	0.00%	63,582,673	0.49%	106,816,433	0.32%
34607	0	0.00%	13,934,869	0.11%	0	0.00%	8,870,558	0.07%	22,805,427	0.07%
32626	0	0.00%	4,083,085	0.03%	0	0.00%	0	0.00%	4,143,556	0.01%
33617	0	0.00%	3,048,055	0.03%	0	0.00%	22,262,900	0.17%	25,310,955	0.08%
32344	0	0.00%	7,990,621	0.07%	0	0.00%	0	0.00%	7,990,621	0.02%
33759	0	0.00%	11,851,898	0.10%	0	0.00%	0	0.00%	11,852,310	0.04%
34608	0	0.00%	42,079,275	0.35%	0	0.00%	27,221,610	0.21%	69,300,886	0.21%
33476	0	0.00%	4,195,675	0.03%	0	0.00%	3,381,765	0.03%	7,577,442	0.02%
33901	29,514,284	0.41%	0	0.00%	0	0.00%	0	0.00%	29,514,380	0.09%
33618	0	0.00%	25,730,357	0.21%	0	0.00%	25,771,088	0.20%	51,501,444	0.16%
33760	0	0.00%	8,997,747	0.07%	0	0.00%	0	0.00%	8,998,215	0.03%
34609	0	0.00%	52,447,194	0.43%	0	0.00%	36,795,124	0.28%	89,242,318	0.27%
32628	0	0.00%	1,379,721	0.01%	0	0.00%	0	0.00%	1,379,721	0.00%
33477	0	0.00%	102,923,436	0.85%	0	0.00%	93,350,647	0.71%	196,274,432	0.59%
33619	0	0.00%	530,152	0.00%	0	0.00%	10,928,515	0.08%	11,458,667	0.03%
32346	0	0.00%	2,311,006	0.02%	0	0.00%	0	0.00%	2,311,006	0.01%
33761	0	0.00%	20,591,964	0.17%	0	0.00%	5,688,253	0.04%	26,280,217	0.08%
34610	0	0.00%	9,585,840	0.08%	0	0.00%	6,682,113	0.05%	16,267,953	0.05%
33478	0	0.00%	46,506,861	0.38%	0	0.00%	43,062,606	0.33%	89,569,468	0.27%
33903	60,571,722	0.85%	0	0.00%	0	0.00%	0	0.00%	60,571,857	0.18%
32771	64,827,709	0.91%	37,494,520	0.31%	0	0.00%	52,169,739	0.40%	154,491,968	0.47%
33762	0	0.00%	8,353,814	0.07%	0	0.00%	0	0.00%	8,354,020	0.03%
32347	0	0.00%	4,182,725	0.03%	0	0.00%	0	0.00%	4,182,725	0.01%
32064	0	0.00%	2,013,474	0.02%	0	0.00%	0	0.00%	2,046,787	0.01%
34470	0	0.00%	10,593,730	0.09%	0	0.00%	10,617,482	0.08%	21,211,212	0.06%
34753	0	0.00%	2,636,091	0.02%	0	0.00%	4,159,498	0.03%	6,795,589	0.02%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33904	111,628,718	1.57%	0	0.00%	0	0.00%	0	0.00%	111,628,943	0.34%
33480	0	0.00%	232,970,410	1.92%	0	0.00%	157,495,941	1.20%	390,467,140	1.18%
32631	0	0.00%	0	0.00%	0	0.00%	537,862	0.00%	554,757	0.00%
33763	0	0.00%	1,436,749	0.01%	0	0.00%	1,427,958	0.01%	2,864,707	0.01%
32348	0	0.00%	3,879,746	0.03%	0	0.00%	0	0.00%	3,879,746	0.01%
34471	0	0.00%	21,808,854	0.18%	0	0.00%	28,309,310	0.22%	50,118,164	0.15%
32773	36,274,206	0.51%	15,252,798	0.13%	0	0.00%	23,247,449	0.18%	74,774,453	0.23%
33905	32,618,626	0.46%	0	0.00%	0	0.00%	0	0.00%	32,618,650	0.10%
34613	0	0.00%	23,095,659	0.19%	0	0.00%	15,310,381	0.12%	38,406,040	0.12%
33764	0	0.00%	21,239,159	0.18%	0	0.00%	0	0.00%	21,239,905	0.06%
32066	0	0.00%	1,892,795	0.02%	0	0.00%	0	0.00%	1,892,795	0.01%
34472	0	0.00%	18,189,752	0.15%	0	0.00%	23,110,690	0.18%	41,300,443	0.13%
34614	0	0.00%	6,181,666	0.05%	0	0.00%	5,053,712	0.04%	11,235,378	0.03%
33765	0	0.00%	7,908,885	0.07%	0	0.00%	0	0.00%	7,909,051	0.02%
32350	0	0.00%	565,247	0.00%	0	0.00%	0	0.00%	565,247	0.00%
34473	0	0.00%	16,214,313	0.13%	0	0.00%	16,259,154	0.12%	32,473,467	0.10%
33624	0	0.00%	30,590,715	0.25%	0	0.00%	30,665,138	0.23%	61,255,853	0.19%
34756	0	0.00%	3,762,776	0.03%	0	0.00%	6,360,734	0.05%	10,123,509	0.03%
33907	28,505,215	0.40%	0	0.00%	0	0.00%	0	0.00%	28,505,469	0.09%
33483	0	0.00%	44,336,394	0.37%	0	0.00%	13,178,458	0.10%	57,515,391	0.17%
32776	0	0.00%	9,569,411	0.08%	0	0.00%	12,788,450	0.10%	22,357,861	0.07%
33908	120,046,272	1.68%	0	0.00%	0	0.00%	0	0.00%	120,047,016	0.36%
33625	0	0.00%	16,814,549	0.14%	0	0.00%	16,857,441	0.13%	33,671,990	0.10%
34474	0	0.00%	10,560,267	0.09%	0	0.00%	10,578,409	0.08%	21,138,677	0.06%
33767	0	0.00%	45,208,740	0.37%	0	0.00%	26,517,202	0.20%	71,725,941	0.22%
33484	0	0.00%	54,046,730	0.45%	0	0.00%	16,851,550	0.13%	70,898,587	0.21%
33060	0	0.00%	4,880,873	0.04%	0	0.00%	0	0.00%	4,881,368	0.01%
33909	42,786,055	0.60%	0	0.00%	0	0.00%	0	0.00%	42,786,116	0.13%
34758	49,875,286	0.70%	23,079,884	0.19%	0	0.00%	40,477,818	0.31%	113,432,988	0.34%
33626	0	0.00%	29,325,090	0.24%	0	0.00%	1,266,887	0.01%	30,591,978	0.09%
34475	0	0.00%	3,682,291	0.03%	0	0.00%	4,700,180	0.04%	8,382,471	0.03%
34759	56,218,027	0.79%	26,854,134	0.22%	0	0.00%	44,257,449	0.34%	127,329,610	0.39%
32778	0	0.00%	15,851,687	0.13%	0	0.00%	22,921,776	0.17%	38,773,478	0.12%
34476	0	0.00%	28,569,105	0.24%	0	0.00%	28,648,140	0.22%	57,217,244	0.17%
32920	16,994,084	0.24%	28,345,097	0.23%	0	0.00%	34,622,020	0.26%	79,961,201	0.24%
33486	0	0.00%	24,082,024	0.20%	0	0.00%	0	0.00%	24,082,394	0.07%
33062	0	0.00%	33,420,318	0.28%	0	0.00%	0	0.00%	33,425,140	0.10%
34760	0	0.00%	0	0.00%	0	0.00%	1,459,223	0.01%	1,471,578	0.00%
32779	76,382,910	1.07%	47,822,877	0.39%	0	0.00%	77,219,805	0.59%	201,425,592	0.61%
33487	0	0.00%	61,106,237	0.50%	0	0.00%	20,120,289	0.15%	81,226,903	0.25%
33770	0	0.00%	18,738,247	0.15%	0	0.00%	2,753,783	0.02%	21,492,030	0.07%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33629	0	0.00%	37,708,912	0.31%	0	0.00%	2,666,567	0.02%	40,375,479	0.12%
32780	45,220,398	0.63%	45,595,729	0.38%	0	0.00%	72,159,051	0.55%	162,975,179	0.49%
34761	30,461,002	0.43%	1,639,320	0.01%	0	0.00%	54,062,437	0.41%	86,162,760	0.26%
33912	50,477,983	0.71%	0	0.00%	0	0.00%	0	0.00%	50,478,062	0.15%
33771	0	0.00%	13,924,259	0.11%	0	0.00%	0	0.00%	13,925,961	0.04%
32922	6,808,238	0.10%	14,040,241	0.12%	0	0.00%	19,290,141	0.15%	40,138,620	0.12%
34479	0	0.00%	7,864,291	0.06%	0	0.00%	10,323,678	0.08%	18,187,969	0.06%
33913	34,949,325	0.49%	0	0.00%	0	0.00%	0	0.00%	34,949,523	0.11%
33064	0	0.00%	8,517,008	0.07%	0	0.00%	0	0.00%	8,518,939	0.03%
34762	0	0.00%	0	0.00%	0	0.00%	633,124	0.00%	1,057,468	0.00%
32640	0	0.00%	0	0.00%	0	0.00%	4,810,634	0.04%	4,900,303	0.01%
33772	0	0.00%	33,584,127	0.28%	0	0.00%	0	0.00%	33,584,774	0.10%
34480	0	0.00%	15,963,082	0.13%	0	0.00%	20,189,880	0.15%	36,152,963	0.11%
33914	127,046,757	1.78%	0	0.00%	0	0.00%	0	0.00%	127,047,073	0.38%
32641	0	0.00%	0	0.00%	0	0.00%	4,059,654	0.03%	4,126,145	0.01%
33773	0	0.00%	11,937,307	0.10%	0	0.00%	0	0.00%	11,937,380	0.04%
34481	0	0.00%	17,099,230	0.14%	0	0.00%	17,143,700	0.13%	34,242,930	0.10%
32359	0	0.00%	1,889,979	0.02%	0	0.00%	0	0.00%	1,889,979	0.01%
33774	0	0.00%	29,493,279	0.24%	0	0.00%	4,724,073	0.04%	34,217,352	0.10%
32501	0	0.00%	0	0.00%	10,106,706	1.51%	0	0.00%	10,106,706	0.03%
34482	0	0.00%	18,888,165	0.16%	0	0.00%	18,939,113	0.14%	37,827,278	0.11%
33916	15,302,329	0.21%	0	0.00%	0	0.00%	0	0.00%	15,303,483	0.05%
33067	0	0.00%	2,654,718	0.02%	0	0.00%	0	0.00%	2,655,441	0.01%
32784	0	0.00%	6,687,916	0.06%	0	0.00%	9,981,799	0.08%	16,669,715	0.05%
32643	0	0.00%	0	0.00%	0	0.00%	6,542,882	0.05%	6,692,494	0.02%
32926	15,031,809	0.21%	24,322,079	0.20%	0	0.00%	39,299,930	0.30%	78,653,818	0.24%
33634	0	0.00%	10,849,277	0.09%	0	0.00%	762,870	0.01%	11,612,147	0.04%
33917	54,462,156	0.76%	0	0.00%	0	0.00%	0	0.00%	54,462,585	0.16%
32502	0	0.00%	0	0.00%	2,677,989	0.40%	0	0.00%	2,677,989	0.01%
33493	0	0.00%	1,093,977	0.01%	0	0.00%	1,096,382	0.01%	2,190,359	0.01%
33776	0	0.00%	33,965,829	0.28%	0	0.00%	955,956	0.01%	34,921,785	0.11%
32927	29,880,647	0.42%	30,131,483	0.25%	0	0.00%	48,105,506	0.37%	108,117,637	0.33%
32503	0	0.00%	0	0.00%	40,043,845	5.97%	0	0.00%	40,043,845	0.12%
34484	0	0.00%	4,043,075	0.03%	0	0.00%	4,051,215	0.03%	8,094,290	0.02%
33635	0	0.00%	8,777,219	0.07%	0	0.00%	551,900	0.00%	9,329,119	0.03%
33069	0	0.00%	17,691,106	0.15%	0	0.00%	0	0.00%	17,693,193	0.05%
33777	0	0.00%	27,954,638	0.23%	0	0.00%	0	0.00%	27,955,644	0.08%
34202	0	0.00%	0	0.00%	0	0.00%	3,231,421	0.02%	3,231,684	0.01%
32504	0	0.00%	0	0.00%	30,507,242	4.55%	0	0.00%	30,507,242	0.09%
33919	85,998,225	1.21%	0	0.00%	0	0.00%	0	0.00%	85,999,230	0.26%
33778	0	0.00%	16,621,887	0.14%	0	0.00%	0	0.00%	16,622,194	0.05%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32080	0	0.00%	37,162,200	0.31%	0	0.00%	37,243,530	0.28%	74,405,885	0.23%
33920	7,127,959	0.10%	0	0.00%	0	0.00%	0	0.00%	7,399,154	0.02%
32505	0	0.00%	0	0.00%	16,660,350	2.48%	0	0.00%	16,660,350	0.05%
33637	0	0.00%	1,312,950	0.01%	0	0.00%	6,926,380	0.05%	8,239,330	0.02%
34769	30,904,837	0.43%	13,397,684	0.11%	0	0.00%	24,582,291	0.19%	68,884,812	0.21%
33496	0	0.00%	58,529,023	0.48%	0	0.00%	7,733,756	0.06%	66,262,779	0.20%
32506	0	0.00%	0	0.00%	36,220,861	5.40%	0	0.00%	36,220,861	0.11%
32789	100,879,387	1.42%	45,396,037	0.37%	0	0.00%	84,941,194	0.65%	231,216,619	0.70%
33921	53,522,993	0.75%	18,438,757	0.15%	0	0.00%	18,464,108	0.14%	90,425,858	0.27%
32648	0	0.00%	596,075	0.00%	0	0.00%	0	0.00%	596,075	0.00%
32931	24,232,830	0.34%	71,804,374	0.59%	0	0.00%	86,650,834	0.66%	182,688,037	0.55%
32082	0	0.00%	67,536,540	0.56%	0	0.00%	67,706,972	0.52%	135,243,512	0.41%
34205	0	0.00%	4,242,362	0.04%	0	0.00%	0	0.00%	4,242,869	0.01%
32507	0	0.00%	0	0.00%	72,528,451	10.82%	0	0.00%	72,528,451	0.22%
34488	0	0.00%	0	0.00%	0	0.00%	4,175,313	0.03%	4,273,258	0.01%
34771	25,728,721	0.36%	12,795,211	0.11%	0	0.00%	22,622,598	0.17%	61,146,530	0.19%
33922	22,530,254	0.32%	0	0.00%	0	0.00%	0	0.00%	22,530,254	0.07%
33073	0	0.00%	4,001,319	0.03%	0	0.00%	0	0.00%	4,001,442	0.01%
33498	0	0.00%	22,352,944	0.18%	0	0.00%	0	0.00%	22,773,332	0.07%
33781	0	0.00%	12,383,419	0.10%	0	0.00%	0	0.00%	12,383,920	0.04%
32932	0	0.00%	0	0.00%	0	0.00%	0	0.00%	620,489	0.00%
34772	38,695,734	0.54%	18,224,302	0.15%	0	0.00%	31,289,272	0.24%	88,209,308	0.27%
33782	0	0.00%	14,938,356	0.12%	0	0.00%	0	0.00%	14,938,823	0.05%
32084	0	0.00%	21,760,039	0.18%	0	0.00%	21,814,573	0.17%	43,574,612	0.13%
34207	0	0.00%	14,304,741	0.12%	0	0.00%	0	0.00%	14,304,992	0.04%
32792	72,765,175	1.02%	29,546,962	0.24%	0	0.00%	60,955,938	0.47%	163,268,075	0.49%
34773	2,557,931	0.04%	1,918,963	0.02%	0	0.00%	2,999,674	0.02%	7,476,568	0.02%
33924	56,821,192	0.80%	10,147,323	0.08%	0	0.00%	8,051,061	0.06%	75,019,577	0.23%
32934	0	0.00%	37,577,442	0.31%	0	0.00%	52,059,520	0.40%	90,120,700	0.27%
34491	0	0.00%	24,186,168	0.20%	0	0.00%	30,944,495	0.24%	55,130,663	0.17%
33076	0	0.00%	1,989,134	0.02%	0	0.00%	0	0.00%	1,989,412	0.01%
34208	0	0.00%	0	0.00%	0	0.00%	582,260	0.00%	582,260	0.00%
32935	0	0.00%	71,450,011	0.59%	0	0.00%	88,980,207	0.68%	160,430,695	0.49%
32086	0	0.00%	1,762,481	0.01%	0	0.00%	22,199,258	0.17%	23,961,740	0.07%
34209	0	0.00%	8,005,633	0.07%	0	0.00%	0	0.00%	8,006,467	0.02%
32653	0	0.00%	0	0.00%	0	0.00%	12,151,501	0.09%	12,633,950	0.04%
33785	0	0.00%	30,382,994	0.25%	0	0.00%	20,208,998	0.15%	50,591,992	0.15%
34210	0	0.00%	18,195,226	0.15%	0	0.00%	0	0.00%	18,196,112	0.06%
33786	0	0.00%	9,671,178	0.08%	0	0.00%	7,301,537	0.06%	16,972,714	0.05%
32937	6,626,497	0.09%	100,381,113	0.83%	0	0.00%	107,902,061	0.82%	214,909,671	0.65%
32796	24,611,851	0.35%	24,816,135	0.20%	0	0.00%	33,151,681	0.25%	82,579,667	0.25%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33928	39,437,576	0.55%	0	0.00%	0	0.00%	0	0.00%	39,438,086	0.12%
34212	0	0.00%	0	0.00%	0	0.00%	1,357,509	0.01%	1,357,580	0.00%
32514	0	0.00%	0	0.00%	36,181,612	5.40%	0	0.00%	36,181,612	0.11%
34637	0	0.00%	6,869,114	0.06%	0	0.00%	6,888,080	0.05%	13,757,194	0.04%
32656	0	0.00%	0	0.00%	0	0.00%	7,458,040	0.06%	7,458,040	0.02%
33647	0	0.00%	58,396,972	0.48%	0	0.00%	80,101,927	0.61%	138,498,899	0.42%
32798	0	0.00%	0	0.00%	0	0.00%	3,566,887	0.03%	3,597,963	0.01%
32940	44,016,431	0.62%	94,378,813	0.78%	0	0.00%	123,697,498	0.94%	262,092,742	0.79%
34638	0	0.00%	20,192,719	0.17%	0	0.00%	20,249,774	0.15%	40,442,494	0.12%
33931	54,876,836	0.77%	11,444,799	0.09%	0	0.00%	0	0.00%	66,321,746	0.20%
32233	0	0.00%	1,705,702	0.01%	0	0.00%	17,324,751	0.13%	19,030,453	0.06%
34639	0	0.00%	23,667,499	0.20%	0	0.00%	23,732,091	0.18%	47,399,589	0.14%
34215	0	0.00%	2,083,153	0.02%	0	0.00%	1,599,914	0.01%	3,683,079	0.01%
34498	0	0.00%	691,724	0.01%	0	0.00%	578,651	0.00%	1,270,375	0.00%
34216	0	0.00%	7,411,327	0.06%	0	0.00%	0	0.00%	7,514,562	0.02%
32801	23,454,274	0.33%	12,510,329	0.10%	0	0.00%	20,055,376	0.15%	56,019,979	0.17%
34217	0	0.00%	33,447,336	0.28%	0	0.00%	4,216,076	0.03%	37,663,646	0.11%
32095	0	0.00%	546,339	0.00%	0	0.00%	8,356,801	0.06%	8,903,140	0.03%
33510	0	0.00%	1,266,894	0.01%	0	0.00%	18,682,079	0.14%	19,948,973	0.06%
34218	0	0.00%	0	0.00%	0	0.00%	0	0.00%	516,783	0.00%
32803	46,075,547	0.65%	18,538,598	0.15%	0	0.00%	38,494,952	0.29%	103,109,097	0.31%
33935	0	0.00%	0	0.00%	0	0.00%	6,902,188	0.05%	7,007,790	0.02%
33511	0	0.00%	1,595,713	0.01%	0	0.00%	34,082,295	0.26%	35,678,008	0.11%
34785	0	0.00%	8,876,471	0.07%	0	0.00%	11,723,720	0.09%	20,600,191	0.06%
33936	14,855,026	0.21%	0	0.00%	0	0.00%	0	0.00%	14,855,132	0.05%
34219	0	0.00%	0	0.00%	0	0.00%	21,027,297	0.16%	21,395,787	0.06%
32804	52,454,853	0.74%	21,739,615	0.18%	0	0.00%	43,957,703	0.34%	118,152,171	0.36%
34786	69,395,245	0.97%	69,982,263	0.58%	0	0.00%	108,787,149	0.83%	248,164,658	0.75%
32805	15,331,922	0.22%	7,580,866	0.06%	0	0.00%	15,490,217	0.12%	38,403,005	0.12%
33513	0	0.00%	7,515,657	0.06%	0	0.00%	7,536,670	0.06%	15,052,326	0.05%
32664	0	0.00%	0	0.00%	0	0.00%	0	0.00%	984,524	0.00%
34787	2,190,710	0.03%	2,206,482	0.02%	0	0.00%	65,487,690	0.50%	69,884,881	0.21%
34221	0	0.00%	2,554,566	0.02%	0	0.00%	2,540,597	0.02%	5,095,175	0.02%
32806	70,928,203	1.00%	24,656,140	0.20%	0	0.00%	51,978,102	0.40%	147,562,445	0.45%
33514	0	0.00%	1,023,030	0.01%	0	0.00%	1,333,646	0.01%	2,356,676	0.01%
32948	0	0.00%	3,576,719	0.03%	0	0.00%	5,361,448	0.04%	8,938,167	0.03%
34788	0	0.00%	11,781,606	0.10%	0	0.00%	18,572,754	0.14%	30,354,360	0.09%
32807	40,312,857	0.57%	13,779,477	0.11%	0	0.00%	29,524,351	0.23%	83,616,685	0.25%
32949	0	0.00%	7,199,841	0.06%	0	0.00%	8,668,205	0.07%	15,868,046	0.05%
32666	0	0.00%	0	0.00%	0	0.00%	4,172,657	0.03%	4,172,657	0.01%
34223	45,286,908	0.64%	4,229,487	0.03%	0	0.00%	0	0.00%	49,516,788	0.15%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32808	28,054,966	0.39%	1,297,793	0.01%	0	0.00%	37,755,957	0.29%	67,108,716	0.20%
32667	0	0.00%	2,990,049	0.02%	0	0.00%	2,998,503	0.02%	5,988,552	0.02%
32950	0	0.00%	13,175,540	0.11%	0	0.00%	16,287,075	0.12%	29,462,615	0.09%
34224	38,510,836	0.54%	0	0.00%	0	0.00%	1,847,830	0.01%	40,358,740	0.12%
32526	0	0.00%	0	0.00%	42,821,795	6.39%	0	0.00%	42,821,795	0.13%
32809	36,710,861	0.52%	13,231,368	0.11%	0	0.00%	26,968,679	0.21%	76,910,908	0.23%
32951	0	0.00%	50,377,621	0.42%	0	0.00%	98,890,204	0.75%	149,268,252	0.45%
32668	0	0.00%	3,603,070	0.03%	0	0.00%	3,137,258	0.02%	6,740,328	0.02%
32102	0	0.00%	0	0.00%	0	0.00%	1,748,186	0.01%	1,748,186	0.01%
32810	32,902,276	0.46%	799,308	0.01%	0	0.00%	33,262,201	0.25%	66,963,785	0.20%
33801	0	0.00%	1,729,314	0.01%	0	0.00%	29,564,499	0.23%	31,293,900	0.09%
32952	22,257,277	0.31%	65,842,184	0.54%	0	0.00%	76,169,949	0.58%	164,269,411	0.50%
32669	0	0.00%	9,574,498	0.08%	0	0.00%	9,599,931	0.07%	19,174,429	0.06%
32811	17,236,875	0.24%	9,260,376	0.08%	0	0.00%	17,355,991	0.13%	43,853,242	0.13%
32953	21,410,388	0.30%	44,937,855	0.37%	0	0.00%	62,050,369	0.47%	128,398,612	0.39%
32812	65,732,446	0.92%	23,190,833	0.19%	0	0.00%	48,243,551	0.37%	137,166,830	0.42%
34652	0	0.00%	19,108,372	0.16%	0	0.00%	28,514,410	0.22%	47,622,782	0.14%
33803	0	0.00%	1,881,277	0.02%	0	0.00%	33,997,254	0.26%	35,878,568	0.11%
34228	0	0.00%	86,712,084	0.72%	0	0.00%	24,982,826	0.19%	111,696,328	0.34%
34653	0	0.00%	18,603,284	0.15%	0	0.00%	18,637,652	0.14%	37,240,937	0.11%
32955	35,524,214	0.50%	65,546,783	0.54%	0	0.00%	89,551,887	0.68%	190,622,883	0.58%
33521	0	0.00%	0	0.00%	0	0.00%	521,444	0.00%	911,979	0.00%
33946	17,276,192	0.24%	8,514,330	0.07%	0	0.00%	8,514,095	0.07%	34,304,617	0.10%
34229	0	0.00%	16,429,864	0.14%	0	0.00%	0	0.00%	16,430,720	0.05%
32814	13,875,681	0.19%	8,663,753	0.07%	0	0.00%	10,665,058	0.08%	33,204,492	0.10%
32531	0	0.00%	0	0.00%	3,033,057	0.45%	0	0.00%	3,033,057	0.01%
34654	0	0.00%	25,030,595	0.21%	0	0.00%	17,730,262	0.14%	42,760,857	0.13%
33805	0	0.00%	0	0.00%	0	0.00%	16,565,088	0.13%	17,028,514	0.05%
33947	27,721,280	0.39%	0	0.00%	0	0.00%	939,236	0.01%	28,660,578	0.09%
34655	0	0.00%	41,701,240	0.34%	0	0.00%	41,794,069	0.32%	83,495,309	0.25%
33523	0	0.00%	12,947,710	0.11%	0	0.00%	12,983,715	0.10%	25,931,425	0.08%
32816	6,530,656	0.09%	3,470,861	0.03%	0	0.00%	4,956,356	0.04%	14,957,872	0.05%
33948	56,479,432	0.79%	0	0.00%	0	0.00%	0	0.00%	56,479,605	0.17%
32250	0	0.00%	5,488,818	0.05%	0	0.00%	27,541,590	0.21%	33,030,408	0.10%
34231	0	0.00%	36,037,875	0.30%	0	0.00%	0	0.00%	36,039,413	0.11%
32533	0	0.00%	0	0.00%	32,719,605	4.88%	0	0.00%	32,719,605	0.10%
34797	0	0.00%	1,762,041	0.01%	0	0.00%	3,934,745	0.03%	5,696,785	0.02%
32958	0	0.00%	63,945,715	0.53%	0	0.00%	80,720,112	0.62%	144,665,884	0.44%
32817	60,272,782	0.85%	21,292,912	0.18%	0	0.00%	44,223,679	0.34%	125,789,373	0.38%
32534	0	0.00%	0	0.00%	12,251,090	1.83%	0	0.00%	12,251,090	0.04%
33525	0	0.00%	12,701,449	0.10%	0	0.00%	17,954,069	0.14%	30,655,518	0.09%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32818	37,101,206	0.52%	1,274,731	0.01%	0	0.00%	49,317,664	0.38%	87,693,601	0.27%
33950	143,527,509	2.01%	0	0.00%	0	0.00%	6,637,673	0.05%	150,165,613	0.45%
32535	0	0.00%	0	0.00%	2,438,559	0.36%	0	0.00%	2,438,559	0.01%
33809	0	0.00%	19,316,699	0.16%	0	0.00%	38,935,494	0.30%	58,252,199	0.18%
32960	0	0.00%	47,122,747	0.39%	0	0.00%	55,758,341	0.43%	102,881,193	0.31%
32819	67,001,565	0.94%	34,153,275	0.28%	0	0.00%	67,696,946	0.52%	168,851,786	0.51%
34234	0	0.00%	1,358,081	0.01%	0	0.00%	0	0.00%	1,358,508	0.00%
33810	0	0.00%	28,206,735	0.23%	0	0.00%	51,261,116	0.39%	79,467,852	0.24%
32112	0	0.00%	0	0.00%	0	0.00%	2,732,036	0.02%	2,795,020	0.01%
33527	0	0.00%	8,260,573	0.07%	0	0.00%	11,878,417	0.09%	20,138,990	0.06%
33952	91,058,592	1.28%	0	0.00%	0	0.00%	0	0.00%	91,058,817	0.28%
32820	12,577,864	0.18%	6,886,589	0.06%	0	0.00%	9,460,677	0.07%	28,925,129	0.09%
33811	0	0.00%	537,497	0.00%	0	0.00%	29,322,706	0.22%	29,860,203	0.09%
32962	0	0.00%	53,691,887	0.44%	0	0.00%	62,279,688	0.48%	115,971,799	0.35%
32113	0	0.00%	2,406,330	0.02%	0	0.00%	3,304,180	0.03%	5,710,510	0.02%
32821	29,588,931	0.42%	2,791,694	0.02%	0	0.00%	21,941,997	0.17%	54,322,622	0.16%
33953	16,273,106	0.23%	0	0.00%	0	0.00%	0	0.00%	16,273,268	0.05%
34236	0	0.00%	31,729,112	0.26%	0	0.00%	0	0.00%	31,732,689	0.10%
33812	0	0.00%	0	0.00%	0	0.00%	18,842,601	0.14%	19,283,061	0.06%
32963	0	0.00%	175,052,877	1.44%	0	0.00%	379,162,362	2.89%	554,215,874	1.68%
32114	27,427,215	0.38%	20,329,839	0.17%	0	0.00%	24,308,306	0.19%	72,065,360	0.22%
32680	0	0.00%	2,960,582	0.02%	0	0.00%	0	0.00%	3,004,188	0.01%
33954	31,003,808	0.43%	0	0.00%	0	0.00%	0	0.00%	31,134,749	0.09%
32822	56,563,625	0.79%	20,971,489	0.17%	0	0.00%	41,298,057	0.32%	118,833,172	0.36%
33813	0	0.00%	1,013,460	0.01%	0	0.00%	72,438,617	0.55%	73,452,132	0.22%
34945	0	0.00%	8,467,071	0.07%	0	0.00%	10,694,018	0.08%	19,161,089	0.06%
33955	62,573,796	0.88%	0	0.00%	0	0.00%	0	0.00%	62,574,161	0.19%
34238	0	0.00%	5,267,485	0.04%	0	0.00%	0	0.00%	5,268,734	0.02%
34946	0	0.00%	9,045,058	0.07%	0	0.00%	11,384,103	0.09%	20,429,160	0.06%
32824	65,157,586	0.91%	26,212,329	0.22%	0	0.00%	53,269,050	0.41%	144,638,965	0.44%
33956	29,721,832	0.42%	0	0.00%	0	0.00%	0	0.00%	29,830,875	0.09%
34239	0	0.00%	1,309,839	0.01%	0	0.00%	0	0.00%	1,309,959	0.00%
32541	0	0.00%	0	0.00%	31,638,957	4.72%	0	0.00%	31,638,957	0.10%
33815	0	0.00%	0	0.00%	0	0.00%	5,497,420	0.04%	5,627,401	0.02%
32966	0	0.00%	32,126,694	0.27%	0	0.00%	46,158,857	0.35%	78,285,611	0.24%
32117	25,474,634	0.36%	16,646,515	0.14%	0	0.00%	21,457,179	0.16%	63,578,328	0.19%
34947	0	0.00%	10,067,625	0.08%	0	0.00%	11,293,855	0.09%	21,361,508	0.06%
32825	95,085,986	1.33%	36,379,106	0.30%	0	0.00%	70,160,048	0.54%	201,625,140	0.61%
33957	131,939,103	1.85%	9,557,454	0.08%	0	0.00%	0	0.00%	141,496,799	0.43%
32967	0	0.00%	39,075,279	0.32%	0	0.00%	49,457,754	0.38%	88,533,187	0.27%
32118	72,249,777	1.01%	56,488,682	0.47%	0	0.00%	62,171,905	0.47%	190,910,364	0.58%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32826	33,726,950	0.47%	12,650,907	0.10%	0	0.00%	18,922,772	0.14%	65,300,629	0.20%
32968	0	0.00%	32,418,708	0.27%	0	0.00%	35,295,480	0.27%	67,714,188	0.21%
32119	41,619,078	0.58%	30,449,812	0.25%	0	0.00%	36,675,073	0.28%	108,743,963	0.33%
34949	0	0.00%	57,553,606	0.47%	0	0.00%	76,639,580	0.58%	134,194,256	0.41%
33534	0	0.00%	0	0.00%	0	0.00%	5,673,744	0.04%	5,805,003	0.02%
32827	21,728,510	0.30%	10,269,227	0.08%	0	0.00%	17,826,216	0.14%	49,823,952	0.15%
34242	0	0.00%	45,991,294	0.38%	0	0.00%	0	0.00%	45,993,989	0.14%
Total	$7,121,653,525		$12,105,192,828		$668,527,112		$13,087,892,511		$33,003,175,376	

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34950	0	0.00%	19,064,152	0.16%	0	0.00%	22,547,054	0.16%	41,611,260	0.12%
34667	0	0.00%	39,675,306	0.33%	0	0.00%	54,143,696	0.38%	93,819,002	0.27%
32686	0	0.00%	4,384,981	0.04%	0	0.00%	4,401,881	0.03%	8,786,862	0.03%
33960	0	0.00%	681,113	0.01%	0	0.00%	1,124,336	0.01%	2,218,684	0.01%
32828	106,055,321	1.29%	45,805,000	0.38%	0	0.00%	78,858,025	0.56%	230,718,346	0.66%
34102	0	0.00%	41,436,616	0.34%	0	0.00%	0	0.00%	41,436,804	0.12%
34951	0	0.00%	42,978,538	0.35%	0	0.00%	58,912,246	0.42%	101,890,794	0.29%
32970	0	0.00%	0	0.00%	0	0.00%	0	0.00%	708,620	0.00%
34668	0	0.00%	26,464,503	0.22%	0	0.00%	46,069,374	0.33%	72,533,877	0.21%
32829	34,025,547	0.41%	14,416,162	0.12%	0	0.00%	23,461,791	0.17%	71,903,499	0.20%
34103	0	0.00%	46,188,156	0.38%	0	0.00%	24,594,249	0.17%	70,783,496	0.20%
34952	0	0.00%	95,003,835	0.78%	0	0.00%	107,245,898	0.76%	202,249,875	0.58%
33820	608,226	0.01%	0	0.00%	0	0.00%	0	0.00%	1,153,882	0.00%
34669	0	0.00%	10,804,110	0.09%	0	0.00%	6,876,895	0.05%	17,681,005	0.05%
32547	0	0.00%	0	0.00%	20,464,699	2.87%	0	0.00%	20,464,699	0.06%
32830	0	0.00%	0	0.00%	0	0.00%	0	0.00%	1,067,313	0.00%
34953	0	0.00%	106,041,759	0.87%	0	0.00%	114,996,976	0.82%	221,038,736	0.63%
33538	0	0.00%	5,885,066	0.05%	0	0.00%	5,908,470	0.04%	11,793,536	0.03%
32548	0	0.00%	0	0.00%	26,082,292	3.66%	0	0.00%	26,082,292	0.07%
32407	0	0.00%	0	0.00%	6,780,282	0.95%	0	0.00%	6,780,328	0.02%
34105	0	0.00%	7,065,318	0.06%	0	0.00%	0	0.00%	7,065,848	0.02%
32124	5,159,109	0.06%	3,547,410	0.03%	0	0.00%	5,209,862	0.04%	13,916,381	0.04%
32266	0	0.00%	7,129,444	0.06%	0	0.00%	7,155,623	0.05%	14,285,067	0.04%
32832	29,400,312	0.36%	17,335,929	0.14%	0	0.00%	24,046,545	0.17%	70,782,785	0.20%
32408	0	0.00%	0	0.00%	8,799,218	1.24%	0	0.00%	8,799,458	0.03%
33823	16,446,776	0.20%	0	0.00%	0	0.00%	37,137,692	0.26%	53,993,622	0.15%
33540	0	0.00%	4,499,519	0.04%	0	0.00%	7,622,254	0.05%	12,121,773	0.03%
32550	0	0.00%	0	0.00%	13,080,206	1.84%	0	0.00%	13,080,206	0.04%
32833	16,279,725	0.20%	7,355,022	0.06%	0	0.00%	13,694,223	0.10%	37,328,970	0.11%
34956	0	0.00%	6,827,748	0.06%	0	0.00%	6,853,953	0.05%	13,681,700	0.04%
32692	0	0.00%	781,518	0.01%	0	0.00%	0	0.00%	781,518	0.00%
33541	0	0.00%	10,284,673	0.08%	0	0.00%	16,801,390	0.12%	27,086,066	0.08%
33966	10,669,172	0.13%	0	0.00%	0	0.00%	0	0.00%	10,669,190	0.03%
34108	21,783,691	0.27%	58,040,331	0.48%	0	0.00%	0	0.00%	79,824,793	0.23%
34957	0	0.00%	58,864,333	0.48%	0	0.00%	59,034,861	0.42%	117,899,269	0.34%
32127	71,679,237	0.87%	72,088,527	0.59%	0	0.00%	87,830,959	0.62%	231,598,723	0.66%
32976	0	0.00%	30,693,823	0.25%	0	0.00%	95,376,069	0.68%	126,069,901	0.36%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33825	44,163,609	0.54%	13,239,304	0.11%	0	0.00%	38,292,541	0.27%	95,695,453	0.27%
32693	0	0.00%	3,414,376	0.03%	0	0.00%	0	0.00%	3,474,895	0.01%
33542	0	0.00%	9,020,882	0.07%	0	0.00%	15,800,109	0.11%	24,821,009	0.07%
33401	0	0.00%	55,577,447	0.46%	0	0.00%	41,725,726	0.30%	97,303,784	0.28%
33967	13,282,399	0.16%	0	0.00%	0	0.00%	0	0.00%	13,282,399	0.04%
32835	66,737,944	0.81%	31,576,646	0.26%	0	0.00%	67,294,115	0.48%	165,608,706	0.47%
34109	0	0.00%	8,940,303	0.07%	0	0.00%	0	0.00%	8,940,610	0.03%
32128	43,858,129	0.53%	18,843,377	0.16%	0	0.00%	36,907,760	0.26%	99,609,266	0.28%
32694	0	0.00%	0	0.00%	0	0.00%	702,057	0.01%	702,057	0.00%
33543	0	0.00%	26,346,085	0.22%	0	0.00%	35,776,491	0.25%	62,122,584	0.18%
34251	0	0.00%	0	0.00%	0	0.00%	4,954,577	0.04%	4,954,577	0.01%
32836	48,970,584	0.60%	24,295,651	0.20%	0	0.00%	59,624,972	0.42%	132,891,206	0.38%
34110	13,750,118	0.17%	13,825,591	0.11%	0	0.00%	0	0.00%	27,576,127	0.08%
32129	35,555,420	0.43%	21,459,656	0.18%	0	0.00%	29,370,499	0.21%	86,385,575	0.25%
33827	9,508,783	0.12%	3,940,728	0.03%	0	0.00%	6,568,919	0.05%	20,018,430	0.06%
33544	0	0.00%	18,205,793	0.15%	0	0.00%	24,402,034	0.17%	42,607,827	0.12%
33403	0	0.00%	18,301,774	0.15%	0	0.00%	13,349,576	0.09%	31,651,406	0.09%
32837	109,573,661	1.33%	35,635,809	0.29%	0	0.00%	80,018,430	0.57%	225,227,900	0.64%
32413	0	0.00%	0	0.00%	29,813,006	4.19%	0	0.00%	29,813,536	0.08%
32130	0	0.00%	0	0.00%	0	0.00%	3,197,792	0.02%	3,253,570	0.01%
34677	0	0.00%	21,440,142	0.18%	0	0.00%	3,187,248	0.02%	24,627,391	0.07%
32696	0	0.00%	5,331,980	0.04%	0	0.00%	4,577,113	0.03%	9,909,093	0.03%
33404	0	0.00%	63,793,626	0.52%	0	0.00%	42,371,086	0.30%	106,165,551	0.30%
33971	32,323,307	0.39%	0	0.00%	0	0.00%	0	0.00%	32,323,307	0.09%
32839	32,444,572	0.40%	11,703,251	0.10%	0	0.00%	26,973,150	0.19%	71,120,973	0.20%
33405	0	0.00%	40,916,678	0.34%	0	0.00%	21,249,462	0.15%	62,166,419	0.18%
32132	14,211,332	0.17%	14,294,543	0.12%	0	0.00%	17,520,979	0.12%	46,026,854	0.13%
33830	21,779,445	0.27%	513,031	0.00%	0	0.00%	30,111,108	0.21%	52,403,584	0.15%
33547	0	0.00%	0	0.00%	0	0.00%	17,207,127	0.12%	17,498,094	0.05%
33972	11,061,769	0.13%	0	0.00%	0	0.00%	0	0.00%	11,396,832	0.03%
33406	0	0.00%	39,551,329	0.33%	0	0.00%	23,495,566	0.17%	63,046,934	0.18%
33548	0	0.00%	6,151,334	0.05%	0	0.00%	6,175,953	0.04%	12,327,287	0.04%
33407	0	0.00%	40,816,865	0.34%	0	0.00%	29,831,676	0.21%	70,648,673	0.20%
32134	0	0.00%	0	0.00%	0	0.00%	3,448,796	0.02%	3,513,433	0.01%
34681	0	0.00%	2,213,618	0.02%	0	0.00%	2,222,471	0.02%	4,436,089	0.01%
33549	0	0.00%	17,882,621	0.15%	0	0.00%	17,952,099	0.13%	35,834,720	0.10%
33408	0	0.00%	95,816,584	0.79%	0	0.00%	76,651,655	0.54%	172,468,896	0.49%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32135	1,044,828	0.01%	888,778	0.01%	0	0.00%	1,048,813	0.01%	2,982,419	0.01%
32701	34,692,223	0.42%	15,379,493	0.13%	0	0.00%	25,764,266	0.18%	75,835,983	0.22%
33409	0	0.00%	40,621,324	0.33%	0	0.00%	23,073,972	0.16%	63,695,448	0.18%
32136	24,796,929	0.30%	20,633,192	0.17%	0	0.00%	25,019,209	0.18%	70,449,331	0.20%
33834	3,648,755	0.04%	0	0.00%	0	0.00%	3,684,843	0.03%	7,367,026	0.02%
34683	0	0.00%	34,412,168	0.28%	0	0.00%	34,515,378	0.25%	68,927,546	0.20%
32702	0	0.00%	0	0.00%	0	0.00%	1,270,125	0.01%	1,270,125	0.00%
32561	0	0.00%	0	0.00%	47,963,142	6.74%	0	0.00%	47,963,142	0.14%
33410	0	0.00%	108,805,761	0.89%	0	0.00%	95,756,755	0.68%	204,562,770	0.58%
32137	7,101,904	0.09%	53,739,265	0.44%	0	0.00%	76,982,448	0.55%	137,823,617	0.39%
34684	0	0.00%	22,832,751	0.19%	0	0.00%	4,364,658	0.03%	27,197,409	0.08%
32703	26,051,984	0.32%	800,658	0.01%	0	0.00%	45,041,897	0.32%	71,894,540	0.20%
33411	0	0.00%	136,417,743	1.12%	0	0.00%	89,050,464	0.63%	225,468,404	0.64%
33836	0	0.00%	0	0.00%	0	0.00%	0	0.00%	589,901	0.00%
34685	0	0.00%	20,774,793	0.17%	0	0.00%	1,597,299	0.01%	22,372,092	0.06%
32563	0	0.00%	0	0.00%	31,279,998	4.39%	0	0.00%	31,279,998	0.09%
33412	0	0.00%	53,695,938	0.44%	0	0.00%	47,532,907	0.34%	101,228,845	0.29%
33837	38,356,418	0.47%	17,540,511	0.14%	0	0.00%	32,138,068	0.23%	88,034,998	0.25%
32564	0	0.00%	0	0.00%	1,044,517	0.15%	0	0.00%	1,044,517	0.00%
33413	0	0.00%	22,932,636	0.19%	0	0.00%	14,798,300	0.11%	37,730,943	0.11%
33838	7,000,355	0.09%	1,833,899	0.02%	0	0.00%	4,805,628	0.03%	13,639,882	0.04%
33980	47,436,424	0.58%	0	0.00%	0	0.00%	1,383,558	0.01%	48,820,019	0.14%
32565	0	0.00%	0	0.00%	3,625,980	0.51%	0	0.00%	3,625,980	0.01%
33414	0	0.00%	132,948,587	1.09%	0	0.00%	101,311,819	0.72%	234,260,481	0.67%
32141	34,058,230	0.41%	34,257,745	0.28%	0	0.00%	42,036,225	0.30%	110,352,200	0.31%
33839	3,514,515	0.04%	1,577,206	0.01%	0	0.00%	3,549,194	0.03%	8,640,915	0.02%
34688	0	0.00%	11,360,541	0.09%	0	0.00%	11,391,882	0.08%	22,752,423	0.06%
32707	72,865,778	0.89%	25,229,578	0.21%	0	0.00%	60,348,632	0.43%	158,443,988	0.45%
33556	0	0.00%	29,082,548	0.24%	0	0.00%	29,193,686	0.21%	58,276,233	0.17%
33981	28,466,345	0.35%	0	0.00%	0	0.00%	0	0.00%	28,466,345	0.08%
32566	0	0.00%	0	0.00%	37,985,622	5.33%	0	0.00%	37,985,622	0.11%
33415	0	0.00%	46,801,691	0.39%	0	0.00%	35,405,638	0.25%	82,207,517	0.23%
34972	0	0.00%	14,119,065	0.12%	0	0.00%	20,183,653	0.14%	34,302,718	0.10%
34689	0	0.00%	23,621,787	0.19%	0	0.00%	23,683,367	0.17%	47,305,154	0.13%
32708	119,546,272	1.46%	41,377,205	0.34%	0	0.00%	99,063,400	0.70%	259,986,877	0.74%
33982	52,431,932	0.64%	0	0.00%	0	0.00%	7,257,232	0.05%	59,689,164	0.17%
33416	0	0.00%	0	0.00%	0	0.00%	0	0.00%	682,271	0.00%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34973	0	0.00%	0	0.00%	0	0.00%	0	0.00%	543,135	0.00%
33841	12,137,658	0.15%	0	0.00%	0	0.00%	7,685,075	0.05%	19,886,108	0.06%
34690	0	0.00%	7,910,537	0.07%	0	0.00%	7,941,965	0.06%	15,852,502	0.05%
32709	3,536,896	0.04%	1,268,504	0.01%	0	0.00%	2,915,515	0.02%	7,720,915	0.02%
33558	0	0.00%	29,989,326	0.25%	0	0.00%	19,963,624	0.14%	49,952,949	0.14%
33983	69,713,731	0.85%	0	0.00%	0	0.00%	1,906,916	0.01%	71,620,710	0.20%
32568	0	0.00%	0	0.00%	2,420,918	0.34%	0	0.00%	2,420,918	0.01%
33417	0	0.00%	50,628,782	0.42%	0	0.00%	32,847,160	0.23%	83,476,225	0.24%
34266	84,730,440	1.03%	0	0.00%	0	0.00%	20,015,979	0.14%	104,746,419	0.30%
34974	0	0.00%	39,597,159	0.33%	0	0.00%	47,244,900	0.34%	87,219,580	0.25%
34691	0	0.00%	11,266,393	0.09%	0	0.00%	18,546,948	0.13%	29,813,341	0.08%
33559	0	0.00%	9,064,704	0.07%	0	0.00%	9,096,231	0.06%	18,160,935	0.05%
33701	0	0.00%	12,159,050	0.10%	0	0.00%	0	0.00%	12,159,265	0.03%
32569	0	0.00%	0	0.00%	10,653,080	1.50%	0	0.00%	10,653,080	0.03%
33418	0	0.00%	183,571,577	1.51%	0	0.00%	149,277,328	1.06%	332,849,069	0.95%
33843	24,623,174	0.30%	4,574,313	0.04%	0	0.00%	13,254,045	0.09%	42,451,532	0.12%
33702	0	0.00%	19,185,458	0.16%	0	0.00%	0	0.00%	19,185,687	0.05%
32570	0	0.00%	0	0.00%	18,636,979	2.62%	0	0.00%	18,636,979	0.05%
34268	0	0.00%	0	0.00%	0	0.00%	0	0.00%	552,063	0.00%
33844	50,134,256	0.61%	19,081,791	0.16%	0	0.00%	50,621,307	0.36%	119,837,354	0.34%
32712	1,413,333	0.02%	1,421,670	0.01%	0	0.00%	58,506,195	0.42%	61,341,198	0.17%
33703	0	0.00%	22,542,721	0.19%	0	0.00%	0	0.00%	22,542,848	0.06%
32571	0	0.00%	0	0.00%	25,786,440	3.62%	0	0.00%	25,786,440	0.07%
34269	15,492,860	0.19%	0	0.00%	0	0.00%	3,611,591	0.03%	19,104,491	0.05%
32713	33,434,100	0.41%	0	0.00%	0	0.00%	33,763,186	0.24%	67,527,318	0.19%
33704	0	0.00%	17,472,557	0.14%	0	0.00%	0	0.00%	17,472,581	0.05%
32148	0	0.00%	0	0.00%	0	0.00%	3,134,945	0.02%	3,134,945	0.01%
33846	0	0.00%	0	0.00%	0	0.00%	509,000	0.00%	516,068	0.00%
34695	0	0.00%	17,304,510	0.14%	0	0.00%	0	0.00%	17,304,608	0.05%
32714	37,826,859	0.46%	3,162,487	0.03%	0	0.00%	38,164,049	0.27%	79,153,395	0.23%
33563	0	0.00%	0	0.00%	0	0.00%	14,104,892	0.10%	14,499,007	0.04%
32856	0	0.00%	0	0.00%	0	0.00%	0	0.00%	629,608	0.00%
33705	0	0.00%	14,885,179	0.12%	0	0.00%	0	0.00%	14,885,684	0.04%
33847	0	0.00%	0	0.00%	0	0.00%	0	0.00%	781,233	0.00%
33706	0	0.00%	60,618,131	0.50%	0	0.00%	0	0.00%	60,620,436	0.17%
33848	1,064,498	0.01%	0	0.00%	0	0.00%	754,557	0.01%	2,112,538	0.01%
33565	0	0.00%	9,207,832	0.08%	0	0.00%	21,452,091	0.15%	30,659,923	0.09%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33990	89,870,048	1.09%	0	0.00%	0	0.00%	0	0.00%	89,870,059	0.26%
33707	0	0.00%	42,223,482	0.35%	0	0.00%	0	0.00%	42,224,254	0.12%
33424	0	0.00%	1,022,934	0.01%	0	0.00%	695,113	0.00%	1,718,063	0.00%
33849	0	0.00%	0	0.00%	0	0.00%	724,368	0.01%	1,047,686	0.00%
34698	0	0.00%	34,135,980	0.28%	0	0.00%	7,266,390	0.05%	41,402,369	0.12%
33566	0	0.00%	12,180,519	0.10%	0	0.00%	21,289,828	0.15%	33,470,347	0.10%
34981	0	0.00%	7,995,361	0.07%	0	0.00%	8,025,929	0.06%	16,021,290	0.05%
33991	50,724,014	0.62%	0	0.00%	0	0.00%	0	0.00%	50,724,113	0.14%
33708	0	0.00%	58,166,192	0.48%	0	0.00%	0	0.00%	58,168,474	0.17%
33425	0	0.00%	0	0.00%	0	0.00%	0	0.00%	715,361	0.00%
33850	6,023,374	0.07%	0	0.00%	0	0.00%	8,313,051	0.06%	14,425,421	0.04%
32718	0	0.00%	0	0.00%	0	0.00%	0	0.00%	702,242	0.00%
33567	0	0.00%	6,350,978	0.05%	0	0.00%	10,311,118	0.07%	16,662,096	0.05%
34982	0	0.00%	56,370,965	0.46%	0	0.00%	52,249,020	0.37%	108,620,096	0.31%
33709	0	0.00%	14,060,854	0.12%	0	0.00%	0	0.00%	14,060,933	0.04%
32577	0	0.00%	0	0.00%	5,274,630	0.74%	0	0.00%	5,274,630	0.02%
33426	0	0.00%	33,437,361	0.28%	0	0.00%	15,677,272	0.11%	49,114,693	0.14%
34275	0	0.00%	1,913,824	0.02%	0	0.00%	0	0.00%	1,914,167	0.01%
33851	2,114,855	0.03%	646,710	0.01%	0	0.00%	1,869,260	0.01%	4,630,824	0.01%
34134	49,652,222	0.60%	49,935,799	0.41%	0	0.00%	0	0.00%	99,588,192	0.28%
34983	0	0.00%	84,117,811	0.69%	0	0.00%	91,331,770	0.65%	175,449,581	0.50%
33993	45,141,542	0.55%	0	0.00%	0	0.00%	0	0.00%	45,141,542	0.13%
33710	0	0.00%	25,252,279	0.21%	0	0.00%	0	0.00%	25,252,423	0.07%
32578	0	0.00%	0	0.00%	1,580,865	0.22%	0	0.00%	1,580,865	0.00%
33852	32,204,091	0.39%	32,391,807	0.27%	0	0.00%	53,955,827	0.38%	118,551,725	0.34%
32720	3,498,170	0.04%	3,516,598	0.03%	0	0.00%	24,757,236	0.18%	31,772,003	0.09%
33569	0	0.00%	3,547,004	0.03%	0	0.00%	54,077,679	0.38%	57,624,752	0.16%
34135	12,413,749	0.15%	0	0.00%	0	0.00%	0	0.00%	12,414,552	0.04%
34984	0	0.00%	34,096,128	0.28%	0	0.00%	36,994,903	0.26%	71,091,031	0.20%
33711	0	0.00%	13,115,040	0.11%	0	0.00%	0	0.00%	13,115,555	0.04%
32579	0	0.00%	0	0.00%	9,639,834	1.35%	0	0.00%	9,639,834	0.03%
33428	0	0.00%	36,381,325	0.30%	0	0.00%	0	0.00%	36,381,631	0.10%
33853	36,298,223	0.44%	8,078,727	0.07%	0	0.00%	23,642,711	0.17%	68,019,662	0.19%
33570	0	0.00%	540,186	0.00%	0	0.00%	9,440,127	0.07%	9,980,339	0.03%
33712	0	0.00%	11,943,616	0.10%	0	0.00%	0	0.00%	11,943,685	0.03%
32580	0	0.00%	0	0.00%	2,670,631	0.38%	0	0.00%	2,670,631	0.01%
34420	0	0.00%	8,823,880	0.07%	0	0.00%	14,960,532	0.11%	23,784,411	0.07%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34986	0	0.00%	61,491,626	0.51%	0	0.00%	66,743,771	0.47%	128,235,618	0.36%
33713	0	0.00%	17,621,544	0.14%	0	0.00%	0	0.00%	17,621,609	0.05%
33430	0	0.00%	6,393,155	0.05%	0	0.00%	6,414,020	0.05%	12,807,188	0.04%
33855	3,033,637	0.04%	1,829,821	0.02%	0	0.00%	3,063,661	0.02%	7,927,119	0.02%
33572	0	0.00%	516,262	0.00%	0	0.00%	18,369,836	0.13%	18,886,098	0.05%
34987	0	0.00%	10,387,078	0.09%	0	0.00%	11,282,598	0.08%	21,669,709	0.06%
33714	0	0.00%	8,500,220	0.07%	0	0.00%	0	0.00%	8,500,638	0.02%
33431	0	0.00%	26,450,333	0.22%	0	0.00%	0	0.00%	26,451,068	0.08%
34705	0	0.00%	1,395,903	0.01%	0	0.00%	2,233,210	0.02%	3,629,113	0.01%
32724	24,628,239	0.30%	2,189,886	0.02%	0	0.00%	24,852,844	0.18%	51,670,969	0.15%
33573	0	0.00%	7,393,809	0.06%	0	0.00%	25,637,950	0.18%	33,031,929	0.09%
33715	0	0.00%	27,714,990	0.23%	0	0.00%	10,015,094	0.07%	37,730,084	0.11%
32583	0	0.00%	0	0.00%	14,320,740	2.01%	0	0.00%	14,320,740	0.04%
33432	0	0.00%	41,656,138	0.34%	0	0.00%	0	0.00%	41,659,937	0.12%
32159	0	0.00%	31,277,863	0.26%	0	0.00%	52,494,084	0.37%	83,771,947	0.24%
33857	1,029,611	0.01%	2,495,401	0.02%	0	0.00%	3,637,043	0.03%	7,162,055	0.02%
32725	65,665,101	0.80%	878,910	0.01%	0	0.00%	66,309,898	0.47%	132,853,909	0.38%
33716	0	0.00%	5,897,141	0.05%	0	0.00%	0	0.00%	5,897,687	0.02%
33433	0	0.00%	50,006,327	0.41%	0	0.00%	0	0.00%	50,007,829	0.14%
32301	0	0.00%	13,877,144	0.11%	0	0.00%	0	0.00%	13,877,144	0.04%
33858	0	0.00%	0	0.00%	0	0.00%	0	0.00%	814,760	0.00%
32726	0	0.00%	15,125,903	0.12%	0	0.00%	25,869,129	0.18%	40,995,049	0.12%
34990	0	0.00%	120,636,615	0.99%	0	0.00%	121,045,761	0.86%	241,682,491	0.69%
33434	0	0.00%	36,049,751	0.30%	0	0.00%	0	0.00%	36,052,189	0.10%
32302	0	0.00%	2,187,810	0.02%	0	0.00%	0	0.00%	2,187,810	0.01%
33859	21,850,769	0.27%	5,700,914	0.05%	0	0.00%	13,476,702	0.10%	41,028,386	0.12%
33576	0	0.00%	5,344,808	0.04%	0	0.00%	5,366,118	0.04%	10,710,926	0.03%
34991	0	0.00%	1,025,629	0.01%	0	0.00%	1,020,114	0.01%	2,045,752	0.01%
33435	0	0.00%	50,795,510	0.42%	0	0.00%	26,727,669	0.19%	77,523,565	0.22%
32303	0	0.00%	27,222,711	0.22%	0	0.00%	0	0.00%	27,222,711	0.08%
33860	0	0.00%	565,995	0.00%	0	0.00%	16,168,464	0.11%	16,734,500	0.05%
34992	0	0.00%	0	0.00%	0	0.00%	0	0.00%	790,861	0.00%
32162	0	0.00%	58,351,499	0.48%	0	0.00%	58,579,798	0.42%	116,931,297	0.33%
33436	0	0.00%	61,420,096	0.51%	0	0.00%	39,462,179	0.28%	100,882,475	0.29%
34285	6,920,039	0.08%	6,958,040	0.06%	0	0.00%	0	0.00%	13,878,947	0.04%
32304	0	0.00%	9,647,122	0.08%	0	0.00%	0	0.00%	9,647,122	0.03%
33437	0	0.00%	109,089,541	0.90%	0	0.00%	75,179,816	0.53%	184,269,632	0.52%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34286	38,960,439	0.47%	0	0.00%	0	0.00%	0	0.00%	39,278,367	0.11%
32305	0	0.00%	4,495,668	0.04%	0	0.00%	0	0.00%	4,495,668	0.01%
34711	0	0.00%	55,699,484	0.46%	0	0.00%	99,964,802	0.71%	155,664,355	0.44%
32730	8,110,916	0.10%	0	0.00%	0	0.00%	5,847,504	0.04%	14,117,491	0.04%
34428	0	0.00%	13,145,996	0.11%	0	0.00%	5,784,498	0.04%	18,930,494	0.05%
34145	0	0.00%	61,802,244	0.51%	0	0.00%	0	0.00%	61,804,611	0.18%
34994	0	0.00%	42,478,944	0.35%	0	0.00%	42,498,335	0.30%	84,977,404	0.24%
32164	0	0.00%	36,511,740	0.30%	0	0.00%	36,656,635	0.26%	73,168,388	0.21%
33438	0	0.00%	1,140,313	0.01%	0	0.00%	979,655	0.01%	2,119,968	0.01%
34287	36,192,263	0.44%	0	0.00%	0	0.00%	0	0.00%	36,192,271	0.10%
34429	0	0.00%	14,485,980	0.12%	0	0.00%	14,524,783	0.10%	29,010,763	0.08%
34288	19,872,220	0.24%	0	0.00%	0	0.00%	665,366	0.00%	20,537,667	0.06%
32732	15,194,637	0.19%	4,799,077	0.04%	0	0.00%	10,944,613	0.08%	30,938,327	0.09%
34430	0	0.00%	0	0.00%	0	0.00%	0	0.00%	619,885	0.00%
34996	0	0.00%	70,416,709	0.58%	0	0.00%	65,411,489	0.46%	135,828,530	0.39%
32025	0	0.00%	0	0.00%	0	0.00%	7,604,864	0.05%	7,604,920	0.02%
33440	0	0.00%	10,523,234	0.09%	0	0.00%	10,561,341	0.08%	21,084,575	0.06%
34289	2,865,593	0.03%	0	0.00%	0	0.00%	0	0.00%	3,205,668	0.01%
32308	0	0.00%	19,729,425	0.16%	0	0.00%	0	0.00%	19,729,428	0.06%
34714	1,747,808	0.02%	1,757,443	0.01%	0	0.00%	15,865,283	0.11%	19,370,534	0.06%
34431	0	0.00%	15,412,741	0.13%	0	0.00%	11,405,928	0.08%	26,818,669	0.08%
34997	0	0.00%	124,081,059	1.02%	0	0.00%	112,956,174	0.80%	237,037,493	0.67%
33865	1,307,080	0.02%	0	0.00%	0	0.00%	741,713	0.01%	2,055,399	0.01%
33441	0	0.00%	17,504,658	0.14%	0	0.00%	0	0.00%	17,506,431	0.05%
32309	0	0.00%	22,905,862	0.19%	0	0.00%	0	0.00%	22,905,869	0.07%
34432	0	0.00%	16,313,875	0.13%	0	0.00%	16,378,776	0.12%	32,692,651	0.09%
32168	58,723,642	0.72%	36,137,703	0.30%	0	0.00%	59,293,584	0.42%	154,154,930	0.44%
34715	0	0.00%	7,890,773	0.06%	0	0.00%	14,513,853	0.10%	22,404,626	0.06%
33442	0	0.00%	29,431,878	0.24%	0	0.00%	0	0.00%	29,433,031	0.08%
32310	0	0.00%	5,536,652	0.05%	0	0.00%	0	0.00%	5,536,652	0.02%
32735	0	0.00%	3,182,364	0.03%	0	0.00%	4,882,008	0.03%	8,064,372	0.02%
33584	0	0.00%	0	0.00%	0	0.00%	15,666,055	0.11%	15,927,591	0.05%
34433	0	0.00%	8,427,419	0.07%	0	0.00%	6,407,983	0.05%	14,835,402	0.04%
32169	59,643,432	0.73%	68,210,008	0.56%	0	0.00%	75,220,761	0.53%	203,074,201	0.58%
33867	720,744	0.01%	0	0.00%	0	0.00%	0	0.00%	1,034,432	0.00%
32311	0	0.00%	13,209,049	0.11%	0	0.00%	0	0.00%	13,209,208	0.04%
32736	0	0.00%	0	0.00%	0	0.00%	11,882,265	0.08%	12,029,431	0.03%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33585	0	0.00%	1,011,208	0.01%	0	0.00%	1,336,379	0.01%	2,347,588	0.01%
34434	0	0.00%	13,302,516	0.11%	0	0.00%	10,238,071	0.07%	23,540,586	0.07%
33868	0	0.00%	4,749,276	0.04%	0	0.00%	10,524,860	0.07%	15,274,136	0.04%
33444	0	0.00%	20,659,466	0.17%	0	0.00%	3,745,383	0.03%	24,405,010	0.07%
34293	39,448,080	0.48%	0	0.00%	0	0.00%	0	0.00%	39,449,494	0.11%
32312	0	0.00%	35,364,762	0.29%	0	0.00%	0	0.00%	35,364,764	0.10%
33445	0	0.00%	53,720,846	0.44%	0	0.00%	12,181,245	0.09%	65,902,445	0.19%
32738	75,722,506	0.92%	0	0.00%	0	0.00%	56,821,841	0.40%	133,021,471	0.38%
34436	0	0.00%	8,306,853	0.07%	0	0.00%	8,340,014	0.06%	16,646,867	0.05%
33870	29,080,530	0.35%	12,744,457	0.10%	0	0.00%	35,793,070	0.25%	77,618,057	0.22%
33446	0	0.00%	65,949,035	0.54%	0	0.00%	48,650,825	0.35%	114,600,205	0.33%
33447	0	0.00%	744,011	0.01%	0	0.00%	0	0.00%	1,232,162	0.00%
32174	83,537,239	1.02%	84,023,159	0.69%	0	0.00%	113,427,350	0.81%	280,987,749	0.80%
33872	40,459,781	0.49%	15,207,033	0.13%	0	0.00%	40,848,549	0.29%	96,515,363	0.27%
33873	21,436,084	0.26%	0	0.00%	0	0.00%	12,471,255	0.09%	34,103,523	0.10%
32317	0	0.00%	10,498,037	0.09%	0	0.00%	0	0.00%	10,498,041	0.03%
32034	0	0.00%	0	0.00%	0	0.00%	6,430,860	0.05%	6,431,102	0.02%
32176	78,052,751	0.95%	62,940,271	0.52%	0	0.00%	71,955,975	0.51%	212,948,997	0.61%
32601	0	0.00%	1,828,558	0.02%	0	0.00%	7,047,241	0.05%	8,875,799	0.03%
33592	0	0.00%	4,512,136	0.04%	0	0.00%	7,467,147	0.05%	11,979,285	0.03%
33875	17,948,560	0.22%	7,915,856	0.07%	0	0.00%	22,156,194	0.16%	48,020,609	0.14%
32744	4,622,412	0.06%	0	0.00%	0	0.00%	4,668,232	0.03%	9,335,443	0.03%
34442	0	0.00%	29,050,701	0.24%	0	0.00%	22,520,139	0.16%	51,570,840	0.15%
33876	8,090,801	0.10%	6,025,176	0.05%	0	0.00%	11,484,095	0.08%	25,600,072	0.07%
32603	0	0.00%	0	0.00%	0	0.00%	1,272,072	0.01%	1,379,342	0.00%
33594	0	0.00%	1,581,481	0.01%	0	0.00%	57,644,512	0.41%	59,226,047	0.17%
32179	0	0.00%	4,119,434	0.03%	0	0.00%	7,144,521	0.05%	11,263,955	0.03%
33877	660,671	0.01%	0	0.00%	0	0.00%	531,659	0.00%	1,363,650	0.00%
32038	0	0.00%	0	0.00%	0	0.00%	3,189,674	0.02%	3,189,674	0.01%
32746	93,168,648	1.13%	42,764,851	0.35%	0	0.00%	70,182,473	0.50%	206,115,972	0.59%
32180	0	0.00%	0	0.00%	0	0.00%	1,577,493	0.01%	1,577,493	0.00%
32605	0	0.00%	1,212,863	0.01%	0	0.00%	18,497,458	0.13%	19,710,321	0.06%
32606	0	0.00%	3,260,613	0.03%	0	0.00%	17,572,856	0.12%	20,833,470	0.06%
33455	0	0.00%	105,993,657	0.87%	0	0.00%	96,407,154	0.68%	202,400,908	0.58%
33597	0	0.00%	4,875,865	0.04%	0	0.00%	6,571,690	0.05%	11,447,555	0.03%
34446	0	0.00%	25,683,364	0.21%	0	0.00%	15,987,620	0.11%	41,670,983	0.12%
33880	31,947,528	0.39%	1,246,988	0.01%	0	0.00%	43,976,546	0.31%	77,171,062	0.22%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32607	0	0.00%	4,020,560	0.03%	0	0.00%	17,189,693	0.12%	21,210,253	0.06%
33598	0	0.00%	0	0.00%	0	0.00%	3,545,669	0.03%	3,602,258	0.01%
34447	0	0.00%	0	0.00%	0	0.00%	0	0.00%	546,074	0.00%
33881	40,408,940	0.49%	17,727,122	0.15%	0	0.00%	49,942,921	0.35%	108,078,983	0.31%
32608	0	0.00%	22,579,934	0.19%	0	0.00%	22,655,203	0.16%	45,235,137	0.13%
34448	0	0.00%	15,899,030	0.13%	0	0.00%	11,655,114	0.08%	27,554,144	0.08%
34731	0	0.00%	13,528,577	0.11%	0	0.00%	18,622,774	0.13%	32,151,354	0.09%
32750	38,634,883	0.47%	21,405,902	0.18%	0	0.00%	39,014,157	0.28%	99,054,942	0.28%
32609	0	0.00%	1,096,227	0.01%	0	0.00%	6,176,979	0.04%	7,273,206	0.02%
33458	0	0.00%	150,313,434	1.24%	0	0.00%	121,999,675	0.87%	272,313,355	0.77%
34449	0	0.00%	1,978,479	0.02%	0	0.00%	1,371,651	0.01%	3,350,129	0.01%
32751	59,268,775	0.72%	25,458,507	0.21%	0	0.00%	43,778,283	0.31%	128,505,564	0.37%
32327	0	0.00%	11,104,861	0.09%	0	0.00%	0	0.00%	11,104,861	0.03%
32893	0	0.00%	0	0.00%	0	0.00%	0	0.00%	573,710	0.00%
34450	0	0.00%	16,253,663	0.13%	0	0.00%	16,314,260	0.12%	32,567,923	0.09%
33884	81,023,621	0.99%	27,379,245	0.23%	0	0.00%	71,122,096	0.51%	179,524,962	0.51%
33601	0	0.00%	0	0.00%	0	0.00%	0	0.00%	539,782	0.00%
33460	0	0.00%	39,738,912	0.33%	0	0.00%	24,266,935	0.17%	64,005,965	0.18%
34734	3,914,000	0.05%	0	0.00%	0	0.00%	8,589,672	0.06%	12,567,440	0.04%
33602	0	0.00%	11,627,462	0.10%	0	0.00%	5,216,426	0.04%	16,843,888	0.05%
33461	0	0.00%	38,616,291	0.32%	0	0.00%	18,482,576	0.13%	57,099,003	0.16%
34452	0	0.00%	14,828,381	0.12%	0	0.00%	14,886,569	0.11%	29,714,950	0.08%
32754	17,000,319	0.21%	12,658,772	0.10%	0	0.00%	20,919,577	0.15%	50,578,668	0.14%
33603	0	0.00%	684,078	0.01%	0	0.00%	681,898	0.00%	1,365,976	0.00%
33462	0	0.00%	58,710,698	0.48%	0	0.00%	28,172,188	0.20%	86,883,061	0.25%
34736	0	0.00%	9,664,499	0.08%	0	0.00%	17,342,401	0.12%	27,006,900	0.08%
33604	0	0.00%	1,034,022	0.01%	0	0.00%	1,031,459	0.01%	2,065,481	0.01%
34453	0	0.00%	15,348,737	0.13%	0	0.00%	11,630,578	0.08%	26,979,315	0.08%
33463	0	0.00%	70,514,277	0.58%	0	0.00%	35,520,434	0.25%	106,034,852	0.30%
32331	0	0.00%	1,553,606	0.01%	0	0.00%	0	0.00%	1,553,606	0.00%
34737	0	0.00%	4,820,285	0.04%	0	0.00%	7,514,673	0.05%	12,334,958	0.04%
33605	0	0.00%	765,706	0.01%	0	0.00%	762,415	0.01%	1,528,121	0.00%
32615	0	0.00%	0	0.00%	0	0.00%	9,038,138	0.06%	9,433,348	0.03%
32757	0	0.00%	1,227,870	0.01%	0	0.00%	31,309,623	0.22%	32,537,524	0.09%
33606	0	0.00%	17,949,098	0.15%	0	0.00%	1,882,173	0.01%	19,831,271	0.06%
32192	0	0.00%	0	0.00%	0	0.00%	528,681	0.00%	825,866	0.00%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33890	8,796,300	0.11%	1,737,328	0.01%	0	0.00%	4,000,482	0.03%	14,534,110	0.04%
34739	951,192	0.01%	1,278,685	0.01%	0	0.00%	2,233,530	0.02%	4,463,407	0.01%
33607	0	0.00%	777,477	0.01%	0	0.00%	774,479	0.01%	1,551,956	0.00%
32617	0	0.00%	1,994,082	0.02%	0	0.00%	3,383,062	0.02%	5,377,144	0.02%
32759	3,317,558	0.04%	3,336,956	0.03%	0	0.00%	4,786,909	0.03%	11,441,422	0.03%
32618	0	0.00%	2,949,882	0.02%	0	0.00%	2,961,689	0.02%	5,911,571	0.02%
33467	0	0.00%	129,697,431	1.07%	0	0.00%	101,197,076	0.72%	230,894,666	0.66%
32052	0	0.00%	1,722,929	0.01%	0	0.00%	0	0.00%	1,806,155	0.01%
32901	0	0.00%	36,913,977	0.30%	0	0.00%	46,974,570	0.33%	83,889,009	0.24%
34741	50,542,483	0.62%	18,995,411	0.16%	0	0.00%	37,131,507	0.26%	106,669,401	0.30%
33609	0	0.00%	1,964,356	0.02%	0	0.00%	1,956,761	0.01%	3,921,117	0.01%
33468	0	0.00%	0	0.00%	0	0.00%	0	0.00%	580,259	0.00%
32053	0	0.00%	980,939	0.01%	0	0.00%	0	0.00%	1,019,548	0.00%
32619	0	0.00%	1,545,461	0.01%	0	0.00%	0	0.00%	1,598,626	0.00%
32195	0	0.00%	2,171,090	0.02%	0	0.00%	3,563,907	0.03%	5,734,997	0.02%
33610	0	0.00%	1,009,067	0.01%	0	0.00%	13,367,421	0.09%	14,376,489	0.04%
33469	0	0.00%	79,450,135	0.65%	0	0.00%	72,995,176	0.52%	152,445,437	0.43%
32903	0	0.00%	66,434,923	0.55%	0	0.00%	80,995,868	0.58%	147,431,456	0.42%
34601	0	0.00%	11,972,480	0.10%	0	0.00%	21,200,170	0.15%	33,172,650	0.09%
34743	78,045,369	0.95%	22,068,009	0.18%	0	0.00%	63,375,970	0.45%	163,489,349	0.47%
33611	0	0.00%	27,214,440	0.22%	0	0.00%	4,419,444	0.03%	31,633,884	0.09%
32055	0	0.00%	0	0.00%	0	0.00%	6,324,281	0.04%	6,324,414	0.02%
32904	0	0.00%	45,540,868	0.37%	0	0.00%	63,517,745	0.45%	109,058,745	0.31%
34602	0	0.00%	8,511,197	0.07%	0	0.00%	8,545,080	0.06%	17,056,277	0.05%
32621	0	0.00%	1,715,190	0.01%	0	0.00%	1,338,416	0.01%	3,053,606	0.01%
33470	0	0.00%	55,071,238	0.45%	0	0.00%	46,073,427	0.33%	101,144,665	0.29%
34744	101,285,965	1.23%	32,459,930	0.27%	0	0.00%	81,680,153	0.58%	215,426,048	0.61%
32763	12,910,710	0.16%	608,895	0.01%	0	0.00%	13,033,739	0.09%	26,553,344	0.08%
33612	0	0.00%	2,702,189	0.02%	0	0.00%	18,217,757	0.13%	20,919,946	0.06%
34461	0	0.00%	14,922,202	0.12%	0	0.00%	14,980,986	0.11%	29,903,188	0.09%
32905	0	0.00%	37,104,411	0.31%	0	0.00%	63,092,877	0.45%	100,197,551	0.29%
32622	0	0.00%	0	0.00%	0	0.00%	640,660	0.00%	640,660	0.00%
33471	0	0.00%	4,999,824	0.04%	0	0.00%	5,019,959	0.04%	10,019,783	0.03%
33896	10,805,123	0.13%	1,292,017	0.01%	0	0.00%	13,313,103	0.09%	25,410,243	0.07%
32764	5,945,978	0.07%	2,162,637	0.02%	0	0.00%	3,672,774	0.03%	11,781,389	0.03%
33613	0	0.00%	18,131,397	0.15%	0	0.00%	18,167,596	0.13%	36,298,993	0.10%
32340	0	0.00%	4,067,093	0.03%	0	0.00%	0	0.00%	4,067,136	0.01%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33755	0	0.00%	14,299,867	0.12%	0	0.00%	1,445,437	0.01%	15,745,303	0.04%
34604	0	0.00%	7,945,604	0.07%	0	0.00%	5,662,961	0.04%	13,608,565	0.04%
33897	0	0.00%	0	0.00%	0	0.00%	29,541,456	0.21%	29,817,510	0.08%
34746	96,106,766	1.17%	35,862,256	0.30%	0	0.00%	76,298,162	0.54%	208,267,185	0.59%
32765	150,916,546	1.84%	47,419,438	0.39%	0	0.00%	109,526,230	0.78%	307,862,214	0.88%
33614	0	0.00%	4,460,715	0.04%	0	0.00%	4,438,780	0.03%	8,899,495	0.03%
32907	0	0.00%	54,047,156	0.44%	0	0.00%	102,437,647	0.73%	156,484,803	0.45%
33756	0	0.00%	26,578,553	0.22%	0	0.00%	5,414,157	0.04%	31,992,710	0.09%
33898	44,685,225	0.54%	8,704,133	0.07%	0	0.00%	25,064,796	0.18%	78,454,155	0.22%
34747	60,412,523	0.74%	37,963,727	0.31%	0	0.00%	60,894,010	0.43%	159,270,260	0.45%
32766	37,584,793	0.46%	15,624,890	0.13%	0	0.00%	27,842,708	0.20%	81,052,391	0.23%
33615	0	0.00%	25,578,225	0.21%	0	0.00%	2,420,267	0.02%	27,998,492	0.08%
32059	0	0.00%	889,044	0.01%	0	0.00%	0	0.00%	889,044	0.00%
32908	0	0.00%	11,175,160	0.09%	0	0.00%	19,431,706	0.14%	30,606,866	0.09%
34606	0	0.00%	40,533,893	0.33%	0	0.00%	22,958,476	0.16%	63,492,369	0.18%
32625	0	0.00%	2,708,106	0.02%	0	0.00%	1,573,115	0.01%	4,281,221	0.01%
34748	0	0.00%	31,054,385	0.26%	0	0.00%	48,701,966	0.35%	79,756,382	0.23%
32767	0	0.00%	0	0.00%	0	0.00%	1,101,141	0.01%	1,117,938	0.00%
33616	0	0.00%	6,630,025	0.05%	0	0.00%	1,017,085	0.01%	7,647,110	0.02%
34465	0	0.00%	27,855,550	0.23%	0	0.00%	16,827,250	0.12%	44,682,800	0.13%
32909	0	0.00%	45,008,911	0.37%	0	0.00%	67,144,949	0.48%	112,153,889	0.32%
34607	0	0.00%	14,696,444	0.12%	0	0.00%	8,476,412	0.06%	23,172,855	0.07%
32626	0	0.00%	3,863,983	0.03%	0	0.00%	0	0.00%	3,928,271	0.01%
33475	0	0.00%	1,004,888	0.01%	0	0.00%	916,946	0.01%	1,921,837	0.01%
33617	0	0.00%	6,180,457	0.05%	0	0.00%	26,670,581	0.19%	32,851,038	0.09%
32344	0	0.00%	6,194,579	0.05%	0	0.00%	0	0.00%	6,194,593	0.02%
33759	0	0.00%	11,282,288	0.09%	0	0.00%	0	0.00%	11,282,682	0.03%
34608	0	0.00%	44,552,068	0.37%	0	0.00%	25,867,752	0.18%	70,419,820	0.20%
33476	0	0.00%	4,948,443	0.04%	0	0.00%	3,976,960	0.03%	8,925,403	0.03%
33901	40,674,788	0.50%	0	0.00%	0	0.00%	0	0.00%	40,674,862	0.12%
33618	0	0.00%	23,685,937	0.19%	0	0.00%	23,755,273	0.17%	47,441,210	0.13%
33477	0	0.00%	91,908,189	0.76%	0	0.00%	83,531,883	0.59%	175,440,445	0.50%
32628	0	0.00%	1,373,601	0.01%	0	0.00%	0	0.00%	1,373,601	0.00%
34609	0	0.00%	51,763,458	0.43%	0	0.00%	33,267,009	0.24%	85,030,466	0.24%
33760	0	0.00%	8,667,098	0.07%	0	0.00%	0	0.00%	8,667,511	0.02%
32062	0	0.00%	870,770	0.01%	0	0.00%	0	0.00%	890,769	0.00%
33619	0	0.00%	736,045	0.01%	0	0.00%	11,740,205	0.08%	12,476,250	0.04%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33478	0	0.00%	47,557,765	0.39%	0	0.00%	44,101,040	0.31%	91,658,805	0.26%
34610	0	0.00%	9,851,476	0.08%	0	0.00%	9,890,959	0.07%	19,742,435	0.06%
33761	0	0.00%	19,561,626	0.16%	0	0.00%	5,372,145	0.04%	24,933,771	0.07%
32346	0	0.00%	2,563,360	0.02%	0	0.00%	0	0.00%	2,563,360	0.01%
32771	68,378,452	0.83%	35,825,159	0.29%	0	0.00%	53,384,309	0.38%	157,587,920	0.45%
33903	79,424,987	0.97%	0	0.00%	0	0.00%	0	0.00%	79,425,133	0.23%
33762	0	0.00%	6,377,294	0.05%	0	0.00%	0	0.00%	6,377,456	0.02%
32064	0	0.00%	1,531,241	0.01%	0	0.00%	0	0.00%	1,559,517	0.00%
32347	0	0.00%	4,395,301	0.04%	0	0.00%	0	0.00%	4,395,301	0.01%
34470	0	0.00%	12,031,408	0.10%	0	0.00%	12,067,378	0.09%	24,098,785	0.07%
34753	0	0.00%	2,536,603	0.02%	0	0.00%	4,494,293	0.03%	7,030,896	0.02%
33904	144,037,842	1.75%	0	0.00%	0	0.00%	0	0.00%	144,038,281	0.41%
33480	0	0.00%	239,431,348	1.97%	0	0.00%	160,674,547	1.14%	400,107,456	1.14%
32631	0	0.00%	0	0.00%	0	0.00%	546,886	0.00%	577,038	0.00%
33763	0	0.00%	1,352,992	0.01%	0	0.00%	1,346,133	0.01%	2,699,125	0.01%
32348	0	0.00%	3,627,915	0.03%	0	0.00%	0	0.00%	3,627,915	0.01%
34471	0	0.00%	21,716,181	0.18%	0	0.00%	38,554,803	0.27%	60,270,985	0.17%
32773	31,908,991	0.39%	14,234,659	0.12%	0	0.00%	23,795,310	0.17%	69,938,959	0.20%
33905	42,584,471	0.52%	0	0.00%	0	0.00%	1,556,114	0.01%	44,140,634	0.13%
34613	0	0.00%	23,868,831	0.20%	0	0.00%	14,670,118	0.10%	38,538,949	0.11%
33764	0	0.00%	20,099,072	0.17%	0	0.00%	0	0.00%	20,099,746	0.06%
32066	0	0.00%	1,476,080	0.01%	0	0.00%	0	0.00%	1,476,080	0.00%
34755	0	0.00%	0	0.00%	0	0.00%	906,430	0.01%	1,301,934	0.00%
34472	0	0.00%	18,329,972	0.15%	0	0.00%	29,558,550	0.21%	47,888,523	0.14%
33765	0	0.00%	8,044,455	0.07%	0	0.00%	0	0.00%	8,044,730	0.02%
34614	0	0.00%	5,244,388	0.04%	0	0.00%	4,020,558	0.03%	9,264,946	0.03%
33907	35,294,463	0.43%	0	0.00%	0	0.00%	0	0.00%	35,294,626	0.10%
34756	0	0.00%	3,225,675	0.03%	0	0.00%	6,488,156	0.05%	9,713,831	0.03%
32775	613,516	0.01%	0	0.00%	0	0.00%	755,678	0.01%	1,817,516	0.01%
33624	0	0.00%	31,360,256	0.26%	0	0.00%	31,476,884	0.22%	62,837,140	0.18%
34473	0	0.00%	17,319,230	0.14%	0	0.00%	17,388,322	0.12%	34,707,552	0.10%
32634	0	0.00%	0	0.00%	0	0.00%	0	0.00%	593,966	0.00%
33483	0	0.00%	43,012,302	0.35%	0	0.00%	13,055,462	0.09%	56,068,547	0.16%
34474	0	0.00%	19,499,285	0.16%	0	0.00%	19,554,240	0.14%	39,053,525	0.11%
33625	0	0.00%	16,164,832	0.13%	0	0.00%	16,228,665	0.12%	32,393,497	0.09%
33908	127,081,973	1.55%	0	0.00%	0	0.00%	0	0.00%	127,082,470	0.36%
32776	0	0.00%	8,850,497	0.07%	0	0.00%	13,122,325	0.09%	22,127,614	0.06%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33484	0	0.00%	45,743,874	0.38%	0	0.00%	10,845,471	0.08%	56,589,537	0.16%
33767	0	0.00%	46,234,897	0.38%	0	0.00%	27,537,318	0.20%	73,772,216	0.21%
33909	46,956,383	0.57%	0	0.00%	0	0.00%	0	0.00%	46,956,383	0.13%
34475	0	0.00%	3,879,447	0.03%	0	0.00%	3,893,592	0.03%	7,773,039	0.02%
33626	0	0.00%	27,675,875	0.23%	0	0.00%	1,133,035	0.01%	28,808,910	0.08%
34758	60,457,739	0.74%	22,452,986	0.18%	0	0.00%	48,695,997	0.35%	131,606,721	0.37%
33060	0	0.00%	6,129,012	0.05%	0	0.00%	0	0.00%	6,129,567	0.02%
32777	0	0.00%	0	0.00%	0	0.00%	679,706	0.00%	686,455	0.00%
34476	0	0.00%	30,907,715	0.25%	0	0.00%	31,030,138	0.22%	61,937,853	0.18%
32778	0	0.00%	15,313,673	0.13%	0	0.00%	24,552,305	0.17%	39,866,016	0.11%
34759	59,015,165	0.72%	24,929,175	0.21%	0	0.00%	46,140,647	0.33%	130,084,987	0.37%
33486	0	0.00%	22,153,561	0.18%	0	0.00%	0	0.00%	22,153,922	0.06%
32920	18,154,115	0.22%	30,836,757	0.25%	0	0.00%	37,857,491	0.27%	86,848,363	0.25%
32779	79,441,164	0.97%	46,028,644	0.38%	0	0.00%	80,186,516	0.57%	205,656,324	0.58%
33062	0	0.00%	32,848,615	0.27%	0	0.00%	0	0.00%	32,851,410	0.09%
34760	0	0.00%	0	0.00%	0	0.00%	1,678,844	0.01%	1,692,963	0.00%
33487	0	0.00%	55,875,880	0.46%	0	0.00%	16,965,790	0.12%	72,842,130	0.21%
33770	0	0.00%	16,264,724	0.13%	0	0.00%	1,097,288	0.01%	17,362,012	0.05%
34478	0	0.00%	0	0.00%	0	0.00%	0	0.00%	507,141	0.00%
33629	0	0.00%	4,127,833	0.03%	0	0.00%	4,110,935	0.03%	8,238,768	0.02%
32780	48,854,871	0.60%	49,136,739	0.40%	0	0.00%	80,905,152	0.57%	178,896,762	0.51%
34761	27,845,302	0.34%	1,253,700	0.01%	0	0.00%	56,968,851	0.40%	86,067,853	0.24%
33912	80,680,248	0.98%	0	0.00%	0	0.00%	0	0.00%	80,680,656	0.23%
33771	0	0.00%	14,749,544	0.12%	0	0.00%	0	0.00%	14,750,061	0.04%
32922	6,893,723	0.08%	16,558,191	0.14%	0	0.00%	23,087,554	0.16%	46,539,468	0.13%
34479	0	0.00%	8,101,372	0.07%	0	0.00%	14,584,323	0.10%	22,685,695	0.06%
33913	32,351,195	0.39%	0	0.00%	0	0.00%	0	0.00%	32,351,415	0.09%
33064	0	0.00%	7,998,526	0.07%	0	0.00%	0	0.00%	7,999,141	0.02%
34762	0	0.00%	0	0.00%	0	0.00%	721,430	0.01%	1,131,086	0.00%
32640	0	0.00%	0	0.00%	0	0.00%	4,298,144	0.03%	4,380,212	0.01%
33772	0	0.00%	32,191,859	0.26%	0	0.00%	0	0.00%	32,192,124	0.09%
34480	0	0.00%	11,855,677	0.10%	0	0.00%	18,736,648	0.13%	30,592,325	0.09%
33914	144,662,958	1.76%	0	0.00%	0	0.00%	0	0.00%	144,663,103	0.41%
32641	0	0.00%	0	0.00%	0	0.00%	3,766,331	0.03%	4,061,114	0.01%
33773	0	0.00%	11,798,190	0.10%	0	0.00%	0	0.00%	11,798,353	0.03%
32924	0	0.00%	0	0.00%	0	0.00%	568,516	0.00%	1,296,119	0.00%
34481	0	0.00%	19,347,045	0.16%	0	0.00%	19,423,977	0.14%	38,771,022	0.11%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32359	0	0.00%	2,011,923	0.02%	0	0.00%	0	0.00%	2,011,923	0.01%
33774	0	0.00%	27,224,037	0.22%	0	0.00%	3,523,475	0.03%	30,747,513	0.09%
33916	19,070,003	0.23%	0	0.00%	0	0.00%	0	0.00%	19,070,003	0.05%
32501	0	0.00%	0	0.00%	12,655,162	1.78%	0	0.00%	12,655,162	0.04%
34482	0	0.00%	21,097,030	0.17%	0	0.00%	21,181,351	0.15%	42,278,381	0.12%
32784	0	0.00%	6,381,774	0.05%	0	0.00%	10,905,965	0.08%	17,287,739	0.05%
33067	0	0.00%	2,498,554	0.02%	0	0.00%	0	0.00%	2,498,642	0.01%
32643	0	0.00%	0	0.00%	0	0.00%	5,579,233	0.04%	5,723,635	0.02%
32926	14,836,715	0.18%	26,555,840	0.22%	0	0.00%	44,868,379	0.32%	86,260,934	0.25%
33634	0	0.00%	10,721,121	0.09%	0	0.00%	0	0.00%	11,154,896	0.03%
33917	72,072,205	0.88%	0	0.00%	0	0.00%	0	0.00%	72,072,240	0.21%
32502	0	0.00%	0	0.00%	1,096,150	0.15%	0	0.00%	1,096,150	0.00%
33776	0	0.00%	33,084,188	0.27%	0	0.00%	635,104	0.00%	33,719,292	0.10%
32927	32,339,921	0.39%	32,528,563	0.27%	0	0.00%	53,914,450	0.38%	118,782,935	0.34%
33493	0	0.00%	1,687,812	0.01%	0	0.00%	1,215,745	0.01%	2,903,557	0.01%
32503	0	0.00%	0	0.00%	41,829,932	5.87%	0	0.00%	41,829,932	0.12%
34484	0	0.00%	3,249,697	0.03%	0	0.00%	3,258,059	0.02%	6,507,756	0.02%
33635	0	0.00%	9,193,884	0.08%	0	0.00%	814,860	0.01%	10,008,743	0.03%
33069	0	0.00%	16,070,446	0.13%	0	0.00%	0	0.00%	16,071,280	0.05%
33777	0	0.00%	26,528,569	0.22%	0	0.00%	0	0.00%	26,529,737	0.08%
34202	0	0.00%	0	0.00%	0	0.00%	3,501,535	0.02%	3,501,643	0.01%
32504	0	0.00%	0	0.00%	31,440,433	4.42%	0	0.00%	31,440,433	0.09%
33919	100,489,442	1.22%	0	0.00%	0	0.00%	0	0.00%	100,490,624	0.29%
33778	0	0.00%	16,676,251	0.14%	0	0.00%	0	0.00%	16,676,643	0.05%
32080	0	0.00%	33,569,509	0.28%	0	0.00%	38,361,738	0.27%	71,931,491	0.20%
33920	8,347,640	0.10%	0	0.00%	0	0.00%	0	0.00%	8,528,016	0.02%
32505	0	0.00%	0	0.00%	18,018,848	2.53%	0	0.00%	18,018,848	0.05%
33637	0	0.00%	1,179,520	0.01%	0	0.00%	6,895,746	0.05%	8,075,266	0.02%
34769	53,037,623	0.65%	14,908,856	0.12%	0	0.00%	42,240,101	0.30%	110,186,580	0.31%
33496	0	0.00%	51,537,309	0.42%	0	0.00%	6,633,154	0.05%	58,170,463	0.17%
32506	0	0.00%	0	0.00%	38,400,499	5.39%	0	0.00%	38,400,499	0.11%
32789	111,799,715	1.36%	40,720,059	0.33%	0	0.00%	92,922,909	0.66%	245,442,683	0.70%
33921	54,338,175	0.66%	18,763,157	0.15%	0	0.00%	18,819,453	0.13%	91,920,785	0.26%
32648	0	0.00%	710,416	0.01%	0	0.00%	0	0.00%	710,416	0.00%
32931	23,735,468	0.29%	76,291,208	0.63%	0	0.00%	94,361,869	0.67%	194,388,545	0.55%
32082	0	0.00%	59,240,922	0.49%	0	0.00%	59,452,594	0.42%	118,693,516	0.34%
33922	30,172,233	0.37%	0	0.00%	0	0.00%	0	0.00%	30,172,233	0.09%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34205	0	0.00%	3,532,440	0.03%	0	0.00%	0	0.00%	3,532,791	0.01%
32507	0	0.00%	0	0.00%	80,485,062	11.30%	0	0.00%	80,485,062	0.23%
34488	0	0.00%	0	0.00%	0	0.00%	4,528,415	0.03%	4,793,216	0.01%
33073	0	0.00%	2,088,120	0.02%	0	0.00%	0	0.00%	2,088,309	0.01%
34771	34,435,116	0.42%	19,087,523	0.16%	0	0.00%	30,169,554	0.21%	83,692,194	0.24%
33498	0	0.00%	21,571,946	0.18%	0	0.00%	0	0.00%	21,940,989	0.06%
33781	0	0.00%	11,941,184	0.10%	0	0.00%	0	0.00%	11,941,294	0.03%
32932	0	0.00%	0	0.00%	0	0.00%	604,977	0.00%	1,222,418	0.00%
34772	43,969,032	0.54%	15,698,242	0.13%	0	0.00%	35,133,828	0.25%	94,801,102	0.27%
33782	0	0.00%	14,894,204	0.12%	0	0.00%	0	0.00%	14,894,714	0.04%
32084	0	0.00%	19,716,696	0.16%	0	0.00%	19,784,804	0.14%	39,501,500	0.11%
34207	0	0.00%	16,148,464	0.13%	0	0.00%	0	0.00%	16,149,369	0.05%
32792	89,619,305	1.09%	30,742,179	0.25%	0	0.00%	74,305,942	0.53%	194,667,426	0.55%
34773	2,124,874	0.03%	2,550,731	0.02%	0	0.00%	4,028,529	0.03%	8,704,133	0.02%
33924	84,846,405	1.03%	13,041,261	0.11%	0	0.00%	9,789,175	0.07%	107,676,842	0.31%
32934	508,744	0.01%	39,305,803	0.32%	0	0.00%	55,054,803	0.39%	94,869,350	0.27%
34208	0	0.00%	0	0.00%	0	0.00%	1,719,096	0.01%	1,719,723	0.00%
34491	0	0.00%	23,391,608	0.19%	0	0.00%	34,775,066	0.25%	58,166,674	0.17%
32793	0	0.00%	0	0.00%	0	0.00%	0	0.00%	883,710	0.00%
33076	0	0.00%	989,312	0.01%	0	0.00%	0	0.00%	989,353	0.00%
32935	0	0.00%	81,479,401	0.67%	0	0.00%	101,992,624	0.72%	183,472,478	0.52%
32086	0	0.00%	2,313,596	0.02%	0	0.00%	20,487,817	0.15%	22,801,413	0.06%
34209	0	0.00%	7,016,131	0.06%	0	0.00%	0	0.00%	7,017,014	0.02%
32794	0	0.00%	0	0.00%	0	0.00%	0	0.00%	586,481	0.00%
32653	0	0.00%	0	0.00%	0	0.00%	10,519,037	0.07%	10,988,637	0.03%
33785	0	0.00%	27,851,070	0.23%	0	0.00%	18,609,063	0.13%	46,460,133	0.13%
34210	0	0.00%	17,792,562	0.15%	0	0.00%	0	0.00%	17,793,767	0.05%
33786	0	0.00%	8,320,994	0.07%	0	0.00%	6,167,014	0.04%	14,488,007	0.04%
32937	5,903,050	0.07%	107,759,779	0.89%	0	0.00%	116,588,045	0.83%	230,250,874	0.65%
32796	26,631,296	0.32%	26,785,719	0.22%	0	0.00%	36,957,859	0.26%	90,374,874	0.26%
33928	36,881,807	0.45%	0	0.00%	0	0.00%	0	0.00%	36,882,064	0.10%
34212	0	0.00%	0	0.00%	0	0.00%	1,461,116	0.01%	1,461,174	0.00%
32514	0	0.00%	0	0.00%	36,878,257	5.18%	0	0.00%	36,878,257	0.10%
34637	0	0.00%	5,274,021	0.04%	0	0.00%	5,294,823	0.04%	10,568,844	0.03%
32656	0	0.00%	0	0.00%	0	0.00%	6,871,076	0.05%	6,871,076	0.02%
33647	0	0.00%	50,959,124	0.42%	0	0.00%	74,033,409	0.53%	124,992,533	0.36%
32798	0	0.00%	0	0.00%	0	0.00%	4,201,160	0.03%	4,233,664	0.01%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34638	0	0.00%	13,316,096	0.11%	0	0.00%	13,369,377	0.10%	26,685,474	0.08%
32940	3,835,492	0.05%	92,636,388	0.76%	0	0.00%	122,677,007	0.87%	219,148,886	0.62%
33931	55,872,793	0.68%	12,611,107	0.10%	0	0.00%	0	0.00%	68,484,177	0.19%
32233	0	0.00%	3,102,963	0.03%	0	0.00%	18,038,220	0.13%	21,141,183	0.06%
34639	0	0.00%	23,741,922	0.20%	0	0.00%	23,834,868	0.17%	47,576,789	0.14%
33932	0	0.00%	0	0.00%	0	0.00%	0	0.00%	575,843	0.00%
34215	0	0.00%	1,622,815	0.01%	0	0.00%	1,107,713	0.01%	2,730,528	0.01%
34498	0	0.00%	726,095	0.01%	0	0.00%	0	0.00%	738,009	0.00%
34216	0	0.00%	9,394,076	0.08%	0	0.00%	0	0.00%	9,578,370	0.03%
32801	18,613,662	0.23%	8,073,946	0.07%	0	0.00%	15,709,160	0.11%	42,396,768	0.12%
34217	0	0.00%	33,003,212	0.27%	0	0.00%	4,388,696	0.03%	37,392,047	0.11%
32802	0	0.00%	0	0.00%	0	0.00%	0	0.00%	948,707	0.00%
32095	0	0.00%	1,421,867	0.01%	0	0.00%	9,446,474	0.07%	10,868,341	0.03%
33510	0	0.00%	592,418	0.00%	0	0.00%	17,549,244	0.12%	18,141,670	0.05%
33935	0	0.00%	0	0.00%	0	0.00%	7,498,001	0.05%	7,659,279	0.02%
34218	0	0.00%	721,937	0.01%	0	0.00%	0	0.00%	940,682	0.00%
32803	58,703,749	0.72%	20,138,854	0.17%	0	0.00%	48,644,933	0.35%	127,487,537	0.36%
33511	0	0.00%	1,991,452	0.02%	0	0.00%	34,069,629	0.24%	36,061,118	0.10%
34785	0	0.00%	9,448,994	0.08%	0	0.00%	12,854,797	0.09%	22,303,790	0.06%
33936	22,552,115	0.27%	0	0.00%	0	0.00%	0	0.00%	22,552,149	0.06%
32804	51,259,165	0.62%	21,046,801	0.17%	0	0.00%	51,744,958	0.37%	124,050,925	0.35%
34219	0	0.00%	0	0.00%	0	0.00%	17,119,661	0.12%	17,119,661	0.05%
32805	19,105,297	0.23%	7,542,841	0.06%	0	0.00%	19,292,319	0.14%	45,940,457	0.13%
34786	76,316,479	0.93%	55,055,916	0.45%	0	0.00%	96,939,439	0.69%	228,311,834	0.65%
33513	0	0.00%	7,938,114	0.07%	0	0.00%	7,969,873	0.06%	15,907,988	0.05%
32664	0	0.00%	589,316	0.00%	0	0.00%	591,684	0.00%	1,181,000	0.00%
33938	0	0.00%	0	0.00%	0	0.00%	0	0.00%	586,329	0.00%
34787	1,912,446	0.02%	1,923,225	0.02%	0	0.00%	63,460,660	0.45%	67,296,331	0.19%
34221	0	0.00%	2,801,945	0.02%	0	0.00%	2,790,450	0.02%	5,592,466	0.02%
32806	76,006,618	0.93%	25,313,863	0.21%	0	0.00%	62,887,297	0.45%	164,207,778	0.47%
33514	0	0.00%	1,032,344	0.01%	0	0.00%	1,397,796	0.01%	2,430,140	0.01%
32948	0	0.00%	4,005,239	0.03%	0	0.00%	6,224,414	0.04%	10,229,653	0.03%
34788	0	0.00%	12,276,261	0.10%	0	0.00%	21,464,484	0.15%	33,740,748	0.10%
32807	54,279,994	0.66%	16,051,949	0.13%	0	0.00%	39,379,557	0.28%	109,711,501	0.31%
32949	0	0.00%	7,791,485	0.06%	0	0.00%	9,800,988	0.07%	17,592,472	0.05%
34223	52,522,965	0.64%	3,824,409	0.03%	0	0.00%	0	0.00%	56,347,510	0.16%
32808	36,565,827	0.45%	1,686,232	0.01%	0	0.00%	51,392,260	0.37%	89,644,319	0.26%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32667	0	0.00%	2,338,715	0.02%	0	0.00%	2,348,131	0.02%	4,686,846	0.01%
32950	0	0.00%	13,834,881	0.11%	0	0.00%	17,501,278	0.12%	31,336,159	0.09%
34224	45,995,741	0.56%	0	0.00%	0	0.00%	2,335,654	0.02%	48,331,395	0.14%
32526	0	0.00%	0	0.00%	42,275,705	5.94%	0	0.00%	42,275,705	0.12%
32809	48,146,755	0.59%	14,699,433	0.12%	0	0.00%	34,995,169	0.25%	97,841,357	0.28%
32951	0	0.00%	63,684,783	0.52%	0	0.00%	124,713,308	0.89%	188,398,523	0.54%
32668	0	0.00%	3,409,333	0.03%	0	0.00%	2,864,289	0.02%	6,273,622	0.02%
32102	0	0.00%	0	0.00%	0	0.00%	1,797,844	0.01%	1,797,844	0.01%
32810	40,495,890	0.49%	1,184,918	0.01%	0	0.00%	40,882,410	0.29%	82,563,218	0.23%
33801	0	0.00%	3,214,530	0.03%	0	0.00%	38,885,245	0.28%	42,100,112	0.12%
32952	22,279,373	0.27%	74,556,597	0.61%	0	0.00%	87,285,464	0.62%	184,121,434	0.52%
32669	0	0.00%	7,490,225	0.06%	0	0.00%	7,518,632	0.05%	15,008,857	0.04%
32811	28,066,025	0.34%	11,517,167	0.09%	0	0.00%	23,584,953	0.17%	63,168,145	0.18%
33802	0	0.00%	644,714	0.01%	0	0.00%	1,107,716	0.01%	1,752,469	0.01%
32953	20,722,228	0.25%	48,329,669	0.40%	0	0.00%	67,412,773	0.48%	136,464,669	0.39%
32812	78,237,267	0.95%	22,924,735	0.19%	0	0.00%	56,712,688	0.40%	157,874,690	0.45%
34652	0	0.00%	18,130,369	0.15%	0	0.00%	30,089,823	0.21%	48,220,193	0.14%
33803	0	0.00%	2,076,330	0.02%	0	0.00%	36,716,190	0.26%	38,792,641	0.11%
33945	904,454	0.01%	0	0.00%	0	0.00%	0	0.00%	904,454	0.00%
34228	0	0.00%	82,620,241	0.68%	0	0.00%	23,304,013	0.17%	105,925,224	0.30%
34653	0	0.00%	18,575,602	0.15%	0	0.00%	18,634,764	0.13%	37,210,366	0.11%
32955	32,743,917	0.40%	67,122,572	0.55%	0	0.00%	92,674,119	0.66%	192,540,609	0.55%
33521	0	0.00%	562,404	0.00%	0	0.00%	784,114	0.01%	1,346,518	0.00%
33946	13,643,738	0.17%	6,739,871	0.06%	0	0.00%	6,754,090	0.05%	27,137,699	0.08%
34229	0	0.00%	15,734,461	0.13%	0	0.00%	0	0.00%	15,734,546	0.04%
32814	8,883,770	0.11%	5,621,636	0.05%	0	0.00%	6,844,234	0.05%	21,349,639	0.06%
34654	0	0.00%	24,726,579	0.20%	0	0.00%	15,736,458	0.11%	40,463,036	0.12%
33805	0	0.00%	860,535	0.01%	0	0.00%	19,125,627	0.14%	19,986,242	0.06%
33947	29,839,883	0.36%	0	0.00%	0	0.00%	944,967	0.01%	30,784,849	0.09%
34655	0	0.00%	39,945,064	0.33%	0	0.00%	40,075,460	0.28%	80,020,524	0.23%
33523	0	0.00%	7,798,833	0.06%	0	0.00%	13,100,452	0.09%	20,899,285	0.06%
32957	0	0.00%	560,966	0.00%	0	0.00%	843,245	0.01%	1,404,211	0.00%
33948	62,420,730	0.76%	0	0.00%	0	0.00%	0	0.00%	62,420,922	0.18%
32250	0	0.00%	7,174,777	0.06%	0	0.00%	27,523,550	0.20%	34,698,327	0.10%
34231	0	0.00%	37,838,315	0.31%	0	0.00%	0	0.00%	37,839,903	0.11%
32533	0	0.00%	0	0.00%	30,556,458	4.29%	0	0.00%	30,556,458	0.09%
34797	0	0.00%	1,724,759	0.01%	0	0.00%	4,589,598	0.03%	6,314,357	0.02%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32958	0	0.00%	73,014,806	0.60%	0	0.00%	96,605,686	0.69%	169,620,525	0.48%
32817	73,177,567	0.89%	21,618,702	0.18%	0	0.00%	53,048,280	0.38%	147,844,549	0.42%
32534	0	0.00%	0	0.00%	11,554,312	1.62%	0	0.00%	11,554,312	0.03%
33525	0	0.00%	11,722,843	0.10%	0	0.00%	18,252,257	0.13%	29,975,106	0.09%
32818	43,078,946	0.52%	1,610,642	0.01%	0	0.00%	59,596,378	0.42%	104,285,966	0.30%
33950	207,603,886	2.53%	0	0.00%	0	0.00%	7,141,905	0.05%	214,746,112	0.61%
32535	0	0.00%	0	0.00%	2,437,717	0.34%	0	0.00%	2,437,717	0.01%
33809	0	0.00%	19,191,122	0.16%	0	0.00%	44,774,250	0.32%	63,965,375	0.18%
32960	0	0.00%	58,454,409	0.48%	0	0.00%	71,289,661	0.51%	129,744,240	0.37%
32819	79,771,931	0.97%	34,911,102	0.29%	0	0.00%	80,496,937	0.57%	195,179,970	0.56%
34234	0	0.00%	1,403,056	0.01%	0	0.00%	0	0.00%	1,403,298	0.00%
33810	0	0.00%	25,224,314	0.21%	0	0.00%	52,297,189	0.37%	77,521,503	0.22%
32961	0	0.00%	0	0.00%	0	0.00%	572,952	0.00%	1,051,443	0.00%
32112	0	0.00%	0	0.00%	0	0.00%	2,384,449	0.02%	2,464,195	0.01%
33527	0	0.00%	7,162,299	0.06%	0	0.00%	11,463,620	0.08%	18,625,919	0.05%
33952	117,324,714	1.43%	0	0.00%	0	0.00%	0	0.00%	117,324,714	0.33%
32820	12,234,218	0.15%	6,067,853	0.05%	0	0.00%	9,091,864	0.06%	27,393,935	0.08%
33811	0	0.00%	0	0.00%	0	0.00%	22,521,426	0.16%	22,780,844	0.06%
32962	0	0.00%	62,770,181	0.52%	0	0.00%	74,921,446	0.53%	137,691,756	0.39%
32113	0	0.00%	2,768,525	0.02%	0	0.00%	4,479,966	0.03%	7,248,491	0.02%
33953	16,830,833	0.21%	0	0.00%	0	0.00%	812,316	0.01%	17,643,149	0.05%
32821	32,654,308	0.40%	1,277,284	0.01%	0	0.00%	23,738,947	0.17%	57,670,539	0.16%
34236	0	0.00%	29,850,381	0.25%	0	0.00%	0	0.00%	29,851,457	0.08%
32963	0	0.00%	212,647,956	1.75%	0	0.00%	512,745,723	3.64%	725,394,917	2.06%
32114	36,200,366	0.44%	22,224,217	0.18%	0	0.00%	36,476,696	0.26%	94,901,278	0.27%
32680	0	0.00%	2,876,231	0.02%	0	0.00%	0	0.00%	2,919,053	0.01%
33812	0	0.00%	0	0.00%	0	0.00%	9,353,719	0.07%	9,437,843	0.03%
33954	38,976,639	0.47%	0	0.00%	0	0.00%	0	0.00%	39,126,334	0.11%
32822	69,765,318	0.85%	22,570,265	0.19%	0	0.00%	50,386,385	0.36%	142,721,969	0.41%
33813	0	0.00%	1,245,168	0.01%	0	0.00%	86,079,832	0.61%	87,325,058	0.25%
32964	0	0.00%	553,470	0.00%	0	0.00%	665,863	0.00%	1,219,335	0.00%
32681	0	0.00%	0	0.00%	0	0.00%	0	0.00%	549,045	0.00%
34945	0	0.00%	8,828,136	0.07%	0	0.00%	11,670,401	0.08%	20,498,536	0.06%
33955	63,365,328	0.77%	0	0.00%	0	0.00%	0	0.00%	63,365,384	0.18%
34238	0	0.00%	5,794,033	0.05%	0	0.00%	0	0.00%	5,794,767	0.02%
32116	599,082	0.01%	764,536	0.01%	0	0.00%	840,958	0.01%	2,204,576	0.01%
34946	0	0.00%	11,143,154	0.09%	0	0.00%	14,444,705	0.10%	25,587,859	0.07%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32824	68,572,643	0.84%	23,549,744	0.19%	0	0.00%	55,947,481	0.40%	148,069,868	0.42%
33956	44,882,899	0.55%	0	0.00%	0	0.00%	0	0.00%	45,013,731	0.13%
34239	0	0.00%	1,392,903	0.01%	0	0.00%	0	0.00%	1,392,970	0.00%
32541	0	0.00%	0	0.00%	44,336,611	6.23%	0	0.00%	44,336,611	0.13%
33815	0	0.00%	0	0.00%	0	0.00%	6,858,723	0.05%	7,203,527	0.02%
32966	0	0.00%	38,736,664	0.32%	0	0.00%	61,159,721	0.43%	99,896,482	0.28%
32117	37,918,760	0.46%	22,727,965	0.19%	0	0.00%	31,391,630	0.22%	92,038,355	0.26%
34947	0	0.00%	31,226,518	0.26%	0	0.00%	39,288,059	0.28%	70,514,925	0.20%
32825	111,944,914	1.36%	36,891,530	0.30%	0	0.00%	81,689,152	0.58%	230,525,595	0.66%
33957	170,982,436	2.08%	9,449,321	0.08%	0	0.00%	0	0.00%	180,431,891	0.51%
32967	0	0.00%	43,779,227	0.36%	0	0.00%	53,597,342	0.38%	97,376,858	0.28%
32118	105,551,668	1.29%	79,945,070	0.66%	0	0.00%	88,385,448	0.63%	273,882,186	0.78%
32826	39,477,250	0.48%	13,297,061	0.11%	0	0.00%	28,365,711	0.20%	81,140,022	0.23%
32119	61,999,946	0.76%	44,446,550	0.37%	0	0.00%	54,385,282	0.39%	160,831,778	0.46%
34949	0	0.00%	67,176,324	0.55%	0	0.00%	94,023,578	0.67%	161,200,789	0.46%
33534	0	0.00%	0	0.00%	0	0.00%	5,498,509	0.04%	5,586,958	0.02%
32968	0	0.00%	34,998,479	0.29%	0	0.00%	38,828,707	0.28%	73,827,186	0.21%
34242	0	0.00%	49,816,772	0.41%	0	0.00%	0	0.00%	49,819,662	0.14%
32827	17,379,709	0.21%	6,926,442	0.06%	0	0.00%	14,216,305	0.10%	38,522,455	0.11%
Total	8,196,278,854		12,127,201,176		709,868,227		14,048,095,558		35,114,010,292	

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34950	0	0.00%	16,513,654	0.14%	0	0	19,432,405	0.15%	35,946,125	0.11%
34667	0	0.00%	37,383,021	0.31%	0	0	48,773,572	0.37%	86,156,592	0.26%
32686	0	0.00%	3,216,067	0.03%	0	0	4,226,372	0.03%	7,442,439	0.02%
33960	0	0.00%	560,548	0.00%	0	0	716,087	0.01%	1,665,206	0.01%
32828	106,111,331	1.49%	49,572,860	0.41%	0	0	79,468,081	0.61%	235,152,272	0.71%
34102	0	0.00%	39,618,305	0.33%	0	0	0	0.00%	39,622,031	0.12%
34951	0	0.00%	36,811,804	0.30%	0	0	47,019,896	0.36%	83,831,710	0.25%
34668	0	0.00%	27,977,735	0.23%	0	0	43,035,052	0.33%	71,012,787	0.22%
32829	31,245,236	0.44%	14,754,462	0.12%	0	0	21,755,182	0.17%	67,754,880	0.21%
34103	0	0.00%	41,302,730	0.34%	0	0	18,667,919	0.14%	59,972,694	0.18%
34952	0	0.00%	86,330,327	0.71%	0	0	96,529,657	0.74%	182,860,238	0.55%
33820	0	0.00%	0	0.00%	0	0	0	0.00%	739,164	0.00%
34669	0	0.00%	11,270,665	0.09%	0	0	8,040,821	0.06%	19,311,487	0.06%
32547	0	0.00%	0	0.00%	21,280,770	0.03181	0	0.00%	21,280,770	0.06%
34953	0	0.00%	105,079,371	0.87%	0	0	113,645,474	0.87%	218,724,846	0.66%
33538	0	0.00%	5,127,780	0.04%	0	0	5,141,823	0.04%	10,269,603	0.03%
32548	0	0.00%	0	0.00%	22,469,343	0.03359	0	0.00%	22,469,343	0.07%
32407	0	0.00%	0	0.00%	2,339,335	0.0035	0	0.00%	2,339,423	0.01%
34105	0	0.00%	7,485,147	0.06%	0	0	0	0.00%	7,488,537	0.02%
32124	4,470,008	0.06%	3,761,227	0.03%	0	0	4,519,995	0.03%	12,751,231	0.04%
32266	0	0.00%	7,199,999	0.06%	0	0	7,220,070	0.06%	14,420,069	0.04%
32832	31,693,063	0.45%	19,291,412	0.16%	0	0	25,911,598	0.20%	76,896,073	0.23%
32408	0	0.00%	0	0.00%	2,710,017	0.00405	0	0.00%	2,710,296	0.01%
33823	17,201,007	0.24%	0	0.00%	0	0	33,132,546	0.25%	50,674,949	0.15%
33540	0	0.00%	4,057,600	0.03%	0	0	6,415,377	0.05%	10,472,977	0.03%
32550	0	0.00%	0	0.00%	5,341,208	0.00799	0	0.00%	5,341,208	0.02%
32833	17,248,986	0.24%	7,999,107	0.07%	0	0	12,851,899	0.10%	38,099,992	0.12%
34956	0	0.00%	5,538,263	0.05%	0	0	5,553,380	0.04%	11,091,643	0.03%
32692	0	0.00%	585,324	0.00%	0	0	0	0.00%	585,324	0.00%
33541	0	0.00%	10,268,256	0.08%	0	0	15,376,250	0.12%	25,644,519	0.08%
33966	17,334,159	0.24%	0	0.00%	0	0	0	0.00%	17,334,307	0.05%
34108	0	0.00%	56,470,538	0.47%	0	0	0	0.00%	56,480,734	0.17%
34957	0	0.00%	73,474,741	0.61%	0	0	73,424,399	0.56%	146,899,253	0.45%
32127	56,249,845	0.79%	56,699,362	0.47%	0	0	67,989,948	0.52%	180,939,154	0.55%
32976	0	0.00%	26,264,666	0.22%	0	0	77,748,894	0.59%	104,013,562	0.32%
33825	43,520,647	0.61%	13,253,629	0.11%	0	0	33,419,437	0.26%	90,193,714	0.27%
32693	0	0.00%	4,692,981	0.04%	0	0	0	0.00%	4,763,158	0.01%
33542	0	0.00%	9,802,490	0.08%	0	0	15,724,709	0.12%	25,527,199	0.08%
33401	0	0.00%	68,999,085	0.57%	0	0	51,929,375	0.40%	120,928,699	0.37%
33967	26,055,851	0.37%	0	0.00%	0	0	0	0.00%	26,055,851	0.08%
32835	52,903,121	0.74%	27,505,452	0.23%	0	0	53,433,321	0.41%	133,841,894	0.41%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
34109	0	0.00%	7,603,244	0.06%	0	0	0	0.00%	7,608,601	0.02%
32128	43,675,324	0.61%	21,855,861	0.18%	0	0	37,165,019	0.28%	102,696,204	0.31%
32694	0	0.00%	0	0.00%	0	0	747,391	0.01%	747,391	0.00%
33543	0	0.00%	22,204,107	0.18%	0	0	29,620,125	0.23%	51,824,232	0.16%
34251	0	0.00%	0	0.00%	0	0	5,120,763	0.04%	5,120,763	0.02%
32836	50,813,484	0.71%	28,533,729	0.24%	0	0	61,374,402	0.47%	140,721,615	0.43%
34110	12,751,792	0.18%	12,836,309	0.11%	0	0	0	0.00%	25,589,500	0.08%
32129	32,603,281	0.46%	21,148,912	0.17%	0	0	27,316,938	0.21%	81,069,131	0.25%
33827	8,678,410	0.12%	3,865,752	0.03%	0	0	6,133,883	0.05%	18,678,044	0.06%
33544	0	0.00%	20,066,163	0.17%	0	0	25,405,825	0.19%	45,471,988	0.14%
33403	0	0.00%	19,551,300	0.16%	0	0	14,355,564	0.11%	33,906,892	0.10%
32837	97,700,896	1.37%	36,334,096	0.30%	0	0	72,093,194	0.55%	206,128,186	0.63%
32413	0	0.00%	0	0.00%	26,186,564	0.03915	0	0.00%	26,186,761	0.08%
32130	0	0.00%	0	0.00%	0	0	3,158,840	0.02%	3,217,579	0.01%
34677	0	0.00%	22,447,124	0.19%	0	0	3,545,411	0.03%	25,992,535	0.08%
32696	0	0.00%	5,786,316	0.05%	0	0	5,154,476	0.04%	10,940,792	0.03%
33545	0	0.00%	9,875,026	0.08%	0	0	11,825,938	0.09%	21,700,965	0.07%
33404	0	0.00%	68,058,428	0.56%	0	0	50,733,772	0.39%	118,792,686	0.36%
33971	18,224,932	0.26%	0	0.00%	0	0	0	0.00%	18,224,992	0.06%
32839	27,666,007	0.39%	12,120,943	0.10%	0	0	23,285,099	0.18%	63,072,049	0.19%
33405	0	0.00%	37,110,040	0.31%	0	0	19,772,284	0.15%	56,882,331	0.17%
32132	10,837,241	0.15%	10,927,867	0.09%	0	0	13,180,615	0.10%	34,945,724	0.11%
33830	21,359,694	0.30%	580,232	0.00%	0	0	28,304,610	0.22%	50,244,536	0.15%
33547	0	0.00%	0	0.00%	0	0	21,251,734	0.16%	21,570,755	0.07%
33972	8,806,302	0.12%	0	0.00%	0	0	0	0.00%	9,158,813	0.03%
33406	0	0.00%	36,858,737	0.30%	0	0	22,886,794	0.17%	59,745,571	0.18%
33548	0	0.00%	7,428,793	0.06%	0	0	7,449,636	0.06%	14,878,429	0.05%
33973	4,128,441	0.06%	0	0.00%	0	0	0	0.00%	4,128,441	0.01%
33407	0	0.00%	41,268,659	0.34%	0	0	30,280,770	0.23%	71,550,062	0.22%
32134	0	0.00%	0	0.00%	0	0	3,397,841	0.03%	3,465,188	0.01%
34681	0	0.00%	2,013,369	0.02%	0	0	2,019,024	0.02%	4,032,393	0.01%
33549	0	0.00%	16,494,070	0.14%	0	0	16,537,402	0.13%	33,031,472	0.10%
33974	7,680,365	0.11%	0	0.00%	0	0	0	0.00%	7,680,365	0.02%
33408	0	0.00%	99,479,706	0.82%	0	0	79,779,374	0.61%	179,259,612	0.54%
32701	29,566,051	0.42%	14,121,483	0.12%	0	0	22,445,559	0.17%	66,133,093	0.20%
33409	0	0.00%	36,911,324	0.31%	0	0	24,704,455	0.19%	61,615,944	0.19%
32136	22,314,108	0.31%	18,796,183	0.16%	0	0	22,553,223	0.17%	63,663,514	0.19%
33834	4,638,533	0.07%	0	0.00%	0	0	2,980,677	0.02%	7,646,143	0.02%
34683	0	0.00%	36,631,929	0.30%	0	0	36,698,758	0.28%	73,330,686	0.22%
32702	0	0.00%	0	0.00%	0	0	1,233,093	0.01%	1,233,093	0.00%
33976	7,270,041	0.10%	0	0.00%	0	0	0	0.00%	7,270,041	0.02%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32561	0	0.00%	0	0.00%	38,362,445	0.05735	0	0.00%	38,362,445	0.12%
33410	0	0.00%	114,101,330	0.94%	0	0	100,533,535	0.77%	214,634,998	0.65%
32137	0	0.00%	60,559,883	0.50%	0	0	80,245,087	0.61%	140,805,688	0.43%
34684	0	0.00%	24,885,511	0.21%	0	0	5,256,182	0.04%	30,141,693	0.09%
32703	26,129,713	0.37%	638,866	0.01%	0	0	39,956,336	0.31%	66,724,916	0.20%
33411	0	0.00%	138,573,673	1.15%	0	0	93,029,911	0.71%	231,603,831	0.70%
34685	0	0.00%	22,689,368	0.19%	0	0	22,731,457	0.17%	45,420,825	0.14%
32563	0	0.00%	0	0.00%	33,058,579	0.04942	0	0.00%	33,058,579	0.10%
33412	0	0.00%	53,881,687	0.45%	0	0	47,704,786	0.36%	101,586,495	0.31%
33837	34,449,475	0.48%	17,743,535	0.15%	0	0	29,128,739	0.22%	81,321,749	0.25%
32564	0	0.00%	0	0.00%	1,059,033	0.00158	0	0.00%	1,059,033	0.00%
33413	0	0.00%	22,097,653	0.18%	0	0	14,700,857	0.11%	36,798,526	0.11%
33838	5,345,818	0.08%	1,826,831	0.02%	0	0	4,158,017	0.03%	11,330,667	0.03%
33980	43,344,693	0.61%	0	0.00%	0	0	0	0.00%	43,345,165	0.13%
32565	0	0.00%	0	0.00%	4,034,964	0.00603	0	0.00%	4,034,964	0.01%
33414	0	0.00%	134,567,052	1.11%	0	0	104,763,662	0.80%	239,330,822	0.73%
32141	27,362,399	0.38%	27,593,007	0.23%	0	0	33,360,537	0.25%	88,315,943	0.27%
33839	3,361,629	0.05%	1,814,167	0.02%	0	0	3,399,408	0.03%	8,575,205	0.03%
34688	0	0.00%	13,076,944	0.11%	0	0	13,099,826	0.10%	26,176,770	0.08%
32707	66,596,058	0.94%	25,882,967	0.21%	0	0	55,528,274	0.42%	148,007,299	0.45%
33556	0	0.00%	32,240,036	0.27%	0	0	32,326,612	0.25%	64,566,648	0.20%
33981	24,852,622	0.35%	0	0.00%	0	0	0	0.00%	24,852,653	0.08%
32566	0	0.00%	0	0.00%	42,480,895	0.06351	0	0.00%	42,480,895	0.13%
33415	0	0.00%	43,756,783	0.36%	0	0	33,785,785	0.26%	77,542,773	0.24%
34972	0	0.00%	12,652,098	0.10%	0	0	17,784,955	0.14%	30,437,052	0.09%
34689	0	0.00%	24,023,488	0.20%	0	0	24,050,101	0.18%	48,073,589	0.15%
32708	109,097,974	1.53%	41,534,382	0.34%	0	0	90,938,673	0.69%	241,571,029	0.73%
33982	40,324,046	0.57%	0	0.00%	0	0	7,170,949	0.05%	47,494,995	0.14%
33841	10,713,199	0.15%	0	0.00%	0	0	6,886,446	0.05%	17,658,012	0.05%
34690	0	0.00%	7,491,906	0.06%	0	0	7,513,155	0.06%	15,005,061	0.05%
32709	3,025,121	0.04%	1,273,827	0.01%	0	0	2,520,086	0.02%	6,819,033	0.02%
33558	0	0.00%	31,645,139	0.26%	0	0	21,878,197	0.17%	53,523,335	0.16%
33983	68,733,582	0.97%	0	0.00%	0	0	1,756,319	0.01%	70,490,124	0.21%
32568	0	0.00%	0	0.00%	2,586,084	0.00387	0	0.00%	2,586,084	0.01%
33417	0	0.00%	49,749,602	0.41%	0	0	32,957,704	0.25%	82,707,523	0.25%
34266	60,585,952	0.85%	0	0.00%	0	0	16,089,839	0.12%	76,675,791	0.23%
34974	0	0.00%	34,420,035	0.28%	0	0	40,948,592	0.31%	75,764,250	0.23%
34691	0	0.00%	12,899,281	0.11%	0	0	18,912,159	0.14%	31,811,440	0.10%
33559	0	0.00%	10,390,838	0.09%	0	0	10,416,323	0.08%	20,807,161	0.06%
33701	0	0.00%	13,528,400	0.11%	0	0	0	0.00%	13,529,183	0.04%
32569	0	0.00%	0	0.00%	10,651,400	0.01592	0	0.00%	10,651,400	0.03%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33418	0	0.00%	192,967,988	1.60%	0	0	156,929,930	1.20%	349,898,083	1.06%
33843	20,643,537	0.29%	4,457,472	0.04%	0	0	11,196,356	0.09%	36,297,365	0.11%
33702	0	0.00%	22,132,667	0.18%	0	0	0	0.00%	22,134,329	0.07%
32570	0	0.00%	0	0.00%	20,533,273	0.0307	0	0.00%	20,533,273	0.06%
33844	43,021,550	0.60%	18,384,929	0.15%	0	0	43,501,509	0.33%	104,907,987	0.32%
32712	958,627	0.01%	966,043	0.01%	0	0	56,669,205	0.43%	58,593,876	0.18%
33703	0	0.00%	23,417,758	0.19%	0	0	0	0.00%	23,418,040	0.07%
32571	0	0.00%	0	0.00%	28,937,273	0.04326	0	0.00%	28,937,273	0.09%
34269	20,289,424	0.28%	0	0.00%	0	0	4,110,746	0.03%	24,400,170	0.07%
32713	28,542,974	0.40%	0	0.00%	0	0	28,862,221	0.22%	57,724,547	0.18%
33704	0	0.00%	18,661,244	0.15%	0	0	0	0.00%	18,662,390	0.06%
32148	0	0.00%	0	0.00%	0	0	3,244,820	0.02%	3,244,820	0.01%
34695	0	0.00%	18,660,092	0.15%	0	0	0	0.00%	18,661,547	0.06%
32714	36,273,852	0.51%	3,506,771	0.03%	0	0	36,636,208	0.28%	76,416,831	0.23%
33563	0	0.00%	0	0.00%	0	0	15,900,700	0.12%	16,319,313	0.05%
33705	0	0.00%	15,452,282	0.13%	0	0	0	0.00%	15,452,527	0.05%
33847	0	0.00%	0	0.00%	0	0	0	0.00%	587,565	0.00%
33706	0	0.00%	64,867,659	0.54%	0	0	0	0.00%	64,869,318	0.20%
33848	0	0.00%	0	0.00%	0	0	0	0.00%	1,013,489	0.00%
33565	0	0.00%	9,265,064	0.08%	0	0	19,092,926	0.15%	28,357,990	0.09%
33990	71,574,504	1.01%	0	0.00%	0	0	0	0.00%	71,574,508	0.22%
33707	0	0.00%	45,079,928	0.37%	0	0	0	0.00%	45,081,535	0.14%
33849	0	0.00%	0	0.00%	0	0	638,998	0.00%	974,266	0.00%
34698	0	0.00%	36,885,215	0.30%	0	0	8,381,977	0.06%	45,267,192	0.14%
33566	0	0.00%	12,575,596	0.10%	0	0	20,113,871	0.15%	32,689,472	0.10%
34981	0	0.00%	7,236,977	0.06%	0	0	7,256,851	0.06%	14,493,828	0.04%
33991	47,350,366	0.67%	0	0.00%	0	0	0	0.00%	47,350,366	0.14%
33708	0	0.00%	64,826,759	0.54%	0	0	0	0.00%	64,828,980	0.20%
33850	6,019,995	0.08%	0	0.00%	0	0	7,985,715	0.06%	14,076,557	0.04%
33567	0	0.00%	5,513,329	0.05%	0	0	7,991,269	0.06%	13,504,599	0.04%
34982	0	0.00%	51,285,920	0.42%	0	0	47,502,550	0.36%	98,788,696	0.30%
33709	0	0.00%	16,739,327	0.14%	0	0	0	0.00%	16,739,800	0.05%
32577	0	0.00%	0	0.00%	5,352,596	0.008	0	0.00%	5,352,596	0.02%
33426	0	0.00%	30,831,766	0.25%	0	0	15,604,750	0.12%	46,436,576	0.14%
34275	0	0.00%	2,026,207	0.02%	0	0	0	0.00%	2,026,607	0.01%
33851	1,505,449	0.02%	563,918	0.00%	0	0	1,333,385	0.01%	3,402,752	0.01%
34134	56,173,512	0.79%	56,604,909	0.47%	0	0	0	0.00%	112,778,529	0.34%
34983	0	0.00%	81,242,192	0.67%	0	0	87,988,436	0.67%	169,230,633	0.51%
33993	42,707,048	0.60%	0	0.00%	0	0	0	0.00%	42,707,048	0.13%
33710	0	0.00%	25,029,047	0.21%	0	0	0	0.00%	25,029,099	0.08%
32578	0	0.00%	0	0.00%	1,594,716	0.00238	0	0.00%	1,594,716	0.00%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33852	29,886,829	0.42%	30,137,485	0.25%	0	0	47,524,083	0.36%	107,548,397	0.33%
32720	922,932	0.01%	929,637	0.01%	0	0	21,409,183	0.16%	23,261,752	0.07%
33569	0	0.00%	565,474	0.00%	0	0	26,522,233	0.20%	27,087,714	0.08%
34135	10,319,108	0.14%	0	0.00%	0	0	0	0.00%	10,319,649	0.03%
34984	0	0.00%	35,097,860	0.29%	0	0	37,991,239	0.29%	73,089,099	0.22%
33711	0	0.00%	13,140,882	0.11%	0	0	0	0.00%	13,141,260	0.04%
32579	0	0.00%	0	0.00%	10,689,147	0.01598	0	0.00%	10,689,147	0.03%
33428	0	0.00%	38,063,630	0.31%	0	0	0	0.00%	38,064,224	0.12%
33853	23,714,383	0.33%	6,504,100	0.05%	0	0	16,260,245	0.12%	46,478,728	0.14%
33570	0	0.00%	0	0.00%	0	0	11,734,482	0.09%	12,005,837	0.04%
33712	0	0.00%	11,493,046	0.10%	0	0	0	0.00%	11,493,214	0.03%
32580	0	0.00%	0	0.00%	2,801,931	0.00419	0	0.00%	2,801,931	0.01%
33854	932,579	0.01%	539,437	0.00%	0	0	744,540	0.01%	2,216,556	0.01%
34420	0	0.00%	8,708,098	0.07%	0	0	11,344,847	0.09%	20,052,944	0.06%
34986	0	0.00%	64,629,104	0.53%	0	0	70,196,615	0.54%	134,825,977	0.41%
33713	0	0.00%	17,198,829	0.14%	0	0	0	0.00%	17,198,892	0.05%
33430	0	0.00%	5,599,679	0.05%	0	0	5,610,477	0.04%	11,210,166	0.03%
33855	2,637,471	0.04%	1,676,840	0.01%	0	0	2,667,182	0.02%	6,981,493	0.02%
33572	0	0.00%	936,153	0.01%	0	0	21,054,745	0.16%	21,990,898	0.07%
34987	0	0.00%	12,968,380	0.11%	0	0	14,040,157	0.11%	27,008,550	0.08%
33714	0	0.00%	8,088,613	0.07%	0	0	0	0.00%	8,088,879	0.02%
33431	0	0.00%	27,711,900	0.23%	0	0	0	0.00%	27,713,801	0.08%
34705	0	0.00%	1,398,148	0.01%	0	0	2,054,641	0.02%	3,452,789	0.01%
32724	24,197,974	0.34%	996,120	0.01%	0	0	24,467,306	0.19%	49,661,400	0.15%
33573	0	0.00%	9,030,887	0.07%	0	0	29,644,179	0.23%	38,675,128	0.12%
33715	0	0.00%	37,121,291	0.31%	0	0	15,620,063	0.12%	52,741,353	0.16%
32583	0	0.00%	0	0.00%	15,560,880	0.02326	0	0.00%	15,560,880	0.05%
33432	0	0.00%	47,496,328	0.39%	0	0	0	0.00%	47,500,439	0.14%
32159	0	0.00%	31,699,138	0.26%	0	0	50,371,345	0.38%	82,070,483	0.25%
33857	988,221	0.01%	2,071,737	0.02%	0	0	2,967,150	0.02%	6,027,109	0.02%
32725	52,340,646	0.74%	788,532	0.01%	0	0	52,924,116	0.40%	106,053,294	0.32%
33716	0	0.00%	5,005,905	0.04%	0	0	0	0.00%	5,006,505	0.02%
33433	0	0.00%	55,043,543	0.46%	0	0	0	0.00%	55,045,587	0.17%
32301	0	0.00%	14,246,931	0.12%	0	0	0	0.00%	14,246,931	0.04%
32726	0	0.00%	14,211,320	0.12%	0	0	21,486,063	0.16%	35,697,382	0.11%
34990	0	0.00%	123,219,773	1.02%	0	0	123,471,695	0.94%	246,691,573	0.75%
33434	0	0.00%	40,055,545	0.33%	0	0	0	0.00%	40,057,859	0.12%
33859	23,818,576	0.33%	6,252,245	0.05%	0	0	12,900,196	0.10%	42,971,017	0.13%
33576	0	0.00%	5,861,857	0.05%	0	0	5,878,142	0.04%	11,739,999	0.04%
33435	0	0.00%	54,090,525	0.45%	0	0	30,082,183	0.23%	84,173,093	0.26%
32303	0	0.00%	28,631,147	0.24%	0	0	0	0.00%	28,631,147	0.09%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33860	0	0.00%	0	0.00%	0	0	16,231,622	0.12%	16,428,904	0.05%
32162	0	0.00%	91,605,138	0.76%	0	0	91,850,349	0.70%	183,455,487	0.56%
33436	0	0.00%	86,959,672	0.72%	0	0	46,139,775	0.35%	133,099,746	0.40%
34285	7,239,028	0.10%	7,286,130	0.06%	0	0	0	0.00%	14,525,810	0.04%
32304	0	0.00%	8,850,033	0.07%	0	0	0	0.00%	8,850,033	0.03%
33578	0	0.00%	4,692,044	0.04%	0	0	22,741,532	0.17%	27,433,603	0.08%
32163	0	0.00%	2,222,008	0.02%	0	0	2,228,119	0.02%	4,450,127	0.01%
33437	0	0.00%	83,926,646	0.69%	0	0	60,527,913	0.46%	144,454,893	0.44%
34286	30,809,054	0.43%	0	0.00%	0	0	0	0.00%	31,014,756	0.09%
32305	0	0.00%	6,360,935	0.05%	0	0	0	0.00%	6,360,935	0.02%
34711	0	0.00%	54,989,528	0.45%	0	0	89,984,859	0.69%	144,974,402	0.44%
32730	7,442,729	0.10%	0	0.00%	0	0	5,548,301	0.04%	13,287,690	0.04%
33579	0	0.00%	645,643	0.01%	0	0	16,225,255	0.12%	16,870,904	0.05%
34428	0	0.00%	12,722,737	0.11%	0	0	6,477,346	0.05%	19,200,084	0.06%
34145	0	0.00%	60,682,812	0.50%	0	0	0	0.00%	60,694,829	0.18%
34994	0	0.00%	40,553,092	0.34%	0	0	40,512,310	0.31%	81,065,466	0.25%
32164	0	0.00%	41,098,016	0.34%	0	0	41,213,226	0.31%	82,311,242	0.25%
33438	0	0.00%	1,124,911	0.01%	0	0	966,917	0.01%	2,091,828	0.01%
34287	30,456,867	0.43%	0	0.00%	0	0	0	0.00%	30,457,613	0.09%
34429	0	0.00%	13,562,135	0.11%	0	0	13,592,007	0.10%	27,154,143	0.08%
32024	0	0.00%	0	0.00%	0	0	9,272,363	0.07%	9,272,363	0.03%
34288	21,232,200	0.30%	0	0.00%	0	0	0	0.00%	21,568,136	0.07%
32732	14,105,884	0.20%	5,108,316	0.04%	0	0	10,268,629	0.08%	29,482,829	0.09%
34996	0	0.00%	64,978,036	0.54%	0	0	60,210,653	0.46%	125,188,816	0.38%
32025	0	0.00%	0	0.00%	0	0	8,758,559	0.07%	8,758,559	0.03%
33440	0	0.00%	9,052,071	0.07%	0	0	9,076,650	0.07%	18,128,721	0.06%
34289	3,178,794	0.04%	0	0.00%	0	0	0	0.00%	3,481,272	0.01%
32308	0	0.00%	18,407,071	0.15%	0	0	0	0.00%	18,407,114	0.06%
34714	1,612,099	0.02%	1,622,908	0.01%	0	0	19,121,657	0.15%	22,356,663	0.07%
34431	0	0.00%	10,454,910	0.09%	0	0	8,668,801	0.07%	19,123,711	0.06%
34997	0	0.00%	118,129,855	0.98%	0	0	107,695,355	0.82%	225,825,303	0.68%
33865	1,182,374	0.02%	0	0.00%	0	0	705,163	0.01%	1,894,731	0.01%
33441	0	0.00%	19,289,844	0.16%	0	0	0	0.00%	19,292,502	0.06%
32309	0	0.00%	29,732,115	0.25%	0	0	0	0.00%	29,732,115	0.09%
34432	0	0.00%	14,097,300	0.12%	0	0	14,135,961	0.11%	28,233,261	0.09%
32168	47,367,131	0.67%	31,947,481	0.26%	0	0	47,885,487	0.37%	127,200,099	0.39%
34715	0	0.00%	11,230,696	0.09%	0	0	18,576,678	0.14%	29,807,374	0.09%
33442	0	0.00%	27,731,167	0.23%	0	0	0	0.00%	27,733,871	0.08%
34291	6,058,001	0.09%	0	0.00%	0	0	0	0.00%	6,058,001	0.02%
32310	0	0.00%	5,173,556	0.04%	0	0	0	0.00%	5,173,556	0.02%
32735	0	0.00%	3,449,819	0.03%	0	0	4,877,141	0.04%	8,326,959	0.03%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33584	0	0.00%	0	0.00%	0	0	16,663,695	0.13%	16,899,705	0.05%
34433	0	0.00%	8,735,334	0.07%	0	0	6,948,080	0.05%	15,683,414	0.05%
32169	49,994,870	0.70%	57,189,731	0.47%	0	0	62,915,286	0.48%	170,099,887	0.52%
33867	0	0.00%	0	0.00%	0	0	0	0.00%	575,302	0.00%
32311	0	0.00%	16,052,686	0.13%	0	0	0	0.00%	16,052,712	0.05%
32736	0	0.00%	0	0.00%	0	0	12,405,422	0.09%	12,562,572	0.04%
33585	0	0.00%	1,086,488	0.01%	0	0	1,382,065	0.01%	2,468,553	0.01%
34434	0	0.00%	12,916,043	0.11%	0	0	10,364,080	0.08%	23,280,124	0.07%
33868	0	0.00%	4,729,329	0.04%	0	0	9,167,376	0.07%	13,896,704	0.04%
33444	0	0.00%	22,364,865	0.18%	0	0	4,520,766	0.03%	26,885,816	0.08%
34293	39,629,613	0.56%	0	0.00%	0	0	0	0.00%	39,632,316	0.12%
32312	0	0.00%	40,000,631	0.33%	0	0	0	0.00%	40,000,654	0.12%
33445	0	0.00%	52,777,287	0.44%	0	0	13,374,173	0.10%	66,151,663	0.20%
32738	58,322,888	0.82%	0	0.00%	0	0	46,473,750	0.36%	105,267,221	0.32%
34436	0	0.00%	7,863,945	0.07%	0	0	7,885,721	0.06%	15,749,666	0.05%
33870	31,086,968	0.44%	12,704,122	0.11%	0	0	31,416,303	0.24%	75,207,393	0.23%
33446	0	0.00%	68,777,470	0.57%	0	0	51,785,128	0.40%	120,562,972	0.37%
32174	92,266,141	1.30%	73,486,132	0.61%	0	0	93,280,765	0.71%	259,033,037	0.79%
33872	29,496,919	0.41%	13,040,247	0.11%	0	0	29,818,977	0.23%	72,356,142	0.22%
33873	18,289,806	0.26%	0	0.00%	0	0	11,016,403	0.08%	29,405,560	0.09%
33449	0	0.00%	21,922,071	0.18%	0	0	17,936,024	0.14%	39,858,095	0.12%
32317	0	0.00%	13,573,578	0.11%	0	0	0	0.00%	13,573,578	0.04%
32034	0	0.00%	0	0.00%	0	0	5,849,603	0.04%	5,849,606	0.02%
32176	57,025,017	0.80%	46,280,518	0.38%	0	0	52,801,800	0.40%	156,107,334	0.47%
32601	0	0.00%	2,792,728	0.02%	0	0	8,114,120	0.06%	10,906,849	0.03%
33592	0	0.00%	5,524,918	0.05%	0	0	5,539,785	0.04%	11,064,702	0.03%
33875	18,563,388	0.26%	9,633,871	0.08%	0	0	22,669,942	0.17%	50,867,201	0.15%
32744	3,616,104	0.05%	0	0.00%	0	0	3,656,751	0.03%	7,318,922	0.02%
34442	0	0.00%	28,427,631	0.24%	0	0	22,646,678	0.17%	51,074,309	0.15%
33876	9,049,893	0.13%	6,974,547	0.06%	0	0	12,841,308	0.10%	28,865,749	0.09%
32603	0	0.00%	0	0.00%	0	0	1,521,028	0.01%	1,673,663	0.01%
33594	0	0.00%	1,066,764	0.01%	0	0	31,597,217	0.24%	32,663,981	0.10%
32179	0	0.00%	4,170,515	0.03%	0	0	5,677,886	0.04%	9,848,401	0.03%
33877	530,622	0.01%	0	0.00%	0	0	0	0.00%	1,134,850	0.00%
32038	0	0.00%	0	0.00%	0	0	3,692,667	0.03%	3,692,667	0.01%
32746	91,476,035	1.28%	46,207,579	0.38%	0	0	70,831,936	0.54%	208,515,550	0.63%
32180	0	0.00%	0	0.00%	0	0	1,697,115	0.01%	1,697,115	0.01%
32605	0	0.00%	1,378,111	0.01%	0	0	20,274,475	0.15%	21,652,587	0.07%
33596	0	0.00%	0	0.00%	0	0	28,000,675	0.21%	28,402,664	0.09%
32606	0	0.00%	2,602,197	0.02%	0	0	17,963,100	0.14%	20,565,298	0.06%
33455	0	0.00%	92,864,629	0.77%	0	0	84,602,006	0.65%	177,466,700	0.54%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
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33597	0	0.00%	4,696,308	0.04%	0	0	4,709,435	0.04%	9,405,743	0.03%
34446	0	0.00%	25,406,128	0.21%	0	0	17,312,063	0.13%	42,718,191	0.13%
33880	28,964,665	0.41%	1,037,617	0.01%	0	0	38,259,003	0.29%	68,261,285	0.21%
32607	0	0.00%	3,228,425	0.03%	0	0	16,550,742	0.13%	19,779,166	0.06%
33598	0	0.00%	0	0.00%	0	0	5,916,459	0.05%	6,191,794	0.02%
33881	35,090,826	0.49%	17,755,002	0.15%	0	0	43,083,312	0.33%	95,929,140	0.29%
32608	0	0.00%	25,580,754	0.21%	0	0	25,644,783	0.20%	51,225,536	0.16%
34448	0	0.00%	14,669,565	0.12%	0	0	11,098,662	0.08%	25,768,228	0.08%
34731	0	0.00%	12,318,544	0.10%	0	0	16,149,012	0.12%	28,467,556	0.09%
32750	35,589,404	0.50%	21,432,084	0.18%	0	0	35,985,408	0.27%	93,006,896	0.28%
32609	0	0.00%	709,121	0.01%	0	0	6,019,577	0.05%	6,728,698	0.02%
33458	0	0.00%	147,860,796	1.22%	0	0	120,252,814	0.92%	268,113,737	0.81%
34449	0	0.00%	1,956,207	0.02%	0	0	1,470,348	0.01%	3,426,555	0.01%
32751	50,148,482	0.70%	24,983,681	0.21%	0	0	38,623,856	0.30%	113,756,018	0.34%
34450	0	0.00%	14,633,127	0.12%	0	0	14,669,010	0.11%	29,302,137	0.09%
33884	75,599,790	1.06%	29,582,603	0.24%	0	0	66,544,219	0.51%	171,726,613	0.52%
33601	0	0.00%	0	0.00%	0	0	0	0.00%	561,047	0.00%
33460	0	0.00%	35,085,978	0.29%	0	0	21,712,745	0.17%	56,798,799	0.17%
34734	4,113,783	0.06%	0	0.00%	0	0	8,103,998	0.06%	12,280,419	0.04%
33602	0	0.00%	15,940,604	0.13%	0	0	8,031,993	0.06%	23,972,597	0.07%
33461	0	0.00%	37,172,868	0.31%	0	0	19,530,453	0.15%	56,703,533	0.17%
34452	0	0.00%	14,756,275	0.12%	0	0	14,796,168	0.11%	29,552,443	0.09%
32754	15,298,001	0.21%	11,805,965	0.10%	0	0	18,684,482	0.14%	45,788,447	0.14%
33462	0	0.00%	53,651,126	0.44%	0	0	27,958,898	0.21%	81,610,210	0.25%
34736	0	0.00%	10,761,927	0.09%	0	0	17,006,111	0.13%	27,768,038	0.08%
33604	0	0.00%	765,613	0.01%	0	0	762,645	0.01%	1,528,257	0.00%
34453	0	0.00%	15,088,671	0.12%	0	0	11,789,260	0.09%	26,877,931	0.08%
33463	0	0.00%	67,160,889	0.56%	0	0	36,109,985	0.28%	103,271,037	0.31%
32331	0	0.00%	1,736,529	0.01%	0	0	0	0.00%	1,736,529	0.01%
32190	0	0.00%	0	0.00%	0	0	521,681	0.00%	521,740	0.00%
34737	0	0.00%	4,815,691	0.04%	0	0	7,119,905	0.05%	11,935,596	0.04%
33605	0	0.00%	920,233	0.01%	0	0	914,836	0.01%	1,835,069	0.01%
32615	0	0.00%	0	0.00%	0	0	10,348,949	0.08%	10,614,821	0.03%
32757	0	0.00%	1,203,980	0.01%	0	0	31,449,052	0.24%	32,653,049	0.10%
33606	0	0.00%	19,330,434	0.16%	0	0	1,912,335	0.01%	21,242,769	0.06%
33890	10,933,850	0.15%	0	0.00%	0	0	3,723,250	0.03%	14,692,584	0.04%
34739	525,026	0.01%	930,078	0.01%	0	0	1,615,631	0.01%	3,070,735	0.01%
33607	0	0.00%	939,535	0.01%	0	0	934,368	0.01%	1,873,903	0.01%
32617	0	0.00%	2,165,391	0.02%	0	0	2,852,197	0.02%	5,017,588	0.02%
32759	2,700,229	0.04%	2,722,932	0.02%	0	0	3,833,302	0.03%	9,256,463	0.03%
32618	0	0.00%	3,302,177	0.03%	0	0	3,311,512	0.03%	6,613,689	0.02%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
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33467	0	0.00%	113,397,066	0.94%	0	0	89,711,918	0.69%	203,109,302	0.62%
32052	0	0.00%	1,881,420	0.02%	0	0	0	0.00%	1,913,626	0.01%
32901	0	0.00%	32,230,473	0.27%	0	0	40,592,510	0.31%	72,823,217	0.22%
34741	33,904,295	0.48%	16,792,061	0.14%	0	0	25,313,437	0.19%	76,009,792	0.23%
33609	0	0.00%	3,015,054	0.02%	0	0	2,998,362	0.02%	6,013,415	0.02%
33468	0	0.00%	936,431	0.01%	0	0	741,516	0.01%	1,677,947	0.01%
32336	0	0.00%	589,646	0.00%	0	0	0	0.00%	589,646	0.00%
32053	0	0.00%	1,114,044	0.01%	0	0	0	0.00%	1,133,340	0.00%
32195	0	0.00%	2,304,560	0.02%	0	0	3,080,578	0.02%	5,385,138	0.02%
33610	0	0.00%	0	0.00%	0	0	12,143,889	0.09%	12,558,490	0.04%
33469	0	0.00%	77,065,251	0.64%	0	0	70,672,282	0.54%	147,737,617	0.45%
32054	0	0.00%	0	0.00%	0	0	3,597,883	0.03%	3,597,883	0.01%
32903	0	0.00%	61,345,811	0.51%	0	0	72,981,751	0.56%	134,328,120	0.41%
34601	0	0.00%	12,112,142	0.10%	0	0	12,145,276	0.09%	24,257,418	0.07%
34743	47,169,109	0.66%	20,384,161	0.17%	0	0	38,699,210	0.30%	106,252,480	0.32%
33611	0	0.00%	26,961,632	0.22%	0	0	3,319,503	0.03%	30,281,135	0.09%
32055	0	0.00%	0	0.00%	0	0	6,741,010	0.05%	6,741,010	0.02%
32904	0	0.00%	43,075,614	0.36%	0	0	59,187,893	0.45%	102,263,588	0.31%
34602	0	0.00%	8,154,135	0.07%	0	0	8,176,924	0.06%	16,331,059	0.05%
32621	0	0.00%	1,642,996	0.01%	0	0	1,406,229	0.01%	3,049,225	0.01%
33470	0	0.00%	52,911,788	0.44%	0	0	44,490,142	0.34%	97,401,930	0.30%
34744	67,846,486	0.95%	31,063,301	0.26%	0	0	55,106,491	0.42%	154,016,278	0.47%
32763	12,728,859	0.18%	736,179	0.01%	0	0	12,865,825	0.10%	26,330,862	0.08%
33612	0	0.00%	1,642,888	0.01%	0	0	16,046,455	0.12%	17,689,342	0.05%
34461	0	0.00%	15,093,188	0.12%	0	0	15,135,274	0.12%	30,228,462	0.09%
32905	0	0.00%	30,323,100	0.25%	0	0	50,479,363	0.39%	80,802,880	0.25%
32622	0	0.00%	0	0.00%	0	0	710,724	0.01%	722,571	0.00%
33471	0	0.00%	4,241,188	0.04%	0	0	4,253,077	0.03%	8,494,265	0.03%
33896	15,233,768	0.21%	2,388,502	0.02%	0	0	15,365,996	0.12%	32,988,265	0.10%
32764	4,912,926	0.07%	2,222,233	0.02%	0	0	3,246,318	0.02%	10,381,477	0.03%
33613	0	0.00%	17,405,592	0.14%	0	0	17,428,369	0.13%	34,833,961	0.11%
32340	0	0.00%	4,290,269	0.04%	0	0	0	0.00%	4,290,269	0.01%
32906	0	0.00%	0	0.00%	0	0	0	0.00%	739,972	0.00%
33755	0	0.00%	15,744,441	0.13%	0	0	2,470,427	0.02%	18,214,868	0.06%
34604	0	0.00%	9,004,603	0.07%	0	0	7,019,590	0.05%	16,024,193	0.05%
33472	0	0.00%	37,083,457	0.31%	0	0	19,288,500	0.15%	56,371,958	0.17%
33897	807,056	0.01%	812,758	0.01%	0	0	25,154,728	0.19%	26,774,542	0.08%
34746	72,183,185	1.01%	33,812,968	0.28%	0	0	57,552,427	0.44%	163,548,580	0.50%
32765	145,200,923	2.04%	51,690,402	0.43%	0	0	106,348,587	0.81%	303,239,912	0.92%
33614	0	0.00%	2,777,555	0.02%	0	0	2,761,091	0.02%	5,538,646	0.02%
32907	0	0.00%	51,035,236	0.42%	0	0	92,852,719	0.71%	143,887,955	0.44%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33756	0	0.00%	28,124,494	0.23%	0	0	6,531,742	0.05%	34,656,236	0.11%
33473	0	0.00%	10,154,574	0.08%	0	0	8,637,657	0.07%	18,792,231	0.06%
33898	43,979,481	0.62%	10,273,891	0.08%	0	0	25,709,275	0.20%	79,962,647	0.24%
34747	47,592,996	0.67%	38,725,916	0.32%	0	0	56,985,020	0.44%	143,303,932	0.43%
32766	38,082,194	0.53%	17,467,318	0.14%	0	0	28,453,413	0.22%	84,002,925	0.25%
33615	0	0.00%	25,688,507	0.21%	0	0	3,214,685	0.02%	28,903,192	0.09%
32059	0	0.00%	928,217	0.01%	0	0	0	0.00%	928,217	0.00%
32908	0	0.00%	11,808,247	0.10%	0	0	19,383,658	0.15%	31,191,905	0.09%
34606	0	0.00%	37,366,880	0.31%	0	0	23,598,720	0.18%	60,965,600	0.18%
32625	0	0.00%	2,881,884	0.02%	0	0	1,507,389	0.01%	4,389,273	0.01%
34748	0	0.00%	31,730,084	0.26%	0	0	45,237,942	0.35%	76,968,026	0.23%
32767	0	0.00%	0	0.00%	0	0	1,007,015	0.01%	1,022,371	0.00%
33616	0	0.00%	7,900,883	0.07%	0	0	1,245,143	0.01%	9,146,026	0.03%
34465	0	0.00%	27,038,252	0.22%	0	0	18,078,160	0.14%	45,116,412	0.14%
32060	0	0.00%	7,781,077	0.06%	0	0	0	0.00%	7,930,745	0.02%
32909	0	0.00%	43,231,564	0.36%	0	0	63,577,541	0.49%	106,809,110	0.32%
34607	0	0.00%	13,934,869	0.12%	0	0	8,870,558	0.07%	22,805,427	0.07%
32626	0	0.00%	4,083,085	0.03%	0	0	0	0.00%	4,143,556	0.01%
33617	0	0.00%	3,025,391	0.03%	0	0	22,233,952	0.17%	25,259,343	0.08%
32344	0	0.00%	7,990,621	0.07%	0	0	0	0.00%	7,990,621	0.02%
33759	0	0.00%	11,831,972	0.10%	0	0	0	0.00%	11,832,401	0.04%
34608	0	0.00%	42,079,281	0.35%	0	0	27,221,652	0.21%	69,300,933	0.21%
33476	0	0.00%	4,188,563	0.03%	0	0	3,377,437	0.03%	7,566,004	0.02%
33901	29,420,746	0.41%	0	0.00%	0	0	0	0.00%	29,420,845	0.09%
33618	0	0.00%	25,695,685	0.21%	0	0	25,737,235	0.20%	51,432,920	0.16%
33760	0	0.00%	8,974,711	0.07%	0	0	0	0.00%	8,975,199	0.03%
34609	0	0.00%	52,447,194	0.43%	0	0	36,795,124	0.28%	89,242,318	0.27%
32628	0	0.00%	1,379,721	0.01%	0	0	0	0.00%	1,379,721	0.00%
33477	0	0.00%	102,248,609	0.85%	0	0	92,835,905	0.71%	195,084,866	0.59%
33619	0	0.00%	529,124	0.00%	0	0	10,928,858	0.08%	11,457,982	0.03%
32346	0	0.00%	2,310,525	0.02%	0	0	0	0.00%	2,310,525	0.01%
33761	0	0.00%	20,531,625	0.17%	0	0	5,644,364	0.04%	26,175,989	0.08%
34610	0	0.00%	9,585,599	0.08%	0	0	6,681,939	0.05%	16,267,538	0.05%
33478	0	0.00%	46,503,749	0.38%	0	0	43,060,908	0.33%	89,564,658	0.27%
33903	60,428,982	0.85%	0	0.00%	0	0	0	0.00%	60,429,121	0.18%
32771	64,820,146	0.91%	37,486,696	0.31%	0	0	52,162,585	0.40%	154,469,428	0.47%
33762	0	0.00%	8,318,299	0.07%	0	0	0	0.00%	8,318,513	0.03%
32347	0	0.00%	4,182,725	0.03%	0	0	0	0.00%	4,182,725	0.01%
32064	0	0.00%	2,013,474	0.02%	0	0	0	0.00%	2,046,787	0.01%
34470	0	0.00%	10,590,160	0.09%	0	0	10,614,049	0.08%	21,204,209	0.06%
34753	0	0.00%	2,636,091	0.02%	0	0	4,159,498	0.03%	6,795,589	0.02%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33904	111,371,802	1.56%	0	0.00%	0	0	0	0.00%	111,372,031	0.34%
33480	0	0.00%	231,591,821	1.91%	0	0	156,776,208	1.20%	388,368,828	1.18%
32631	0	0.00%	0	0.00%	0	0	537,904	0.00%	554,799	0.00%
33763	0	0.00%	1,425,850	0.01%	0	0	1,417,390	0.01%	2,843,240	0.01%
32348	0	0.00%	3,879,746	0.03%	0	0	0	0.00%	3,879,746	0.01%
34471	0	0.00%	21,810,917	0.18%	0	0	28,314,213	0.22%	50,125,130	0.15%
32773	36,257,348	0.51%	15,245,853	0.13%	0	0	23,238,756	0.18%	74,741,956	0.23%
33905	32,588,805	0.46%	0	0.00%	0	0	0	0.00%	32,588,829	0.10%
34613	0	0.00%	23,092,845	0.19%	0	0	15,308,760	0.12%	38,401,605	0.12%
33764	0	0.00%	21,197,055	0.18%	0	0	0	0.00%	21,197,833	0.06%
32066	0	0.00%	1,892,795	0.02%	0	0	0	0.00%	1,892,795	0.01%
34472	0	0.00%	18,187,630	0.15%	0	0	23,106,829	0.18%	41,294,459	0.13%
34614	0	0.00%	6,181,666	0.05%	0	0	5,053,712	0.04%	11,235,378	0.03%
33765	0	0.00%	7,894,887	0.07%	0	0	0	0.00%	7,895,060	0.02%
32350	0	0.00%	565,247	0.00%	0	0	0	0.00%	565,247	0.00%
34473	0	0.00%	16,214,012	0.13%	0	0	16,258,860	0.12%	32,472,873	0.10%
33624	0	0.00%	30,586,572	0.25%	0	0	30,661,240	0.23%	61,247,812	0.19%
34756	0	0.00%	3,762,781	0.03%	0	0	6,360,751	0.05%	10,123,533	0.03%
33907	28,337,602	0.40%	0	0.00%	0	0	0	0.00%	28,337,861	0.09%
33483	0	0.00%	44,095,474	0.36%	0	0	13,036,410	0.10%	57,132,430	0.17%
32776	0	0.00%	9,569,411	0.08%	0	0	12,788,450	0.10%	22,357,861	0.07%
33908	119,570,060	1.68%	0	0.00%	0	0	0	0.00%	119,570,817	0.36%
33625	0	0.00%	16,810,110	0.14%	0	0	16,853,140	0.13%	33,663,249	0.10%
34474	0	0.00%	10,563,211	0.09%	0	0	10,581,544	0.08%	21,144,755	0.06%
33767	0	0.00%	44,937,107	0.37%	0	0	26,357,615	0.20%	71,294,722	0.22%
33484	0	0.00%	53,712,370	0.44%	0	0	16,658,489	0.13%	70,371,170	0.21%
33060	0	0.00%	4,811,814	0.04%	0	0	0	0.00%	4,812,341	0.01%
33909	42,745,114	0.60%	0	0.00%	0	0	0	0.00%	42,745,177	0.13%
34758	49,875,817	0.70%	23,079,815	0.19%	0	0	40,478,845	0.31%	113,434,478	0.34%
33626	0	0.00%	29,306,313	0.24%	0	0	1,249,964	0.01%	30,556,278	0.09%
34475	0	0.00%	3,682,139	0.03%	0	0	4,700,210	0.04%	8,382,349	0.03%
34759	56,212,150	0.79%	26,853,668	0.22%	0	0	44,256,202	0.34%	127,322,020	0.39%
32778	0	0.00%	15,848,924	0.13%	0	0	22,916,940	0.18%	38,765,880	0.12%
34476	0	0.00%	28,568,830	0.24%	0	0	28,647,872	0.22%	57,216,702	0.17%
32920	16,852,535	0.24%	28,077,304	0.23%	0	0	34,273,567	0.26%	79,203,406	0.24%
33486	0	0.00%	24,037,503	0.20%	0	0	0	0.00%	24,037,893	0.07%
33062	0	0.00%	32,975,384	0.27%	0	0	0	0.00%	32,980,519	0.10%
34760	0	0.00%	0	0.00%	0	0	1,459,223	0.01%	1,471,578	0.00%
32779	76,370,096	1.07%	47,813,704	0.40%	0	0	77,206,983	0.59%	201,390,783	0.61%
33487	0	0.00%	60,778,894	0.50%	0	0	19,933,408	0.15%	80,712,683	0.24%
33770	0	0.00%	18,713,987	0.15%	0	0	2,735,369	0.02%	21,449,356	0.07%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
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33629	0	0.00%	37,685,847	0.31%	0	0	2,648,530	0.02%	40,334,377	0.12%
32780	45,180,778	0.63%	45,554,964	0.38%	0	0	72,091,337	0.55%	162,827,078	0.49%
34761	30,458,465	0.43%	1,636,729	0.01%	0	0	54,055,278	0.41%	86,150,472	0.26%
33912	50,357,015	0.71%	0	0.00%	0	0	0	0.00%	50,357,095	0.15%
33771	0	0.00%	13,892,244	0.11%	0	0	0	0.00%	13,894,017	0.04%
32922	6,788,227	0.10%	14,010,399	0.12%	0	0	19,243,891	0.15%	40,042,517	0.12%
34479	0	0.00%	7,863,928	0.07%	0	0	10,323,007	0.08%	18,186,935	0.06%
33913	34,891,542	0.49%	0	0.00%	0	0	0	0.00%	34,891,745	0.11%
33064	0	0.00%	8,414,436	0.07%	0	0	0	0.00%	8,416,473	0.03%
34762	0	0.00%	0	0.00%	0	0	633,124	0.00%	1,057,468	0.00%
32640	0	0.00%	0	0.00%	0	0	4,810,589	0.04%	4,900,224	0.01%
33772	0	0.00%	33,543,978	0.28%	0	0	0	0.00%	33,544,651	0.10%
34480	0	0.00%	15,962,640	0.13%	0	0	20,189,204	0.15%	36,151,844	0.11%
33914	126,887,503	1.78%	0	0.00%	0	0	0	0.00%	126,887,825	0.38%
32641	0	0.00%	0	0.00%	0	0	4,059,600	0.03%	4,126,045	0.01%
33773	0	0.00%	11,931,477	0.10%	0	0	0	0.00%	11,931,552	0.04%
34481	0	0.00%	17,100,682	0.14%	0	0	17,145,193	0.13%	34,245,875	0.10%
32359	0	0.00%	1,891,206	0.02%	0	0	0	0.00%	1,891,206	0.01%
33774	0	0.00%	29,413,937	0.24%	0	0	4,680,483	0.04%	34,094,419	0.10%
32501	0	0.00%	0	0.00%	10,089,163	0.01508	0	0.00%	10,089,163	0.03%
34482	0	0.00%	18,887,081	0.16%	0	0	18,938,065	0.14%	37,825,146	0.11%
33916	15,257,105	0.21%	0	0.00%	0	0	0	0.00%	15,258,293	0.05%
33067	0	0.00%	2,608,183	0.02%	0	0	0	0.00%	2,608,953	0.01%
32784	0	0.00%	6,687,916	0.06%	0	0	9,981,799	0.08%	16,669,715	0.05%
32643	0	0.00%	0	0.00%	0	0	6,542,816	0.05%	6,692,324	0.02%
32926	15,030,735	0.21%	24,320,301	0.20%	0	0	39,295,475	0.30%	78,646,512	0.24%
33634	0	0.00%	10,848,352	0.09%	0	0	761,054	0.01%	11,609,406	0.04%
33917	54,411,565	0.76%	0	0.00%	0	0	0	0.00%	54,412,007	0.17%
32502	0	0.00%	0	0.00%	2,674,046	0.004	0	0.00%	2,674,046	0.01%
33493	0	0.00%	1,093,065	0.01%	0	0	1,095,480	0.01%	2,188,545	0.01%
33776	0	0.00%	33,951,394	0.28%	0	0	951,187	0.01%	34,902,580	0.11%
32927	29,876,055	0.42%	30,126,682	0.25%	0	0	48,095,915	0.37%	108,098,652	0.33%
32503	0	0.00%	0	0.00%	39,974,385	0.05976	0	0.00%	39,974,385	0.12%
34484	0	0.00%	4,044,469	0.03%	0	0	4,052,656	0.03%	8,097,125	0.02%
33635	0	0.00%	8,776,581	0.07%	0	0	551,052	0.00%	9,327,633	0.03%
33069	0	0.00%	17,484,423	0.14%	0	0	0	0.00%	17,486,646	0.05%
33777	0	0.00%	27,888,718	0.23%	0	0	0	0.00%	27,889,766	0.08%
34202	0	0.00%	0	0.00%	0	0	3,220,085	0.02%	3,220,349	0.01%
32504	0	0.00%	0	0.00%	30,469,233	0.04555	0	0.00%	30,469,233	0.09%
33919	85,612,039	1.20%	0	0.00%	0	0	0	0.00%	85,613,065	0.26%
33778	0	0.00%	16,609,114	0.14%	0	0	0	0.00%	16,609,433	0.05%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
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32080	0	0.00%	37,095,572	0.31%	0	0	37,177,165	0.28%	74,272,895	0.23%
33920	7,124,963	0.10%	0	0.00%	0	0	0	0.00%	7,393,356	0.02%
32505	0	0.00%	0	0.00%	16,650,676	0.02489	0	0.00%	16,650,676	0.05%
33637	0	0.00%	1,309,572	0.01%	0	0	6,923,335	0.05%	8,232,907	0.03%
34769	30,904,108	0.43%	13,396,080	0.11%	0	0	24,587,878	0.19%	68,888,066	0.21%
33496	0	0.00%	58,435,689	0.48%	0	0	7,663,153	0.06%	66,098,842	0.20%
32506	0	0.00%	0	0.00%	36,208,125	0.05413	0	0.00%	36,208,125	0.11%
32789	100,841,975	1.42%	45,373,977	0.38%	0	0	84,913,599	0.65%	231,129,551	0.70%
33921	53,377,420	0.75%	18,400,595	0.15%	0	0	18,426,706	0.14%	90,204,722	0.27%
32648	0	0.00%	597,099	0.00%	0	0	0	0.00%	597,099	0.00%
32931	24,048,025	0.34%	71,141,518	0.59%	0	0	86,376,820	0.66%	181,566,363	0.55%
32082	0	0.00%	67,521,062	0.56%	0	0	67,691,572	0.52%	135,212,634	0.41%
34205	0	0.00%	4,206,031	0.03%	0	0	0	0.00%	4,206,556	0.01%
32507	0	0.00%	0	0.00%	71,894,193	0.10748	0	0.00%	71,894,193	0.22%
34488	0	0.00%	0	0.00%	0	0	4,175,101	0.03%	4,272,819	0.01%
34771	25,728,668	0.36%	12,795,204	0.11%	0	0	22,622,563	0.17%	61,146,435	0.19%
33922	22,409,213	0.31%	0	0.00%	0	0	0	0.00%	22,409,213	0.07%
33073	0	0.00%	3,916,505	0.03%	0	0	0	0.00%	3,916,636	0.01%
33498	0	0.00%	22,352,160	0.18%	0	0	0	0.00%	22,771,994	0.07%
33781	0	0.00%	12,370,530	0.10%	0	0	0	0.00%	12,371,052	0.04%
32932	0	0.00%	0	0.00%	0	0	0	0.00%	611,312	0.00%
34772	38,696,261	0.54%	18,224,367	0.15%	0	0	31,290,087	0.24%	88,210,716	0.27%
33782	0	0.00%	14,924,627	0.12%	0	0	0	0.00%	14,925,114	0.05%
32084	0	0.00%	21,757,391	0.18%	0	0	21,811,958	0.17%	43,569,349	0.13%
34207	0	0.00%	14,275,796	0.12%	0	0	0	0.00%	14,276,056	0.04%
32792	72,701,073	1.02%	29,522,652	0.24%	0	0	60,910,479	0.47%	163,134,203	0.49%
34773	2,558,534	0.04%	1,919,258	0.02%	0	0	3,000,235	0.02%	7,478,028	0.02%
33924	56,710,943	0.80%	10,142,367	0.08%	0	0	8,047,733	0.06%	74,901,043	0.23%
32934	0	0.00%	37,572,898	0.31%	0	0	52,053,432	0.40%	90,106,267	0.27%
34491	0	0.00%	24,185,879	0.20%	0	0	30,943,945	0.24%	55,129,824	0.17%
33076	0	0.00%	1,959,560	0.02%	0	0	0	0.00%	1,959,855	0.01%
34208	0	0.00%	0	0.00%	0	0	579,506	0.00%	579,506	0.00%
32935	0	0.00%	71,386,575	0.59%	0	0	88,887,309	0.68%	160,274,367	0.49%
32086	0	0.00%	1,753,816	0.01%	0	0	22,190,850	0.17%	23,944,666	0.07%
34209	0	0.00%	7,947,104	0.07%	0	0	0	0.00%	7,947,967	0.02%
32653	0	0.00%	0	0.00%	0	0	12,147,379	0.09%	12,626,688	0.04%
33785	0	0.00%	30,208,504	0.25%	0	0	20,119,065	0.15%	50,327,569	0.15%
34210	0	0.00%	18,120,491	0.15%	0	0	0	0.00%	18,121,409	0.05%
33786	0	0.00%	9,653,025	0.08%	0	0	7,289,220	0.06%	16,942,245	0.05%
32937	6,546,650	0.09%	100,074,893	0.83%	0	0	107,566,121	0.82%	214,187,664	0.65%
32796	24,604,318	0.35%	24,808,140	0.21%	0	0	33,143,255	0.25%	82,555,712	0.25%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
33928	39,314,277	0.55%	0	0.00%	0	0	0	0.00%	39,314,794	0.12%
34212	0	0.00%	0	0.00%	0	0	1,351,403	0.01%	1,351,475	0.00%
32514	0	0.00%	0	0.00%	36,163,236	0.05406	0	0.00%	36,163,236	0.11%
34637	0	0.00%	6,869,114	0.06%	0	0	6,888,080	0.05%	13,757,194	0.04%
32656	0	0.00%	0	0.00%	0	0	7,458,050	0.06%	7,458,050	0.02%
33647	0	0.00%	58,385,343	0.48%	0	0	80,092,816	0.61%	138,478,159	0.42%
32798	0	0.00%	0	0.00%	0	0	3,566,887	0.03%	3,597,963	0.01%
32940	43,981,912	0.62%	94,304,749	0.78%	0	0	123,503,958	0.94%	261,790,618	0.79%
34638	0	0.00%	20,192,719	0.17%	0	0	20,249,774	0.15%	40,442,494	0.12%
33931	54,424,538	0.76%	11,285,392	0.09%	0	0	0	0.00%	65,710,047	0.20%
32233	0	0.00%	1,690,699	0.01%	0	0	17,306,973	0.13%	18,997,672	0.06%
34639	0	0.00%	23,667,260	0.20%	0	0	23,731,891	0.18%	47,399,151	0.14%
34215	0	0.00%	2,074,876	0.02%	0	0	1,593,835	0.01%	3,668,723	0.01%
34498	0	0.00%	691,724	0.01%	0	0	578,651	0.00%	1,270,375	0.00%
34216	0	0.00%	7,409,211	0.06%	0	0	0	0.00%	7,512,182	0.02%
32801	23,483,520	0.33%	12,501,294	0.10%	0	0	20,083,883	0.15%	56,068,696	0.17%
34217	0	0.00%	33,318,106	0.28%	0	0	4,193,982	0.03%	37,512,324	0.11%
32095	0	0.00%	543,289	0.00%	0	0	8,354,674	0.06%	8,897,964	0.03%
33510	0	0.00%	1,261,545	0.01%	0	0	18,675,096	0.14%	19,936,641	0.06%
34218	0	0.00%	0	0.00%	0	0	0	0.00%	508,533	0.00%
32803	46,057,492	0.65%	18,531,666	0.15%	0	0	38,482,737	0.29%	103,071,895	0.31%
33935	0	0.00%	0	0.00%	0	0	6,902,188	0.05%	7,007,790	0.02%
33511	0	0.00%	1,590,076	0.01%	0	0	34,077,866	0.26%	35,667,941	0.11%
34785	0	0.00%	8,875,550	0.07%	0	0	11,722,113	0.09%	20,597,662	0.06%
33936	14,844,830	0.21%	0	0.00%	0	0	0	0.00%	14,844,939	0.05%
34219	0	0.00%	0	0.00%	0	0	21,026,169	0.16%	21,393,371	0.06%
32804	52,436,226	0.74%	21,730,165	0.18%	0	0	43,944,677	0.34%	118,111,068	0.36%
34786	69,396,261	0.97%	69,982,998	0.58%	0	0	108,796,715	0.83%	248,175,974	0.75%
32805	15,331,995	0.22%	7,578,892	0.06%	0	0	15,490,198	0.12%	38,401,085	0.12%
32664	0	0.00%	0	0.00%	0	0	0	0.00%	984,524	0.00%
33513	0	0.00%	7,515,657	0.06%	0	0	7,536,670	0.06%	15,052,326	0.05%
34787	2,179,382	0.03%	2,194,630	0.02%	0	0	65,475,927	0.50%	69,849,938	0.21%
34221	0	0.00%	2,532,032	0.02%	0	0	2,518,605	0.02%	5,050,649	0.02%
32806	70,888,109	1.00%	24,642,675	0.20%	0	0	51,955,817	0.40%	147,486,601	0.45%
33514	0	0.00%	1,023,030	0.01%	0	0	1,333,646	0.01%	2,356,676	0.01%
32948	0	0.00%	3,576,719	0.03%	0	0	5,361,448	0.04%	8,938,167	0.03%
34788	0	0.00%	11,781,604	0.10%	0	0	18,572,750	0.14%	30,354,353	0.09%
32807	40,292,873	0.57%	13,774,653	0.11%	0	0	29,514,700	0.23%	83,582,226	0.25%
32949	0	0.00%	7,199,841	0.06%	0	0	8,668,205	0.07%	15,868,046	0.05%
32666	0	0.00%	0	0.00%	0	0	4,172,657	0.03%	4,172,657	0.01%
34223	45,239,899	0.64%	4,206,850	0.03%	0	0	0	0.00%	49,447,158	0.15%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32808	28,050,648	0.39%	1,293,427	0.01%	0	0	37,749,793	0.29%	67,093,868	0.20%
32667	0	0.00%	2,990,049	0.02%	0	0	2,998,503	0.02%	5,988,552	0.02%
32950	0	0.00%	13,175,540	0.11%	0	0	16,287,075	0.12%	29,462,615	0.09%
34224	38,494,487	0.54%	0	0.00%	0	0	1,838,362	0.01%	40,332,922	0.12%
32526	0	0.00%	0	0.00%	42,810,127	0.064	0	0.00%	42,810,127	0.13%
32809	36,685,765	0.52%	13,225,192	0.11%	0	0	26,956,761	0.21%	76,867,718	0.23%
32951	0	0.00%	50,248,848	0.42%	0	0	99,049,238	0.76%	149,298,519	0.45%
32668	0	0.00%	3,603,070	0.03%	0	0	3,137,258	0.02%	6,740,328	0.02%
32102	0	0.00%	0	0.00%	0	0	1,748,186	0.01%	1,748,186	0.01%
32810	32,899,218	0.46%	797,435	0.01%	0	0	33,259,105	0.25%	66,955,758	0.20%
33801	0	0.00%	1,723,937	0.01%	0	0	29,546,319	0.23%	31,270,344	0.09%
32952	22,250,277	0.31%	65,826,052	0.54%	0	0	76,150,069	0.58%	164,226,398	0.50%
32669	0	0.00%	9,572,468	0.08%	0	0	9,597,913	0.07%	19,170,381	0.06%
32811	17,216,978	0.24%	9,250,191	0.08%	0	0	17,336,543	0.13%	43,803,712	0.13%
32953	21,384,399	0.30%	44,901,743	0.37%	0	0	61,998,878	0.47%	128,285,020	0.39%
32812	65,678,885	0.92%	23,178,670	0.19%	0	0	48,219,285	0.37%	137,076,840	0.42%
34652	0	0.00%	19,098,519	0.16%	0	0	28,498,370	0.22%	47,596,889	0.14%
33803	0	0.00%	1,877,442	0.02%	0	0	33,991,461	0.26%	35,868,941	0.11%
34228	0	0.00%	86,015,419	0.71%	0	0	24,795,871	0.19%	110,812,725	0.34%
34653	0	0.00%	18,597,680	0.15%	0	0	18,632,425	0.14%	37,230,106	0.11%
32955	35,487,098	0.50%	65,491,529	0.54%	0	0	89,473,728	0.68%	190,452,354	0.58%
33521	0	0.00%	0	0.00%	0	0	521,444	0.00%	911,979	0.00%
33946	17,235,709	0.24%	8,490,276	0.07%	0	0	8,490,678	0.06%	34,216,662	0.10%
34229	0	0.00%	16,432,995	0.14%	0	0	0	0.00%	16,433,873	0.05%
32814	13,879,993	0.19%	8,658,371	0.07%	0	0	10,670,828	0.08%	33,209,191	0.10%
32531	0	0.00%	0	0.00%	3,033,057	0.00453	0	0.00%	3,033,057	0.01%
34654	0	0.00%	25,025,437	0.21%	0	0	17,727,487	0.14%	42,752,924	0.13%
33805	0	0.00%	0	0.00%	0	0	16,566,723	0.13%	17,029,611	0.05%
33947	27,713,152	0.39%	0	0.00%	0	0	931,455	0.01%	28,644,670	0.09%
34655	0	0.00%	41,692,676	0.34%	0	0	41,786,031	0.32%	83,478,707	0.25%
33523	0	0.00%	12,947,700	0.11%	0	0	12,983,709	0.10%	25,931,409	0.08%
32816	6,360,093	0.09%	3,398,347	0.03%	0	0	4,840,956	0.04%	14,599,396	0.04%
33948	56,438,090	0.79%	0	0.00%	0	0	0	0.00%	56,438,268	0.17%
32250	0	0.00%	5,447,860	0.05%	0	0	27,503,028	0.21%	32,950,888	0.10%
34231	0	0.00%	35,991,112	0.30%	0	0	0	0.00%	35,992,697	0.11%
32533	0	0.00%	0	0.00%	32,717,211	0.04891	0	0.00%	32,717,211	0.10%
34797	0	0.00%	1,762,041	0.01%	0	0	3,934,745	0.03%	5,696,785	0.02%
32958	0	0.00%	63,903,796	0.53%	0	0	80,635,031	0.62%	144,538,885	0.44%
32817	60,266,452	0.85%	21,291,787	0.18%	0	0	44,221,232	0.34%	125,779,471	0.38%
32534	0	0.00%	0	0.00%	12,250,551	0.01831	0	0.00%	12,250,551	0.04%
33525	0	0.00%	12,701,329	0.11%	0	0	17,953,797	0.14%	30,655,126	0.09%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32818	37,105,582	0.52%	1,273,435	0.01%	0	0	49,323,826	0.38%	87,702,843	0.27%
33950	143,079,499	2.01%	0	0.00%	0	0	6,582,123	0.05%	149,662,053	0.45%
32535	0	0.00%	0	0.00%	2,438,559	0.00365	0	0.00%	2,438,559	0.01%
33809	0	0.00%	19,318,023	0.16%	0	0	38,942,390	0.30%	58,260,420	0.18%
32960	0	0.00%	46,885,175	0.39%	0	0	55,516,393	0.42%	102,401,675	0.31%
32819	66,985,980	0.94%	34,133,493	0.28%	0	0	67,681,496	0.52%	168,800,969	0.51%
34234	0	0.00%	1,351,968	0.01%	0	0	0	0.00%	1,352,411	0.00%
33810	0	0.00%	28,206,861	0.23%	0	0	51,260,676	0.39%	79,467,536	0.24%
32112	0	0.00%	0	0.00%	0	0	2,732,044	0.02%	2,795,035	0.01%
33527	0	0.00%	8,260,573	0.07%	0	0	11,878,417	0.09%	20,138,990	0.06%
33952	90,984,409	1.28%	0	0.00%	0	0	0	0.00%	90,984,641	0.28%
32820	12,577,864	0.18%	6,886,589	0.06%	0	0	9,460,677	0.07%	28,925,129	0.09%
33811	0	0.00%	536,913	0.00%	0	0	29,325,805	0.22%	29,862,718	0.09%
32962	0	0.00%	53,490,470	0.44%	0	0	62,066,101	0.47%	115,556,798	0.35%
32113	0	0.00%	2,406,330	0.02%	0	0	3,304,180	0.03%	5,710,510	0.02%
32821	29,577,116	0.42%	2,774,461	0.02%	0	0	21,940,504	0.17%	54,292,081	0.16%
33953	16,263,333	0.23%	0	0.00%	0	0	0	0.00%	16,263,499	0.05%
34236	0	0.00%	31,670,884	0.26%	0	0	0	0.00%	31,674,572	0.10%
33812	0	0.00%	0	0.00%	0	0	18,846,635	0.14%	19,286,834	0.06%
32963	0	0.00%	174,312,030	1.44%	0	0	381,352,191	2.91%	555,664,863	1.68%
32114	27,343,441	0.38%	20,285,482	0.17%	0	0	24,248,477	0.19%	71,877,399	0.22%
32680	0	0.00%	2,960,582	0.02%	0	0	0	0.00%	3,004,188	0.01%
33954	31,003,808	0.44%	0	0.00%	0	0	0	0.00%	31,134,749	0.09%
32822	56,464,823	0.79%	20,951,836	0.17%	0	0	41,263,886	0.32%	118,680,545	0.36%
33813	0	0.00%	1,012,857	0.01%	0	0	72,438,771	0.55%	73,451,683	0.22%
34945	0	0.00%	8,467,071	0.07%	0	0	10,694,018	0.08%	19,161,089	0.06%
33955	62,288,289	0.87%	0	0.00%	0	0	0	0.00%	62,288,662	0.19%
34238	0	0.00%	5,245,526	0.04%	0	0	0	0.00%	5,246,812	0.02%
34946	0	0.00%	9,043,184	0.07%	0	0	11,381,543	0.09%	20,424,728	0.06%
32824	65,105,622	0.91%	26,201,674	0.22%	0	0	53,250,074	0.41%	144,557,370	0.44%
33956	29,725,141	0.42%	0	0.00%	0	0	0	0.00%	29,833,403	0.09%
34239	0	0.00%	1,303,774	0.01%	0	0	0	0.00%	1,303,898	0.00%
32541	0	0.00%	0	0.00%	31,572,541	0.0472	0	0.00%	31,572,541	0.10%
33815	0	0.00%	0	0.00%	0	0	5,496,298	0.04%	5,625,308	0.02%
32966	0	0.00%	32,078,924	0.27%	0	0	46,055,217	0.35%	78,134,202	0.24%
32117	25,465,249	0.36%	16,641,282	0.14%	0	0	21,450,308	0.16%	63,556,838	0.19%
34947	0	0.00%	10,062,703	0.08%	0	0	11,300,189	0.09%	21,362,920	0.06%
32825	95,079,641	1.34%	36,375,438	0.30%	0	0	70,157,548	0.54%	201,612,627	0.61%
33957	131,948,131	1.85%	9,432,907	0.08%	0	0	0	0.00%	141,381,289	0.43%
32967	0	0.00%	39,038,963	0.32%	0	0	49,384,410	0.38%	88,423,529	0.27%
32118	71,737,474	1.01%	56,162,752	0.46%	0	0	61,756,652	0.47%	189,656,878	0.58%

	Hurricane Charley		Hurricane Frances		Hurricane Ivan		Hurricane Jeanne		Total	
ZIP Code	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)	Personal and Commercial Residential Monetary Contribution($)	Percent of Losses (%)
32826	33,718,948	0.47%	12,649,767	0.10%	0	0	18,921,609	0.14%	65,290,323	0.20%
32968	0	0.00%	32,418,698	0.27%	0	0	35,295,467	0.27%	67,714,164	0.21%
32119	41,560,436	0.58%	30,422,531	0.25%	0	0	36,636,468	0.28%	108,619,435	0.33%
34949	0	0.00%	56,928,839	0.47%	0	0	76,757,548	0.59%	133,687,469	0.41%
33534	0	0.00%	0	0.00%	0	0	5,673,793	0.04%	5,804,881	0.02%
32827	21,730,887	0.31%	10,269,921	0.08%	0	0	17,829,694	0.14%	49,830,502	0.15%
34242	0	0.00%	45,828,000	0.38%	0	0	0	0.00%	45,830,771	0.14%
Total	7,113,640,368		12,077,773,460		666,975,555		13,075,266,846		32,953,503,140	

Appendix E – Form A-4: Output Ranges

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
~~November 1, 2014~~ March 15, 2015

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Alachua	LOW	0.830	0.909	0.957	0.192	0.173	0.232	0.229	1.394
	AVERAGE	0.973	0.982	2.936	0.211	0.197	0.297	0.278	1.868
	HIGH	1.207	1.229	5.425	0.299	0.245	0.345	0.369	2.780
Baker	LOW	0.604	0.597	1.339	0.131	0.120	0.168	0.202	1.749
	AVERAGE	0.694	0.687	1.739	0.151	0.143	0.168	0.202	1.749
	HIGH	0.715	0.699	1.930	0.158	0.149	0.168	0.202	1.749
Bay	LOW	1.394	1.391	3.991	0.388	0.304	0.573	0.454	3.053
	AVERAGE	2.752	2.442	8.863	0.732	0.675	1.825	1.174	6.446
	HIGH	4.251	4.196	20.417	1.761	1.311	2.415	1.355	7.515
Bradford	LOW	0.743	0.730	1.859	0.167	0.149	0.324	0.272	1.855
	AVERAGE	0.877	0.864	2.337	0.197	0.174	0.324	0.272	2.028
	HIGH	1.112	1.102	3.407	0.274	0.245	0.324	0.272	2.117
Brevard	LOW	3.017	1.949	6.497	0.251	0.298	0.443	0.590	3.306
	AVERAGE	4.855	4.467	15.697	0.742	0.811	1.522	1.977	6.824
	HIGH	13.083	10.767	33.138	3.338	3.403	5.119	4.449	11.750
Broward	LOW	2.941	2.360	2.734	0.509	0.568	0.648	0.701	3.880
	AVERAGE	6.840	5.914	25.198	2.197	1.411	2.411	2.358	9.463
	HIGH	15.076	13.909	50.428	4.606	4.003	7.436	6.472	15.677
Calhoun	LOW	1.135	1.096	2.498	0.260	0.226	0.000	0.000	1.620
	AVERAGE	1.160	1.128	3.004	0.264	0.235	0.000	0.000	1.620
	HIGH	1.196	1.196	3.176	0.277	0.245	0.000	0.000	1.620
Charlotte	LOW	4.005	2.718	8.930	0.486	0.493	0.581	0.943	4.989

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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
	AVERAGE	5.383	4.586	12.411	0.825	0.631	1.578	1.077	5.692
	HIGH	6.991	6.683	31.282	1.396	1.067	2.201	1.593	8.063
Citrus	LOW	2.108	2.183	4.200	0.272	0.251	0.349	0.480	0.506
	AVERAGE	2.905	2.490	5.368	0.343	0.303	0.635	0.615	2.605
	HIGH	3.601	3.089	7.202	0.417	0.414	0.697	0.769	3.018
Clay	LOW	0.756	0.775	1.681	0.162	0.154	0.204	0.181	1.506
	AVERAGE	0.834	0.832	2.345	0.185	0.174	0.240	0.216	1.796
	HIGH	1.079	1.016	4.134	0.225	0.211	0.290	0.304	2.090
Collier	LOW	4.472	2.406	6.406	0.570	0.521	0.831	0.782	4.792
	AVERAGE	6.221	5.244	19.826	1.001	0.817	1.564	1.528	6.530
	HIGH	10.002	9.218	44.602	2.900	1.536	3.081	2.836	9.662
Columbia	LOW	0.676	0.653	1.582	0.163	0.149	0.224	0.232	1.464
	AVERAGE	0.818	0.799	2.179	0.176	0.163	0.256	0.239	1.679
	HIGH	0.894	0.854	2.423	0.198	0.195	0.262	0.247	1.705
De Soto	LOW	4.124	4.120	5.709	0.494	0.352	1.045	0.511	4.167
	AVERAGE	4.640	4.447	8.962	0.592	0.588	1.048	0.954	4.275
	HIGH	5.088	4.507	11.200	0.715	0.682	1.053	0.970	4.284
Dixie	LOW	1.074	1.040	3.205	0.250	0.197	0.289	0.293	1.571
	AVERAGE	1.228	1.067	3.452	0.276	0.215	0.520	0.491	2.645
	HIGH	2.168	1.996	11.124	0.529	0.216	0.653	0.576	2.950
Duval	LOW	0.579	0.633	0.626	0.123	0.132	0.153	0.142	0.055
	AVERAGE	0.848	0.803	2.229	0.186	0.171	0.246	0.259	2.138

Modeling Organization: Florida International University
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
	HIGH	1.849	1.599	5.829	0.520	0.385	0.528	0.431	3.561
Escambia	LOW	1.727	1.723	2.434	0.430	0.401	0.591	0.432	2.655
	AVERAGE	2.949	2.866	11.129	0.850	0.805	1.514	1.472	6.042
	HIGH	5.038	4.070	22.669	1.837	1.381	2.465	3.135	7.338
Flagler	LOW	1.849	1.539	3.731	0.249	0.243	0.558	0.272	1.950
	AVERAGE	2.838	2.062	7.088	0.458	0.338	0.944	0.573	3.241
	HIGH	5.686	4.742	17.515	1.213	1.019	1.880	1.442	5.232
Franklin	LOW	2.393	2.460	7.775	0.750	0.697	0.565	0.885	3.231
	AVERAGE	2.866	2.764	9.779	0.958	0.864	0.742	1.200	5.062
	HIGH	3.045	3.112	10.862	1.051	1.243	1.376	1.288	5.851
Gadsen	LOW	0.598	0.625	1.663	0.157	0.137	0.000	0.000	1.520
	AVERAGE	0.768	0.761	2.027	0.178	0.162	0.000	0.000	1.647
	HIGH	0.913	0.895	2.941	0.209	0.179	0.000	0.000	1.891
Gilchrist	LOW	0.974	0.935	2.552	0.217	0.200	0.000	0.295	1.914
	AVERAGE	1.071	1.047	2.993	0.253	0.226	0.000	0.295	1.914
	HIGH	1.111	1.101	3.208	0.260	0.243	0.000	0.295	1.914
Glades	LOW	6.079	5.619	14.693	0.992	0.805	0.000	0.000	0.000
	AVERAGE	6.082	5.627	14.795	0.992	0.805	0.000	0.000	0.000
	HIGH	6.141	5.960	18.156	0.992	0.805	0.000	0.000	0.000
Gulf	LOW	1.446	1.470	5.100	0.362	0.308	0.690	0.538	4.188
	AVERAGE	2.061	2.094	6.629	0.585	0.542	0.690	0.763	4.188
	HIGH	2.184	2.335	9.643	0.654	0.636	0.690	0.817	4.188

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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Hamilton	LOW	0.538	0.523	1.286	0.115	0.109	0.000	0.138	1.134
	AVERAGE	0.597	0.584	1.357	0.133	0.128	0.000	0.138	1.247
	HIGH	0.644	0.621	1.425	0.137	0.130	0.000	0.138	1.290
Hardee	LOW	4.273	4.106	8.554	0.520	0.479	0.000	0.000	4.425
	AVERAGE	4.377	4.191	8.683	0.554	0.508	0.000	0.000	4.472
	HIGH	4.925	4.535	10.198	0.818	0.715	0.000	0.000	5.372
Hendry	LOW	5.012	3.572	10.234	0.619	0.553	1.141	0.978	6.449
	AVERAGE	5.634	5.346	14.403	0.769	0.877	1.574	1.619	6.777
	HIGH	6.976	6.529	16.579	1.159	1.049	1.832	1.721	6.808
Hernando	LOW	2.239	1.749	1.522	0.322	0.276	0.601	0.643	2.498
	AVERAGE	3.008	2.693	6.748	0.344	0.320	0.645	0.703	2.628
	HIGH	4.060	3.536	8.733	0.415	0.447	0.715	0.771	4.783
Highlands	LOW	3.788	3.582	7.197	0.424	0.453	0.743	0.835	4.266
	AVERAGE	4.265	4.060	9.876	0.511	0.494	0.918	0.917	4.460
	HIGH	5.484	5.015	15.110	0.621	0.727	1.131	1.128	5.606
Hillsborough	LOW	1.503	1.544	1.542	0.332	0.276	0.462	0.471	0.601
	AVERAGE	3.340	3.202	9.228	0.414	0.404	0.713	0.731	3.342
	HIGH	5.031	5.393	16.697	0.690	0.725	1.168	1.658	4.369
Holmes	LOW	1.306	1.152	4.071	0.328	0.307	0.360	0.000	2.481
	AVERAGE	1.453	1.412	4.135	0.342	0.307	0.360	0.000	2.481
	HIGH	1.485	1.415	6.194	0.390	0.307	0.360	0.000	2.481

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Indian River	LOW	3.982	2.717	10.405	0.596	0.463	1.091	1.006	3.841
	AVERAGE	7.521	5.757	16.246	2.116	1.596	2.807	3.113	10.349
	HIGH	12.967	10.142	38.392	3.684	3.089	5.652	4.584	13.939
Jackson	LOW	0.893	0.878	2.269	0.195	0.167	0.295	0.276	1.611
	AVERAGE	1.105	1.065	3.071	0.248	0.227	0.295	0.299	2.353
	HIGH	1.411	1.390	4.478	0.329	0.320	0.295	0.336	2.625
Jefferson	LOW	0.593	0.578	1.565	0.133	0.123	0.000	0.000	1.353
	AVERAGE	0.629	0.617	1.671	0.142	0.129	0.000	0.000	1.353
	HIGH	0.747	0.691	2.718	0.168	0.144	0.000	0.000	1.353
Lafayette	LOW	0.797	0.741	2.114	0.186	0.167	0.000	0.000	0.000
	AVERAGE	0.815	0.803	2.119	0.186	0.167	0.000	0.000	0.000
	HIGH	0.815	0.803	2.643	0.186	0.167	0.000	0.000	0.000
Lake	LOW	1.947	1.353	1.409	0.212	0.220	0.294	0.338	1.962
	AVERAGE	2.556	2.315	6.073	0.310	0.295	0.641	0.596	2.522
	HIGH	3.543	3.097	8.940	0.463	0.425	0.712	0.729	3.309
Lee	LOW	2.101	2.930	8.202	0.526	0.389	0.619	0.688	4.465
	AVERAGE	6.633	4.377	17.274	1.048	0.654	1.692	1.284	6.353
	HIGH	10.389	9.339	32.992	2.350	1.905	3.994	3.669	14.232
Leon	LOW	0.727	0.699	0.758	0.143	0.120	0.180	0.166	1.087
	AVERAGE	0.770	0.752	2.414	0.167	0.155	0.216	0.211	1.381
	HIGH	1.005	0.923	4.299	0.200	0.185	0.309	0.245	1.767
Levy	LOW	1.066	1.044	3.282	0.226	0.220	0.293	0.661	1.959

Form A-4A Output Ranges (2007 FHCF Exposure Data)
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
	AVERAGE	1.450	1.264	3.889	0.382	0.273	0.803	0.661	3.864
	HIGH	2.495	2.477	8.339	0.810	0.356	0.811	0.661	4.108
Liberty	LOW	1.068	1.024	2.880	0.239	0.229	0.000	0.000	0.000
	AVERAGE	1.073	1.048	2.934	0.247	0.230	0.000	0.000	0.000
	HIGH	1.088	1.071	3.490	0.249	0.235	0.000	0.000	0.000
Madison	LOW	0.519	0.511	1.256	0.131	0.109	0.000	0.000	1.310
	AVERAGE	0.628	0.615	1.532	0.143	0.128	0.000	0.000	1.310
	HIGH	0.656	0.634	1.773	0.150	0.132	0.000	0.000	1.310
Manatee	LOW	3.505	2.654	1.929	0.457	0.335	0.525	0.597	3.360
	AVERAGE	4.831	3.679	13.348	0.800	0.699	1.863	1.725	6.881
	HIGH	9.448	8.403	36.712	2.258	1.982	3.487	2.930	9.974
Marion	LOW	2.041	1.668	3.700	0.220	0.207	0.386	0.396	1.925
	AVERAGE	2.500	2.117	4.615	0.269	0.255	0.515	0.500	2.411
	HIGH	2.974	2.809	5.844	0.347	0.342	0.710	0.651	3.398
Martin	LOW	5.060	4.393	17.581	0.689	0.694	1.458	1.770	6.453
	AVERAGE	9.115	7.004	31.698	2.652	1.893	3.849	2.995	9.615
	HIGH	12.634	10.970	49.969	3.941	3.672	4.951	3.994	12.567
Miami-Dade	LOW	3.394	2.321	2.904	0.515	0.484	0.719	0.646	4.155
	AVERAGE	7.395	6.662	21.055	2.776	2.253	3.557	3.536	10.856
	HIGH	23.291	14.812	58.917	7.214	5.744	7.321	8.601	19.087
Monroe	LOW	8.884	6.417	11.712	3.576	2.297	4.389	3.446	16.076
	AVERAGE	11.610	11.671	62.843	4.399	3.590	4.944	4.721	19.597

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
	HIGH	18.053	15.712	88.510	6.778	8.810	10.747	6.887	26.001
Nassau	LOW	0.528	0.518	1.293	0.121	0.102	0.124	0.165	1.342
	AVERAGE	0.952	0.873	2.047	0.259	0.238	0.388	0.345	2.586
	HIGH	1.118	1.105	6.457	0.282	0.264	0.389	0.345	3.478
Okaloosa	LOW	1.595	1.590	5.539	0.395	0.372	0.383	0.861	2.410
	AVERAGE	3.332	3.278	9.108	1.081	1.006	1.989	1.684	7.851
	HIGH	5.189	5.436	23.985	1.982	2.627	2.633	2.073	8.560
Okeechobee	LOW	4.762	4.570	11.814	0.650	0.669	1.486	0.701	5.009
	AVERAGE	5.560	5.052	15.696	0.780	0.732	1.486	1.396	6.206
	HIGH	6.106	5.315	16.700	0.902	0.770	1.486	1.407	6.309
Orange	LOW	1.306	1.299	1.465	0.231	0.240	0.318	0.336	0.322
	AVERAGE	2.709	2.634	5.957	0.321	0.312	0.558	0.547	2.492
	HIGH	3.889	3.728	10.129	0.412	0.362	0.685	0.749	3.582
Osceola	LOW	2.079	2.105	6.288	0.307	0.307	0.490	0.413	2.789
	AVERAGE	2.657	2.745	7.456	0.339	0.343	0.611	0.534	3.016
	HIGH	4.688	3.878	11.071	0.810	0.420	0.726	0.760	3.859
Palm Beach	LOW	4.307	4.387	2.869	0.888	0.773	0.797	0.973	4.118
	AVERAGE	8.893	7.009	25.381	3.724	2.074	3.125	2.844	9.876
	HIGH	18.317	13.741	68.802	5.717	4.442	6.934	5.954	16.537
Pasco	LOW	1.724	1.652	2.338	0.313	0.281	0.436	0.375	2.575
	AVERAGE	2.808	2.773	7.607	0.382	0.370	0.729	0.706	3.143
	HIGH	3.813	3.796	10.956	0.537	0.448	0.831	0.791	3.475

Form A-4A Output Ranges (2007 FHCF Exposure Data)
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Pinellas	LOW	2.284	2.186	1.618	0.344	0.333	0.646	0.584	2.769
	AVERAGE	3.824	3.666	11.009	0.510	0.493	1.010	1.005	4.782
	HIGH	6.022	6.327	19.846	1.261	1.045	1.767	1.937	6.796
Polk	LOW	1.640	1.641	1.881	0.291	0.258	0.371	0.365	2.288
	AVERAGE	3.507	3.163	8.377	0.428	0.408	0.730	0.738	3.425
	HIGH	5.379	4.674	12.057	0.633	0.613	0.955	1.136	4.359
Putnam	LOW	0.957	0.916	1.031	0.211	0.194	0.344	0.218	1.340
	AVERAGE	1.107	1.067	3.439	0.247	0.228	0.375	0.317	2.138
	HIGH	1.341	1.282	4.652	0.304	0.293	0.402	0.370	2.594
St. Johns	LOW	0.732	0.754	2.364	0.168	0.155	0.194	0.214	1.483
	AVERAGE	1.256	1.366	4.258	0.383	0.343	0.597	0.601	3.545
	HIGH	2.278	2.002	10.379	0.671	0.565	0.915	0.834	4.501
St. Lucie	LOW	4.820	2.443	5.292	0.652	0.507	0.554	0.786	3.268
	AVERAGE	6.306	4.130	22.273	1.232	0.921	2.637	2.748	8.313
	HIGH	11.701	10.393	33.551	3.838	2.753	5.055	3.870	10.669
Santa Rosa	LOW	1.848	1.804	7.467	0.477	0.441	0.455	0.563	3.789
	AVERAGE	3.289	3.236	11.744	1.199	1.219	3.089	1.811	8.711
	HIGH	6.601	6.092	33.022	2.718	2.106	3.596	2.036	17.609
Sarasota	LOW	1.872	1.909	9.592	0.463	0.397	0.705	0.558	3.309
	AVERAGE	4.984	4.161	16.143	0.731	0.643	1.271	1.197	5.662
	HIGH	7.254	6.357	22.814	1.416	1.140	2.057	1.843	7.440

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Seminole	LOW	2.457	2.087	4.808	0.230	0.247	0.331	0.411	1.905
	AVERAGE	2.826	2.639	6.153	0.307	0.300	0.570	0.547	2.497
	HIGH	3.664	3.246	7.706	0.364	0.367	0.725	0.603	2.729
Sumter	LOW	1.379	1.321	3.976	0.254	0.268	0.518	0.376	1.836
	AVERAGE	1.584	1.721	6.353	0.287	0.301	0.660	0.457	2.168
	HIGH	3.475	3.201	7.633	0.385	0.381	0.682	0.674	3.055
Suwanee	LOW	0.629	0.615	1.533	0.140	0.130	0.191	0.000	1.398
	AVERAGE	0.712	0.696	1.776	0.151	0.144	0.191	0.000	1.513
	HIGH	0.898	0.875	2.303	0.205	0.200	0.191	0.000	1.651
Taylor	LOW	0.726	0.683	1.565	0.156	0.167	0.182	0.330	1.782
	AVERAGE	0.866	0.808	2.569	0.184	0.168	0.261	0.330	2.082
	HIGH	1.228	1.246	4.189	0.359	0.170	0.343	0.330	2.185
Union	LOW	0.847	0.824	0.802	0.191	0.136	0.198	0.238	1.816
	AVERAGE	0.857	0.835	2.321	0.204	0.184	0.198	0.238	1.816
	HIGH	0.987	0.967	2.934	0.233	0.217	0.198	0.238	1.816
Volusia	LOW	1.109	1.431	1.261	0.265	0.267	0.410	0.311	2.077
	AVERAGE	3.723	3.200	7.534	0.502	0.562	1.227	1.615	4.975
	HIGH	7.539	7.342	24.971	1.444	1.415	2.338	2.165	6.181
Wakulla	LOW	0.802	0.808	2.593	0.176	0.180	0.283	0.293	1.441
	AVERAGE	1.012	0.974	2.938	0.230	0.260	0.504	0.642	2.648
	HIGH	1.964	2.024	8.737	0.647	0.665	0.740	0.689	4.139
Walton	LOW	1.622	1.551	4.552	0.383	0.361	0.559	0.367	2.544

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 for 0% Deductible

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Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
	AVERAGE	3.072	2.690	7.575	1.046	1.091	1.993	1.475	7.114
	HIGH	4.321	3.895	31.234	1.585	1.690	2.393	1.686	7.790
Washington	LOW	1.410	1.391	4.025	0.324	0.313	0.487	0.000	2.716
	AVERAGE	1.426	1.408	4.258	0.343	0.319	0.507	0.000	2.716
	HIGH	1.525	1.500	5.133	0.358	0.371	0.544	0.000	2.716
Statewide	LOW	0.519	0.511	0.626	0.115	0.102	0.124	0.138	0.055
	AVERAGE	3.141	4.292	10.002	0.893	1.025	1.427	1.994	7.653
	HIGH	23.291	15.712	88.510	7.214	8.810	10.747	8.601	26.001

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Alachua	LOW	0.130	0.197	0.210	0.039	0.031	0.032	0.033	0.213
	AVERAGE	0.228	0.234	1.952	0.048	0.044	0.052	0.047	0.364
	HIGH	0.336	0.357	4.130	0.086	0.061	0.074	0.087	0.678
Baker	LOW	0.123	0.120	0.755	0.028	0.023	0.022	0.029	0.369
	AVERAGE	0.142	0.138	1.054	0.033	0.031	0.022	0.029	0.369
	HIGH	0.166	0.142	1.202	0.035	0.036	0.022	0.029	0.369
Bay	LOW	0.368	0.369	2.724	0.102	0.078	0.126	0.095	0.584
	AVERAGE	1.232	0.991	7.107	0.353	0.312	1.129	0.575	2.793
	HIGH	2.465	2.400	18.042	1.254	0.847	1.669	0.720	3.691
Bradford	LOW	0.164	0.156	1.128	0.037	0.031	0.071	0.040	0.349
	AVERAGE	0.208	0.198	1.492	0.046	0.039	0.071	0.040	0.416
	HIGH	0.279	0.273	2.332	0.074	0.061	0.071	0.040	0.450
Brevard	LOW	1.560	0.479	4.947	0.048	0.072	0.064	0.125	0.681
	AVERAGE	2.965	2.625	13.563	0.388	0.445	0.753	1.104	3.127
	HIGH	10.072	7.896	30.126	2.620	2.698	3.851	3.189	6.981
Broward	LOW	0.926	0.562	0.791	0.118	0.146	0.100	0.120	0.778
	AVERAGE	4.254	3.431	22.311	1.597	0.876	1.375	1.290	4.727
	HIGH	11.814	10.545	46.721	3.878	3.246	5.980	4.842	10.213
Calhoun	LOW	0.283	0.254	1.461	0.061	0.046	0.000	0.000	0.216
	AVERAGE	0.292	0.272	1.929	0.063	0.047	0.000	0.000	0.216
	HIGH	0.303	0.305	2.070	0.067	0.049	0.000	0.000	0.216

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Charlotte	LOW	1.878	1.034	6.869	0.134	0.112	0.086	0.220	1.403
	AVERAGE	3.160	2.514	10.201	0.383	0.237	0.643	0.300	1.895
	HIGH	4.406	4.135	28.225	0.839	0.565	1.100	0.685	3.551
Citrus	LOW	0.902	0.984	2.939	0.067	0.053	0.048	0.071	0.097
	AVERAGE	1.521	1.208	3.965	0.099	0.078	0.127	0.120	0.496
	HIGH	2.070	1.611	5.640	0.140	0.141	0.147	0.183	0.576
Clay	LOW	0.148	0.151	0.941	0.036	0.034	0.031	0.025	0.248
	AVERAGE	0.177	0.181	1.490	0.043	0.041	0.041	0.036	0.381
	HIGH	0.298	0.251	3.014	0.056	0.055	0.060	0.058	0.483
Collier	LOW	2.369	0.615	4.418	0.157	0.117	0.165	0.157	1.098
	AVERAGE	3.693	2.887	17.144	0.490	0.346	0.586	0.555	2.214
	HIGH	6.916	6.206	41.148	2.119	0.924	1.800	1.599	4.550
Columbia	LOW	0.143	0.129	0.921	0.036	0.031	0.038	0.037	0.240
	AVERAGE	0.184	0.172	1.371	0.040	0.036	0.047	0.039	0.351
	HIGH	0.200	0.188	1.584	0.045	0.044	0.048	0.041	0.377
De Soto	LOW	2.155	2.155	3.924	0.119	0.067	0.256	0.072	0.889
	AVERAGE	2.575	2.405	6.941	0.202	0.203	0.267	0.216	1.024
	HIGH	2.957	2.472	9.026	0.278	0.267	0.274	0.224	1.034
Dixie	LOW	0.287	0.246	2.182	0.067	0.042	0.045	0.047	0.226
	AVERAGE	0.363	0.271	2.406	0.081	0.051	0.168	0.147	0.513
	HIGH	0.901	0.746	9.355	0.215	0.052	0.239	0.187	0.594

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Duval	LOW	0.085	0.102	0.125	0.023	0.025	0.020	0.018	0.003
	AVERAGE	0.212	0.191	1.469	0.047	0.040	0.048	0.051	0.607
	HIGH	0.915	0.599	4.590	0.281	0.151	0.196	0.120	1.423
Escambia	LOW	0.602	0.602	0.841	0.119	0.133	0.129	0.064	0.517
	AVERAGE	1.367	1.288	9.263	0.431	0.400	0.828	0.796	2.428
	HIGH	3.058	2.362	20.199	1.284	0.875	1.654	2.270	3.320
Flagler	LOW	0.821	0.578	2.670	0.067	0.066	0.141	0.038	0.350
	AVERAGE	1.573	0.957	5.732	0.228	0.130	0.423	0.195	1.094
	HIGH	3.866	3.021	15.568	0.849	0.664	1.177	0.786	2.332
Franklin	LOW	1.010	1.040	6.027	0.393	0.351	0.150	0.390	0.570
	AVERAGE	1.323	1.261	7.952	0.545	0.490	0.279	0.621	1.809
	HIGH	1.449	1.506	8.912	0.624	0.803	0.737	0.675	2.295
Gadsen	LOW	0.087	0.107	0.992	0.034	0.029	0.000	0.000	0.257
	AVERAGE	0.172	0.166	1.267	0.040	0.035	0.000	0.000	0.297
	HIGH	0.219	0.207	1.992	0.047	0.040	0.000	0.000	0.375
Gilchrist	LOW	0.239	0.213	1.644	0.052	0.047	0.000	0.051	0.348
	AVERAGE	0.271	0.257	1.996	0.064	0.055	0.000	0.051	0.348
	HIGH	0.284	0.279	2.168	0.067	0.061	0.000	0.051	0.348
Glades	LOW	3.581	3.194	12.197	0.474	0.332	0.000	0.000	0.000
	AVERAGE	3.582	3.202	12.293	0.474	0.332	0.000	0.000	0.000
	HIGH	3.612	3.541	15.484	0.474	0.332	0.000	0.000	0.000

Form A-4A Output Ranges (2007 FHCF Exposure Data)
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Gulf	LOW	0.386	0.406	3.716	0.099	0.065	0.215	0.101	1.273
	AVERAGE	0.701	0.755	5.076	0.239	0.217	0.215	0.265	1.273
	HIGH	0.765	0.890	7.758	0.283	0.278	0.215	0.305	1.273
Hamilton	LOW	0.106	0.096	0.738	0.024	0.021	0.000	0.018	0.185
	AVERAGE	0.128	0.119	0.785	0.028	0.027	0.000	0.018	0.221
	HIGH	0.148	0.132	0.827	0.030	0.028	0.000	0.018	0.234
Hardee	LOW	2.348	2.210	6.555	0.164	0.139	0.000	0.000	1.214
	AVERAGE	2.411	2.262	6.717	0.186	0.157	0.000	0.000	1.241
	HIGH	2.779	2.469	8.099	0.369	0.308	0.000	0.000	1.757
Hendry	LOW	2.781	1.672	8.072	0.210	0.165	0.278	0.142	2.405
	AVERAGE	3.310	3.061	11.995	0.326	0.407	0.589	0.630	2.409
	HIGH	4.376	3.985	14.050	0.629	0.536	0.775	0.706	2.444
Hernando	LOW	0.965	0.590	0.392	0.087	0.056	0.115	0.131	0.439
	AVERAGE	1.575	1.333	5.201	0.097	0.086	0.130	0.154	0.506
	HIGH	2.427	1.909	6.992	0.134	0.163	0.158	0.182	1.783
Highlands	LOW	1.867	1.678	5.372	0.107	0.130	0.120	0.145	1.067
	AVERAGE	2.303	2.138	7.824	0.154	0.149	0.196	0.199	1.129
	HIGH	3.148	2.749	12.609	0.216	0.304	0.297	0.311	1.885
Hillsborough	LOW	0.351	0.344	0.404	0.078	0.053	0.067	0.071	0.111
	AVERAGE	1.751	1.631	7.408	0.126	0.120	0.152	0.158	0.798
	HIGH	3.177	3.379	14.489	0.312	0.333	0.392	0.778	1.290

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Holmes	LOW	0.408	0.285	2.870	0.088	0.082	0.060	0.000	0.443
	AVERAGE	0.450	0.423	2.920	0.098	0.082	0.060	0.000	0.443
	HIGH	0.473	0.425	4.839	0.131	0.082	0.060	0.000	0.443
Indian River	LOW	2.176	1.174	8.461	0.223	0.151	0.330	0.339	1.860
	AVERAGE	5.137	3.605	13.982	1.577	1.104	1.813	2.048	5.986
	HIGH	9.974	7.347	35.307	2.995	2.413	4.336	3.314	8.963
Jackson	LOW	0.215	0.203	1.421	0.040	0.032	0.045	0.041	0.235
	AVERAGE	0.290	0.269	2.044	0.060	0.053	0.045	0.047	0.497
	HIGH	0.436	0.426	3.251	0.094	0.098	0.045	0.054	0.572
Jefferson	LOW	0.125	0.115	0.970	0.026	0.027	0.000	0.000	0.273
	AVERAGE	0.138	0.131	1.048	0.033	0.028	0.000	0.000	0.273
	HIGH	0.184	0.158	1.923	0.042	0.028	0.000	0.000	0.273
Lafayette	LOW	0.193	0.171	1.342	0.046	0.039	0.000	0.000	0.000
	AVERAGE	0.194	0.186	1.347	0.046	0.039	0.000	0.000	0.000
	HIGH	0.216	0.186	1.853	0.046	0.039	0.000	0.000	0.000
Lake	LOW	0.719	0.268	0.341	0.039	0.050	0.039	0.045	0.292
	AVERAGE	1.229	1.017	4.573	0.077	0.071	0.118	0.107	0.433
	HIGH	1.945	1.579	7.204	0.166	0.142	0.141	0.167	0.736
Lee	LOW	0.490	1.110	6.123	0.135	0.075	0.092	0.099	0.949
	AVERAGE	4.136	2.272	14.717	0.562	0.242	0.739	0.430	2.249
	HIGH	7.238	6.390	29.879	1.674	1.269	2.569	2.257	8.512

Form A-4A Output Ranges (2007 FHCF Exposure Data)
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Leon	LOW	0.154	0.135	0.161	0.029	0.021	0.025	0.022	0.165
	AVERAGE	0.180	0.168	1.611	0.038	0.034	0.035	0.034	0.227
	HIGH	0.270	0.238	3.236	0.052	0.048	0.073	0.053	0.367
Levy	LOW	0.261	0.247	2.245	0.043	0.045	0.041	0.215	0.316
	AVERAGE	0.454	0.330	2.703	0.143	0.067	0.325	0.215	1.116
	HIGH	1.102	1.087	6.641	0.437	0.114	0.329	0.215	1.226
Liberty	LOW	0.257	0.230	1.858	0.054	0.046	0.000	0.000	0.000
	AVERAGE	0.258	0.242	1.911	0.059	0.049	0.000	0.000	0.000
	HIGH	0.269	0.259	2.409	0.060	0.061	0.000	0.000	0.000
Madison	LOW	0.107	0.102	0.735	0.026	0.023	0.000	0.000	0.249
	AVERAGE	0.143	0.134	0.928	0.033	0.028	0.000	0.000	0.249
	HIGH	0.151	0.138	1.137	0.036	0.031	0.000	0.000	0.249
Manatee	LOW	1.811	1.070	0.521	0.132	0.067	0.080	0.113	0.660
	AVERAGE	2.861	1.917	11.290	0.415	0.331	0.995	0.854	3.068
	HIGH	6.713	5.727	33.635	1.669	1.401	2.311	1.806	5.353
Marion	LOW	0.955	0.643	2.611	0.046	0.038	0.060	0.048	0.331
	AVERAGE	1.273	0.972	3.330	0.068	0.062	0.093	0.090	0.482
	HIGH	1.623	1.472	4.425	0.105	0.106	0.163	0.141	0.942
Martin	LOW	2.837	2.184	15.050	0.270	0.279	0.514	0.793	2.415
	AVERAGE	6.357	4.485	28.764	2.027	1.317	2.661	1.860	5.028
	HIGH	9.590	8.001	46.543	3.178	2.920	3.659	2.730	7.691

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

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Miami-Dade	LOW	1.363	0.584	0.955	0.121	0.109	0.131	0.102	0.710
	AVERAGE	4.827	4.156	18.508	2.145	1.643	2.475	2.387	6.051
	HIGH	19.369	11.409	55.139	6.322	4.811	5.913	6.841	13.116
Monroe	LOW	6.221	3.816	9.034	2.868	1.688	3.165	2.289	10.763
	AVERAGE	8.711	8.616	59.444	3.645	2.832	3.701	3.405	14.244
	HIGH	14.760	12.360	84.765	5.885	7.793	9.022	5.395	20.026
Nassau	LOW	0.108	0.101	0.768	0.026	0.023	0.016	0.022	0.322
	AVERAGE	0.266	0.230	1.356	0.093	0.083	0.125	0.101	0.838
	HIGH	0.362	0.349	5.276	0.105	0.097	0.125	0.101	1.420
Okaloosa	LOW	0.460	0.447	4.192	0.120	0.114	0.057	0.324	0.359
	AVERAGE	1.709	1.648	7.443	0.632	0.568	1.238	0.967	3.776
	HIGH	3.204	3.451	21.446	1.413	1.979	1.785	1.307	4.287
Okeechobee	LOW	2.631	2.472	9.600	0.248	0.261	0.507	0.129	1.323
	AVERAGE	3.215	2.789	13.173	0.334	0.296	0.507	0.440	1.974
	HIGH	3.614	2.962	14.097	0.415	0.317	0.507	0.445	2.030
Orange	LOW	0.267	0.251	0.360	0.043	0.047	0.043	0.045	0.027
	AVERAGE	1.305	1.250	4.439	0.078	0.076	0.092	0.090	0.436
	HIGH	2.269	2.070	8.183	0.133	0.111	0.127	0.175	0.921
Osceola	LOW	0.774	0.662	4.620	0.068	0.069	0.071	0.057	0.487
	AVERAGE	1.222	1.289	5.710	0.085	0.089	0.112	0.092	0.615
	HIGH	2.689	2.023	8.965	0.388	0.126	0.152	0.162	0.978

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

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Palm Beach	LOW	1.933	2.087	0.861	0.397	0.318	0.159	0.220	1.093
	AVERAGE	5.988	4.324	22.440	2.962	1.435	1.930	1.645	4.941
	HIGH	14.425	10.135	64.571	4.783	3.541	5.433	4.334	10.810
Pasco	LOW	0.466	0.413	1.034	0.067	0.053	0.062	0.052	0.441
	AVERAGE	1.342	1.314	5.930	0.110	0.105	0.159	0.154	0.676
	HIGH	2.094	2.132	9.045	0.207	0.154	0.203	0.189	0.797
Pinellas	LOW	0.925	0.850	0.402	0.087	0.077	0.127	0.080	0.461
	AVERAGE	2.108	1.975	9.141	0.194	0.181	0.341	0.320	1.637
	HIGH	3.767	4.048	17.449	0.782	0.574	0.866	0.960	2.912
Polk	LOW	0.385	0.346	0.498	0.065	0.052	0.050	0.051	0.353
	AVERAGE	1.808	1.540	6.532	0.121	0.112	0.143	0.143	0.773
	HIGH	3.339	2.711	9.898	0.250	0.246	0.214	0.334	1.247
Putnam	LOW	0.238	0.207	0.236	0.045	0.041	0.065	0.030	0.440
	AVERAGE	0.307	0.282	2.411	0.065	0.059	0.085	0.063	0.515
	HIGH	0.425	0.382	3.470	0.088	0.089	0.098	0.089	0.621
St. Johns	LOW	0.118	0.134	1.470	0.034	0.031	0.027	0.032	0.235
	AVERAGE	0.414	0.460	3.190	0.167	0.137	0.249	0.250	1.336
	HIGH	1.081	0.841	8.794	0.373	0.289	0.475	0.411	1.919
St. Lucie	LOW	2.658	0.731	3.472	0.250	0.154	0.082	0.192	0.501
	AVERAGE	3.908	2.117	19.653	0.751	0.492	1.610	1.703	4.030
	HIGH	8.712	7.460	30.536	3.149	2.096	3.781	2.675	5.876

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Santa Rosa	LOW	0.596	0.567	5.859	0.156	0.108	0.069	0.100	1.030
	AVERAGE	1.665	1.594	9.823	0.743	0.756	2.214	1.081	4.519
	HIGH	4.470	3.978	30.183	2.076	1.506	2.668	1.270	12.553
Sarasota	LOW	0.440	0.475	7.607	0.122	0.090	0.154	0.090	0.539
	AVERAGE	2.915	2.233	13.821	0.337	0.266	0.473	0.416	2.074
	HIGH	4.741	3.973	20.074	0.911	0.660	1.083	0.880	3.323
Seminole	LOW	1.153	0.859	3.431	0.043	0.050	0.045	0.058	0.326
	AVERAGE	1.445	1.298	4.629	0.076	0.074	0.099	0.094	0.449
	HIGH	2.146	1.793	6.019	0.110	0.110	0.156	0.108	0.525
Sumter	LOW	0.365	0.324	2.708	0.055	0.062	0.069	0.051	0.264
	AVERAGE	0.512	0.607	4.788	0.069	0.075	0.133	0.072	0.359
	HIGH	1.832	1.660	5.949	0.110	0.108	0.138	0.129	0.690
Suwanee	LOW	0.134	0.125	0.916	0.029	0.028	0.029	0.000	0.221
	AVERAGE	0.159	0.149	1.087	0.034	0.032	0.029	0.000	0.269
	HIGH	0.217	0.203	1.508	0.051	0.052	0.029	0.000	0.292
Taylor	LOW	0.179	0.146	0.924	0.038	0.040	0.025	0.065	0.403
	AVERAGE	0.221	0.198	1.743	0.047	0.042	0.044	0.065	0.403
	HIGH	0.334	0.348	3.036	0.130	0.044	0.063	0.065	0.404
Union	LOW	0.190	0.168	0.165	0.042	0.024	0.027	0.034	0.328
	AVERAGE	0.194	0.179	1.471	0.048	0.042	0.027	0.034	0.328
	HIGH	0.247	0.235	1.983	0.060	0.058	0.027	0.034	0.328

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Volusia	LOW	0.223	0.325	0.299	0.067	0.069	0.069	0.042	0.344
	AVERAGE	2.221	1.802	6.037	0.231	0.290	0.569	0.871	2.012
	HIGH	5.285	5.091	22.492	0.988	0.994	1.400	1.264	2.784
Wakulla	LOW	0.155	0.160	1.728	0.038	0.041	0.051	0.069	0.229
	AVERAGE	0.261	0.245	2.027	0.069	0.092	0.168	0.236	0.721
	HIGH	0.754	0.806	7.118	0.326	0.355	0.292	0.259	1.329
Walton	LOW	0.529	0.495	3.251	0.098	0.100	0.142	0.053	0.452
	AVERAGE	1.447	1.132	5.997	0.608	0.657	1.252	0.813	3.203
	HIGH	2.510	2.110	28.486	1.057	1.152	1.596	0.977	3.681
Washington	LOW	0.407	0.397	2.801	0.071	0.085	0.088	0.000	0.548
	AVERAGE	0.413	0.402	3.003	0.095	0.088	0.109	0.000	0.548
	HIGH	0.480	0.439	3.772	0.101	0.126	0.147	0.000	0.548
Statewide	LOW	0.085	0.096	0.125	0.023	0.021	0.016	0.018	0.003
	AVERAGE	1.711	2.407	8.234	0.560	0.616	0.737	1.062	3.620
	HIGH	19.369	12.360	84.765	6.322	7.793	9.022	6.841	20.026

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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Alachua	LOW	0.844	0.792	0.956	0.178	0.152	0.241	0.217	0.674
	AVERAGE	0.915	0.924	2.738	0.194	0.182	0.274	0.255	1.631
	HIGH	1.161	1.076	5.736	0.252	0.261	0.288	0.318	2.371
Baker	LOW	0.596	0.597	1.222	0.125	0.120	0.000	0.000	0.000
	AVERAGE	0.655	0.650	1.490	0.133	0.125	0.000	0.000	0.000
	HIGH	0.674	0.659	1.708	0.139	0.134	0.000	0.000	0.000
Bay	LOW	1.280	1.296	2.628	0.371	0.269	0.486	0.470	2.365
	AVERAGE	2.481	2.318	6.758	0.583	0.542	1.565	0.964	5.575
	HIGH	3.699	3.780	19.664	1.100	0.977	2.034	1.332	6.541
Bradford	LOW	0.725	0.714	1.601	0.158	0.137	0.000	0.000	0.000
	AVERAGE	0.839	0.832	2.069	0.185	0.164	0.000	0.000	0.000
	HIGH	1.085	1.089	2.844	0.227	0.256	0.000	0.000	0.000
Brevard	LOW	2.596	1.668	1.853	0.314	0.297	0.524	0.455	2.726
	AVERAGE	4.045	3.734	14.167	0.639	0.701	1.135	1.599	5.450
	HIGH	10.565	9.221	33.012	3.038	2.733	4.382	3.325	9.557
Broward	LOW	2.501	2.580	2.778	0.485	0.524	0.692	0.689	3.849
	AVERAGE	5.795	5.056	20.579	1.178	1.124	1.600	2.052	7.555
	HIGH	13.073	14.515	50.336	4.494	2.975	7.916	4.207	12.482
Calhoun	LOW	1.059	1.027	1.966	0.247	0.224	0.000	0.000	0.000
	AVERAGE	1.098	1.066	2.706	0.252	0.254	0.000	0.000	0.000
	HIGH	1.163	1.125	2.940	0.262	0.302	0.000	0.000	0.000
Charlotte	LOW	3.605	3.305	7.145	0.448	0.438	0.592	0.794	3.432
	AVERAGE	4.405	3.830	9.575	0.605	0.531	1.338	0.889	3.758
	HIGH	6.265	5.476	31.102	0.968	0.659	1.816	1.356	5.020
Citrus	LOW	2.155	1.192	4.447	0.253	0.251	0.314	0.315	1.992
	AVERAGE	2.409	2.108	5.213	0.302	0.276	0.563	0.537	2.279
	HIGH	3.665	2.808	7.560	0.370	0.332	0.611	0.718	2.708

LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Clay	LOW	0.705	0.691	1.919	0.141	0.137	0.188	0.176	1.133
	AVERAGE	0.784	0.777	2.193	0.166	0.156	0.218	0.198	1.247
	HIGH	0.958	1.010	4.002	0.200	0.191	0.240	0.262	1.576
Collier	LOW	2.692	2.397	2.468	0.574	0.451	0.584	0.576	3.756
	AVERAGE	4.902	4.316	16.763	0.799	0.727	1.260	1.223	4.816
	HIGH	10.811	7.505	42.921	1.285	1.096	2.383	1.848	7.437
Columbia	LOW	0.641	0.602	1.122	0.149	0.131	0.206	0.209	1.063
	AVERAGE	0.764	0.750	1.853	0.162	0.152	0.224	0.215	1.063
	HIGH	0.838	0.818	2.130	0.183	0.167	0.232	0.226	1.063
De Soto	LOW	2.224	3.488	2.108	0.484	0.429	0.862	0.820	3.418
	AVERAGE	3.680	3.574	8.089	0.495	0.472	0.873	0.881	3.825
	HIGH	4.769	3.814	13.302	0.572	0.670	0.898	0.883	3.834
Dixie	LOW	0.997	0.973	2.852	0.229	0.184	0.256	0.261	1.346
	AVERAGE	1.123	0.994	2.969	0.237	0.220	0.382	0.346	2.077
	HIGH	1.990	1.703	10.438	0.241	0.221	0.446	0.420	2.282
Duval	LOW	0.563	0.572	0.725	0.132	0.110	0.169	0.142	0.972
	AVERAGE	0.832	0.796	2.066	0.177	0.164	0.228	0.244	1.594
	HIGH	1.740	1.602	6.017	0.458	0.354	0.516	0.428	3.526
Escambia	LOW	1.592	1.586	2.986	0.385	0.368	0.456	0.431	3.103
	AVERAGE	2.670	2.650	9.139	0.741	0.714	1.292	1.152	5.663
	HIGH	4.753	3.684	17.757	2.717	1.506	2.716	1.820	6.030
Flagler	LOW	1.552	1.430	3.515	0.217	0.207	0.478	0.261	1.521
	AVERAGE	2.314	1.797	7.074	0.386	0.285	0.700	0.479	2.381
	HIGH	3.921	4.373	10.119	0.925	0.778	1.400	1.146	4.633
Franklin	LOW	2.067	1.815	6.541	0.669	0.395	0.526	0.480	4.075
	AVERAGE	2.485	2.419	8.977	0.739	0.466	0.833	0.687	4.075

Form A-4B Output Ranges (2012 FHCF Exposure Data)

LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
	HIGH	2.685	2.679	11.377	0.768	0.669	1.456	0.885	4.075
Gadsen	LOW	0.585	0.610	1.306	0.140	0.130	0.000	0.157	1.160
	AVERAGE	0.714	0.715	1.851	0.160	0.151	0.000	0.157	1.204
	HIGH	0.861	0.859	2.954	0.202	0.163	0.000	0.157	1.433
Gilchrist	LOW	0.892	0.879	2.161	0.182	0.157	0.000	0.286	0.000
	AVERAGE	1.000	0.982	2.710	0.240	0.219	0.000	0.286	0.000
	HIGH	1.046	1.028	3.015	0.247	0.231	0.000	0.286	0.000
Glades	LOW	5.091	3.499	12.894	0.818	0.684	0.000	0.000	0.000
	AVERAGE	5.112	4.100	12.916	0.818	0.684	0.000	0.000	0.000
	HIGH	5.787	4.108	13.666	0.818	0.684	0.000	0.000	0.000
Gulf	LOW	1.340	1.388	1.937	0.311	0.287	0.564	0.459	3.122
	AVERAGE	1.879	1.934	5.374	0.493	0.465	0.564	0.465	3.122
	HIGH	1.991	2.110	9.030	0.522	0.492	0.564	0.922	3.122
Hamilton	LOW	0.513	0.506	1.174	0.112	0.079	0.000	0.000	0.000
	AVERAGE	0.558	0.558	1.258	0.125	0.113	0.000	0.000	0.000
	HIGH	0.600	0.596	1.353	0.129	0.116	0.000	0.000	0.000
Hardee	LOW	3.311	3.221	7.399	0.423	0.405	0.000	1.071	2.305
	AVERAGE	3.455	3.300	7.772	0.471	0.427	0.000	1.071	2.322
	HIGH	4.178	3.675	8.079	0.703	0.494	0.000	1.071	2.468
Hendry	LOW	4.002	3.509	8.216	0.423	0.498	1.040	0.960	4.393
	AVERAGE	4.510	4.175	12.433	0.765	0.646	1.254	1.115	4.393
	HIGH	5.455	4.923	14.305	0.897	0.764	1.331	1.144	4.393
Hernando	LOW	2.214	1.584	4.832	0.282	0.251	0.572	0.368	1.925
	AVERAGE	2.514	2.247	6.315	0.303	0.288	0.622	0.600	2.300
	HIGH	4.275	3.159	8.518	0.344	0.358	0.763	0.833	2.462
Highlands	LOW	2.955	2.720	6.663	0.379	0.369	0.691	0.676	3.183

Form A-4B Output Ranges (2012 FHCF Exposure Data)

LOSS COSTS PER $1000 for 0% Deductible

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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
	AVERAGE	3.478	3.307	8.895	0.458	0.422	0.780	0.767	3.437
	HIGH	4.490	4.568	15.554	0.770	0.600	0.970	0.816	4.139
Hillsborough	LOW	1.592	1.438	1.672	0.303	0.265	0.410	0.391	2.371
	AVERAGE	2.816	2.800	8.282	0.365	0.362	0.606	0.631	2.924
	HIGH	3.690	4.901	14.202	0.643	0.635	0.983	1.101	5.086
Holmes	LOW	1.052	1.263	3.738	0.301	0.284	0.518	0.000	1.921
	AVERAGE	1.324	1.307	3.795	0.308	0.284	0.518	0.000	1.921
	HIGH	1.351	1.342	4.504	0.330	0.284	0.518	0.000	1.921
Indian River	LOW	2.186	2.362	8.407	0.416	0.397	0.990	0.645	3.618
	AVERAGE	5.654	4.619	14.589	1.657	1.393	2.067	2.412	7.010
	HIGH	10.103	7.741	50.833	2.914	2.394	4.191	3.442	9.166
Jackson	LOW	0.867	0.849	1.900	0.202	0.155	0.000	0.265	1.409
	AVERAGE	1.020	1.016	2.666	0.228	0.219	0.000	0.312	1.690
	HIGH	1.282	1.292	3.860	0.296	0.282	0.000	0.457	1.804
Jefferson	LOW	0.541	0.523	1.223	0.113	0.110	0.184	0.000	0.000
	AVERAGE	0.603	0.599	1.501	0.131	0.123	0.184	0.000	0.000
	HIGH	0.728	0.690	1.878	0.131	0.144	0.184	0.000	0.000
Lafayette	LOW	0.715	0.724	0.690	0.182	0.141	0.290	0.000	0.000
	AVERAGE	0.753	0.745	1.968	0.182	0.141	0.290	0.000	0.000
	HIGH	0.754	0.745	1.978	0.182	0.141	0.290	0.000	0.000
Lake	LOW	1.623	1.504	3.878	0.198	0.196	0.365	0.337	1.699
	AVERAGE	2.093	1.970	5.877	0.278	0.263	0.497	0.478	2.164
	HIGH	2.771	2.609	8.767	0.385	0.410	0.608	0.517	2.577
Lee	LOW	2.055	2.036	2.306	0.387	0.376	0.603	0.540	2.746
	AVERAGE	4.958	3.424	16.231	0.691	0.529	1.258	0.957	4.865
	HIGH	7.741	6.906	29.662	2.142	1.648	2.772	2.396	9.444

Form A-4B Output Ranges (2012 FHCF Exposure Data)

LOSS COSTS PER $1000 for 0% Deductible

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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Leon	LOW	0.691	0.636	0.768	0.132	0.123	0.175	0.170	0.930
	AVERAGE	0.731	0.720	2.185	0.154	0.143	0.199	0.195	1.197
	HIGH	0.974	0.937	3.173	0.189	0.169	0.229	0.251	1.557
Levy	LOW	0.989	0.966	3.027	0.228	0.164	0.764	0.746	1.960
	AVERAGE	1.309	1.182	3.518	0.357	0.261	0.764	0.746	2.999
	HIGH	2.174	2.125	7.550	0.833	1.094	0.764	0.746	3.074
Liberty	LOW	0.989	0.876	2.218	0.224	0.219	0.000	0.000	0.000
	AVERAGE	0.991	0.984	2.494	0.231	0.219	0.000	0.000	0.000
	HIGH	1.042	0.986	3.350	0.232	0.219	0.000	0.000	0.000
Madison	LOW	0.491	0.485	1.250	0.105	0.090	0.000	0.000	0.000
	AVERAGE	0.587	0.576	1.433	0.128	0.117	0.000	0.000	0.000
	HIGH	0.602	0.587	1.554	0.134	0.131	0.000	0.000	0.000
Manatee	LOW	2.523	1.866	1.900	0.391	0.312	0.483	0.498	2.741
	AVERAGE	3.755	3.045	11.076	0.562	0.570	1.242	1.364	5.038
	HIGH	8.171	6.360	32.328	1.674	1.631	2.507	2.368	7.474
Marion	LOW	1.165	1.060	1.160	0.180	0.169	0.275	0.276	1.585
	AVERAGE	1.526	1.444	4.411	0.221	0.212	0.342	0.361	1.781
	HIGH	2.594	1.965	6.209	0.263	0.247	0.495	0.433	3.009
Martin	LOW	4.669	3.928	16.715	0.743	0.491	0.800	1.576	5.351
	AVERAGE	7.896	6.254	30.083	2.400	1.566	3.435	2.708	7.940
	HIGH	10.568	12.882	41.822	3.936	4.427	5.065	4.895	9.588
Miami-Dade	LOW	2.625	2.387	3.029	0.447	0.484	0.665	0.633	2.370
	AVERAGE	6.320	5.713	15.845	2.200	1.733	2.914	2.735	9.310
	HIGH	13.338	12.270	40.289	5.900	5.862	5.056	6.648	17.058
Monroe	LOW	8.750	7.550	48.897	1.931	2.362	4.136	1.924	13.769
	AVERAGE	10.131	10.270	58.142	3.737	2.828	4.277	4.137	16.277
	HIGH	15.980	13.196	83.006	7.980	7.882	9.938	6.004	24.264

Form A-4B Output Ranges (2012 FHCF Exposure Data)

LOSS COSTS PER $1000 for 0% Deductible

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Nassau	LOW	0.505	0.500	1.159	0.113	0.102	0.321	0.130	0.990
	AVERAGE	0.880	0.823	1.871	0.230	0.204	0.321	0.317	1.736
	HIGH	1.021	1.001	3.731	0.249	0.220	0.321	0.317	1.740
Okaloosa	LOW	1.427	1.427	2.351	0.358	0.340	0.397	0.704	2.462
	AVERAGE	2.892	2.827	6.457	0.825	0.808	1.476	1.408	6.187
	HIGH	4.949	4.083	22.265	2.210	2.080	2.157	1.997	6.795
Okeechobee	LOW	3.910	3.082	10.638	0.549	0.540	0.852	0.766	3.097
	AVERAGE	4.429	4.027	14.346	0.648	0.610	0.852	1.119	3.596
	HIGH	4.812	4.292	16.271	0.761	0.658	0.852	1.168	3.602
Orange	LOW	1.291	1.268	1.346	0.238	0.207	0.318	0.337	1.742
	AVERAGE	2.175	2.187	5.448	0.292	0.280	0.471	0.459	2.178
	HIGH	3.434	2.945	10.092	0.376	0.404	0.590	0.696	2.719
Osceola	LOW	1.786	1.370	5.540	0.285	0.247	0.394	0.380	2.070
	AVERAGE	1.930	1.925	6.952	0.299	0.286	0.470	0.434	2.159
	HIGH	2.456	2.700	10.309	0.486	0.334	0.496	0.518	2.731
Palm Beach	LOW	2.890	2.685	2.781	0.648	0.628	0.711	0.659	4.009
	AVERAGE	7.904	6.251	21.400	4.028	2.035	2.692	2.610	8.325
	HIGH	14.703	13.733	48.759	5.372	3.826	6.021	7.147	14.658
Pasco	LOW	1.612	1.635	2.666	0.265	0.247	0.402	0.416	1.927
	AVERAGE	2.333	2.349	6.796	0.321	0.322	0.580	0.592	2.648
	HIGH	3.769	3.179	10.937	0.427	0.378	0.783	0.645	3.076
Pinellas	LOW	1.480	1.435	1.616	0.328	0.277	0.422	0.388	2.330
	AVERAGE	3.682	3.543	10.428	0.451	0.474	0.957	0.965	4.071
	HIGH	6.052	8.090	19.186	1.436	0.983	1.752	1.562	5.860
Polk	LOW	1.576	1.680	1.769	0.274	0.252	0.359	0.344	1.699
	AVERAGE	2.815	2.658	7.841	0.366	0.365	0.560	0.602	2.555

Form A-4B Output Ranges (2012 FHCF Exposure Data)

LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
	HIGH	4.055	4.494	12.032	0.536	0.553	1.076	0.940	3.650
Putnam	LOW	0.902	0.869	2.256	0.187	0.186	0.242	0.230	1.550
	AVERAGE	1.025	1.001	3.499	0.234	0.212	0.317	0.263	1.703
	HIGH	1.213	1.174	5.213	0.313	0.255	0.489	0.345	1.739
St. Johns	LOW	0.718	0.723	1.706	0.152	0.147	0.190	0.178	0.996
	AVERAGE	1.140	1.257	3.468	0.310	0.302	0.496	0.526	2.834
	HIGH	2.015	1.827	10.631	0.594	0.477	0.767	0.716	3.609
St. Lucie	LOW	4.164	2.278	10.184	0.503	0.424	0.566	0.574	2.996
	AVERAGE	5.637	3.626	21.027	0.920	0.718	2.062	2.340	6.457
	HIGH	10.480	9.462	34.346	2.937	2.022	3.802	3.109	7.919
Santa Rosa	LOW	1.728	1.699	1.800	0.412	0.398	0.528	1.024	3.012
	AVERAGE	2.837	2.827	8.901	0.969	0.872	2.492	1.450	6.758
	HIGH	5.751	5.199	33.220	2.432	1.537	4.088	1.600	7.963
Sarasota	LOW	1.885	1.853	2.000	0.427	0.370	0.514	0.481	2.472
	AVERAGE	3.954	3.461	14.782	0.591	0.534	1.015	0.966	3.919
	HIGH	6.453	7.120	22.557	1.177	0.989	1.616	1.469	4.988
Seminole	LOW	2.035	1.724	4.397	0.265	0.218	0.344	0.337	1.899
	AVERAGE	2.377	2.313	5.748	0.287	0.276	0.495	0.483	2.121
	HIGH	3.593	3.419	7.525	0.364	0.353	0.602	0.691	2.368
Sumter	LOW	1.350	1.276	4.567	0.228	0.222	0.367	0.354	1.863
	AVERAGE	1.454	1.476	6.020	0.251	0.252	0.486	0.387	2.228
	HIGH	2.912	2.637	7.798	0.376	0.440	0.643	0.516	2.600
Suwanee	LOW	0.608	0.599	1.465	0.137	0.124	0.194	0.195	0.766
	AVERAGE	0.683	0.673	1.671	0.150	0.134	0.194	0.195	1.053
	HIGH	0.852	0.816	2.129	0.188	0.156	0.194	0.195	1.643
Taylor	LOW	0.659	0.568	1.740	0.150	0.145	0.181	0.293	1.398

Form A-4B Output Ranges (2012 FHCF Exposure Data)

LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
	AVERAGE	0.804	0.759	2.170	0.181	0.154	0.300	0.293	1.398
	HIGH	1.121	1.151	4.061	0.250	0.233	0.307	0.293	1.398
Union	LOW	0.806	0.794	0.805	0.176	0.162	0.200	0.189	1.003
	AVERAGE	0.811	0.801	1.949	0.183	0.164	0.200	0.189	1.003
	HIGH	0.920	0.893	3.539	0.206	0.190	0.200	0.189	1.003
Volusia	LOW	1.179	1.077	1.259	0.183	0.190	0.322	0.345	1.182
	AVERAGE	2.546	2.287	6.064	0.353	0.554	0.809	1.062	4.040
	HIGH	7.548	4.713	23.005	1.031	1.066	2.332	1.394	5.010
Wakulla	LOW	0.774	0.785	2.414	0.168	0.147	0.259	0.591	1.336
	AVERAGE	0.930	0.915	2.639	0.184	0.183	0.382	0.591	2.623
	HIGH	1.789	1.804	8.230	0.379	0.525	0.468	0.591	2.882
Walton	LOW	1.384	1.453	3.823	0.339	0.326	0.492	0.319	2.216
	AVERAGE	2.707	2.456	6.133	0.803	0.814	1.699	1.192	5.768
	HIGH	3.895	3.369	31.221	1.355	1.110	2.039	1.343	6.720
Washington	LOW	1.236	1.242	3.102	0.291	0.230	0.340	0.000	1.763
	AVERAGE	1.257	1.263	3.253	0.300	0.273	0.340	0.000	1.763
	HIGH	1.431	1.398	4.315	0.349	0.308	0.340	0.000	1.763
Statewide	LOW	0.491	0.485	0.690	0.105	0.079	0.169	0.130	0.674
	AVERAGE	2.551	3.659	7.719	0.809	0.896	0.977	1.662	6.423
	HIGH	15.980	14.515	83.006	7.980	7.882	9.938	7.147	24.264

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Alachua	LOW	0.120	0.097	0.210	0.036	0.027	0.033	0.029	0.003
	AVERAGE	0.186	0.193	1.778	0.041	0.039	0.045	0.040	0.104
	HIGH	0.316	0.307	4.392	0.068	0.074	0.050	0.048	0.301
Baker	LOW	0.102	0.103	0.633	0.025	0.022	0.000	0.000	0.000
	AVERAGE	0.115	0.112	0.837	0.027	0.024	0.000	0.000	0.000
	HIGH	0.136	0.119	1.010	0.028	0.031	0.000	0.000	0.000
Bay	LOW	0.254	0.303	1.496	0.097	0.055	0.087	0.071	0.037
	AVERAGE	1.004	0.885	5.153	0.236	0.210	0.917	0.411	1.992
	HIGH	1.963	2.029	17.317	0.664	0.561	1.337	0.706	2.753
Bradford	LOW	0.148	0.141	0.902	0.034	0.027	0.000	0.000	0.000
	AVERAGE	0.182	0.177	1.260	0.041	0.034	0.000	0.000	0.000
	HIGH	0.255	0.259	1.825	0.047	0.069	0.000	0.000	0.000
Brevard	LOW	1.240	0.364	0.513	0.077	0.072	0.095	0.066	0.210
	AVERAGE	2.295	2.023	12.128	0.307	0.360	0.482	0.834	1.825
	HIGH	7.796	6.518	30.003	2.406	2.091	3.216	2.254	4.714
Broward	LOW	0.657	0.700	0.830	0.118	0.147	0.109	0.110	0.423
	AVERAGE	3.369	2.725	17.890	0.697	0.632	0.723	1.089	3.022
	HIGH	9.769	11.272	46.633	3.701	2.297	6.403	2.927	6.973
Calhoun	LOW	0.208	0.211	0.977	0.054	0.050	0.000	0.000	0.000
	AVERAGE	0.243	0.222	1.656	0.056	0.065	0.000	0.000	0.000
	HIGH	0.264	0.239	1.865	0.057	0.090	0.000	0.000	0.000

Form A-4B Output Ranges (2012 FHCF Exposure Data)

LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Charlotte	LOW	1.761	1.480	5.192	0.115	0.110	0.087	0.160	0.308
	AVERAGE	2.363	1.909	7.513	0.218	0.168	0.504	0.218	0.476
	HIGH	3.880	3.112	28.048	0.479	0.237	0.843	0.423	1.041
Citrus	LOW	0.972	0.232	3.137	0.056	0.053	0.042	0.042	0.051
	AVERAGE	1.136	0.914	3.827	0.077	0.066	0.106	0.097	0.129
	HIGH	2.149	1.453	5.983	0.116	0.090	0.122	0.164	0.205
Clay	LOW	0.114	0.081	1.136	0.028	0.028	0.026	0.023	0.000
	AVERAGE	0.142	0.142	1.360	0.035	0.033	0.033	0.029	0.043
	HIGH	0.277	0.251	2.898	0.044	0.043	0.042	0.053	0.082
Collier	LOW	0.846	0.611	0.687	0.176	0.109	0.085	0.083	0.387
	AVERAGE	2.619	2.136	14.247	0.332	0.283	0.410	0.390	0.902
	HIGH	7.635	4.745	39.518	0.714	0.557	1.157	0.818	2.647
Columbia	LOW	0.121	0.096	0.532	0.029	0.025	0.030	0.033	0.006
	AVERAGE	0.149	0.142	1.089	0.035	0.032	0.035	0.034	0.006
	HIGH	0.186	0.160	1.325	0.038	0.037	0.038	0.036	0.006
De Soto	LOW	0.662	1.671	0.579	0.139	0.106	0.180	0.131	0.266
	AVERAGE	1.817	1.718	6.130	0.144	0.133	0.182	0.193	0.380
	HIGH	2.672	1.974	10.981	0.183	0.267	0.183	0.194	0.383
Dixie	LOW	0.229	0.196	1.861	0.052	0.033	0.036	0.036	0.001
	AVERAGE	0.285	0.216	1.971	0.060	0.058	0.066	0.057	0.061
	HIGH	0.738	0.494	8.696	0.063	0.058	0.082	0.073	0.078

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Duval	LOW	0.081	0.066	0.151	0.025	0.020	0.023	0.018	0.015
	AVERAGE	0.202	0.187	1.325	0.043	0.038	0.040	0.045	0.164
	HIGH	0.726	0.605	4.830	0.218	0.129	0.187	0.117	1.289
Escambia	LOW	0.504	0.500	1.237	0.116	0.110	0.070	0.064	0.464
	AVERAGE	1.135	1.108	7.384	0.345	0.330	0.652	0.540	1.986
	HIGH	2.808	1.896	15.514	2.060	1.004	1.877	1.112	2.218
Flagler	LOW	0.589	0.496	2.478	0.048	0.044	0.097	0.036	0.036
	AVERAGE	1.143	0.746	5.703	0.169	0.094	0.260	0.138	0.405
	HIGH	2.338	2.723	8.460	0.594	0.459	0.791	0.575	1.733
Franklin	LOW	0.731	0.463	4.864	0.325	0.111	0.123	0.080	0.734
	AVERAGE	0.999	0.967	7.186	0.363	0.149	0.352	0.222	0.734
	HIGH	1.204	1.133	9.532	0.393	0.316	0.817	0.330	0.734
Gadsen	LOW	0.077	0.095	0.685	0.028	0.026	0.000	0.020	0.015
	AVERAGE	0.139	0.138	1.119	0.034	0.031	0.000	0.020	0.019
	HIGH	0.183	0.183	2.003	0.044	0.035	0.000	0.020	0.039
Gilchrist	LOW	0.185	0.177	1.304	0.036	0.030	0.000	0.043	0.000
	AVERAGE	0.218	0.207	1.745	0.057	0.052	0.000	0.043	0.000
	HIGH	0.232	0.220	1.990	0.059	0.057	0.000	0.043	0.000
Glades	LOW	2.789	1.532	10.498	0.349	0.231	0.000	0.000	0.000
	AVERAGE	2.808	1.964	10.519	0.349	0.231	0.000	0.000	0.000
	HIGH	3.414	1.969	11.241	0.349	0.231	0.000	0.000	0.000

Form A-4B Output Ranges (2012 FHCF Exposure Data)

LOSS COSTS PER $1000 with Specified Deductibles

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Gulf	LOW	0.304	0.342	0.527	0.074	0.073	0.132	0.072	0.512
	AVERAGE	0.556	0.619	3.929	0.170	0.166	0.132	0.076	0.512
	HIGH	0.609	0.708	7.184	0.185	0.180	0.132	0.398	0.512
Hamilton	LOW	0.087	0.083	0.645	0.022	0.013	0.000	0.000	0.000
	AVERAGE	0.103	0.102	0.701	0.026	0.022	0.000	0.000	0.000
	HIGH	0.117	0.114	0.763	0.027	0.022	0.000	0.000	0.000
Hardee	LOW	1.570	1.498	5.558	0.107	0.104	0.000	0.281	0.003
	AVERAGE	1.683	1.567	5.878	0.136	0.114	0.000	0.281	0.003
	HIGH	2.186	1.770	6.136	0.283	0.142	0.000	0.281	0.004
Hendry	LOW	2.011	1.622	6.172	0.084	0.131	0.230	0.220	0.589
	AVERAGE	2.395	2.109	10.132	0.329	0.234	0.407	0.351	0.589
	HIGH	3.109	2.658	11.882	0.428	0.316	0.470	0.375	0.589
Hernando	LOW	0.948	0.462	3.448	0.065	0.049	0.110	0.050	0.022
	AVERAGE	1.189	0.988	4.807	0.075	0.069	0.125	0.120	0.144
	HIGH	2.627	1.772	6.791	0.093	0.102	0.174	0.226	0.193
Highlands	LOW	1.258	1.064	4.885	0.074	0.086	0.132	0.103	0.234
	AVERAGE	1.682	1.548	6.923	0.125	0.110	0.159	0.148	0.291
	HIGH	2.338	2.491	13.049	0.332	0.191	0.231	0.165	0.372
Hillsborough	LOW	0.372	0.315	0.457	0.063	0.051	0.057	0.054	0.087
	AVERAGE	1.339	1.317	6.528	0.097	0.096	0.114	0.124	0.310
	HIGH	1.977	3.031	12.070	0.261	0.272	0.279	0.361	1.399

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

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Holmes	LOW	0.216	0.338	2.555	0.078	0.068	0.146	0.000	0.058
	AVERAGE	0.354	0.343	2.606	0.081	0.068	0.146	0.000	0.058
	HIGH	0.367	0.386	3.315	0.091	0.068	0.146	0.000	0.058
Indian River	LOW	0.607	0.902	6.602	0.126	0.111	0.316	0.134	0.561
	AVERAGE	3.524	2.624	12.414	1.166	0.928	1.230	1.503	2.881
	HIGH	7.381	5.202	47.441	2.272	1.785	3.070	2.366	4.472
Jackson	LOW	0.189	0.178	1.093	0.041	0.029	0.000	0.038	0.011
	AVERAGE	0.232	0.228	1.684	0.052	0.050	0.000	0.064	0.051
	HIGH	0.337	0.346	2.730	0.077	0.073	0.000	0.130	0.066
Jefferson	LOW	0.097	0.085	0.684	0.020	0.021	0.028	0.000	0.000
	AVERAGE	0.116	0.113	0.896	0.027	0.027	0.028	0.000	0.000
	HIGH	0.169	0.149	1.184	0.028	0.028	0.028	0.000	0.000
Lafayette	LOW	0.159	0.153	0.146	0.045	0.027	0.070	0.000	0.000
	AVERAGE	0.159	0.153	1.227	0.045	0.027	0.070	0.000	0.000
	HIGH	0.167	0.167	1.235	0.045	0.027	0.070	0.000	0.000
Lake	LOW	0.651	0.514	2.748	0.041	0.041	0.050	0.045	0.001
	AVERAGE	0.869	0.757	4.393	0.061	0.057	0.078	0.074	0.083
	HIGH	1.373	1.292	6.964	0.105	0.136	0.106	0.084	0.156
Lee	LOW	0.474	0.459	0.618	0.081	0.080	0.089	0.077	0.024
	AVERAGE	2.729	1.509	13.740	0.280	0.156	0.468	0.259	0.922
	HIGH	4.908	4.301	26.647	1.458	1.021	1.616	1.287	4.050

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

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Leon	LOW	0.126	0.076	0.163	0.024	0.023	0.023	0.023	0.000
	AVERAGE	0.153	0.145	1.412	0.033	0.030	0.030	0.029	0.050
	HIGH	0.266	0.256	2.278	0.041	0.037	0.039	0.041	0.092
Levy	LOW	0.213	0.129	1.925	0.046	0.029	0.295	0.292	0.080
	AVERAGE	0.345	0.268	2.366	0.129	0.067	0.295	0.292	0.393
	HIGH	0.821	0.785	5.891	0.463	0.697	0.295	0.292	0.415
Liberty	LOW	0.202	0.120	1.283	0.050	0.049	0.000	0.000	0.000
	AVERAGE	0.203	0.200	1.524	0.053	0.049	0.000	0.000	0.000
	HIGH	0.236	0.201	2.284	0.053	0.049	0.000	0.000	0.000
Madison	LOW	0.087	0.084	0.730	0.020	0.015	0.000	0.000	0.000
	AVERAGE	0.117	0.110	0.849	0.027	0.024	0.000	0.000	0.000
	HIGH	0.123	0.115	0.949	0.029	0.029	0.000	0.000	0.000
Manatee	LOW	1.031	0.527	0.514	0.092	0.061	0.071	0.074	0.131
	AVERAGE	1.979	1.409	9.139	0.227	0.236	0.554	0.622	1.433
	HIGH	5.513	3.980	29.349	1.145	1.101	1.508	1.384	3.002
Marion	LOW	0.278	0.195	0.266	0.034	0.031	0.036	0.036	0.010
	AVERAGE	0.519	0.453	3.147	0.045	0.042	0.049	0.051	0.057
	HIGH	1.339	0.843	4.756	0.069	0.056	0.094	0.059	0.471
Martin	LOW	2.541	1.927	14.260	0.324	0.122	0.186	0.674	1.200
	AVERAGE	5.292	3.842	27.201	1.803	1.033	2.322	1.649	3.304
	HIGH	7.672	9.805	38.621	3.173	3.608	3.677	3.523	4.613

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Miami-Dade	LOW	0.743	0.609	1.073	0.095	0.108	0.108	0.099	0.255
	AVERAGE	3.901	3.361	13.536	1.631	1.178	1.950	1.737	4.615
	HIGH	10.221	9.123	36.997	5.055	4.824	3.909	5.141	10.754
Monroe	LOW	6.177	4.722	45.810	1.358	1.702	3.006	0.937	8.580
	AVERAGE	7.379	7.364	54.854	3.031	2.134	3.101	2.905	10.934
	HIGH	12.812	10.070	79.337	6.973	6.830	8.386	4.639	18.025
Nassau	LOW	0.091	0.088	0.651	0.023	0.022	0.083	0.017	0.030
	AVERAGE	0.213	0.191	1.202	0.074	0.061	0.083	0.084	0.179
	HIGH	0.291	0.260	2.768	0.082	0.068	0.083	0.084	0.217
Okaloosa	LOW	0.277	0.365	0.780	0.089	0.090	0.059	0.210	0.169
	AVERAGE	1.329	1.277	4.965	0.424	0.409	0.809	0.744	2.307
	HIGH	3.008	2.253	19.755	1.582	1.501	1.370	1.247	2.658
Okeechobee	LOW	1.938	1.215	8.486	0.172	0.170	0.184	0.133	0.069
	AVERAGE	2.299	1.963	11.903	0.237	0.210	0.184	0.322	0.194
	HIGH	2.565	2.137	13.784	0.312	0.238	0.184	0.349	0.196
Orange	LOW	0.262	0.242	0.320	0.048	0.038	0.043	0.045	0.002
	AVERAGE	0.897	0.910	3.979	0.065	0.061	0.070	0.069	0.127
	HIGH	1.961	1.535	8.147	0.114	0.131	0.103	0.145	0.290
Osceola	LOW	0.535	0.269	3.941	0.056	0.046	0.054	0.052	0.043
	AVERAGE	0.644	0.643	5.246	0.063	0.059	0.067	0.062	0.091
	HIGH	0.854	1.052	8.251	0.147	0.072	0.071	0.076	0.181

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Palm Beach	LOW	0.834	0.678	0.813	0.214	0.189	0.118	0.104	0.381
	AVERAGE	5.135	3.690	18.620	3.257	1.408	1.584	1.492	3.353
	HIGH	11.114	10.243	45.064	4.486	2.986	4.661	5.395	8.426
Pasco	LOW	0.376	0.402	1.340	0.050	0.046	0.056	0.058	0.024
	AVERAGE	0.972	0.988	5.182	0.075	0.078	0.108	0.117	0.211
	HIGH	2.056	1.620	9.029	0.149	0.113	0.198	0.138	0.293
Pinellas	LOW	0.333	0.303	0.401	0.079	0.053	0.061	0.055	0.049
	AVERAGE	1.997	1.881	8.603	0.155	0.169	0.315	0.303	1.012
	HIGH	3.788	5.674	16.822	0.929	0.529	0.859	0.683	2.039
Polk	LOW	0.366	0.359	0.465	0.057	0.050	0.049	0.046	0.010
	AVERAGE	1.271	1.150	6.043	0.089	0.090	0.095	0.106	0.134
	HIGH	2.217	2.551	9.877	0.196	0.206	0.301	0.232	0.420
Putnam	LOW	0.196	0.174	1.390	0.035	0.037	0.033	0.031	0.012
	AVERAGE	0.248	0.233	2.468	0.060	0.052	0.061	0.042	0.083
	HIGH	0.334	0.306	3.986	0.100	0.069	0.151	0.056	0.087
St. Johns	LOW	0.106	0.096	0.867	0.031	0.028	0.026	0.024	0.013
	AVERAGE	0.327	0.373	2.473	0.119	0.112	0.179	0.199	0.694
	HIGH	0.857	0.696	9.039	0.311	0.223	0.358	0.321	1.073
St. Lucie	LOW	2.126	0.602	8.053	0.141	0.098	0.084	0.090	0.042
	AVERAGE	3.354	1.708	18.452	0.484	0.329	1.141	1.397	2.412
	HIGH	7.628	6.643	31.308	2.317	1.434	2.691	2.052	3.433

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Santa Rosa	LOW	0.489	0.471	0.514	0.114	0.106	0.111	0.459	0.574
	AVERAGE	1.263	1.245	7.132	0.546	0.464	1.687	0.777	2.713
	HIGH	3.683	3.162	30.377	1.815	1.007	3.096	0.899	3.618
Sarasota	LOW	0.452	0.433	0.528	0.094	0.080	0.074	0.072	0.040
	AVERAGE	2.087	1.677	12.544	0.233	0.192	0.329	0.300	0.688
	HIGH	4.189	4.617	19.850	0.709	0.543	0.777	0.647	1.267
Seminole	LOW	0.820	0.584	3.062	0.057	0.040	0.047	0.047	0.079
	AVERAGE	1.102	1.052	4.267	0.067	0.064	0.079	0.077	0.151
	HIGH	2.092	1.942	5.866	0.108	0.107	0.110	0.147	0.209
Sumter	LOW	0.341	0.288	3.174	0.044	0.043	0.051	0.048	0.018
	AVERAGE	0.411	0.419	4.477	0.053	0.054	0.079	0.056	0.108
	HIGH	1.398	1.183	6.094	0.107	0.138	0.099	0.089	0.169
Suwanee	LOW	0.114	0.108	0.851	0.030	0.025	0.029	0.026	0.000
	AVERAGE	0.134	0.126	0.993	0.033	0.028	0.029	0.026	0.030
	HIGH	0.185	0.161	1.381	0.043	0.030	0.029	0.026	0.124
Taylor	LOW	0.131	0.067	1.050	0.031	0.032	0.025	0.048	0.002
	AVERAGE	0.178	0.165	1.395	0.043	0.035	0.049	0.048	0.002
	HIGH	0.290	0.318	2.918	0.062	0.057	0.050	0.048	0.002
Union	LOW	0.155	0.144	0.166	0.038	0.035	0.027	0.025	0.000
	AVERAGE	0.157	0.150	1.140	0.040	0.035	0.027	0.025	0.000
	HIGH	0.219	0.184	2.511	0.048	0.038	0.027	0.025	0.000

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Volusia	LOW	0.240	0.204	0.298	0.034	0.035	0.045	0.052	0.058
	AVERAGE	1.261	1.061	4.671	0.125	0.288	0.302	0.479	1.172
	HIGH	5.289	2.862	20.601	0.629	0.692	1.399	0.709	1.687
Wakulla	LOW	0.133	0.142	1.568	0.035	0.028	0.041	0.191	0.013
	AVERAGE	0.205	0.201	1.760	0.041	0.045	0.081	0.191	0.268
	HIGH	0.613	0.627	6.643	0.114	0.248	0.109	0.191	0.317
Walton	LOW	0.334	0.376	2.578	0.089	0.083	0.080	0.045	0.105
	AVERAGE	1.124	0.924	4.636	0.408	0.418	1.004	0.580	2.029
	HIGH	2.040	1.561	28.474	0.856	0.648	1.291	0.699	2.727
Washington	LOW	0.298	0.305	1.982	0.067	0.042	0.049	0.000	0.036
	AVERAGE	0.306	0.312	2.105	0.076	0.069	0.049	0.000	0.036
	HIGH	0.393	0.397	3.049	0.100	0.084	0.049	0.000	0.036
Statewide	LOW	0.077	0.066	0.146	0.020	0.013	0.023	0.017	0.000
	AVERAGE	1.240	1.896	6.130	0.500	0.513	0.427	0.840	2.443
	HIGH	12.812	11.272	79.337	6.973	6.830	8.386	5.395	18.025

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: January 22, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Alachua	LOW	0.830	0.909	0.957	0.192	0.173	0.232	0.229	1.393
	AVERAGE	0.973	0.982	2.936	0.211	0.197	0.297	0.278	1.843
	HIGH	1.207	1.229	5.425	0.299	0.245	0.345	0.369	2.712
Baker	LOW	0.604	0.597	1.339	0.131	0.120	0.168	0.202	1.708
	AVERAGE	0.694	0.687	1.739	0.151	0.143	0.168	0.202	1.708
	HIGH	0.715	0.699	1.930	0.158	0.149	0.168	0.202	1.708
Bay	LOW	1.394	1.391	3.991	0.388	0.304	0.573	0.454	3.049
	AVERAGE	2.752	2.442	8.863	0.732	0.675	1.825	1.174	6.316
	HIGH	4.251	4.196	20.417	1.761	1.311	2.415	1.355	7.369
Bradford	LOW	0.743	0.730	1.859	0.167	0.149	0.324	0.272	1.823
	AVERAGE	0.877	0.864	2.337	0.197	0.174	0.324	0.272	1.982
	HIGH	1.112	1.102	3.407	0.274	0.245	0.324	0.272	2.064
Brevard	LOW	3.017	1.949	6.497	0.251	0.298	0.443	0.590	3.258
	AVERAGE	4.855	4.467	15.697	0.742	0.811	1.522	1.977	6.778
	HIGH	13.083	10.767	33.138	3.338	3.403	5.119	4.449	11.797
Broward	LOW	2.941	2.360	2.734	0.509	0.568	0.648	0.701	3.575
	AVERAGE	6.840	5.914	25.198	2.197	1.411	2.411	2.358	9.480
	HIGH	15.076	13.909	50.428	4.606	4.003	7.436	6.472	15.927
Calhoun	LOW	1.135	1.096	2.498	0.260	0.226	0.000	0.000	1.614
	AVERAGE	1.160	1.128	3.004	0.264	0.235	0.000	0.000	1.614
	HIGH	1.196	1.196	3.176	0.277	0.245	0.000	0.000	1.614
Charlotte	LOW	4.005	2.718	8.930	0.486	0.493	0.581	0.943	4.901

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: January 22, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
	AVERAGE	5.383	4.586	12.411	0.825	0.631	1.578	1.077	5.582
	HIGH	6.991	6.683	31.282	1.396	1.067	2.201	1.593	7.871
Citrus	LOW	2.108	2.183	4.200	0.272	0.251	0.349	0.480	0.506
	AVERAGE	2.905	2.490	5.368	0.343	0.303	0.635	0.615	2.566
	HIGH	3.601	3.089	7.202	0.417	0.414	0.697	0.769	2.979
Clay	LOW	0.756	0.775	1.681	0.162	0.154	0.204	0.181	1.506
	AVERAGE	0.834	0.832	2.345	0.185	0.174	0.240	0.216	1.772
	HIGH	1.079	1.016	4.134	0.225	0.211	0.290	0.304	2.058
Collier	LOW	4.472	2.406	6.406	0.570	0.521	0.831	0.782	4.723
	AVERAGE	6.221	5.244	19.826	1.001	0.817	1.564	1.528	6.456
	HIGH	10.002	9.218	44.602	2.900	1.536	3.081	2.836	9.474
Columbia	LOW	0.676	0.653	1.582	0.163	0.149	0.224	0.232	1.463
	AVERAGE	0.818	0.799	2.179	0.176	0.163	0.256	0.239	1.652
	HIGH	0.894	0.854	2.423	0.198	0.195	0.262	0.247	1.674
De Soto	LOW	4.124	4.120	5.709	0.494	0.352	1.045	0.511	4.091
	AVERAGE	4.640	4.447	8.962	0.592	0.588	1.048	0.954	4.225
	HIGH	5.088	4.507	11.200	0.715	0.682	1.053	0.970	4.235
Dixie	LOW	1.074	1.040	3.205	0.250	0.197	0.289	0.293	1.580
	AVERAGE	1.228	1.067	3.452	0.276	0.215	0.520	0.491	2.641
	HIGH	2.168	1.996	11.124	0.529	0.216	0.653	0.576	2.942
Duval	LOW	0.579	0.633	0.626	0.123	0.132	0.153	0.142	0.055
	AVERAGE	0.848	0.803	2.229	0.186	0.171	0.246	0.259	2.106

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: January 22, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
	HIGH	1.849	1.599	5.829	0.520	0.385	0.528	0.431	3.487
Escambia	LOW	1.727	1.723	2.434	0.430	0.401	0.591	0.432	2.601
	AVERAGE	2.949	2.866	11.129	0.850	0.805	1.514	1.472	5.919
	HIGH	5.038	4.070	22.669	1.837	1.381	2.465	3.135	7.411
Flagler	LOW	1.849	1.539	3.731	0.249	0.243	0.558	0.272	1.941
	AVERAGE	2.838	2.062	7.088	0.458	0.338	0.944	0.573	3.191
	HIGH	5.686	4.742	17.515	1.213	1.019	1.880	1.442	5.201
Franklin	LOW	2.393	2.460	7.775	0.750	0.697	0.565	0.885	3.193
	AVERAGE	2.866	2.764	9.779	0.958	0.864	0.742	1.200	4.978
	HIGH	3.045	3.112	10.862	1.051	1.243	1.376	1.288	5.736
Gadsen	LOW	0.598	0.625	1.663	0.157	0.137	0.000	0.000	1.498
	AVERAGE	0.768	0.761	2.027	0.178	0.162	0.000	0.000	1.620
	HIGH	0.913	0.895	2.941	0.209	0.179	0.000	0.000	1.856
Gilchrist	LOW	0.974	0.935	2.552	0.217	0.200	0.000	0.295	1.890
	AVERAGE	1.071	1.047	2.993	0.253	0.226	0.000	0.295	1.890
	HIGH	1.111	1.101	3.208	0.260	0.243	0.000	0.295	1.890
Glades	LOW	6.179	5.619	14.693	0.992	0.805	0.000	0.000	0.000
	AVERAGE	6.182	5.627	14.795	0.992	0.805	0.000	0.000	0.000
	HIGH	6.141	5.960	18.156	0.992	0.805	0.000	0.000	0.000
Gulf	LOW	1.446	1.470	5.100	0.362	0.308	0.690	0.538	4.114
	AVERAGE	2.061	2.094	6.629	0.585	0.542	0.690	0.763	4.114
	HIGH	2.184	2.335	9.643	0.654	0.636	0.690	0.817	4.114

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: January 22, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Hamilton	LOW	0.538	0.523	1.286	0.115	0.109	0.000	0.138	1.119
	AVERAGE	0.597	0.584	1.357	0.133	0.128	0.000	0.138	1.233
	HIGH	0.644	0.621	1.425	0.137	0.130	0.000	0.138	1.276
Hardee	LOW	4.273	4.106	8.554	0.520	0.479	0.000	0.000	4.366
	AVERAGE	4.377	4.191	8.683	0.554	0.508	0.000	0.000	4.410
	HIGH	4.925	4.535	10.198	0.818	0.715	0.000	0.000	5.245
Hendry	LOW	5.012	3.572	10.234	0.619	0.553	1.141	0.978	6.186
	AVERAGE	5.634	5.346	14.403	0.769	0.877	1.574	1.619	6.699
	HIGH	6.976	6.529	16.579	1.159	1.049	1.832	1.721	6.748
Hernando	LOW	2.239	1.749	1.522	0.322	0.276	0.601	0.643	2.468
	AVERAGE	3.008	2.693	6.748	0.344	0.320	0.645	0.703	2.596
	HIGH	4.060	3.536	8.733	0.415	0.447	0.715	0.771	4.634
Highlands	LOW	3.788	3.582	7.197	0.424	0.453	0.743	0.835	4.201
	AVERAGE	4.265	4.060	9.876	0.511	0.494	0.918	0.917	4.391
	HIGH	5.484	5.015	15.110	0.621	0.727	1.131	1.128	5.593
Hillsborough	LOW	1.503	1.544	1.542	0.332	0.276	0.462	0.471	0.601
	AVERAGE	3.340	3.202	9.228	0.414	0.404	0.713	0.731	3.285
	HIGH	5.031	5.393	16.697	0.690	0.725	1.168	1.658	4.275
Holmes	LOW	1.306	1.152	4.071	0.328	0.307	0.360	0.000	2.446
	AVERAGE	1.453	1.412	4.135	0.342	0.307	0.360	0.000	2.446
	HIGH	1.485	1.415	6.194	0.390	0.307	0.360	0.000	2.446

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: January 22, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Indian River	LOW	3.982	2.717	10.405	0.596	0.463	1.091	1.006	3.841
	AVERAGE	7.521	5.757	16.246	2.116	1.596	2.807	3.113	10.308
	HIGH	12.967	10.142	38.392	3.684	3.089	5.652	4.584	13.938
Jackson	LOW	0.893	0.878	2.269	0.195	0.167	0.295	0.276	1.586
	AVERAGE	1.105	1.065	3.071	0.248	0.227	0.295	0.299	2.315
	HIGH	1.411	1.390	4.478	0.329	0.320	0.295	0.336	2.579
Jefferson	LOW	0.593	0.578	1.565	0.133	0.123	0.000	0.000	1.334
	AVERAGE	0.629	0.617	1.671	0.142	0.129	0.000	0.000	1.334
	HIGH	0.747	0.691	2.718	0.168	0.144	0.000	0.000	1.334
Lafayette	LOW	0.797	0.741	2.114	0.186	0.167	0.000	0.000	0.000
	AVERAGE	0.815	0.803	2.119	0.186	0.167	0.000	0.000	0.000
	HIGH	0.815	0.803	2.643	0.186	0.167	0.000	0.000	0.000
Lake	LOW	1.947	1.353	1.409	0.212	0.220	0.294	0.338	1.965
	AVERAGE	2.556	2.315	6.173	0.310	0.295	0.641	0.596	2.495
	HIGH	3.543	3.097	8.940	0.463	0.425	0.712	0.729	3.256
Lee	LOW	2.101	2.930	8.202	0.526	0.389	0.619	0.688	4.393
	AVERAGE	6.633	4.377	17.274	1.048	0.654	1.692	1.284	6.249
	HIGH	10.389	9.339	32.992	2.350	1.905	3.994	3.669	13.762
Leon	LOW	0.727	0.699	0.758	0.143	0.120	0.180	0.166	1.089
	AVERAGE	0.770	0.752	2.414	0.167	0.155	0.216	0.211	1.368
	HIGH	1.005	0.923	4.299	0.200	0.185	0.309	0.245	1.739
Levy	LOW	1.066	1.044	3.282	0.226	0.220	0.293	0.661	1.926

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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
	AVERAGE	1.450	1.264	3.889	0.382	0.273	0.803	0.661	3.794
	HIGH	2.495	2.477	8.339	0.810	0.356	0.811	0.661	4.034
Liberty	LOW	1.068	1.024	2.880	0.239	0.229	0.000	0.000	0.000
	AVERAGE	1.073	1.048	2.934	0.247	0.230	0.000	0.000	0.000
	HIGH	1.088	1.071	3.490	0.249	0.235	0.000	0.000	0.000
Madison	LOW	0.519	0.511	1.256	0.131	0.109	0.000	0.000	1.291
	AVERAGE	0.628	0.615	1.532	0.143	0.128	0.000	0.000	1.291
	HIGH	0.656	0.634	1.773	0.150	0.132	0.000	0.000	1.291
Manatee	LOW	3.505	2.654	1.929	0.457	0.335	0.525	0.597	3.351
	AVERAGE	4.831	3.679	13.348	0.800	0.699	1.863	1.725	6.824
	HIGH	9.448	8.403	36.712	2.258	1.982	3.487	2.930	10.011
Marion	LOW	2.041	1.668	3.700	0.220	0.207	0.386	0.396	1.885
	AVERAGE	2.500	2.117	4.615	0.269	0.255	0.515	0.500	2.379
	HIGH	2.974	2.809	5.844	0.347	0.342	0.710	0.651	3.363
Martin	LOW	5.060	4.393	17.581	0.689	0.694	1.458	1.770	6.331
	AVERAGE	9.115	7.004	31.698	2.652	1.893	3.849	2.995	9.626
	HIGH	12.634	10.970	49.969	3.941	3.672	4.951	3.994	12.566
Miami-Dade	LOW	3.394	2.321	2.904	0.515	0.484	0.719	0.646	3.724
	AVERAGE	7.395	6.662	21.055	2.776	2.253	3.557	3.536	10.894
	HIGH	23.291	14.812	58.917	7.214	5.744	7.321	8.601	19.391
Monroe	LOW	8.884	6.417	11.712	3.576	2.297	4.389	3.446	16.204
	AVERAGE	11.610	11.671	62.843	4.399	3.590	4.944	4.721	19.869

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
	HIGH	18.053	15.712	88.510	6.778	8.810	10.747	6.887	26.462
Nassau	LOW	0.528	0.518	1.293	0.121	0.102	0.124	0.165	1.321
	AVERAGE	0.952	0.873	2.047	0.259	0.238	0.388	0.345	2.544
	HIGH	1.118	1.105	6.457	0.282	0.264	0.389	0.345	3.392
Okaloosa	LOW	1.595	1.590	5.539	0.395	0.372	0.383	0.861	2.388
	AVERAGE	3.332	3.278	9.108	1.081	1.006	1.989	1.684	7.683
	HIGH	5.189	5.436	23.985	1.982	2.627	2.633	2.073	8.347
Okeechobee	LOW	4.762	4.570	11.814	0.650	0.669	1.486	0.701	4.917
	AVERAGE	5.560	5.052	15.696	0.780	0.732	1.486	1.396	6.104
	HIGH	6.106	5.315	16.700	0.902	0.770	1.486	1.407	6.206
Orange	LOW	1.306	1.299	1.465	0.231	0.240	0.318	0.336	0.322
	AVERAGE	2.709	2.634	5.957	0.321	0.312	0.558	0.547	2.462
	HIGH	3.889	3.728	10.129	0.412	0.362	0.685	0.749	3.497
Osceola	LOW	2.079	2.105	6.288	0.307	0.307	0.490	0.413	2.760
	AVERAGE	2.657	2.745	7.456	0.339	0.343	0.611	0.534	2.975
	HIGH	4.688	3.878	11.071	0.810	0.420	0.726	0.760	3.764
Palm Beach	LOW	4.307	4.387	2.869	0.888	0.773	0.797	0.973	4.118
	AVERAGE	8.893	7.009	25.381	3.724	2.074	3.125	2.844	9.886
	HIGH	18.317	13.741	68.802	5.717	4.442	6.934	5.954	16.800
Pasco	LOW	1.724	1.652	2.338	0.313	0.281	0.436	0.375	2.568
	AVERAGE	2.808	2.773	7.607	0.382	0.370	0.729	0.706	3.107
	HIGH	3.813	3.796	10.956	0.537	0.448	0.831	0.791	3.431

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Pinellas	LOW	2.284	2.186	1.618	0.344	0.333	0.646	0.584	2.737
	AVERAGE	3.824	3.666	11.009	0.510	0.493	1.010	1.005	4.740
	HIGH	6.122	6.327	19.846	1.261	1.045	1.767	1.937	6.789
Polk	LOW	1.640	1.641	1.881	0.291	0.258	0.371	0.365	2.286
	AVERAGE	3.507	3.163	8.377	0.428	0.408	0.730	0.738	3.377
	HIGH	5.379	4.674	12.057	0.633	0.613	0.955	1.136	4.298
Putnam	LOW	0.957	0.916	1.031	0.211	0.194	0.344	0.218	1.331
	AVERAGE	1.107	1.067	3.439	0.247	0.228	0.375	0.317	2.102
	HIGH	1.341	1.282	4.652	0.304	0.293	0.402	0.370	2.550
St. Johns	LOW	0.732	0.754	2.364	0.168	0.155	0.194	0.214	1.481
	AVERAGE	1.256	1.366	4.258	0.383	0.343	0.597	0.601	3.510
	HIGH	2.278	2.002	10.379	0.671	0.565	0.915	0.834	4.481
St. Lucie	LOW	4.820	2.443	5.292	0.652	0.507	0.554	0.786	3.259
	AVERAGE	6.306	4.130	22.273	1.232	0.921	2.637	2.748	8.285
	HIGH	11.701	10.393	33.551	3.838	2.753	5.055	3.870	10.690
Santa Rosa	LOW	1.848	1.804	7.467	0.477	0.441	0.455	0.563	3.693
	AVERAGE	3.289	3.236	11.744	1.199	1.219	3.089	1.811	8.532
	HIGH	6.601	6.192	33.022	2.718	2.106	3.596	2.036	16.504
Sarasota	LOW	1.872	1.909	9.592	0.463	0.397	0.705	0.558	3.285
	AVERAGE	4.984	4.161	16.143	0.731	0.643	1.271	1.197	5.586
	HIGH	7.254	6.357	22.814	1.416	1.140	2.057	1.843	7.369

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Seminole	LOW	2.457	2.087	4.808	0.230	0.247	0.331	0.411	1.879
	AVERAGE	2.826	2.639	6.153	0.307	0.300	0.570	0.547	2.458
	HIGH	3.664	3.246	7.706	0.364	0.367	0.725	0.603	2.677
Sumter	LOW	1.379	1.321	3.976	0.254	0.268	0.518	0.376	1.845
	AVERAGE	1.584	1.721	6.353	0.287	0.301	0.660	0.457	2.164
	HIGH	3.475	3.201	7.633	0.385	0.381	0.682	0.674	3.003
Suwanee	LOW	0.629	0.615	1.533	0.140	0.130	0.191	0.000	1.387
	AVERAGE	0.712	0.696	1.776	0.151	0.144	0.191	0.000	1.497
	HIGH	0.898	0.875	2.303	0.205	0.200	0.191	0.000	1.631
Taylor	LOW	0.726	0.683	1.565	0.156	0.167	0.182	0.330	1.750
	AVERAGE	0.866	0.808	2.569	0.184	0.168	0.261	0.330	2.060
	HIGH	1.228	1.246	4.189	0.359	0.170	0.343	0.330	2.165
Union	LOW	0.847	0.824	0.802	0.191	0.136	0.198	0.238	1.790
	AVERAGE	0.857	0.835	2.321	0.204	0.184	0.198	0.238	1.790
	HIGH	0.987	0.967	2.934	0.233	0.217	0.198	0.238	1.790
Volusia	LOW	1.109	1.431	1.261	0.265	0.267	0.410	0.311	2.040
	AVERAGE	3.723	3.200	7.534	0.502	0.562	1.227	1.615	4.936
	HIGH	7.539	7.342	24.971	1.444	1.415	2.338	2.165	6.151
Wakulla	LOW	0.802	0.808	2.593	0.176	0.180	0.283	0.293	1.443
	AVERAGE	1.012	0.974	2.938	0.230	0.260	0.504	0.642	2.617
	HIGH	1.964	2.024	8.737	0.647	0.665	0.740	0.689	4.067
Walton	LOW	1.622	1.551	4.552	0.383	0.361	0.559	0.367	2.521

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
	AVERAGE	3.072	2.690	7.575	1.046	1.091	1.993	1.475	6.989
	HIGH	4.321	3.895	31.234	1.585	1.690	2.393	1.686	7.662
Washington	LOW	1.410	1.391	4.025	0.324	0.313	0.487	0.000	2.675
	AVERAGE	1.426	1.408	4.258	0.343	0.319	0.507	0.000	2.675
	HIGH	1.525	1.500	5.133	0.358	0.371	0.544	0.000	2.675
Statewide	LOW	0.519	0.511	0.626	0.115	0.102	0.124	0.138	0.055
	AVERAGE	3.141	4.292	10.002	0.893	1.025	1.427	1.994	7.633
	HIGH	23.291	15.712	88.510	7.214	8.810	10.747	8.601	26.462

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Alachua	LOW	0.130	0.197	0.210	0.039	0.031	0.032	0.033	0.211
	AVERAGE	0.228	0.234	1.952	0.048	0.044	0.052	0.047	0.348
	HIGH	0.336	0.357	4.130	0.086	0.061	0.074	0.087	0.633
Baker	LOW	0.123	0.120	0.755	0.028	0.023	0.022	0.029	0.345
	AVERAGE	0.142	0.138	1.054	0.033	0.031	0.022	0.029	0.345
	HIGH	0.166	0.142	1.202	0.035	0.036	0.022	0.029	0.345
Bay	LOW	0.368	0.369	2.724	0.102	0.078	0.126	0.095	0.580
	AVERAGE	1.232	0.991	7.107	0.353	0.312	1.129	0.575	2.681
	HIGH	2.465	2.400	18.042	1.254	0.847	1.669	0.720	3.562
Bradford	LOW	0.164	0.156	1.128	0.037	0.031	0.071	0.040	0.332
	AVERAGE	0.208	0.198	1.492	0.046	0.039	0.071	0.040	0.389
	HIGH	0.279	0.273	2.332	0.074	0.061	0.071	0.040	0.419
Brevard	LOW	1.560	0.479	4.947	0.048	0.072	0.064	0.125	0.653
	AVERAGE	2.965	2.625	13.563	0.388	0.445	0.753	1.104	3.106
	HIGH	10.072	7.896	30.126	2.620	2.698	3.851	3.189	7.057
Broward	LOW	0.926	0.562	0.791	0.118	0.146	0.100	0.120	0.661
	AVERAGE	4.254	3.431	22.311	1.597	0.876	1.375	1.290	4.788
	HIGH	11.814	10.545	46.721	3.878	3.246	5.980	4.842	10.497
Calhoun	LOW	0.283	0.254	1.461	0.061	0.046	0.000	0.000	0.215
	AVERAGE	0.292	0.272	1.929	0.063	0.047	0.000	0.000	0.215
	HIGH	0.303	0.305	2.070	0.067	0.049	0.000	0.000	0.215

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Charlotte	LOW	1.878	1.034	6.869	0.134	0.112	0.086	0.220	1.340
	AVERAGE	3.160	2.514	10.201	0.383	0.237	0.643	0.300	1.810
	HIGH	4.406	4.135	28.225	0.839	0.565	1.100	0.685	3.380
Citrus	LOW	0.902	0.984	2.939	0.067	0.053	0.048	0.071	0.097
	AVERAGE	1.521	1.208	3.965	0.099	0.078	0.127	0.120	0.473
	HIGH	2.070	1.611	5.640	0.140	0.141	0.147	0.183	0.555
Clay	LOW	0.148	0.151	0.941	0.036	0.034	0.031	0.025	0.247
	AVERAGE	0.177	0.181	1.490	0.043	0.041	0.041	0.036	0.364
	HIGH	0.298	0.251	3.014	0.056	0.055	0.060	0.058	0.459
Collier	LOW	2.369	0.615	4.418	0.157	0.117	0.165	0.157	1.046
	AVERAGE	3.693	2.887	17.144	0.490	0.346	0.586	0.555	2.166
	HIGH	6.916	6.206	41.148	2.119	0.924	1.800	1.599	4.384
Columbia	LOW	0.143	0.129	0.921	0.036	0.031	0.038	0.037	0.240
	AVERAGE	0.184	0.172	1.371	0.040	0.036	0.047	0.039	0.335
	HIGH	0.200	0.188	1.584	0.045	0.044	0.048	0.041	0.359
De Soto	LOW	2.155	2.155	3.924	0.119	0.067	0.256	0.072	0.840
	AVERAGE	2.575	2.405	6.941	0.202	0.203	0.267	0.216	0.987
	HIGH	2.957	2.472	9.026	0.278	0.267	0.274	0.224	0.998
Dixie	LOW	0.287	0.246	2.182	0.067	0.042	0.045	0.047	0.225
	AVERAGE	0.363	0.271	2.406	0.081	0.051	0.168	0.147	0.503
	HIGH	0.901	0.746	9.355	0.215	0.052	0.239	0.187	0.582

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
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Duval	LOW	0.085	0.102	0.125	0.023	0.025	0.020	0.018	0.003
	AVERAGE	0.212	0.191	1.469	0.047	0.040	0.048	0.051	0.581
	HIGH	0.915	0.599	4.590	0.281	0.151	0.196	0.120	1.356
Escambia	LOW	0.602	0.602	0.841	0.119	0.133	0.129	0.064	0.482
	AVERAGE	1.367	1.288	9.263	0.431	0.400	0.828	0.796	2.321
	HIGH	3.058	2.362	20.199	1.284	0.875	1.654	2.270	3.395
Flagler	LOW	0.821	0.578	2.670	0.067	0.066	0.141	0.038	0.343
	AVERAGE	1.573	0.957	5.732	0.228	0.130	0.423	0.195	1.052
	HIGH	3.866	3.021	15.568	0.849	0.664	1.177	0.786	2.316
Franklin	LOW	1.010	1.040	6.127	0.393	0.351	0.150	0.390	0.547
	AVERAGE	1.323	1.261	7.952	0.545	0.490	0.279	0.621	1.740
	HIGH	1.449	1.506	8.912	0.624	0.803	0.737	0.675	2.202
Gadsen	LOW	0.087	0.107	0.992	0.034	0.029	0.000	0.000	0.246
	AVERAGE	0.172	0.166	1.267	0.040	0.035	0.000	0.000	0.283
	HIGH	0.219	0.207	1.992	0.047	0.040	0.000	0.000	0.353
Gilchrist	LOW	0.239	0.213	1.644	0.052	0.047	0.000	0.051	0.334
	AVERAGE	0.271	0.257	1.996	0.064	0.055	0.000	0.051	0.334
	HIGH	0.284	0.279	2.168	0.067	0.061	0.000	0.051	0.334
Glades	LOW	3.581	3.194	12.197	0.474	0.332	0.000	0.000	0.000
	AVERAGE	3.582	3.202	12.293	0.474	0.332	0.000	0.000	0.000
	HIGH	3.612	3.541	15.484	0.474	0.332	0.000	0.000	0.000

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

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Gulf	LOW	0.386	0.406	3.716	0.099	0.065	0.215	0.101	1.210
	AVERAGE	0.701	0.755	5.076	0.239	0.217	0.215	0.265	1.210
	HIGH	0.765	0.890	7.758	0.283	0.278	0.215	0.305	1.210
Hamilton	LOW	0.106	0.096	0.738	0.024	0.021	0.000	0.018	0.178
	AVERAGE	0.128	0.119	0.785	0.028	0.027	0.000	0.018	0.213
	HIGH	0.148	0.132	0.827	0.030	0.028	0.000	0.018	0.226
Hardee	LOW	2.348	2.210	6.555	0.164	0.139	0.000	0.000	1.175
	AVERAGE	2.411	2.262	6.717	0.186	0.157	0.000	0.000	1.199
	HIGH	2.779	2.469	8.099	0.369	0.308	0.000	0.000	1.651
Hendry	LOW	2.781	1.672	8.072	0.210	0.165	0.278	0.142	2.199
	AVERAGE	3.310	3.061	11.995	0.326	0.407	0.589	0.630	2.357
	HIGH	4.376	3.985	14.050	0.629	0.536	0.775	0.706	2.372
Hernando	LOW	0.965	0.590	0.392	0.087	0.056	0.115	0.131	0.423
	AVERAGE	1.575	1.333	5.201	0.097	0.086	0.130	0.154	0.485
	HIGH	2.427	1.909	6.992	0.134	0.163	0.158	0.182	1.637
Highlands	LOW	1.867	1.678	5.372	0.107	0.130	0.120	0.145	1.023
	AVERAGE	2.303	2.138	7.824	0.154	0.149	0.196	0.199	1.081
	HIGH	3.148	2.749	12.609	0.216	0.304	0.297	0.311	1.816
Hillsborough	LOW	0.351	0.344	0.404	0.078	0.053	0.067	0.071	0.111
	AVERAGE	1.751	1.631	7.408	0.126	0.120	0.152	0.158	0.758
	HIGH	3.177	3.379	14.489	0.312	0.333	0.392	0.778	1.224

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Model Release Date: January 22, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Holmes	LOW	0.408	0.285	2.870	0.088	0.082	0.060	0.000	0.426
	AVERAGE	0.450	0.423	2.920	0.098	0.082	0.060	0.000	0.426
	HIGH	0.473	0.425	4.839	0.131	0.082	0.060	0.000	0.426
Indian River	LOW	2.176	1.174	8.461	0.223	0.151	0.330	0.339	1.817
	AVERAGE	5.137	3.605	13.982	1.577	1.104	1.813	2.048	5.971
	HIGH	9.974	7.347	35.307	2.995	2.413	4.336	3.314	8.989
Jackson	LOW	0.215	0.203	1.421	0.040	0.032	0.045	0.041	0.224
	AVERAGE	0.290	0.269	2.044	0.060	0.053	0.045	0.047	0.473
	HIGH	0.436	0.426	3.251	0.094	0.098	0.045	0.054	0.544
Jefferson	LOW	0.125	0.115	0.970	0.026	0.027	0.000	0.000	0.263
	AVERAGE	0.138	0.131	1.048	0.033	0.028	0.000	0.000	0.263
	HIGH	0.184	0.158	1.923	0.042	0.028	0.000	0.000	0.263
Lafayette	LOW	0.193	0.171	1.342	0.046	0.039	0.000	0.000	0.000
	AVERAGE	0.194	0.186	1.347	0.046	0.039	0.000	0.000	0.000
	HIGH	0.216	0.186	1.853	0.046	0.039	0.000	0.000	0.000
Lake	LOW	0.719	0.268	0.341	0.039	0.050	0.039	0.045	0.290
	AVERAGE	1.229	1.017	4.573	0.077	0.071	0.118	0.107	0.418
	HIGH	1.945	1.579	7.204	0.166	0.142	0.141	0.167	0.697
Lee	LOW	0.490	1.110	6.123	0.135	0.075	0.092	0.099	0.904
	AVERAGE	4.136	2.272	14.717	0.562	0.242	0.739	0.430	2.171
	HIGH	7.238	6.390	29.879	1.674	1.269	2.569	2.257	8.066

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: January 22, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Leon	LOW	0.154	0.135	0.161	0.029	0.021	0.025	0.022	0.165
	AVERAGE	0.180	0.168	1.611	0.038	0.034	0.035	0.034	0.220
	HIGH	0.270	0.238	3.236	0.052	0.048	0.073	0.053	0.348
Levy	LOW	0.261	0.247	2.245	0.043	0.045	0.041	0.215	0.299
	AVERAGE	0.454	0.330	2.703	0.143	0.067	0.325	0.215	1.061
	HIGH	1.102	1.087	6.641	0.437	0.114	0.329	0.215	1.167
Liberty	LOW	0.257	0.230	1.858	0.054	0.046	0.000	0.000	0.000
	AVERAGE	0.258	0.242	1.911	0.059	0.049	0.000	0.000	0.000
	HIGH	0.269	0.259	2.409	0.060	0.061	0.000	0.000	0.000
Madison	LOW	0.107	0.102	0.735	0.026	0.023	0.000	0.000	0.239
	AVERAGE	0.143	0.134	0.928	0.033	0.028	0.000	0.000	0.239
	HIGH	0.151	0.138	1.137	0.036	0.031	0.000	0.000	0.239
Manatee	LOW	1.811	1.070	0.521	0.132	0.067	0.080	0.113	0.649
	AVERAGE	2.861	1.917	11.290	0.415	0.331	0.995	0.854	3.032
	HIGH	6.713	5.727	33.635	1.669	1.401	2.311	1.806	5.416
Marion	LOW	0.955	0.643	2.611	0.046	0.038	0.060	0.048	0.324
	AVERAGE	1.273	0.972	3.330	0.068	0.062	0.093	0.090	0.463
	HIGH	1.623	1.472	4.425	0.105	0.106	0.163	0.141	0.916
Martin	LOW	2.837	2.184	15.050	0.270	0.279	0.514	0.793	2.316
	AVERAGE	6.357	4.485	28.764	2.027	1.317	2.661	1.860	5.065
	HIGH	9.590	8.001	46.543	3.178	2.920	3.659	2.730	7.714

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Miami-Dade	LOW	1.363	0.584	0.955	0.121	0.109	0.131	0.102	0.557
	AVERAGE	4.827	4.156	18.508	2.145	1.643	2.475	2.387	6.145
	HIGH	19.369	11.409	55.139	6.322	4.811	5.913	6.841	13.473
Monroe	LOW	6.221	3.816	9.034	2.868	1.688	3.165	2.289	11.244
	AVERAGE	8.711	8.616	59.444	3.645	2.832	3.701	3.405	14.531
	HIGH	14.760	12.360	84.765	5.885	7.793	9.022	5.395	20.509
Nassau	LOW	0.108	0.101	0.768	0.026	0.023	0.016	0.022	0.304
	AVERAGE	0.266	0.230	1.356	0.093	0.083	0.125	0.101	0.804
	HIGH	0.362	0.349	5.276	0.105	0.097	0.125	0.101	1.340
Okaloosa	LOW	0.460	0.447	4.192	0.120	0.114	0.057	0.324	0.351
	AVERAGE	1.709	1.648	7.443	0.632	0.568	1.238	0.967	3.630
	HIGH	3.204	3.451	21.446	1.413	1.979	1.785	1.307	4.096
Okeechobee	LOW	2.631	2.472	9.600	0.248	0.261	0.507	0.129	1.263
	AVERAGE	3.215	2.789	13.173	0.334	0.296	0.507	0.440	1.907
	HIGH	3.614	2.962	14.097	0.415	0.317	0.507	0.445	1.963
Orange	LOW	0.267	0.251	0.360	0.043	0.047	0.043	0.045	0.027
	AVERAGE	1.305	1.250	4.439	0.078	0.076	0.092	0.090	0.418
	HIGH	2.269	2.070	8.183	0.133	0.111	0.127	0.175	0.864
Osceola	LOW	0.774	0.662	4.620	0.068	0.069	0.071	0.057	0.483
	AVERAGE	1.222	1.289	5.710	0.085	0.089	0.112	0.092	0.588
	HIGH	2.689	2.023	8.965	0.388	0.126	0.152	0.162	0.913

Modeling Organization: Florida International University
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Model Release Date: January 22, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Palm Beach	LOW	1.933	2.087	0.861	0.397	0.318	0.159	0.220	1.036
	AVERAGE	5.988	4.324	22.440	2.962	1.435	1.930	1.645	4.978
	HIGH	14.425	10.135	64.571	4.783	3.541	5.433	4.334	10.833
Pasco	LOW	0.466	0.413	1.034	0.067	0.053	0.062	0.052	0.431
	AVERAGE	1.342	1.314	5.930	0.110	0.105	0.159	0.154	0.653
	HIGH	2.094	2.132	9.045	0.207	0.154	0.203	0.189	0.770
Pinellas	LOW	0.925	0.850	0.402	0.087	0.077	0.127	0.080	0.446
	AVERAGE	2.108	1.975	9.141	0.194	0.181	0.341	0.320	1.610
	HIGH	3.767	4.048	17.449	0.782	0.574	0.866	0.960	2.922
Polk	LOW	0.385	0.346	0.498	0.065	0.052	0.050	0.051	0.349
	AVERAGE	1.808	1.540	6.532	0.121	0.112	0.143	0.143	0.739
	HIGH	3.339	2.711	9.898	0.250	0.246	0.214	0.334	1.215
Putnam	LOW	0.238	0.207	0.236	0.045	0.041	0.065	0.030	0.429
	AVERAGE	0.307	0.282	2.411	0.065	0.059	0.085	0.063	0.491
	HIGH	0.425	0.382	3.470	0.088	0.089	0.098	0.089	0.586
St. Johns	LOW	0.118	0.134	1.470	0.034	0.031	0.027	0.032	0.233
	AVERAGE	0.414	0.460	3.190	0.167	0.137	0.249	0.250	1.307
	HIGH	1.081	0.841	8.794	0.373	0.289	0.475	0.411	1.907
St. Lucie	LOW	2.658	0.731	3.472	0.250	0.154	0.082	0.192	0.482
	AVERAGE	3.908	2.117	19.653	0.751	0.492	1.610	1.703	4.026
	HIGH	8.712	7.460	30.536	3.149	2.096	3.781	2.675	5.922

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: January 22, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Santa Rosa	LOW	0.596	0.567	5.859	0.156	0.108	0.069	0.100	0.967
	AVERAGE	1.665	1.594	9.823	0.743	0.756	2.214	1.081	4.362
	HIGH	4.470	3.978	30.183	2.076	1.506	2.668	1.270	11.454
Sarasota	LOW	0.440	0.475	7.607	0.122	0.090	0.154	0.090	0.519
	AVERAGE	2.915	2.233	13.821	0.337	0.266	0.473	0.416	2.020
	HIGH	4.741	3.973	20.074	0.911	0.660	1.083	0.880	3.276
Seminole	LOW	1.153	0.859	3.431	0.043	0.050	0.045	0.058	0.316
	AVERAGE	1.445	1.298	4.629	0.076	0.074	0.099	0.094	0.426
	HIGH	2.146	1.793	6.119	0.110	0.110	0.156	0.108	0.494
Sumter	LOW	0.365	0.324	2.708	0.055	0.062	0.069	0.051	0.263
	AVERAGE	0.512	0.607	4.788	0.069	0.075	0.133	0.072	0.351
	HIGH	1.832	1.660	5.949	0.110	0.108	0.138	0.129	0.663
Suwanee	LOW	0.134	0.125	0.916	0.029	0.028	0.029	0.000	0.211
	AVERAGE	0.159	0.149	1.087	0.034	0.032	0.029	0.000	0.261
	HIGH	0.217	0.203	1.508	0.051	0.052	0.029	0.000	0.284
Taylor	LOW	0.179	0.146	0.924	0.038	0.040	0.025	0.065	0.380
	AVERAGE	0.221	0.198	1.743	0.047	0.042	0.044	0.065	0.387
	HIGH	0.334	0.348	3.036	0.130	0.044	0.063	0.065	0.389
Union	LOW	0.190	0.168	0.165	0.042	0.024	0.027	0.034	0.314
	AVERAGE	0.194	0.179	1.471	0.048	0.042	0.027	0.034	0.314
	HIGH	0.247	0.235	1.983	0.060	0.058	0.027	0.034	0.314

Form A-4A Output Ranges (2007 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Volusia	LOW	0.223	0.325	0.299	0.067	0.069	0.069	0.042	0.325
	AVERAGE	2.221	1.802	6.137	0.231	0.290	0.569	0.871	1.991
	HIGH	5.285	5.091	22.492	0.988	0.994	1.400	1.264	2.776
Wakulla	LOW	0.155	0.160	1.728	0.038	0.041	0.051	0.069	0.228
	AVERAGE	0.261	0.245	2.027	0.069	0.092	0.168	0.236	0.701
	HIGH	0.754	0.806	7.118	0.326	0.355	0.292	0.259	1.284
Walton	LOW	0.529	0.495	3.251	0.098	0.100	0.142	0.053	0.436
	AVERAGE	1.447	1.132	5.997	0.608	0.657	1.252	0.813	3.095
	HIGH	2.510	2.110	28.486	1.057	1.152	1.596	0.977	3.573
Washington	LOW	0.407	0.397	2.801	0.071	0.085	0.088	0.000	0.525
	AVERAGE	0.413	0.402	3.003	0.095	0.088	0.109	0.000	0.525
	HIGH	0.480	0.439	3.772	0.101	0.126	0.147	0.000	0.525
Statewide	LOW	0.085	0.096	0.125	0.023	0.021	0.016	0.018	0.003
	AVERAGE	1.711	2.407	8.234	0.560	0.616	0.737	1.062	3.631
	HIGH	19.369	12.360	84.765	6.322	7.793	9.022	6.841	20.509

Form A-4B Output Ranges (2012 FHCF Exposure Data)

LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: March 15, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Alachua	LOW	0.844	0.792	0.956	0.178	0.152	0.241	0.217	0.773
	AVERAGE	0.915	0.924	2.738	0.194	0.182	0.274	0.255	1.891
	HIGH	1.161	1.076	5.736	0.252	0.261	0.288	0.318	2.736
Baker	LOW	0.596	0.597	1.222	0.125	0.120	0.000	0.000	0.000
	AVERAGE	0.655	0.650	1.490	0.133	0.125	0.000	0.000	0.000
	HIGH	0.674	0.659	1.708	0.139	0.134	0.000	0.000	0.000
Bay	LOW	1.280	1.296	2.628	0.371	0.269	0.486	0.470	2.726
	AVERAGE	2.481	2.319	6.758	0.583	0.542	1.566	0.964	6.492
	HIGH	3.699	3.783	19.664	1.100	0.977	2.034	1.332	7.679
Bradford	LOW	0.725	0.714	1.601	0.158	0.137	0.000	0.000	0.000
	AVERAGE	0.839	0.832	2.069	0.185	0.164	0.000	0.000	0.000
	HIGH	1.085	1.089	2.844	0.227	0.256	0.000	0.000	0.000
Brevard	LOW	2.597	1.668	1.853	0.314	0.297	0.524	0.455	3.157
	AVERAGE	4.046	3.734	14.169	0.639	0.701	1.135	1.599	6.469
	HIGH	10.565	9.221	33.012	3.038	2.733	4.382	3.325	11.649
Broward	LOW	2.501	2.580	2.778	0.485	0.524	0.692	0.689	4.155
	AVERAGE	5.796	5.056	20.581	1.178	1.124	1.598	2.052	8.993
	HIGH	13.073	14.515	50.336	4.494	2.975	7.916	4.208	15.242
Calhoun	LOW	1.059	1.027	1.966	0.247	0.224	0.000	0.000	0.000
	AVERAGE	1.098	1.066	2.706	0.252	0.254	0.000	0.000	0.000
	HIGH	1.163	1.125	2.940	0.262	0.302	0.000	0.000	0.000
Charlotte	LOW	3.605	3.302	7.149	0.448	0.438	0.592	0.794	3.988
	AVERAGE	4.406	3.830	9.580	0.605	0.531	1.338	0.889	4.355
	HIGH	6.265	5.477	31.102	0.968	0.659	1.816	1.356	5.829
Citrus	LOW	2.155	1.192	4.447	0.253	0.251	0.314	0.315	2.280
	AVERAGE	2.408	2.108	5.215	0.302	0.276	0.563	0.537	2.620
	HIGH	3.665	2.808	7.560	0.370	0.332	0.611	0.718	3.115

Form A-4B Output Ranges (2012 FHCF Exposure Data)

LOSS COSTS PER $1000 for 0% Deductible

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Clay	LOW	0.705	0.691	1.919	0.141	0.137	0.188	0.176	1.324
	AVERAGE	0.784	0.777	2.193	0.166	0.156	0.218	0.198	1.451
	HIGH	0.958	1.010	4.002	0.200	0.191	0.240	0.262	1.816
Collier	LOW	2.692	2.397	2.468	0.574	0.451	0.584	0.576	4.271
	AVERAGE	4.901	4.316	16.761	0.798	0.727	1.259	1.223	5.567
	HIGH	10.811	7.505	42.921	1.285	1.096	2.383	1.848	8.590
Columbia	LOW	0.641	0.602	1.122	0.149	0.131	0.206	0.209	1.242
	AVERAGE	0.764	0.750	1.853	0.162	0.152	0.224	0.215	1.242
	HIGH	0.838	0.818	2.130	0.183	0.167	0.232	0.226	1.242
De Soto	LOW	2.224	3.488	2.108	0.484	0.429	0.862	0.820	3.963
	AVERAGE	3.681	3.574	8.089	0.495	0.472	0.873	0.880	4.422
	HIGH	4.769	3.814	13.302	0.572	0.670	0.898	0.882	4.433
Dixie	LOW	0.997	0.973	2.852	0.229	0.184	0.256	0.261	1.578
	AVERAGE	1.123	0.994	2.969	0.237	0.220	0.382	0.346	2.431
	HIGH	1.990	1.703	10.438	0.241	0.221	0.446	0.420	2.672
Duval	LOW	0.563	0.572	0.725	0.132	0.110	0.169	0.142	1.140
	AVERAGE	0.832	0.796	2.067	0.177	0.164	0.228	0.244	1.862
	HIGH	1.739	1.602	6.017	0.458	0.354	0.516	0.428	4.123
Escambia	LOW	1.592	1.586	2.986	0.385	0.368	0.456	0.431	3.600
	AVERAGE	2.671	2.650	9.143	0.741	0.714	1.292	1.152	6.566
	HIGH	4.753	3.685	17.757	2.717	1.506	2.716	1.820	7.013
Flagler	LOW	1.552	1.430	3.515	0.217	0.207	0.478	0.261	1.789
	AVERAGE	2.315	1.797	7.080	0.386	0.285	0.702	0.479	2.779
	HIGH	3.925	4.373	10.131	0.925	0.778	1.411	1.147	5.491
Franklin	LOW	2.067	1.815	6.541	0.669	0.395	0.526	0.480	4.744
	AVERAGE	2.485	2.420	8.977	0.739	0.466	0.833	0.687	4.744

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	HIGH	2.685	2.679	11.377	0.768	0.669	1.456	0.885	4.744
Gadsen	LOW	0.585	0.610	1.306	0.140	0.130	0.000	0.157	1.334
	AVERAGE	0.714	0.715	1.851	0.160	0.151	0.000	0.157	1.384
	HIGH	0.861	0.859	2.954	0.202	0.163	0.000	0.157	1.645
Gilchrist	LOW	0.892	0.879	2.161	0.182	0.157	0.000	0.286	0.000
	AVERAGE	1.000	0.982	2.710	0.240	0.219	0.000	0.286	0.000
	HIGH	1.046	1.028	3.015	0.247	0.231	0.000	0.286	0.000
Glades	LOW	5.091	3.499	12.894	0.818	0.684	0.000	0.000	0.000
	AVERAGE	5.112	4.100	12.916	0.818	0.684	0.000	0.000	0.000
	HIGH	5.787	4.108	13.666	0.818	0.684	0.000	0.000	0.000
Gulf	LOW	1.340	1.388	1.937	0.311	0.287	0.564	0.459	3.563
	AVERAGE	1.879	1.933	5.374	0.493	0.465	0.564	0.465	3.563
	HIGH	1.991	2.109	9.030	0.522	0.492	0.564	0.922	3.563
Hamilton	LOW	0.513	0.506	1.174	0.112	0.079	0.000	0.000	0.000
	AVERAGE	0.558	0.558	1.258	0.125	0.113	0.000	0.000	0.000
	HIGH	0.600	0.596	1.353	0.129	0.116	0.000	0.000	0.000
Hardee	LOW	3.311	3.224	7.399	0.423	0.405	0.000	1.071	2.698
	AVERAGE	3.455	3.300	7.777	0.471	0.427	0.000	1.071	2.717
	HIGH	4.178	3.675	8.091	0.703	0.494	0.000	1.071	2.889
Hendry	LOW	4.002	3.509	8.216	0.423	0.498	1.040	0.960	5.125
	AVERAGE	4.511	4.173	12.433	0.765	0.646	1.254	1.115	5.125
	HIGH	5.459	4.921	14.305	0.897	0.764	1.331	1.144	5.125
Hernando	LOW	2.214	1.584	4.832	0.282	0.251	0.572	0.368	2.224
	AVERAGE	2.514	2.247	6.315	0.303	0.288	0.622	0.600	2.655
	HIGH	4.275	3.159	8.518	0.344	0.358	0.763	0.833	2.842
Highlands	LOW	2.955	2.720	6.663	0.379	0.369	0.691	0.676	3.684

Form A-4B Output Ranges (2012 FHCF Exposure Data)

LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: March 15, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
	AVERAGE	3.479	3.307	8.896	0.458	0.422	0.780	0.767	3.967
	HIGH	4.490	4.568	15.554	0.770	0.600	0.970	0.816	4.727
Hillsborough	LOW	1.592	1.438	1.672	0.303	0.265	0.410	0.391	2.762
	AVERAGE	2.816	2.800	8.283	0.365	0.362	0.606	0.631	3.392
	HIGH	3.690	4.901	14.202	0.643	0.635	0.983	1.101	5.913
Holmes	LOW	1.052	1.263	3.738	0.301	0.284	0.518	0.000	2.202
	AVERAGE	1.324	1.307	3.795	0.308	0.284	0.518	0.000	2.202
	HIGH	1.351	1.342	4.504	0.330	0.284	0.518	0.000	2.202
Indian River	LOW	2.186	2.362	8.407	0.416	0.397	0.990	0.645	4.183
	AVERAGE	5.655	4.620	14.589	1.657	1.393	2.066	2.413	8.412
	HIGH	10.105	7.743	50.833	2.914	2.394	4.191	3.444	11.121
Jackson	LOW	0.867	0.849	1.900	0.202	0.155	0.000	0.265	1.625
	AVERAGE	1.020	1.016	2.666	0.228	0.221	0.000	0.312	1.952
	HIGH	1.282	1.292	3.860	0.296	0.288	0.000	0.457	2.084
Jefferson	LOW	0.541	0.523	1.223	0.113	0.110	0.184	0.000	0.000
	AVERAGE	0.603	0.599	1.501	0.131	0.124	0.184	0.000	0.000
	HIGH	0.728	0.690	1.878	0.131	0.144	0.184	0.000	0.000
Lafayette	LOW	0.715	0.724	0.690	0.182	0.141	0.290	0.000	0.000
	AVERAGE	0.753	0.745	1.968	0.182	0.141	0.290	0.000	0.000
	HIGH	0.754	0.745	1.978	0.182	0.141	0.290	0.000	0.000
Lake	LOW	1.623	1.504	3.878	0.198	0.196	0.365	0.337	1.991
	AVERAGE	2.093	1.970	5.877	0.278	0.263	0.497	0.478	2.498
	HIGH	2.771	2.609	8.767	0.385	0.410	0.608	0.517	2.965
Lee	LOW	2.055	2.036	2.306	0.387	0.376	0.603	0.540	3.174
	AVERAGE	4.958	3.424	16.232	0.691	0.529	1.258	0.957	5.682
	HIGH	7.741	6.906	29.662	2.142	1.648	2.772	2.396	11.109

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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Leon	LOW	0.691	0.636	0.768	0.132	0.123	0.175	0.170	1.089
	AVERAGE	0.731	0.721	2.185	0.154	0.143	0.199	0.195	1.395
	HIGH	0.974	0.937	3.173	0.189	0.169	0.229	0.251	1.767
Levy	LOW	0.989	0.966	3.027	0.228	0.164	0.764	0.746	2.248
	AVERAGE	1.309	1.182	3.518	0.357	0.261	0.764	0.746	3.472
	HIGH	2.174	2.125	7.550	0.833	1.094	0.764	0.746	3.561
Liberty	LOW	0.989	0.876	2.218	0.224	0.219	0.000	0.000	0.000
	AVERAGE	0.991	0.984	2.494	0.231	0.219	0.000	0.000	0.000
	HIGH	1.042	0.986	3.350	0.232	0.219	0.000	0.000	0.000
Madison	LOW	0.491	0.485	1.250	0.105	0.090	0.000	0.000	0.000
	AVERAGE	0.587	0.576	1.433	0.128	0.117	0.000	0.000	0.000
	HIGH	0.602	0.587	1.554	0.134	0.131	0.000	0.000	0.000
Manatee	LOW	2.523	1.866	1.900	0.391	0.312	0.483	0.498	3.187
	AVERAGE	3.755	3.045	11.080	0.562	0.570	1.242	1.364	5.943
	HIGH	8.171	6.360	32.328	1.674	1.631	2.507	2.368	8.962
Marion	LOW	1.165	1.060	1.160	0.180	0.169	0.275	0.276	1.837
	AVERAGE	1.526	1.444	4.412	0.221	0.212	0.342	0.361	2.059
	HIGH	2.594	1.965	6.209	0.263	0.247	0.495	0.433	3.494
Martin	LOW	4.669	3.928	16.715	0.743	0.491	0.800	1.575	6.310
	AVERAGE	7.897	6.254	30.083	2.400	1.566	3.435	2.708	9.586
	HIGH	10.584	12.882	41.822	3.936	4.427	5.065	4.895	11.751
Miami-Dade	LOW	2.625	2.387	3.029	0.447	0.484	0.665	0.633	2.370
	AVERAGE	6.319	5.713	15.845	2.200	1.733	2.914	2.735	11.177
	HIGH	13.338	12.270	40.289	5.900	5.862	5.056	6.648	21.103
Monroe	LOW	8.750	7.550	48.897	1.931	2.362	4.136	1.924	16.996
	AVERAGE	10.131	10.270	58.142	3.737	2.828	4.277	4.137	20.165
	HIGH	15.980	13.196	83.006	7.980	7.882	9.938	6.004	31.146

Form A-4B Output Ranges (2012 FHCF Exposure Data)

LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Nassau	LOW	0.505	0.500	1.159	0.113	0.102	0.321	0.130	1.140
	AVERAGE	0.880	0.823	1.871	0.230	0.204	0.321	0.317	2.020
	HIGH	1.021	1.001	3.731	0.249	0.220	0.321	0.317	2.025
Okaloosa	LOW	1.428	1.427	2.351	0.358	0.340	0.397	0.704	2.878
	AVERAGE	2.892	2.827	6.457	0.825	0.808	1.476	1.408	7.224
	HIGH	4.949	4.083	22.265	2.210	2.080	2.157	1.997	7.988
Okeechobee	LOW	3.910	3.082	10.638	0.549	0.540	0.852	0.766	3.573
	AVERAGE	4.429	4.028	14.346	0.648	0.610	0.852	1.119	4.181
	HIGH	4.812	4.294	16.271	0.761	0.658	0.852	1.168	4.189
Orange	LOW	1.291	1.268	1.346	0.238	0.207	0.318	0.337	2.015
	AVERAGE	2.175	2.187	5.448	0.292	0.280	0.471	0.459	2.523
	HIGH	3.434	2.945	10.092	0.376	0.404	0.590	0.696	3.137
Osceola	LOW	1.786	1.370	5.540	0.285	0.247	0.394	0.380	2.427
	AVERAGE	1.929	1.925	6.952	0.299	0.286	0.470	0.434	2.507
	HIGH	2.447	2.686	10.309	0.486	0.334	0.496	0.518	3.115
Palm Beach	LOW	2.890	2.685	2.781	0.648	0.628	0.711	0.659	4.661
	AVERAGE	7.906	6.251	21.400	4.028	2.035	2.693	2.610	10.001
	HIGH	14.703	13.733	48.759	5.372	3.826	6.021	7.147	18.076
Pasco	LOW	1.612	1.635	2.666	0.265	0.247	0.402	0.416	2.247
	AVERAGE	2.333	2.349	6.797	0.321	0.322	0.580	0.592	3.063
	HIGH	3.769	3.179	10.937	0.427	0.378	0.783	0.645	3.538
Pinellas	LOW	1.480	1.435	1.616	0.328	0.277	0.422	0.388	2.691
	AVERAGE	3.682	3.543	10.429	0.451	0.474	0.957	0.964	4.752
	HIGH	6.052	8.090	19.186	1.436	0.983	1.753	1.562	6.909
Polk	LOW	1.576	1.680	1.769	0.274	0.252	0.359	0.344	1.994
	AVERAGE	2.815	2.658	7.842	0.366	0.364	0.559	0.602	2.963

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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
	HIGH	4.055	4.494	12.032	0.536	0.553	1.076	0.940	4.240
Putnam	LOW	0.902	0.869	2.256	0.187	0.186	0.242	0.230	1.793
	AVERAGE	1.025	1.002	3.499	0.234	0.212	0.317	0.263	1.954
	HIGH	1.213	1.174	5.213	0.313	0.255	0.489	0.345	1.993
St. Johns	LOW	0.718	0.723	1.707	0.152	0.147	0.190	0.178	1.169
	AVERAGE	1.141	1.258	3.469	0.310	0.302	0.496	0.526	3.325
	HIGH	2.016	1.827	10.631	0.594	0.477	0.767	0.716	4.268
St. Lucie	LOW	4.169	2.278	10.184	0.503	0.424	0.566	0.574	3.490
	AVERAGE	5.638	3.626	21.028	0.921	0.718	2.062	2.340	7.575
	HIGH	10.480	9.462	34.346	2.937	2.022	3.802	3.109	9.368
Santa Rosa	LOW	1.728	1.700	1.800	0.412	0.398	0.528	1.024	3.416
	AVERAGE	2.837	2.828	8.905	0.969	0.872	2.492	1.450	7.902
	HIGH	5.751	5.199	33.220	2.432	1.537	4.088	1.600	9.341
Sarasota	LOW	1.885	1.853	2.000	0.427	0.370	0.514	0.481	2.880
	AVERAGE	3.955	3.461	14.784	0.591	0.534	1.015	0.966	4.542
	HIGH	6.453	7.120	22.560	1.177	0.989	1.616	1.469	5.826
Seminole	LOW	2.037	1.724	4.397	0.265	0.218	0.344	0.337	2.211
	AVERAGE	2.378	2.313	5.748	0.287	0.276	0.495	0.483	2.447
	HIGH	3.593	3.419	7.525	0.364	0.353	0.602	0.691	2.723
Sumter	LOW	1.350	1.276	4.567	0.228	0.222	0.367	0.354	2.179
	AVERAGE	1.454	1.476	6.021	0.251	0.252	0.486	0.387	2.576
	HIGH	2.912	2.637	7.798	0.376	0.440	0.643	0.516	2.977
Suwanee	LOW	0.608	0.599	1.465	0.137	0.124	0.194	0.195	0.900
	AVERAGE	0.683	0.673	1.671	0.150	0.134	0.194	0.195	1.223
	HIGH	0.852	0.816	2.129	0.188	0.156	0.194	0.195	1.902
Taylor	LOW	0.659	0.568	1.740	0.150	0.145	0.181	0.293	1.638

Form A-4B Output Ranges (2012 FHCF Exposure Data)

LOSS COSTS PER $1000 for 0% Deductible

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
	AVERAGE	0.804	0.759	2.170	0.181	0.154	0.300	0.293	1.638
	HIGH	1.121	1.151	4.061	0.250	0.233	0.307	0.293	1.638
Union	LOW	0.806	0.794	0.805	0.176	0.162	0.200	0.189	1.180
	AVERAGE	0.811	0.801	1.949	0.183	0.164	0.200	0.189	1.180
	HIGH	0.920	0.893	3.539	0.206	0.190	0.200	0.189	1.180
Volusia	LOW	1.179	1.077	1.259	0.183	0.190	0.322	0.345	1.359
	AVERAGE	2.546	2.287	6.064	0.353	0.554	0.810	1.062	4.736
	HIGH	7.548	4.713	23.005	1.031	1.066	2.332	1.394	5.887
Wakulla	LOW	0.774	0.785	2.414	0.168	0.147	0.259	0.591	1.553
	AVERAGE	0.930	0.915	2.639	0.184	0.183	0.382	0.591	3.023
	HIGH	1.790	1.804	8.230	0.379	0.525	0.468	0.591	3.318
Walton	LOW	1.384	1.453	3.823	0.339	0.326	0.492	0.319	2.571
	AVERAGE	2.708	2.458	6.133	0.803	0.813	1.699	1.192	6.716
	HIGH	3.896	3.374	31.221	1.355	1.108	2.039	1.343	7.840
Washington	LOW	1.236	1.242	3.102	0.291	0.230	0.340	0.000	2.045
	AVERAGE	1.257	1.263	3.253	0.300	0.273	0.340	0.000	2.045
	HIGH	1.431	1.398	4.315	0.349	0.308	0.340	0.000	2.045
Statewide	LOW	0.491	0.485	0.690	0.105	0.079	0.169	0.130	0.773
	AVERAGE	2.551	3.659	7.720	0.809	0.896	0.977	1.662	7.641
	HIGH	15.980	14.515	83.006	7.980	7.882	9.938	7.147	31.146

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: March 15, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Alachua	LOW	0.120	0.097	0.210	0.036	0.027	0.033	0.029	0.108
	AVERAGE	0.186	0.193	1.778	0.041	0.039	0.045	0.040	0.371
	HIGH	0.316	0.307	4.392	0.068	0.074	0.050	0.048	0.695
Baker	LOW	0.102	0.103	0.633	0.025	0.022	0.000	0.000	0.000
	AVERAGE	0.115	0.112	0.837	0.027	0.024	0.000	0.000	0.000
	HIGH	0.136	0.119	1.010	0.028	0.031	0.000	0.000	0.000
Bay	LOW	0.254	0.303	1.496	0.097	0.055	0.087	0.071	0.411
	AVERAGE	1.004	0.886	5.153	0.236	0.210	0.917	0.411	2.923
	HIGH	1.964	2.032	17.317	0.664	0.561	1.337	0.706	3.905
Bradford	LOW	0.148	0.141	0.902	0.034	0.027	0.000	0.000	0.000
	AVERAGE	0.182	0.177	1.260	0.041	0.034	0.000	0.000	0.000
	HIGH	0.255	0.259	1.825	0.047	0.069	0.000	0.000	0.000
Brevard	LOW	1.241	0.364	0.513	0.077	0.072	0.095	0.066	0.734
	AVERAGE	2.296	2.023	12.130	0.307	0.360	0.482	0.835	2.868
	HIGH	7.796	6.518	30.003	2.406	2.091	3.216	2.254	6.835
Broward	LOW	0.657	0.700	0.830	0.118	0.147	0.109	0.110	0.915
	AVERAGE	3.369	2.725	17.892	0.697	0.632	0.722	1.089	4.512
	HIGH	9.769	11.272	46.633	3.701	2.297	6.403	2.927	9.738
Calhoun	LOW	0.208	0.211	0.977	0.054	0.050	0.000	0.000	0.000
	AVERAGE	0.243	0.222	1.656	0.056	0.065	0.000	0.000	0.000
	HIGH	0.264	0.239	1.865	0.057	0.090	0.000	0.000	0.000

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: March 15, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Charlotte	LOW	1.761	1.477	5.196	0.115	0.110	0.087	0.160	0.875
	AVERAGE	2.363	1.909	7.518	0.218	0.168	0.504	0.218	1.090
	HIGH	3.880	3.113	28.048	0.479	0.237	0.843	0.423	1.869
Citrus	LOW	0.972	0.232	3.137	0.056	0.053	0.042	0.042	0.356
	AVERAGE	1.134	0.914	3.829	0.077	0.066	0.106	0.097	0.485
	HIGH	2.149	1.453	5.983	0.116	0.090	0.122	0.164	0.634
Clay	LOW	0.115	0.081	1.136	0.028	0.028	0.026	0.023	0.192
	AVERAGE	0.142	0.142	1.360	0.035	0.033	0.033	0.029	0.249
	HIGH	0.277	0.251	2.898	0.044	0.043	0.042	0.053	0.333
Collier	LOW	0.846	0.611	0.687	0.176	0.109	0.085	0.083	0.994
	AVERAGE	2.618	2.136	14.245	0.331	0.283	0.410	0.390	1.676
	HIGH	7.635	4.745	39.518	0.714	0.557	1.157	0.818	3.815
Columbia	LOW	0.121	0.096	0.532	0.029	0.025	0.030	0.033	0.182
	AVERAGE	0.149	0.142	1.089	0.035	0.032	0.035	0.034	0.182
	HIGH	0.186	0.160	1.325	0.038	0.037	0.038	0.036	0.182
De Soto	LOW	0.662	1.671	0.579	0.139	0.106	0.180	0.131	0.820
	AVERAGE	1.818	1.718	6.130	0.144	0.133	0.182	0.193	1.000
	HIGH	2.672	1.974	10.981	0.183	0.267	0.183	0.194	1.004
Dixie	LOW	0.229	0.196	1.861	0.052	0.033	0.036	0.036	0.225
	AVERAGE	0.285	0.216	1.971	0.060	0.058	0.066	0.057	0.407
	HIGH	0.738	0.494	8.696	0.063	0.058	0.082	0.073	0.458

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
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County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Duval	LOW	0.081	0.066	0.151	0.025	0.020	0.023	0.018	0.181
	AVERAGE	0.202	0.187	1.325	0.043	0.038	0.040	0.045	0.437
	HIGH	0.725	0.605	4.830	0.218	0.129	0.187	0.117	1.891
Escambia	LOW	0.504	0.500	1.237	0.116	0.110	0.070	0.064	0.961
	AVERAGE	1.135	1.109	7.388	0.345	0.330	0.652	0.540	2.903
	HIGH	2.808	1.897	15.514	2.060	1.004	1.877	1.112	3.217
Flagler	LOW	0.589	0.496	2.478	0.048	0.044	0.097	0.036	0.301
	AVERAGE	1.144	0.746	5.709	0.169	0.094	0.261	0.138	0.809
	HIGH	2.341	2.723	8.471	0.594	0.459	0.800	0.576	2.603
Franklin	LOW	0.731	0.463	4.864	0.325	0.111	0.123	0.080	1.423
	AVERAGE	0.999	0.967	7.186	0.363	0.149	0.352	0.222	1.423
	HIGH	1.204	1.133	9.532	0.393	0.316	0.817	0.330	1.423
Gadsen	LOW	0.077	0.095	0.685	0.028	0.026	0.000	0.020	0.194
	AVERAGE	0.139	0.138	1.119	0.034	0.031	0.000	0.020	0.205
	HIGH	0.183	0.183	2.003	0.044	0.035	0.000	0.020	0.258
Gilchrist	LOW	0.185	0.178	1.304	0.036	0.030	0.000	0.043	0.000
	AVERAGE	0.218	0.207	1.745	0.057	0.052	0.000	0.043	0.000
	HIGH	0.232	0.220	1.990	0.059	0.057	0.000	0.043	0.000
Glades	LOW	2.789	1.532	10.498	0.349	0.231	0.000	0.000	0.000
	AVERAGE	2.808	1.964	10.519	0.349	0.231	0.000	0.000	0.000
	HIGH	3.414	1.969	11.241	0.349	0.231	0.000	0.000	0.000

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: March 15, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Gulf	LOW	0.304	0.342	0.527	0.074	0.073	0.132	0.072	0.955
	AVERAGE	0.556	0.618	3.929	0.170	0.166	0.132	0.076	0.955
	HIGH	0.609	0.707	7.184	0.185	0.180	0.132	0.398	0.955
Hamilton	LOW	0.087	0.083	0.645	0.022	0.013	0.000	0.000	0.000
	AVERAGE	0.103	0.102	0.701	0.026	0.022	0.000	0.000	0.000
	HIGH	0.117	0.114	0.763	0.027	0.022	0.000	0.000	0.000
Hardee	LOW	1.570	1.501	5.558	0.107	0.104	0.000	0.281	0.384
	AVERAGE	1.683	1.567	5.882	0.136	0.114	0.000	0.281	0.387
	HIGH	2.186	1.770	6.147	0.283	0.142	0.000	0.281	0.412
Hendry	LOW	2.011	1.622	6.172	0.084	0.131	0.230	0.220	1.329
	AVERAGE	2.396	2.108	10.132	0.329	0.234	0.407	0.351	1.329
	HIGH	3.112	2.655	11.882	0.428	0.316	0.470	0.375	1.329
Hernando	LOW	0.948	0.462	3.448	0.065	0.049	0.110	0.050	0.331
	AVERAGE	1.189	0.988	4.807	0.075	0.069	0.125	0.120	0.511
	HIGH	2.627	1.772	6.791	0.093	0.102	0.174	0.226	0.587
Highlands	LOW	1.258	1.064	4.884	0.074	0.086	0.132	0.103	0.752
	AVERAGE	1.683	1.548	6.924	0.125	0.110	0.159	0.148	0.839
	HIGH	2.338	2.491	13.049	0.332	0.191	0.231	0.165	0.992
Hillsborough	LOW	0.372	0.315	0.457	0.063	0.051	0.057	0.054	0.495
	AVERAGE	1.339	1.317	6.529	0.097	0.096	0.114	0.124	0.793
	HIGH	1.977	3.031	12.070	0.261	0.272	0.279	0.361	2.272

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: March 15, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Holmes	LOW	0.216	0.338	2.555	0.078	0.068	0.146	0.000	0.350
	AVERAGE	0.354	0.343	2.606	0.081	0.068	0.146	0.000	0.350
	HIGH	0.367	0.386	3.315	0.091	0.068	0.146	0.000	0.350
Indian River	LOW	0.607	0.902	6.602	0.126	0.111	0.316	0.134	1.154
	AVERAGE	3.525	2.625	12.414	1.166	0.928	1.229	1.504	4.306
	HIGH	7.382	5.203	47.441	2.272	1.785	3.070	2.367	6.451
Jackson	LOW	0.189	0.178	1.093	0.041	0.029	0.000	0.038	0.239
	AVERAGE	0.232	0.228	1.684	0.052	0.051	0.000	0.064	0.318
	HIGH	0.337	0.346	2.730	0.077	0.076	0.000	0.130	0.352
Jefferson	LOW	0.097	0.085	0.684	0.020	0.021	0.028	0.000	0.000
	AVERAGE	0.116	0.113	0.896	0.027	0.027	0.028	0.000	0.000
	HIGH	0.169	0.149	1.184	0.028	0.028	0.028	0.000	0.000
Lafayette	LOW	0.159	0.153	0.146	0.045	0.027	0.070	0.000	0.000
	AVERAGE	0.159	0.153	1.227	0.045	0.027	0.070	0.000	0.000
	HIGH	0.167	0.167	1.235	0.045	0.027	0.070	0.000	0.000
Lake	LOW	0.651	0.514	2.748	0.041	0.041	0.050	0.045	0.283
	AVERAGE	0.869	0.757	4.394	0.061	0.057	0.078	0.074	0.426
	HIGH	1.373	1.292	6.964	0.105	0.136	0.106	0.084	0.558
Lee	LOW	0.474	0.459	0.618	0.081	0.080	0.089	0.077	0.469
	AVERAGE	2.729	1.509	13.740	0.280	0.156	0.468	0.259	1.763
	HIGH	4.908	4.301	26.647	1.458	1.021	1.616	1.287	5.731

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: March 15, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Leon	LOW	0.126	0.076	0.163	0.024	0.023	0.023	0.023	0.153
	AVERAGE	0.153	0.145	1.412	0.033	0.030	0.030	0.029	0.250
	HIGH	0.266	0.256	2.278	0.041	0.037	0.039	0.041	0.318
Levy	LOW	0.213	0.129	1.925	0.046	0.029	0.295	0.292	0.384
	AVERAGE	0.345	0.268	2.366	0.129	0.067	0.295	0.292	0.877
	HIGH	0.821	0.785	5.891	0.463	0.697	0.295	0.292	0.913
Liberty	LOW	0.202	0.120	1.283	0.050	0.049	0.000	0.000	0.000
	AVERAGE	0.203	0.200	1.524	0.053	0.049	0.000	0.000	0.000
	HIGH	0.236	0.201	2.284	0.053	0.049	0.000	0.000	0.000
Madison	LOW	0.087	0.084	0.730	0.020	0.015	0.000	0.000	0.000
	AVERAGE	0.117	0.110	0.849	0.027	0.024	0.000	0.000	0.000
	HIGH	0.123	0.115	0.949	0.029	0.029	0.000	0.000	0.000
Manatee	LOW	1.031	0.527	0.514	0.092	0.061	0.071	0.074	0.583
	AVERAGE	1.979	1.409	9.143	0.227	0.236	0.554	0.622	2.357
	HIGH	5.513	3.980	29.349	1.145	1.101	1.508	1.384	4.520
Marion	LOW	0.278	0.195	0.266	0.034	0.031	0.036	0.036	0.275
	AVERAGE	0.519	0.453	3.147	0.045	0.042	0.049	0.051	0.341
	HIGH	1.339	0.843	4.756	0.069	0.056	0.094	0.059	0.965
Martin	LOW	2.541	1.927	14.260	0.324	0.122	0.186	0.673	2.192
	AVERAGE	5.293	3.842	27.201	1.803	1.033	2.322	1.649	4.974
	HIGH	7.686	9.805	38.621	3.173	3.608	3.677	3.523	6.802

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: March 15, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Miami-Dade	LOW	0.743	0.609	1.073	0.095	0.108	0.108	0.099	0.703
	AVERAGE	3.900	3.361	13.536	1.631	1.179	1.950	1.738	6.540
	HIGH	10.221	9.123	36.997	5.055	4.824	3.909	5.141	14.879
Monroe	LOW	6.177	4.722	45.810	1.358	1.702	3.006	0.937	11.825
	AVERAGE	7.379	7.364	54.854	3.031	2.134	3.101	2.905	14.839
	HIGH	12.812	10.070	79.337	6.973	6.830	8.386	4.639	24.927
Nassau	LOW	0.091	0.088	0.651	0.023	0.022	0.083	0.017	0.183
	AVERAGE	0.213	0.191	1.202	0.074	0.061	0.083	0.084	0.465
	HIGH	0.291	0.260	2.768	0.082	0.068	0.083	0.084	0.489
Okaloosa	LOW	0.277	0.365	0.780	0.089	0.090	0.059	0.210	0.584
	AVERAGE	1.329	1.277	4.965	0.424	0.409	0.809	0.744	3.361
	HIGH	3.008	2.253	19.755	1.582	1.501	1.370	1.247	3.870
Okeechobee	LOW	1.938	1.215	8.486	0.172	0.170	0.184	0.133	0.562
	AVERAGE	2.299	1.965	11.903	0.237	0.210	0.184	0.322	0.783
	HIGH	2.565	2.139	13.784	0.312	0.238	0.184	0.349	0.786
Orange	LOW	0.262	0.242	0.320	0.048	0.038	0.043	0.045	0.316
	AVERAGE	0.896	0.910	3.979	0.065	0.061	0.070	0.069	0.480
	HIGH	1.961	1.535	8.147	0.114	0.131	0.103	0.145	0.733
Osceola	LOW	0.536	0.269	3.941	0.056	0.046	0.054	0.052	0.399
	AVERAGE	0.644	0.643	5.246	0.063	0.059	0.067	0.062	0.445
	HIGH	0.847	1.041	8.251	0.147	0.072	0.071	0.076	0.530

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: March 15, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Palm Beach	LOW	0.834	0.678	0.813	0.214	0.189	0.118	0.104	1.050
	AVERAGE	5.136	3.690	18.620	3.257	1.408	1.584	1.492	5.055
	HIGH	11.114	10.243	45.064	4.486	2.986	4.661	5.395	11.873
Pasco	LOW	0.376	0.401	1.340	0.050	0.046	0.056	0.058	0.343
	AVERAGE	0.972	0.988	5.183	0.075	0.078	0.108	0.117	0.638
	HIGH	2.056	1.620	9.029	0.149	0.113	0.198	0.138	0.764
Pinellas	LOW	0.333	0.303	0.401	0.079	0.053	0.061	0.055	0.424
	AVERAGE	1.997	1.881	8.604	0.155	0.169	0.315	0.303	1.709
	HIGH	3.788	5.674	16.822	0.929	0.529	0.861	0.683	3.106
Polk	LOW	0.366	0.359	0.465	0.057	0.050	0.049	0.046	0.295
	AVERAGE	1.271	1.150	6.044	0.089	0.090	0.094	0.106	0.547
	HIGH	2.217	2.551	9.877	0.196	0.206	0.301	0.232	1.012
Putnam	LOW	0.196	0.174	1.390	0.035	0.037	0.033	0.031	0.262
	AVERAGE	0.248	0.234	2.468	0.060	0.052	0.061	0.042	0.347
	HIGH	0.334	0.306	3.986	0.100	0.069	0.151	0.056	0.355
St. Johns	LOW	0.106	0.096	0.869	0.031	0.028	0.026	0.024	0.183
	AVERAGE	0.328	0.373	2.473	0.119	0.112	0.179	0.199	1.190
	HIGH	0.858	0.696	9.039	0.311	0.223	0.358	0.321	1.739
St. Lucie	LOW	2.130	0.602	8.053	0.141	0.098	0.084	0.090	0.527
	AVERAGE	3.354	1.708	18.453	0.485	0.329	1.141	1.397	3.547
	HIGH	7.628	6.643	31.308	2.317	1.434	2.691	2.052	4.899

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: March 15, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Santa Rosa	LOW	0.489	0.471	0.514	0.114	0.106	0.111	0.459	1.020
	AVERAGE	1.263	1.245	7.136	0.546	0.464	1.687	0.777	3.875
	HIGH	3.683	3.162	30.377	1.815	1.007	3.096	0.899	5.016
Sarasota	LOW	0.452	0.433	0.528	0.094	0.080	0.074	0.072	0.446
	AVERAGE	2.087	1.677	12.545	0.233	0.192	0.329	0.300	1.328
	HIGH	4.189	4.617	19.852	0.709	0.543	0.777	0.647	2.124
Seminole	LOW	0.821	0.584	3.062	0.057	0.040	0.047	0.047	0.393
	AVERAGE	1.102	1.052	4.267	0.067	0.064	0.079	0.077	0.488
	HIGH	2.092	1.942	5.866	0.108	0.107	0.110	0.147	0.584
Sumter	LOW	0.341	0.288	3.174	0.044	0.043	0.051	0.048	0.329
	AVERAGE	0.411	0.419	4.478	0.053	0.054	0.079	0.056	0.464
	HIGH	1.398	1.183	6.094	0.107	0.138	0.099	0.089	0.553
Suwanee	LOW	0.114	0.108	0.851	0.030	0.025	0.029	0.026	0.128
	AVERAGE	0.134	0.126	0.993	0.033	0.028	0.029	0.026	0.202
	HIGH	0.185	0.161	1.381	0.043	0.030	0.029	0.026	0.401
Taylor	LOW	0.131	0.067	1.050	0.031	0.032	0.025	0.048	0.233
	AVERAGE	0.178	0.165	1.395	0.043	0.035	0.049	0.048	0.233
	HIGH	0.290	0.318	2.918	0.062	0.057	0.050	0.048	0.233
Union	LOW	0.155	0.144	0.166	0.038	0.035	0.027	0.025	0.169
	AVERAGE	0.157	0.150	1.140	0.040	0.035	0.027	0.025	0.169
	HIGH	0.219	0.184	2.511	0.048	0.038	0.027	0.025	0.169

Form A-4B Output Ranges (2012 FHCF Exposure Data)
LOSS COSTS PER $1000 with Specified Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: March 15, 2015

County	Loss Costs	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Volusia	LOW	0.240	0.204	0.298	0.034	0.035	0.045	0.052	0.245
	AVERAGE	1.261	1.061	4.670	0.125	0.288	0.302	0.479	1.884
	HIGH	5.289	2.862	20.601	0.629	0.692	1.399	0.709	2.553
Wakulla	LOW	0.133	0.142	1.568	0.035	0.028	0.041	0.191	0.255
	AVERAGE	0.205	0.201	1.760	0.041	0.045	0.081	0.191	0.684
	HIGH	0.614	0.627	6.643	0.114	0.248	0.109	0.191	0.772
Walton	LOW	0.334	0.376	2.578	0.089	0.083	0.080	0.045	0.471
	AVERAGE	1.124	0.925	4.636	0.408	0.418	1.004	0.580	2.990
	HIGH	2.040	1.566	28.474	0.856	0.647	1.291	0.699	3.863
Washington	LOW	0.298	0.305	1.982	0.067	0.042	0.049	0.000	0.320
	AVERAGE	0.306	0.312	2.105	0.076	0.069	0.049	0.000	0.320
	HIGH	0.393	0.397	3.049	0.100	0.084	0.049	0.000	0.320
Statewide	LOW	0.077	0.066	0.146	0.020	0.013	0.023	0.017	0.108
	AVERAGE	1.240	1.896	6.131	0.500	0.513	0.427	0.840	3.693
	HIGH	12.812	11.272	79.337	6.973	6.830	8.386	5.395	24.927

Appendix F – Form A-5: Percentage Change in Output Ranges

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
~~November 1, 2014~~ January 22, 2015

Form A-5 Percentage Change in Output Ranges

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Percentage Change in $0 Deductible Output Ranges

Region	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Coastal	-3.43%	-5.18%	-4.39%	-11.92%	-10.19%	-7.48%	-8.70%	-10.23%
Inland	-6.52%	-6.87%	-8.73%	-7.87%	-8.72%	-8.42%	-7.57%	345.36%
North	-2.85%	-3.73%	-1.42%	-4.04%	-5.18%	-6.97%	-6.17%	19.37%
Central	-4.22%	-3.09%	-6.35%	-6.96%	-4.78%	-4.34%	-2.49%	38.32%
South	-4.53%	-6.77%	-5.19%	-15.03%	-11.55%	-9.50%	-9.97%	-15.77%
Statewide	-4.02%	-5.40%	-5.66%	-11.55%	-10.09%	-7.54%	-8.69%	-8.36%

Percentage Change in Specified Deductible Output Ranges

Region	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masrony Condo Unit	Commercial Residential
Coastal	-5.01%	-7.02%	-5.17%	-15.45%	-13.49%	-11.74%	-12.94%	-46.37%
Inland	-9.45%	-9.56%	-10.59%	-13.32%	-14.47%	-14.75%	-13.53%	165.82%
North	-4.50%	-5.70%	-1.97%	-5.43%	-6.81%	-9.28%	-8.84%	-33.89%
Central	-5.94%	-5.07%	-7.64%	-11.42%	-8.11%	-7.78%	-5.81%	-29.10%
South	-5.94%	-8.43%	-5.78%	-17.68%	-14.45%	-14.10%	-14.18%	-48.41%
Statewide	-5.70%	-7.30%	-6.65%	-15.38%	-13.52%	-11.81%	-12.94%	-45.92%

Form A-5 Percentage Change in Output Ranges

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: January 22, 2015

Percentage Change in $0 Deductible Output Ranges

Region	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Coastal	-3.43%	-5.18%	-4.39%	-11.92%	-10.19%	-7.48%	-8.70%	-10.44%
Inland	-6.52%	-6.87%	-8.73%	-7.87%	-8.72%	-8.42%	-7.57%	339.59%
North	-2.85%	-3.73%	-1.42%	-4.04%	-5.18%	-6.97%	-6.17%	17.09%
Central	-4.22%	-3.09%	-6.35%	-6.96%	-4.78%	-4.34%	-2.49%	37.06%
South	-4.53%	-6.77%	-5.19%	-15.03%	-11.55%	-9.50%	-9.97%	-15.79%
Statewide	-4.02%	-5.40%	-5.66%	-11.55%	-10.09%	-7.54%	-8.69%	-8.60%

Percentage Change in Specified Deductible Output Ranges

Region	Frame Owners	Masonry Owners	Mobile Homes	Frame Renters	Masonry Renters	Frame Condo Unit	Masonry Condo Unit	Commercial Residential
Coastal	-5.01%	-7.02%	-5.17%	-15.45%	-13.49%	-11.74%	-12.94%	-46.19%
Inland	-9.45%	-9.56%	-10.59%	-13.32%	-14.47%	-14.75%	-13.53%	154.58%
North	-4.50%	-5.70%	-1.97%	-5.43%	-6.81%	-9.28%	-8.84%	-36.44%
Central	-5.94%	-5.07%	-7.64%	-11.42%	-8.11%	-7.78%	-5.81%	-30.15%
South	-5.94%	-8.43%	-5.78%	-17.68%	-14.45%	-14.10%	-14.18%	-48.00%
Statewide	-5.70%	-7.30%	-6.65%	-15.38%	-13.52%	-11.81%	-12.94%	-45.77%

Appendix G – Form A-6: Logical Relationship to Risk

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
~~November 1, 2014~~February 3, 2015

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
Frame Owners	1	BAY	4.190	3.537	2.885	2.329	1.615	1.296	1.000	0.844	0.689	0.556	0.386	0.309
	2	BREVARD	4.590	4.015	3.442	2.731	1.302	0.648	1.000	0.875	0.750	0.595	0.284	0.141
	3	BREVARD	4.429	3.858	3.290	2.589	1.187	0.561	1.000	0.871	0.743	0.585	0.268	0.127
	4	BROWARD	8.204	7.427	6.652	5.591	3.295	1.919	1.000	0.905	0.811	0.681	0.402	0.234
	5	BROWARD	14.045	13.096	12.150	10.813	7.858	5.781	1.000	0.932	0.865	0.770	0.560	0.412
	6	CITRUS	3.324	2.865	2.407	1.863	0.784	0.311	1.000	0.862	0.724	0.560	0.236	0.093
	7	CLAY	0.895	0.629	0.365	0.214	0.088	0.064	1.000	0.703	0.407	0.239	0.098	0.072
	8	COLLIER	6.141	5.387	4.634	3.713	1.854	0.966	1.000	0.877	0.755	0.605	0.302	0.157
	9	COLUMBIA	0.885	0.627	0.371	0.219	0.080	0.059	1.000	0.709	0.419	0.247	0.091	0.067
	10	DIXIE	2.882	2.357	1.834	1.427	0.954	0.767	1.000	0.818	0.637	0.495	0.331	0.266
	11	DUVAL	1.878	1.491	1.105	0.823	0.467	0.378	1.000	0.794	0.589	0.438	0.249	0.201
	12	FRANKLIN	6.099	5.342	4.588	3.931	3.054	2.546	1.000	0.876	0.752	0.645	0.501	0.417
	13	GLADES	5.349	4.649	3.952	3.124	1.463	0.670	1.000	0.869	0.739	0.584	0.273	0.125
	14	HAMILTON	0.786	0.557	0.330	0.193	0.071	0.053	1.000	0.709	0.420	0.246	0.090	0.067
	15	HERNANDO	4.396	3.829	3.265	2.584	1.221	0.602	1.000	0.871	0.743	0.588	0.278	0.137
	16	HILLSBOROUGH	3.943	3.412	2.883	2.227	0.927	0.377	1.000	0.865	0.731	0.565	0.235	0.096
	17	HOLMES	1.527	1.138	0.750	0.482	0.198	0.143	1.000	0.745	0.491	0.315	0.129	0.094
	18	INDIAN RIVER	11.844	10.985	10.129	9.050	6.795	5.307	1.000	0.928	0.855	0.764	0.574	0.448
	19	JACKSON	1.167	0.843	0.520	0.313	0.112	0.082	1.000	0.722	0.446	0.268	0.096	0.070
	20	LEE	5.439	4.724	4.011	3.208	1.599	0.772	1.000	0.869	0.737	0.590	0.294	0.142
	21	LEON	1.114	0.814	0.515	0.322	0.130	0.095	1.000	0.731	0.463	0.289	0.117	0.086
	22	MARION	1.621	1.225	0.831	0.582	0.227	0.100	1.000	0.756	0.513	0.359	0.140	0.062
	23	MARTIN	6.002	5.314	4.628	3.693	1.749	0.837	1.000	0.885	0.771	0.615	0.291	0.140
	24	MARTIN	13.648	12.772	11.898	10.606	7.745	5.853	1.000	0.936	0.872	0.777	0.567	0.429
	25	MIAMI-DADE	7.451	6.740	6.031	5.061	2.956	1.690	1.000	0.905	0.809	0.679	0.397	0.227
	26	MIAMI-DADE	10.846	10.008	9.171	8.007	5.456	3.739	1.000	0.923	0.846	0.738	0.503	0.345
	27	MONROE	13.480	12.664	11.849	10.575	7.745	5.731	1.000	0.939	0.879	0.784	0.575	0.425
	28	MONROE	21.049	20.082	19.116	17.607	14.219	11.501	1.000	0.954	0.908	0.836	0.676	0.546
	29	OKALOOSA	3.048	2.475	1.904	1.434	0.869	0.657	1.000	0.812	0.624	0.470	0.285	0.216
	30	OSCEOLA	2.095	1.602	1.112	0.787	0.311	0.137	1.000	0.765	0.531	0.376	0.149	0.065
	31	OSCEOLA	2.660	2.075	1.493	1.085	0.470	0.230	1.000	0.780	0.561	0.408	0.177	0.086
	32	PALM BEACH	7.732	6.945	6.160	5.049	2.678	1.453	1.000	0.898	0.797	0.653	0.346	0.188
	33	PALM BEACH	11.134	10.230	9.328	8.009	5.125	3.417	1.000	0.919	0.838	0.719	0.460	0.307
	34	PINELLAS	4.486	3.963	3.442	2.701	1.180	0.546	1.000	0.883	0.767	0.602	0.263	0.122
	35	SAINT JOHNS	1.424	1.087	0.752	0.514	0.272	0.207	1.000	0.763	0.528	0.361	0.191	0.145
	36	SANTA ROSA	2.216	1.748	1.281	0.910	0.483	0.352	1.000	0.789	0.578	0.411	0.218	0.159
	37	SEMINOLE	3.279	2.833	2.387	1.839	0.745	0.282	1.000	0.864	0.728	0.561	0.227	0.086
	38	TAYLOR	1.002	0.731	0.461	0.283	0.120	0.089	1.000	0.729	0.460	0.282	0.119	0.089
	39	VOLUSIA	2.997	2.520	2.043	1.592	0.760	0.386	1.000	0.841	0.682	0.531	0.254	0.129
	40	WAKULLA	2.541	2.035	1.531	1.145	0.712	0.557	1.000	0.801	0.602	0.450	0.280	0.219

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
	1	BAY	3.863	3.212	2.562	2.039	1.394	1.094	1.000	0.831	0.663	0.528	0.361	0.283
	2	BREVARD	4.521	3.947	3.374	2.670	1.258	0.614	1.000	0.873	0.746	0.590	0.278	0.136
	3	BREVARD	4.369	3.798	3.230	2.536	1.151	0.535	1.000	0.869	0.739	0.580	0.264	0.122
	4	BROWARD	7.980	7.203	6.429	5.380	3.128	1.802	1.000	0.903	0.806	0.674	0.392	0.226
	5	BROWARD	13.289	12.342	11.397	10.074	7.170	5.159	1.000	0.929	0.858	0.758	0.540	0.388
	6	CITRUS	3.273	2.814	2.356	1.820	0.762	0.301	1.000	0.860	0.720	0.556	0.233	0.092
	7	CLAY	0.862	0.596	0.331	0.191	0.082	0.060	1.000	0.691	0.385	0.222	0.096	0.070
	8	COLLIER	6.018	5.263	4.511	3.603	1.778	0.910	1.000	0.875	0.750	0.599	0.295	0.151
	9	COLUMBIA	0.850	0.593	0.337	0.195	0.075	0.055	1.000	0.697	0.396	0.230	0.088	0.065
	10	DIXIE	2.684	2.160	1.638	1.255	0.832	0.656	1.000	0.805	0.610	0.468	0.310	0.245
	11	DUVAL	1.690	1.304	0.919	0.659	0.343	0.262	1.000	0.771	0.543	0.390	0.203	0.155
	12	FRANKLIN	5.539	4.785	4.032	3.411	2.609	2.127	1.000	0.864	0.728	0.616	0.471	0.384
	13	GLADES	5.240	4.541	3.843	3.031	1.407	0.640	1.000	0.867	0.734	0.578	0.269	0.122
	14	HAMILTON	0.755	0.526	0.299	0.172	0.066	0.049	1.000	0.697	0.396	0.228	0.088	0.065
	15	HERNANDO	4.327	3.761	3.197	2.524	1.183	0.576	1.000	0.869	0.739	0.583	0.273	0.133
	16	HILLSBOROUGH	3.878	3.347	2.818	2.172	0.896	0.361	1.000	0.863	0.727	0.560	0.231	0.093
	17	HOLMES	1.456	1.068	0.681	0.431	0.182	0.133	1.000	0.733	0.467	0.296	0.125	0.091
	18	INDIAN RIVER	11.084	10.226	9.370	8.306	6.088	4.625	1.000	0.923	0.845	0.749	0.549	0.417
	19	JACKSON	1.117	0.793	0.471	0.278	0.104	0.076	1.000	0.710	0.421	0.249	0.093	0.068
Masonry Owners	20	LEE	5.311	4.597	3.884	3.096	1.530	0.739	1.000	0.866	0.731	0.583	0.288	0.139
	21	LEON	1.066	0.766	0.467	0.288	0.121	0.088	1.000	0.719	0.439	0.270	0.113	0.083
	22	MARION	1.585	1.189	0.795	0.557	0.220	0.098	1.000	0.750	0.502	0.351	0.139	0.062
	23	MARTIN	5.758	5.071	4.387	3.475	1.614	0.784	1.000	0.881	0.762	0.603	0.280	0.136
	24	MARTIN	12.669	11.794	10.922	9.653	6.886	5.107	1.000	0.931	0.862	0.762	0.544	0.403
	25	MIAMI-DADE	7.257	6.547	5.838	4.878	2.814	1.594	1.000	0.902	0.804	0.672	0.388	0.220
	26	MIAMI-DADE	10.416	9.579	8.743	7.591	5.085	3.424	1.000	0.920	0.839	0.729	0.488	0.329
	27	MONROE	12.698	11.883	11.069	9.809	7.039	5.110	1.000	0.936	0.872	0.772	0.554	0.402
	28	MONROE	19.295	18.329	17.365	15.871	12.549	9.930	1.000	0.950	0.900	0.823	0.650	0.515
	29	OKALOOSA	2.870	2.298	1.728	1.286	0.779	0.582	1.000	0.801	0.602	0.448	0.272	0.203
	30	OSCEOLA	2.046	1.554	1.063	0.752	0.302	0.134	1.000	0.759	0.520	0.368	0.148	0.065
	31	OSCEOLA	2.594	2.010	1.428	1.038	0.455	0.223	1.000	0.775	0.551	0.400	0.175	0.086
	32	PALM BEACH	7.405	6.619	5.836	4.750	2.473	1.348	1.000	0.894	0.788	0.641	0.334	0.182
	33	PALM BEACH	10.532	9.629	8.728	7.436	4.654	3.065	1.000	0.914	0.829	0.706	0.442	0.291
	34	PINELLAS	4.383	3.861	3.340	2.608	1.118	0.511	1.000	0.881	0.762	0.595	0.255	0.117
	35	SAINT JOHNS	1.354	1.018	0.683	0.460	0.246	0.186	1.000	0.752	0.504	0.340	0.182	0.137
	36	SANTA ROSA	2.101	1.634	1.168	0.820	0.439	0.318	1.000	0.778	0.556	0.390	0.209	0.152
	37	SEMINOLE	3.221	2.775	2.330	1.789	0.719	0.273	1.000	0.861	0.723	0.555	0.223	0.085
	38	TAYLOR	0.960	0.689	0.419	0.253	0.110	0.082	1.000	0.718	0.437	0.264	0.115	0.085
	39	VOLUSIA	2.950	2.472	1.996	1.553	0.737	0.369	1.000	0.838	0.677	0.526	0.250	0.125
	40	WAKULLA	2.386	1.881	1.377	1.015	0.629	0.485	1.000	0.788	0.577	0.426	0.264	0.203

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Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
	1	BAY	16.793	15.392	15.392	14.488	12.396	10.398	1.000	0.917	0.917	0.863	0.738	0.619
	2	BREVARD	13.278	12.093	12.093	11.191	9.023	7.106	1.000	0.911	0.911	0.843	0.680	0.535
	3	BREVARD	12.433	11.256	11.256	10.364	8.225	6.363	1.000	0.905	0.905	0.834	0.662	0.512
	4	BROWARD	25.412	23.803	23.803	22.494	19.229	15.979	1.000	0.937	0.937	0.885	0.757	0.629
	5	BROWARD	45.049	43.085	43.085	41.455	37.334	32.950	1.000	0.956	0.956	0.920	0.829	0.731
	6	CITRUS	7.795	6.842	6.842	6.135	4.469	3.130	1.000	0.878	0.878	0.787	0.573	0.402
	7	CLAY	2.717	2.146	2.146	1.832	1.164	0.705	1.000	0.790	0.790	0.674	0.428	0.259
	8	COLLIER	23.503	21.931	21.931	20.651	17.454	14.286	1.000	0.933	0.933	0.879	0.743	0.608
	9	COLUMBIA	2.556	2.004	2.004	1.702	1.058	0.622	1.000	0.784	0.784	0.666	0.414	0.243
	10	DIXIE	11.053	9.924	9.924	9.224	7.634	6.209	1.000	0.898	0.898	0.835	0.691	0.562
	11	DUVAL	6.873	6.056	6.056	5.567	4.477	3.561	1.000	0.881	0.881	0.810	0.651	0.518
	12	FRANKLIN	22.352	20.731	20.731	19.695	17.309	14.974	1.000	0.927	0.927	0.881	0.774	0.670
	13	GLADES	16.872	15.428	15.428	14.311	11.587	9.067	1.000	0.914	0.914	0.848	0.687	0.537
	14	HAMILTON	2.250	1.759	1.759	1.491	0.919	0.535	1.000	0.782	0.782	0.663	0.408	0.238
	15	HERNANDO	13.231	12.054	12.054	11.150	8.962	7.008	1.000	0.911	0.911	0.843	0.677	0.530
	16	HILLSBOROUGH	10.744	9.652	9.652	8.818	6.803	5.029	1.000	0.898	0.898	0.821	0.633	0.468
	17	HOLMES	5.109	4.274	4.274	3.780	2.677	1.805	1.000	0.836	0.836	0.740	0.524	0.353
Mobile Homes	18	INDIAN RIVER	44.321	42.532	42.532	41.088	37.463	33.562	1.000	0.960	0.960	0.927	0.845	0.757
	19	JACKSON	3.531	2.835	2.835	2.440	1.576	0.947	1.000	0.803	0.803	0.691	0.446	0.268
	20	LEE	19.542	18.057	18.057	16.857	13.881	11.043	1.000	0.924	0.924	0.863	0.710	0.565
	21	LEON	3.505	2.861	2.861	2.496	1.700	1.111	1.000	0.816	0.816	0.712	0.485	0.317
	22	MARION	6.004	5.178	5.178	4.574	3.167	2.097	1.000	0.862	0.862	0.762	0.527	0.349
	23	MARTIN	17.543	16.109	16.109	14.969	12.181	9.628	1.000	0.918	0.918	0.853	0.694	0.549
	24	MARTIN	46.080	44.258	44.258	42.738	38.879	34.685	1.000	0.960	0.960	0.927	0.844	0.753
	25	MIAMI-DADE	23.249	21.778	21.778	20.580	17.596	14.608	1.000	0.937	0.937	0.885	0.757	0.628
	26	MIAMI-DADE	35.836	34.099	34.099	32.671	29.089	25.348	1.000	0.952	0.952	0.912	0.812	0.707
	27	MONROE	53.076	51.355	51.355	49.892	46.124	41.763	1.000	0.968	0.968	0.940	0.869	0.787
	28	MONROE	75.209	73.190	73.190	71.474	67.035	61.774	1.000	0.973	0.973	0.950	0.891	0.821
	29	OKALOOSA	13.223	11.991	11.991	11.202	9.388	7.689	1.000	0.907	0.907	0.847	0.710	0.581
	30	OSCEOLA	8.262	7.234	7.234	6.468	4.654	3.186	1.000	0.875	0.875	0.783	0.563	0.386
	31	OSCEOLA	12.499	11.277	11.277	10.346	8.106	6.164	1.000	0.902	0.902	0.828	0.649	0.493
	32	PALM BEACH	25.089	23.449	23.449	22.119	18.810	15.557	1.000	0.935	0.935	0.882	0.750	0.620
	33	PALM BEACH	38.748	36.866	36.866	35.309	31.376	27.244	1.000	0.951	0.951	0.911	0.810	0.703
	34	PINELLAS	12.119	11.031	11.031	10.169	8.069	6.194	1.000	0.910	0.910	0.839	0.666	0.511
	35	SAINT JOHNS	5.268	4.542	4.542	4.113	3.162	2.389	1.000	0.862	0.862	0.781	0.600	0.453
	36	SANTA ROSA	9.466	8.459	8.459	7.826	6.382	5.062	1.000	0.894	0.894	0.827	0.674	0.535
	37	SEMINOLE	6.956	6.028	6.028	5.342	3.730	2.468	1.000	0.867	0.867	0.768	0.536	0.355
	38	TAYLOR	3.292	2.707	2.707	2.374	1.648	1.110	1.000	0.822	0.822	0.721	0.501	0.337
	39	VOLUSIA	11.403	10.420	10.420	9.676	7.894	6.318	1.000	0.914	0.914	0.849	0.692	0.554
	40	WAKULLA	9.882	8.792	8.792	8.122	6.601	5.274	1.000	0.890	0.890	0.822	0.668	0.534

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Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
Frame Renters	1	BAY	1.622	0.967	1.037	0.967	0.845	0.761	1.000	0.597	0.639	0.597	0.521	0.469
	2	BREVARD	0.923	0.448	0.519	0.448	0.310	0.264	1.000	0.485	0.563	0.485	0.335	0.286
	3	BREVARD	0.860	0.394	0.463	0.394	0.262	0.222	1.000	0.458	0.539	0.458	0.305	0.259
	4	BROWARD	2.168	1.373	1.532	1.373	1.034	0.853	1.000	0.633	0.707	0.633	0.477	0.393
	5	BROWARD	5.181	4.099	4.346	4.099	3.540	3.110	1.000	0.791	0.839	0.791	0.683	0.600
	6	CITRUS	0.645	0.254	0.321	0.254	0.130	0.102	1.000	0.395	0.498	0.395	0.201	0.158
	7	CLAY	0.239	0.053	0.061	0.053	0.043	0.041	1.000	0.222	0.255	0.222	0.179	0.171
	8	COLLIER	1.384	0.697	0.816	0.697	0.461	0.380	1.000	0.504	0.590	0.504	0.333	0.275
	9	COLUMBIA	0.230	0.049	0.057	0.049	0.040	0.038	1.000	0.214	0.246	0.214	0.174	0.167
	10	DIXIE	1.070	0.580	0.624	0.580	0.505	0.457	1.000	0.541	0.583	0.541	0.471	0.426
	11	DUVAL	0.534	0.271	0.283	0.271	0.249	0.235	1.000	0.507	0.530	0.507	0.466	0.439
	12	FRANKLIN	2.682	1.890	1.990	1.890	1.697	1.533	1.000	0.705	0.742	0.705	0.633	0.572
	13	GLADES	1.020	0.455	0.536	0.455	0.297	0.247	1.000	0.446	0.526	0.446	0.291	0.242
	14	HAMILTON	0.210	0.044	0.051	0.044	0.036	0.034	1.000	0.211	0.244	0.211	0.170	0.163
	15	HERNANDO	0.970	0.460	0.552	0.460	0.284	0.232	1.000	0.475	0.569	0.475	0.293	0.239
	16	HILLSBOROUGH	0.671	0.264	0.315	0.264	0.165	0.138	1.000	0.393	0.470	0.393	0.246	0.206
	17	HOLMES	0.430	0.117	0.135	0.117	0.093	0.087	1.000	0.272	0.315	0.272	0.216	0.203
	18	INDIAN RIVER	4.577	3.662	3.862	3.662	3.218	2.885	1.000	0.800	0.844	0.800	0.703	0.630
	19	JACKSON	0.316	0.070	0.081	0.070	0.055	0.052	1.000	0.220	0.258	0.220	0.173	0.166
	20	LEE	1.120	0.525	0.616	0.525	0.345	0.284	1.000	0.469	0.550	0.469	0.308	0.254
	21	LEON	0.307	0.078	0.089	0.078	0.062	0.059	1.000	0.253	0.292	0.253	0.203	0.191
	22	MARION	0.257	0.060	0.067	0.060	0.049	0.046	1.000	0.235	0.262	0.235	0.190	0.181
	23	MARTIN	1.412	0.697	0.850	0.697	0.397	0.309	1.000	0.494	0.602	0.494	0.281	0.219
	24	MARTIN	5.372	4.272	4.551	4.272	3.653	3.236	1.000	0.795	0.847	0.795	0.680	0.602
	25	MIAMI-DADE	1.938	1.210	1.355	1.210	0.901	0.737	1.000	0.624	0.699	0.624	0.465	0.380
	26	MIAMI-DADE	3.542	2.621	2.822	2.621	2.178	1.871	1.000	0.740	0.797	0.740	0.615	0.528
	27	MONROE	5.645	4.405	4.780	4.405	3.562	3.022	1.000	0.780	0.847	0.780	0.631	0.535
	28	MONROE	10.136	8.584	9.073	8.584	7.451	6.552	1.000	0.847	0.895	0.847	0.735	0.646
	29	OKALOOSA	1.056	0.499	0.552	0.499	0.416	0.372	1.000	0.473	0.523	0.473	0.394	0.352
	30	OSCEOLA	0.335	0.082	0.091	0.082	0.066	0.062	1.000	0.243	0.271	0.243	0.196	0.186
	31	OSCEOLA	0.449	0.131	0.145	0.131	0.108	0.101	1.000	0.293	0.322	0.293	0.241	0.224
	32	PALM BEACH	1.990	1.128	1.320	1.128	0.738	0.597	1.000	0.567	0.663	0.567	0.371	0.300
	33	PALM BEACH	3.579	2.511	2.766	2.511	1.970	1.692	1.000	0.702	0.773	0.702	0.551	0.473
	34	PINELLAS	1.013	0.467	0.580	0.467	0.252	0.196	1.000	0.461	0.573	0.461	0.248	0.193
	35	SAINT JOHNS	0.451	0.162	0.182	0.162	0.133	0.122	1.000	0.360	0.403	0.360	0.296	0.270
	36	SANTA ROSA	0.715	0.275	0.313	0.275	0.221	0.199	1.000	0.385	0.437	0.385	0.309	0.279
	37	SEMINOLE	0.654	0.254	0.328	0.254	0.117	0.088	1.000	0.388	0.501	0.388	0.178	0.135
	38	TAYLOR	0.295	0.074	0.086	0.074	0.059	0.055	1.000	0.250	0.290	0.250	0.198	0.186
	39	VOLUSIA	0.510	0.223	0.241	0.223	0.190	0.173	1.000	0.438	0.472	0.438	0.373	0.340
	40	WAKULLA	0.889	0.423	0.463	0.423	0.361	0.325	1.000	0.476	0.520	0.476	0.405	0.366

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
Masonry Renters	1	BAY	1.425	0.806	0.866	0.806	0.697	0.621	1.000	0.565	0.608	0.565	0.489	0.436
	2	BREVARD	0.895	0.428	0.499	0.428	0.292	0.248	1.000	0.478	0.557	0.478	0.326	0.277
	3	BREVARD	0.836	0.379	0.447	0.379	0.249	0.210	1.000	0.453	0.534	0.453	0.298	0.252
	4	BROWARD	2.088	1.303	1.461	1.303	0.971	0.797	1.000	0.624	0.700	0.624	0.465	0.382
	5	BROWARD	4.703	3.637	3.879	3.637	3.094	2.689	1.000	0.773	0.825	0.773	0.658	0.572
	6	CITRUS	0.631	0.247	0.313	0.247	0.125	0.098	1.000	0.392	0.496	0.392	0.198	0.156
	7	CLAY	0.219	0.046	0.053	0.046	0.038	0.037	1.000	0.212	0.240	0.212	0.175	0.167
	8	COLLIER	1.337	0.662	0.779	0.662	0.430	0.352	1.000	0.495	0.583	0.495	0.321	0.263
	9	COLUMBIA	0.212	0.043	0.049	0.043	0.036	0.034	1.000	0.203	0.230	0.203	0.169	0.162
	10	DIXIE	0.952	0.488	0.527	0.488	0.422	0.379	1.000	0.513	0.553	0.513	0.444	0.398
	11	DUVAL	0.410	0.163	0.172	0.163	0.146	0.137	1.000	0.397	0.421	0.397	0.356	0.335
	12	FRANKLIN	2.305	1.554	1.642	1.554	1.380	1.232	1.000	0.674	0.712	0.674	0.599	0.534
	13	GLADES	0.992	0.437	0.517	0.437	0.282	0.235	1.000	0.441	0.521	0.441	0.285	0.237
	14	HAMILTON	0.193	0.039	0.044	0.039	0.032	0.031	1.000	0.200	0.228	0.200	0.165	0.158
	15	HERNANDO	0.948	0.446	0.537	0.446	0.271	0.221	1.000	0.470	0.566	0.470	0.286	0.233
	16	HILLSBOROUGH	0.653	0.254	0.305	0.254	0.157	0.131	1.000	0.389	0.467	0.389	0.240	0.201
	17	HOLMES	0.396	0.104	0.118	0.104	0.084	0.078	1.000	0.262	0.299	0.262	0.212	0.198
	18	INDIAN RIVER	4.078	3.177	3.372	3.177	2.745	2.433	1.000	0.779	0.827	0.779	0.673	0.597
	19	JACKSON	0.291	0.061	0.070	0.061	0.049	0.047	1.000	0.209	0.241	0.209	0.169	0.161
	20	LEE	1.091	0.507	0.596	0.507	0.330	0.274	1.000	0.464	0.547	0.464	0.303	0.251
	21	LEON	0.283	0.069	0.078	0.069	0.056	0.053	1.000	0.243	0.276	0.243	0.199	0.187
	22	MARION	0.249	0.058	0.065	0.058	0.047	0.045	1.000	0.233	0.260	0.233	0.188	0.179
	23	MARTIN	1.357	0.655	0.804	0.655	0.369	0.291	1.000	0.482	0.592	0.482	0.272	0.214
	24	MARTIN	4.674	3.594	3.864	3.594	3.006	2.630	1.000	0.769	0.827	0.769	0.643	0.563
	25	MIAMI-DADE	1.873	1.154	1.298	1.154	0.851	0.693	1.000	0.616	0.693	0.616	0.454	0.370
	26	MIAMI-DADE	3.317	2.409	2.606	2.409	1.976	1.683	1.000	0.726	0.786	0.726	0.596	0.507
	27	MONROE	5.156	3.931	4.301	3.931	3.111	2.603	1.000	0.762	0.834	0.762	0.603	0.505
	28	MONROE	8.856	7.325	7.803	7.325	6.228	5.387	1.000	0.827	0.881	0.827	0.703	0.608
	29	OKALOOSA	0.968	0.442	0.487	0.442	0.368	0.327	1.000	0.456	0.503	0.456	0.380	0.338
	30	OSCEOLA	0.324	0.078	0.087	0.078	0.063	0.060	1.000	0.242	0.269	0.242	0.195	0.185
	31	OSCEOLA	0.436	0.127	0.140	0.127	0.104	0.097	1.000	0.292	0.321	0.292	0.240	0.223
	32	PALM BEACH	1.894	1.045	1.233	1.045	0.673	0.546	1.000	0.552	0.651	0.552	0.355	0.288
	33	PALM BEACH	3.258	2.208	2.456	2.208	1.692	1.440	1.000	0.678	0.754	0.678	0.519	0.442
	34	PINELLAS	0.979	0.441	0.552	0.441	0.231	0.179	1.000	0.451	0.564	0.451	0.236	0.182
	35	SAINT JOHNS	0.414	0.143	0.159	0.143	0.118	0.107	1.000	0.344	0.383	0.344	0.285	0.259
	36	SANTA ROSA	0.662	0.247	0.279	0.247	0.200	0.180	1.000	0.373	0.421	0.373	0.302	0.271
	37	SEMINOLE	0.640	0.246	0.319	0.246	0.112	0.086	1.000	0.384	0.498	0.384	0.175	0.134
	38	TAYLOR	0.273	0.065	0.074	0.065	0.053	0.049	1.000	0.238	0.273	0.238	0.193	0.180
	39	VOLUSIA	0.497	0.217	0.234	0.217	0.184	0.168	1.000	0.437	0.471	0.437	0.371	0.338
	40	WAKULLA	0.804	0.364	0.397	0.364	0.309	0.277	1.000	0.453	0.494	0.453	0.385	0.345

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
Frame Condo Unit	1	BAY	1.942	1.304	1.304	1.094	0.937	0.828	1.000	0.672	0.672	0.563	0.483	0.426
	2	BREVARD	1.324	0.784	0.784	0.545	0.385	0.292	1.000	0.593	0.593	0.412	0.291	0.221
	3	BREVARD	1.248	0.717	0.717	0.483	0.331	0.246	1.000	0.574	0.574	0.387	0.265	0.197
	4	BROWARD	2.860	2.045	2.045	1.642	1.239	0.958	1.000	0.715	0.715	0.574	0.433	0.335
	5	BROWARD	6.298	5.255	5.255	4.715	4.049	3.461	1.000	0.834	0.834	0.749	0.643	0.550
	6	CITRUS	0.936	0.510	0.510	0.321	0.182	0.114	1.000	0.545	0.545	0.343	0.195	0.122
	7	CLAY	0.312	0.114	0.114	0.059	0.045	0.042	1.000	0.364	0.364	0.188	0.144	0.134
	8	COLLIER	1.910	1.178	1.178	0.848	0.580	0.422	1.000	0.617	0.617	0.444	0.304	0.221
	9	COLUMBIA	0.303	0.109	0.109	0.054	0.042	0.039	1.000	0.358	0.358	0.179	0.138	0.129
	10	DIXIE	1.293	0.806	0.806	0.654	0.557	0.495	1.000	0.624	0.624	0.506	0.431	0.383
	11	DUVAL	0.687	0.390	0.390	0.301	0.271	0.250	1.000	0.568	0.568	0.439	0.394	0.364
	12	FRANKLIN	3.136	2.384	2.384	2.124	1.883	1.678	1.000	0.760	0.760	0.677	0.601	0.535
	13	GLADES	1.488	0.846	0.846	0.560	0.378	0.275	1.000	0.569	0.569	0.376	0.254	0.185
	14	HAMILTON	0.274	0.098	0.098	0.049	0.037	0.035	1.000	0.357	0.357	0.179	0.136	0.127
	15	HERNANDO	1.349	0.808	0.808	0.565	0.367	0.258	1.000	0.599	0.599	0.419	0.272	0.191
	16	HILLSBOROUGH	1.021	0.534	0.534	0.325	0.214	0.154	1.000	0.523	0.523	0.318	0.210	0.151
	17	HOLMES	0.553	0.227	0.227	0.132	0.099	0.090	1.000	0.410	0.410	0.238	0.179	0.163
	18	INDIAN RIVER	5.505	4.621	4.621	4.190	3.671	3.202	1.000	0.839	0.839	0.761	0.667	0.582
	19	JACKSON	0.411	0.151	0.151	0.077	0.057	0.053	1.000	0.367	0.367	0.188	0.139	0.130
	20	LEE	1.591	0.940	0.940	0.645	0.437	0.316	1.000	0.591	0.591	0.406	0.274	0.198
	21	LEON	0.397	0.156	0.156	0.087	0.066	0.061	1.000	0.393	0.393	0.219	0.166	0.153
	22	MARION	0.399	0.156	0.156	0.068	0.054	0.048	1.000	0.391	0.391	0.171	0.134	0.120
	23	MARTIN	1.926	1.202	1.202	0.862	0.525	0.343	1.000	0.624	0.624	0.447	0.273	0.178
	24	MARTIN	6.439	5.427	5.427	4.895	4.184	3.585	1.000	0.843	0.843	0.760	0.650	0.557
	25	MIAMI-DADE	2.568	1.821	1.821	1.452	1.083	0.829	1.000	0.709	0.709	0.565	0.422	0.323
	26	MIAMI-DADE	4.425	3.520	3.520	3.060	2.526	2.089	1.000	0.795	0.795	0.691	0.571	0.472
	27	MONROE	6.680	5.672	5.672	5.073	4.133	3.384	1.000	0.849	0.849	0.759	0.619	0.507
	28	MONROE	11.694	10.473	10.473	9.733	8.466	7.301	1.000	0.896	0.896	0.832	0.724	0.624
	29	OKALOOSA	1.293	0.744	0.744	0.570	0.459	0.401	1.000	0.575	0.575	0.441	0.355	0.310
	30	OSCEOLA	0.518	0.206	0.206	0.093	0.073	0.064	1.000	0.398	0.398	0.178	0.140	0.124
	31	OSCEOLA	0.680	0.294	0.294	0.151	0.120	0.105	1.000	0.432	0.432	0.222	0.177	0.155
	32	PALM BEACH	2.644	1.788	1.788	1.370	0.929	0.664	1.000	0.676	0.676	0.518	0.351	0.251
	33	PALM BEACH	4.487	3.466	3.466	2.946	2.325	1.878	1.000	0.772	0.772	0.657	0.518	0.419
	34	PINELLAS	1.400	0.837	0.837	0.580	0.342	0.218	1.000	0.598	0.598	0.415	0.244	0.155
	35	SAINT JOHNS	0.563	0.269	0.269	0.183	0.145	0.129	1.000	0.477	0.477	0.325	0.258	0.229
	36	SANTA ROSA	0.890	0.450	0.450	0.315	0.242	0.212	1.000	0.506	0.506	0.354	0.272	0.238
	37	SEMINOLE	0.941	0.513	0.513	0.323	0.171	0.099	1.000	0.545	0.545	0.343	0.181	0.105
	38	TAYLOR	0.376	0.147	0.147	0.083	0.062	0.057	1.000	0.390	0.390	0.221	0.166	0.151
	39	VOLUSIA	0.774	0.411	0.411	0.261	0.217	0.187	1.000	0.532	0.532	0.338	0.280	0.242
	40	WAKULLA	1.087	0.622	0.622	0.480	0.398	0.351	1.000	0.572	0.572	0.441	0.366	0.323

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
Masonry Condo Unit	1	BAY	1.723	1.112	1.112	0.914	0.775	0.675	1.000	0.646	0.646	0.531	0.450	0.392
	2	BREVARD	1.290	0.758	0.758	0.522	0.364	0.274	1.000	0.587	0.587	0.405	0.282	0.213
	3	BREVARD	1.219	0.695	0.695	0.465	0.315	0.233	1.000	0.570	0.570	0.381	0.258	0.191
	4	BROWARD	2.762	1.956	1.956	1.561	1.166	0.894	1.000	0.708	0.708	0.565	0.422	0.324
	5	BROWARD	5.768	4.737	4.737	4.208	3.558	2.995	1.000	0.821	0.821	0.729	0.617	0.519
	6	CITRUS	0.918	0.498	0.498	0.312	0.176	0.109	1.000	0.542	0.542	0.340	0.191	0.119
	7	CLAY	0.290	0.103	0.103	0.051	0.040	0.037	1.000	0.355	0.355	0.176	0.138	0.129
	8	COLLIER	1.853	1.132	1.132	0.807	0.544	0.390	1.000	0.611	0.611	0.435	0.294	0.211
	9	COLUMBIA	0.282	0.098	0.098	0.047	0.037	0.035	1.000	0.348	0.348	0.167	0.132	0.124
	10	DIXIE	1.160	0.695	0.695	0.552	0.467	0.410	1.000	0.599	0.599	0.476	0.403	0.354
	11	DUVAL	0.550	0.269	0.269	0.184	0.159	0.145	1.000	0.488	0.488	0.335	0.290	0.264
	12	FRANKLIN	2.721	2.000	2.000	1.755	1.536	1.349	1.000	0.735	0.735	0.645	0.565	0.496
	13	GLADES	1.450	0.819	0.819	0.538	0.360	0.261	1.000	0.565	0.565	0.371	0.248	0.180
	14	HAMILTON	0.255	0.088	0.088	0.042	0.033	0.031	1.000	0.346	0.346	0.166	0.130	0.122
	15	HERNANDO	1.321	0.786	0.786	0.548	0.351	0.245	1.000	0.595	0.595	0.415	0.266	0.185
	16	HILLSBOROUGH	0.997	0.518	0.518	0.313	0.204	0.146	1.000	0.520	0.520	0.314	0.205	0.146
	17	HOLMES	0.515	0.205	0.205	0.116	0.089	0.081	1.000	0.399	0.399	0.226	0.173	0.158
	18	INDIAN RIVER	4.954	4.081	4.081	3.656	3.149	2.700	1.000	0.824	0.824	0.738	0.636	0.545
	19	JACKSON	0.383	0.136	0.136	0.067	0.051	0.048	1.000	0.356	0.356	0.176	0.134	0.125
	20	LEE	1.550	0.910	0.910	0.622	0.418	0.303	1.000	0.587	0.587	0.401	0.270	0.195
	21	LEON	0.370	0.141	0.141	0.077	0.059	0.055	1.000	0.382	0.382	0.207	0.161	0.147
	22	MARION	0.388	0.151	0.151	0.066	0.052	0.046	1.000	0.390	0.390	0.170	0.133	0.118
	23	MARTIN	1.850	1.138	1.138	0.810	0.487	0.321	1.000	0.615	0.615	0.438	0.263	0.174
	24	MARTIN	5.678	4.679	4.679	4.164	3.482	2.928	1.000	0.824	0.824	0.733	0.613	0.516
	25	MIAMI-DADE	2.487	1.749	1.749	1.386	1.025	0.778	1.000	0.703	0.703	0.557	0.412	0.313
	26	MIAMI-DADE	4.168	3.273	3.273	2.821	2.299	1.878	1.000	0.785	0.785	0.677	0.551	0.450
	27	MONROE	6.138	5.139	5.139	4.552	3.634	2.921	1.000	0.837	0.837	0.742	0.592	0.476
	28	MONROE	10.302	9.092	9.092	8.367	7.134	6.028	1.000	0.883	0.883	0.812	0.692	0.585
	29	OKALOOSA	1.192	0.666	0.666	0.503	0.406	0.351	1.000	0.559	0.559	0.422	0.340	0.295
	30	OSCEOLA	0.503	0.200	0.200	0.089	0.070	0.062	1.000	0.397	0.397	0.178	0.138	0.123
	31	OSCEOLA	0.661	0.285	0.285	0.146	0.116	0.102	1.000	0.431	0.431	0.221	0.175	0.154
	32	PALM BEACH	2.519	1.677	1.677	1.273	0.849	0.605	1.000	0.665	0.665	0.505	0.337	0.240
	33	PALM BEACH	4.122	3.116	3.116	2.612	2.015	1.602	1.000	0.756	0.756	0.634	0.489	0.389
	34	PINELLAS	1.357	0.801	0.801	0.550	0.317	0.198	1.000	0.590	0.590	0.405	0.233	0.146
	35	SAINT JOHNS	0.522	0.241	0.241	0.161	0.129	0.114	1.000	0.462	0.462	0.308	0.246	0.218
	36	SANTA ROSA	0.828	0.408	0.408	0.281	0.218	0.191	1.000	0.493	0.493	0.340	0.264	0.230
	37	SEMINOLE	0.922	0.500	0.500	0.313	0.164	0.095	1.000	0.542	0.542	0.339	0.178	0.103
	38	TAYLOR	0.350	0.132	0.132	0.073	0.056	0.051	1.000	0.378	0.378	0.207	0.159	0.145
	39	VOLUSIA	0.757	0.400	0.400	0.253	0.210	0.180	1.000	0.529	0.529	0.335	0.277	0.238
	40	WAKULLA	0.991	0.546	0.546	0.413	0.341	0.298	1.000	0.551	0.551	0.416	0.344	0.301

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
			Loss Cost at different Deductibles						Ratios relative $0					
			0%	2%	3%	5%	10%		0%	2%	3%	5%	10%	
	1	BAY	14.376	12.227	11.425	10.044	7.376							
	2	BREVARD	8.598	6.915	6.334	5.376	3.639		1.000	0.804	0.737	0.625	0.423	
	3	BREVARD	7.822	6.182	5.619	4.701	3.054		1.000	0.790	0.718	0.601	0.390	
	4	BROWARD	15.876	13.425	12.492	10.872	7.703		1.000	0.846	0.787	0.685	0.485	
	5	BROWARD	27.147	24.063	22.847	20.697	16.314		1.000	0.886	0.842	0.762	0.601	
	6	CITRUS	4.341	3.123	2.745	2.153	1.208		1.000	0.719	0.632	0.496	0.278	
	7	CLAY	1.750	1.102	0.938	0.692	0.340		1.000	0.630	0.536	0.396	0.194	
	8	COLLIER	14.760	12.347	11.419	9.811	6.699		1.000	0.836	0.774	0.665	0.454	
	9	COLUMBIA	1.554	0.946	0.793	0.577	0.274		1.000	0.609	0.510	0.371	0.176	
	10	DIXIE	7.182	5.768	5.287	4.500	3.110		1.000	0.803	0.736	0.627	0.433	
	11	DUVAL	5.207	4.145	3.802	3.237	2.253		1.000	0.796	0.730	0.622	0.433	
	12	FRANKLIN	16.460	14.199	13.360	11.909	9.091		1.000	0.863	0.812	0.723	0.552	
	13	GLADES	10.698	8.597	7.839	6.559	4.195		1.000	0.804	0.733	0.613	0.392	
	14	HAMILTON	1.300	0.773	0.643	0.467	0.224		1.000	0.595	0.494	0.359	0.172	
	15	HERNANDO	7.346	5.777	5.245	4.373	2.810		1.000	0.786	0.714	0.595	0.382	
	16	HILLSBOROUGH	6.921	5.348	4.800	3.891	2.301		1.000	0.773	0.693	0.562	0.332	
	17	HOLMES	4.254	3.091	2.717	2.122	1.163		1.000	0.726	0.639	0.499	0.273	
	18	INDIAN RIVER	25.061	22.406	21.376	19.540	15.788		1.000	0.894	0.853	0.780	0.630	
	19	JACKSON	2.678	1.764	1.497	1.097	0.508		1.000	0.659	0.559	0.409	0.190	
Commercial Residential	20	LEE	11.515	9.341	8.542	7.178	4.630		1.000	0.811	0.742	0.623	0.402	
	21	LEON	2.289	1.535	1.324	1.008	0.528		1.000	0.671	0.578	0.441	0.231	
	22	MARION	3.291	2.259	1.957	1.493	0.777		1.000	0.686	0.595	0.454	0.236	
	23	MARTIN	11.263	9.146	8.382	7.106	4.740		1.000	0.812	0.744	0.631	0.421	
	24	MARTIN	27.686	24.791	23.630	21.567	17.334		1.000	0.895	0.854	0.779	0.626	
	25	MIAMI-DADE	15.877	13.545	12.638	11.041	7.878		1.000	0.853	0.796	0.695	0.496	
	26	MIAMI-DADE	22.893	20.163	19.095	17.194	13.303		1.000	0.881	0.834	0.751	0.581	
	27	MONROE	33.413	30.507	29.285	27.053	22.267		1.000	0.913	0.876	0.810	0.666	
	28	MONROE	40.034	36.987	35.707	33.367	28.220		1.000	0.924	0.892	0.833	0.705	
	29	OKALOOSA	11.959	10.046	9.336	8.096	5.695		1.000	0.840	0.781	0.677	0.476	
	30	OSCEOLA	4.856	3.487	3.052	2.367	1.262		1.000	0.718	0.629	0.487	0.260	
	31	OSCEOLA	7.123	5.475	4.923	4.042	2.512		1.000	0.769	0.691	0.568	0.353	
	32	PALM BEACH	15.494	13.008	12.072	10.470	7.392		1.000	0.840	0.779	0.676	0.477	
	33	PALM BEACH	24.399	21.415	20.233	18.137	13.935		1.000	0.878	0.829	0.743	0.571	
	34	PINELLAS	7.913	6.300	5.730	4.789	3.072		1.000	0.796	0.724	0.605	0.388	
	35	SAINT JOHNS	4.217	3.237	2.928	2.432	1.581		1.000	0.767	0.694	0.577	0.375	
	36	SANTA ROSA	7.934	6.468	5.934	5.009	3.247		1.000	0.815	0.748	0.631	0.409	
	37	SEMINOLE	3.945	2.743	2.373	1.808	0.933		1.000	0.695	0.602	0.458	0.237	
	38	TAYLOR	2.295	1.586	1.381	1.079	0.605		1.000	0.691	0.602	0.470	0.264	
	39	VOLUSIA	7.142	5.780	5.311	4.535	3.131		1.000	0.809	0.744	0.635	0.438	
	40	WAKULLA	7.038	5.639	5.171	4.404	3.043		1.000	0.801	0.735	0.626	0.432	

Form A-6: Logical Relationship to Risk - Construction

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Policy	Location	County	Loss Cost per Construction		Frame / Masonry
			Masonry	Frame	
	1	BAY	3.863	4.190	1.085
	2	BREVARD	4.521	4.590	1.015
	3	BREVARD	4.369	4.429	1.014
	4	BROWARD	7.980	8.204	1.028
	5	BROWARD	13.289	14.045	1.057
	6	CITRUS	3.273	3.324	1.016
	7	CLAY	0.862	0.895	1.038
	8	COLLIER	6.018	6.141	1.021
	9	COLUMBIA	0.850	0.885	1.041
	10	DIXIE	2.684	2.882	1.074
	11	DUVAL	1.690	1.878	1.111
	12	FRANKLIN	5.539	6.099	1.101
	13	GLADES	5.240	5.349	1.021
	14	HAMILTON	0.755	0.786	1.041
	15	HERNANDO	4.327	4.396	1.016
	16	HILLSBOROUGH	3.878	3.943	1.017
	17	HOLMES	1.456	1.527	1.049
	18	INDIAN RIVER	11.084	11.844	1.068
	19	JACKSON	1.117	1.167	1.045
Owners	20	LEE	5.311	5.439	1.024
	21	LEON	1.066	1.114	1.045
	22	MARION	1.585	1.621	1.023
	23	MARTIN	5.758	6.002	1.042
	24	MARTIN	12.669	13.648	1.077
	25	MIAMI-DADE	7.257	7.451	1.027
	26	MIAMI-DADE	10.416	10.846	1.041
	27	MONROE	12.698	13.480	1.062
	28	MONROE	19.295	21.049	1.091
	29	OKALOOSA	2.870	3.048	1.062
	30	OSCEOLA	2.046	2.095	1.024
	31	OSCEOLA	2.594	2.660	1.025
	32	PALM BEACH	7.405	7.732	1.044
	33	PALM BEACH	10.532	11.134	1.057
	34	PINELLAS	4.383	4.486	1.024
	35	SAINT JOHNS	1.354	1.424	1.052
	36	SANTA ROSA	2.101	2.216	1.055
	37	SEMINOLE	3.221	3.279	1.018
	38	TAYLOR	0.960	1.002	1.044
	39	VOLUSIA	2.950	2.997	1.016
	40	WAKULLA	2.386	2.541	1.065

Form A-6: Logical Relationship to Risk - Construction

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Policy	Location	County	Loss Cost per Construction		Frame / Masonry
			Masonry	Frame	
	1	BAY	1.425	1.622	1.138
	2	BREVARD	0.895	0.923	1.031
	3	BREVARD	0.836	0.860	1.028
	4	BROWARD	2.088	2.168	1.038
	5	BROWARD	4.703	5.181	1.102
	6	CITRUS	0.631	0.645	1.022
	7	CLAY	0.219	0.239	1.089
	8	COLLIER	1.337	1.384	1.034
	9	COLUMBIA	0.212	0.230	1.086
	10	DIXIE	0.952	1.070	1.124
	11	DUVAL	0.410	0.534	1.304
	12	FRANKLIN	2.305	2.682	1.164
	13	GLADES	0.992	1.020	1.028
	14	HAMILTON	0.193	0.210	1.087
	15	HERNANDO	0.948	0.970	1.024
	16	HILLSBOROUGH	0.653	0.671	1.027
	17	HOLMES	0.396	0.430	1.084
	18	INDIAN RIVER	4.078	4.577	1.123
	19	JACKSON	0.291	0.316	1.086
Renters	20	LEE	1.091	1.120	1.027
	21	LEON	0.283	0.307	1.085
	22	MARION	0.249	0.257	1.033
	23	MARTIN	1.357	1.412	1.041
	24	MARTIN	4.674	5.372	1.149
	25	MIAMI-DADE	1.873	1.938	1.035
	26	MIAMI-DADE	3.317	3.542	1.068
	27	MONROE	5.156	5.645	1.095
	28	MONROE	8.856	10.136	1.145
	29	OKALOOSA	0.968	1.056	1.091
	30	OSCEOLA	0.324	0.335	1.033
	31	OSCEOLA	0.436	0.449	1.030
	32	PALM BEACH	1.894	1.990	1.051
	33	PALM BEACH	3.258	3.579	1.098
	34	PINELLAS	0.979	1.013	1.035
	35	SAINT JOHNS	0.414	0.451	1.088
	36	SANTA ROSA	0.662	0.715	1.080
	37	SEMINOLE	0.640	0.654	1.023
	38	TAYLOR	0.273	0.295	1.082
	39	VOLUSIA	0.497	0.510	1.026
	40	WAKULLA	0.804	0.889	1.106

Form A-6: Logical Relationship to Risk - Construction

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Policy	Location	County	Loss Cost per Construction		Frame / Masonry
			Masonry	Frame	
	1	BAY	1.723	1.942	1.127
	2	BREVARD	1.290	1.324	1.026
	3	BREVARD	1.219	1.248	1.023
	4	BROWARD	2.762	2.860	1.036
	5	BROWARD	5.768	6.298	1.092
	6	CITRUS	0.918	0.936	1.020
	7	CLAY	0.290	0.312	1.075
	8	COLLIER	1.853	1.910	1.030
	9	COLUMBIA	0.282	0.303	1.074
	10	DIXIE	1.160	1.293	1.114
	11	DUVAL	0.550	0.687	1.249
	12	FRANKLIN	2.721	3.136	1.152
	13	GLADES	1.450	1.488	1.026
	14	HAMILTON	0.255	0.274	1.075
	15	HERNANDO	1.321	1.349	1.021
	16	HILLSBOROUGH	0.997	1.021	1.023
	17	HOLMES	0.515	0.553	1.075
	18	INDIAN RIVER	4.954	5.505	1.111
	19	JACKSON	0.383	0.411	1.075
Condo Unit	20	LEE	1.550	1.591	1.026
	21	LEON	0.370	0.397	1.075
	22	MARION	0.388	0.399	1.030
	23	MARTIN	1.850	1.926	1.041
	24	MARTIN	5.678	6.439	1.134
	25	MIAMI-DADE	2.487	2.568	1.033
	26	MIAMI-DADE	4.168	4.425	1.062
	27	MONROE	6.138	6.680	1.088
	28	MONROE	10.302	11.694	1.135
	29	OKALOOSA	1.192	1.293	1.085
	30	OSCEOLA	0.503	0.518	1.030
	31	OSCEOLA	0.661	0.680	1.029
	32	PALM BEACH	2.519	2.644	1.049
	33	PALM BEACH	4.122	4.487	1.088
	34	PINELLAS	1.357	1.400	1.032
	35	SAINT JOHNS	0.522	0.563	1.080
	36	SANTA ROSA	0.828	0.890	1.074
	37	SEMINOLE	0.922	0.941	1.021
	38	TAYLOR	0.350	0.376	1.073
	39	VOLUSIA	0.757	0.774	1.023
	40	WAKULLA	0.991	1.087	1.098

Form A-6: Logical Relationship to Risk - Construction

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Policy	Location	County	Loss Cost per Construction		Frame / Masonry
			Masonry	Frame	
Construction / Policy	Location	County	Loss Cost per Construction Concrete		
	1	BAY	14.376		
	2	BREVARD	8.598		
	3	BREVARD	7.822		
	4	BROWARD	15.876		
	5	BROWARD	27.147		
	6	CITRUS	4.341		
	7	CLAY	1.750		
	8	COLLIER	14.760		
	9	COLUMBIA	1.554		
	10	DIXIE	7.182		
	11	DUVAL	5.207		
	12	FRANKLIN	16.460		
	13	GLADES	10.698		
	14	HAMILTON	1.300		
	15	HERNANDO	7.346		
	16	HILLSBOROUGH	6.921		
	17	HOLMES	4.254		
	18	INDIAN RIVER	25.061		
	19	JACKSON	2.678		
Commercial Residential	20	LEE	11.515		
	21	LEON	2.289		
	22	MARION	3.291		
	23	MARTIN	11.263		
	24	MARTIN	27.686		
	25	MIAMI-DADE	15.877		
	26	MIAMI-DADE	22.893		
	27	MONROE	33.413		
	28	MONROE	40.034		
	29	OKALOOSA	11.959		
	30	OSCEOLA	4.856		
	31	OSCEOLA	7.123		
	32	PALM BEACH	15.494		
	33	PALM BEACH	24.399		
	34	PINELLAS	7.913		
	35	SAINT JOHNS	4.217		
	36	SANTA ROSA	7.934		
	37	SEMINOLE	3.945		
	38	TAYLOR	2.295		
	39	VOLUSIA	7.142		
	40	WAKULLA	7.038		

Form A-6: Logical Relationship to Risk - Policy Form

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Location	County	Loss Cost per Policy Type			Mobile homes / Frame Owners	Mobile Homes / Masonry Owners
		Frame Owners	Masonry Owners	Mobile Homes		
1	BAY	4.190	3.863	16.793	4.008	4.347
2	BREVARD	4.590	4.521	13.278	2.893	2.937
3	BREVARD	4.429	4.369	12.433	2.807	2.846
4	BROWARD	8.204	7.980	25.412	3.098	3.184
5	BROWARD	14.045	13.289	45.049	3.208	3.390
6	CITRUS	3.324	3.273	7.795	2.345	2.381
7	CLAY	0.895	0.862	2.717	3.036	3.152
8	COLLIER	6.141	6.018	23.503	3.827	3.906
9	COLUMBIA	0.885	0.850	2.556	2.888	3.005
10	DIXIE	2.882	2.684	11.053	3.835	4.118
11	DUVAL	1.878	1.690	6.873	3.660	4.067
12	FRANKLIN	6.099	5.539	22.352	3.665	4.035
13	GLADES	5.349	5.240	16.872	3.154	3.220
14	HAMILTON	0.786	0.755	2.250	2.862	2.981
15	HERNANDO	4.396	4.327	13.231	3.010	3.058
16	HILLSBOROUGH	3.943	3.878	10.744	2.725	2.771
17	HOLMES	1.527	1.456	5.109	3.346	3.508
18	INDIAN RIVER	11.844	11.084	44.321	3.742	3.999
19	JACKSON	1.167	1.117	3.531	3.025	3.160
20	LEE	5.439	5.311	19.542	3.593	3.680
21	LEON	1.114	1.066	3.505	3.146	3.289
22	MARION	1.621	1.585	6.004	3.704	3.789
23	MARTIN	6.002	5.758	17.543	2.923	3.047
24	MARTIN	13.648	12.669	46.080	3.376	3.637
25	MIAMI-DADE	7.451	7.257	23.249	3.120	3.204
26	MIAMI-DADE	10.846	10.416	35.836	3.304	3.440
27	MONROE	13.480	12.698	53.076	3.937	4.180
28	MONROE	21.049	19.295	75.209	3.573	3.898
29	OKALOOSA	3.048	2.870	13.223	4.338	4.608
30	OSCEOLA	2.095	2.046	8.262	3.945	4.039
31	OSCEOLA	2.660	2.594	12.499	4.700	4.819
32	PALM BEACH	7.732	7.405	25.089	3.245	3.388
33	PALM BEACH	11.134	10.532	38.748	3.480	3.679
34	PINELLAS	4.486	4.383	12.119	2.701	2.765
35	SAINT JOHNS	1.424	1.354	5.268	3.699	3.891
36	SANTA ROSA	2.216	2.101	9.466	4.271	4.506
37	SEMINOLE	3.279	3.221	6.956	2.121	2.160
38	TAYLOR	1.002	0.960	3.292	3.285	3.431
39	VOLUSIA	2.997	2.950	11.403	3.804	3.866
40	WAKULLA	2.541	2.386	9.882	3.889	4.142

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
	1	BAY	3.204	0.175	0.647	0.164	1.000	0.054	0.202	0.051
	2	BREVARD	4.005	0.123	0.385	0.076	1.000	0.031	0.096	0.019
	3	BREVARD	3.880	0.119	0.362	0.068	1.000	0.031	0.093	0.018
	4	BROWARD	6.921	0.199	0.885	0.199	1.000	0.029	0.128	0.029
	5	BROWARD	11.164	0.290	2.017	0.574	1.000	0.026	0.181	0.051
	6	CITRUS	2.916	0.086	0.284	0.039	1.000	0.029	0.097	0.013
	7	CLAY	0.731	0.044	0.102	0.018	1.000	0.060	0.139	0.024
	8	COLLIER	5.259	0.190	0.583	0.108	1.000	0.036	0.111	0.021
	9	COLUMBIA	0.728	0.042	0.098	0.017	1.000	0.058	0.135	0.024
	10	DIXIE	2.222	0.125	0.432	0.104	1.000	0.056	0.194	0.047
	11	DUVAL	1.528	0.083	0.210	0.057	1.000	0.054	0.138	0.037
	12	FRANKLIN	4.538	0.219	1.052	0.289	1.000	0.048	0.232	0.064
	13	GLADES	4.682	0.157	0.429	0.081	1.000	0.034	0.092	0.017
	14	HAMILTON	0.644	0.037	0.089	0.015	1.000	0.058	0.139	0.024
	15	HERNANDO	3.787	0.124	0.416	0.069	1.000	0.033	0.110	0.018
	16	HILLSBOROUGH	3.498	0.109	0.284	0.051	1.000	0.031	0.081	0.015
	17	HOLMES	1.238	0.075	0.180	0.034	1.000	0.060	0.146	0.028
	18	INDIAN RIVER	9.273	0.282	1.780	0.509	1.000	0.030	0.192	0.055
	19	JACKSON	0.953	0.056	0.135	0.024	1.000	0.059	0.141	0.025
Frame Owners	20	LEE	4.710	0.169	0.471	0.089	1.000	0.036	0.100	0.019
	21	LEON	0.907	0.054	0.129	0.024	1.000	0.059	0.143	0.026
	22	MARION	1.422	0.070	0.107	0.022	1.000	0.049	0.075	0.015
	23	MARTIN	5.141	0.155	0.613	0.093	1.000	0.030	0.119	0.018
	24	MARTIN	10.678	0.285	2.111	0.575	1.000	0.027	0.198	0.054
	25	MIAMI-DADE	6.300	0.183	0.793	0.176	1.000	0.029	0.126	0.028
	26	MIAMI-DADE	8.833	0.242	1.398	0.373	1.000	0.027	0.158	0.042
	27	MONROE	10.355	0.303	2.291	0.531	1.000	0.029	0.221	0.051
	28	MONROE	15.575	0.406	3.986	1.082	1.000	0.026	0.256	0.069
	29	OKALOOSA	2.373	0.147	0.429	0.099	1.000	0.062	0.181	0.042
	30	OSCEOLA	1.835	0.092	0.139	0.029	1.000	0.050	0.076	0.016
	31	OSCEOLA	2.313	0.123	0.184	0.041	1.000	0.053	0.079	0.018
	32	PALM BEACH	6.539	0.198	0.843	0.152	1.000	0.030	0.129	0.023
	33	PALM BEACH	9.081	0.264	1.448	0.341	1.000	0.029	0.159	0.038
	34	PINELLAS	3.867	0.112	0.444	0.062	1.000	0.029	0.115	0.016
	35	SAINT JOHNS	1.129	0.069	0.187	0.038	1.000	0.061	0.166	0.033
	36	SANTA ROSA	1.745	0.113	0.295	0.062	1.000	0.065	0.169	0.036
	37	SEMINOLE	2.872	0.080	0.292	0.036	1.000	0.028	0.102	0.012
	38	TAYLOR	0.805	0.049	0.126	0.022	1.000	0.061	0.156	0.027
	39	VOLUSIA	2.639	0.103	0.202	0.053	1.000	0.039	0.077	0.020
	40	WAKULLA	1.980	0.116	0.363	0.082	1.000	0.059	0.183	0.042

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
	1	BAY	2.976	0.175	0.569	0.143	1.000	0.059	0.191	0.048
	2	BREVARD	3.950	0.123	0.374	0.074	1.000	0.031	0.095	0.019
	3	BREVARD	3.832	0.119	0.352	0.066	1.000	0.031	0.092	0.017
	4	BROWARD	6.737	0.199	0.852	0.192	1.000	0.030	0.126	0.029
	5	BROWARD	10.647	0.290	1.833	0.519	1.000	0.027	0.172	0.049
	6	CITRUS	2.872	0.086	0.278	0.038	1.000	0.030	0.097	0.013
	7	CLAY	0.708	0.044	0.094	0.016	1.000	0.062	0.133	0.022
	8	COLLIER	5.159	0.190	0.564	0.104	1.000	0.037	0.109	0.020
	9	COLUMBIA	0.702	0.042	0.091	0.015	1.000	0.060	0.129	0.022
	10	DIXIE	2.083	0.125	0.385	0.091	1.000	0.060	0.185	0.044
	11	DUVAL	1.402	0.083	0.163	0.042	1.000	0.059	0.116	0.030
	12	FRANKLIN	4.168	0.219	0.905	0.247	1.000	0.053	0.217	0.059
	13	GLADES	4.587	0.157	0.417	0.078	1.000	0.034	0.091	0.017
	14	HAMILTON	0.621	0.037	0.083	0.014	1.000	0.060	0.133	0.022
	15	HERNANDO	3.730	0.124	0.407	0.067	1.000	0.033	0.109	0.018
	16	HILLSBOROUGH	3.442	0.109	0.277	0.049	1.000	0.032	0.081	0.014
	17	HOLMES	1.184	0.075	0.167	0.031	1.000	0.063	0.141	0.026
	18	INDIAN RIVER	8.764	0.282	1.588	0.451	1.000	0.032	0.181	0.051
	19	JACKSON	0.915	0.056	0.124	0.021	1.000	0.061	0.136	0.023
Masonry Owners	20	LEE	4.597	0.169	0.458	0.087	1.000	0.037	0.100	0.019
	21	LEON	0.871	0.054	0.120	0.022	1.000	0.062	0.138	0.025
	22	MARION	1.390	0.070	0.104	0.021	1.000	0.050	0.075	0.015
	23	MARTIN	4.924	0.155	0.589	0.090	1.000	0.031	0.120	0.018
	24	MARTIN	10.048	0.285	1.845	0.492	1.000	0.028	0.184	0.049
	25	MIAMI-DADE	6.138	0.183	0.766	0.170	1.000	0.030	0.125	0.028
	26	MIAMI-DADE	8.516	0.242	1.309	0.349	1.000	0.028	0.154	0.041
	27	MONROE	9.817	0.303	2.099	0.479	1.000	0.031	0.214	0.049
	28	MONROE	14.461	0.406	3.500	0.928	1.000	0.028	0.242	0.064
	29	OKALOOSA	2.239	0.147	0.394	0.090	1.000	0.066	0.176	0.040
	30	OSCEOLA	1.791	0.092	0.135	0.027	1.000	0.052	0.075	0.015
	31	OSCEOLA	2.254	0.123	0.178	0.039	1.000	0.054	0.079	0.017
	32	PALM BEACH	6.260	0.198	0.801	0.145	1.000	0.032	0.128	0.023
	33	PALM BEACH	8.638	0.264	1.323	0.307	1.000	0.031	0.153	0.035
	34	PINELLAS	3.781	0.112	0.430	0.059	1.000	0.030	0.114	0.016
	35	SAINT JOHNS	1.078	0.069	0.173	0.034	1.000	0.064	0.160	0.032
	36	SANTA ROSA	1.657	0.113	0.274	0.057	1.000	0.068	0.165	0.035
	37	SEMINOLE	2.821	0.080	0.285	0.035	1.000	0.029	0.101	0.012
	38	TAYLOR	0.774	0.049	0.117	0.020	1.000	0.064	0.151	0.025
	39	VOLUSIA	2.598	0.103	0.197	0.051	1.000	0.040	0.076	0.020
	40	WAKULLA	1.867	0.116	0.328	0.074	1.000	0.062	0.176	0.039

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
	1	BAY	13.094	0.175	2.543	0.981	1.000	0.013	0.194	0.075
	2	BREVARD	10.845	0.123	1.640	0.671	1.000	0.011	0.151	0.062
	3	BREVARD	10.257	0.119	1.460	0.597	1.000	0.012	0.142	0.058
	4	BROWARD	20.062	0.199	3.606	1.544	1.000	0.010	0.180	0.077
	5	BROWARD	33.758	0.290	7.754	3.246	1.000	0.009	0.230	0.096
	6	CITRUS	6.702	0.086	0.721	0.286	1.000	0.013	0.108	0.043
	7	CLAY	2.412	0.044	0.195	0.066	1.000	0.018	0.081	0.027
	8	COLLIER	18.724	0.190	3.239	1.349	1.000	0.010	0.173	0.072
	9	COLUMBIA	2.278	0.042	0.177	0.058	1.000	0.019	0.078	0.026
	10	DIXIE	8.819	0.125	1.520	0.588	1.000	0.014	0.172	0.067
	11	DUVAL	5.573	0.083	0.878	0.339	1.000	0.015	0.157	0.061
	12	FRANKLIN	17.004	0.219	3.700	1.429	1.000	0.013	0.218	0.084
	13	GLADES	13.828	0.157	2.039	0.848	1.000	0.011	0.147	0.061
	14	HAMILTON	2.008	0.037	0.154	0.050	1.000	0.019	0.077	0.025
	15	HERNANDO	10.840	0.124	1.609	0.658	1.000	0.011	0.148	0.061
	16	HILLSBOROUGH	9.049	0.109	1.123	0.462	1.000	0.012	0.124	0.051
	17	HOLMES	4.419	0.075	0.452	0.163	1.000	0.017	0.102	0.037
	18	INDIAN RIVER	32.629	0.282	8.140	3.271	1.000	0.009	0.249	0.100
	19	JACKSON	3.134	0.056	0.255	0.086	1.000	0.018	0.081	0.027
Mobile Homes	20	LEE	15.860	0.169	2.479	1.034	1.000	0.011	0.156	0.065
	21	LEON	3.059	0.054	0.290	0.102	1.000	0.018	0.095	0.033
	22	MARION	5.253	0.070	0.492	0.190	1.000	0.013	0.094	0.036
	23	MARTIN	14.321	0.155	2.150	0.917	1.000	0.011	0.150	0.064
	24	MARTIN	34.141	0.285	8.231	3.423	1.000	0.008	0.241	0.100
	25	MIAMI-DADE	18.375	0.183	3.285	1.407	1.000	0.010	0.179	0.077
	26	MIAMI-DADE	27.203	0.242	5.901	2.489	1.000	0.009	0.217	0.091
	27	MONROE	38.724	0.303	9.872	4.177	1.000	0.008	0.255	0.108
	28	MONROE	53.783	0.406	14.798	6.221	1.000	0.008	0.275	0.116
	29	OKALOOSA	10.501	0.147	1.854	0.721	1.000	0.014	0.177	0.069
	30	OSCEOLA	7.159	0.092	0.723	0.288	1.000	0.013	0.101	0.040
	31	OSCEOLA	10.393	0.123	1.407	0.576	1.000	0.012	0.135	0.055
	32	PALM BEACH	19.870	0.198	3.516	1.503	1.000	0.010	0.177	0.076
	33	PALM BEACH	29.486	0.264	6.326	2.673	1.000	0.009	0.215	0.091
	34	PINELLAS	10.035	0.112	1.396	0.576	1.000	0.011	0.139	0.057
	35	SAINT JOHNS	4.384	0.069	0.589	0.225	1.000	0.016	0.134	0.051
	36	SANTA ROSA	7.679	0.113	1.206	0.469	1.000	0.015	0.157	0.061
	37	SEMINOLE	6.084	0.080	0.570	0.222	1.000	0.013	0.094	0.037
	38	TAYLOR	2.852	0.049	0.287	0.103	1.000	0.017	0.101	0.036
	39	VOLUSIA	9.226	0.103	1.477	0.597	1.000	0.011	0.160	0.065
	40	WAKULLA	7.973	0.116	1.295	0.498	1.000	0.015	0.162	0.062

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
	1	BAY	0.000	0.000	1.294	0.328			1.000	0.253
	2	BREVARD	0.000	0.000	0.770	0.153			1.000	0.198
	3	BREVARD	0.000	0.000	0.723	0.136			1.000	0.189
	4	BROWARD	0.000	0.000	1.769	0.398			1.000	0.225
	5	BROWARD	0.000	0.000	4.034	1.147			1.000	0.284
	6	CITRUS	0.000	0.000	0.567	0.078			1.000	0.137
	7	CLAY	0.000	0.000	0.203	0.036			1.000	0.175
	8	COLLIER	0.000	0.000	1.167	0.217			1.000	0.186
	9	COLUMBIA	0.000	0.000	0.196	0.035			1.000	0.177
	10	DIXIE	0.000	0.000	0.863	0.207			1.000	0.240
	11	DUVAL	0.000	0.000	0.420	0.114			1.000	0.272
	12	FRANKLIN	0.000	0.000	2.104	0.577			1.000	0.274
	13	GLADES	0.000	0.000	0.859	0.161			1.000	0.188
	14	HAMILTON	0.000	0.000	0.179	0.031			1.000	0.173
	15	HERNANDO	0.000	0.000	0.832	0.138			1.000	0.166
	16	HILLSBOROUGH	0.000	0.000	0.569	0.102			1.000	0.179
	17	HOLMES	0.000	0.000	0.361	0.069			1.000	0.191
	18	INDIAN RIVER	0.000	0.000	3.560	1.017			1.000	0.286
	19	JACKSON	0.000	0.000	0.269	0.047			1.000	0.175
Frame Renters	20	LEE	0.000	0.000	0.943	0.177			1.000	0.188
	21	LEON	0.000	0.000	0.259	0.048			1.000	0.185
	22	MARION	0.000	0.000	0.214	0.043			1.000	0.201
	23	MARTIN	0.000	0.000	1.226	0.186			1.000	0.152
	24	MARTIN	0.000	0.000	4.221	1.150			1.000	0.272
	25	MIAMI-DADE	0.000	0.000	1.587	0.351			1.000	0.221
	26	MIAMI-DADE	0.000	0.000	2.796	0.746			1.000	0.267
	27	MONROE	0.000	0.000	4.582	1.063			1.000	0.232
	28	MONROE	0.000	0.000	7.972	2.164			1.000	0.272
	29	OKALOOSA	0.000	0.000	0.859	0.197			1.000	0.229
	30	OSCEOLA	0.000	0.000	0.278	0.057			1.000	0.206
	31	OSCEOLA	0.000	0.000	0.367	0.081			1.000	0.222
	32	PALM BEACH	0.000	0.000	1.685	0.305			1.000	0.181
	33	PALM BEACH	0.000	0.000	2.897	0.682			1.000	0.235
	34	PINELLAS	0.000	0.000	0.889	0.124			1.000	0.140
	35	SAINT JOHNS	0.000	0.000	0.375	0.076			1.000	0.202
	36	SANTA ROSA	0.000	0.000	0.591	0.124			1.000	0.210
	37	SEMINOLE	0.000	0.000	0.583	0.071			1.000	0.122
	38	TAYLOR	0.000	0.000	0.251	0.044			1.000	0.174
	39	VOLUSIA	0.000	0.000	0.405	0.105			1.000	0.260
	40	WAKULLA	0.000	0.000	0.725	0.164			1.000	0.227

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
	1	BAY	0.000	0.000	1.138	0.287			1.000	0.252
	2	BREVARD	0.000	0.000	0.747	0.148			1.000	0.198
	3	BREVARD	0.000	0.000	0.704	0.133			1.000	0.188
	4	BROWARD	0.000	0.000	1.703	0.385			1.000	0.226
	5	BROWARD	0.000	0.000	3.665	1.038			1.000	0.283
	6	CITRUS	0.000	0.000	0.555	0.076			1.000	0.136
	7	CLAY	0.000	0.000	0.188	0.032			1.000	0.169
	8	COLLIER	0.000	0.000	1.129	0.209			1.000	0.185
	9	COLUMBIA	0.000	0.000	0.181	0.031			1.000	0.171
	10	DIXIE	0.000	0.000	0.770	0.182			1.000	0.237
	11	DUVAL	0.000	0.000	0.325	0.085			1.000	0.260
	12	FRANKLIN	0.000	0.000	1.810	0.495			1.000	0.273
	13	GLADES	0.000	0.000	0.835	0.157			1.000	0.188
	14	HAMILTON	0.000	0.000	0.165	0.027			1.000	0.166
	15	HERNANDO	0.000	0.000	0.814	0.134			1.000	0.165
	16	HILLSBOROUGH	0.000	0.000	0.554	0.099			1.000	0.178
	17	HOLMES	0.000	0.000	0.334	0.062			1.000	0.187
	18	INDIAN RIVER	0.000	0.000	3.177	0.901			1.000	0.284
	19	JACKSON	0.000	0.000	0.249	0.042			1.000	0.170
Masonry Renters	20	LEE	0.000	0.000	0.917	0.174			1.000	0.190
	21	LEON	0.000	0.000	0.240	0.043			1.000	0.180
	22	MARION	0.000	0.000	0.208	0.041			1.000	0.199
	23	MARTIN	0.000	0.000	1.177	0.180			1.000	0.153
	24	MARTIN	0.000	0.000	3.689	0.984			1.000	0.267
	25	MIAMI-DADE	0.000	0.000	1.532	0.341			1.000	0.222
	26	MIAMI-DADE	0.000	0.000	2.619	0.698			1.000	0.267
	27	MONROE	0.000	0.000	4.199	0.957			1.000	0.228
	28	MONROE	0.000	0.000	7.001	1.855			1.000	0.265
	29	OKALOOSA	0.000	0.000	0.788	0.180			1.000	0.229
	30	OSCEOLA	0.000	0.000	0.269	0.055			1.000	0.204
	31	OSCEOLA	0.000	0.000	0.357	0.079			1.000	0.221
	32	PALM BEACH	0.000	0.000	1.603	0.291			1.000	0.181
	33	PALM BEACH	0.000	0.000	2.645	0.613			1.000	0.232
	34	PINELLAS	0.000	0.000	0.860	0.119			1.000	0.138
	35	SAINT JOHNS	0.000	0.000	0.346	0.068			1.000	0.198
	36	SANTA ROSA	0.000	0.000	0.548	0.114			1.000	0.209
	37	SEMINOLE	0.000	0.000	0.570	0.069			1.000	0.122
	38	TAYLOR	0.000	0.000	0.233	0.039			1.000	0.169
	39	VOLUSIA	0.000	0.000	0.394	0.103			1.000	0.260
	40	WAKULLA	0.000	0.000	0.657	0.147			1.000	0.224

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
	1	BAY	0.320	0.000	1.294	0.328	0.248		1.000	0.253
	2	BREVARD	0.400	0.000	0.770	0.153	0.520		1.000	0.198
	3	BREVARD	0.388	0.000	0.723	0.136	0.536		1.000	0.189
	4	BROWARD	0.692	0.000	1.769	0.398	0.391		1.000	0.225
	5	BROWARD	1.116	0.000	4.034	1.147	0.277		1.000	0.284
	6	CITRUS	0.292	0.000	0.567	0.078	0.514		1.000	0.137
	7	CLAY	0.073	0.000	0.203	0.036	0.360		1.000	0.175
	8	COLLIER	0.526	0.000	1.167	0.217	0.451		1.000	0.186
	9	COLUMBIA	0.073	0.000	0.196	0.035	0.372		1.000	0.177
	10	DIXIE	0.222	0.000	0.863	0.207	0.257		1.000	0.240
	11	DUVAL	0.153	0.000	0.420	0.114	0.364		1.000	0.272
	12	FRANKLIN	0.454	0.000	2.104	0.577	0.216		1.000	0.274
	13	GLADES	0.468	0.000	0.859	0.161	0.545		1.000	0.188
	14	HAMILTON	0.064	0.000	0.179	0.031	0.360		1.000	0.173
	15	HERNANDO	0.379	0.000	0.832	0.138	0.455		1.000	0.166
	16	HILLSBOROUGH	0.350	0.000	0.569	0.102	0.615		1.000	0.179
	17	HOLMES	0.124	0.000	0.361	0.069	0.343		1.000	0.191
	18	INDIAN RIVER	0.927	0.000	3.560	1.017	0.260		1.000	0.286
	19	JACKSON	0.095	0.000	0.269	0.047	0.354		1.000	0.175
Frame Condo Unit	20	LEE	0.471	0.000	0.943	0.177	0.500		1.000	0.188
	21	LEON	0.091	0.000	0.259	0.048	0.350		1.000	0.185
	22	MARION	0.142	0.000	0.214	0.043	0.664		1.000	0.201
	23	MARTIN	0.514	0.000	1.226	0.186	0.419		1.000	0.152
	24	MARTIN	1.068	0.000	4.221	1.150	0.253		1.000	0.272
	25	MIAMI-DADE	0.630	0.000	1.587	0.351	0.397		1.000	0.221
	26	MIAMI-DADE	0.883	0.000	2.796	0.746	0.316		1.000	0.267
	27	MONROE	1.035	0.000	4.582	1.063	0.226		1.000	0.232
	28	MONROE	1.557	0.000	7.972	2.164	0.195		1.000	0.272
	29	OKALOOSA	0.237	0.000	0.859	0.197	0.276		1.000	0.229
	30	OSCEOLA	0.183	0.000	0.278	0.057	0.661		1.000	0.206
	31	OSCEOLA	0.231	0.000	0.367	0.081	0.630		1.000	0.222
	32	PALM BEACH	0.654	0.000	1.685	0.305	0.388		1.000	0.181
	33	PALM BEACH	0.908	0.000	2.897	0.682	0.313		1.000	0.235
	34	PINELLAS	0.387	0.000	0.889	0.124	0.435		1.000	0.140
	35	SAINT JOHNS	0.113	0.000	0.375	0.076	0.301		1.000	0.202
	36	SANTA ROSA	0.175	0.000	0.591	0.124	0.295		1.000	0.210
	37	SEMINOLE	0.287	0.000	0.583	0.071	0.493		1.000	0.122
	38	TAYLOR	0.081	0.000	0.251	0.044	0.320		1.000	0.174
	39	VOLUSIA	0.264	0.000	0.405	0.105	0.652		1.000	0.260
	40	WAKULLA	0.198	0.000	0.725	0.164	0.273		1.000	0.227

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
	1	BAY	0.298	0.000	1.138	0.287	0.261		1.000	0.252
	2	BREVARD	0.395	0.000	0.747	0.148	0.529		1.000	0.198
	3	BREVARD	0.383	0.000	0.704	0.133	0.545		1.000	0.188
	4	BROWARD	0.674	0.000	1.703	0.385	0.396		1.000	0.226
	5	BROWARD	1.065	0.000	3.665	1.038	0.290		1.000	0.283
	6	CITRUS	0.287	0.000	0.555	0.076	0.517		1.000	0.136
	7	CLAY	0.071	0.000	0.188	0.032	0.377		1.000	0.169
	8	COLLIER	0.516	0.000	1.129	0.209	0.457		1.000	0.185
	9	COLUMBIA	0.070	0.000	0.181	0.031	0.388		1.000	0.171
	10	DIXIE	0.208	0.000	0.770	0.182	0.271		1.000	0.237
	11	DUVAL	0.140	0.000	0.325	0.085	0.431		1.000	0.260
	12	FRANKLIN	0.417	0.000	1.810	0.495	0.230		1.000	0.273
	13	GLADES	0.459	0.000	0.835	0.157	0.549		1.000	0.188
	14	HAMILTON	0.062	0.000	0.165	0.027	0.376		1.000	0.166
	15	HERNANDO	0.373	0.000	0.814	0.134	0.458		1.000	0.165
	16	HILLSBOROUGH	0.344	0.000	0.554	0.099	0.621		1.000	0.178
	17	HOLMES	0.118	0.000	0.334	0.062	0.354		1.000	0.187
	18	INDIAN RIVER	0.876	0.000	3.177	0.901	0.276		1.000	0.284
	19	JACKSON	0.092	0.000	0.249	0.042	0.368		1.000	0.170
Masonry Condo Unit	20	LEE	0.460	0.000	0.917	0.174	0.501		1.000	0.190
	21	LEON	0.087	0.000	0.240	0.043	0.363		1.000	0.180
	22	MARION	0.139	0.000	0.208	0.041	0.670		1.000	0.199
	23	MARTIN	0.492	0.000	1.177	0.180	0.418		1.000	0.153
	24	MARTIN	1.005	0.000	3.689	0.984	0.272		1.000	0.267
	25	MIAMI-DADE	0.614	0.000	1.532	0.341	0.401		1.000	0.222
	26	MIAMI-DADE	0.852	0.000	2.619	0.698	0.325		1.000	0.267
	27	MONROE	0.982	0.000	4.199	0.957	0.234		1.000	0.228
	28	MONROE	1.446	0.000	7.001	1.855	0.207		1.000	0.265
	29	OKALOOSA	0.224	0.000	0.788	0.180	0.284		1.000	0.229
	30	OSCEOLA	0.179	0.000	0.269	0.055	0.665		1.000	0.204
	31	OSCEOLA	0.225	0.000	0.357	0.079	0.632		1.000	0.221
	32	PALM BEACH	0.626	0.000	1.603	0.291	0.391		1.000	0.181
	33	PALM BEACH	0.864	0.000	2.645	0.613	0.327		1.000	0.232
	34	PINELLAS	0.378	0.000	0.860	0.119	0.440		1.000	0.138
	35	SAINT JOHNS	0.108	0.000	0.346	0.068	0.312		1.000	0.198
	36	SANTA ROSA	0.166	0.000	0.548	0.114	0.302		1.000	0.209
	37	SEMINOLE	0.282	0.000	0.570	0.069	0.495		1.000	0.122
	38	TAYLOR	0.077	0.000	0.233	0.039	0.332		1.000	0.169
	39	VOLUSIA	0.260	0.000	0.394	0.103	0.659		1.000	0.260
	40	WAKULLA	0.187	0.000	0.657	0.147	0.284		1.000	0.224

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
	1	BAY	14.076	0.000	0.300	0.000	1.000		0.021	0.000
	2	BREVARD	8.433	0.000	0.164	0.000	1.000		0.020	0.000
	3	BREVARD	7.675	0.000	0.147	0.000	1.000		0.019	0.000
	4	BROWARD	15.566	0.000	0.310	0.000	1.000		0.020	0.000
	5	BROWARD	26.572	0.000	0.575	0.000	1.000		0.022	0.000
	6	CITRUS	4.265	0.000	0.077	0.000	1.000		0.018	0.000
	7	CLAY	1.720	0.000	0.030	0.000	1.000		0.017	0.000
	8	COLLIER	14.473	0.000	0.287	0.000	1.000		0.020	0.000
	9	COLUMBIA	1.528	0.000	0.026	0.000	1.000		0.017	0.000
	10	DIXIE	7.039	0.000	0.143	0.000	1.000		0.020	0.000
	11	DUVAL	5.105	0.000	0.102	0.000	1.000		0.020	0.000
	12	FRANKLIN	16.108	0.000	0.353	0.000	1.000		0.022	0.000
	13	GLADES	10.498	0.000	0.201	0.000	1.000		0.019	0.000
	14	HAMILTON	1.279	0.000	0.022	0.000	1.000		0.017	0.000
	15	HERNANDO	7.208	0.000	0.138	0.000	1.000		0.019	0.000
	16	HILLSBOROUGH	6.796	0.000	0.126	0.000	1.000		0.018	0.000
	17	HOLMES	4.178	0.000	0.077	0.000	1.000		0.018	0.000
	18	INDIAN RIVER	24.505	0.000	0.556	0.000	1.000		0.023	0.000
	19	JACKSON	2.632	0.000	0.046	0.000	1.000		0.017	0.000
Commercial Residential	20	LEE	11.297	0.000	0.218	0.000	1.000		0.019	0.000
	21	LEON	2.249	0.000	0.040	0.000	1.000		0.018	0.000
	22	MARION	3.235	0.000	0.057	0.000	1.000		0.018	0.000
	23	MARTIN	11.052	0.000	0.211	0.000	1.000		0.019	0.000
	24	MARTIN	27.087	0.000	0.599	0.000	1.000		0.022	0.000
	25	MIAMI-DADE	15.567	0.000	0.311	0.000	1.000		0.020	0.000
	26	MIAMI-DADE	22.418	0.000	0.475	0.000	1.000		0.021	0.000
	27	MONROE	32.683	0.000	0.731	0.000	1.000		0.022	0.000
	28	MONROE	39.136	0.000	0.898	0.000	1.000		0.023	0.000
	29	OKALOOSA	11.717	0.000	0.242	0.000	1.000		0.021	0.000
	30	OSCEOLA	4.771	0.000	0.085	0.000	1.000		0.018	0.000
	31	OSCEOLA	6.992	0.000	0.131	0.000	1.000		0.019	0.000
	32	PALM BEACH	15.193	0.000	0.301	0.000	1.000		0.020	0.000
	33	PALM BEACH	23.895	0.000	0.505	0.000	1.000		0.021	0.000
	34	PINELLAS	7.765	0.000	0.148	0.000	1.000		0.019	0.000
	35	SAINT JOHNS	4.138	0.000	0.080	0.000	1.000		0.019	0.000
	36	SANTA ROSA	7.780	0.000	0.155	0.000	1.000		0.020	0.000
	37	SEMINOLE	3.877	0.000	0.068	0.000	1.000		0.017	0.000
	38	TAYLOR	2.254	0.000	0.041	0.000	1.000		0.018	0.000
	39	VOLUSIA	7.003	0.000	0.139	0.000	1.000		0.020	0.000
	40	WAKULLA	6.899	0.000	0.140	0.000	1.000		0.020	0.000

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
	1	BAY	4.682	2.941	2.015	1.000	0.628	0.430
	2	BREVARD	5.351	3.266	1.813	1.000	0.610	0.339
	3	BREVARD	5.152	3.171	1.784	1.000	0.615	0.346
	4	BROWARD	10.281	2.573	2.571	1.000	0.250	0.250
	5	BROWARD	17.945	3.749	3.756	1.000	0.209	0.209
	6	CITRUS	4.001	2.395	1.411	1.000	0.599	0.353
	7	CLAY	0.973	0.794	0.720	1.000	0.816	0.740
	8	COLLIER	8.089	4.787	2.556	1.000	0.592	0.316
	9	COLUMBIA	0.925	0.762	0.694	1.000	0.824	0.750
	10	DIXIE	3.202	2.149	1.589	1.000	0.671	0.496
	11	DUVAL	2.022	1.409	1.091	1.000	0.697	0.539
	12	FRANKLIN	7.044	4.257	2.618	1.000	0.604	0.372
	13	GLADES	7.255	4.141	2.303	1.000	0.571	0.317
	14	HAMILTON	0.817	0.673	0.616	1.000	0.824	0.754
	15	HERNANDO	5.360	3.251	1.819	1.000	0.606	0.339
	16	HILLSBOROUGH	5.088	2.976	1.704	1.000	0.585	0.335
	17	HOLMES	1.644	1.238	1.081	1.000	0.753	0.658
	18	INDIAN RIVER	15.737	9.125	4.447	1.000	0.580	0.283
	19	JACKSON	1.224	0.982	0.885	1.000	0.803	0.723
Frame Owners	20	LEE	7.808	4.378	2.357	1.000	0.561	0.302
	21	LEON	1.184	0.923	0.821	1.000	0.779	0.693
	22	MARION	3.294	2.007	1.198	1.000	0.609	0.364
	23	MARTIN	6.683	4.021	2.179	1.000	0.602	0.326
	24	MARTIN	15.325	8.685	3.993	1.000	0.567	0.261
	25	MIAMI-DADE	9.333	2.353	2.351	1.000	0.252	0.252
	26	MIAMI-DADE	13.934	2.965	2.967	1.000	0.213	0.213
	27	MONROE	14.103	7.376	3.726	1.000	0.523	0.264
	28	MONROE	21.957	12.092	6.583	1.000	0.551	0.300
	29	OKALOOSA	3.420	2.242	1.690	1.000	0.656	0.494
	30	OSCEOLA	4.338	2.609	1.528	1.000	0.601	0.352
	31	OSCEOLA	5.759	3.342	1.887	1.000	0.580	0.328
	32	PALM BEACH	8.691	5.064	2.605	1.000	0.583	0.300
	33	PALM BEACH	12.595	7.113	3.374	1.000	0.565	0.268
	34	PINELLAS	4.809	2.949	1.672	1.000	0.613	0.348
	35	SAINT JOHNS	1.531	1.139	0.950	1.000	0.744	0.620
	36	SANTA ROSA	2.425	1.684	1.354	1.000	0.694	0.558
	37	SEMINOLE	3.793	2.300	1.358	1.000	0.606	0.358
	38	TAYLOR	1.032	0.837	0.746	1.000	0.811	0.722
	39	VOLUSIA	4.504	2.735	1.516	1.000	0.607	0.337
	40	WAKULLA	2.804	1.923	1.476	1.000	0.686	0.527

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
	1	BAY	4.373	2.804	1.982	1.000	0.641	0.453
	2	BREVARD	5.271	3.210	1.774	1.000	0.609	0.337
	3	BREVARD	5.079	3.118	1.746	1.000	0.614	0.344
	4	BROWARD	10.036	2.532	2.532	1.000	0.252	0.252
	5	BROWARD	17.261	3.514	3.525	1.000	0.204	0.204
	6	CITRUS	3.941	2.354	1.370	1.000	0.597	0.348
	7	CLAY	0.925	0.770	0.712	1.000	0.833	0.770
	8	COLLIER	7.963	4.692	2.477	1.000	0.589	0.311
	9	COLUMBIA	0.879	0.739	0.686	1.000	0.840	0.780
	10	DIXIE	2.992	2.052	1.559	1.000	0.686	0.521
	11	DUVAL	1.903	1.357	1.080	1.000	0.713	0.568
	12	FRANKLIN	6.498	4.002	2.535	1.000	0.616	0.390
	13	GLADES	7.127	4.048	2.229	1.000	0.568	0.313
	14	HAMILTON	0.776	0.655	0.610	1.000	0.843	0.785
	15	HERNANDO	5.293	3.204	1.773	1.000	0.605	0.335
	16	HILLSBOROUGH	5.028	2.927	1.645	1.000	0.582	0.327
	17	HOLMES	1.554	1.196	1.068	1.000	0.770	0.687
	18	INDIAN RIVER	14.840	8.675	4.179	1.000	0.585	0.282
	19	JACKSON	1.159	0.952	0.875	1.000	0.821	0.755
Masonry Owners	20	LEE	7.625	4.240	2.305	1.000	0.556	0.302
	21	LEON	1.123	0.894	0.812	1.000	0.796	0.723
	22	MARION	3.242	1.968	1.164	1.000	0.607	0.359
	23	MARTIN	6.435	3.812	2.105	1.000	0.592	0.327
	24	MARTIN	14.539	8.251	3.797	1.000	0.568	0.261
	25	MIAMI-DADE	9.117	2.317	2.317	1.000	0.254	0.254
	26	MIAMI-DADE	13.512	2.882	2.886	1.000	0.213	0.214
	27	MONROE	13.409	7.034	3.386	1.000	0.525	0.252
	28	MONROE	20.292	11.104	5.433	1.000	0.547	0.268
	29	OKALOOSA	3.221	2.155	1.668	1.000	0.669	0.518
	30	OSCEOLA	4.262	2.553	1.485	1.000	0.599	0.348
	31	OSCEOLA	5.656	3.268	1.830	1.000	0.578	0.324
	32	PALM BEACH	8.379	4.812	2.513	1.000	0.574	0.300
	33	PALM BEACH	12.088	6.779	3.245	1.000	0.561	0.268
	34	PINELLAS	4.724	2.876	1.613	1.000	0.609	0.341
	35	SAINT JOHNS	1.448	1.100	0.939	1.000	0.760	0.649
	36	SANTA ROSA	2.289	1.622	1.337	1.000	0.708	0.584
	37	SEMINOLE	3.725	2.251	1.321	1.000	0.604	0.355
	38	TAYLOR	0.980	0.812	0.738	1.000	0.829	0.753
	39	VOLUSIA	4.446	2.690	1.477	1.000	0.605	0.332
	40	WAKULLA	2.636	1.848	1.458	1.000	0.701	0.553

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
Construction / Policy	**Location**	**County**	**Loss Cost per Year Built**			**Ratios Relative to 1974 Year Built**		
			Year Built 1974	**Year Built 1992**	**Year Built 2004**	**Year Built 1974**	**Year Built 1992**	**Year Built 2004**
	1	BAY	26.886	26.886	2.856	1.000	1.000	0.106
	2	BREVARD	15.919	15.919	2.021	1.000	1.000	0.127
	3	BREVARD	14.886	14.886	1.977	1.000	1.000	0.133
	4	BROWARD	29.392	29.392	2.853	1.000	1.000	0.097
	5	BROWARD	52.031	52.031	5.188	1.000	1.000	0.100
	6	CITRUS	9.268	9.268	1.515	1.000	1.000	0.164
	7	CLAY	4.072	4.072	0.847	1.000	1.000	0.208
	8	COLLIER	27.169	27.169	2.732	1.000	1.000	0.101
	9	COLUMBIA	3.818	3.818	0.812	1.000	1.000	0.213
	10	DIXIE	17.480	17.480	2.176	1.000	1.000	0.125
	11	DUVAL	10.818	10.818	1.426	1.000	1.000	0.132
	12	FRANKLIN	36.933	36.933	3.791	1.000	1.000	0.103
	13	GLADES	20.245	20.245	2.494	1.000	1.000	0.123
	14	HAMILTON	3.358	3.358	0.720	1.000	1.000	0.214
	15	HERNANDO	15.858	15.858	2.036	1.000	1.000	0.128
	16	HILLSBOROUGH	12.835	12.835	1.827	1.000	1.000	0.142
	17	HOLMES	7.840	7.840	1.339	1.000	1.000	0.171
	18	INDIAN RIVER	53.038	53.038	6.913	1.000	1.000	0.130
	19	JACKSON	5.318	5.318	1.064	1.000	1.000	0.200
Mobile Homes	20	LEE	22.549	22.549	2.497	1.000	1.000	0.111
	21	LEON	5.326	5.326	0.989	1.000	1.000	0.186
	22	MARION	7.113	7.113	1.278	1.000	1.000	0.180
	23	MARTIN	20.223	20.223	2.354	1.000	1.000	0.116
	24	MARTIN	53.156	53.156	5.571	1.000	1.000	0.105
	25	MIAMI-DADE	26.895	26.895	2.589	1.000	1.000	0.096
	26	MIAMI-DADE	41.516	41.516	3.608	1.000	1.000	0.087
	27	MONROE	61.402	61.402	5.540	1.000	1.000	0.090
	28	MONROE	86.089	86.089	11.131	1.000	1.000	0.129
	29	OKALOOSA	21.162	21.162	2.261	1.000	1.000	0.107
	30	OSCEOLA	9.815	9.815	1.641	1.000	1.000	0.167
	31	OSCEOLA	14.953	14.953	2.039	1.000	1.000	0.136
	32	PALM BEACH	29.011	29.011	2.859	1.000	1.000	0.099
	33	PALM BEACH	44.852	44.852	4.076	1.000	1.000	0.091
	34	PINELLAS	13.946	13.946	1.768	1.000	1.000	0.127
	35	SAINT JOHNS	8.219	8.219	1.194	1.000	1.000	0.145
	36	SANTA ROSA	15.038	15.038	1.773	1.000	1.000	0.118
	37	SEMINOLE	8.246	8.246	1.456	1.000	1.000	0.177
	38	TAYLOR	5.023	5.023	0.901	1.000	1.000	0.179
	39	VOLUSIA	13.681	13.681	1.689	1.000	1.000	0.123
	40	WAKULLA	15.629	15.629	1.947	1.000	1.000	0.125

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
Frame Renters	1	BAY	1.667	0.870	0.498	1.000	0.522	0.299
	2	BREVARD	0.766	0.480	0.345	1.000	0.627	0.450
	3	BREVARD	0.694	0.451	0.337	1.000	0.650	0.486
	4	BROWARD	2.151	0.503	0.504	1.000	0.234	0.234
	5	BROWARD	5.846	0.916	0.920	1.000	0.157	0.157
	6	CITRUS	0.442	0.304	0.257	1.000	0.688	0.583
	7	CLAY	0.224	0.169	0.143	1.000	0.751	0.639
	8	COLLIER	1.240	0.729	0.498	1.000	0.588	0.402
	9	COLUMBIA	0.212	0.162	0.138	1.000	0.762	0.651
	10	DIXIE	1.077	0.608	0.386	1.000	0.565	0.359
	11	DUVAL	0.631	0.368	0.242	1.000	0.583	0.384
	12	FRANKLIN	2.864	1.446	0.729	1.000	0.505	0.254
	13	GLADES	0.969	0.564	0.432	1.000	0.582	0.446
	14	HAMILTON	0.187	0.143	0.123	1.000	0.765	0.656
	15	HERNANDO	0.749	0.469	0.348	1.000	0.625	0.464
	16	HILLSBOROUGH	0.605	0.390	0.315	1.000	0.644	0.521
	17	HOLMES	0.407	0.284	0.239	1.000	0.699	0.587
	18	INDIAN RIVER	5.267	2.781	1.360	1.000	0.528	0.258
	19	JACKSON	0.285	0.216	0.186	1.000	0.760	0.654
	20	LEE	1.096	0.613	0.457	1.000	0.560	0.417
	21	LEON	0.285	0.204	0.172	1.000	0.716	0.603
	22	MARION	0.351	0.254	0.216	1.000	0.724	0.615
	23	MARTIN	0.908	0.562	0.418	1.000	0.619	0.460
	24	MARTIN	4.798	2.431	1.104	1.000	0.507	0.230
	25	MIAMI-DADE	1.905	0.458	0.459	1.000	0.241	0.241
	26	MIAMI-DADE	3.946	0.613	0.615	1.000	0.155	0.156
	27	MONROE	4.236	2.141	1.037	1.000	0.506	0.245
	28	MONROE	8.332	4.667	2.729	1.000	0.560	0.328
	29	OKALOOSA	1.069	0.595	0.401	1.000	0.556	0.375
	30	OSCEOLA	0.468	0.330	0.280	1.000	0.706	0.598
	31	OSCEOLA	0.747	0.454	0.352	1.000	0.607	0.471
	32	PALM BEACH	1.429	0.790	0.511	1.000	0.553	0.358
	33	PALM BEACH	2.986	1.501	0.747	1.000	0.503	0.250
	34	PINELLAS	0.620	0.411	0.317	1.000	0.663	0.511
	35	SAINT JOHNS	0.423	0.274	0.207	1.000	0.647	0.489
	36	SANTA ROSA	0.688	0.419	0.316	1.000	0.610	0.460
	37	SEMINOLE	0.400	0.290	0.248	1.000	0.725	0.619
	38	TAYLOR	0.253	0.188	0.156	1.000	0.741	0.617
	39	VOLUSIA	0.687	0.419	0.288	1.000	0.610	0.419
	40	WAKULLA	0.883	0.511	0.340	1.000	0.578	0.386

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
	1	BAY	1.546	0.809	0.459	1.000	0.524	0.297
	2	BREVARD	0.749	0.469	0.333	1.000	0.627	0.444
	3	BREVARD	0.679	0.440	0.326	1.000	0.648	0.480
	4	BROWARD	2.114	0.483	0.484	1.000	0.229	0.229
	5	BROWARD	5.585	0.741	0.748	1.000	0.133	0.134
	6	CITRUS	0.431	0.294	0.248	1.000	0.682	0.575
	7	CLAY	0.201	0.152	0.132	1.000	0.754	0.659
	8	COLLIER	1.215	0.714	0.482	1.000	0.588	0.396
	9	COLUMBIA	0.190	0.145	0.128	1.000	0.765	0.672
	10	DIXIE	0.987	0.558	0.351	1.000	0.566	0.355
	11	DUVAL	0.581	0.339	0.225	1.000	0.583	0.386
	12	FRANKLIN	2.607	1.325	0.639	1.000	0.508	0.245
	13	GLADES	0.951	0.551	0.417	1.000	0.579	0.439
	14	HAMILTON	0.167	0.129	0.113	1.000	0.770	0.678
	15	HERNANDO	0.730	0.456	0.336	1.000	0.625	0.460
	16	HILLSBOROUGH	0.594	0.379	0.304	1.000	0.638	0.512
	17	HOLMES	0.370	0.258	0.221	1.000	0.696	0.598
	18	INDIAN RIVER	4.760	2.600	1.226	1.000	0.546	0.258
	19	JACKSON	0.255	0.195	0.172	1.000	0.763	0.674
Masonry Renters	20	LEE	1.076	0.600	0.442	1.000	0.558	0.411
	21	LEON	0.258	0.185	0.159	1.000	0.715	0.617
	22	MARION	0.343	0.246	0.208	1.000	0.717	0.607
	23	MARTIN	0.891	0.549	0.403	1.000	0.616	0.453
	24	MARTIN	4.381	2.263	0.998	1.000	0.517	0.228
	25	MIAMI-DADE	1.876	0.441	0.441	1.000	0.235	0.235
	26	MIAMI-DADE	3.832	0.566	0.568	1.000	0.148	0.148
	27	MONROE	3.926	1.973	0.813	1.000	0.502	0.207
	28	MONROE	7.310	3.949	1.763	1.000	0.540	0.241
	29	OKALOOSA	1.002	0.555	0.374	1.000	0.554	0.373
	30	OSCEOLA	0.458	0.320	0.270	1.000	0.699	0.589
	31	OSCEOLA	0.732	0.443	0.340	1.000	0.605	0.464
	32	PALM BEACH	1.399	0.775	0.495	1.000	0.554	0.354
	33	PALM BEACH	2.834	1.447	0.711	1.000	0.510	0.251
	34	PINELLAS	0.608	0.401	0.306	1.000	0.660	0.503
	35	SAINT JOHNS	0.389	0.251	0.192	1.000	0.646	0.494
	36	SANTA ROSA	0.640	0.387	0.294	1.000	0.605	0.460
	37	SEMINOLE	0.391	0.281	0.239	1.000	0.718	0.610
	38	TAYLOR	0.229	0.170	0.145	1.000	0.743	0.632
	39	VOLUSIA	0.670	0.409	0.278	1.000	0.610	0.415
	40	WAKULLA	0.816	0.472	0.316	1.000	0.579	0.387

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
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Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
	1	BAY	2.035	1.103	0.657	1.000	0.542	0.323
	2	BREVARD	1.250	0.771	0.496	1.000	0.616	0.397
	3	BREVARD	1.162	0.733	0.487	1.000	0.631	0.419
	4	BROWARD	3.052	0.715	0.716	1.000	0.234	0.235
	5	BROWARD	7.319	1.216	1.220	1.000	0.166	0.167
	6	CITRUS	0.811	0.520	0.377	1.000	0.641	0.465
	7	CLAY	0.306	0.235	0.204	1.000	0.768	0.666
	8	COLLIER	1.968	1.152	0.710	1.000	0.585	0.361
	9	COLUMBIA	0.290	0.226	0.196	1.000	0.778	0.678
	10	DIXIE	1.331	0.780	0.513	1.000	0.586	0.386
	11	DUVAL	0.793	0.482	0.331	1.000	0.608	0.417
	12	FRANKLIN	3.403	1.777	0.932	1.000	0.522	0.274
	13	GLADES	1.630	0.934	0.625	1.000	0.573	0.383
	14	HAMILTON	0.256	0.200	0.174	1.000	0.781	0.682
	15	HERNANDO	1.235	0.758	0.500	1.000	0.613	0.405
	16	HILLSBOROUGH	1.073	0.657	0.459	1.000	0.612	0.428
	17	HOLMES	0.543	0.386	0.327	1.000	0.711	0.603
	18	INDIAN RIVER	6.549	3.526	1.709	1.000	0.538	0.261
	19	JACKSON	0.387	0.298	0.260	1.000	0.770	0.671
Frame Condo Unit	20	LEE	1.805	1.004	0.653	1.000	0.556	0.362
	21	LEON	0.384	0.281	0.240	1.000	0.731	0.625
	22	MARION	0.656	0.435	0.318	1.000	0.663	0.485
	23	MARTIN	1.515	0.920	0.599	1.000	0.607	0.395
	24	MARTIN	6.062	3.149	1.420	1.000	0.519	0.234
	25	MIAMI-DADE	2.725	0.653	0.653	1.000	0.239	0.240
	26	MIAMI-DADE	5.118	0.855	0.857	1.000	0.167	0.167
	27	MONROE	5.404	2.742	1.328	1.000	0.507	0.246
	28	MONROE	10.070	5.602	3.211	1.000	0.556	0.319
	29	OKALOOSA	1.343	0.775	0.535	1.000	0.577	0.399
	30	OSCEOLA	0.869	0.566	0.409	1.000	0.651	0.471
	31	OSCEOLA	1.274	0.753	0.511	1.000	0.591	0.401
	32	PALM BEACH	2.206	1.237	0.726	1.000	0.561	0.329
	33	PALM BEACH	4.070	2.111	1.021	1.000	0.519	0.251
	34	PINELLAS	1.059	0.674	0.457	1.000	0.637	0.432
	35	SAINT JOHNS	0.548	0.367	0.285	1.000	0.669	0.519
	36	SANTA ROSA	0.884	0.555	0.425	1.000	0.628	0.480
	37	SEMINOLE	0.751	0.497	0.363	1.000	0.662	0.483
	38	TAYLOR	0.339	0.257	0.218	1.000	0.759	0.643
	39	VOLUSIA	1.093	0.661	0.415	1.000	0.605	0.380
	40	WAKULLA	1.108	0.666	0.459	1.000	0.601	0.415

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
Masonry Condo Unit	1	BAY	1.888	1.032	0.617	1.000	0.547	0.327
	2	BREVARD	1.226	0.755	0.481	1.000	0.615	0.392
	3	BREVARD	1.141	0.718	0.472	1.000	0.629	0.414
	4	BROWARD	2.992	0.692	0.693	1.000	0.231	0.232
	5	BROWARD	7.003	1.026	1.034	1.000	0.147	0.148
	6	CITRUS	0.795	0.506	0.364	1.000	0.637	0.458
	7	CLAY	0.279	0.217	0.193	1.000	0.776	0.690
	8	COLLIER	1.931	1.129	0.686	1.000	0.584	0.355
	9	COLUMBIA	0.264	0.208	0.186	1.000	0.786	0.703
	10	DIXIE	1.224	0.723	0.476	1.000	0.591	0.389
	11	DUVAL	0.734	0.449	0.313	1.000	0.612	0.426
	12	FRANKLIN	3.105	1.637	0.839	1.000	0.527	0.270
	13	GLADES	1.600	0.912	0.604	1.000	0.570	0.377
	14	HAMILTON	0.233	0.184	0.165	1.000	0.790	0.709
	15	HERNANDO	1.211	0.741	0.484	1.000	0.612	0.400
	16	HILLSBOROUGH	1.056	0.642	0.443	1.000	0.608	0.419
	17	HOLMES	0.499	0.357	0.310	1.000	0.715	0.620
	18	INDIAN RIVER	5.978	3.310	1.554	1.000	0.554	0.260
	19	JACKSON	0.353	0.275	0.245	1.000	0.778	0.695
	20	LEE	1.768	0.977	0.633	1.000	0.553	0.358
	21	LEON	0.352	0.259	0.227	1.000	0.737	0.645
	22	MARION	0.643	0.423	0.307	1.000	0.658	0.478
	23	MARTIN	1.475	0.887	0.578	1.000	0.601	0.392
	24	MARTIN	5.587	2.946	1.299	1.000	0.527	0.232
	25	MIAMI-DADE	2.675	0.632	0.633	1.000	0.236	0.237
	26	MIAMI-DADE	4.967	0.802	0.804	1.000	0.161	0.162
	27	MONROE	5.041	2.547	1.081	1.000	0.505	0.214
	28	MONROE	8.933	4.821	2.178	1.000	0.540	0.244
	29	OKALOOSA	1.259	0.728	0.507	1.000	0.578	0.403
	30	OSCEOLA	0.852	0.550	0.395	1.000	0.646	0.464
	31	OSCEOLA	1.249	0.735	0.494	1.000	0.589	0.395
	32	PALM BEACH	2.147	1.198	0.701	1.000	0.558	0.327
	33	PALM BEACH	3.875	2.026	0.973	1.000	0.523	0.251
	34	PINELLAS	1.038	0.657	0.441	1.000	0.633	0.424
	35	SAINT JOHNS	0.507	0.342	0.269	1.000	0.673	0.531
	36	SANTA ROSA	0.826	0.519	0.402	1.000	0.628	0.487
	37	SEMINOLE	0.736	0.484	0.351	1.000	0.657	0.477
	38	TAYLOR	0.310	0.238	0.206	1.000	0.766	0.664
	39	VOLUSIA	1.070	0.647	0.401	1.000	0.604	0.375
	40	WAKULLA	1.028	0.622	0.434	1.000	0.605	0.423

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
	1	BAY	14.376	14.376	11.959	1.000	1.000	0.832
	2	BREVARD	8.598	8.598	6.715	1.000	1.000	0.781
	3	BREVARD	7.822	7.822	5.970	1.000	1.000	0.763
	4	BROWARD	15.876	12.878	12.878	1.000	0.811	0.811
	5	BROWARD	27.147	24.078	24.078	1.000	0.887	0.887
	6	CITRUS	4.341	4.341	3.052	1.000	1.000	0.703
	7	CLAY	1.750	1.750	1.149	1.000	1.000	0.657
	8	COLLIER	14.760	14.760	11.704	1.000	1.000	0.793
	9	COLUMBIA	1.554	1.554	1.000	1.000	1.000	0.644
	10	DIXIE	7.182	7.182	5.579	1.000	1.000	0.777
	11	DUVAL	5.207	5.207	4.085	1.000	1.000	0.785
	12	FRANKLIN	16.460	16.460	13.946	1.000	1.000	0.847
	13	GLADES	10.698	10.698	8.137	1.000	1.000	0.761
	14	HAMILTON	1.300	1.300	0.828	1.000	1.000	0.637
	15	HERNANDO	7.346	7.346	5.537	1.000	1.000	0.754
	16	HILLSBOROUGH	6.921	6.921	5.087	1.000	1.000	0.735
	17	HOLMES	4.254	4.254	3.016	1.000	1.000	0.709
	18	INDIAN RIVER	25.061	25.061	22.759	1.000	1.000	0.908
	19	JACKSON	2.678	2.678	1.781	1.000	1.000	0.665
Commercial Residential	20	LEE	11.515	11.515	8.803	1.000	1.000	0.765
	21	LEON	2.289	2.289	1.546	1.000	1.000	0.676
	22	MARION	3.291	3.291	2.246	1.000	1.000	0.682
	23	MARTIN	11.263	11.263	8.767	1.000	1.000	0.778
	24	MARTIN	27.686	27.686	25.100	1.000	1.000	0.907
	25	MIAMI-DADE	15.877	13.018	13.018	1.000	0.820	0.820
	26	MIAMI-DADE	22.893	19.934	19.934	1.000	0.871	0.871
	27	MONROE	33.413	33.413	31.116	1.000	1.000	0.931
	28	MONROE	40.034	40.034	38.029	1.000	1.000	0.950
	29	OKALOOSA	11.959	11.959	9.665	1.000	1.000	0.808
	30	OSCEOLA	4.856	4.856	3.388	1.000	1.000	0.698
	31	OSCEOLA	7.123	7.123	5.258	1.000	1.000	0.738
	32	PALM BEACH	15.494	15.494	12.492	1.000	1.000	0.806
	33	PALM BEACH	24.399	24.399	21.141	1.000	1.000	0.866
	34	PINELLAS	7.913	7.913	6.050	1.000	1.000	0.765
	35	SAINT JOHNS	4.217	4.217	3.179	1.000	1.000	0.754
	36	SANTA ROSA	7.934	7.934	6.096	1.000	1.000	0.768
	37	SEMINOLE	3.945	3.945	2.703	1.000	1.000	0.685
	38	TAYLOR	2.295	2.295	1.591	1.000	1.000	0.693
	39	VOLUSIA	7.142	7.142	5.615	1.000	1.000	0.786
	40	WAKULLA	7.038	7.038	5.491	1.000	1.000	0.780

Form A-6: Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
Frame Owners	1	BAY	6.153	2.941	1.995	1.000	0.478	0.324
	2	BREVARD	6.660	3.266	1.817	1.000	0.490	0.273
	3	BREVARD	6.374	3.171	1.791	1.000	0.497	0.281
	4	BROWARD	11.034	2.573	2.564	1.000	0.233	0.232
	5	BROWARD	19.111	3.749	3.550	1.000	0.196	0.186
	6	CITRUS	4.618	2.395	1.419	1.000	0.519	0.307
	7	CLAY	1.064	0.794	0.720	1.000	0.746	0.676
	8	COLLIER	10.265	4.787	2.557	1.000	0.466	0.249
	9	COLUMBIA	1.009	0.762	0.693	1.000	0.755	0.687
	10	DIXIE	4.090	2.149	1.573	1.000	0.525	0.385
	11	DUVAL	2.543	1.409	1.091	1.000	0.554	0.429
	12	FRANKLIN	9.236	4.257	2.547	1.000	0.461	0.276
	13	GLADES	8.490	4.141	2.307	1.000	0.488	0.272
	14	HAMILTON	0.890	0.673	0.616	1.000	0.756	0.692
	15	HERNANDO	6.679	3.251	1.822	1.000	0.487	0.273
	16	HILLSBOROUGH	5.869	2.976	1.702	1.000	0.507	0.290
	17	HOLMES	1.851	1.238	1.081	1.000	0.669	0.584
	18	INDIAN RIVER	20.268	9.125	4.025	1.000	0.450	0.199
	19	JACKSON	1.346	0.982	0.886	1.000	0.730	0.658
	20	LEE	9.172	4.378	2.409	1.000	0.477	0.263
	21	LEON	1.318	0.923	0.821	1.000	0.700	0.623
	22	MARION	3.759	2.007	1.202	1.000	0.534	0.320
	23	MARTIN	8.350	4.021	2.198	1.000	0.482	0.263
	24	MARTIN	19.614	8.685	3.737	1.000	0.443	0.191
	25	MIAMI-DADE	10.015	2.353	2.348	1.000	0.235	0.234
	26	MIAMI-DADE	14.817	2.965	2.896	1.000	0.200	0.195
	27	MONROE	21.595	7.376	3.642	1.000	0.342	0.169
	28	MONROE	32.451	12.092	5.986	1.000	0.373	0.184
	29	OKALOOSA	4.490	2.242	1.684	1.000	0.499	0.375
	30	OSCEOLA	4.966	2.609	1.538	1.000	0.525	0.310
	31	OSCEOLA	6.692	3.342	1.889	1.000	0.499	0.282
	32	PALM BEACH	11.060	5.064	2.614	1.000	0.458	0.236
	33	PALM BEACH	16.244	7.113	3.295	1.000	0.438	0.203
	34	PINELLAS	5.956	2.949	1.674	1.000	0.495	0.281
	35	SAINT JOHNS	1.883	1.139	0.949	1.000	0.605	0.504
	36	SANTA ROSA	3.118	1.684	1.354	1.000	0.540	0.434
	37	SEMINOLE	4.328	2.300	1.368	1.000	0.531	0.316
	38	TAYLOR	1.208	0.837	0.746	1.000	0.693	0.618
	39	VOLUSIA	5.626	2.735	1.506	1.000	0.486	0.268
	40	WAKULLA	3.569	1.923	1.470	1.000	0.539	0.412

Form A-6: Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
	1	BAY	5.724	2.804	1.962	1.000	0.490	0.343
	2	BREVARD	6.584	3.210	1.761	1.000	0.488	0.267
	3	BREVARD	6.310	3.118	1.737	1.000	0.494	0.275
	4	BROWARD	10.739	2.532	2.509	1.000	0.236	0.234
	5	BROWARD	18.196	3.514	3.381	1.000	0.193	0.186
	6	CITRUS	4.566	2.354	1.375	1.000	0.516	0.301
	7	CLAY	1.008	0.770	0.712	1.000	0.764	0.706
	8	COLLIER	10.140	4.692	2.468	1.000	0.463	0.243
	9	COLUMBIA	0.955	0.739	0.686	1.000	0.774	0.718
	10	DIXIE	3.806	2.052	1.547	1.000	0.539	0.406
	11	DUVAL	2.381	1.357	1.077	1.000	0.570	0.452
	12	FRANKLIN	8.490	4.002	2.487	1.000	0.471	0.293
	13	GLADES	8.383	4.048	2.225	1.000	0.483	0.265
	14	HAMILTON	0.843	0.655	0.609	1.000	0.777	0.723
	15	HERNANDO	6.602	3.204	1.764	1.000	0.485	0.267
	16	HILLSBOROUGH	5.803	2.927	1.643	1.000	0.504	0.283
	17	HOLMES	1.745	1.196	1.067	1.000	0.685	0.611
	18	INDIAN RIVER	18.990	8.675	3.813	1.000	0.457	0.201
	19	JACKSON	1.271	0.952	0.875	1.000	0.749	0.688
Masonry Owners	20	LEE	9.051	4.240	2.320	1.000	0.468	0.256
	21	LEON	1.245	0.894	0.811	1.000	0.718	0.651
	22	MARION	3.718	1.968	1.167	1.000	0.529	0.314
	23	MARTIN	8.152	3.812	2.156	1.000	0.468	0.264
	24	MARTIN	18.536	8.251	3.604	1.000	0.445	0.194
	25	MIAMI-DADE	9.758	2.317	2.299	1.000	0.237	0.236
	26	MIAMI-DADE	14.284	2.882	2.818	1.000	0.202	0.197
	27	MONROE	20.415	7.034	3.516	1.000	0.345	0.172
	28	MONROE	29.979	11.104	5.636	1.000	0.370	0.188
	29	OKALOOSA	4.215	2.155	1.660	1.000	0.511	0.394
	30	OSCEOLA	4.913	2.553	1.490	1.000	0.520	0.303
	31	OSCEOLA	6.608	3.268	1.826	1.000	0.495	0.276
	32	PALM BEACH	10.772	4.812	2.561	1.000	0.447	0.238
	33	PALM BEACH	15.629	6.779	3.212	1.000	0.434	0.206
	34	PINELLAS	5.903	2.876	1.621	1.000	0.487	0.275
	35	SAINT JOHNS	1.773	1.100	0.937	1.000	0.620	0.529
	36	SANTA ROSA	2.937	1.622	1.335	1.000	0.552	0.454
	37	SEMINOLE	4.281	2.251	1.326	1.000	0.526	0.310
	38	TAYLOR	1.141	0.812	0.738	1.000	0.712	0.647
	39	VOLUSIA	5.551	2.690	1.460	1.000	0.485	0.263
	40	WAKULLA	3.339	1.848	1.450	1.000	0.553	0.434

Form A-6: Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
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Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
	1	BAY	26.886	26.886	2.856	1.000	1.000	0.106
	2	BREVARD	15.919	15.919	2.021	1.000	1.000	0.127
	3	BREVARD	14.886	14.886	1.977	1.000	1.000	0.133
	4	BROWARD	29.392	29.392	2.853	1.000	1.000	0.097
	5	BROWARD	52.031	52.031	5.188	1.000	1.000	0.100
	6	CITRUS	9.268	9.268	1.515	1.000	1.000	0.164
	7	CLAY	4.072	4.072	0.847	1.000	1.000	0.208
	8	COLLIER	27.169	27.169	2.732	1.000	1.000	0.101
	9	COLUMBIA	3.818	3.818	0.812	1.000	1.000	0.213
	10	DIXIE	17.480	17.480	2.176	1.000	1.000	0.125
	11	DUVAL	10.818	10.818	1.426	1.000	1.000	0.132
	12	FRANKLIN	36.933	36.933	3.791	1.000	1.000	0.103
	13	GLADES	20.245	20.245	2.494	1.000	1.000	0.123
	14	HAMILTON	3.358	3.358	0.720	1.000	1.000	0.214
	15	HERNANDO	15.858	15.858	2.036	1.000	1.000	0.128
	16	HILLSBOROUGH	12.835	12.835	1.827	1.000	1.000	0.142
	17	HOLMES	7.840	7.840	1.339	1.000	1.000	0.171
	18	INDIAN RIVER	53.038	53.038	6.913	1.000	1.000	0.130
	19	JACKSON	5.318	5.318	1.064	1.000	1.000	0.200
Mobile Homes	20	LEE	22.549	22.549	2.497	1.000	1.000	0.111
	21	LEON	5.326	5.326	0.989	1.000	1.000	0.186
	22	MARION	7.113	7.113	1.278	1.000	1.000	0.180
	23	MARTIN	20.223	20.223	2.354	1.000	1.000	0.116
	24	MARTIN	53.156	53.156	5.571	1.000	1.000	0.105
	25	MIAMI-DADE	26.895	26.895	2.589	1.000	1.000	0.096
	26	MIAMI-DADE	41.516	41.516	3.608	1.000	1.000	0.087
	27	MONROE	61.402	61.402	5.540	1.000	1.000	0.090
	28	MONROE	86.089	86.089	11.131	1.000	1.000	0.129
	29	OKALOOSA	21.162	21.162	2.261	1.000	1.000	0.107
	30	OSCEOLA	9.815	9.815	1.641	1.000	1.000	0.167
	31	OSCEOLA	14.953	14.953	2.039	1.000	1.000	0.136
	32	PALM BEACH	29.011	29.011	2.859	1.000	1.000	0.099
	33	PALM BEACH	44.852	44.852	4.076	1.000	1.000	0.091
	34	PINELLAS	13.946	13.946	1.768	1.000	1.000	0.127
	35	SAINT JOHNS	8.219	8.219	1.194	1.000	1.000	0.145
	36	SANTA ROSA	15.038	15.038	1.773	1.000	1.000	0.118
	37	SEMINOLE	8.246	8.246	1.456	1.000	1.000	0.177
	38	TAYLOR	5.023	5.023	0.901	1.000	1.000	0.179
	39	VOLUSIA	13.681	13.681	1.689	1.000	1.000	0.123
	40	WAKULLA	15.629	15.629	1.947	1.000	1.000	0.125

Form A-6: Logical Relationship to Risk - Building Strength

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Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
Frame Renters	1	BAY	2.355	0.870	0.481	1.000	0.369	0.204
	2	BREVARD	1.072	0.480	0.339	1.000	0.448	0.316
	3	BREVARD	0.955	0.451	0.333	1.000	0.472	0.349
	4	BROWARD	2.125	0.503	0.493	1.000	0.237	0.232
	5	BROWARD	5.992	0.916	0.768	1.000	0.153	0.128
	6	CITRUS	0.536	0.304	0.257	1.000	0.568	0.479
	7	CLAY	0.248	0.169	0.143	1.000	0.679	0.577
	8	COLLIER	1.811	0.729	0.489	1.000	0.402	0.270
	9	COLUMBIA	0.233	0.162	0.138	1.000	0.693	0.591
	10	DIXIE	1.476	0.608	0.371	1.000	0.412	0.251
	11	DUVAL	0.856	0.368	0.240	1.000	0.430	0.280
	12	FRANKLIN	4.015	1.446	0.662	1.000	0.360	0.165
	13	GLADES	1.215	0.564	0.426	1.000	0.464	0.351
	14	HAMILTON	0.206	0.143	0.123	1.000	0.696	0.596
	15	HERNANDO	1.058	0.469	0.344	1.000	0.443	0.325
	16	HILLSBOROUGH	0.724	0.390	0.313	1.000	0.538	0.432
	17	HOLMES	0.464	0.284	0.238	1.000	0.613	0.513
	18	INDIAN RIVER	7.511	2.781	1.069	1.000	0.370	0.142
	19	JACKSON	0.314	0.216	0.186	1.000	0.689	0.592
	20	LEE	1.390	0.613	0.449	1.000	0.441	0.323
	21	LEON	0.322	0.204	0.172	1.000	0.634	0.532
	22	MARION	0.404	0.254	0.215	1.000	0.629	0.533
	23	MARTIN	1.265	0.562	0.414	1.000	0.444	0.327
	24	MARTIN	6.752	2.431	0.900	1.000	0.360	0.133
	25	MIAMI-DADE	1.876	0.458	0.451	1.000	0.244	0.240
	26	MIAMI-DADE	3.962	0.613	0.569	1.000	0.155	0.144
	27	MONROE	7.928	2.141	0.856	1.000	0.270	0.108
	28	MONROE	14.363	4.667	1.911	1.000	0.325	0.133
	29	OKALOOSA	1.513	0.595	0.396	1.000	0.393	0.262
	30	OSCEOLA	0.545	0.330	0.279	1.000	0.606	0.511
	31	OSCEOLA	0.920	0.454	0.347	1.000	0.493	0.377
	32	PALM BEACH	2.075	0.790	0.500	1.000	0.381	0.241
	33	PALM BEACH	4.342	1.501	0.678	1.000	0.346	0.156
	34	PINELLAS	0.849	0.411	0.314	1.000	0.484	0.370
	35	SAINT JOHNS	0.555	0.274	0.206	1.000	0.493	0.370
	36	SANTA ROSA	0.945	0.419	0.314	1.000	0.444	0.333
	37	SEMINOLE	0.460	0.290	0.247	1.000	0.630	0.536
	38	TAYLOR	0.308	0.188	0.156	1.000	0.610	0.506
	39	VOLUSIA	0.971	0.419	0.280	1.000	0.431	0.289
	40	WAKULLA	1.209	0.511	0.335	1.000	0.422	0.277

Form A-6: Logical Relationship to Risk - Building Strength

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Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
	1	BAY	2.171	0.809	0.450	1.000	0.373	0.207
	2	BREVARD	1.039	0.469	0.327	1.000	0.452	0.315
	3	BREVARD	0.928	0.440	0.322	1.000	0.474	0.346
	4	BROWARD	2.048	0.483	0.476	1.000	0.236	0.232
	5	BROWARD	5.483	0.741	0.681	1.000	0.135	0.124
	6	CITRUS	0.523	0.294	0.247	1.000	0.563	0.473
	7	CLAY	0.224	0.152	0.132	1.000	0.677	0.591
	8	COLLIER	1.756	0.714	0.473	1.000	0.407	0.269
	9	COLUMBIA	0.210	0.145	0.127	1.000	0.690	0.606
	10	DIXIE	1.350	0.558	0.345	1.000	0.414	0.256
	11	DUVAL	0.788	0.339	0.223	1.000	0.430	0.283
	12	FRANKLIN	3.631	1.325	0.617	1.000	0.365	0.170
	13	GLADES	1.186	0.551	0.412	1.000	0.464	0.347
	14	HAMILTON	0.185	0.129	0.113	1.000	0.696	0.612
	15	HERNANDO	1.023	0.456	0.333	1.000	0.446	0.325
	16	HILLSBOROUGH	0.710	0.379	0.302	1.000	0.534	0.425
	17	HOLMES	0.425	0.258	0.220	1.000	0.606	0.519
	18	INDIAN RIVER	6.653	2.600	1.000	1.000	0.391	0.150
	19	JACKSON	0.284	0.195	0.172	1.000	0.686	0.605
Masonry Renters	20	LEE	1.354	0.600	0.434	1.000	0.443	0.320
	21	LEON	0.293	0.185	0.159	1.000	0.629	0.541
	22	MARION	0.396	0.246	0.208	1.000	0.622	0.525
	23	MARTIN	1.235	0.549	0.400	1.000	0.444	0.324
	24	MARTIN	6.102	2.263	0.855	1.000	0.371	0.140
	25	MIAMI-DADE	1.815	0.441	0.435	1.000	0.243	0.240
	26	MIAMI-DADE	3.734	0.566	0.543	1.000	0.152	0.145
	27	MONROE	7.227	1.973	0.820	1.000	0.273	0.114
	28	MONROE	12.619	3.949	1.757	1.000	0.313	0.139
	29	OKALOOSA	1.415	0.555	0.370	1.000	0.392	0.261
	30	OSCEOLA	0.535	0.320	0.269	1.000	0.598	0.502
	31	OSCEOLA	0.898	0.443	0.335	1.000	0.493	0.373
	32	PALM BEACH	2.009	0.775	0.484	1.000	0.386	0.241
	33	PALM BEACH	4.066	1.447	0.655	1.000	0.356	0.161
	34	PINELLAS	0.828	0.401	0.303	1.000	0.484	0.366
	35	SAINT JOHNS	0.513	0.251	0.191	1.000	0.489	0.372
	36	SANTA ROSA	0.885	0.387	0.293	1.000	0.438	0.331
	37	SEMINOLE	0.451	0.281	0.238	1.000	0.623	0.528
	38	TAYLOR	0.281	0.170	0.144	1.000	0.605	0.513
	39	VOLUSIA	0.937	0.409	0.271	1.000	0.436	0.289
	40	WAKULLA	1.116	0.472	0.312	1.000	0.423	0.280

Form A-6: Logical Relationship to Risk - Building Strength

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Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
	1	BAY	4.870	1.103	0.736	1.000	0.227	0.151
	2	BREVARD	2.588	0.771	0.507	1.000	0.298	0.196
	3	BREVARD	2.326	0.733	0.495	1.000	0.315	0.213
	4	BROWARD	5.144	0.715	0.732	1.000	0.139	0.142
	5	BROWARD	12.259	1.216	1.384	1.000	0.099	0.113
	6	CITRUS	1.307	0.520	0.380	1.000	0.398	0.291
	7	CLAY	0.431	0.235	0.204	1.000	0.545	0.474
	8	COLLIER	4.505	1.152	0.730	1.000	0.256	0.162
	9	COLUMBIA	0.399	0.226	0.197	1.000	0.566	0.493
	10	DIXIE	2.967	0.780	0.561	1.000	0.263	0.189
	11	DUVAL	1.719	0.482	0.353	1.000	0.280	0.206
	12	FRANKLIN	7.881	1.777	1.137	1.000	0.225	0.144
	13	GLADES	3.088	0.934	0.630	1.000	0.302	0.204
	14	HAMILTON	0.349	0.200	0.175	1.000	0.572	0.500
	15	HERNANDO	2.547	0.758	0.516	1.000	0.298	0.203
	16	HILLSBOROUGH	1.830	0.657	0.460	1.000	0.359	0.251
	17	HOLMES	0.878	0.386	0.328	1.000	0.440	0.374
	18	INDIAN RIVER	14.316	3.526	2.205	1.000	0.246	0.154
	19	JACKSON	0.547	0.298	0.260	1.000	0.545	0.475
Frame Condo Unit	20	LEE	3.531	1.004	0.665	1.000	0.284	0.188
	21	LEON	0.591	0.281	0.241	1.000	0.475	0.407
	22	MARION	0.970	0.435	0.319	1.000	0.449	0.329
	23	MARTIN	3.169	0.920	0.611	1.000	0.290	0.193
	24	MARTIN	13.373	3.149	1.837	1.000	0.235	0.137
	25	MIAMI-DADE	4.612	0.653	0.666	1.000	0.141	0.144
	26	MIAMI-DADE	8.756	0.855	0.915	1.000	0.098	0.105
	27	MONROE	15.814	2.742	1.821	1.000	0.173	0.115
	28	MONROE	26.093	5.602	4.070	1.000	0.215	0.156
	29	OKALOOSA	3.296	0.775	0.557	1.000	0.235	0.169
	30	OSCEOLA	1.339	0.566	0.411	1.000	0.422	0.307
	31	OSCEOLA	2.280	0.753	0.515	1.000	0.330	0.226
	32	PALM BEACH	5.053	1.237	0.758	1.000	0.245	0.150
	33	PALM BEACH	9.501	2.111	1.188	1.000	0.222	0.125
	34	PINELLAS	2.104	0.674	0.462	1.000	0.320	0.220
	35	SAINT JOHNS	1.120	0.367	0.291	1.000	0.328	0.260
	36	SANTA ROSA	2.067	0.555	0.431	1.000	0.269	0.209
	37	SEMINOLE	1.113	0.497	0.364	1.000	0.447	0.327
	38	TAYLOR	0.574	0.257	0.219	1.000	0.448	0.381
	39	VOLUSIA	2.309	0.661	0.425	1.000	0.286	0.184
	40	WAKULLA	2.471	0.666	0.486	1.000	0.269	0.197

Form A-6: Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
	1	BAY	4.472	1.032	0.696	1.000	0.231	0.156
	2	BREVARD	2.510	0.755	0.491	1.000	0.301	0.195
	3	BREVARD	2.270	0.718	0.479	1.000	0.316	0.211
	4	BROWARD	4.935	0.692	0.707	1.000	0.140	0.143
	5	BROWARD	11.138	1.026	1.139	1.000	0.092	0.102
	6	CITRUS	1.291	0.506	0.366	1.000	0.392	0.284
	7	CLAY	0.404	0.217	0.193	1.000	0.536	0.477
	8	COLLIER	4.373	1.129	0.705	1.000	0.258	0.161
	9	COLUMBIA	0.373	0.208	0.186	1.000	0.556	0.497
	10	DIXIE	2.726	0.723	0.529	1.000	0.265	0.194
	11	DUVAL	1.589	0.449	0.335	1.000	0.283	0.211
	12	FRANKLIN	7.080	1.637	1.064	1.000	0.231	0.150
	13	GLADES	3.034	0.912	0.608	1.000	0.301	0.200
	14	HAMILTON	0.326	0.184	0.165	1.000	0.563	0.505
	15	HERNANDO	2.472	0.741	0.498	1.000	0.300	0.202
	16	HILLSBOROUGH	1.813	0.642	0.444	1.000	0.354	0.245
	17	HOLMES	0.833	0.357	0.310	1.000	0.429	0.373
	18	INDIAN RIVER	12.659	3.310	2.036	1.000	0.261	0.161
	19	JACKSON	0.516	0.275	0.246	1.000	0.533	0.476
Masonry Condo Unit	20	LEE	3.458	0.977	0.641	1.000	0.283	0.185
	21	LEON	0.558	0.259	0.227	1.000	0.465	0.408
	22	MARION	0.961	0.423	0.308	1.000	0.441	0.320
	23	MARTIN	3.089	0.887	0.593	1.000	0.287	0.192
	24	MARTIN	11.983	2.946	1.717	1.000	0.246	0.143
	25	MIAMI-DADE	4.444	0.632	0.644	1.000	0.142	0.145
	26	MIAMI-DADE	8.152	0.802	0.849	1.000	0.098	0.104
	27	MONROE	14.196	2.547	1.715	1.000	0.179	0.121
	28	MONROE	22.887	4.821	3.726	1.000	0.211	0.163
	29	OKALOOSA	3.096	0.728	0.529	1.000	0.235	0.171
	30	OSCEOLA	1.329	0.550	0.397	1.000	0.414	0.298
	31	OSCEOLA	2.234	0.735	0.497	1.000	0.329	0.223
	32	PALM BEACH	4.855	1.198	0.736	1.000	0.247	0.152
	33	PALM BEACH	8.812	2.026	1.139	1.000	0.230	0.129
	34	PINELLAS	2.066	0.657	0.447	1.000	0.318	0.216
	35	SAINT JOHNS	1.050	0.342	0.276	1.000	0.325	0.263
	36	SANTA ROSA	1.966	0.519	0.409	1.000	0.264	0.208
	37	SEMINOLE	1.105	0.484	0.352	1.000	0.438	0.318
	38	TAYLOR	0.541	0.238	0.207	1.000	0.439	0.382
	39	VOLUSIA	2.228	0.647	0.411	1.000	0.290	0.185
	40	WAKULLA	2.293	0.622	0.461	1.000	0.271	0.201

Form A-6: Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
	1	BAY	15.132	14.376	8.052	1.000	0.950	0.532
	2	BREVARD	9.126	8.598	4.185	1.000	0.942	0.459
	3	BREVARD	8.327	7.822	3.617	1.000	0.939	0.434
	4	BROWARD	16.771	12.878	8.258	1.000	0.768	0.492
	5	BROWARD	28.302	24.078	17.667	1.000	0.851	0.624
	6	CITRUS	4.666	4.341	1.717	1.000	0.930	0.368
	7	CLAY	1.894	1.750	0.630	1.000	0.924	0.333
	8	COLLIER	15.641	14.760	7.236	1.000	0.944	0.463
	9	COLUMBIA	1.685	1.554	0.544	1.000	0.922	0.323
	10	DIXIE	7.628	7.182	3.564	1.000	0.941	0.467
	11	DUVAL	5.523	5.207	2.637	1.000	0.943	0.477
	12	FRANKLIN	17.280	16.460	9.756	1.000	0.953	0.565
	13	GLADES	11.397	10.698	4.806	1.000	0.939	0.422
	14	HAMILTON	1.412	1.300	0.449	1.000	0.921	0.318
	15	HERNANDO	7.833	7.346	3.333	1.000	0.938	0.426
	16	HILLSBOROUGH	7.403	6.921	2.904	1.000	0.935	0.392
	17	HOLMES	4.569	4.254	1.685	1.000	0.931	0.369
	18	INDIAN RIVER	26.027	25.061	17.456	1.000	0.963	0.671
	19	JACKSON	2.895	2.678	0.952	1.000	0.925	0.329
Commercial Residential	20	LEE	12.259	11.515	5.233	1.000	0.939	0.427
	21	LEON	2.471	2.289	0.856	1.000	0.926	0.347
	22	MARION	3.549	3.291	1.234	1.000	0.927	0.348
	23	MARTIN	11.962	11.263	5.349	1.000	0.942	0.447
	24	MARTIN	28.763	27.686	19.012	1.000	0.963	0.661
	25	MIAMI-DADE	16.748	13.018	8.386	1.000	0.777	0.501
	26	MIAMI-DADE	23.939	19.934	14.110	1.000	0.833	0.589
	27	MONROE	34.569	33.413	24.136	1.000	0.967	0.698
	28	MONROE	41.275	40.034	30.479	1.000	0.970	0.738
	29	OKALOOSA	12.639	11.959	6.158	1.000	0.946	0.487
	30	OSCEOLA	5.224	4.856	1.868	1.000	0.929	0.358
	31	OSCEOLA	7.614	7.123	3.077	1.000	0.935	0.404
	32	PALM BEACH	16.380	15.494	7.966	1.000	0.946	0.486
	33	PALM BEACH	25.531	24.399	14.927	1.000	0.956	0.585
	34	PINELLAS	8.423	7.913	3.624	1.000	0.939	0.430
	35	SAINT JOHNS	4.496	4.217	1.932	1.000	0.938	0.430
	36	SANTA ROSA	8.442	7.934	3.628	1.000	0.940	0.430
	37	SEMINOLE	4.252	3.945	1.480	1.000	0.928	0.348
	38	TAYLOR	2.470	2.295	0.901	1.000	0.929	0.365
	39	VOLUSIA	7.575	7.142	3.539	1.000	0.943	0.467
	40	WAKULLA	7.472	7.038	3.495	1.000	0.942	0.468

Form A-6: Logical Relationship to Risk - Condo Unit Floor Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County / City	Loss Cost by Floor of Interest				Ratios Relative to 3rd Floor			
			3rd Floor	9th Floor	15th Floor	18th Floor	3rd Floor	9th Floor	15th Floor	18th Floor
Condo Unit A	1	BAY	1.279	4.426	7.867	16.582	1.000	3.461	6.151	12.966
	2	BREVARD	0.468	1.860	3.701	8.314	1.000	3.977	7.915	17.780
	3	BREVARD	0.404	1.532	3.155	7.198	1.000	3.790	7.804	17.804
	4	BROWARD	0.998	3.655	7.036	15.609	1.000	3.662	7.049	15.639
	5	BROWARD	3.550	10.163	16.212	31.172	1.000	2.863	4.567	8.782
	6	CITRUS	0.192	0.660	1.420	3.373	1.000	3.441	7.406	17.601
	7	CLAY	0.068	0.233	0.531	1.301	1.000	3.412	7.768	19.043
	8	COLLIER	0.838	3.103	6.187	14.006	1.000	3.704	7.386	16.720
	9	COLUMBIA	0.062	0.199	0.455	1.117	1.000	3.206	7.314	17.965
	10	DIXIE	0.815	2.043	3.306	6.734	1.000	2.506	4.056	8.261
	11	DUVAL	0.457	1.544	2.776	5.898	1.000	3.375	6.068	12.895
	12	FRANKLIN	2.430	6.080	9.422	17.488	1.000	2.502	3.877	7.195
	13	GLADES	0.502	1.857	3.906	9.044	1.000	3.701	7.784	18.024
	14	HAMILTON	0.057	0.177	0.395	0.969	1.000	3.111	6.953	17.039
	15	HERNANDO	0.477	1.510	2.898	6.461	1.000	3.164	6.073	13.536
	16	HILLSBOROUGH	0.283	1.088	2.419	5.820	1.000	3.837	8.534	20.532
	17	HOLMES	0.159	0.643	1.504	3.703	1.000	4.048	9.460	23.298
	18	INDIAN RIVER	5.168	12.044	17.169	28.885	1.000	2.331	3.322	5.589
	19	JACKSON	0.089	0.337	0.823	2.088	1.000	3.801	9.270	23.525
	20	LEE	0.613	2.149	4.413	10.129	1.000	3.502	7.193	16.511
	21	LEON	0.107	0.352	0.778	1.862	1.000	3.300	7.289	17.447
	22	MARION	0.129	0.446	1.001	2.430	1.000	3.446	7.733	18.780
	23	MARTIN	0.536	2.147	4.478	10.326	1.000	4.002	8.346	19.247
	24	MARTIN	4.297	11.802	18.140	33.562	1.000	2.746	4.221	7.810
	25	MIAMI-DADE	0.890	3.662	7.355	16.564	1.000	4.112	8.260	18.602
	26	MIAMI-DADE	2.092	7.392	12.810	26.300	1.000	3.534	6.124	12.573
	27	MONROE	5.444	16.656	25.864	48.109	1.000	3.060	4.751	8.838
	28	MONROE	10.502	21.775	29.639	52.847	1.000	2.073	2.822	5.032
	29	OKALOOSA	0.739	2.957	5.920	13.254	1.000	4.000	8.008	17.927
	30	OSCEOLA	0.179	0.644	1.462	3.544	1.000	3.591	8.156	19.771
	31	OSCEOLA	0.378	1.270	2.589	5.961	1.000	3.357	6.845	15.759
	32	PALM BEACH	0.950	3.483	6.737	15.067	1.000	3.667	7.093	15.861
	33	PALM BEACH	2.358	7.881	13.392	27.061	1.000	3.342	5.678	11.474
	34	PINELLAS	0.362	1.500	3.197	7.442	1.000	4.140	8.822	20.538
	35	SAINT JOHNS	0.254	0.931	1.880	4.198	1.000	3.672	7.410	16.550
	36	SANTA ROSA	0.461	1.692	3.511	8.140	1.000	3.667	7.608	17.642
	37	SEMINOLE	0.146	0.514	1.171	2.846	1.000	3.531	8.046	19.556
	38	TAYLOR	0.105	0.366	0.811	1.931	1.000	3.486	7.735	18.408
	39	VOLUSIA	0.466	1.743	3.323	7.382	1.000	3.742	7.135	15.852
	40	WAKULLA	0.594	1.847	3.298	7.023	1.000	3.111	5.554	11.830

Form A-6: Logical Relationship to Risk - Condo Unit Floor Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County / City	Loss Cost by Floor of Interest				Ratios Relative to 3rd Floor			
			3rd Floor	9th Floor	15th Floor	18th Floor	3rd Floor	9th Floor	15th Floor	18th Floor
	1	BAY	4.557	14.741	21.817	41.424	1.000	3.235	4.788	9.090
	2	BREVARD	1.868	7.883	12.615	25.270	1.000	4.220	6.754	13.530
	3	BREVARD	1.631	7.083	11.566	23.508	1.000	4.344	7.094	14.418
	4	BROWARD	3.980	14.382	21.990	43.163	1.000	3.614	5.525	10.845
	5	BROWARD	10.567	26.615	36.663	64.338	1.000	2.519	3.470	6.089
	6	CITRUS	0.748	3.842	6.671	14.062	1.000	5.137	8.921	18.803
	7	CLAY	0.227	1.552	2.970	6.521	1.000	6.845	13.095	28.753
	8	COLLIER	3.469	13.556	21.096	42.056	1.000	3.908	6.081	12.124
	9	COLUMBIA	0.200	1.348	2.623	5.794	1.000	6.744	13.119	28.977
	10	DIXIE	2.740	7.652	10.955	20.726	1.000	2.793	3.998	7.564
	11	DUVAL	1.665	5.777	8.859	17.196	1.000	3.470	5.321	10.328
	12	FRANKLIN	7.464	17.803	24.233	42.393	1.000	2.385	3.247	5.680
	13	GLADES	2.119	9.275	15.012	30.624	1.000	4.378	7.086	14.456
	14	HAMILTON	0.182	1.202	2.301	5.078	1.000	6.595	12.628	27.869
	15	HERNANDO	1.879	7.015	11.019	22.085	1.000	3.734	5.865	11.755
	16	HILLSBOROUGH	1.173	6.136	10.475	21.926	1.000	5.232	8.933	18.698
	17	HOLMES	0.613	3.893	7.073	15.183	1.000	6.348	11.534	24.760
	18	INDIAN RIVER	13.225	27.470	35.439	57.054	1.000	2.077	2.680	4.314
	19	JACKSON	0.309	2.370	4.583	10.150	1.000	7.661	14.817	32.811
Condo Unit B	20	LEE	2.595	10.535	16.744	33.827	1.000	4.060	6.452	13.036
	21	LEON	0.393	2.184	3.948	8.471	1.000	5.559	10.048	21.559
	22	MARION	0.485	2.846	5.166	11.099	1.000	5.862	10.642	22.861
	23	MARTIN	2.238	9.796	15.827	32.198	1.000	4.377	7.071	14.386
	24	MARTIN	12.020	28.337	38.150	64.985	1.000	2.357	3.174	5.406
	25	MIAMI-DADE	3.614	14.451	22.517	44.550	1.000	3.999	6.231	12.329
	26	MIAMI-DADE	7.301	21.889	31.326	57.347	1.000	2.998	4.290	7.854
	27	MONROE	15.937	38.307	50.896	84.583	1.000	2.404	3.194	5.307
	28	MONROE	25.336	45.858	55.842	88.861	1.000	1.810	2.204	3.507
	29	OKALOOSA	3.013	11.972	18.716	37.046	1.000	3.974	6.213	12.297
	30	OSCEOLA	0.700	4.071	7.276	15.521	1.000	5.816	10.394	22.171
	31	OSCEOLA	1.547	6.486	10.556	21.633	1.000	4.193	6.823	13.983
	32	PALM BEACH	3.846	13.917	21.337	42.105	1.000	3.619	5.548	10.948
	33	PALM BEACH	7.907	23.107	32.933	60.056	1.000	2.922	4.165	7.595
	34	PINELLAS	1.488	7.174	11.903	24.381	1.000	4.822	8.000	16.386
	35	SAINT JOHNS	0.958	4.204	6.954	14.128	1.000	4.387	7.257	14.744
	36	SANTA ROSA	1.949	8.289	13.267	26.894	1.000	4.252	6.806	13.797
	37	SEMINOLE	0.551	3.308	6.003	12.902	1.000	6.002	10.891	23.408
	38	TAYLOR	0.391	2.156	3.877	8.290	1.000	5.506	9.904	21.176
	39	VOLUSIA	1.824	7.080	11.025	21.885	1.000	3.882	6.045	12.000
	40	WAKULLA	2.204	7.204	10.854	21.060	1.000	3.269	4.926	9.557

Form A-6: Logical Relationship to Risk - Number of Stories Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County / City	Loss Cost by Number of Stories		Ratios Relative to 1 Story		Loss Cost by Number of Stories			Ratios Relative to 5 Story		
			1 Story	2 Story	1 Story	2 Story	5 Story	10 Story	20 Story	5 Story	10 Story	20 Story
Frame Owners	1	BAY	3.841	5.364	1.000	1.397						
	2	BREVARD	4.681	5.286	1.000	1.129						
	3	BREVARD	4.526	5.028	1.000	1.111						
	4	BROWARD	7.409	8.803	1.000	1.188						
	5	BROWARD	12.721	16.507	1.000	1.298						
	6	CITRUS	3.417	3.600	1.000	1.053						
	7	CLAY	0.867	0.909	1.000	1.049						
	8	COLLIER	6.962	8.127	1.000	1.167						
	9	COLUMBIA	0.829	0.864	1.000	1.043						
	10	DIXIE	2.698	3.549	1.000	1.315						
	11	DUVAL	1.724	2.203	1.000	1.278						
	12	FRANKLIN	5.849	8.391	1.000	1.435						
	13	GLADES	5.955	6.602	1.000	1.109						
	14	HAMILTON	0.733	0.763	1.000	1.041						
	15	HERNANDO	4.717	5.291	1.000	1.122						
	16	HILLSBOROUGH	4.248	4.551	1.000	1.071						
	17	HOLMES	1.398	1.526	1.000	1.092						
	18	INDIAN RIVER	13.923	18.263	1.000	1.312						
	19	JACKSON	1.084	1.132	1.000	1.045						
	20	LEE	6.354	7.149	1.000	1.125						
	21	LEON	1.027	1.106	1.000	1.077						
	22	MARION	2.825	2.931	1.000	1.037						
	23	MARTIN	5.838	6.544	1.000	1.121						
	24	MARTIN	13.175	17.280	1.000	1.312						
	25	MIAMI-DADE	6.714	7.947	1.000	1.184						
	26	MIAMI-DADE	9.648	12.396	1.000	1.285						
	27	MONROE	14.067	19.259	1.000	1.369						
	28	MONROE	22.019	30.061	1.000	1.365						
	29	OKALOOSA	2.781	3.748	1.000	1.347						
	30	OSCEOLA	3.698	3.861	1.000	1.044						
	31	OSCEOLA	4.785	5.231	1.000	1.093						
	32	PALM BEACH	7.462	8.816	1.000	1.181						
	33	PALM BEACH	10.706	13.605	1.000	1.271						
	34	PINELLAS	4.238	4.667	1.000	1.101						
	35	SAINT JOHNS	1.326	1.590	1.000	1.199						
	36	SANTA ROSA	2.008	2.538	1.000	1.264						
	37	SEMINOLE	3.251	3.369	1.000	1.036						
	38	TAYLOR	0.931	1.021	1.000	1.096						
	39	VOLUSIA	3.922	4.493	1.000	1.146						
	40	WAKULLA	2.362	3.057	1.000	1.294						

Form A-6: Logical Relationship to Risk - Number of Stories Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County / City	Loss Cost by Number of Stories		Ratios Relative to 1 Story		Loss Cost by Number of Stories			Ratios Relative to 5 Story		
			1 Story	2 Story	1 Story	2 Story	5 Story	10 Story	20 Story	5 Story	10 Story	20 Story
	1	BAY	3.576	4.936	1.000	1.380						
	2	BREVARD	4.630	5.191	1.000	1.121						
	3	BREVARD	4.482	4.951	1.000	1.105						
	4	BROWARD	7.226	8.546	1.000	1.183						
	5	BROWARD	12.139	15.469	1.000	1.274						
	6	CITRUS	3.376	3.554	1.000	1.052						
	7	CLAY	0.827	0.868	1.000	1.050						
	8	COLLIER	6.881	7.967	1.000	1.158						
	9	COLUMBIA	0.791	0.826	1.000	1.044						
	10	DIXIE	2.515	3.275	1.000	1.302						
	11	DUVAL	1.622	2.053	1.000	1.265						
	12	FRANKLIN	5.363	7.570	1.000	1.412						
	13	GLADES	5.881	6.508	1.000	1.107						
	14	HAMILTON	0.700	0.729	1.000	1.042						
	15	HERNANDO	4.664	5.196	1.000	1.114						
	16	HILLSBOROUGH	4.198	4.493	1.000	1.070						
	17	HOLMES	1.323	1.447	1.000	1.094						
	18	INDIAN RIVER	12.926	16.547	1.000	1.280						
	19	JACKSON	1.029	1.077	1.000	1.047						
Masonry Owners	20	LEE	6.274	7.038	1.000	1.122						
	21	LEON	0.976	1.052	1.000	1.079						
	22	MARION	2.792	2.895	1.000	1.037						
	23	MARTIN	5.700	6.383	1.000	1.120						
	24	MARTIN	12.466	16.021	1.000	1.285						
	25	MIAMI-DADE	6.552	7.727	1.000	1.179						
	26	MIAMI-DADE	9.341	11.840	1.000	1.268						
	27	MONROE	13.346	17.863	1.000	1.338						
	28	MONROE	20.238	27.027	1.000	1.335						
	29	OKALOOSA	2.615	3.507	1.000	1.341						
	30	OSCEOLA	3.655	3.814	1.000	1.044						
	31	OSCEOLA	4.725	5.155	1.000	1.091						
	32	PALM BEACH	7.279	8.564	1.000	1.177						
	33	PALM BEACH	10.332	12.957	1.000	1.254						
	34	PINELLAS	4.201	4.605	1.000	1.096						
	35	SAINT JOHNS	1.254	1.499	1.000	1.195						
	36	SANTA ROSA	1.892	2.391	1.000	1.264						
	37	SEMINOLE	3.212	3.329	1.000	1.036						
	38	TAYLOR	0.886	0.970	1.000	1.096						
	39	VOLUSIA	3.874	4.399	1.000	1.136						
	40	WAKULLA	2.219	2.847	1.000	1.283						

Form A-6: Logical Relationship to Risk - Number of Stories Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County / City	Loss Cost by Number of Stories		Ratios Relative to 1 Story		Loss Cost by Number of Stories			Ratios Relative to 5 Story		
			1 Story	2 Story	1 Story	2 Story	5 Story	10 Story	20 Story	5 Story	10 Story	20 Story
Frame Renters	1	BAY	1.211	2.137	1.000	1.765						
	2	BREVARD	0.584	0.913	1.000	1.562						
	3	BREVARD	0.538	0.803	1.000	1.491						
	4	BROWARD	1.038	1.830	1.000	1.764						
	5	BROWARD	3.159	5.733	1.000	1.815						
	6	CITRUS	0.355	0.442	1.000	1.247						
	7	CLAY	0.193	0.211	1.000	1.096						
	8	COLLIER	0.898	1.519	1.000	1.690						
	9	COLUMBIA	0.184	0.199	1.000	1.082						
	10	DIXIE	0.816	1.330	1.000	1.630						
	11	DUVAL	0.482	0.764	1.000	1.586						
	12	FRANKLIN	2.130	3.800	1.000	1.784						
	13	GLADES	0.662	0.989	1.000	1.493						
	14	HAMILTON	0.163	0.175	1.000	1.075						
	15	HERNANDO	0.588	0.899	1.000	1.529						
	16	HILLSBOROUGH	0.442	0.585	1.000	1.324						
	17	HOLMES	0.327	0.386	1.000	1.180						
	18	INDIAN RIVER	4.261	7.383	1.000	1.733						
	19	JACKSON	0.246	0.265	1.000	1.078						
	20	LEE	0.729	1.138	1.000	1.562						
	21	LEON	0.235	0.271	1.000	1.156						
	22	MARION	0.287	0.335	1.000	1.168						
	23	MARTIN	0.677	1.051	1.000	1.551						
	24	MARTIN	3.675	6.579	1.000	1.790						
	25	MIAMI-DADE	0.909	1.596	1.000	1.756						
	26	MIAMI-DADE	1.916	3.655	1.000	1.908						
	27	MONROE	4.110	7.744	1.000	1.884						
	28	MONROE	8.234	14.270	1.000	1.733						
	29	OKALOOSA	0.765	1.294	1.000	1.691						
	30	OSCEOLA	0.373	0.446	1.000	1.196						
	31	OSCEOLA	0.533	0.764	1.000	1.435						
	32	PALM BEACH	1.015	1.784	1.000	1.758						
	33	PALM BEACH	2.159	4.025	1.000	1.864						
	34	PINELLAS	0.484	0.701	1.000	1.448						
	35	SAINT JOHNS	0.335	0.476	1.000	1.422						
	36	SANTA ROSA	0.510	0.776	1.000	1.522						
	37	SEMINOLE	0.327	0.379	1.000	1.161						
	38	TAYLOR	0.217	0.260	1.000	1.195						
	39	VOLUSIA	0.517	0.837	1.000	1.617						
	40	WAKULLA	0.666	1.069	1.000	1.606						

Form A-6: Logical Relationship to Risk - Number of Stories Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County / City	Loss Cost by Number of Stories		Ratios Relative to 1 Story		Loss Cost by Number of Stories			Ratios Relative to 5 Story		
			1 Story	2 Story	1 Story	2 Story	5 Story	10 Story	20 Story	5 Story	10 Story	20 Story
Masonry Renters	1	BAY	1.113	1.953	1.000	1.754						
	2	BREVARD	0.572	0.890	1.000	1.556						
	3	BREVARD	0.527	0.786	1.000	1.492						
	4	BROWARD	1.017	1.777	1.000	1.748						
	5	BROWARD	2.898	5.116	1.000	1.765						
	6	CITRUS	0.345	0.434	1.000	1.259						
	7	CLAY	0.170	0.189	1.000	1.114						
	8	COLLIER	0.882	1.490	1.000	1.691						
	9	COLUMBIA	0.162	0.178	1.000	1.097						
	10	DIXIE	0.737	1.204	1.000	1.633						
	11	DUVAL	0.440	0.698	1.000	1.588						
	12	FRANKLIN	1.914	3.365	1.000	1.758						
	13	GLADES	0.649	0.980	1.000	1.511						
	14	HAMILTON	0.143	0.156	1.000	1.091						
	15	HERNANDO	0.573	0.873	1.000	1.524						
	16	HILLSBOROUGH	0.431	0.579	1.000	1.343						
	17	HOLMES	0.291	0.352	1.000	1.207						
	18	INDIAN RIVER	3.726	6.315	1.000	1.695						
	19	JACKSON	0.216	0.237	1.000	1.095						
	20	LEE	0.716	1.126	1.000	1.574						
	21	LEON	0.208	0.246	1.000	1.179						
	22	MARION	0.279	0.328	1.000	1.179						
	23	MARTIN	0.665	1.037	1.000	1.560						
	24	MARTIN	3.308	5.765	1.000	1.743						
	25	MIAMI-DADE	0.892	1.559	1.000	1.748						
	26	MIAMI-DADE	1.840	3.408	1.000	1.853						
	27	MONROE	3.772	6.859	1.000	1.819						
	28	MONROE	7.087	12.073	1.000	1.704						
	29	OKALOOSA	0.709	1.219	1.000	1.719						
	30	OSCEOLA	0.363	0.438	1.000	1.209						
	31	OSCEOLA	0.521	0.752	1.000	1.444						
	32	PALM BEACH	0.998	1.741	1.000	1.745						
	33	PALM BEACH	2.047	3.718	1.000	1.816						
	34	PINELLAS	0.473	0.690	1.000	1.458						
	35	SAINT JOHNS	0.303	0.440	1.000	1.451						
	36	SANTA ROSA	0.466	0.734	1.000	1.573						
	37	SEMINOLE	0.317	0.372	1.000	1.173						
	38	TAYLOR	0.193	0.236	1.000	1.222						
	39	VOLUSIA	0.506	0.811	1.000	1.602						
	40	WAKULLA	0.609	0.983	1.000	1.613						

Form A-6: Logical Relationship to Risk - Number of Stories Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Location	County / City	Loss Cost by Number of Stories		Ratios Relative to 1 Story		Loss Cost by Number of Stories			Ratios Relative to 5 Story		
			1 Story	2 Story	1 Story	2 Story	5 Story	10 Story	20 Story	5 Story	10 Story	20 Story
	1	BAY					2.259	5.534	14.376	1.000	2.450	6.365
	2	BREVARD					1.041	2.948	8.598	1.000	2.830	8.255
	3	BREVARD					0.901	2.608	7.822	1.000	2.894	8.678
	4	BROWARD					2.246	5.841	15.876	1.000	2.601	7.068
	5	BROWARD					5.492	11.691	27.147	1.000	2.129	4.943
	6	CITRUS					0.418	1.323	4.341	1.000	3.161	10.375
	7	CLAY					0.130	0.471	1.750	1.000	3.622	13.456
	8	COLLIER					1.932	5.260	14.760	1.000	2.723	7.642
	9	COLUMBIA					0.114	0.413	1.554	1.000	3.634	13.671
	10	DIXIE					1.237	2.854	7.182	1.000	2.307	5.805
	11	DUVAL					0.745	1.900	5.207	1.000	2.551	6.992
	12	FRANKLIN					3.436	7.213	16.460	1.000	2.099	4.790
	13	GLADES					1.231	3.586	10.698	1.000	2.914	8.693
	14	HAMILTON					0.097	0.348	1.300	1.000	3.587	13.407
	15	HERNANDO					0.964	2.581	7.346	1.000	2.678	7.623
	16	HILLSBOROUGH					0.679	2.158	6.921	1.000	3.179	10.195
	17	HOLMES					0.344	1.212	4.254	1.000	3.518	12.353
	18	INDIAN RIVER					6.170	11.868	25.061	1.000	1.924	4.062
	19	JACKSON					0.179	0.697	2.678	1.000	3.891	14.951
Commercial Residential	20	LEE					1.420	3.970	11.515	1.000	2.796	8.110
	21	LEON					0.199	0.664	2.289	1.000	3.330	11.487
	22	MARION					0.279	0.945	3.291	1.000	3.382	11.778
	23	MARTIN					1.310	3.797	11.263	1.000	2.898	8.596
	24	MARTIN					6.041	12.341	27.686	1.000	2.043	4.583
	25	MIAMI-DADE					2.090	5.695	15.877	1.000	2.726	7.598
	26	MIAMI-DADE					3.988	9.306	22.893	1.000	2.333	5.740
	27	MONROE					7.388	15.167	33.413	1.000	2.053	4.523
	28	MONROE					11.444	20.432	40.034	1.000	1.785	3.498
	29	OKALOOSA					1.560	4.259	11.959	1.000	2.729	7.664
	30	OSCEOLA					0.425	1.426	4.856	1.000	3.359	11.436
	31	OSCEOLA					0.832	2.373	7.123	1.000	2.851	8.557
	32	PALM BEACH					2.169	5.671	15.494	1.000	2.615	7.144
	33	PALM BEACH					4.322	9.956	24.399	1.000	2.304	5.646
	34	PINELLAS					0.841	2.564	7.913	1.000	3.048	9.407
	35	SAINT JOHNS					0.477	1.386	4.217	1.000	2.909	8.851
	36	SANTA ROSA					0.946	2.708	7.934	1.000	2.863	8.391
	37	SEMINOLE					0.331	1.133	3.945	1.000	3.429	11.935
	38	TAYLOR					0.206	0.674	2.295	1.000	3.266	11.124
	39	VOLUSIA					0.937	2.536	7.142	1.000	2.707	7.624
	40	WAKULLA					1.055	2.632	7.038	1.000	2.495	6.671

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
Frame Owners	1	BAY	4.190	3.537	2.885	2.329	1.615	1.296	1.000	0.844	0.689	0.556	0.386	0.309
	2	BREVARD	4.590	4.015	3.442	2.731	1.302	0.648	1.000	0.875	0.750	0.595	0.284	0.141
	3	BREVARD	4.429	3.858	3.290	2.589	1.187	0.561	1.000	0.871	0.743	0.585	0.268	0.127
	4	BROWARD	8.204	7.427	6.652	5.591	3.295	1.919	1.000	0.905	0.811	0.681	0.402	0.234
	5	BROWARD	14.045	13.096	12.150	10.813	7.858	5.781	1.000	0.932	0.865	0.770	0.560	0.412
	6	CITRUS	3.324	2.865	2.407	1.863	0.784	0.311	1.000	0.862	0.724	0.560	0.236	0.093
	7	CLAY	0.895	0.629	0.365	0.214	0.088	0.064	1.000	0.703	0.407	0.239	0.098	0.072
	8	COLLIER	6.141	5.387	4.634	3.713	1.854	0.966	1.000	0.877	0.755	0.605	0.302	0.157
	9	COLUMBIA	0.885	0.627	0.371	0.219	0.080	0.059	1.000	0.709	0.419	0.247	0.091	0.067
	10	DIXIE	2.882	2.357	1.834	1.427	0.954	0.767	1.000	0.818	0.637	0.495	0.331	0.266
	11	DUVAL	1.878	1.491	1.105	0.823	0.467	0.378	1.000	0.794	0.589	0.438	0.249	0.201
	12	FRANKLIN	6.099	5.342	4.588	3.931	3.054	2.546	1.000	0.876	0.752	0.645	0.501	0.417
	13	GLADES	5.349	4.649	3.952	3.124	1.463	0.670	1.000	0.869	0.739	0.584	0.273	0.125
	14	HAMILTON	0.786	0.557	0.330	0.193	0.071	0.053	1.000	0.709	0.420	0.246	0.090	0.067
	15	HERNANDO	4.396	3.829	3.265	2.584	1.221	0.602	1.000	0.871	0.743	0.588	0.278	0.137
	16	HILLSBOROUGH	3.943	3.412	2.883	2.227	0.927	0.377	1.000	0.865	0.731	0.565	0.235	0.096
	17	HOLMES	1.527	1.138	0.750	0.482	0.198	0.143	1.000	0.745	0.491	0.315	0.129	0.094
	18	INDIAN RIVER	11.844	10.985	10.129	9.050	6.795	5.307	1.000	0.928	0.855	0.764	0.574	0.448
	19	JACKSON	1.167	0.843	0.520	0.313	0.112	0.082	1.000	0.722	0.446	0.268	0.096	0.070
	20	LEE	5.439	4.724	4.011	3.208	1.599	0.772	1.000	0.869	0.737	0.590	0.294	0.142
	21	LEON	1.114	0.814	0.515	0.322	0.130	0.095	1.000	0.731	0.463	0.289	0.117	0.086
	22	MARION	1.621	1.225	0.831	0.582	0.227	0.100	1.000	0.756	0.513	0.359	0.140	0.062
	23	MARTIN	6.002	5.314	4.628	3.693	1.749	0.837	1.000	0.885	0.771	0.615	0.291	0.140
	24	MARTIN	13.648	12.772	11.898	10.606	7.745	5.853	1.000	0.936	0.872	0.777	0.567	0.429
	25	MIAMI-DADE	7.451	6.740	6.031	5.061	2.956	1.690	1.000	0.905	0.809	0.679	0.397	0.227
	26	MIAMI-DADE	10.846	10.008	9.171	8.007	5.456	3.739	1.000	0.923	0.846	0.738	0.503	0.345
	27	MONROE	13.480	12.664	11.849	10.575	7.745	5.731	1.000	0.939	0.879	0.784	0.575	0.425
	28	MONROE	21.049	20.082	19.116	17.607	14.219	11.501	1.000	0.954	0.908	0.836	0.676	0.546
	29	OKALOOSA	3.048	2.475	1.904	1.434	0.869	0.657	1.000	0.812	0.624	0.470	0.285	0.216
	30	OSCEOLA	2.095	1.602	1.112	0.787	0.311	0.137	1.000	0.765	0.531	0.376	0.149	0.065
	31	OSCEOLA	2.660	2.075	1.493	1.085	0.470	0.230	1.000	0.780	0.561	0.408	0.177	0.086
	32	PALM BEACH	7.732	6.945	6.160	5.049	2.678	1.453	1.000	0.898	0.797	0.653	0.346	0.188
	33	PALM BEACH	11.134	10.230	9.328	8.009	5.125	3.417	1.000	0.919	0.838	0.719	0.460	0.307
	34	PINELLAS	4.486	3.963	3.442	2.701	1.180	0.546	1.000	0.883	0.767	0.602	0.263	0.122
	35	SAINT JOHNS	1.424	1.087	0.752	0.514	0.272	0.207	1.000	0.763	0.528	0.361	0.191	0.145
	36	SANTA ROSA	2.216	1.748	1.281	0.910	0.483	0.352	1.000	0.789	0.578	0.411	0.218	0.159
	37	SEMINOLE	3.279	2.833	2.387	1.839	0.745	0.282	1.000	0.864	0.728	0.561	0.227	0.086
	38	TAYLOR	1.002	0.731	0.461	0.283	0.120	0.089	1.000	0.729	0.460	0.282	0.119	0.089
	39	VOLUSIA	2.997	2.520	2.043	1.592	0.760	0.386	1.000	0.841	0.682	0.531	0.254	0.129
	40	WAKULLA	2.541	2.035	1.531	1.145	0.712	0.557	1.000	0.801	0.602	0.450	0.280	0.219

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
Masonry Owners	1	BAY	3.863	3.212	2.562	2.039	1.394	1.094	1.000	0.831	0.663	0.528	0.361	0.283
	2	BREVARD	4.521	3.947	3.374	2.670	1.258	0.614	1.000	0.873	0.746	0.590	0.278	0.136
	3	BREVARD	4.369	3.798	3.230	2.536	1.151	0.535	1.000	0.869	0.739	0.580	0.264	0.122
	4	BROWARD	7.980	7.203	6.429	5.380	3.128	1.802	1.000	0.903	0.806	0.674	0.392	0.226
	5	BROWARD	13.289	12.342	11.397	10.074	7.170	5.159	1.000	0.929	0.858	0.758	0.540	0.388
	6	CITRUS	3.273	2.814	2.356	1.820	0.762	0.301	1.000	0.860	0.720	0.556	0.233	0.092
	7	CLAY	0.862	0.596	0.331	0.191	0.082	0.060	1.000	0.691	0.385	0.222	0.096	0.070
	8	COLLIER	6.018	5.263	4.511	3.603	1.778	0.910	1.000	0.875	0.750	0.599	0.295	0.151
	9	COLUMBIA	0.850	0.593	0.337	0.195	0.075	0.055	1.000	0.697	0.396	0.230	0.088	0.065
	10	DIXIE	2.684	2.160	1.638	1.255	0.832	0.656	1.000	0.805	0.610	0.468	0.310	0.245
	11	DUVAL	1.690	1.304	0.919	0.659	0.343	0.262	1.000	0.771	0.543	0.390	0.203	0.155
	12	FRANKLIN	5.539	4.785	4.032	3.411	2.609	2.127	1.000	0.864	0.728	0.616	0.471	0.384
	13	GLADES	5.240	4.541	3.843	3.031	1.407	0.640	1.000	0.867	0.734	0.578	0.269	0.122
	14	HAMILTON	0.755	0.526	0.299	0.172	0.066	0.049	1.000	0.697	0.396	0.228	0.088	0.065
	15	HERNANDO	4.327	3.761	3.197	2.524	1.183	0.576	1.000	0.869	0.739	0.583	0.273	0.133
	16	HILLSBOROUGH	3.878	3.347	2.818	2.172	0.896	0.361	1.000	0.863	0.727	0.560	0.231	0.093
	17	HOLMES	1.456	1.068	0.681	0.431	0.182	0.133	1.000	0.733	0.467	0.296	0.125	0.091
	18	INDIAN RIVER	11.084	10.226	9.370	8.306	6.088	4.625	1.000	0.923	0.845	0.749	0.549	0.417
	19	JACKSON	1.117	0.793	0.471	0.278	0.104	0.076	1.000	0.710	0.421	0.249	0.093	0.068
	20	LEE	5.311	4.597	3.884	3.096	1.530	0.739	1.000	0.866	0.731	0.583	0.288	0.139
	21	LEON	1.066	0.766	0.467	0.288	0.121	0.088	1.000	0.719	0.439	0.270	0.113	0.083
	22	MARION	1.585	1.189	0.795	0.557	0.220	0.098	1.000	0.750	0.502	0.351	0.139	0.062
	23	MARTIN	5.758	5.071	4.387	3.475	1.614	0.784	1.000	0.881	0.762	0.603	0.280	0.136
	24	MARTIN	12.669	11.794	10.922	9.653	6.886	5.107	1.000	0.931	0.862	0.762	0.544	0.403
	25	MIAMI-DADE	7.257	6.547	5.838	4.878	2.814	1.594	1.000	0.902	0.804	0.672	0.388	0.220
	26	MIAMI-DADE	10.416	9.579	8.743	7.591	5.085	3.424	1.000	0.920	0.839	0.729	0.488	0.329
	27	MONROE	12.698	11.883	11.069	9.809	7.039	5.110	1.000	0.936	0.872	0.772	0.554	0.402
	28	MONROE	19.295	18.329	17.365	15.871	12.549	9.930	1.000	0.950	0.900	0.823	0.650	0.515
	29	OKALOOSA	2.870	2.298	1.728	1.286	0.779	0.582	1.000	0.801	0.602	0.448	0.272	0.203
	30	OSCEOLA	2.046	1.554	1.063	0.752	0.302	0.134	1.000	0.759	0.520	0.368	0.148	0.065
	31	OSCEOLA	2.594	2.010	1.428	1.038	0.455	0.223	1.000	0.775	0.551	0.400	0.175	0.086
	32	PALM BEACH	7.405	6.619	5.836	4.750	2.473	1.348	1.000	0.894	0.788	0.641	0.334	0.182
	33	PALM BEACH	10.532	9.629	8.728	7.436	4.654	3.065	1.000	0.914	0.829	0.706	0.442	0.291
	34	PINELLAS	4.383	3.861	3.340	2.608	1.118	0.511	1.000	0.881	0.762	0.595	0.255	0.117
	35	SAINT JOHNS	1.354	1.018	0.683	0.460	0.246	0.186	1.000	0.752	0.504	0.340	0.182	0.137
	36	SANTA ROSA	2.101	1.634	1.168	0.820	0.439	0.318	1.000	0.778	0.556	0.390	0.209	0.152
	37	SEMINOLE	3.221	2.775	2.330	1.789	0.719	0.273	1.000	0.861	0.723	0.555	0.223	0.085
	38	TAYLOR	0.960	0.689	0.419	0.253	0.110	0.082	1.000	0.718	0.437	0.264	0.115	0.085
	39	VOLUSIA	2.950	2.472	1.996	1.553	0.737	0.369	1.000	0.838	0.677	0.526	0.250	0.125
	40	WAKULLA	2.386	1.881	1.377	1.015	0.629	0.485	1.000	0.788	0.577	0.426	0.264	0.203

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
Mobile Homes	1	BAY	16.793	15.392	15.392	14.488	12.396	10.398	1.000	0.917	0.917	0.863	0.738	0.619
	2	BREVARD	13.278	12.093	12.093	11.191	9.023	7.106	1.000	0.911	0.911	0.843	0.680	0.535
	3	BREVARD	12.433	11.256	11.256	10.364	8.225	6.363	1.000	0.905	0.905	0.834	0.662	0.512
	4	BROWARD	25.412	23.803	23.803	22.494	19.229	15.979	1.000	0.937	0.937	0.885	0.757	0.629
	5	BROWARD	45.049	43.085	43.085	41.455	37.334	32.950	1.000	0.956	0.956	0.920	0.829	0.731
	6	CITRUS	7.795	6.842	6.842	6.135	4.469	3.130	1.000	0.878	0.878	0.787	0.573	0.402
	7	CLAY	2.717	2.146	2.146	1.832	1.164	0.705	1.000	0.790	0.790	0.674	0.428	0.259
	8	COLLIER	23.503	21.931	21.931	20.651	17.454	14.286	1.000	0.933	0.933	0.879	0.743	0.608
	9	COLUMBIA	2.556	2.004	2.004	1.702	1.058	0.622	1.000	0.784	0.784	0.666	0.414	0.243
	10	DIXIE	11.053	9.924	9.924	9.224	7.634	6.209	1.000	0.898	0.898	0.835	0.691	0.562
	11	DUVAL	6.873	6.056	6.056	5.567	4.477	3.561	1.000	0.881	0.881	0.810	0.651	0.518
	12	FRANKLIN	22.352	20.731	20.731	19.695	17.309	14.974	1.000	0.927	0.927	0.881	0.774	0.670
	13	GLADES	16.872	15.428	15.428	14.311	11.587	9.067	1.000	0.914	0.914	0.848	0.687	0.537
	14	HAMILTON	2.250	1.759	1.759	1.491	0.919	0.535	1.000	0.782	0.782	0.663	0.408	0.238
	15	HERNANDO	13.231	12.054	12.054	11.150	8.962	7.008	1.000	0.911	0.911	0.843	0.677	0.530
	16	HILLSBOROUGH	10.744	9.652	9.652	8.818	6.803	5.029	1.000	0.898	0.898	0.821	0.633	0.468
	17	HOLMES	5.109	4.274	4.274	3.780	2.677	1.805	1.000	0.836	0.836	0.740	0.524	0.353
	18	INDIAN RIVER	44.321	42.532	42.532	41.088	37.463	33.562	1.000	0.960	0.960	0.927	0.845	0.757
	19	JACKSON	3.531	2.835	2.835	2.440	1.576	0.947	1.000	0.803	0.803	0.691	0.446	0.268
	20	LEE	19.542	18.057	18.057	16.857	13.881	11.043	1.000	0.924	0.924	0.863	0.710	0.565
	21	LEON	3.505	2.861	2.861	2.496	1.700	1.111	1.000	0.816	0.816	0.712	0.485	0.317
	22	MARION	6.004	5.178	5.178	4.574	3.167	2.097	1.000	0.862	0.862	0.762	0.527	0.349
	23	MARTIN	17.543	16.109	16.109	14.969	12.181	9.628	1.000	0.918	0.918	0.853	0.694	0.549
	24	MARTIN	46.080	44.258	44.258	42.738	38.879	34.685	1.000	0.960	0.960	0.927	0.844	0.753
	25	MIAMI-DADE	23.249	21.778	21.778	20.580	17.596	14.608	1.000	0.937	0.937	0.885	0.757	0.628
	26	MIAMI-DADE	35.836	34.099	34.099	32.671	29.089	25.348	1.000	0.952	0.952	0.912	0.812	0.707
	27	MONROE	53.076	51.355	51.355	49.892	46.124	41.763	1.000	0.968	0.968	0.940	0.869	0.787
	28	MONROE	75.209	73.190	73.190	71.474	67.035	61.774	1.000	0.973	0.973	0.950	0.891	0.821
	29	OKALOOSA	13.223	11.991	11.991	11.202	9.388	7.689	1.000	0.907	0.907	0.847	0.710	0.581
	30	OSCEOLA	8.262	7.234	7.234	6.468	4.654	3.186	1.000	0.875	0.875	0.783	0.563	0.386
	31	OSCEOLA	12.499	11.277	11.277	10.346	8.106	6.164	1.000	0.902	0.902	0.828	0.649	0.493
	32	PALM BEACH	25.089	23.449	23.449	22.119	18.810	15.557	1.000	0.935	0.935	0.882	0.750	0.620
	33	PALM BEACH	38.748	36.866	36.866	35.309	31.376	27.244	1.000	0.951	0.951	0.911	0.810	0.703
	34	PINELLAS	12.119	11.031	11.031	10.169	8.069	6.194	1.000	0.910	0.910	0.839	0.666	0.511
	35	SAINT JOHNS	5.268	4.542	4.542	4.113	3.162	2.389	1.000	0.862	0.862	0.781	0.600	0.453
	36	SANTA ROSA	9.466	8.459	8.459	7.826	6.382	5.062	1.000	0.894	0.894	0.827	0.674	0.535
	37	SEMINOLE	6.956	6.028	6.028	5.342	3.730	2.468	1.000	0.867	0.867	0.768	0.536	0.355
	38	TAYLOR	3.292	2.707	2.707	2.374	1.648	1.110	1.000	0.822	0.822	0.721	0.501	0.337
	39	VOLUSIA	11.403	10.420	10.420	9.676	7.894	6.318	1.000	0.914	0.914	0.849	0.692	0.554
	40	WAKULLA	9.882	8.792	8.792	8.122	6.601	5.274	1.000	0.890	0.890	0.822	0.668	0.534

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
Frame Renters	1	BAY	1.622	0.967	1.037	0.967	0.845	0.761	1.000	0.597	0.639	0.597	0.521	0.469
	2	BREVARD	0.923	0.448	0.519	0.448	0.310	0.264	1.000	0.485	0.563	0.485	0.335	0.286
	3	BREVARD	0.860	0.394	0.463	0.394	0.262	0.222	1.000	0.458	0.539	0.458	0.305	0.259
	4	BROWARD	2.168	1.373	1.532	1.373	1.034	0.853	1.000	0.633	0.707	0.633	0.477	0.393
	5	BROWARD	5.181	4.099	4.346	4.099	3.540	3.110	1.000	0.791	0.839	0.791	0.683	0.600
	6	CITRUS	0.645	0.254	0.321	0.254	0.130	0.102	1.000	0.395	0.498	0.395	0.201	0.158
	7	CLAY	0.239	0.053	0.061	0.053	0.043	0.041	1.000	0.222	0.255	0.222	0.179	0.171
	8	COLLIER	1.384	0.697	0.816	0.697	0.461	0.380	1.000	0.504	0.590	0.504	0.333	0.275
	9	COLUMBIA	0.230	0.049	0.057	0.049	0.040	0.038	1.000	0.214	0.246	0.214	0.174	0.167
	10	DIXIE	1.070	0.580	0.624	0.580	0.505	0.457	1.000	0.541	0.583	0.541	0.471	0.426
	11	DUVAL	0.534	0.271	0.283	0.271	0.249	0.235	1.000	0.507	0.530	0.507	0.466	0.439
	12	FRANKLIN	2.682	1.890	1.990	1.890	1.697	1.533	1.000	0.705	0.742	0.705	0.633	0.572
	13	GLADES	1.020	0.455	0.536	0.455	0.297	0.247	1.000	0.446	0.526	0.446	0.291	0.242
	14	HAMILTON	0.210	0.044	0.051	0.044	0.036	0.034	1.000	0.211	0.244	0.211	0.170	0.163
	15	HERNANDO	0.970	0.460	0.552	0.460	0.284	0.232	1.000	0.475	0.569	0.475	0.293	0.239
	16	HILLSBOROUGH	0.671	0.264	0.315	0.264	0.165	0.138	1.000	0.393	0.470	0.393	0.246	0.206
	17	HOLMES	0.430	0.117	0.135	0.117	0.093	0.087	1.000	0.272	0.315	0.272	0.216	0.203
	18	INDIAN RIVER	4.577	3.662	3.862	3.662	3.218	2.885	1.000	0.800	0.844	0.800	0.703	0.630
	19	JACKSON	0.316	0.070	0.081	0.070	0.055	0.052	1.000	0.220	0.258	0.220	0.173	0.166
	20	LEE	1.120	0.525	0.616	0.525	0.345	0.284	1.000	0.469	0.550	0.469	0.308	0.254
	21	LEON	0.307	0.078	0.089	0.078	0.062	0.059	1.000	0.253	0.292	0.253	0.203	0.191
	22	MARION	0.257	0.060	0.067	0.060	0.049	0.046	1.000	0.235	0.262	0.235	0.190	0.181
	23	MARTIN	1.412	0.697	0.850	0.697	0.397	0.309	1.000	0.494	0.602	0.494	0.281	0.219
	24	MARTIN	5.372	4.272	4.551	4.272	3.653	3.236	1.000	0.795	0.847	0.795	0.680	0.602
	25	MIAMI-DADE	1.938	1.210	1.355	1.210	0.901	0.737	1.000	0.624	0.699	0.624	0.465	0.380
	26	MIAMI-DADE	3.542	2.621	2.822	2.621	2.178	1.871	1.000	0.740	0.797	0.740	0.615	0.528
	27	MONROE	5.645	4.405	4.780	4.405	3.562	3.022	1.000	0.780	0.847	0.780	0.631	0.535
	28	MONROE	10.136	8.584	9.073	8.584	7.451	6.552	1.000	0.847	0.895	0.847	0.735	0.646
	29	OKALOOSA	1.056	0.499	0.552	0.499	0.416	0.372	1.000	0.473	0.523	0.473	0.394	0.352
	30	OSCEOLA	0.335	0.082	0.091	0.082	0.066	0.062	1.000	0.243	0.271	0.243	0.196	0.186
	31	OSCEOLA	0.449	0.131	0.145	0.131	0.108	0.101	1.000	0.293	0.322	0.293	0.241	0.224
	32	PALM BEACH	1.990	1.128	1.320	1.128	0.738	0.597	1.000	0.567	0.663	0.567	0.371	0.300
	33	PALM BEACH	3.579	2.511	2.766	2.511	1.970	1.692	1.000	0.702	0.773	0.702	0.551	0.473
	34	PINELLAS	1.013	0.467	0.580	0.467	0.252	0.196	1.000	0.461	0.573	0.461	0.248	0.193
	35	SAINT JOHNS	0.451	0.162	0.182	0.162	0.133	0.122	1.000	0.360	0.403	0.360	0.296	0.270
	36	SANTA ROSA	0.715	0.275	0.313	0.275	0.221	0.199	1.000	0.385	0.437	0.385	0.309	0.279
	37	SEMINOLE	0.654	0.254	0.328	0.254	0.117	0.088	1.000	0.388	0.501	0.388	0.178	0.135
	38	TAYLOR	0.295	0.074	0.086	0.074	0.059	0.055	1.000	0.250	0.290	0.250	0.198	0.186
	39	VOLUSIA	0.510	0.223	0.241	0.223	0.190	0.173	1.000	0.438	0.472	0.438	0.373	0.340
	40	WAKULLA	0.889	0.423	0.463	0.423	0.361	0.325	1.000	0.476	0.520	0.476	0.405	0.366

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
Masonry Renters	1	BAY	1.425	0.806	0.866	0.806	0.697	0.621	1.000	0.565	0.608	0.565	0.489	0.436
	2	BREVARD	0.895	0.428	0.499	0.428	0.292	0.248	1.000	0.478	0.557	0.478	0.326	0.277
	3	BREVARD	0.836	0.379	0.447	0.379	0.249	0.210	1.000	0.453	0.534	0.453	0.298	0.252
	4	BROWARD	2.088	1.303	1.461	1.303	0.971	0.797	1.000	0.624	0.700	0.624	0.465	0.382
	5	BROWARD	4.703	3.637	3.879	3.637	3.094	2.689	1.000	0.773	0.825	0.773	0.658	0.572
	6	CITRUS	0.631	0.247	0.313	0.247	0.125	0.098	1.000	0.392	0.496	0.392	0.198	0.156
	7	CLAY	0.219	0.046	0.053	0.046	0.038	0.037	1.000	0.212	0.240	0.212	0.175	0.167
	8	COLLIER	1.337	0.662	0.779	0.662	0.430	0.352	1.000	0.495	0.583	0.495	0.321	0.263
	9	COLUMBIA	0.212	0.043	0.049	0.043	0.036	0.034	1.000	0.203	0.230	0.203	0.169	0.162
	10	DIXIE	0.952	0.488	0.527	0.488	0.422	0.379	1.000	0.513	0.553	0.513	0.444	0.398
	11	DUVAL	0.410	0.163	0.172	0.163	0.146	0.137	1.000	0.397	0.421	0.397	0.356	0.335
	12	FRANKLIN	2.305	1.554	1.642	1.554	1.380	1.232	1.000	0.674	0.712	0.674	0.599	0.534
	13	GLADES	0.992	0.437	0.517	0.437	0.282	0.235	1.000	0.441	0.521	0.441	0.285	0.237
	14	HAMILTON	0.193	0.039	0.044	0.039	0.032	0.031	1.000	0.200	0.228	0.200	0.165	0.158
	15	HERNANDO	0.948	0.446	0.537	0.446	0.271	0.221	1.000	0.470	0.566	0.470	0.286	0.233
	16	HILLSBOROUGH	0.653	0.254	0.305	0.254	0.157	0.131	1.000	0.389	0.467	0.389	0.240	0.201
	17	HOLMES	0.396	0.104	0.118	0.104	0.084	0.078	1.000	0.262	0.299	0.262	0.212	0.198
	18	INDIAN RIVER	4.078	3.177	3.372	3.177	2.745	2.433	1.000	0.779	0.827	0.779	0.673	0.597
	19	JACKSON	0.291	0.061	0.070	0.061	0.049	0.047	1.000	0.209	0.241	0.209	0.169	0.161
	20	LEE	1.091	0.507	0.596	0.507	0.330	0.274	1.000	0.464	0.547	0.464	0.303	0.251
	21	LEON	0.283	0.069	0.078	0.069	0.056	0.053	1.000	0.243	0.276	0.243	0.199	0.187
	22	MARION	0.249	0.058	0.065	0.058	0.047	0.045	1.000	0.233	0.260	0.233	0.188	0.179
	23	MARTIN	1.357	0.655	0.804	0.655	0.369	0.291	1.000	0.482	0.592	0.482	0.272	0.214
	24	MARTIN	4.674	3.594	3.864	3.594	3.006	2.630	1.000	0.769	0.827	0.769	0.643	0.563
	25	MIAMI-DADE	1.873	1.154	1.298	1.154	0.851	0.693	1.000	0.616	0.693	0.616	0.454	0.370
	26	MIAMI-DADE	3.317	2.409	2.606	2.409	1.976	1.683	1.000	0.726	0.786	0.726	0.596	0.507
	27	MONROE	5.156	3.931	4.301	3.931	3.111	2.603	1.000	0.762	0.834	0.762	0.603	0.505
	28	MONROE	8.856	7.325	7.803	7.325	6.228	5.387	1.000	0.827	0.881	0.827	0.703	0.608
	29	OKALOOSA	0.968	0.442	0.487	0.442	0.368	0.327	1.000	0.456	0.503	0.456	0.380	0.338
	30	OSCEOLA	0.324	0.078	0.087	0.078	0.063	0.060	1.000	0.242	0.269	0.242	0.195	0.185
	31	OSCEOLA	0.436	0.127	0.140	0.127	0.104	0.097	1.000	0.292	0.321	0.292	0.240	0.223
	32	PALM BEACH	1.894	1.045	1.233	1.045	0.673	0.546	1.000	0.552	0.651	0.552	0.355	0.288
	33	PALM BEACH	3.258	2.208	2.456	2.208	1.692	1.440	1.000	0.678	0.754	0.678	0.519	0.442
	34	PINELLAS	0.979	0.441	0.552	0.441	0.231	0.179	1.000	0.451	0.564	0.451	0.236	0.182
	35	SAINT JOHNS	0.414	0.143	0.159	0.143	0.118	0.107	1.000	0.344	0.383	0.344	0.285	0.259
	36	SANTA ROSA	0.662	0.247	0.279	0.247	0.200	0.180	1.000	0.373	0.421	0.373	0.302	0.271
	37	SEMINOLE	0.640	0.246	0.319	0.246	0.112	0.086	1.000	0.384	0.498	0.384	0.175	0.134
	38	TAYLOR	0.273	0.065	0.074	0.065	0.053	0.049	1.000	0.238	0.273	0.238	0.193	0.180
	39	VOLUSIA	0.497	0.217	0.234	0.217	0.184	0.168	1.000	0.437	0.471	0.437	0.371	0.338
	40	WAKULLA	0.804	0.364	0.397	0.364	0.309	0.277	1.000	0.453	0.494	0.453	0.385	0.345

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
Frame Condo Unit	1	BAY	1.942	1.304	1.304	1.094	0.937	0.828	1.000	0.672	0.672	0.563	0.483	0.426
	2	BREVARD	1.324	0.784	0.784	0.545	0.385	0.292	1.000	0.593	0.593	0.412	0.291	0.221
	3	BREVARD	1.248	0.717	0.717	0.483	0.331	0.246	1.000	0.574	0.574	0.387	0.265	0.197
	4	BROWARD	2.860	2.045	2.045	1.642	1.239	0.958	1.000	0.715	0.715	0.574	0.433	0.335
	5	BROWARD	6.298	5.255	5.255	4.715	4.049	3.461	1.000	0.834	0.834	0.749	0.643	0.550
	6	CITRUS	0.936	0.510	0.510	0.321	0.182	0.114	1.000	0.545	0.545	0.343	0.195	0.122
	7	CLAY	0.312	0.114	0.114	0.059	0.045	0.042	1.000	0.364	0.364	0.188	0.144	0.134
	8	COLLIER	1.910	1.178	1.178	0.848	0.580	0.422	1.000	0.617	0.617	0.444	0.304	0.221
	9	COLUMBIA	0.303	0.109	0.109	0.054	0.042	0.039	1.000	0.358	0.358	0.179	0.138	0.129
	10	DIXIE	1.293	0.806	0.806	0.654	0.557	0.495	1.000	0.624	0.624	0.506	0.431	0.383
	11	DUVAL	0.687	0.390	0.390	0.301	0.271	0.250	1.000	0.568	0.568	0.439	0.394	0.364
	12	FRANKLIN	3.136	2.384	2.384	2.124	1.883	1.678	1.000	0.760	0.760	0.677	0.601	0.535
	13	GLADES	1.488	0.846	0.846	0.560	0.378	0.275	1.000	0.569	0.569	0.376	0.254	0.185
	14	HAMILTON	0.274	0.098	0.098	0.049	0.037	0.035	1.000	0.357	0.357	0.179	0.136	0.127
	15	HERNANDO	1.349	0.808	0.808	0.565	0.367	0.258	1.000	0.599	0.599	0.419	0.272	0.191
	16	HILLSBOROUGH	1.021	0.534	0.534	0.325	0.214	0.154	1.000	0.523	0.523	0.318	0.210	0.151
	17	HOLMES	0.553	0.227	0.227	0.132	0.099	0.090	1.000	0.410	0.410	0.238	0.179	0.163
	18	INDIAN RIVER	5.505	4.621	4.621	4.190	3.671	3.202	1.000	0.839	0.839	0.761	0.667	0.582
	19	JACKSON	0.411	0.151	0.151	0.077	0.057	0.053	1.000	0.367	0.367	0.188	0.139	0.130
	20	LEE	1.591	0.940	0.940	0.645	0.437	0.316	1.000	0.591	0.591	0.406	0.274	0.198
	21	LEON	0.397	0.156	0.156	0.087	0.066	0.061	1.000	0.393	0.393	0.219	0.166	0.153
	22	MARION	0.399	0.156	0.156	0.068	0.054	0.048	1.000	0.391	0.391	0.171	0.134	0.120
	23	MARTIN	1.926	1.202	1.202	0.862	0.525	0.343	1.000	0.624	0.624	0.447	0.273	0.178
	24	MARTIN	6.439	5.427	5.427	4.895	4.184	3.585	1.000	0.843	0.843	0.760	0.650	0.557
	25	MIAMI-DADE	2.568	1.821	1.821	1.452	1.083	0.829	1.000	0.709	0.709	0.565	0.422	0.323
	26	MIAMI-DADE	4.425	3.520	3.520	3.060	2.526	2.089	1.000	0.795	0.795	0.691	0.571	0.472
	27	MONROE	6.680	5.672	5.672	5.073	4.133	3.384	1.000	0.849	0.849	0.759	0.619	0.507
	28	MONROE	11.694	10.473	10.473	9.733	8.466	7.301	1.000	0.896	0.896	0.832	0.724	0.624
	29	OKALOOSA	1.293	0.744	0.744	0.570	0.459	0.401	1.000	0.575	0.575	0.441	0.355	0.310
	30	OSCEOLA	0.518	0.206	0.206	0.093	0.073	0.064	1.000	0.398	0.398	0.178	0.140	0.124
	31	OSCEOLA	0.680	0.294	0.294	0.151	0.120	0.105	1.000	0.432	0.432	0.222	0.177	0.155
	32	PALM BEACH	2.644	1.788	1.788	1.370	0.929	0.664	1.000	0.676	0.676	0.518	0.351	0.251
	33	PALM BEACH	4.487	3.466	3.466	2.946	2.325	1.878	1.000	0.772	0.772	0.657	0.518	0.419
	34	PINELLAS	1.400	0.837	0.837	0.580	0.342	0.218	1.000	0.598	0.598	0.415	0.244	0.155
	35	SAINT JOHNS	0.563	0.269	0.269	0.183	0.145	0.129	1.000	0.477	0.477	0.325	0.258	0.229
	36	SANTA ROSA	0.890	0.450	0.450	0.315	0.242	0.212	1.000	0.506	0.506	0.354	0.272	0.238
	37	SEMINOLE	0.941	0.513	0.513	0.323	0.171	0.099	1.000	0.545	0.545	0.343	0.181	0.105
	38	TAYLOR	0.376	0.147	0.147	0.083	0.062	0.057	1.000	0.390	0.390	0.221	0.166	0.151
	39	VOLUSIA	0.774	0.411	0.411	0.261	0.217	0.187	1.000	0.532	0.532	0.338	0.280	0.242
	40	WAKULLA	1.087	0.622	0.622	0.480	0.398	0.351	1.000	0.572	0.572	0.441	0.366	0.323

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			$0	$500	1%	2%	5%	10%	$0	$500	1%	2%	5%	10%
Masonry Condo Unit	1	BAY	1.723	1.112	1.112	0.914	0.775	0.675	1.000	0.646	0.646	0.531	0.450	0.392
	2	BREVARD	1.290	0.758	0.758	0.522	0.364	0.274	1.000	0.587	0.587	0.405	0.282	0.213
	3	BREVARD	1.219	0.695	0.695	0.465	0.315	0.233	1.000	0.570	0.570	0.381	0.258	0.191
	4	BROWARD	2.762	1.956	1.956	1.561	1.166	0.894	1.000	0.708	0.708	0.565	0.422	0.324
	5	BROWARD	5.768	4.737	4.737	4.208	3.558	2.995	1.000	0.821	0.821	0.729	0.617	0.519
	6	CITRUS	0.918	0.498	0.498	0.312	0.176	0.109	1.000	0.542	0.542	0.340	0.191	0.119
	7	CLAY	0.290	0.103	0.103	0.051	0.040	0.037	1.000	0.355	0.355	0.176	0.138	0.129
	8	COLLIER	1.853	1.132	1.132	0.807	0.544	0.390	1.000	0.611	0.611	0.435	0.294	0.211
	9	COLUMBIA	0.282	0.098	0.098	0.047	0.037	0.035	1.000	0.348	0.348	0.167	0.132	0.124
	10	DIXIE	1.160	0.695	0.695	0.552	0.467	0.410	1.000	0.599	0.599	0.476	0.403	0.354
	11	DUVAL	0.550	0.269	0.269	0.184	0.159	0.145	1.000	0.488	0.488	0.335	0.290	0.264
	12	FRANKLIN	2.721	2.000	2.000	1.755	1.536	1.349	1.000	0.735	0.735	0.645	0.565	0.496
	13	GLADES	1.450	0.819	0.819	0.538	0.360	0.261	1.000	0.565	0.565	0.371	0.248	0.180
	14	HAMILTON	0.255	0.088	0.088	0.042	0.033	0.031	1.000	0.346	0.346	0.166	0.130	0.122
	15	HERNANDO	1.321	0.786	0.786	0.548	0.351	0.245	1.000	0.595	0.595	0.415	0.266	0.185
	16	HILLSBOROUGH	0.997	0.518	0.518	0.313	0.204	0.146	1.000	0.520	0.520	0.314	0.205	0.146
	17	HOLMES	0.515	0.205	0.205	0.116	0.089	0.081	1.000	0.399	0.399	0.226	0.173	0.158
	18	INDIAN RIVER	4.954	4.081	4.081	3.656	3.149	2.700	1.000	0.824	0.824	0.738	0.636	0.545
	19	JACKSON	0.383	0.136	0.136	0.067	0.051	0.048	1.000	0.356	0.356	0.176	0.134	0.125
	20	LEE	1.550	0.910	0.910	0.622	0.418	0.303	1.000	0.587	0.587	0.401	0.270	0.195
	21	LEON	0.370	0.141	0.141	0.077	0.059	0.055	1.000	0.382	0.382	0.207	0.161	0.147
	22	MARION	0.388	0.151	0.151	0.066	0.052	0.046	1.000	0.390	0.390	0.170	0.133	0.118
	23	MARTIN	1.850	1.138	1.138	0.810	0.487	0.321	1.000	0.615	0.615	0.438	0.263	0.174
	24	MARTIN	5.678	4.679	4.679	4.164	3.482	2.928	1.000	0.824	0.824	0.733	0.613	0.516
	25	MIAMI-DADE	2.487	1.749	1.749	1.386	1.025	0.778	1.000	0.703	0.703	0.557	0.412	0.313
	26	MIAMI-DADE	4.168	3.273	3.273	2.821	2.299	1.878	1.000	0.785	0.785	0.677	0.551	0.450
	27	MONROE	6.138	5.139	5.139	4.552	3.634	2.921	1.000	0.837	0.837	0.742	0.592	0.476
	28	MONROE	10.302	9.092	9.092	8.367	7.134	6.028	1.000	0.883	0.883	0.812	0.692	0.585
	29	OKALOOSA	1.192	0.666	0.666	0.503	0.406	0.351	1.000	0.559	0.559	0.422	0.340	0.295
	30	OSCEOLA	0.503	0.200	0.200	0.089	0.070	0.062	1.000	0.397	0.397	0.178	0.138	0.123
	31	OSCEOLA	0.661	0.285	0.285	0.146	0.116	0.102	1.000	0.431	0.431	0.221	0.175	0.154
	32	PALM BEACH	2.519	1.677	1.677	1.273	0.849	0.605	1.000	0.665	0.665	0.505	0.337	0.240
	33	PALM BEACH	4.122	3.116	3.116	2.612	2.015	1.602	1.000	0.756	0.756	0.634	0.489	0.389
	34	PINELLAS	1.357	0.801	0.801	0.550	0.317	0.198	1.000	0.590	0.590	0.405	0.233	0.146
	35	SAINT JOHNS	0.522	0.241	0.241	0.161	0.129	0.114	1.000	0.462	0.462	0.308	0.246	0.218
	36	SANTA ROSA	0.828	0.408	0.408	0.281	0.218	0.191	1.000	0.493	0.493	0.340	0.264	0.230
	37	SEMINOLE	0.922	0.500	0.500	0.313	0.164	0.095	1.000	0.542	0.542	0.339	0.178	0.103
	38	TAYLOR	0.350	0.132	0.132	0.073	0.056	0.051	1.000	0.378	0.378	0.207	0.159	0.145
	39	VOLUSIA	0.757	0.400	0.400	0.253	0.210	0.180	1.000	0.529	0.529	0.335	0.277	0.238
	40	WAKULLA	0.991	0.546	0.546	0.413	0.341	0.298	1.000	0.551	0.551	0.416	0.344	0.301

Form A-6: Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost at different Deductibles						Ratios relative $0					
			0%	2%	3%	5%	10%		0%	2%	3%	5%	10%	
Commercial Residential	1	BAY	14.376	12.227	11.425	10.044	7.376		1.000	0.851	0.795	0.699	0.513	
	2	BREVARD	8.598	6.915	6.334	5.376	3.639		1.000	0.804	0.737	0.625	0.423	
	3	BREVARD	7.822	6.182	5.619	4.701	3.054		1.000	0.790	0.718	0.601	0.390	
	4	BROWARD	15.876	13.425	12.492	10.872	7.703		1.000	0.846	0.787	0.685	0.485	
	5	BROWARD	27.147	24.063	22.847	20.697	16.314		1.000	0.886	0.842	0.762	0.601	
	6	CITRUS	4.341	3.123	2.745	2.153	1.208		1.000	0.719	0.632	0.496	0.278	
	7	CLAY	1.750	1.102	0.938	0.692	0.340		1.000	0.630	0.536	0.396	0.194	
	8	COLLIER	14.760	12.347	11.419	9.811	6.699		1.000	0.836	0.774	0.665	0.454	
	9	COLUMBIA	1.554	0.946	0.793	0.577	0.274		1.000	0.609	0.510	0.371	0.176	
	10	DIXIE	7.182	5.768	5.287	4.500	3.110		1.000	0.803	0.736	0.627	0.433	
	11	DUVAL	5.207	4.145	3.802	3.237	2.253		1.000	0.796	0.730	0.622	0.433	
	12	FRANKLIN	16.460	14.199	13.360	11.909	9.091		1.000	0.863	0.812	0.723	0.552	
	13	GLADES	10.698	8.597	7.839	6.559	4.195		1.000	0.804	0.733	0.613	0.392	
	14	HAMILTON	1.300	0.773	0.643	0.467	0.224		1.000	0.595	0.494	0.359	0.172	
	15	HERNANDO	7.346	5.777	5.245	4.373	2.810		1.000	0.786	0.714	0.595	0.382	
	16	HILLSBOROUGH	6.921	5.348	4.800	3.891	2.301		1.000	0.773	0.693	0.562	0.332	
	17	HOLMES	4.254	3.091	2.717	2.122	1.163		1.000	0.726	0.639	0.499	0.273	
	18	INDIAN RIVER	25.061	22.406	21.376	19.540	15.788		1.000	0.894	0.853	0.780	0.630	
	19	JACKSON	2.678	1.764	1.497	1.097	0.508		1.000	0.659	0.559	0.409	0.190	
	20	LEE	11.515	9.341	8.542	7.178	4.630		1.000	0.811	0.742	0.623	0.402	
	21	LEON	2.289	1.535	1.324	1.008	0.528		1.000	0.671	0.578	0.441	0.231	
	22	MARION	3.291	2.259	1.957	1.493	0.777		1.000	0.686	0.595	0.454	0.236	
	23	MARTIN	11.263	9.146	8.382	7.106	4.740		1.000	0.812	0.744	0.631	0.421	
	24	MARTIN	27.686	24.791	23.630	21.567	17.334		1.000	0.895	0.854	0.779	0.626	
	25	MIAMI-DADE	15.877	13.545	12.638	11.041	7.878		1.000	0.853	0.796	0.695	0.496	
	26	MIAMI-DADE	22.893	20.163	19.095	17.194	13.303		1.000	0.881	0.834	0.751	0.581	
	27	MONROE	33.413	30.507	29.285	27.053	22.267		1.000	0.913	0.876	0.810	0.666	
	28	MONROE	40.034	36.987	35.707	33.367	28.220		1.000	0.924	0.892	0.833	0.705	
	29	OKALOOSA	11.959	10.046	9.336	8.096	5.695		1.000	0.840	0.781	0.677	0.476	
	30	OSCEOLA	4.856	3.487	3.052	2.367	1.262		1.000	0.718	0.629	0.487	0.260	
	31	OSCEOLA	7.123	5.475	4.923	4.042	2.512		1.000	0.769	0.691	0.568	0.353	
	32	PALM BEACH	15.494	13.008	12.072	10.470	7.392		1.000	0.840	0.779	0.676	0.477	
	33	PALM BEACH	24.399	21.415	20.233	18.137	13.935		1.000	0.878	0.829	0.743	0.571	
	34	PINELLAS	7.913	6.300	5.730	4.789	3.072		1.000	0.796	0.724	0.605	0.388	
	35	SAINT JOHNS	4.217	3.237	2.928	2.432	1.581		1.000	0.767	0.694	0.577	0.375	
	36	SANTA ROSA	7.934	6.468	5.934	5.009	3.247		1.000	0.815	0.748	0.631	0.409	
	37	SEMINOLE	3.945	2.743	2.373	1.808	0.933		1.000	0.695	0.602	0.458	0.237	
	38	TAYLOR	2.295	1.586	1.381	1.079	0.605		1.000	0.691	0.602	0.470	0.264	
	39	VOLUSIA	7.142	5.780	5.311	4.535	3.131		1.000	0.809	0.744	0.635	0.438	
	40	WAKULLA	7.038	5.639	5.171	4.404	3.043		1.000	0.801	0.735	0.626	0.432	

Form A-6: Logical Relationship to Risk - Construction

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Policy	Location	County	Loss Cost per Construction		Frame / Masonry
			Masonry	Frame	
Owners	1	BAY	3.863	4.190	1.085
	2	BREVARD	4.521	4.590	1.015
	3	BREVARD	4.369	4.429	1.014
	4	BROWARD	7.980	8.204	1.028
	5	BROWARD	13.289	14.045	1.057
	6	CITRUS	3.273	3.324	1.016
	7	CLAY	0.862	0.895	1.038
	8	COLLIER	6.018	6.141	1.021
	9	COLUMBIA	0.850	0.885	1.041
	10	DIXIE	2.684	2.882	1.074
	11	DUVAL	1.690	1.878	1.111
	12	FRANKLIN	5.539	6.099	1.101
	13	GLADES	5.240	5.349	1.021
	14	HAMILTON	0.755	0.786	1.041
	15	HERNANDO	4.327	4.396	1.016
	16	HILLSBOROUGH	3.878	3.943	1.017
	17	HOLMES	1.456	1.527	1.049
	18	INDIAN RIVER	11.084	11.844	1.068
	19	JACKSON	1.117	1.167	1.045
	20	LEE	5.311	5.439	1.024
	21	LEON	1.066	1.114	1.045
	22	MARION	1.585	1.621	1.023
	23	MARTIN	5.758	6.002	1.042
	24	MARTIN	12.669	13.648	1.077
	25	MIAMI-DADE	7.257	7.451	1.027
	26	MIAMI-DADE	10.416	10.846	1.041
	27	MONROE	12.698	13.480	1.062
	28	MONROE	19.295	21.049	1.091
	29	OKALOOSA	2.870	3.048	1.062
	30	OSCEOLA	2.046	2.095	1.024
	31	OSCEOLA	2.594	2.660	1.025
	32	PALM BEACH	7.405	7.732	1.044
	33	PALM BEACH	10.532	11.134	1.057
	34	PINELLAS	4.383	4.486	1.024
	35	SAINT JOHNS	1.354	1.424	1.052
	36	SANTA ROSA	2.101	2.216	1.055
	37	SEMINOLE	3.221	3.279	1.018
	38	TAYLOR	0.960	1.002	1.044
	39	VOLUSIA	2.950	2.997	1.016
	40	WAKULLA	2.386	2.541	1.065

Form A-6: Logical Relationship to Risk - Construction

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Policy	Location	County	Loss Cost per Construction		Frame / Masonry
			Masonry	Frame	
Renters	1	BAY	1.425	1.622	1.138
	2	BREVARD	0.895	0.923	1.031
	3	BREVARD	0.836	0.860	1.028
	4	BROWARD	2.088	2.168	1.038
	5	BROWARD	4.703	5.181	1.102
	6	CITRUS	0.631	0.645	1.022
	7	CLAY	0.219	0.239	1.089
	8	COLLIER	1.337	1.384	1.034
	9	COLUMBIA	0.212	0.230	1.086
	10	DIXIE	0.952	1.070	1.124
	11	DUVAL	0.410	0.534	1.304
	12	FRANKLIN	2.305	2.682	1.164
	13	GLADES	0.992	1.020	1.028
	14	HAMILTON	0.193	0.210	1.087
	15	HERNANDO	0.948	0.970	1.024
	16	HILLSBOROUGH	0.653	0.671	1.027
	17	HOLMES	0.396	0.430	1.084
	18	INDIAN RIVER	4.078	4.577	1.123
	19	JACKSON	0.291	0.316	1.086
	20	LEE	1.091	1.120	1.027
	21	LEON	0.283	0.307	1.085
	22	MARION	0.249	0.257	1.033
	23	MARTIN	1.357	1.412	1.041
	24	MARTIN	4.674	5.372	1.149
	25	MIAMI-DADE	1.873	1.938	1.035
	26	MIAMI-DADE	3.317	3.542	1.068
	27	MONROE	5.156	5.645	1.095
	28	MONROE	8.856	10.136	1.145
	29	OKALOOSA	0.968	1.056	1.091
	30	OSCEOLA	0.324	0.335	1.033
	31	OSCEOLA	0.436	0.449	1.030
	32	PALM BEACH	1.894	1.990	1.051
	33	PALM BEACH	3.258	3.579	1.098
	34	PINELLAS	0.979	1.013	1.035
	35	SAINT JOHNS	0.414	0.451	1.088
	36	SANTA ROSA	0.662	0.715	1.080
	37	SEMINOLE	0.640	0.654	1.023
	38	TAYLOR	0.273	0.295	1.082
	39	VOLUSIA	0.497	0.510	1.026
	40	WAKULLA	0.804	0.889	1.106

Form A-6: Logical Relationship to Risk - Construction

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Construction		Construction / Policy
			Concrete	Frame	
Condo Unit	1	BAY	1.723	1.942	1.127
	2	BREVARD	1.290	1.324	1.026
	3	BREVARD	1.219	1.248	1.023
	4	BROWARD	2.762	2.860	1.036
	5	BROWARD	5.768	6.298	1.092
	6	CITRUS	0.918	0.936	1.020
	7	CLAY	0.290	0.312	1.075
	8	COLLIER	1.853	1.910	1.030
	9	COLUMBIA	0.282	0.303	1.074
	10	DIXIE	1.160	1.293	1.114
	11	DUVAL	0.550	0.687	1.249
	12	FRANKLIN	2.721	3.136	1.152
	13	GLADES	1.450	1.488	1.026
	14	HAMILTON	0.255	0.274	1.075
	15	HERNANDO	1.321	1.349	1.021
	16	HILLSBOROUGH	0.997	1.021	1.023
	17	HOLMES	0.515	0.553	1.075
	18	INDIAN RIVER	4.954	5.505	1.111
	19	JACKSON	0.383	0.411	1.075
	20	LEE	1.550	1.591	1.026
	21	LEON	0.370	0.397	1.075
	22	MARION	0.388	0.399	1.030
	23	MARTIN	1.850	1.926	1.041
	24	MARTIN	5.678	6.439	1.134
	25	MIAMI-DADE	2.487	2.568	1.033
	26	MIAMI-DADE	4.168	4.425	1.062
	27	MONROE	6.138	6.680	1.088
	28	MONROE	10.302	11.694	1.135
	29	OKALOOSA	1.192	1.293	1.085
	30	OSCEOLA	0.503	0.518	1.030
	31	OSCEOLA	0.661	0.680	1.029
	32	PALM BEACH	2.519	2.644	1.049
	33	PALM BEACH	4.122	4.487	1.088
	34	PINELLAS	1.357	1.400	1.032
	35	SAINT JOHNS	0.522	0.563	1.080
	36	SANTA ROSA	0.828	0.890	1.074
	37	SEMINOLE	0.922	0.941	1.021
	38	TAYLOR	0.350	0.376	1.073
	39	VOLUSIA	0.757	0.774	1.023
	40	WAKULLA	0.991	1.087	1.098

Form A-6: Logical Relationship to Risk - Construction

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Construction		
			Concrete		
Commercial Residential	1	BAY	14.376		
	2	BREVARD	8.598		
	3	BREVARD	7.822		
	4	BROWARD	15.876		
	5	BROWARD	27.147		
	6	CITRUS	4.341		
	7	CLAY	1.750		
	8	COLLIER	14.760		
	9	COLUMBIA	1.554		
	10	DIXIE	7.182		
	11	DUVAL	5.207		
	12	FRANKLIN	16.460		
	13	GLADES	10.698		
	14	HAMILTON	1.300		
	15	HERNANDO	7.346		
	16	HILLSBOROUGH	6.921		
	17	HOLMES	4.254		
	18	INDIAN RIVER	25.061		
	19	JACKSON	2.678		
	20	LEE	11.515		
	21	LEON	2.289		
	22	MARION	3.291		
	23	MARTIN	11.263		
	24	MARTIN	27.686		
	25	MIAMI-DADE	15.877		
	26	MIAMI-DADE	22.893		
	27	MONROE	33.413		
	28	MONROE	40.034		
	29	OKALOOSA	11.959		
	30	OSCEOLA	4.856		
	31	OSCEOLA	7.123		
	32	PALM BEACH	15.494		
	33	PALM BEACH	24.399		
	34	PINELLAS	7.913		
	35	SAINT JOHNS	4.217		
	36	SANTA ROSA	7.934		
	37	SEMINOLE	3.945		
	38	TAYLOR	2.295		
	39	VOLUSIA	7.142		
	40	WAKULLA	7.038		

Form A-6: Logical Relationship to Risk - Policy Form

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Location	County	Loss Cost per Policy Type			Mobile homes / Frame Owners	Mobile Homes / Masonry Owners
		Frame Owners	Masonry Owners	Mobile Homes		
1	BAY	4.190	3.863	16.793	4.008	4.347
2	BREVARD	4.590	4.521	13.278	2.893	2.937
3	BREVARD	4.429	4.369	12.433	2.807	2.846
4	BROWARD	8.204	7.980	25.412	3.098	3.184
5	BROWARD	14.045	13.289	45.049	3.208	3.390
6	CITRUS	3.324	3.273	7.795	2.345	2.381
7	CLAY	0.895	0.862	2.717	3.036	3.152
8	COLLIER	6.141	6.018	23.503	3.827	3.906
9	COLUMBIA	0.885	0.850	2.556	2.888	3.005
10	DIXIE	2.882	2.684	11.053	3.835	4.118
11	DUVAL	1.878	1.690	6.873	3.660	4.067
12	FRANKLIN	6.099	5.539	22.352	3.665	4.035
13	GLADES	5.349	5.240	16.872	3.154	3.220
14	HAMILTON	0.786	0.755	2.250	2.862	2.981
15	HERNANDO	4.396	4.327	13.231	3.010	3.058
16	HILLSBOROUGH	3.943	3.878	10.744	2.725	2.771
17	HOLMES	1.527	1.456	5.109	3.346	3.508
18	INDIAN RIVER	11.844	11.084	44.321	3.742	3.999
19	JACKSON	1.167	1.117	3.531	3.025	3.160
20	LEE	5.439	5.311	19.542	3.593	3.680
21	LEON	1.114	1.066	3.505	3.146	3.289
22	MARION	1.621	1.585	6.004	3.704	3.789
23	MARTIN	6.002	5.758	17.543	2.923	3.047
24	MARTIN	13.648	12.669	46.080	3.376	3.637
25	MIAMI-DADE	7.451	7.257	23.249	3.120	3.204
26	MIAMI-DADE	10.846	10.416	35.836	3.304	3.440
27	MONROE	13.480	12.698	53.076	3.937	4.180
28	MONROE	21.049	19.295	75.209	3.573	3.898
29	OKALOOSA	3.048	2.870	13.223	4.338	4.608
30	OSCEOLA	2.095	2.046	8.262	3.945	4.039
31	OSCEOLA	2.660	2.594	12.499	4.700	4.819
32	PALM BEACH	7.732	7.405	25.089	3.245	3.388
33	PALM BEACH	11.134	10.532	38.748	3.480	3.679
34	PINELLAS	4.486	4.383	12.119	2.701	2.765
35	SAINT JOHNS	1.424	1.354	5.268	3.699	3.891
36	SANTA ROSA	2.216	2.101	9.466	4.271	4.506
37	SEMINOLE	3.279	3.221	6.956	2.121	2.160
38	TAYLOR	1.002	0.960	3.292	3.285	3.431
39	VOLUSIA	2.997	2.950	11.403	3.804	3.866
40	WAKULLA	2.541	2.386	9.882	3.889	4.142

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
Frame Owners	1	BAY	3.204	0.175	0.647	0.164	1.000	0.054	0.202	0.051
	2	BREVARD	4.005	0.123	0.385	0.076	1.000	0.031	0.096	0.019
	3	BREVARD	3.880	0.119	0.362	0.068	1.000	0.031	0.093	0.018
	4	BROWARD	6.921	0.199	0.885	0.199	1.000	0.029	0.128	0.029
	5	BROWARD	11.164	0.290	2.017	0.574	1.000	0.026	0.181	0.051
	6	CITRUS	2.916	0.086	0.284	0.039	1.000	0.029	0.097	0.013
	7	CLAY	0.731	0.044	0.102	0.018	1.000	0.060	0.139	0.024
	8	COLLIER	5.259	0.190	0.583	0.108	1.000	0.036	0.111	0.021
	9	COLUMBIA	0.728	0.042	0.098	0.017	1.000	0.058	0.135	0.024
	10	DIXIE	2.222	0.125	0.432	0.104	1.000	0.056	0.194	0.047
	11	DUVAL	1.528	0.083	0.210	0.057	1.000	0.054	0.138	0.037
	12	FRANKLIN	4.538	0.219	1.052	0.289	1.000	0.048	0.232	0.064
	13	GLADES	4.682	0.157	0.429	0.081	1.000	0.034	0.092	0.017
	14	HAMILTON	0.644	0.037	0.089	0.015	1.000	0.058	0.139	0.024
	15	HERNANDO	3.787	0.124	0.416	0.069	1.000	0.033	0.110	0.018
	16	HILLSBOROUGH	3.498	0.109	0.284	0.051	1.000	0.031	0.081	0.015
	17	HOLMES	1.238	0.075	0.180	0.034	1.000	0.060	0.146	0.028
	18	INDIAN RIVER	9.273	0.282	1.780	0.509	1.000	0.030	0.192	0.055
	19	JACKSON	0.953	0.056	0.135	0.024	1.000	0.059	0.141	0.025
	20	LEE	4.710	0.169	0.471	0.089	1.000	0.036	0.100	0.019
	21	LEON	0.907	0.054	0.129	0.024	1.000	0.059	0.143	0.026
	22	MARION	1.422	0.070	0.107	0.022	1.000	0.049	0.075	0.015
	23	MARTIN	5.141	0.155	0.613	0.093	1.000	0.030	0.119	0.018
	24	MARTIN	10.678	0.285	2.111	0.575	1.000	0.027	0.198	0.054
	25	MIAMI-DADE	6.300	0.183	0.793	0.176	1.000	0.029	0.126	0.028
	26	MIAMI-DADE	8.833	0.242	1.398	0.373	1.000	0.027	0.158	0.042
	27	MONROE	10.355	0.303	2.291	0.531	1.000	0.029	0.221	0.051
	28	MONROE	15.575	0.406	3.986	1.082	1.000	0.026	0.256	0.069
	29	OKALOOSA	2.373	0.147	0.429	0.099	1.000	0.062	0.181	0.042
	30	OSCEOLA	1.835	0.092	0.139	0.029	1.000	0.050	0.076	0.016
	31	OSCEOLA	2.313	0.123	0.184	0.041	1.000	0.053	0.079	0.018
	32	PALM BEACH	6.539	0.198	0.843	0.152	1.000	0.030	0.129	0.023
	33	PALM BEACH	9.081	0.264	1.448	0.341	1.000	0.029	0.159	0.038
	34	PINELLAS	3.867	0.112	0.444	0.062	1.000	0.029	0.115	0.016
	35	SAINT JOHNS	1.129	0.069	0.187	0.038	1.000	0.061	0.166	0.033
	36	SANTA ROSA	1.745	0.113	0.295	0.062	1.000	0.065	0.169	0.036
	37	SEMINOLE	2.872	0.080	0.292	0.036	1.000	0.028	0.102	0.012
	38	TAYLOR	0.805	0.049	0.126	0.022	1.000	0.061	0.156	0.027
	39	VOLUSIA	2.639	0.103	0.202	0.053	1.000	0.039	0.077	0.020
	40	WAKULLA	1.980	0.116	0.363	0.082	1.000	0.059	0.183	0.042

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
Masonry Owners	1	BAY	2.976	0.175	0.569	0.143	1.000	0.059	0.191	0.048
	2	BREVARD	3.950	0.123	0.374	0.074	1.000	0.031	0.095	0.019
	3	BREVARD	3.832	0.119	0.352	0.066	1.000	0.031	0.092	0.017
	4	BROWARD	6.737	0.199	0.852	0.192	1.000	0.030	0.126	0.029
	5	BROWARD	10.647	0.290	1.833	0.519	1.000	0.027	0.172	0.049
	6	CITRUS	2.872	0.086	0.278	0.038	1.000	0.030	0.097	0.013
	7	CLAY	0.708	0.044	0.094	0.016	1.000	0.062	0.133	0.022
	8	COLLIER	5.159	0.190	0.564	0.104	1.000	0.037	0.109	0.020
	9	COLUMBIA	0.702	0.042	0.091	0.015	1.000	0.060	0.129	0.022
	10	DIXIE	2.083	0.125	0.385	0.091	1.000	0.060	0.185	0.044
	11	DUVAL	1.402	0.083	0.163	0.042	1.000	0.059	0.116	0.030
	12	FRANKLIN	4.168	0.219	0.905	0.247	1.000	0.053	0.217	0.059
	13	GLADES	4.587	0.157	0.417	0.078	1.000	0.034	0.091	0.017
	14	HAMILTON	0.621	0.037	0.083	0.014	1.000	0.060	0.133	0.022
	15	HERNANDO	3.730	0.124	0.407	0.067	1.000	0.033	0.109	0.018
	16	HILLSBOROUGH	3.442	0.109	0.277	0.049	1.000	0.032	0.081	0.014
	17	HOLMES	1.184	0.075	0.167	0.031	1.000	0.063	0.141	0.026
	18	INDIAN RIVER	8.764	0.282	1.588	0.451	1.000	0.032	0.181	0.051
	19	JACKSON	0.915	0.056	0.124	0.021	1.000	0.061	0.136	0.023
	20	LEE	4.597	0.169	0.458	0.087	1.000	0.037	0.100	0.019
	21	LEON	0.871	0.054	0.120	0.022	1.000	0.062	0.138	0.025
	22	MARION	1.390	0.070	0.104	0.021	1.000	0.050	0.075	0.015
	23	MARTIN	4.924	0.155	0.589	0.090	1.000	0.031	0.120	0.018
	24	MARTIN	10.048	0.285	1.845	0.492	1.000	0.028	0.184	0.049
	25	MIAMI-DADE	6.138	0.183	0.766	0.170	1.000	0.030	0.125	0.028
	26	MIAMI-DADE	8.516	0.242	1.309	0.349	1.000	0.028	0.154	0.041
	27	MONROE	9.817	0.303	2.099	0.479	1.000	0.031	0.214	0.049
	28	MONROE	14.461	0.406	3.500	0.928	1.000	0.028	0.242	0.064
	29	OKALOOSA	2.239	0.147	0.394	0.090	1.000	0.066	0.176	0.040
	30	OSCEOLA	1.791	0.092	0.135	0.027	1.000	0.052	0.075	0.015
	31	OSCEOLA	2.254	0.123	0.178	0.039	1.000	0.054	0.079	0.017
	32	PALM BEACH	6.260	0.198	0.801	0.145	1.000	0.032	0.128	0.023
	33	PALM BEACH	8.638	0.264	1.323	0.307	1.000	0.031	0.153	0.035
	34	PINELLAS	3.781	0.112	0.430	0.059	1.000	0.030	0.114	0.016
	35	SAINT JOHNS	1.078	0.069	0.173	0.034	1.000	0.064	0.160	0.032
	36	SANTA ROSA	1.657	0.113	0.274	0.057	1.000	0.068	0.165	0.035
	37	SEMINOLE	2.821	0.080	0.285	0.035	1.000	0.029	0.101	0.012
	38	TAYLOR	0.774	0.049	0.117	0.020	1.000	0.064	0.151	0.025
	39	VOLUSIA	2.598	0.103	0.197	0.051	1.000	0.040	0.076	0.020
	40	WAKULLA	1.867	0.116	0.328	0.074	1.000	0.062	0.176	0.039

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
Mobile Homes	1	BAY	13.094	0.175	2.543	0.981	1.000	0.013	0.194	0.075
	2	BREVARD	10.845	0.123	1.640	0.671	1.000	0.011	0.151	0.062
	3	BREVARD	10.257	0.119	1.460	0.597	1.000	0.012	0.142	0.058
	4	BROWARD	20.062	0.199	3.606	1.544	1.000	0.010	0.180	0.077
	5	BROWARD	33.758	0.290	7.754	3.246	1.000	0.009	0.230	0.096
	6	CITRUS	6.702	0.086	0.721	0.286	1.000	0.013	0.108	0.043
	7	CLAY	2.412	0.044	0.195	0.066	1.000	0.018	0.081	0.027
	8	COLLIER	18.724	0.190	3.239	1.349	1.000	0.010	0.173	0.072
	9	COLUMBIA	2.278	0.042	0.177	0.058	1.000	0.019	0.078	0.026
	10	DIXIE	8.819	0.125	1.520	0.588	1.000	0.014	0.172	0.067
	11	DUVAL	5.573	0.083	0.878	0.339	1.000	0.015	0.157	0.061
	12	FRANKLIN	17.004	0.219	3.700	1.429	1.000	0.013	0.218	0.084
	13	GLADES	13.828	0.157	2.039	0.848	1.000	0.011	0.147	0.061
	14	HAMILTON	2.008	0.037	0.154	0.050	1.000	0.019	0.077	0.025
	15	HERNANDO	10.840	0.124	1.609	0.658	1.000	0.011	0.148	0.061
	16	HILLSBOROUGH	9.049	0.109	1.123	0.462	1.000	0.012	0.124	0.051
	17	HOLMES	4.419	0.075	0.452	0.163	1.000	0.017	0.102	0.037
	18	INDIAN RIVER	32.629	0.282	8.140	3.271	1.000	0.009	0.249	0.100
	19	JACKSON	3.134	0.056	0.255	0.086	1.000	0.018	0.081	0.027
	20	LEE	15.860	0.169	2.479	1.034	1.000	0.011	0.156	0.065
	21	LEON	3.059	0.054	0.290	0.102	1.000	0.018	0.095	0.033
	22	MARION	5.253	0.070	0.492	0.190	1.000	0.013	0.094	0.036
	23	MARTIN	14.321	0.155	2.150	0.917	1.000	0.011	0.150	0.064
	24	MARTIN	34.141	0.285	8.231	3.423	1.000	0.008	0.241	0.100
	25	MIAMI-DADE	18.375	0.183	3.285	1.407	1.000	0.010	0.179	0.077
	26	MIAMI-DADE	27.203	0.242	5.901	2.489	1.000	0.009	0.217	0.091
	27	MONROE	38.724	0.303	9.872	4.177	1.000	0.008	0.255	0.108
	28	MONROE	53.783	0.406	14.798	6.221	1.000	0.008	0.275	0.116
	29	OKALOOSA	10.501	0.147	1.854	0.721	1.000	0.014	0.177	0.069
	30	OSCEOLA	7.159	0.092	0.723	0.288	1.000	0.013	0.101	0.040
	31	OSCEOLA	10.393	0.123	1.407	0.576	1.000	0.012	0.135	0.055
	32	PALM BEACH	19.870	0.198	3.516	1.503	1.000	0.010	0.177	0.076
	33	PALM BEACH	29.486	0.264	6.326	2.673	1.000	0.009	0.215	0.091
	34	PINELLAS	10.035	0.112	1.396	0.576	1.000	0.011	0.139	0.057
	35	SAINT JOHNS	4.384	0.069	0.589	0.225	1.000	0.016	0.134	0.051
	36	SANTA ROSA	7.679	0.113	1.206	0.469	1.000	0.015	0.157	0.061
	37	SEMINOLE	6.084	0.080	0.570	0.222	1.000	0.013	0.094	0.037
	38	TAYLOR	2.852	0.049	0.287	0.103	1.000	0.017	0.101	0.036
	39	VOLUSIA	9.226	0.103	1.477	0.597	1.000	0.011	0.160	0.065
	40	WAKULLA	7.973	0.116	1.295	0.498	1.000	0.015	0.162	0.062

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
Frame Renters	1	BAY	0.000	0.000	1.294	0.328			1.000	0.253
	2	BREVARD	0.000	0.000	0.770	0.153			1.000	0.198
	3	BREVARD	0.000	0.000	0.723	0.136			1.000	0.189
	4	BROWARD	0.000	0.000	1.769	0.398			1.000	0.225
	5	BROWARD	0.000	0.000	4.034	1.147			1.000	0.284
	6	CITRUS	0.000	0.000	0.567	0.078			1.000	0.137
	7	CLAY	0.000	0.000	0.203	0.036			1.000	0.175
	8	COLLIER	0.000	0.000	1.167	0.217			1.000	0.186
	9	COLUMBIA	0.000	0.000	0.196	0.035			1.000	0.177
	10	DIXIE	0.000	0.000	0.863	0.207			1.000	0.240
	11	DUVAL	0.000	0.000	0.420	0.114			1.000	0.272
	12	FRANKLIN	0.000	0.000	2.104	0.577			1.000	0.274
	13	GLADES	0.000	0.000	0.859	0.161			1.000	0.188
	14	HAMILTON	0.000	0.000	0.179	0.031			1.000	0.173
	15	HERNANDO	0.000	0.000	0.832	0.138			1.000	0.166
	16	HILLSBOROUGH	0.000	0.000	0.569	0.102			1.000	0.179
	17	HOLMES	0.000	0.000	0.361	0.069			1.000	0.191
	18	INDIAN RIVER	0.000	0.000	3.560	1.017			1.000	0.286
	19	JACKSON	0.000	0.000	0.269	0.047			1.000	0.175
	20	LEE	0.000	0.000	0.943	0.177			1.000	0.188
	21	LEON	0.000	0.000	0.259	0.048			1.000	0.185
	22	MARION	0.000	0.000	0.214	0.043			1.000	0.201
	23	MARTIN	0.000	0.000	1.226	0.186			1.000	0.152
	24	MARTIN	0.000	0.000	4.221	1.150			1.000	0.272
	25	MIAMI-DADE	0.000	0.000	1.587	0.351			1.000	0.221
	26	MIAMI-DADE	0.000	0.000	2.796	0.746			1.000	0.267
	27	MONROE	0.000	0.000	4.582	1.063			1.000	0.232
	28	MONROE	0.000	0.000	7.972	2.164			1.000	0.272
	29	OKALOOSA	0.000	0.000	0.859	0.197			1.000	0.229
	30	OSCEOLA	0.000	0.000	0.278	0.057			1.000	0.206
	31	OSCEOLA	0.000	0.000	0.367	0.081			1.000	0.222
	32	PALM BEACH	0.000	0.000	1.685	0.305			1.000	0.181
	33	PALM BEACH	0.000	0.000	2.897	0.682			1.000	0.235
	34	PINELLAS	0.000	0.000	0.889	0.124			1.000	0.140
	35	SAINT JOHNS	0.000	0.000	0.375	0.076			1.000	0.202
	36	SANTA ROSA	0.000	0.000	0.591	0.124			1.000	0.210
	37	SEMINOLE	0.000	0.000	0.583	0.071			1.000	0.122
	38	TAYLOR	0.000	0.000	0.251	0.044			1.000	0.174
	39	VOLUSIA	0.000	0.000	0.405	0.105			1.000	0.260
	40	WAKULLA	0.000	0.000	0.725	0.164			1.000	0.227

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
Masonry Renters	1	BAY	0.000	0.000	1.138	0.287			1.000	0.252
	2	BREVARD	0.000	0.000	0.747	0.148			1.000	0.198
	3	BREVARD	0.000	0.000	0.704	0.133			1.000	0.188
	4	BROWARD	0.000	0.000	1.703	0.385			1.000	0.226
	5	BROWARD	0.000	0.000	3.665	1.038			1.000	0.283
	6	CITRUS	0.000	0.000	0.555	0.076			1.000	0.136
	7	CLAY	0.000	0.000	0.188	0.032			1.000	0.169
	8	COLLIER	0.000	0.000	1.129	0.209			1.000	0.185
	9	COLUMBIA	0.000	0.000	0.181	0.031			1.000	0.171
	10	DIXIE	0.000	0.000	0.770	0.182			1.000	0.237
	11	DUVAL	0.000	0.000	0.325	0.085			1.000	0.260
	12	FRANKLIN	0.000	0.000	1.810	0.495			1.000	0.273
	13	GLADES	0.000	0.000	0.835	0.157			1.000	0.188
	14	HAMILTON	0.000	0.000	0.165	0.027			1.000	0.166
	15	HERNANDO	0.000	0.000	0.814	0.134			1.000	0.165
	16	HILLSBOROUGH	0.000	0.000	0.554	0.099			1.000	0.178
	17	HOLMES	0.000	0.000	0.334	0.062			1.000	0.187
	18	INDIAN RIVER	0.000	0.000	3.177	0.901			1.000	0.284
	19	JACKSON	0.000	0.000	0.249	0.042			1.000	0.170
	20	LEE	0.000	0.000	0.917	0.174			1.000	0.190
	21	LEON	0.000	0.000	0.240	0.043			1.000	0.180
	22	MARION	0.000	0.000	0.208	0.041			1.000	0.199
	23	MARTIN	0.000	0.000	1.177	0.180			1.000	0.153
	24	MARTIN	0.000	0.000	3.689	0.984			1.000	0.267
	25	MIAMI-DADE	0.000	0.000	1.532	0.341			1.000	0.222
	26	MIAMI-DADE	0.000	0.000	2.619	0.698			1.000	0.267
	27	MONROE	0.000	0.000	4.199	0.957			1.000	0.228
	28	MONROE	0.000	0.000	7.001	1.855			1.000	0.265
	29	OKALOOSA	0.000	0.000	0.788	0.180			1.000	0.229
	30	OSCEOLA	0.000	0.000	0.269	0.055			1.000	0.204
	31	OSCEOLA	0.000	0.000	0.357	0.079			1.000	0.221
	32	PALM BEACH	0.000	0.000	1.603	0.291			1.000	0.181
	33	PALM BEACH	0.000	0.000	2.645	0.613			1.000	0.232
	34	PINELLAS	0.000	0.000	0.860	0.119			1.000	0.138
	35	SAINT JOHNS	0.000	0.000	0.346	0.068			1.000	0.198
	36	SANTA ROSA	0.000	0.000	0.548	0.114			1.000	0.209
	37	SEMINOLE	0.000	0.000	0.570	0.069			1.000	0.122
	38	TAYLOR	0.000	0.000	0.233	0.039			1.000	0.169
	39	VOLUSIA	0.000	0.000	0.394	0.103			1.000	0.260
	40	WAKULLA	0.000	0.000	0.657	0.147			1.000	0.224

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
Frame Condo Unit	1	BAY	0.320	0.000	1.294	0.328	0.248		1.000	0.253
	2	BREVARD	0.400	0.000	0.770	0.153	0.520		1.000	0.198
	3	BREVARD	0.388	0.000	0.723	0.136	0.536		1.000	0.189
	4	BROWARD	0.692	0.000	1.769	0.398	0.391		1.000	0.225
	5	BROWARD	1.116	0.000	4.034	1.147	0.277		1.000	0.284
	6	CITRUS	0.292	0.000	0.567	0.078	0.514		1.000	0.137
	7	CLAY	0.073	0.000	0.203	0.036	0.360		1.000	0.175
	8	COLLIER	0.526	0.000	1.167	0.217	0.451		1.000	0.186
	9	COLUMBIA	0.073	0.000	0.196	0.035	0.372		1.000	0.177
	10	DIXIE	0.222	0.000	0.863	0.207	0.257		1.000	0.240
	11	DUVAL	0.153	0.000	0.420	0.114	0.364		1.000	0.272
	12	FRANKLIN	0.454	0.000	2.104	0.577	0.216		1.000	0.274
	13	GLADES	0.468	0.000	0.859	0.161	0.545		1.000	0.188
	14	HAMILTON	0.064	0.000	0.179	0.031	0.360		1.000	0.173
	15	HERNANDO	0.379	0.000	0.832	0.138	0.455		1.000	0.166
	16	HILLSBOROUGH	0.350	0.000	0.569	0.102	0.615		1.000	0.179
	17	HOLMES	0.124	0.000	0.361	0.069	0.343		1.000	0.191
	18	INDIAN RIVER	0.927	0.000	3.560	1.017	0.260		1.000	0.286
	19	JACKSON	0.095	0.000	0.269	0.047	0.354		1.000	0.175
	20	LEE	0.471	0.000	0.943	0.177	0.500		1.000	0.188
	21	LEON	0.091	0.000	0.259	0.048	0.350		1.000	0.185
	22	MARION	0.142	0.000	0.214	0.043	0.664		1.000	0.201
	23	MARTIN	0.514	0.000	1.226	0.186	0.419		1.000	0.152
	24	MARTIN	1.068	0.000	4.221	1.150	0.253		1.000	0.272
	25	MIAMI-DADE	0.630	0.000	1.587	0.351	0.397		1.000	0.221
	26	MIAMI-DADE	0.883	0.000	2.796	0.746	0.316		1.000	0.267
	27	MONROE	1.035	0.000	4.582	1.063	0.226		1.000	0.232
	28	MONROE	1.557	0.000	7.972	2.164	0.195		1.000	0.272
	29	OKALOOSA	0.237	0.000	0.859	0.197	0.276		1.000	0.229
	30	OSCEOLA	0.183	0.000	0.278	0.057	0.661		1.000	0.206
	31	OSCEOLA	0.231	0.000	0.367	0.081	0.630		1.000	0.222
	32	PALM BEACH	0.654	0.000	1.685	0.305	0.388		1.000	0.181
	33	PALM BEACH	0.908	0.000	2.897	0.682	0.313		1.000	0.235
	34	PINELLAS	0.387	0.000	0.889	0.124	0.435		1.000	0.140
	35	SAINT JOHNS	0.113	0.000	0.375	0.076	0.301		1.000	0.202
	36	SANTA ROSA	0.175	0.000	0.591	0.124	0.295		1.000	0.210
	37	SEMINOLE	0.287	0.000	0.583	0.071	0.493		1.000	0.122
	38	TAYLOR	0.081	0.000	0.251	0.044	0.320		1.000	0.174
	39	VOLUSIA	0.264	0.000	0.405	0.105	0.652		1.000	0.260
	40	WAKULLA	0.198	0.000	0.725	0.164	0.273		1.000	0.227

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
Masonry Condo Unit	1	BAY	0.298	0.000	1.138	0.287	0.261		1.000	0.252
	2	BREVARD	0.395	0.000	0.747	0.148	0.529		1.000	0.198
	3	BREVARD	0.383	0.000	0.704	0.133	0.545		1.000	0.188
	4	BROWARD	0.674	0.000	1.703	0.385	0.396		1.000	0.226
	5	BROWARD	1.065	0.000	3.665	1.038	0.290		1.000	0.283
	6	CITRUS	0.287	0.000	0.555	0.076	0.517		1.000	0.136
	7	CLAY	0.071	0.000	0.188	0.032	0.377		1.000	0.169
	8	COLLIER	0.516	0.000	1.129	0.209	0.457		1.000	0.185
	9	COLUMBIA	0.070	0.000	0.181	0.031	0.388		1.000	0.171
	10	DIXIE	0.208	0.000	0.770	0.182	0.271		1.000	0.237
	11	DUVAL	0.140	0.000	0.325	0.085	0.431		1.000	0.260
	12	FRANKLIN	0.417	0.000	1.810	0.495	0.230		1.000	0.273
	13	GLADES	0.459	0.000	0.835	0.157	0.549		1.000	0.188
	14	HAMILTON	0.062	0.000	0.165	0.027	0.376		1.000	0.166
	15	HERNANDO	0.373	0.000	0.814	0.134	0.458		1.000	0.165
	16	HILLSBOROUGH	0.344	0.000	0.554	0.099	0.621		1.000	0.178
	17	HOLMES	0.118	0.000	0.334	0.062	0.354		1.000	0.187
	18	INDIAN RIVER	0.876	0.000	3.177	0.901	0.276		1.000	0.284
	19	JACKSON	0.092	0.000	0.249	0.042	0.368		1.000	0.170
	20	LEE	0.460	0.000	0.917	0.174	0.501		1.000	0.190
	21	LEON	0.087	0.000	0.240	0.043	0.363		1.000	0.180
	22	MARION	0.139	0.000	0.208	0.041	0.670		1.000	0.199
	23	MARTIN	0.492	0.000	1.177	0.180	0.418		1.000	0.153
	24	MARTIN	1.005	0.000	3.689	0.984	0.272		1.000	0.267
	25	MIAMI-DADE	0.614	0.000	1.532	0.341	0.401		1.000	0.222
	26	MIAMI-DADE	0.852	0.000	2.619	0.698	0.325		1.000	0.267
	27	MONROE	0.982	0.000	4.199	0.957	0.234		1.000	0.228
	28	MONROE	1.446	0.000	7.001	1.855	0.207		1.000	0.265
	29	OKALOOSA	0.224	0.000	0.788	0.180	0.284		1.000	0.229
	30	OSCEOLA	0.179	0.000	0.269	0.055	0.665		1.000	0.204
	31	OSCEOLA	0.225	0.000	0.357	0.079	0.632		1.000	0.221
	32	PALM BEACH	0.626	0.000	1.603	0.291	0.391		1.000	0.181
	33	PALM BEACH	0.864	0.000	2.645	0.613	0.327		1.000	0.232
	34	PINELLAS	0.378	0.000	0.860	0.119	0.440		1.000	0.138
	35	SAINT JOHNS	0.108	0.000	0.346	0.068	0.312		1.000	0.198
	36	SANTA ROSA	0.166	0.000	0.548	0.114	0.302		1.000	0.209
	37	SEMINOLE	0.282	0.000	0.570	0.069	0.495		1.000	0.122
	38	TAYLOR	0.077	0.000	0.233	0.039	0.332		1.000	0.169
	39	VOLUSIA	0.260	0.000	0.394	0.103	0.659		1.000	0.260
	40	WAKULLA	0.187	0.000	0.657	0.147	0.284		1.000	0.224

Form A-6: Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Coverage				Ratios Relative to Dominant Coverage			
			Coverage A	Coverage B	Coverage C	Coverage D	Coverage A	Coverage B	Coverage C	Coverage D
Commercial Residential	1	BAY	14.076	0.000	0.300	0.000	1.000		0.021	0.000
	2	BREVARD	8.433	0.000	0.164	0.000	1.000		0.020	0.000
	3	BREVARD	7.675	0.000	0.147	0.000	1.000		0.019	0.000
	4	BROWARD	15.566	0.000	0.310	0.000	1.000		0.020	0.000
	5	BROWARD	26.572	0.000	0.575	0.000	1.000		0.022	0.000
	6	CITRUS	4.265	0.000	0.077	0.000	1.000		0.018	0.000
	7	CLAY	1.720	0.000	0.030	0.000	1.000		0.017	0.000
	8	COLLIER	14.473	0.000	0.287	0.000	1.000		0.020	0.000
	9	COLUMBIA	1.528	0.000	0.026	0.000	1.000		0.017	0.000
	10	DIXIE	7.039	0.000	0.143	0.000	1.000		0.020	0.000
	11	DUVAL	5.105	0.000	0.102	0.000	1.000		0.020	0.000
	12	FRANKLIN	16.108	0.000	0.353	0.000	1.000		0.022	0.000
	13	GLADES	10.498	0.000	0.201	0.000	1.000		0.019	0.000
	14	HAMILTON	1.279	0.000	0.022	0.000	1.000		0.017	0.000
	15	HERNANDO	7.208	0.000	0.138	0.000	1.000		0.019	0.000
	16	HILLSBOROUGH	6.796	0.000	0.126	0.000	1.000		0.018	0.000
	17	HOLMES	4.178	0.000	0.077	0.000	1.000		0.018	0.000
	18	INDIAN RIVER	24.505	0.000	0.556	0.000	1.000		0.023	0.000
	19	JACKSON	2.632	0.000	0.046	0.000	1.000		0.017	0.000
	20	LEE	11.297	0.000	0.218	0.000	1.000		0.019	0.000
	21	LEON	2.249	0.000	0.040	0.000	1.000		0.018	0.000
	22	MARION	3.235	0.000	0.057	0.000	1.000		0.018	0.000
	23	MARTIN	11.052	0.000	0.211	0.000	1.000		0.019	0.000
	24	MARTIN	27.087	0.000	0.599	0.000	1.000		0.022	0.000
	25	MIAMI-DADE	15.567	0.000	0.311	0.000	1.000		0.020	0.000
	26	MIAMI-DADE	22.418	0.000	0.475	0.000	1.000		0.021	0.000
	27	MONROE	32.683	0.000	0.731	0.000	1.000		0.022	0.000
	28	MONROE	39.136	0.000	0.898	0.000	1.000		0.023	0.000
	29	OKALOOSA	11.717	0.000	0.242	0.000	1.000		0.021	0.000
	30	OSCEOLA	4.771	0.000	0.085	0.000	1.000		0.018	0.000
	31	OSCEOLA	6.992	0.000	0.131	0.000	1.000		0.019	0.000
	32	PALM BEACH	15.193	0.000	0.301	0.000	1.000		0.020	0.000
	33	PALM BEACH	23.895	0.000	0.505	0.000	1.000		0.021	0.000
	34	PINELLAS	7.765	0.000	0.148	0.000	1.000		0.019	0.000
	35	SAINT JOHNS	4.138	0.000	0.080	0.000	1.000		0.019	0.000
	36	SANTA ROSA	7.780	0.000	0.155	0.000	1.000		0.020	0.000
	37	SEMINOLE	3.877	0.000	0.068	0.000	1.000		0.017	0.000
	38	TAYLOR	2.254	0.000	0.041	0.000	1.000		0.018	0.000
	39	VOLUSIA	7.003	0.000	0.139	0.000	1.000		0.020	0.000
	40	WAKULLA	6.899	0.000	0.140	0.000	1.000		0.020	0.000

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
Frame Owners	1	BAY	4.682	2.941	2.015	1.000	0.628	0.430
	2	BREVARD	5.351	3.266	1.813	1.000	0.610	0.339
	3	BREVARD	5.152	3.171	1.784	1.000	0.615	0.346
	4	BROWARD	10.281	2.573	2.571	1.000	0.250	0.250
	5	BROWARD	17.945	3.749	3.756	1.000	0.209	0.209
	6	CITRUS	4.001	2.395	1.411	1.000	0.599	0.353
	7	CLAY	0.973	0.794	0.720	1.000	0.816	0.740
	8	COLLIER	8.089	4.787	2.556	1.000	0.592	0.316
	9	COLUMBIA	0.925	0.762	0.694	1.000	0.824	0.750
	10	DIXIE	3.202	2.149	1.589	1.000	0.671	0.496
	11	DUVAL	2.022	1.409	1.091	1.000	0.697	0.539
	12	FRANKLIN	7.044	4.257	2.618	1.000	0.604	0.372
	13	GLADES	7.255	4.141	2.303	1.000	0.571	0.317
	14	HAMILTON	0.817	0.673	0.616	1.000	0.824	0.754
	15	HERNANDO	5.360	3.251	1.819	1.000	0.606	0.339
	16	HILLSBOROUGH	5.088	2.976	1.704	1.000	0.585	0.335
	17	HOLMES	1.644	1.238	1.081	1.000	0.753	0.658
	18	INDIAN RIVER	15.737	9.125	4.447	1.000	0.580	0.283
	19	JACKSON	1.224	0.982	0.885	1.000	0.803	0.723
	20	LEE	7.808	4.378	2.357	1.000	0.561	0.302
	21	LEON	1.184	0.923	0.821	1.000	0.779	0.693
	22	MARION	3.294	2.007	1.198	1.000	0.609	0.364
	23	MARTIN	6.683	4.021	2.179	1.000	0.602	0.326
	24	MARTIN	15.325	8.685	3.993	1.000	0.567	0.261
	25	MIAMI-DADE	9.333	2.353	2.351	1.000	0.252	0.252
	26	MIAMI-DADE	13.934	2.965	2.967	1.000	0.213	0.213
	27	MONROE	14.103	7.376	3.726	1.000	0.523	0.264
	28	MONROE	21.957	12.092	6.583	1.000	0.551	0.300
	29	OKALOOSA	3.420	2.242	1.690	1.000	0.656	0.494
	30	OSCEOLA	4.338	2.609	1.528	1.000	0.601	0.352
	31	OSCEOLA	5.759	3.342	1.887	1.000	0.580	0.328
	32	PALM BEACH	8.691	5.064	2.605	1.000	0.583	0.300
	33	PALM BEACH	12.595	7.113	3.374	1.000	0.565	0.268
	34	PINELLAS	4.809	2.949	1.672	1.000	0.613	0.348
	35	SAINT JOHNS	1.531	1.139	0.950	1.000	0.744	0.620
	36	SANTA ROSA	2.425	1.684	1.354	1.000	0.694	0.558
	37	SEMINOLE	3.793	2.300	1.358	1.000	0.606	0.358
	38	TAYLOR	1.032	0.837	0.746	1.000	0.811	0.722
	39	VOLUSIA	4.504	2.735	1.516	1.000	0.607	0.337
	40	WAKULLA	2.804	1.923	1.476	1.000	0.686	0.527

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
Masonry Owners	1	BAY	4.373	2.804	1.982	1.000	0.641	0.453
	2	BREVARD	5.271	3.210	1.774	1.000	0.609	0.337
	3	BREVARD	5.079	3.118	1.746	1.000	0.614	0.344
	4	BROWARD	10.036	2.532	2.532	1.000	0.252	0.252
	5	BROWARD	17.261	3.514	3.525	1.000	0.204	0.204
	6	CITRUS	3.941	2.354	1.370	1.000	0.597	0.348
	7	CLAY	0.925	0.770	0.712	1.000	0.833	0.770
	8	COLLIER	7.963	4.692	2.477	1.000	0.589	0.311
	9	COLUMBIA	0.879	0.739	0.686	1.000	0.840	0.780
	10	DIXIE	2.992	2.052	1.559	1.000	0.686	0.521
	11	DUVAL	1.903	1.357	1.080	1.000	0.713	0.568
	12	FRANKLIN	6.498	4.002	2.535	1.000	0.616	0.390
	13	GLADES	7.127	4.048	2.229	1.000	0.568	0.313
	14	HAMILTON	0.776	0.655	0.610	1.000	0.843	0.785
	15	HERNANDO	5.293	3.204	1.773	1.000	0.605	0.335
	16	HILLSBOROUGH	5.028	2.927	1.645	1.000	0.582	0.327
	17	HOLMES	1.554	1.196	1.068	1.000	0.770	0.687
	18	INDIAN RIVER	14.840	8.675	4.179	1.000	0.585	0.282
	19	JACKSON	1.159	0.952	0.875	1.000	0.821	0.755
	20	LEE	7.625	4.240	2.305	1.000	0.556	0.302
	21	LEON	1.123	0.894	0.812	1.000	0.796	0.723
	22	MARION	3.242	1.968	1.164	1.000	0.607	0.359
	23	MARTIN	6.435	3.812	2.105	1.000	0.592	0.327
	24	MARTIN	14.539	8.251	3.797	1.000	0.568	0.261
	25	MIAMI-DADE	9.117	2.317	2.317	1.000	0.254	0.254
	26	MIAMI-DADE	13.512	2.882	2.886	1.000	0.213	0.214
	27	MONROE	13.409	7.034	3.386	1.000	0.525	0.252
	28	MONROE	20.292	11.104	5.433	1.000	0.547	0.268
	29	OKALOOSA	3.221	2.155	1.668	1.000	0.669	0.518
	30	OSCEOLA	4.262	2.553	1.485	1.000	0.599	0.348
	31	OSCEOLA	5.656	3.268	1.830	1.000	0.578	0.324
	32	PALM BEACH	8.379	4.812	2.513	1.000	0.574	0.300
	33	PALM BEACH	12.088	6.779	3.245	1.000	0.561	0.268
	34	PINELLAS	4.724	2.876	1.613	1.000	0.609	0.341
	35	SAINT JOHNS	1.448	1.100	0.939	1.000	0.760	0.649
	36	SANTA ROSA	2.289	1.622	1.337	1.000	0.708	0.584
	37	SEMINOLE	3.725	2.251	1.321	1.000	0.604	0.355
	38	TAYLOR	0.980	0.812	0.738	1.000	0.829	0.753
	39	VOLUSIA	4.446	2.690	1.477	1.000	0.605	0.332
	40	WAKULLA	2.636	1.848	1.458	1.000	0.701	0.553

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1974 Year Built		
			Year Built 1974	Year Built 1992	Year Built 2004	Year Built 1974	Year Built 1992	Year Built 2004
Mobile Homes	1	BAY	26.886	26.886	2.856	1.000	1.000	0.106
	2	BREVARD	15.919	15.919	2.021	1.000	1.000	0.127
	3	BREVARD	14.886	14.886	1.977	1.000	1.000	0.133
	4	BROWARD	29.392	29.392	2.853	1.000	1.000	0.097
	5	BROWARD	52.031	52.031	5.188	1.000	1.000	0.100
	6	CITRUS	9.268	9.268	1.515	1.000	1.000	0.164
	7	CLAY	4.072	4.072	0.847	1.000	1.000	0.208
	8	COLLIER	27.169	27.169	2.732	1.000	1.000	0.101
	9	COLUMBIA	3.818	3.818	0.812	1.000	1.000	0.213
	10	DIXIE	17.480	17.480	2.176	1.000	1.000	0.125
	11	DUVAL	10.818	10.818	1.426	1.000	1.000	0.132
	12	FRANKLIN	36.933	36.933	3.791	1.000	1.000	0.103
	13	GLADES	20.245	20.245	2.494	1.000	1.000	0.123
	14	HAMILTON	3.358	3.358	0.720	1.000	1.000	0.214
	15	HERNANDO	15.858	15.858	2.036	1.000	1.000	0.128
	16	HILLSBOROUGH	12.835	12.835	1.827	1.000	1.000	0.142
	17	HOLMES	7.840	7.840	1.339	1.000	1.000	0.171
	18	INDIAN RIVER	53.038	53.038	6.913	1.000	1.000	0.130
	19	JACKSON	5.318	5.318	1.064	1.000	1.000	0.200
	20	LEE	22.549	22.549	2.497	1.000	1.000	0.111
	21	LEON	5.326	5.326	0.989	1.000	1.000	0.186
	22	MARION	7.113	7.113	1.278	1.000	1.000	0.180
	23	MARTIN	20.223	20.223	2.354	1.000	1.000	0.116
	24	MARTIN	53.156	53.156	5.571	1.000	1.000	0.105
	25	MIAMI-DADE	26.895	26.895	2.589	1.000	1.000	0.096
	26	MIAMI-DADE	41.516	41.516	3.608	1.000	1.000	0.087
	27	MONROE	61.402	61.402	5.540	1.000	1.000	0.090
	28	MONROE	86.089	86.089	11.131	1.000	1.000	0.129
	29	OKALOOSA	21.162	21.162	2.261	1.000	1.000	0.107
	30	OSCEOLA	9.815	9.815	1.641	1.000	1.000	0.167
	31	OSCEOLA	14.953	14.953	2.039	1.000	1.000	0.136
	32	PALM BEACH	29.011	29.011	2.859	1.000	1.000	0.099
	33	PALM BEACH	44.852	44.852	4.076	1.000	1.000	0.091
	34	PINELLAS	13.946	13.946	1.768	1.000	1.000	0.127
	35	SAINT JOHNS	8.219	8.219	1.194	1.000	1.000	0.145
	36	SANTA ROSA	15.038	15.038	1.773	1.000	1.000	0.118
	37	SEMINOLE	8.246	8.246	1.456	1.000	1.000	0.177
	38	TAYLOR	5.023	5.023	0.901	1.000	1.000	0.179
	39	VOLUSIA	13.681	13.681	1.689	1.000	1.000	0.123
	40	WAKULLA	15.629	15.629	1.947	1.000	1.000	0.125

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
Frame Renters	1	BAY	1.667	0.870	0.498	1.000	0.522	0.299
	2	BREVARD	0.766	0.480	0.345	1.000	0.627	0.450
	3	BREVARD	0.694	0.451	0.337	1.000	0.650	0.486
	4	BROWARD	2.151	0.503	0.504	1.000	0.234	0.234
	5	BROWARD	5.846	0.916	0.920	1.000	0.157	0.157
	6	CITRUS	0.442	0.304	0.257	1.000	0.688	0.583
	7	CLAY	0.224	0.169	0.143	1.000	0.751	0.639
	8	COLLIER	1.240	0.729	0.498	1.000	0.588	0.402
	9	COLUMBIA	0.212	0.162	0.138	1.000	0.762	0.651
	10	DIXIE	1.077	0.608	0.386	1.000	0.565	0.359
	11	DUVAL	0.631	0.368	0.242	1.000	0.583	0.384
	12	FRANKLIN	2.864	1.446	0.729	1.000	0.505	0.254
	13	GLADES	0.969	0.564	0.432	1.000	0.582	0.446
	14	HAMILTON	0.187	0.143	0.123	1.000	0.765	0.656
	15	HERNANDO	0.749	0.469	0.348	1.000	0.625	0.464
	16	HILLSBOROUGH	0.605	0.390	0.315	1.000	0.644	0.521
	17	HOLMES	0.407	0.284	0.239	1.000	0.699	0.587
	18	INDIAN RIVER	5.267	2.781	1.360	1.000	0.528	0.258
	19	JACKSON	0.285	0.216	0.186	1.000	0.760	0.654
	20	LEE	1.096	0.613	0.457	1.000	0.560	0.417
	21	LEON	0.285	0.204	0.172	1.000	0.716	0.603
	22	MARION	0.351	0.254	0.216	1.000	0.724	0.615
	23	MARTIN	0.908	0.562	0.418	1.000	0.619	0.460
	24	MARTIN	4.798	2.431	1.104	1.000	0.507	0.230
	25	MIAMI-DADE	1.905	0.458	0.459	1.000	0.241	0.241
	26	MIAMI-DADE	3.946	0.613	0.615	1.000	0.155	0.156
	27	MONROE	4.236	2.141	1.037	1.000	0.506	0.245
	28	MONROE	8.332	4.667	2.729	1.000	0.560	0.328
	29	OKALOOSA	1.069	0.595	0.401	1.000	0.556	0.375
	30	OSCEOLA	0.468	0.330	0.280	1.000	0.706	0.598
	31	OSCEOLA	0.747	0.454	0.352	1.000	0.607	0.471
	32	PALM BEACH	1.429	0.790	0.511	1.000	0.553	0.358
	33	PALM BEACH	2.986	1.501	0.747	1.000	0.503	0.250
	34	PINELLAS	0.620	0.411	0.317	1.000	0.663	0.511
	35	SAINT JOHNS	0.423	0.274	0.207	1.000	0.647	0.489
	36	SANTA ROSA	0.688	0.419	0.316	1.000	0.610	0.460
	37	SEMINOLE	0.400	0.290	0.248	1.000	0.725	0.619
	38	TAYLOR	0.253	0.188	0.156	1.000	0.741	0.617
	39	VOLUSIA	0.687	0.419	0.288	1.000	0.610	0.419
	40	WAKULLA	0.883	0.511	0.340	1.000	0.578	0.386

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
Masonry Renters	1	BAY	1.546	0.809	0.459	1.000	0.524	0.297
	2	BREVARD	0.749	0.469	0.333	1.000	0.627	0.444
	3	BREVARD	0.679	0.440	0.326	1.000	0.648	0.480
	4	BROWARD	2.114	0.483	0.484	1.000	0.229	0.229
	5	BROWARD	5.585	0.741	0.748	1.000	0.133	0.134
	6	CITRUS	0.431	0.294	0.248	1.000	0.682	0.575
	7	CLAY	0.201	0.152	0.132	1.000	0.754	0.659
	8	COLLIER	1.215	0.714	0.482	1.000	0.588	0.396
	9	COLUMBIA	0.190	0.145	0.128	1.000	0.765	0.672
	10	DIXIE	0.987	0.558	0.351	1.000	0.566	0.355
	11	DUVAL	0.581	0.339	0.225	1.000	0.583	0.386
	12	FRANKLIN	2.607	1.325	0.639	1.000	0.508	0.245
	13	GLADES	0.951	0.551	0.417	1.000	0.579	0.439
	14	HAMILTON	0.167	0.129	0.113	1.000	0.770	0.678
	15	HERNANDO	0.730	0.456	0.336	1.000	0.625	0.460
	16	HILLSBOROUGH	0.594	0.379	0.304	1.000	0.638	0.512
	17	HOLMES	0.370	0.258	0.221	1.000	0.696	0.598
	18	INDIAN RIVER	4.760	2.600	1.226	1.000	0.546	0.258
	19	JACKSON	0.255	0.195	0.172	1.000	0.763	0.674
	20	LEE	1.076	0.600	0.442	1.000	0.558	0.411
	21	LEON	0.258	0.185	0.159	1.000	0.715	0.617
	22	MARION	0.343	0.246	0.208	1.000	0.717	0.607
	23	MARTIN	0.891	0.549	0.403	1.000	0.616	0.453
	24	MARTIN	4.381	2.263	0.998	1.000	0.517	0.228
	25	MIAMI-DADE	1.876	0.441	0.441	1.000	0.235	0.235
	26	MIAMI-DADE	3.832	0.566	0.568	1.000	0.148	0.148
	27	MONROE	3.926	1.973	0.813	1.000	0.502	0.207
	28	MONROE	7.310	3.949	1.763	1.000	0.540	0.241
	29	OKALOOSA	1.002	0.555	0.374	1.000	0.554	0.373
	30	OSCEOLA	0.458	0.320	0.270	1.000	0.699	0.589
	31	OSCEOLA	0.732	0.443	0.340	1.000	0.605	0.464
	32	PALM BEACH	1.399	0.775	0.495	1.000	0.554	0.354
	33	PALM BEACH	2.834	1.447	0.711	1.000	0.510	0.251
	34	PINELLAS	0.608	0.401	0.306	1.000	0.660	0.503
	35	SAINT JOHNS	0.389	0.251	0.192	1.000	0.646	0.494
	36	SANTA ROSA	0.640	0.387	0.294	1.000	0.605	0.460
	37	SEMINOLE	0.391	0.281	0.239	1.000	0.718	0.610
	38	TAYLOR	0.229	0.170	0.145	1.000	0.743	0.632
	39	VOLUSIA	0.670	0.409	0.278	1.000	0.610	0.415
	40	WAKULLA	0.816	0.472	0.316	1.000	0.579	0.387

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
Frame Condo Unit	1	BAY	2.035	1.103	0.657	1.000	0.542	0.323
	2	BREVARD	1.250	0.771	0.496	1.000	0.616	0.397
	3	BREVARD	1.162	0.733	0.487	1.000	0.631	0.419
	4	BROWARD	3.052	0.715	0.716	1.000	0.234	0.235
	5	BROWARD	7.319	1.216	1.220	1.000	0.166	0.167
	6	CITRUS	0.811	0.520	0.377	1.000	0.641	0.465
	7	CLAY	0.306	0.235	0.204	1.000	0.768	0.666
	8	COLLIER	1.968	1.152	0.710	1.000	0.585	0.361
	9	COLUMBIA	0.290	0.226	0.196	1.000	0.778	0.678
	10	DIXIE	1.331	0.780	0.513	1.000	0.586	0.386
	11	DUVAL	0.793	0.482	0.331	1.000	0.608	0.417
	12	FRANKLIN	3.403	1.777	0.932	1.000	0.522	0.274
	13	GLADES	1.630	0.934	0.625	1.000	0.573	0.383
	14	HAMILTON	0.256	0.200	0.174	1.000	0.781	0.682
	15	HERNANDO	1.235	0.758	0.500	1.000	0.613	0.405
	16	HILLSBOROUGH	1.073	0.657	0.459	1.000	0.612	0.428
	17	HOLMES	0.543	0.386	0.327	1.000	0.711	0.603
	18	INDIAN RIVER	6.549	3.526	1.709	1.000	0.538	0.261
	19	JACKSON	0.387	0.298	0.260	1.000	0.770	0.671
	20	LEE	1.805	1.004	0.653	1.000	0.556	0.362
	21	LEON	0.384	0.281	0.240	1.000	0.731	0.625
	22	MARION	0.656	0.435	0.318	1.000	0.663	0.485
	23	MARTIN	1.515	0.920	0.599	1.000	0.607	0.395
	24	MARTIN	6.062	3.149	1.420	1.000	0.519	0.234
	25	MIAMI-DADE	2.725	0.653	0.653	1.000	0.239	0.240
	26	MIAMI-DADE	5.118	0.855	0.857	1.000	0.167	0.167
	27	MONROE	5.404	2.742	1.328	1.000	0.507	0.246
	28	MONROE	10.070	5.602	3.211	1.000	0.556	0.319
	29	OKALOOSA	1.343	0.775	0.535	1.000	0.577	0.399
	30	OSCEOLA	0.869	0.566	0.409	1.000	0.651	0.471
	31	OSCEOLA	1.274	0.753	0.511	1.000	0.591	0.401
	32	PALM BEACH	2.206	1.237	0.726	1.000	0.561	0.329
	33	PALM BEACH	4.070	2.111	1.021	1.000	0.519	0.251
	34	PINELLAS	1.059	0.674	0.457	1.000	0.637	0.432
	35	SAINT JOHNS	0.548	0.367	0.285	1.000	0.669	0.519
	36	SANTA ROSA	0.884	0.555	0.425	1.000	0.628	0.480
	37	SEMINOLE	0.751	0.497	0.363	1.000	0.662	0.483
	38	TAYLOR	0.339	0.257	0.218	1.000	0.759	0.643
	39	VOLUSIA	1.093	0.661	0.415	1.000	0.605	0.380
	40	WAKULLA	1.108	0.666	0.459	1.000	0.601	0.415

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
Masonry Condo Unit	1	BAY	1.888	1.032	0.617	1.000	0.547	0.327
	2	BREVARD	1.226	0.755	0.481	1.000	0.615	0.392
	3	BREVARD	1.141	0.718	0.472	1.000	0.629	0.414
	4	BROWARD	2.992	0.692	0.693	1.000	0.231	0.232
	5	BROWARD	7.003	1.026	1.034	1.000	0.147	0.148
	6	CITRUS	0.795	0.506	0.364	1.000	0.637	0.458
	7	CLAY	0.279	0.217	0.193	1.000	0.776	0.690
	8	COLLIER	1.931	1.129	0.686	1.000	0.584	0.355
	9	COLUMBIA	0.264	0.208	0.186	1.000	0.786	0.703
	10	DIXIE	1.224	0.723	0.476	1.000	0.591	0.389
	11	DUVAL	0.734	0.449	0.313	1.000	0.612	0.426
	12	FRANKLIN	3.105	1.637	0.839	1.000	0.527	0.270
	13	GLADES	1.600	0.912	0.604	1.000	0.570	0.377
	14	HAMILTON	0.233	0.184	0.165	1.000	0.790	0.709
	15	HERNANDO	1.211	0.741	0.484	1.000	0.612	0.400
	16	HILLSBOROUGH	1.056	0.642	0.443	1.000	0.608	0.419
	17	HOLMES	0.499	0.357	0.310	1.000	0.715	0.620
	18	INDIAN RIVER	5.978	3.310	1.554	1.000	0.554	0.260
	19	JACKSON	0.353	0.275	0.245	1.000	0.778	0.695
	20	LEE	1.768	0.977	0.633	1.000	0.553	0.358
	21	LEON	0.352	0.259	0.227	1.000	0.737	0.645
	22	MARION	0.643	0.423	0.307	1.000	0.658	0.478
	23	MARTIN	1.475	0.887	0.578	1.000	0.601	0.392
	24	MARTIN	5.587	2.946	1.299	1.000	0.527	0.232
	25	MIAMI-DADE	2.675	0.632	0.633	1.000	0.236	0.237
	26	MIAMI-DADE	4.967	0.802	0.804	1.000	0.161	0.162
	27	MONROE	5.041	2.547	1.081	1.000	0.505	0.214
	28	MONROE	8.933	4.821	2.178	1.000	0.540	0.244
	29	OKALOOSA	1.259	0.728	0.507	1.000	0.578	0.403
	30	OSCEOLA	0.852	0.550	0.395	1.000	0.646	0.464
	31	OSCEOLA	1.249	0.735	0.494	1.000	0.589	0.395
	32	PALM BEACH	2.147	1.198	0.701	1.000	0.558	0.327
	33	PALM BEACH	3.875	2.026	0.973	1.000	0.523	0.251
	34	PINELLAS	1.038	0.657	0.441	1.000	0.633	0.424
	35	SAINT JOHNS	0.507	0.342	0.269	1.000	0.673	0.531
	36	SANTA ROSA	0.826	0.519	0.402	1.000	0.628	0.487
	37	SEMINOLE	0.736	0.484	0.351	1.000	0.657	0.477
	38	TAYLOR	0.310	0.238	0.206	1.000	0.766	0.664
	39	VOLUSIA	1.070	0.647	0.401	1.000	0.604	0.375
	40	WAKULLA	1.028	0.622	0.434	1.000	0.605	0.423

Form A-6: Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost per Year Built			Ratios Relative to 1980 Year Built		
			Year Built 1980	Year Built 1998	Year Built 2004	Year Built 1980	Year Built 1998	Year Built 2004
Commercial Residential	1	BAY	14.376	14.376	11.959	1.000	1.000	0.832
	2	BREVARD	8.598	8.598	6.715	1.000	1.000	0.781
	3	BREVARD	7.822	7.822	5.970	1.000	1.000	0.763
	4	BROWARD	15.876	12.878	12.878	1.000	0.811	0.811
	5	BROWARD	27.147	24.078	24.078	1.000	0.887	0.887
	6	CITRUS	4.341	4.341	3.052	1.000	1.000	0.703
	7	CLAY	1.750	1.750	1.149	1.000	1.000	0.657
	8	COLLIER	14.760	14.760	11.704	1.000	1.000	0.793
	9	COLUMBIA	1.554	1.554	1.000	1.000	1.000	0.644
	10	DIXIE	7.182	7.182	5.579	1.000	1.000	0.777
	11	DUVAL	5.207	5.207	4.085	1.000	1.000	0.785
	12	FRANKLIN	16.460	16.460	13.946	1.000	1.000	0.847
	13	GLADES	10.698	10.698	8.137	1.000	1.000	0.761
	14	HAMILTON	1.300	1.300	0.828	1.000	1.000	0.637
	15	HERNANDO	7.346	7.346	5.537	1.000	1.000	0.754
	16	HILLSBOROUGH	6.921	6.921	5.087	1.000	1.000	0.735
	17	HOLMES	4.254	4.254	3.016	1.000	1.000	0.709
	18	INDIAN RIVER	25.061	25.061	22.759	1.000	1.000	0.908
	19	JACKSON	2.678	2.678	1.781	1.000	1.000	0.665
	20	LEE	11.515	11.515	8.803	1.000	1.000	0.765
	21	LEON	2.289	2.289	1.546	1.000	1.000	0.676
	22	MARION	3.291	3.291	2.246	1.000	1.000	0.682
	23	MARTIN	11.263	11.263	8.767	1.000	1.000	0.778
	24	MARTIN	27.686	27.686	25.100	1.000	1.000	0.907
	25	MIAMI-DADE	15.877	13.018	13.018	1.000	0.820	0.820
	26	MIAMI-DADE	22.893	19.934	19.934	1.000	0.871	0.871
	27	MONROE	33.413	33.413	31.116	1.000	1.000	0.931
	28	MONROE	40.034	40.034	38.029	1.000	1.000	0.950
	29	OKALOOSA	11.959	11.959	9.665	1.000	1.000	0.808
	30	OSCEOLA	4.856	4.856	3.388	1.000	1.000	0.698
	31	OSCEOLA	7.123	7.123	5.258	1.000	1.000	0.738
	32	PALM BEACH	15.494	15.494	12.492	1.000	1.000	0.806
	33	PALM BEACH	24.399	24.399	21.141	1.000	1.000	0.866
	34	PINELLAS	7.913	7.913	6.050	1.000	1.000	0.765
	35	SAINT JOHNS	4.217	4.217	3.179	1.000	1.000	0.754
	36	SANTA ROSA	7.934	7.934	6.096	1.000	1.000	0.768
	37	SEMINOLE	3.945	3.945	2.703	1.000	1.000	0.685
	38	TAYLOR	2.295	2.295	1.591	1.000	1.000	0.693
	39	VOLUSIA	7.142	7.142	5.615	1.000	1.000	0.786
	40	WAKULLA	7.038	7.038	5.491	1.000	1.000	0.780

Form A-6: Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
Frame Owners	1	BAY	6.153	2.941	1.995	1.000	0.478	0.324
	2	BREVARD	6.660	3.266	1.817	1.000	0.490	0.273
	3	BREVARD	6.374	3.171	1.791	1.000	0.497	0.281
	4	BROWARD	11.034	2.573	2.564	1.000	0.233	0.232
	5	BROWARD	19.111	3.749	3.550	1.000	0.196	0.186
	6	CITRUS	4.618	2.395	1.419	1.000	0.519	0.307
	7	CLAY	1.064	0.794	0.720	1.000	0.746	0.676
	8	COLLIER	10.265	4.787	2.557	1.000	0.466	0.249
	9	COLUMBIA	1.009	0.762	0.693	1.000	0.755	0.687
	10	DIXIE	4.090	2.149	1.573	1.000	0.525	0.385
	11	DUVAL	2.543	1.409	1.091	1.000	0.554	0.429
	12	FRANKLIN	9.236	4.257	2.547	1.000	0.461	0.276
	13	GLADES	8.490	4.141	2.307	1.000	0.488	0.272
	14	HAMILTON	0.890	0.673	0.616	1.000	0.756	0.692
	15	HERNANDO	6.679	3.251	1.822	1.000	0.487	0.273
	16	HILLSBOROUGH	5.869	2.976	1.702	1.000	0.507	0.290
	17	HOLMES	1.851	1.238	1.081	1.000	0.669	0.584
	18	INDIAN RIVER	20.268	9.125	4.025	1.000	0.450	0.199
	19	JACKSON	1.346	0.982	0.886	1.000	0.730	0.658
	20	LEE	9.172	4.378	2.409	1.000	0.477	0.263
	21	LEON	1.318	0.923	0.821	1.000	0.700	0.623
	22	MARION	3.759	2.007	1.202	1.000	0.534	0.320
	23	MARTIN	8.350	4.021	2.198	1.000	0.482	0.263
	24	MARTIN	19.614	8.685	3.737	1.000	0.443	0.191
	25	MIAMI-DADE	10.015	2.353	2.348	1.000	0.235	0.234
	26	MIAMI-DADE	14.817	2.965	2.896	1.000	0.200	0.195
	27	MONROE	21.595	7.376	3.642	1.000	0.342	0.169
	28	MONROE	32.451	12.092	5.986	1.000	0.373	0.184
	29	OKALOOSA	4.490	2.242	1.684	1.000	0.499	0.375
	30	OSCEOLA	4.966	2.609	1.538	1.000	0.525	0.310
	31	OSCEOLA	6.692	3.342	1.889	1.000	0.499	0.282
	32	PALM BEACH	11.060	5.064	2.614	1.000	0.458	0.236
	33	PALM BEACH	16.244	7.113	3.295	1.000	0.438	0.203
	34	PINELLAS	5.956	2.949	1.674	1.000	0.495	0.281
	35	SAINT JOHNS	1.883	1.139	0.949	1.000	0.605	0.504
	36	SANTA ROSA	3.118	1.684	1.354	1.000	0.540	0.434
	37	SEMINOLE	4.328	2.300	1.368	1.000	0.531	0.316
	38	TAYLOR	1.208	0.837	0.746	1.000	0.693	0.618
	39	VOLUSIA	5.626	2.735	1.506	1.000	0.486	0.268
	40	WAKULLA	3.569	1.923	1.470	1.000	0.539	0.412

Form A-6: Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
Masonry Owners	1	BAY	5.724	2.804	1.962	1.000	0.490	0.343
	2	BREVARD	6.584	3.210	1.761	1.000	0.488	0.267
	3	BREVARD	6.310	3.118	1.737	1.000	0.494	0.275
	4	BROWARD	10.739	2.532	2.509	1.000	0.236	0.234
	5	BROWARD	18.196	3.514	3.381	1.000	0.193	0.186
	6	CITRUS	4.566	2.354	1.375	1.000	0.516	0.301
	7	CLAY	1.008	0.770	0.712	1.000	0.764	0.706
	8	COLLIER	10.140	4.692	2.468	1.000	0.463	0.243
	9	COLUMBIA	0.955	0.739	0.686	1.000	0.774	0.718
	10	DIXIE	3.806	2.052	1.547	1.000	0.539	0.406
	11	DUVAL	2.381	1.357	1.077	1.000	0.570	0.452
	12	FRANKLIN	8.490	4.002	2.487	1.000	0.471	0.293
	13	GLADES	8.383	4.048	2.225	1.000	0.483	0.265
	14	HAMILTON	0.843	0.655	0.609	1.000	0.777	0.723
	15	HERNANDO	6.602	3.204	1.764	1.000	0.485	0.267
	16	HILLSBOROUGH	5.803	2.927	1.643	1.000	0.504	0.283
	17	HOLMES	1.745	1.196	1.067	1.000	0.685	0.611
	18	INDIAN RIVER	18.990	8.675	3.813	1.000	0.457	0.201
	19	JACKSON	1.271	0.952	0.875	1.000	0.749	0.688
	20	LEE	9.051	4.240	2.320	1.000	0.468	0.256
	21	LEON	1.245	0.894	0.811	1.000	0.718	0.651
	22	MARION	3.718	1.968	1.167	1.000	0.529	0.314
	23	MARTIN	8.152	3.812	2.156	1.000	0.468	0.264
	24	MARTIN	18.536	8.251	3.604	1.000	0.445	0.194
	25	MIAMI-DADE	9.758	2.317	2.299	1.000	0.237	0.236
	26	MIAMI-DADE	14.284	2.882	2.818	1.000	0.202	0.197
	27	MONROE	20.415	7.034	3.516	1.000	0.345	0.172
	28	MONROE	29.979	11.104	5.636	1.000	0.370	0.188
	29	OKALOOSA	4.215	2.155	1.660	1.000	0.511	0.394
	30	OSCEOLA	4.913	2.553	1.490	1.000	0.520	0.303
	31	OSCEOLA	6.608	3.268	1.826	1.000	0.495	0.276
	32	PALM BEACH	10.772	4.812	2.561	1.000	0.447	0.238
	33	PALM BEACH	15.629	6.779	3.212	1.000	0.434	0.206
	34	PINELLAS	5.903	2.876	1.621	1.000	0.487	0.275
	35	SAINT JOHNS	1.773	1.100	0.937	1.000	0.620	0.529
	36	SANTA ROSA	2.937	1.622	1.335	1.000	0.552	0.454
	37	SEMINOLE	4.281	2.251	1.326	1.000	0.526	0.310
	38	TAYLOR	1.141	0.812	0.738	1.000	0.712	0.647
	39	VOLUSIA	5.551	2.690	1.460	1.000	0.485	0.263
	40	WAKULLA	3.339	1.848	1.450	1.000	0.553	0.434

Form A-6: Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
Mobile Homes	1	BAY	26.886	26.886	2.856	1.000	1.000	0.106
	2	BREVARD	15.919	15.919	2.021	1.000	1.000	0.127
	3	BREVARD	14.886	14.886	1.977	1.000	1.000	0.133
	4	BROWARD	29.392	29.392	2.853	1.000	1.000	0.097
	5	BROWARD	52.031	52.031	5.188	1.000	1.000	0.100
	6	CITRUS	9.268	9.268	1.515	1.000	1.000	0.164
	7	CLAY	4.072	4.072	0.847	1.000	1.000	0.208
	8	COLLIER	27.169	27.169	2.732	1.000	1.000	0.101
	9	COLUMBIA	3.818	3.818	0.812	1.000	1.000	0.213
	10	DIXIE	17.480	17.480	2.176	1.000	1.000	0.125
	11	DUVAL	10.818	10.818	1.426	1.000	1.000	0.132
	12	FRANKLIN	36.933	36.933	3.791	1.000	1.000	0.103
	13	GLADES	20.245	20.245	2.494	1.000	1.000	0.123
	14	HAMILTON	3.358	3.358	0.720	1.000	1.000	0.214
	15	HERNANDO	15.858	15.858	2.036	1.000	1.000	0.128
	16	HILLSBOROUGH	12.835	12.835	1.827	1.000	1.000	0.142
	17	HOLMES	7.840	7.840	1.339	1.000	1.000	0.171
	18	INDIAN RIVER	53.038	53.038	6.913	1.000	1.000	0.130
	19	JACKSON	5.318	5.318	1.064	1.000	1.000	0.200
	20	LEE	22.549	22.549	2.497	1.000	1.000	0.111
	21	LEON	5.326	5.326	0.989	1.000	1.000	0.186
	22	MARION	7.113	7.113	1.278	1.000	1.000	0.180
	23	MARTIN	20.223	20.223	2.354	1.000	1.000	0.116
	24	MARTIN	53.156	53.156	5.571	1.000	1.000	0.105
	25	MIAMI-DADE	26.895	26.895	2.589	1.000	1.000	0.096
	26	MIAMI-DADE	41.516	41.516	3.608	1.000	1.000	0.087
	27	MONROE	61.402	61.402	5.540	1.000	1.000	0.090
	28	MONROE	86.089	86.089	11.131	1.000	1.000	0.129
	29	OKALOOSA	21.162	21.162	2.261	1.000	1.000	0.107
	30	OSCEOLA	9.815	9.815	1.641	1.000	1.000	0.167
	31	OSCEOLA	14.953	14.953	2.039	1.000	1.000	0.136
	32	PALM BEACH	29.011	29.011	2.859	1.000	1.000	0.099
	33	PALM BEACH	44.852	44.852	4.076	1.000	1.000	0.091
	34	PINELLAS	13.946	13.946	1.768	1.000	1.000	0.127
	35	SAINT JOHNS	8.219	8.219	1.194	1.000	1.000	0.145
	36	SANTA ROSA	15.038	15.038	1.773	1.000	1.000	0.118
	37	SEMINOLE	8.246	8.246	1.456	1.000	1.000	0.177
	38	TAYLOR	5.023	5.023	0.901	1.000	1.000	0.179
	39	VOLUSIA	13.681	13.681	1.689	1.000	1.000	0.123
	40	WAKULLA	15.629	15.629	1.947	1.000	1.000	0.125

Form A-6: Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
Frame Renters	1	BAY	2.355	0.870	0.481	1.000	0.369	0.204
	2	BREVARD	1.072	0.480	0.339	1.000	0.448	0.316
	3	BREVARD	0.955	0.451	0.333	1.000	0.472	0.349
	4	BROWARD	2.125	0.503	0.493	1.000	0.237	0.232
	5	BROWARD	5.992	0.916	0.768	1.000	0.153	0.128
	6	CITRUS	0.536	0.304	0.257	1.000	0.568	0.479
	7	CLAY	0.248	0.169	0.143	1.000	0.679	0.577
	8	COLLIER	1.811	0.729	0.489	1.000	0.402	0.270
	9	COLUMBIA	0.233	0.162	0.138	1.000	0.693	0.591
	10	DIXIE	1.476	0.608	0.371	1.000	0.412	0.251
	11	DUVAL	0.856	0.368	0.240	1.000	0.430	0.280
	12	FRANKLIN	4.015	1.446	0.662	1.000	0.360	0.165
	13	GLADES	1.215	0.564	0.426	1.000	0.464	0.351
	14	HAMILTON	0.206	0.143	0.123	1.000	0.696	0.596
	15	HERNANDO	1.058	0.469	0.344	1.000	0.443	0.325
	16	HILLSBOROUGH	0.724	0.390	0.313	1.000	0.538	0.432
	17	HOLMES	0.464	0.284	0.238	1.000	0.613	0.513
	18	INDIAN RIVER	7.511	2.781	1.069	1.000	0.370	0.142
	19	JACKSON	0.314	0.216	0.186	1.000	0.689	0.592
	20	LEE	1.390	0.613	0.449	1.000	0.441	0.323
	21	LEON	0.322	0.204	0.172	1.000	0.634	0.532
	22	MARION	0.404	0.254	0.215	1.000	0.629	0.533
	23	MARTIN	1.265	0.562	0.414	1.000	0.444	0.327
	24	MARTIN	6.752	2.431	0.900	1.000	0.360	0.133
	25	MIAMI-DADE	1.876	0.458	0.451	1.000	0.244	0.240
	26	MIAMI-DADE	3.962	0.613	0.569	1.000	0.155	0.144
	27	MONROE	7.928	2.141	0.856	1.000	0.270	0.108
	28	MONROE	14.363	4.667	1.911	1.000	0.325	0.133
	29	OKALOOSA	1.513	0.595	0.396	1.000	0.393	0.262
	30	OSCEOLA	0.545	0.330	0.279	1.000	0.606	0.511
	31	OSCEOLA	0.920	0.454	0.347	1.000	0.493	0.377
	32	PALM BEACH	2.075	0.790	0.500	1.000	0.381	0.241
	33	PALM BEACH	4.342	1.501	0.678	1.000	0.346	0.156
	34	PINELLAS	0.849	0.411	0.314	1.000	0.484	0.370
	35	SAINT JOHNS	0.555	0.274	0.206	1.000	0.493	0.370
	36	SANTA ROSA	0.945	0.419	0.314	1.000	0.444	0.333
	37	SEMINOLE	0.460	0.290	0.247	1.000	0.630	0.536
	38	TAYLOR	0.308	0.188	0.156	1.000	0.610	0.506
	39	VOLUSIA	0.971	0.419	0.280	1.000	0.431	0.289
	40	WAKULLA	1.209	0.511	0.335	1.000	0.422	0.277

Form A-6: Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
Masonry Renters	1	BAY	2.171	0.809	0.450	1.000	0.373	0.207
	2	BREVARD	1.039	0.469	0.327	1.000	0.452	0.315
	3	BREVARD	0.928	0.440	0.322	1.000	0.474	0.346
	4	BROWARD	2.048	0.483	0.476	1.000	0.236	0.232
	5	BROWARD	5.483	0.741	0.681	1.000	0.135	0.124
	6	CITRUS	0.523	0.294	0.247	1.000	0.563	0.473
	7	CLAY	0.224	0.152	0.132	1.000	0.677	0.591
	8	COLLIER	1.756	0.714	0.473	1.000	0.407	0.269
	9	COLUMBIA	0.210	0.145	0.127	1.000	0.690	0.606
	10	DIXIE	1.350	0.558	0.345	1.000	0.414	0.256
	11	DUVAL	0.788	0.339	0.223	1.000	0.430	0.283
	12	FRANKLIN	3.631	1.325	0.617	1.000	0.365	0.170
	13	GLADES	1.186	0.551	0.412	1.000	0.464	0.347
	14	HAMILTON	0.185	0.129	0.113	1.000	0.696	0.612
	15	HERNANDO	1.023	0.456	0.333	1.000	0.446	0.325
	16	HILLSBOROUGH	0.710	0.379	0.302	1.000	0.534	0.425
	17	HOLMES	0.425	0.258	0.220	1.000	0.606	0.519
	18	INDIAN RIVER	6.653	2.600	1.000	1.000	0.391	0.150
	19	JACKSON	0.284	0.195	0.172	1.000	0.686	0.605
	20	LEE	1.354	0.600	0.434	1.000	0.443	0.320
	21	LEON	0.293	0.185	0.159	1.000	0.629	0.541
	22	MARION	0.396	0.246	0.208	1.000	0.622	0.525
	23	MARTIN	1.235	0.549	0.400	1.000	0.444	0.324
	24	MARTIN	6.102	2.263	0.855	1.000	0.371	0.140
	25	MIAMI-DADE	1.815	0.441	0.435	1.000	0.243	0.240
	26	MIAMI-DADE	3.734	0.566	0.543	1.000	0.152	0.145
	27	MONROE	7.227	1.973	0.820	1.000	0.273	0.114
	28	MONROE	12.619	3.949	1.757	1.000	0.313	0.139
	29	OKALOOSA	1.415	0.555	0.370	1.000	0.392	0.261
	30	OSCEOLA	0.535	0.320	0.269	1.000	0.598	0.502
	31	OSCEOLA	0.898	0.443	0.335	1.000	0.493	0.373
	32	PALM BEACH	2.009	0.775	0.484	1.000	0.386	0.241
	33	PALM BEACH	4.066	1.447	0.655	1.000	0.356	0.161
	34	PINELLAS	0.828	0.401	0.303	1.000	0.484	0.366
	35	SAINT JOHNS	0.513	0.251	0.191	1.000	0.489	0.372
	36	SANTA ROSA	0.885	0.387	0.293	1.000	0.438	0.331
	37	SEMINOLE	0.451	0.281	0.238	1.000	0.623	0.528
	38	TAYLOR	0.281	0.170	0.144	1.000	0.605	0.513
	39	VOLUSIA	0.937	0.409	0.271	1.000	0.436	0.289
	40	WAKULLA	1.116	0.472	0.312	1.000	0.423	0.280

Form A-6: Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
Frame Condo Unit	1	BAY	4.870	1.103	0.736	1.000	0.227	0.151
	2	BREVARD	2.588	0.771	0.507	1.000	0.298	0.196
	3	BREVARD	2.326	0.733	0.495	1.000	0.315	0.213
	4	BROWARD	5.144	0.715	0.732	1.000	0.139	0.142
	5	BROWARD	12.259	1.216	1.384	1.000	0.099	0.113
	6	CITRUS	1.307	0.520	0.380	1.000	0.398	0.291
	7	CLAY	0.431	0.235	0.204	1.000	0.545	0.474
	8	COLLIER	4.505	1.152	0.730	1.000	0.256	0.162
	9	COLUMBIA	0.399	0.226	0.197	1.000	0.566	0.493
	10	DIXIE	2.967	0.780	0.561	1.000	0.263	0.189
	11	DUVAL	1.719	0.482	0.353	1.000	0.280	0.206
	12	FRANKLIN	7.881	1.777	1.137	1.000	0.225	0.144
	13	GLADES	3.088	0.934	0.630	1.000	0.302	0.204
	14	HAMILTON	0.349	0.200	0.175	1.000	0.572	0.500
	15	HERNANDO	2.547	0.758	0.516	1.000	0.298	0.203
	16	HILLSBOROUGH	1.830	0.657	0.460	1.000	0.359	0.251
	17	HOLMES	0.878	0.386	0.328	1.000	0.440	0.374
	18	INDIAN RIVER	14.316	3.526	2.205	1.000	0.246	0.154
	19	JACKSON	0.547	0.298	0.260	1.000	0.545	0.475
	20	LEE	3.531	1.004	0.665	1.000	0.284	0.188
	21	LEON	0.591	0.281	0.241	1.000	0.475	0.407
	22	MARION	0.970	0.435	0.319	1.000	0.449	0.329
	23	MARTIN	3.169	0.920	0.611	1.000	0.290	0.193
	24	MARTIN	13.373	3.149	1.837	1.000	0.235	0.137
	25	MIAMI-DADE	4.612	0.653	0.666	1.000	0.141	0.144
	26	MIAMI-DADE	8.756	0.855	0.915	1.000	0.098	0.105
	27	MONROE	15.814	2.742	1.821	1.000	0.173	0.115
	28	MONROE	26.093	5.602	4.070	1.000	0.215	0.156
	29	OKALOOSA	3.296	0.775	0.557	1.000	0.235	0.169
	30	OSCEOLA	1.339	0.566	0.411	1.000	0.422	0.307
	31	OSCEOLA	2.280	0.753	0.515	1.000	0.330	0.226
	32	PALM BEACH	5.053	1.237	0.758	1.000	0.245	0.150
	33	PALM BEACH	9.501	2.111	1.188	1.000	0.222	0.125
	34	PINELLAS	2.104	0.674	0.462	1.000	0.320	0.220
	35	SAINT JOHNS	1.120	0.367	0.291	1.000	0.328	0.260
	36	SANTA ROSA	2.067	0.555	0.431	1.000	0.269	0.209
	37	SEMINOLE	1.113	0.497	0.364	1.000	0.447	0.327
	38	TAYLOR	0.574	0.257	0.219	1.000	0.448	0.381
	39	VOLUSIA	2.309	0.661	0.425	1.000	0.286	0.184
	40	WAKULLA	2.471	0.666	0.486	1.000	0.269	0.197

Form A-6: Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
Masonry Condo Unit	1	BAY	4.472	1.032	0.696	1.000	0.231	0.156
	2	BREVARD	2.510	0.755	0.491	1.000	0.301	0.195
	3	BREVARD	2.270	0.718	0.479	1.000	0.316	0.211
	4	BROWARD	4.935	0.692	0.707	1.000	0.140	0.143
	5	BROWARD	11.138	1.026	1.139	1.000	0.092	0.102
	6	CITRUS	1.291	0.506	0.366	1.000	0.392	0.284
	7	CLAY	0.404	0.217	0.193	1.000	0.536	0.477
	8	COLLIER	4.373	1.129	0.705	1.000	0.258	0.161
	9	COLUMBIA	0.373	0.208	0.186	1.000	0.556	0.497
	10	DIXIE	2.726	0.723	0.529	1.000	0.265	0.194
	11	DUVAL	1.589	0.449	0.335	1.000	0.283	0.211
	12	FRANKLIN	7.080	1.637	1.064	1.000	0.231	0.150
	13	GLADES	3.034	0.912	0.608	1.000	0.301	0.200
	14	HAMILTON	0.326	0.184	0.165	1.000	0.563	0.505
	15	HERNANDO	2.472	0.741	0.498	1.000	0.300	0.202
	16	HILLSBOROUGH	1.813	0.642	0.444	1.000	0.354	0.245
	17	HOLMES	0.833	0.357	0.310	1.000	0.429	0.373
	18	INDIAN RIVER	12.659	3.310	2.036	1.000	0.261	0.161
	19	JACKSON	0.516	0.275	0.246	1.000	0.533	0.476
	20	LEE	3.458	0.977	0.641	1.000	0.283	0.185
	21	LEON	0.558	0.259	0.227	1.000	0.465	0.408
	22	MARION	0.961	0.423	0.308	1.000	0.441	0.320
	23	MARTIN	3.089	0.887	0.593	1.000	0.287	0.192
	24	MARTIN	11.983	2.946	1.717	1.000	0.246	0.143
	25	MIAMI-DADE	4.444	0.632	0.644	1.000	0.142	0.145
	26	MIAMI-DADE	8.152	0.802	0.849	1.000	0.098	0.104
	27	MONROE	14.196	2.547	1.715	1.000	0.179	0.121
	28	MONROE	22.887	4.821	3.726	1.000	0.211	0.163
	29	OKALOOSA	3.096	0.728	0.529	1.000	0.235	0.171
	30	OSCEOLA	1.329	0.550	0.397	1.000	0.414	0.298
	31	OSCEOLA	2.234	0.735	0.497	1.000	0.329	0.223
	32	PALM BEACH	4.855	1.198	0.736	1.000	0.247	0.152
	33	PALM BEACH	8.812	2.026	1.139	1.000	0.230	0.129
	34	PINELLAS	2.066	0.657	0.447	1.000	0.318	0.216
	35	SAINT JOHNS	1.050	0.342	0.276	1.000	0.325	0.263
	36	SANTA ROSA	1.966	0.519	0.409	1.000	0.264	0.208
	37	SEMINOLE	1.105	0.484	0.352	1.000	0.438	0.318
	38	TAYLOR	0.541	0.238	0.207	1.000	0.439	0.382
	39	VOLUSIA	2.228	0.647	0.411	1.000	0.290	0.185
	40	WAKULLA	2.293	0.622	0.461	1.000	0.271	0.201

Form A-6: Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County	Loss Cost by Building Strength			Ratio Relative to Weak		
			Weak	Medium	Strong	Weak	Medium	Strong
Commercial Residential	1	BAY	15.132	14.376	8.052	1.000	0.950	0.532
	2	BREVARD	9.126	8.598	4.185	1.000	0.942	0.459
	3	BREVARD	8.327	7.822	3.617	1.000	0.939	0.434
	4	BROWARD	16.771	12.878	8.258	1.000	0.768	0.492
	5	BROWARD	28.302	24.078	17.667	1.000	0.851	0.624
	6	CITRUS	4.666	4.341	1.717	1.000	0.930	0.368
	7	CLAY	1.894	1.750	0.630	1.000	0.924	0.333
	8	COLLIER	15.641	14.760	7.236	1.000	0.944	0.463
	9	COLUMBIA	1.685	1.554	0.544	1.000	0.922	0.323
	10	DIXIE	7.628	7.182	3.564	1.000	0.941	0.467
	11	DUVAL	5.523	5.207	2.637	1.000	0.943	0.477
	12	FRANKLIN	17.280	16.460	9.756	1.000	0.953	0.565
	13	GLADES	11.397	10.698	4.806	1.000	0.939	0.422
	14	HAMILTON	1.412	1.300	0.449	1.000	0.921	0.318
	15	HERNANDO	7.833	7.346	3.333	1.000	0.938	0.426
	16	HILLSBOROUGH	7.403	6.921	2.904	1.000	0.935	0.392
	17	HOLMES	4.569	4.254	1.685	1.000	0.931	0.369
	18	INDIAN RIVER	26.027	25.061	17.456	1.000	0.963	0.671
	19	JACKSON	2.895	2.678	0.952	1.000	0.925	0.329
	20	LEE	12.259	11.515	5.233	1.000	0.939	0.427
	21	LEON	2.471	2.289	0.856	1.000	0.926	0.347
	22	MARION	3.549	3.291	1.234	1.000	0.927	0.348
	23	MARTIN	11.962	11.263	5.349	1.000	0.942	0.447
	24	MARTIN	28.763	27.686	19.012	1.000	0.963	0.661
	25	MIAMI-DADE	16.748	13.018	8.386	1.000	0.777	0.501
	26	MIAMI-DADE	23.939	19.934	14.110	1.000	0.833	0.589
	27	MONROE	34.569	33.413	24.136	1.000	0.967	0.698
	28	MONROE	41.275	40.034	30.479	1.000	0.970	0.738
	29	OKALOOSA	12.639	11.959	6.158	1.000	0.946	0.487
	30	OSCEOLA	5.224	4.856	1.868	1.000	0.929	0.358
	31	OSCEOLA	7.614	7.123	3.077	1.000	0.935	0.404
	32	PALM BEACH	16.380	15.494	7.966	1.000	0.946	0.486
	33	PALM BEACH	25.531	24.399	14.927	1.000	0.956	0.585
	34	PINELLAS	8.423	7.913	3.624	1.000	0.939	0.430
	35	SAINT JOHNS	4.496	4.217	1.932	1.000	0.938	0.430
	36	SANTA ROSA	8.442	7.934	3.628	1.000	0.940	0.430
	37	SEMINOLE	4.252	3.945	1.480	1.000	0.928	0.348
	38	TAYLOR	2.470	2.295	0.901	1.000	0.929	0.365
	39	VOLUSIA	7.575	7.142	3.539	1.000	0.943	0.467
	40	WAKULLA	7.472	7.038	3.495	1.000	0.942	0.468

Form A-6: Logical Relationship to Risk - Condo Unit Floor Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County / City	Loss Cost by Floor of Interest				Ratios Relative to 3rd Floor			
			3rd Floor	9th Floor	15th Floor	18th Floor	3rd Floor	9th Floor	15th Floor	18th Floor
Condo Unit A	1	BAY	1.279	4.426	7.867	16.582	1.000	3.461	6.151	12.966
	2	BREVARD	0.468	1.860	3.701	8.314	1.000	3.977	7.915	17.780
	3	BREVARD	0.404	1.532	3.155	7.198	1.000	3.790	7.804	17.804
	4	BROWARD	0.998	3.655	7.036	15.609	1.000	3.662	7.049	15.639
	5	BROWARD	3.550	10.163	16.212	31.172	1.000	2.863	4.567	8.782
	6	CITRUS	0.192	0.660	1.420	3.373	1.000	3.441	7.406	17.601
	7	CLAY	0.068	0.233	0.531	1.301	1.000	3.412	7.768	19.043
	8	COLLIER	0.838	3.103	6.187	14.006	1.000	3.704	7.386	16.720
	9	COLUMBIA	0.062	0.199	0.455	1.117	1.000	3.206	7.314	17.965
	10	DIXIE	0.815	2.043	3.306	6.734	1.000	2.506	4.056	8.261
	11	DUVAL	0.457	1.544	2.776	5.898	1.000	3.375	6.068	12.895
	12	FRANKLIN	2.430	6.080	9.422	17.488	1.000	2.502	3.877	7.195
	13	GLADES	0.502	1.857	3.906	9.044	1.000	3.701	7.784	18.024
	14	HAMILTON	0.057	0.177	0.395	0.969	1.000	3.111	6.953	17.039
	15	HERNANDO	0.477	1.510	2.898	6.461	1.000	3.164	6.073	13.536
	16	HILLSBOROUGH	0.283	1.088	2.419	5.820	1.000	3.837	8.534	20.532
	17	HOLMES	0.159	0.643	1.504	3.703	1.000	4.048	9.460	23.298
	18	INDIAN RIVER	5.168	12.044	17.169	28.885	1.000	2.331	3.322	5.589
	19	JACKSON	0.089	0.337	0.823	2.088	1.000	3.801	9.270	23.525
	20	LEE	0.613	2.149	4.413	10.129	1.000	3.502	7.193	16.511
	21	LEON	0.107	0.352	0.778	1.862	1.000	3.300	7.289	17.447
	22	MARION	0.129	0.446	1.001	2.430	1.000	3.446	7.733	18.780
	23	MARTIN	0.536	2.147	4.478	10.326	1.000	4.002	8.346	19.247
	24	MARTIN	4.297	11.802	18.140	33.562	1.000	2.746	4.221	7.810
	25	MIAMI-DADE	0.890	3.662	7.355	16.564	1.000	4.112	8.260	18.602
	26	MIAMI-DADE	2.092	7.392	12.810	26.300	1.000	3.534	6.124	12.573
	27	MONROE	5.444	16.656	25.864	48.109	1.000	3.060	4.751	8.838
	28	MONROE	10.502	21.775	29.639	52.847	1.000	2.073	2.822	5.032
	29	OKALOOSA	0.739	2.957	5.920	13.254	1.000	4.000	8.008	17.927
	30	OSCEOLA	0.179	0.644	1.462	3.544	1.000	3.591	8.156	19.771
	31	OSCEOLA	0.378	1.270	2.589	5.961	1.000	3.357	6.845	15.759
	32	PALM BEACH	0.950	3.483	6.737	15.067	1.000	3.667	7.093	15.861
	33	PALM BEACH	2.358	7.881	13.392	27.061	1.000	3.342	5.678	11.474
	34	PINELLAS	0.362	1.500	3.197	7.442	1.000	4.140	8.822	20.538
	35	SAINT JOHNS	0.254	0.931	1.880	4.198	1.000	3.672	7.410	16.550
	36	SANTA ROSA	0.461	1.692	3.511	8.140	1.000	3.667	7.608	17.642
	37	SEMINOLE	0.146	0.514	1.171	2.846	1.000	3.531	8.046	19.556
	38	TAYLOR	0.105	0.366	0.811	1.931	1.000	3.486	7.735	18.408
	39	VOLUSIA	0.466	1.743	3.323	7.382	1.000	3.742	7.135	15.852
	40	WAKULLA	0.594	1.847	3.298	7.023	1.000	3.111	5.554	11.830

Form A-6: Logical Relationship to Risk - Condo Unit Floor Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County / City	Loss Cost by Floor of Interest				Ratios Relative to 3rd Floor			
			3rd Floor	9th Floor	15th Floor	18th Floor	3rd Floor	9th Floor	15th Floor	18th Floor
Condo Unit B	1	BAY	4.557	14.741	21.817	41.424	1.000	3.235	4.788	9.090
	2	BREVARD	1.868	7.883	12.615	25.270	1.000	4.220	6.754	13.530
	3	BREVARD	1.631	7.083	11.566	23.508	1.000	4.344	7.094	14.418
	4	BROWARD	3.980	14.382	21.990	43.163	1.000	3.614	5.525	10.845
	5	BROWARD	10.567	26.615	36.663	64.338	1.000	2.519	3.470	6.089
	6	CITRUS	0.748	3.842	6.671	14.062	1.000	5.137	8.921	18.803
	7	CLAY	0.227	1.552	2.970	6.521	1.000	6.845	13.095	28.753
	8	COLLIER	3.469	13.556	21.096	42.056	1.000	3.908	6.081	12.124
	9	COLUMBIA	0.200	1.348	2.623	5.794	1.000	6.744	13.119	28.977
	10	DIXIE	2.740	7.652	10.955	20.726	1.000	2.793	3.998	7.564
	11	DUVAL	1.665	5.777	8.859	17.196	1.000	3.470	5.321	10.328
	12	FRANKLIN	7.464	17.803	24.233	42.393	1.000	2.385	3.247	5.680
	13	GLADES	2.119	9.275	15.012	30.624	1.000	4.378	7.086	14.456
	14	HAMILTON	0.182	1.202	2.301	5.078	1.000	6.595	12.628	27.869
	15	HERNANDO	1.879	7.015	11.019	22.085	1.000	3.734	5.865	11.755
	16	HILLSBOROUGH	1.173	6.136	10.475	21.926	1.000	5.232	8.933	18.698
	17	HOLMES	0.613	3.893	7.073	15.183	1.000	6.348	11.534	24.760
	18	INDIAN RIVER	13.225	27.470	35.439	57.054	1.000	2.077	2.680	4.314
	19	JACKSON	0.309	2.370	4.583	10.150	1.000	7.661	14.817	32.811
	20	LEE	2.595	10.535	16.744	33.827	1.000	4.060	6.452	13.036
	21	LEON	0.393	2.184	3.948	8.471	1.000	5.559	10.048	21.559
	22	MARION	0.485	2.846	5.166	11.099	1.000	5.862	10.642	22.861
	23	MARTIN	2.238	9.796	15.827	32.198	1.000	4.377	7.071	14.386
	24	MARTIN	12.020	28.337	38.150	64.985	1.000	2.357	3.174	5.406
	25	MIAMI-DADE	3.614	14.451	22.517	44.550	1.000	3.999	6.231	12.329
	26	MIAMI-DADE	7.301	21.889	31.326	57.347	1.000	2.998	4.290	7.854
	27	MONROE	15.937	38.307	50.896	84.583	1.000	2.404	3.194	5.307
	28	MONROE	25.336	45.858	55.842	88.861	1.000	1.810	2.204	3.507
	29	OKALOOSA	3.013	11.972	18.716	37.046	1.000	3.974	6.213	12.297
	30	OSCEOLA	0.700	4.071	7.276	15.521	1.000	5.816	10.394	22.171
	31	OSCEOLA	1.547	6.486	10.556	21.633	1.000	4.193	6.823	13.983
	32	PALM BEACH	3.846	13.917	21.337	42.105	1.000	3.619	5.548	10.948
	33	PALM BEACH	7.907	23.107	32.933	60.056	1.000	2.922	4.165	7.595
	34	PINELLAS	1.488	7.174	11.903	24.381	1.000	4.822	8.000	16.386
	35	SAINT JOHNS	0.958	4.204	6.954	14.128	1.000	4.387	7.257	14.744
	36	SANTA ROSA	1.949	8.289	13.267	26.894	1.000	4.252	6.806	13.797
	37	SEMINOLE	0.551	3.308	6.003	12.902	1.000	6.002	10.891	23.408
	38	TAYLOR	0.391	2.156	3.877	8.290	1.000	5.506	9.904	21.176
	39	VOLUSIA	1.824	7.080	11.025	21.885	1.000	3.882	6.045	12.000
	40	WAKULLA	2.204	7.204	10.854	21.060	1.000	3.269	4.926	9.557

Form A-6: Logical Relationship to Risk - Number of Stories Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County / City	Loss Cost by Number of Stories		Ratios Relative to 1 Story		Loss Cost by Number of Stories			Ratios Relative to 5 Story		
			1 Story	2 Story	1 Story	2 Story	5 Story	10 Story	20 Story	5 Story	10 Story	20 Story
Frame Owners	1	BAY	3.841	5.364	1.000	1.397						
	2	BREVARD	4.681	5.286	1.000	1.129						
	3	BREVARD	4.526	5.028	1.000	1.111						
	4	BROWARD	7.409	8.803	1.000	1.188						
	5	BROWARD	12.721	16.507	1.000	1.298						
	6	CITRUS	3.417	3.600	1.000	1.053						
	7	CLAY	0.867	0.909	1.000	1.049						
	8	COLLIER	6.962	8.127	1.000	1.167						
	9	COLUMBIA	0.829	0.864	1.000	1.043						
	10	DIXIE	2.698	3.549	1.000	1.315						
	11	DUVAL	1.724	2.203	1.000	1.278						
	12	FRANKLIN	5.849	8.391	1.000	1.435						
	13	GLADES	5.955	6.602	1.000	1.109						
	14	HAMILTON	0.733	0.763	1.000	1.041						
	15	HERNANDO	4.717	5.291	1.000	1.122						
	16	HILLSBOROUGH	4.248	4.551	1.000	1.071						
	17	HOLMES	1.398	1.526	1.000	1.092						
	18	INDIAN RIVER	13.923	18.263	1.000	1.312						
	19	JACKSON	1.084	1.132	1.000	1.045						
	20	LEE	6.354	7.149	1.000	1.125						
	21	LEON	1.027	1.106	1.000	1.077						
	22	MARION	2.825	2.931	1.000	1.037						
	23	MARTIN	5.838	6.544	1.000	1.121						
	24	MARTIN	13.175	17.280	1.000	1.312						
	25	MIAMI-DADE	6.714	7.947	1.000	1.184						
	26	MIAMI-DADE	9.648	12.396	1.000	1.285						
	27	MONROE	14.067	19.259	1.000	1.369						
	28	MONROE	22.019	30.061	1.000	1.365						
	29	OKALOOSA	2.781	3.748	1.000	1.347						
	30	OSCEOLA	3.698	3.861	1.000	1.044						
	31	OSCEOLA	4.785	5.231	1.000	1.093						
	32	PALM BEACH	7.462	8.816	1.000	1.181						
	33	PALM BEACH	10.706	13.605	1.000	1.271						
	34	PINELLAS	4.238	4.667	1.000	1.101						
	35	SAINT JOHNS	1.326	1.590	1.000	1.199						
	36	SANTA ROSA	2.008	2.538	1.000	1.264						
	37	SEMINOLE	3.251	3.369	1.000	1.036						
	38	TAYLOR	0.931	1.021	1.000	1.096						
	39	VOLUSIA	3.922	4.493	1.000	1.146						
	40	WAKULLA	2.362	3.057	1.000	1.294						

Form A-6: Logical Relationship to Risk - Number of Stories Sensitivity

Modeling Organization: Florida International University
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Model Release Date: February 3, 2015

Construction / Policy	Location	County / City	Loss Cost by Number of Stories		Ratios Relative to 1 Story		Loss Cost by Number of Stories			Ratios Relative to 5 Story		
			1 Story	2 Story	1 Story	2 Story	5 Story	10 Story	20 Story	5 Story	10 Story	20 Story
Masonry Owners	1	BAY	3.576	4.936	1.000	1.380						
	2	BREVARD	4.630	5.191	1.000	1.121						
	3	BREVARD	4.482	4.951	1.000	1.105						
	4	BROWARD	7.226	8.546	1.000	1.183						
	5	BROWARD	12.139	15.469	1.000	1.274						
	6	CITRUS	3.376	3.554	1.000	1.052						
	7	CLAY	0.827	0.868	1.000	1.050						
	8	COLLIER	6.881	7.967	1.000	1.158						
	9	COLUMBIA	0.791	0.826	1.000	1.044						
	10	DIXIE	2.515	3.275	1.000	1.302						
	11	DUVAL	1.622	2.053	1.000	1.265						
	12	FRANKLIN	5.363	7.570	1.000	1.412						
	13	GLADES	5.881	6.508	1.000	1.107						
	14	HAMILTON	0.700	0.729	1.000	1.042						
	15	HERNANDO	4.664	5.196	1.000	1.114						
	16	HILLSBOROUGH	4.198	4.493	1.000	1.070						
	17	HOLMES	1.323	1.447	1.000	1.094						
	18	INDIAN RIVER	12.926	16.547	1.000	1.280						
	19	JACKSON	1.029	1.077	1.000	1.047						
	20	LEE	6.274	7.038	1.000	1.122						
	21	LEON	0.976	1.052	1.000	1.079						
	22	MARION	2.792	2.895	1.000	1.037						
	23	MARTIN	5.700	6.383	1.000	1.120						
	24	MARTIN	12.466	16.021	1.000	1.285						
	25	MIAMI-DADE	6.552	7.727	1.000	1.179						
	26	MIAMI-DADE	9.341	11.840	1.000	1.268						
	27	MONROE	13.346	17.863	1.000	1.338						
	28	MONROE	20.238	27.027	1.000	1.335						
	29	OKALOOSA	2.615	3.507	1.000	1.341						
	30	OSCEOLA	3.655	3.814	1.000	1.044						
	31	OSCEOLA	4.725	5.155	1.000	1.091						
	32	PALM BEACH	7.279	8.564	1.000	1.177						
	33	PALM BEACH	10.332	12.957	1.000	1.254						
	34	PINELLAS	4.201	4.605	1.000	1.096						
	35	SAINT JOHNS	1.254	1.499	1.000	1.195						
	36	SANTA ROSA	1.892	2.391	1.000	1.264						
	37	SEMINOLE	3.212	3.329	1.000	1.036						
	38	TAYLOR	0.886	0.970	1.000	1.096						
	39	VOLUSIA	3.874	4.399	1.000	1.136						
	40	WAKULLA	2.219	2.847	1.000	1.283						

Form A-6: Logical Relationship to Risk - Number of Stories Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.1
Model Release Date: February 3, 2015

Construction / Policy	Location	County / City	Loss Cost by Number of Stories		Ratios Relative to 1 Story		Loss Cost by Number of Stories			Ratios Relative to 5 Story		
			1 Story	2 Story	1 Story	2 Story	5 Story	10 Story	20 Story	5 Story	10 Story	20 Story
Frame Renters	1	BAY	1.211	2.137	1.000	1.765						
	2	BREVARD	0.584	0.913	1.000	1.562						
	3	BREVARD	0.538	0.803	1.000	1.491						
	4	BROWARD	1.038	1.830	1.000	1.764						
	5	BROWARD	3.159	5.733	1.000	1.815						
	6	CITRUS	0.355	0.442	1.000	1.247						
	7	CLAY	0.193	0.211	1.000	1.096						
	8	COLLIER	0.898	1.519	1.000	1.690						
	9	COLUMBIA	0.184	0.199	1.000	1.082						
	10	DIXIE	0.816	1.330	1.000	1.630						
	11	DUVAL	0.482	0.764	1.000	1.586						
	12	FRANKLIN	2.130	3.800	1.000	1.784						
	13	GLADES	0.662	0.989	1.000	1.493						
	14	HAMILTON	0.163	0.175	1.000	1.075						
	15	HERNANDO	0.588	0.899	1.000	1.529						
	16	HILLSBOROUGH	0.442	0.585	1.000	1.324						
	17	HOLMES	0.327	0.386	1.000	1.180						
	18	INDIAN RIVER	4.261	7.383	1.000	1.733						
	19	JACKSON	0.246	0.265	1.000	1.078						
	20	LEE	0.729	1.138	1.000	1.562						
	21	LEON	0.235	0.271	1.000	1.156						
	22	MARION	0.287	0.335	1.000	1.168						
	23	MARTIN	0.677	1.051	1.000	1.551						
	24	MARTIN	3.675	6.579	1.000	1.790						
	25	MIAMI-DADE	0.909	1.596	1.000	1.756						
	26	MIAMI-DADE	1.916	3.655	1.000	1.908						
	27	MONROE	4.110	7.744	1.000	1.884						
	28	MONROE	8.234	14.270	1.000	1.733						
	29	OKALOOSA	0.765	1.294	1.000	1.691						
	30	OSCEOLA	0.373	0.446	1.000	1.196						
	31	OSCEOLA	0.533	0.764	1.000	1.435						
	32	PALM BEACH	1.015	1.784	1.000	1.758						
	33	PALM BEACH	2.159	4.025	1.000	1.864						
	34	PINELLAS	0.484	0.701	1.000	1.448						
	35	SAINT JOHNS	0.335	0.476	1.000	1.422						
	36	SANTA ROSA	0.510	0.776	1.000	1.522						
	37	SEMINOLE	0.327	0.379	1.000	1.161						
	38	TAYLOR	0.217	0.260	1.000	1.195						
	39	VOLUSIA	0.517	0.837	1.000	1.617						
	40	WAKULLA	0.666	1.069	1.000	1.606						

Form A-6: Logical Relationship to Risk - Number of Stories Sensitivity

Modeling Organization: Florida International University
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Model Release Date: February 3, 2015

Construction / Policy	Location	County / City	Loss Cost by Number of Stories		Ratios Relative to 1 Story		Loss Cost by Number of Stories			Ratios Relative to 5 Story		
			1 Story	2 Story	1 Story	2 Story	5 Story	10 Story	20 Story	5 Story	10 Story	20 Story
Masonry Renters	1	BAY	1.113	1.953	1.000	1.754						
	2	BREVARD	0.572	0.890	1.000	1.556						
	3	BREVARD	0.527	0.786	1.000	1.492						
	4	BROWARD	1.017	1.777	1.000	1.748						
	5	BROWARD	2.898	5.116	1.000	1.765						
	6	CITRUS	0.345	0.434	1.000	1.259						
	7	CLAY	0.170	0.189	1.000	1.114						
	8	COLLIER	0.882	1.490	1.000	1.691						
	9	COLUMBIA	0.162	0.178	1.000	1.097						
	10	DIXIE	0.737	1.204	1.000	1.633						
	11	DUVAL	0.440	0.698	1.000	1.588						
	12	FRANKLIN	1.914	3.365	1.000	1.758						
	13	GLADES	0.649	0.980	1.000	1.511						
	14	HAMILTON	0.143	0.156	1.000	1.091						
	15	HERNANDO	0.573	0.873	1.000	1.524						
	16	HILLSBOROUGH	0.431	0.579	1.000	1.343						
	17	HOLMES	0.291	0.352	1.000	1.207						
	18	INDIAN RIVER	3.726	6.315	1.000	1.695						
	19	JACKSON	0.216	0.237	1.000	1.095						
	20	LEE	0.716	1.126	1.000	1.574						
	21	LEON	0.208	0.246	1.000	1.179						
	22	MARION	0.279	0.328	1.000	1.179						
	23	MARTIN	0.665	1.037	1.000	1.560						
	24	MARTIN	3.308	5.765	1.000	1.743						
	25	MIAMI-DADE	0.892	1.559	1.000	1.748						
	26	MIAMI-DADE	1.840	3.408	1.000	1.853						
	27	MONROE	3.772	6.859	1.000	1.819						
	28	MONROE	7.087	12.073	1.000	1.704						
	29	OKALOOSA	0.709	1.219	1.000	1.719						
	30	OSCEOLA	0.363	0.438	1.000	1.209						
	31	OSCEOLA	0.521	0.752	1.000	1.444						
	32	PALM BEACH	0.998	1.741	1.000	1.745						
	33	PALM BEACH	2.047	3.718	1.000	1.816						
	34	PINELLAS	0.473	0.690	1.000	1.458						
	35	SAINT JOHNS	0.303	0.440	1.000	1.451						
	36	SANTA ROSA	0.466	0.734	1.000	1.573						
	37	SEMINOLE	0.317	0.372	1.000	1.173						
	38	TAYLOR	0.193	0.236	1.000	1.222						
	39	VOLUSIA	0.506	0.811	1.000	1.602						
	40	WAKULLA	0.609	0.983	1.000	1.613						

Form A-6: Logical Relationship to Risk - Number of Stories Sensitivity

Modeling Organization: Florida International University
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Construction / Policy	Location	County / City	Loss Cost by Number of Stories		Ratios Relative to 1 Story		Loss Cost by Number of Stories			Ratios Relative to 5 Story		
			1 Story	2 Story	1 Story	2 Story	5 Story	10 Story	20 Story	5 Story	10 Story	20 Story
Commercial Residential	1	BAY					2.259	5.534	14.376	1.000	2.450	6.365
	2	BREVARD					1.041	2.948	8.598	1.000	2.830	8.255
	3	BREVARD					0.901	2.608	7.822	1.000	2.894	8.678
	4	BROWARD					2.246	5.841	15.876	1.000	2.601	7.068
	5	BROWARD					5.492	11.691	27.147	1.000	2.129	4.943
	6	CITRUS					0.418	1.323	4.341	1.000	3.161	10.375
	7	CLAY					0.130	0.471	1.750	1.000	3.622	13.456
	8	COLLIER					1.932	5.260	14.760	1.000	2.723	7.642
	9	COLUMBIA					0.114	0.413	1.554	1.000	3.634	13.671
	10	DIXIE					1.237	2.854	7.182	1.000	2.307	5.805
	11	DUVAL					0.745	1.900	5.207	1.000	2.551	6.992
	12	FRANKLIN					3.436	7.213	16.460	1.000	2.099	4.790
	13	GLADES					1.231	3.586	10.698	1.000	2.914	8.693
	14	HAMILTON					0.097	0.348	1.300	1.000	3.587	13.407
	15	HERNANDO					0.964	2.581	7.346	1.000	2.678	7.623
	16	HILLSBOROUGH					0.679	2.158	6.921	1.000	3.179	10.195
	17	HOLMES					0.344	1.212	4.254	1.000	3.518	12.353
	18	INDIAN RIVER					6.170	11.868	25.061	1.000	1.924	4.062
	19	JACKSON					0.179	0.697	2.678	1.000	3.891	14.951
	20	LEE					1.420	3.970	11.515	1.000	2.796	8.110
	21	LEON					0.199	0.664	2.289	1.000	3.330	11.487
	22	MARION					0.279	0.945	3.291	1.000	3.382	11.778
	23	MARTIN					1.310	3.797	11.263	1.000	2.898	8.596
	24	MARTIN					6.041	12.341	27.686	1.000	2.043	4.583
	25	MIAMI-DADE					2.090	5.695	15.877	1.000	2.726	7.598
	26	MIAMI-DADE					3.988	9.306	22.893	1.000	2.333	5.740
	27	MONROE					7.388	15.167	33.413	1.000	2.053	4.523
	28	MONROE					11.444	20.432	40.034	1.000	1.785	3.498
	29	OKALOOSA					1.560	4.259	11.959	1.000	2.729	7.664
	30	OSCEOLA					0.425	1.426	4.856	1.000	3.359	11.436
	31	OSCEOLA					0.832	2.373	7.123	1.000	2.851	8.557
	32	PALM BEACH					2.169	5.671	15.494	1.000	2.615	7.144
	33	PALM BEACH					4.322	9.956	24.399	1.000	2.304	5.646
	34	PINELLAS					0.841	2.564	7.913	1.000	3.048	9.407
	35	SAINT JOHNS					0.477	1.386	4.217	1.000	2.909	8.851
	36	SANTA ROSA					0.946	2.708	7.934	1.000	2.863	8.391
	37	SEMINOLE					0.331	1.133	3.945	1.000	3.429	11.935
	38	TAYLOR					0.206	0.674	2.295	1.000	3.266	11.124
	39	VOLUSIA					0.937	2.536	7.142	1.000	2.707	7.624
	40	WAKULLA					1.055	2.632	7.038	1.000	2.495	6.671

Appendix H - Form A-7: Percentage Change in Logical Relationship to Risk

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
November 1, 2014

Form A-7: Percent Change in Logical Relationship to Risk - Deductibles

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Region	Percent Change in Loss Cost					
		$0	$500	1%	2%	5%	10%
Frame Owners	Coastal	-5.2%	-5.6%	-6.0%	-6.7%	-9.0%	-11.2%
	Inland	-5.3%	-5.9%	-7.0%	-7.9%	-10.7%	-12.3%
	North	-1.4%	-1.8%	-2.5%	-3.1%	-3.7%	-3.9%
	Central	-2.8%	-3.2%	-3.7%	-4.3%	-6.6%	-8.4%
	South	-7.3%	-7.5%	-7.9%	-8.4%	-10.5%	-12.8%
	Statewide	-5.2%	-5.6%	-6.2%	-6.8%	-9.1%	-11.3%
Masonry Owners	Coastal	-5.0%	-5.4%	-5.8%	-6.5%	-8.7%	-10.8%
	Inland	-5.2%	-5.9%	-7.0%	-7.9%	-10.7%	-12.1%
	North	-1.2%	-1.7%	-2.4%	-3.0%	-3.5%	-3.6%
	Central	-2.8%	-3.1%	-3.6%	-4.2%	-6.5%	-8.2%
	South	-7.0%	-7.3%	-7.6%	-8.2%	-10.2%	-12.4%
	Statewide	-5.1%	-5.5%	-6.0%	-6.6%	-8.9%	-10.9%
Mobile Homes	Coastal	-6.1%	-6.4%	-6.4%	-6.6%	-7.3%	-8.0%
	Inland	-9.4%	-10.4%	-10.4%	-11.1%	-13.6%	-16.1%
	North	-1.8%	-2.1%	-2.1%	-2.3%	-2.7%	-3.1%
	Central	-6.5%	-7.1%	-7.1%	-7.6%	-9.1%	-10.5%
	South	-7.3%	-7.4%	-7.4%	-7.6%	-8.0%	-8.6%
	Statewide	-6.5%	-6.9%	-6.9%	-7.2%	-7.9%	-8.7%
Frame Renters	Coastal	-8.0%	-10.7%	-10.1%	-10.7%	-12.2%	-12.7%
	Inland	-6.0%	-10.4%	-9.6%	-10.4%	-12.0%	-11.8%
	North	-1.9%	-3.7%	-3.5%	-3.7%	-3.9%	-4.0%
	Central	-4.0%	-6.8%	-6.0%	-6.8%	-8.7%	-9.0%
	South	-10.3%	-12.5%	-12.0%	-12.5%	-14.0%	-14.6%
	Statewide	-7.8%	-10.6%	-10.0%	-10.6%	-12.2%	-12.7%
Masonry Renters	Coastal	-7.5%	-10.1%	-9.5%	-10.1%	-11.7%	-12.2%
	Inland	-6.0%	-10.3%	-9.6%	-10.3%	-11.9%	-11.6%
	North	-1.6%	-3.2%	-3.1%	-3.2%	-3.4%	-3.5%
	Central	-3.9%	-6.6%	-5.9%	-6.6%	-8.6%	-8.8%
	South	-9.7%	-11.9%	-11.4%	-11.9%	-13.4%	-14.0%
	Statewide	-7.3%	-10.2%	-9.6%	-10.2%	-11.7%	-12.1%
Frame Condo Unit	Coastal	-7.3%	-9.1%	-9.1%	-10.3%	-11.7%	-12.7%
	Inland	-5.7%	-8.1%	-8.1%	-10.3%	-11.7%	-12.1%
	North	-1.8%	-3.0%	-3.0%	-3.7%	-4.0%	-4.0%
	Central	-3.6%	-5.2%	-5.2%	-6.6%	-8.1%	-9.0%
	South	-9.5%	-11.1%	-11.1%	-12.2%	-13.5%	-14.6%
	Statewide	-7.1%	-9.0%	-9.0%	-10.3%	-11.7%	-12.6%
Masonry Condo Unit	Coastal	-6.9%	-8.6%	-8.6%	-9.9%	-11.2%	-12.2%
	Inland	-5.7%	-8.1%	-8.1%	-10.3%	-11.6%	-11.9%
	North	-1.5%	-2.6%	-2.6%	-3.3%	-3.5%	-3.5%
	Central	-3.5%	-5.1%	-5.1%	-6.4%	-7.9%	-8.9%
	South	-9.0%	-10.5%	-10.5%	-11.6%	-12.9%	-13.9%
	Statewide	-6.7%	-8.6%	-8.6%	-9.9%	-11.3%	-12.1%

Construction / Policy	Region	Percent Change in Loss Cost					
		$0	2%	3%	5%	10%	
Commercial Residential	Coastal	-6.3%	-6.9%	-7.1%	-7.6%	-8.7%	
	Inland	-12.4%	-15.0%	-15.8%	-17.1%	-19.4%	
	North	-3.3%	-3.7%	-3.8%	-3.9%	-4.1%	
	Central	-8.5%	-10.0%	-10.5%	-11.2%	-12.6%	
	South	-7.0%	-7.4%	-7.6%	-8.0%	-9.3%	
	Statewide	-7.1%	-7.8%	-8.1%	-8.6%	-9.6%	

Form A-7: Percent Change in Logical Relationship to Risk - Construction

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Policy	Region	Percent Change in Loss Cost	
		Masonry	Frame
Owners	Coastal	-5.0%	-5.2%
	Inland	-5.2%	-5.3%
	North	-1.2%	-1.4%
	Central	-2.8%	-2.8%
	South	-7.0%	-7.3%
	Statewide	-5.1%	-5.2%
Renters	Coastal	-7.5%	-8.0%
	Inland	-6.0%	-6.0%
	North	-1.6%	-1.9%
	Central	-3.9%	-4.0%
	South	-9.7%	-10.3%
	Statewide	-7.3%	-7.8%
Condo Unit	Coastal	-6.9%	-7.3%
	Inland	-5.7%	-5.7%
	North	-1.5%	-1.8%
	Central	-3.5%	-3.6%
	South	-9.0%	-9.5%
	Statewide	-6.7%	-7.1%
Policy	**Region**	**Percent Change in Loss Cost**	
		Concrete	
Commercial Residential	Coastal	85.4%	
	Inland	195.4%	
	North	122.2%	
	Central	137.3%	
	South	72.4%	
	Statewide	94.4%	

Form A-7: Percent Change in Logical Relationship to Risk - Policy Form

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Region	Percent Change in Loss Cost		
	Frame Owners	Masonry Owners	Mobile Homes
Coastal	-5.2%	-5.0%	-6.1%
Inland	-5.3%	-5.2%	-9.4%
North	-1.4%	-1.2%	-1.8%
Central	-2.8%	-2.8%	-6.5%
South	-7.3%	-7.0%	-7.3%
Statewide	-5.2%	-5.1%	-6.5%

Form A-7: Percent Change in Logical Relationship to Risk - Coverage

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Region	Percent Change in Loss Cost			
		Coverage A	Coverage B	Coverage C	Coverage D
Frame Owners	Coastal	-4.8%	-4.0%	-7.4%	-10.5%
	Inland	-5.2%	-5.9%	-5.5%	-8.5%
	North	-1.3%	-1.0%	-1.6%	-2.9%
	Central	-2.6%	-3.9%	-3.5%	-6.9%
	South	-6.8%	-5.5%	-9.7%	-12.6%
	Statewide	-4.8%	-4.4%	-7.2%	-10.3%
Masonry Owners	Coastal	-4.6%	-4.0%	-6.9%	-10.0%
	Inland	-5.1%	-5.9%	-5.5%	-8.6%
	North	-1.2%	-1.0%	-1.3%	-2.6%
	Central	-2.6%	-3.9%	-3.3%	-6.9%
	South	-6.6%	-5.5%	-9.2%	-11.9%
	Statewide	-4.7%	-4.4%	-6.8%	-9.9%
Mobile Homes	Coastal	-5.6%	-4.0%	-8.1%	-8.3%
	Inland	-8.5%	-5.9%	-14.8%	-16.4%
	North	-1.6%	-1.0%	-2.8%	-3.1%
	Central	-5.7%	-3.9%	-10.1%	-10.9%
	South	-6.8%	-5.5%	-8.9%	-9.0%
	Statewide	-6.0%	-4.4%	-8.8%	-9.1%
Frame Renters	Coastal	NA	NA	-7.4%	-10.5%
	Inland	NA	NA	-5.5%	-8.5%
	North	NA	NA	-1.6%	-2.9%
	Central	NA	NA	-3.5%	-6.9%
	South	NA	NA	-9.7%	-12.6%
	Statewide	NA	NA	-7.2%	-10.3%
Masonry Renters	Coastal	NA	NA	-6.9%	-10.0%
	Inland	NA	NA	-5.5%	-8.6%
	North	NA	NA	-1.3%	-2.6%
	Central	NA	NA	-3.3%	-6.9%
	South	NA	NA	-9.2%	-11.9%
	Statewide	NA	NA	-6.8%	-9.9%
Frame Condo Unit	Coastal	-4.8%	NA	-7.4%	-10.5%
	Inland	-5.2%	NA	-5.5%	-8.5%
	North	-1.3%	NA	-1.6%	-2.9%
	Central	-2.6%	NA	-3.5%	-6.9%
	South	-6.8%	NA	-9.7%	-12.6%
	Statewide	-4.8%	NA	-7.2%	-10.3%
Masonry Condo Unit	Coastal	-4.6%	NA	-6.9%	-10.0%
	Inland	-5.1%	NA	-5.5%	-8.6%
	North	-1.2%	NA	-1.3%	-2.6%
	Central	-2.6%	NA	-3.3%	-6.9%
	South	-6.6%	NA	-9.2%	-11.9%
	Statewide	-4.7%	NA	-6.8%	-9.9%
Commercial Residential	Coastal	-6.3%	NA	-6.2%	NA
	Inland	-12.4%	NA	-12.5%	NA
	North	-3.3%	NA	-2.6%	NA
	Central	-8.5%	NA	-8.3%	NA
	South	-7.0%	NA	-7.1%	NA
	Statewide	-7.1%	NA	-7.0%	NA

Form A-7: Percent Change in Logical Relationship to Risk - Building Code / Enforcement (Year Built) Sensitivity

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Region	Percent Change in Loss Cost		
		Year Built 1980	Year Built 1998	Year Built 2004
Frame Owners	Coastal	-5.5%	-3.6%	-3.5%
	Inland	-6.2%	-5.0%	-3.7%
	North	-1.8%	-0.9%	-0.1%
	Central	-3.6%	-2.8%	-1.9%
	South	-7.4%	-5.5%	-5.8%
	Statewide	-5.6%	-3.9%	-3.6%
Masonry Owners	Coastal	-5.4%	-3.5%	-3.4%
	Inland	-6.2%	-5.0%	-3.6%
	North	-1.8%	-0.8%	-0.1%
	Central	-3.6%	-2.8%	-1.8%
	South	-7.3%	-5.4%	-5.7%
	Statewide	-5.5%	-3.8%	-3.4%

Construction / Policy	Region	Percent Change in Loss Cost		
		Year Built 1974	Year Built 1992	Year Built 2004
Mobile Homes	Coastal	-6.1%	-6.1%	-4.7%
	Inland	-9.3%	-9.3%	-3.9%
	North	-1.9%	-1.9%	-0.7%
	Central	-6.8%	-6.8%	-2.3%
	South	-7.3%	-7.3%	-7.4%
	Statewide	-6.5%	-6.5%	-4.6%

Construction / Policy	Region	Percent Change in Loss Cost		
		Year Built 1980	Year Built 1998	Year Built 2004
Frame Renters	Coastal	-9.8%	-6.5%	-4.9%
	Inland	-8.4%	-5.5%	-4.3%
	North	-2.8%	-1.7%	-0.9%
	Central	-6.8%	-4.3%	-2.5%
	South	-11.9%	-9.1%	-7.5%
	Statewide	-9.7%	-6.3%	-4.8%
Masonry Renters	Coastal	-9.7%	-6.2%	-4.4%
	Inland	-8.6%	-5.7%	-4.3%
	North	-2.8%	-1.7%	-0.8%
	Central	-6.8%	-4.4%	-2.5%
	South	-11.6%	-8.6%	-6.9%
	Statewide	-9.5%	-6.1%	-4.4%
Frame Condo Unit	Coastal	-8.5%	-5.5%	-4.4%
	Inland	-7.4%	-5.3%	-4.0%
	North	-2.6%	-1.5%	-0.7%
	Central	-5.5%	-3.7%	-2.3%
	South	-10.6%	-7.9%	-7.0%
	Statewide	-8.4%	-5.5%	-4.4%
Masonry Condo Unit	Coastal	-8.3%	-5.2%	-4.1%
	Inland	-7.5%	-5.4%	-4.0%
	North	-2.6%	-1.4%	-0.5%
	Central	-5.5%	-3.7%	-2.2%
	South	-10.3%	-7.6%	-6.5%
	Statewide	-8.2%	-5.3%	-4.1%
Commercial Residential	Coastal	-6.3%	-6.4%	-7.3%
	Inland	-12.4%	-12.4%	-14.2%
	North	-3.3%	-3.3%	-3.7%
	Central	-8.5%	-8.5%	-9.7%
	South	-7.0%	-7.3%	-8.1%
	Statewide	-7.1%	-7.2%	-8.1%

Form A-7: Percent Change in Logical Relationship to Risk - Building Strength

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Region	Percent Change in Loss Cost		
		Weak	Medium	Strong
Frame Owners	Coastal	-5.4%	-3.6%	-3.2%
	Inland	-6.6%	-5.0%	-3.6%
	North	-1.8%	-0.9%	0.0%
	Central	-4.0%	-2.8%	-1.8%
	South	-7.2%	-5.5%	-5.3%
	Statewide	-5.6%	-3.9%	-3.3%
Masonry Owners	Coastal	-5.2%	-3.5%	-3.1%
	Inland	-6.6%	-5.0%	-3.5%
	North	-1.7%	-0.8%	0.0%
	Central	-3.9%	-2.8%	-1.7%
	South	-7.0%	-5.4%	-5.2%
	Statewide	-5.5%	-3.8%	-3.2%
Mobile Homes	Coastal	-6.1%	-6.1%	-4.7%
	Inland	-9.3%	-9.3%	-3.9%
	North	-1.9%	-1.9%	-0.7%
	Central	-6.8%	-6.8%	-2.3%
	South	-7.3%	-7.3%	-7.4%
	Statewide	-6.5%	-6.5%	-4.6%
Frame Renters	Coastal	-9.5%	-6.5%	-3.9%
	Inland	-9.6%	-5.5%	-4.1%
	North	-2.6%	-1.7%	-0.7%
	Central	-7.7%	-4.3%	-2.2%
	South	-11.5%	-9.1%	-6.3%
	Statewide	-9.5%	-6.3%	-3.9%
Masonry Renters	Coastal	-9.2%	-6.2%	-3.8%
	Inland	-9.6%	-5.7%	-4.2%
	North	-2.6%	-1.7%	-0.7%
	Central	-7.6%	-4.4%	-2.2%
	South	-11.1%	-8.6%	-6.0%
	Statewide	-9.3%	-6.1%	-3.8%
Frame Condo Unit	Coastal	-8.5%	-5.5%	-5.2%
	Inland	-10.9%	-5.3%	-4.1%
	North	-2.7%	-1.5%	-0.9%
	Central	-8.1%	-3.7%	-2.4%
	South	-10.1%	-7.9%	-8.0%
	Statewide	-8.8%	-5.5%	-5.0%
Masonry Condo Unit	Coastal	-8.2%	-5.2%	-4.8%
	Inland	-10.9%	-5.4%	-4.1%
	North	-2.7%	-1.4%	-0.9%
	Central	-8.0%	-3.7%	-2.4%
	South	-9.6%	-7.6%	-7.4%
	Statewide	-8.4%	-5.3%	-4.7%
Commercial Residential	Coastal	-6.1%	-6.4%	-8.7%
	Inland	-12.2%	-12.4%	-15.1%
	North	-3.2%	-3.3%	-4.0%
	Central	-8.3%	-8.5%	-10.5%
	South	-6.9%	-7.3%	-9.7%
	Statewide	-6.9%	-7.2%	-9.4%

Form A-7: Percent Change in Logical Relationship to Risk - Condo Unit Floor

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Region	Percent Change in Loss Cost			
		3rd Floor	9th Floor	15th Floor	18th Floor
Condo Unit A	Coastal	-10.6%	-10.4%	-9.3%	60.5%
	Inland	-16.0%	-17.8%	-17.4%	53.9%
	North	-5.1%	-5.9%	-5.9%	68.0%
	Central	-11.0%	-12.3%	-12.2%	59.5%
	South	-12.0%	-11.3%	-9.9%	58.4%
	Statewide	-11.0%	-11.0%	-10.1%	59.8%
Condo Unit B	Coastal	-9.1%	-6.9%	-6.2%	59.1%
	Inland	-17.5%	-14.8%	-13.5%	54.6%
	North	-4.8%	-5.1%	-5.1%	63.3%
	Central	-11.9%	-10.5%	-9.6%	58.2%
	South	-9.8%	-6.9%	-6.1%	57.4%
	Statewide	-9.8%	-7.9%	-7.2%	58.4%

Form A-7: Percent Change in Logical Relationship to Risk - Number of Stories

Modeling Organization: Florida International University
Model Name & Version Number: Florida Public Hurricane Loss Model 6.0
Model Release Date: November 1, 2014

Construction / Policy	Region	Percent Change in Loss Cost		
		1 Story	2 Story	
Frame Owners	Coastal	-5.0%	-6.0%	
	Inland	-5.5%	-6.4%	
	North	-1.3%	-1.7%	
	Central	-3.1%	-4.1%	
	South	-7.1%	-7.9%	
	Statewide	-5.1%	-6.1%	
Masonry Owners	Coastal	-4.9%	-5.7%	
	Inland	-5.5%	-6.4%	
	North	-1.2%	-1.7%	
	Central	-3.0%	-4.0%	
	South	-6.9%	-7.6%	
	Statewide	-5.0%	-5.8%	
Frame Renters	Coastal	-9.2%	-10.3%	
	Inland	-6.0%	-9.2%	
	North	-2.1%	-2.7%	
	Central	-5.0%	-7.6%	
	South	-12.0%	-12.4%	
	Statewide	-8.8%	-10.2%	
Masonry Renters	Coastal	-8.9%	-9.8%	
	Inland	-6.2%	-9.4%	
	North	-2.1%	-2.7%	
	Central	-5.0%	-7.7%	
	South	-11.5%	-11.8%	
	Statewide	-8.6%	-9.8%	

Construction / Policy	Region	Percent Change in Loss Cost		
		5 Story	10 Story	20 Story
Commercial Residential	Coastal	-6.4%	-6.8%	-6.3%
	Inland	-13.3%	-13.4%	-12.4%
	North	-0.4%	-2.5%	-3.3%
	Central	-8.1%	-8.9%	-8.5%
	South	-7.5%	-7.7%	-7.0%
	Statewide	-7.0%	-7.5%	-7.1%

Appendix I – Form A-8: Probable Maximum Loss for Florida

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
~~November 1, 2014~~March 15, 2015

Form A-8: Probable Maximum Loss for Florida

Part A - Personal and Commercial Residential Probable Maximum Loss for Florida

RangeStart (Millions)	RangeEnd (Millions)	Total Loss (Millions)	Average Loss per Year (Millions)	Number of Hurricanes	Expected Annual Hurricane Losses (Millions)	Return Period (Years)
0	500	1,176,154.44	33.18	9,698	20.63	2.14
501	1000	1,925,811.27	727.82	3,823	33.79	2.83
1001	1500	2,112,738.74	1,231.20	2,655	37.07	3.17
1501	2000	1,900,556.43	1,724.64	1,798	33.34	3.43
2001	2500	1,765,014.86	2,242.71	1,295	30.97	3.63
2501	3000	1,661,859.45	2,724.36	1,052	29.16	3.80
3001	3500	1,640,601.43	3,261.63	847	28.78	3.94
3501	4000	1,671,228.64	3,764.03	802	29.32	4.08
4001	4500	1,870,412.52	4,260.62	779	32.81	4.21
4501	5000	1,879,267.50	4,757.64	718	32.97	4.35
5001	6000	4,128,376.03	5,497.17	1,368	72.43	4.55
6001	7000	4,325,310.77	6,494.46	1,211	75.88	4.82
7001	8000	5,079,935.07	7,481.49	1,264	89.12	5.11
8001	9000	5,362,632.19	8,498.62	1,184	94.08	5.42
9001	10000	5,586,039.05	9,500.07	1,111	98.00	5.77
10001	11000	6,133,309.48	10,520.26	1,116	107.60	6.12
11001	12000	6,101,729.99	11,491.02	981	107.05	6.53
12001	13000	6,264,766.42	12,504.52	989	109.91	6.92
13001	14000	6,809,929.78	13,485.01	1,008	119.47	7.39
14001	15000	7,041,085.97	14,487.83	955	123.53	7.88
15001	16000	6,420,104.03	15,507.50	847	112.63	8.40
16001	17000	6,801,481.35	16,468.48	815	119.32	8.95
17001	18000	7,685,142.10	17,466.23	863	134.83	9.60
18001	19000	7,934,118.31	18,494.45	849	139.20	10.34
19001	20000	7,049,678.79	19,474.25	746	123.68	11.14
20001	21000	6,318,196.67	20,513.63	639	110.85	11.92
21001	22000	5,678,505.63	21,509.49	559	99.62	12.68
22001	23000	6,348,297.33	22,511.69	572	111.37	13.48
23001	24000	6,294,809.11	23,488.09	559	110.44	14.45
24001	25000	6,103,482.26	24,511.98	502	107.08	15.46
25001	26000	6,495,576.82	25,472.85	544	113.96	16.58
26001	27000	5,295,454.41	26,477.27	456	92.90	17.75
27001	28000	5,605,858.20	27,479.70	437	98.35	18.98
28001	29000	5,416,840.15	28,509.69	415	95.03	20.29
29001	30000	4,834,605.60	29,479.30	367	84.82	21.62
30001	35000	23,515,796.34	32,390.90	1,710	412.56	26.15
35001	40000	19,667,139.58	37,390.00	1,258	345.04	36.52
40001	45000	17,393,005.39	42,318.75	1,056	305.14	52.63
45001	50000	12,455,422.84	47,359.02	694	218.52	76.00
50001	55000	11,151,657.84	52,355.20	559	195.64	111.33
55001	60000	7,520,154.66	57,405.76	342	131.93	167.16
60001	65000	5,690,986.51	62,538.31	246	99.84	242.55
65001	70000	4,730,917.54	67,584.54	194	83.00	363.06
70001	75000	2,678,541.35	72,393.01	112	46.99	558.82
75001	80000	1,853,756.63	77,239.86	70	32.52	826.09
80001	90000	2,728,345.54	85,260.80	98	47.87	1,390.24
90001	100000	1,030,587.56	93,689.78	29	18.08	2,850.00
100001	Maximum	1,799,340.25	112,458.77	49	31.57	14,250.00
Total		280,934,562.82		50,241		

Form A-8: Probable Maximum Loss for Florida

Part B - Personal and Commercial Residential Probable Maximum Loss for Florida

Return Period (Years)	Estimated Loss Level (Billion)	Uncertainty Interval (Billion)		
Top Event	$149.08		-	$1,838.70
1000	$81.93	$76.92	-	$85.00
500	$70.59	$68.96	-	$73.57
250	$63.01	$61.30	-	$64.41
100	$51.22	$50.21	-	$52.09
50	$41.66	$40.94	-	$42.39
20	$28.32	$27.76	-	$28.83
10	$18.05	$17.73	-	$18.37
5	$7.13	$6.85	-	$7.39

Form A-8: Probable Maximum Loss for Florida

Part A - Personal and Commercial Residential Probable Maximum Loss for Florida

RangeStart (Millions)	RangeEnd (Millions)	Total Loss (Millions)	Average Loss per Year (Millions)	Number of Hurricanes	Expected Annual Hurricane Losses (Millions)	Return Period (Years)
0	500	1,176,897.80	33.20	9,709	20.65	2.14
501	1000	1,929,450.99	727.82	3,828	33.85	2.83
1001	1500	2,110,045.75	1,231.07	2,653	37.02	3.17
1501	2000	1,890,287.17	1,723.14	1,794	33.16	3.44
2001	2500	1,774,733.36	2,240.82	1,300	31.14	3.64
2501	3000	1,637,690.15	2,720.42	1,041	28.73	3.81
3001	3500	1,646,498.15	3,253.95	851	28.89	3.95
3501	4000	1,711,853.55	3,762.32	822	30.03	4.08
4001	4500	1,849,235.48	4,260.91	768	32.44	4.22
4501	5000	1,888,178.69	4,756.12	726	33.13	4.35
5001	6000	4,109,039.44	5,493.37	1,360	72.09	4.55
6001	7000	4,362,449.07	6,491.74	1,220	76.53	4.83
7001	8000	5,061,486.22	7,476.35	1,259	88.80	5.12
8001	9000	5,386,200.74	8,495.58	1,197	94.49	5.43
9001	10000	5,615,454.34	9,501.61	1,106	98.52	5.78
10001	11000	6,088,709.97	10,515.91	1,120	106.82	6.12
11001	12000	6,087,054.20	11,485.01	983	106.79	6.54
12001	13000	6,402,733.80	12,505.34	1,003	112.33	6.94
13001	14000	6,827,398.82	13,492.88	1,017	119.78	7.41
14001	15000	6,851,831.68	14,485.90	927	120.21	7.90
15001	16000	6,717,917.12	15,514.82	881	117.86	8.43
16001	17000	6,866,033.59	16,504.89	830	120.46	9.00
17001	18000	7,466,840.37	17,486.75	836	131.00	9.64
18001	19000	7,841,783.61	18,494.77	843	137.58	10.37
19001	20000	7,108,977.78	19,476.65	752	124.72	11.20
20001	21000	6,051,552.60	20,513.74	606	106.17	11.95
21001	22000	5,760,846.45	21,495.70	566	101.07	12.70
22001	23000	6,462,152.41	22,516.21	590	113.37	13.53
23001	24000	6,083,646.95	23,488.98	529	106.73	14.48
24001	25000	6,273,633.38	24,506.38	520	110.06	15.48
25001	26000	6,525,628.91	25,490.74	554	114.48	16.69
26001	27000	5,168,450.84	26,504.88	445	90.67	17.82
27001	28000	5,525,215.13	27,488.63	425	96.93	19.03
28001	29000	5,330,741.51	28,506.64	411	93.52	20.33
29001	30000	5,042,837.76	29,490.28	389	88.47	21.67
30001	35000	23,130,796.86	32,396.07	1,674	405.80	26.19
35001	40000	19,660,922.34	37,378.18	1,260	344.93	36.59
40001	45000	17,298,620.00	42,294.91	1,048	303.48	52.68
45001	50000	12,249,982.03	47,297.23	688	214.91	75.60
50001	55000	11,457,546.70	52,317.56	572	201.01	110.68
55001	60000	7,290,348.50	57,404.32	333	127.90	166.18
60001	65000	5,939,347.04	62,519.44	256	104.20	240.51
65001	70000	4,597,805.71	67,614.79	184	80.66	358.49
70001	75000	2,749,155.67	72,346.20	116	48.23	553.40
75001	80000	1,932,198.16	77,287.93	73	33.90	826.09
80001	90000	2,645,265.03	85,331.13	96	46.41	1,357.14
90001	100000	1,123,601.19	93,633.43	31	19.71	2,850.00
100001	Maximum	1,804,482.15	112,780.13	49	31.66	14,250.00
Total		280,513,559.16		50,241		

Form A-8: Probable Maximum Loss for Florida

Part B - Personal and Commercial Residential Probable Maximum Loss for Florida

Return Period (Years)	Estimated Loss Level (Billion)	Uncertainty Interval (Billion)		
Top Event	$149.24		-	$1,838.34
1000	$82.08	$77.09	-	$85.04
500	$70.67	$69.16	-	$73.79
250	$63.09	$61.31	-	$64.47
100	$51.22	$50.27	-	$52.14
50	$41.62	$40.89	-	$42.38
20	$28.28	$27.71	-	$28.79
10	$18.01	$17.67	-	$18.32
5	$7.11	$6.82	-	$7.36

Appendix J – Form G1 − G7

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
~~January 5, 2015~~March 15, 2015

Form G-1

I hereby certify that I have reviewed the current submission of FLORIDA PUBLIC HURRICANE LOSS MODEL (Name of Model)
Version 6-0 for compliance with the 2013 Standards adopted by the Florida Commission on Hurricane Loss Projection Methodology and hereby certify that:

1) The model meets the Statistical Standards (S1 – S6);
2) The disclosures and forms related to the Statistical Standards section are editorially and technically accurate, reliable, unbiased, and complete;
3) My review was completed in accordance with the professional standards and code of ethical conduct for my profession; and
4) In expressing my opinion I have not been influenced by any other party in order to bias or prejudice my opinion.

SHAHID HAMID	PHD (Financial Economics)
Name	Professional Credentials (area of expertise)
S. Hamid	10/28/2014
Signature (original submission)	Date
Signature (response to deficiencies, if any)	Date
Signature (revisions to submission, if any)	Date
Signature (final submission)	Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

Signature (revisions to submission)	Date

Note: A facsimile or any properly reproduced signature will be acceptable to meet this requirement.

Form G-1

I hereby certify that I have reviewed the current submission of FLORIDA PUBLIC HURRICANE LOSS MODEL
(Name of Model)
Version 6.0 6.1 for compliance with the 2013 Standards adopted by the Florida Commission on Hurricane Loss Projection Methodology and hereby certify that:

1) The model meets the Statistical Standards (S1 – S6);
2) The disclosures and forms related to the Statistical Standards section are editorially and technically accurate, reliable, unbiased, and complete;
3) My review was completed in accordance with the professional standards and code of ethical conduct for my profession; and
4) In expressing my opinion I have not been influenced by any other party in order to bias or prejudice my opinion.

SHAHID HAMID	PHD (Financial Economics)
Name	Professional Credentials (area of expertise)
S. Hamid	10/23/2014
Signature (original submission)	Date
S. Hamid	1/5/2015
Signature (response to deficiencies, if any)	Date
S. Hamid	4/5/2015
Signature (revisions to submission, if any)	Date
Signature (final submission)	Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

Signature (revisions to submission)	Date

Note: A facsimile or any properly reproduced signature will be acceptable to meet this requirement.

Form G-2

Form G-2: Meteorological Standards Expert Certification

I hereby certify that I have reviewed the current submission of _____
(Name of Model)
Version 6.0 for compliance with the 2013 Standards adopted by the Florida Commission on Hurricane Loss Projection Methodology and hereby certify that:

1) The model meets the Meteorological Standards (M1 – M6);
2) The disclosures and forms related to the Meteorological Standards section are editorially and technically accurate, reliable, unbiased, and complete;
3) My review was completed in accordance with the professional standards and code of ethical conduct for my profession; and
4) In expressing my opinion I have not been influenced by any other party in order to bias or prejudice my opinion.

MARK D. POWELL	Meteorology
Name	Professional Credentials (area of expertise)
[signature]	10-23-2014
Signature (original submission)	Date
Signature (response to deficiencies, if any)	Date
Signature (revisions to submission, if any)	Date
Signature (final submission)	Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

Signature (revisions to submission)	Date

Note: A facsimile or any properly reproduced signature will be acceptable to meet this requirement.

Form G-2: Meteorological Standards Expert Certification

I hereby certify that I have reviewed the current submission of Florida Public Hurricane Loss Model (Name of Model)
Version 6.0 for compliance with the 2013 Standards adopted by the Florida Commission on Hurricane Loss Projection Methodology and hereby certify that:

1) The model meets the Meteorological Standards (M1 – M6);
2) The disclosures and forms related to the Meteorological Standards section are editorially and technically accurate, reliable, unbiased, and complete;
3) My review was completed in accordance with the professional standards and code of ethical conduct for my profession; and
4) In expressing my opinion I have not been influenced by any other party in order to bias or prejudice my opinion.

MARK D. POWELL	Meteorology PhD in Meteorology Certified Consulting Meteorologist AMS Fellow
Name	Professional Credentials (area of expertise)
[signature]	10-23-2014
Signature (original submission)	Date
[signature]	12-22-2014
Signature (response to deficiencies, if any)	Date
[signature]	3-30-2015
Signature (revisions to submission, if any)	Date
Signature (final submission)	Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

Signature (revisions to submission)	Date

Note: A facsimile or any properly reproduced signature will be acceptable to meet this requirement.

Form G-3

Form G-3: Statistical Standards Expert Certification

I hereby certify that I have reviewed the current submission of FPHLM (Name of Model) Version 6.0 for compliance with the 2013 Standards adopted by the Florida Commission on Hurricane Loss Projection Methodology and hereby certify that:

1) The model meets the Statistical Standards (S1 – S6);
2) The disclosures and forms related to the Statistical Standards section are editorially and technically accurate, reliable, unbiased, and complete;
3) My review was completed in accordance with the professional standards and code of ethical conduct for my profession; and
4) In expressing my opinion I have not been influenced by any other party in order to bias or prejudice my opinion.

SNEH GULATI	STATISTICS
Name	Professional Credentials (area of expertise)
[signature]	10/27/2014
Signature (original submission)	Date
Signature (response to deficiencies, if any)	Date
Signature (revisions to submission, if any)	Date
Signature (final submission)	Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

Signature (revisions to submission)	Date

Note: A facsimile or any properly reproduced signature will be acceptable to meet this requirement.

Form G-3: Statistical Standards Expert Certification

I hereby certify that I have reviewed the current submission of FPHLM
(Name of Model)
Version 6.0 for compliance with the 2013 Standards adopted by the Florida Commission on Hurricane Loss Projection Methodology and hereby certify that:

1) The model meets the Statistical Standards (S1 – S6);
2) The disclosures and forms related to the Statistical Standards section are editorially and technically accurate, reliable, unbiased, and complete;
3) My review was completed in accordance with the professional standards and code of ethical conduct for my profession; and
4) In expressing my opinion I have not been influenced by any other party in order to bias or prejudice my opinion.

SNEH GULATI	Ph.D in STATISTICS
Name	Professional Credentials (area of expertise)
[signature]	10/27/2014
Signature (original submission)	Date
[signature]	12/22/2014
Signature (response to deficiencies, if any)	Date
[signature]	04/03/2015
Signature (revisions to submission, if any)	Date
Signature (final submission)	Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

Signature (revisions to submission)	Date

Note: A facsimile or any properly reproduced signature will be acceptable to meet this requirement.

Form G-4

Form G-4: Vulnerability Standards Expert Certification

I hereby certify that I have reviewed the current submission of FPHLM (Name of Model)
Version 6.0 for compliance with the 2013 Standards adopted by the Florida Commission on Hurricane Loss Projection Methodology and hereby certify that:

1) The model meets the Vulnerability Standards (V1 – V3);
2) The disclosures and forms related to the Vulnerability Standards section are editorially and technically accurate, reliable, unbiased, and complete;
3) My review was completed in accordance with the professional standards and code of ethical conduct for my profession; and
4) In expressing my opinion I have not been influenced by any other party in order to bias or prejudice my opinion.

JEAN-PAUL PINELLI	structural & wind engineering Ph.D., Florida P.E. license 53310
Name	Professional Credentials (area of expertise)
Signature (original submission)	10/17/14 Date
Signature (response to deficiencies, if any)	Date
Signature (revisions to submission, if any)	Date
Signature (final submission)	Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

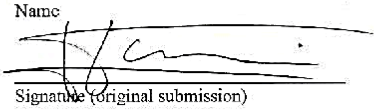
Signature (revisions to submission)	Date

Note: A facsimile or any properly reproduced signature will be acceptable to meet this requirement.

Form G-4: Vulnerability Standards Expert Certification

I hereby certify that I have reviewed the current submission of FPHLM (Name of Model)
Version 6.0 for compliance with the 2013 Standards adopted by the Florida Commission on Hurricane Loss Projection Methodology and hereby certify that:

1) The model meets the Vulnerability Standards (V1 – V3);
2) The disclosures and forms related to the Vulnerability Standards section are editorially and technically accurate, reliable, unbiased, and complete;
3) My review was completed in accordance with the professional standards and code of ethical conduct for my profession; and
4) In expressing my opinion I have not been influenced by any other party in order to bias or prejudice my opinion.

JEAN-PAUL PINELLI Name	structural & wind engineering Ph.D., Florida P.E. license 53310 Professional Credentials (area of expertise)
 Signature (original submission)	10/17/14 Date
Signature (response to deficiencies, if any)	Date
 Signature (revisions to submission, if any)	4/1/2015 Date
Signature (final submission)	Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

Signature (revisions to submission)	Date

Note: A facsimile or any properly reproduced signature will be acceptable to meet this requirement.

Form G-5

Form G-5: Actuarial Standards Expert Certification

I hereby certify that I have reviewed the current submission of FLORIDA PUBLIC HURRICANE LOSS MODEL
(Name of Model)
Version 6.0 for compliance with the 2013 Standards adopted by the Florida Commission on Hurricane Loss Projection Methodology and hereby certify that:

1) The model meets the Actuarial Standards (A1 – A6);
2) The disclosures and forms related to the Actuarial Standards section are editorially and technically accurate, reliable, unbiased, and complete;
3) My review was completed in accordance with the Actuarial Standards of Practice and Code of Conduct; and
4) In expressing my opinion I have not been influenced by any other party in order to bias or prejudice my opinion.

GAIL FLANNERY Name	FCAS, MAAA Professional Credentials (area of expertise)
Gail Flannery ~~Signature (original submission)~~	10/28/14 ~~Date~~
Signature (response to deficiencies, if any)	Date
Signature (revisions to submission, if any)	Date
Signature (final submission)	Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

Signature (revisions to submission)	Date

Note: A facsimile or any properly reproduced signature will be acceptable to meet this requirement.

Form G-5: Actuarial Standards Expert Certification

I hereby certify that I have reviewed the current submission of FLORIDA PUBLIC HURRICANE LOSS MODEL
(Name of Model)
Version 6.0 for compliance with the 2013 Standards adopted by the Florida Commission on Hurricane Loss Projection Methodology and hereby certify that:

1) The model meets the Actuarial Standards (A1 – A6);
2) The disclosures and forms related to the Actuarial Standards section are editorially and technically accurate, reliable, unbiased, and complete;
3) My review was completed in accordance with the Actuarial Standards of Practice and Code of Conduct; and
4) In expressing my opinion I have not been influenced by any other party in order to bias or prejudice my opinion.

GAIL FLANNERY Name	FCAS, MAAA Professional Credentials (area of expertise)
Gail Flannery Signature (original submission)	10/28/14 Date
 Signature (response to deficiencies, if any)	 Date
Gail Flannery Signature (revisions to submission, if any)	4/1/15 Date
 Signature (final submission)	 Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

 Signature (revisions to submission)	 Date

Note: A facsimile or any properly reproduced signature will be acceptable to meet this requirement.

Form G-6

Form G-6: Computer Standards Expert Certification

I hereby certify that I have reviewed the current submission of Florida Public Hurricane ~~Loss Model~~ (Name of Model)

Version 6.0 for compliance with the 2013 Standards adopted by the Florida Commission on Hurricane Loss Projection Methodology and hereby certify that:

1) The model meets the Computer Standards (C1 – C7);
2) The disclosures and forms related to the Computer Standards section are editorially and technically accurate, reliable, unbiased, and complete;
3) My review was completed in accordance with the professional standards and code of ethical conduct for my profession; and
4) In expressing my opinion I have not been influenced by any other party in order to bias or prejudice my opinion.

Shu-Ching Chen	Computer Science
Name	Professional Credentials (area of expertise)
[signature]	10/22/2014
Signature (original submission)	Date
Signature (response to deficiencies, if any)	Date
Signature (revisions to submission, if any)	Date
Signature (final submission)	Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

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3) My review was completed in accordance with the professional standards and code of ethical conduct for my profession; and
4) In expressing my opinion I have not been influenced by any other party in order to bias or prejudice my opinion.

Shu-Ching Chen	Ph.D. in Electrical and Computer Engineering & M.S. in Computer Science
Name	Professional Credentials (area of expertise)
[signature]	10/22/2014
Signature (original submission)	Date
[signature]	12/22/2014
Signature (response to deficiencies, if any)	Date
[signature]	4/6/2015
Signature (revisions to submission, if any)	Date
Signature (final submission)	Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

Signature (revisions to submission)	Date

Note: A facsimile or any properly reproduced signature will be acceptable to meet this requirement.

Form G-7

Form G-7: Editorial Certification

I/We hereby certify that I/we have reviewed the current submission of FPHLM (Name of Model) Version 6.0 for compliance with the "Process for Determining the Acceptability of a Computer Simulation Model" adopted by the Florida Commission on Hurricane Loss Projection Methodology in its *Report of Activities as of November 1, 2013*, and hereby certify that:

1) The model submission is in compliance with the Commission's Notification Requirements and General Standard G-5 (Editorial Compliance);
2) The disclosures and forms related to each standards section are editorially accurate and contain complete information and any changes that have been made to the submission during the review process have been reviewed for completeness, grammatical correctness, and typographical errors;
3) There are no incomplete responses, inaccurate citations, charts or graphs, or extraneous text or references;
4) The current version of the model submission has been reviewed for grammatical correctness, typographical errors, completeness, the exclusion of extraneous data/ information and is otherwise acceptable for publication; and
5) In expressing my/our opinion I/we have not been influenced by any other party in order to bias or prejudice my/our opinion.

Steven Cocke	Ph.D. Physics
Name	Professional Credentials (area of expertise)
Signature (original submission)	10/29/2014 Date
Signature (response to deficiencies, if any)	Date
Signature (revisions to submission, if any)	Date
Signature (final submission)	Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

Signature (revisions to submission)	Date

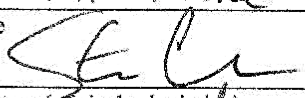
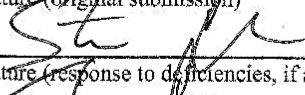
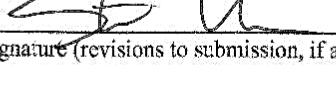
Note: A facsimile or any properly reproduced signature will be acceptable to meet this requirement.

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Form G-7: Editorial Certification

I/We hereby certify that I/we have reviewed the current submission of FPHLM (Name of Model)
Version 6.0 for compliance with the "Process for Determining the Acceptability of a Computer Simulation Model" adopted by the Florida Commission on Hurricane Loss Projection Methodology in its *Report of Activities as of November 1, 2013*, and hereby certify that:

1) The model submission is in compliance with the Commission's Notification Requirements and General Standard G-5 (Editorial Compliance);
2) The disclosures and forms related to each standards section are editorially accurate and contain complete information and any changes that have been made to the submission during the review process have been reviewed for completeness, grammatical correctness, and typographical errors;
3) There are no incomplete responses, inaccurate citations, charts or graphs, or extraneous text or references;
4) The current version of the model submission has been reviewed for grammatical correctness, typographical errors, completeness, the exclusion of extraneous data/ information and is otherwise acceptable for publication; and
5) In expressing my/our opinion I/we have not been influenced by any other party in order to bias or prejudice my/our opinion.

Steven Cocke	Ph.D Physics
Name	Professional Credentials (area of expertise)
	10/29/2014
Signature (original submission)	Date
	1/5/2015
Signature (response to deficiencies, if any)	Date
	4/6/2015
Signature (revisions to submission, if any)	Date
Signature (final submission)	Date

An updated signature and form is required following any modification of the model and any revision of the original submission. If a signatory differs from the original signatory, provide the printed name and professional credentials for any new signatories. Additional signature lines shall be added as necessary with the following format:

Signature (revisions to submission)	Date

Note: A facsimile or any properly reproduced signature will be acceptable to meet this requirement.

Appendix K – Form M-1: Annual Occurrence Rates

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
November 1, 2014

Form M-1. Modeled Annual Occurrence Rates

	Entire State				Region A – NW Florida			
	Historical		Modeled		Historical		Modeled	
Category	Number	Rate	Number	Rate	Number	Rate	Number	Rate
1	22	0.19	27.08	0.24	13	0.11	15.27	0.13
2	17	0.15	13.79	0.12	6	0.05	5.98	0.05
3	15	0.13	13.56	0.12	6	0.05	4.14	0.04
4	8	0.07	7.35	0.06	0	0.00	1.65	0.01
5	2	0.02	1.79	0.02	0	0.00	0.18	0.00
	Region B – SW Florida				Region C – SE Florida			
	Historical		Modeled		Historical		Modeled	
Category	Number	Rate	Number	Rate	Number	Rate	Number	Rate
1	7	0.06	8.21	0.07	6	0.05	6.91	0.06
2	4	0.04	4.97	0.04	6	0.05	4.01	0.04
3	6	0.05	4.81	0.04	4	0.04	5.14	0.05
4	2	0.02	2.07	0.02	6	0.05	3.85	0.03
5	1	0.01	0.45	0.00	1	0.01	1.22	0.01
	Region D – NE Florida				Florida By-Passing Hurricanes			
	Historical		Modeled		Historical		Modeled	
Category	Number	Rate	Number	Rate	Number	Rate	Number	Rate
1	1	0.01	1.17	0.01	5	0.04	5.76	0.05
2	2	0.02	0.63	0.01	3	0.03	2.78	0.02
3	0	0.00	0.54	0.00	5	0.04	3.17	0.03
4	0	0.00	0.16	0.00	1	0.01	1.68	0.01
5	0	0.00	0.02	0.00	0	0.00	0.51	0.00
	Region E – Georgia				Region F – Alabama/Mississippi			
	Historical		Modeled		Historical		Modeled	
Category	Number	Rate	Number	Rate	Number	Rate	Number	Rate
1	2	0.02	1.36	0.01	6	0.05	5.21	0.05
2	1	0.01	0.70	0.01	2	0.02	2.58	0.02
3	0	0.00	0.38	0.00	3	0.03	2.91	0.03
4	0	0.00	0.17	0.00	1	0.01	1.38	0.01
5	0	0.00	0.08	0.00	1	0.01	0.33	0.00

	Entire State Landfalls			
	Historical		Modeled	
Category	Number	Rate	Number	Rate
1	27	0.24	31.56	0.28
2	18	0.16	15.59	0.14
3	16	0.14	14.64	0.13
4	8	0.07	7.72	0.07
5	2	0.02	1.87	0.02

Appendix L – Form M-2: Maps of Maximum Winds

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
November 1, 2014

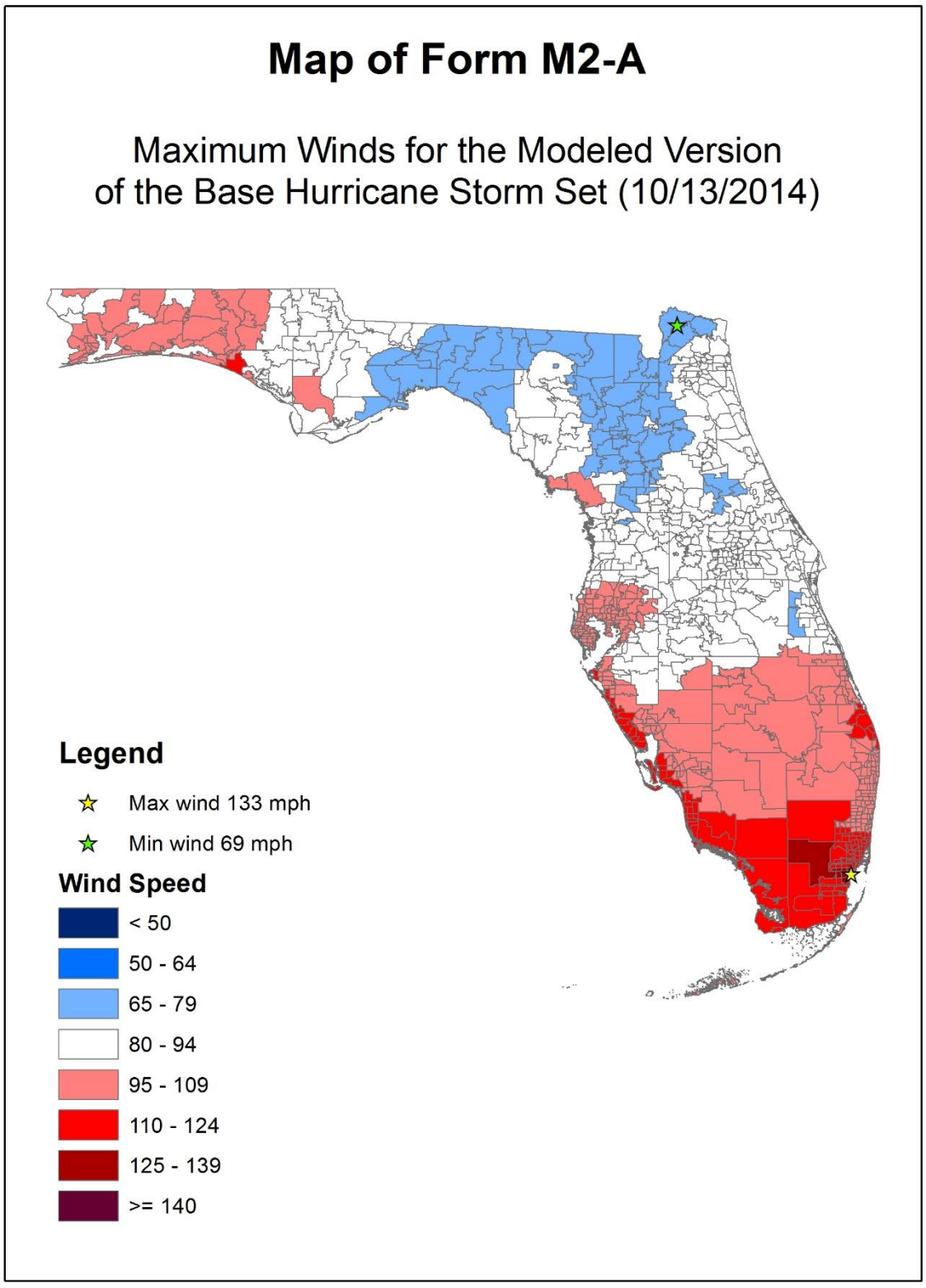


Figure 94. Maximum ZIP Code wind speed for open terrain wind exposure based on simulations of the historical storm set.

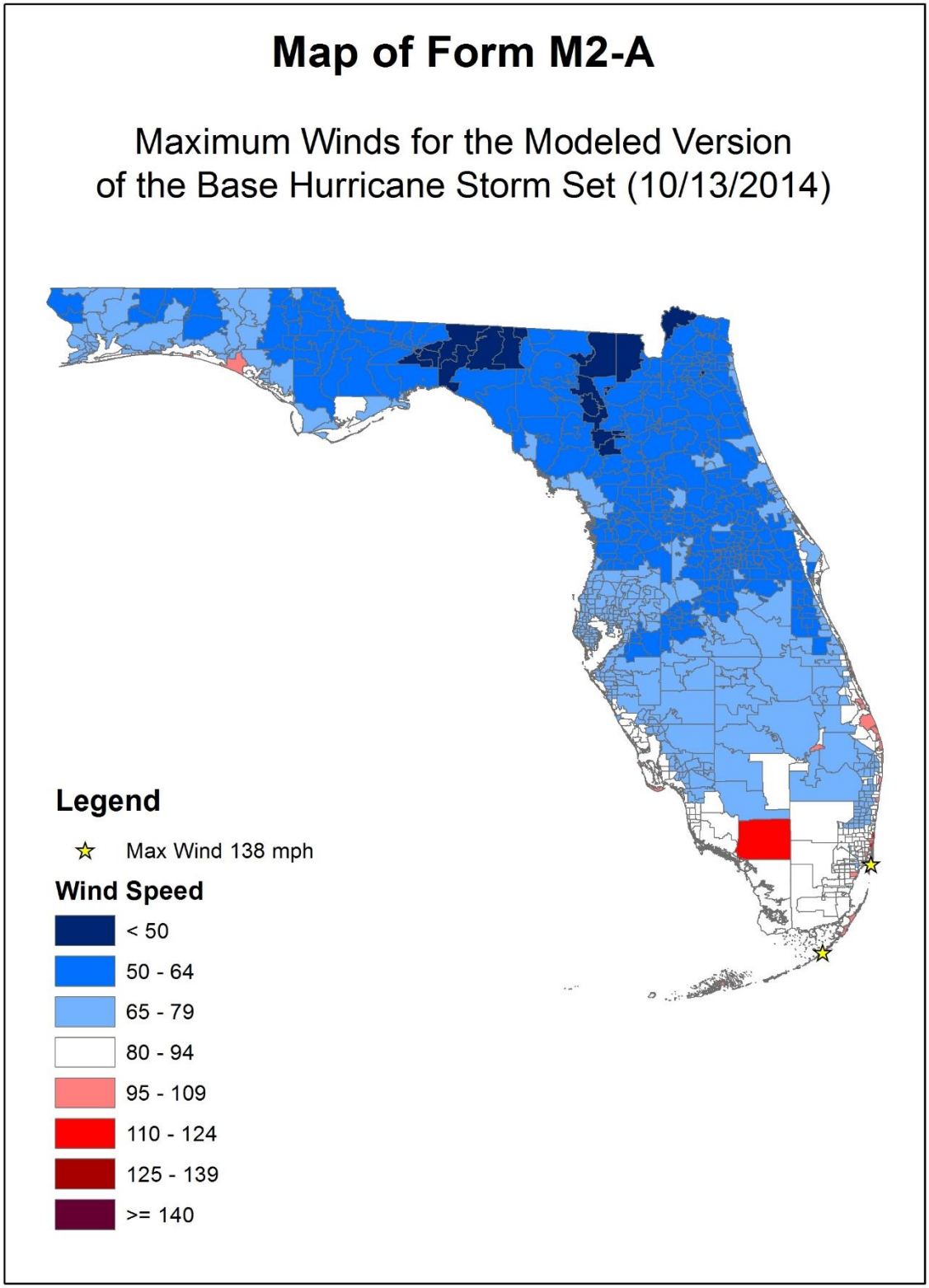
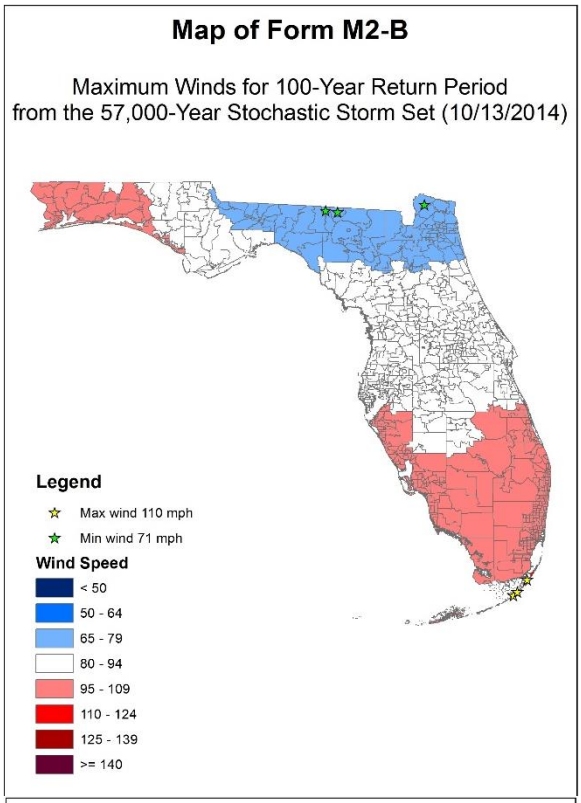


Figure 95. Maximum ZIP Code wind speed for actual terrain wind exposure based on simulations of the historical storm set. Note that winds below 50 mph were not saved for this calculation, and thus the minimum wind cannot be determined.



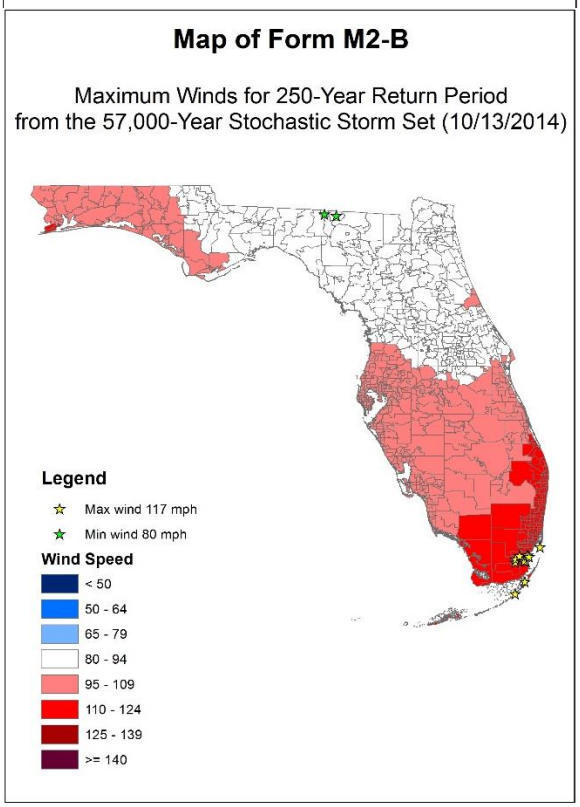
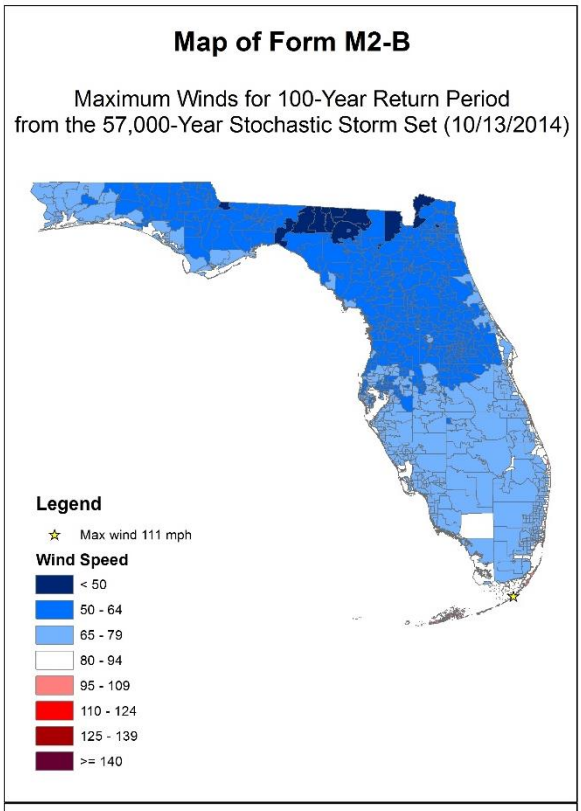


Figure 96. 100- and 250-year return period wind speeds at Florida ZIP Codes for open terrain wind exposure.



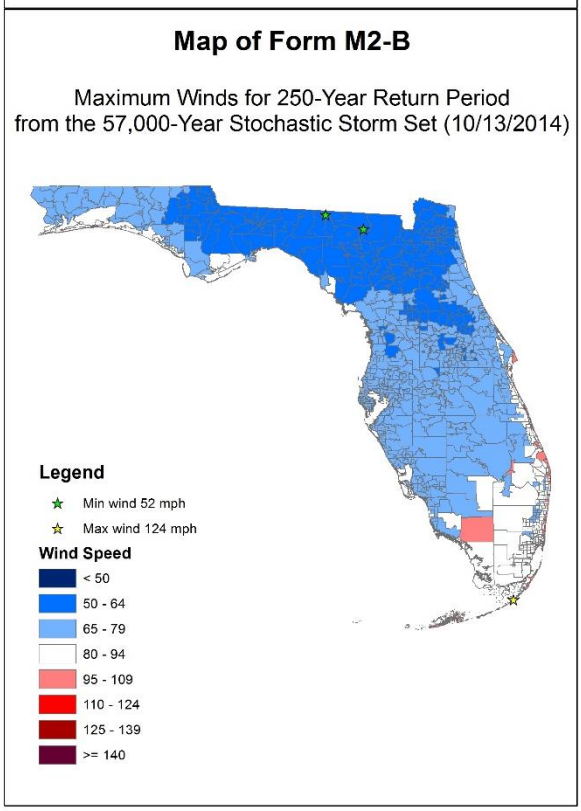


Figure 97. 100- and 250-year return period wind speeds at Florida ZIP Codes for actual terrain wind exposure. Note that winds below 50 mph were not saved for this calculation, and thus the minimum wind cannot be determined.

Appendix M – Form M-3: Radius of Maximum Winds and Radii of Standard Wind Thresholds

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
November 1, 2014

Table 32. Range of outer wind radii (sm) as a function of central sea level pressure (mb).

Central Pressure (mb)	Rmax (sm)		Outer Radii (>110 mph) (sm)		Outer Radii (>73 mph) (sm)		Outer Radii (>40 mph) (sm)	
	Min	Max	Min	Max	Min	Max	Min	Max
900	3.52	19.69	8.21	41.80	13.30	78.24	25.03	186.22
910	3.35	20.83	7.82	41.80	12.91	81.86	25.03	211.03
920	3.52	36.04	7.43	66.07	12.12	129.15	24.25	258.21
930	5.19	72.12	10.96	84.85	18.08	199.40	35.62	448.46
940	6.66	91.91	12.11	72.93	20.59	202.85	44.83	433.95
950	7.57	87.32	12.29	69.30	21.34	169.51	47.06	394.28
960	6.14	82.27	8.94	60.36	16.76	144.86	39.68	356.91
970	6.14	81.37	6.14	51.30	12.29	137.32	28.50	367.30
980	6.14	75.11	7.26	29.50	9.83	97.69	24.59	318.57
990	6.14	71.98	NA	NA	6.76	76.23	20.12	314.32

Table 33. Extended Best-Track and H*Wind wind radii ranges based on Atlantic basin hurricanes.

Storms Ext. Best Track (DeMaria 2010)	Central Pressure (mb)	Rmax (sm)		Outer Radii (>110 mph) (sm) From H*Wind		Outer Radii (>73 mph) (sm)		Outer Radii (>40 mph) (sm)	
		min	max	min	max	min	max	min	max
Katrina, Rita, Wilma 2005	900	6	23	21 Rita	31 Wilma	69	103	102	230
Mitch 98, Ivan 04, Katrina, Wilma 05	910	14	29	33 Wilma	34.5 Mitch	57	115	172	230
Isabel 03, Ivan 04, Rita 05, Dean 07	920	11	29	26 Ivan	34.5 Isabel	34	144	161	287
Andrew 92, Floyd 99, Ivan 04, Dennis 05, Dean 07	930	11	34	22 Andrew	32 Ivan	29	126.5	115	287
Luis 95, Lili 02, Floyd 99	940	6	40	16Lili	55 Isabel	29	138.0	115	287
Gabrielle 89, Iris 01, Bret 99	950	6	63	7 Iris	46 Wilma	17	172	98	345
Gustav 02, Dennis 05, Gilbert 88, Claudette 01	960	6	86	13 Gustav	n/a	23	161	86	402
Joan 88, Felix 95, Lili 96, Karl 10	970	6	103	6 Karl	n/a	17	287	57	621
Gabrielle 01, Emily 05, Noel 07, Beta 05	980	11	138	n/a	n/a	17	144	57	690
Lili 02, Olga 01, Lisa 04	990	11	207	n/a	n/a	17	138.0	34	632

Appendix N – Form S-1: Probability and Frequency of Florida Landfalling Hurricanes per Year

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
November 1, 2014

Model Results
Probability and Frequency of Florida Landfalling Hurricanes per Year

Number Of Hurricanes Per Year	Historical Probabilities	Modeled Probabilities	Historical Frequencies	Modeled Frequencies
0	0.6140	0.6281	70	72
1	0.2368	0.2372	27	27
2	0.1228	0.0924	14	10
3	0.0263	0.0341	3	4
4	0.0000	0.0078	0	1
5	0.0000	0.0004	0	0
6	0.0000	0.0000	0	0
7	0.0000	0.0000	0	0
8	0.0000	0.0000	0	0
9	0.0000	0.0000	0	0
10 or more	0.0000	0.0000	0	0

Note: Historical and modeled frequencies are the number of occurrences in a 114 year

Appendix O – Form S-2A: Examples of Loss Exceedance Estimates (2007 FHCF Exposure Data)

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
~~November 1, 2014~~February 3, 2015

Part A

Return Period (Years)	Probability of Exceedance	Estimated Loss Notional Risk Data Set	Estimated Personal and Commercial Residential Loss FHCF Data Set
Top Event	NA	$74,944,743	$161,188,324,301
10000	0.01%	$62,108,674	$119,926,973,207
5000	0.02%	$56,415,817	$112,079,258,013
2000	0.05%	$48,481,152	$94,752,851,302
1000	0.10%	$43,545,403	$87,006,706,865
500	0.20%	$38,311,517	$75,122,095,408
250	0.40%	$33,498,855	$67,066,612,989
100	1.00%	$26,507,007	$54,736,801,014
50	2.00%	$21,362,926	$44,361,285,219
20	5.00%	$14,137,594	$30,028,683,413
10	10.00%	$8,525,494	$19,087,304,505
5	20.00%	$3,067,170	$7,426,284,740

Part B

Mean (Total Average Annual Loss)	$2,418,765	$5,215,999,807
Median	$0	$1,537
Standard Deviation	$5,549,370	$11,566,844,815
Interquartile Range	$1,625,907	$3,583,103,977
Sample Size	57000	57000

Part A

Return Period (Years)	Probability of Exceedance	Estimated Loss Notional Risk Data Set	Estimated Personal and Commercial Residential Loss FHCF Data Set
Top Event	NA	$74,944,743	$161,881,976,026
10000	0.01%	$62,108,674	$120,159,927,763
5000	0.02%	$56,415,817	$112,302,882,613
2000	0.05%	$48,481,152	$95,256,160,840
1000	0.10%	$43,545,403	$87,408,015,497
500	0.20%	$38,311,517	$75,462,336,175
250	0.40%	$33,498,855	$67,186,332,606
100	1.00%	$26,507,007	$54,816,137,220
50	2.00%	$21,362,926	$44,374,357,779
20	5.00%	$14,137,594	$30,013,955,633
10	10.00%	$8,525,494	$19,038,406,531
5	20.00%	$3,067,170	$7,401,398,632

Part B

Mean (Total Average Annual Loss)	$2,418,765	$5,210,672,383
Median	$0	$1,537
Standard Deviation	$5,549,370	$11,569,797,644
Interquartile Range	$1,625,907	$3,571,588,480
Sample Size	57000	57000

Appendix P – Form S-2B: Examples of Loss Exceedance Estimates (2012 FHCF Exposure Data)

Florida International University

Florida Public Hurricane Loss Model ~~6.0~~6.1

~~November 1, 2014~~March 15, 2015

Part A

Return Period (Years)	Probability of Exceedance	Estimated Loss Notional Risk Data Set	Estimated Personal and Commercial Residential Loss FHCF Data Set
Top Event	NA	$74,944,743	$149,082,126,453
10000	0.01%	$62,108,674	$110,236,675,637
5000	0.02%	$56,415,817	$103,093,769,689
2000	0.05%	$48,481,152	$88,726,426,770
1000	0.10%	$43,545,403	$81,929,996,926
500	0.20%	$38,311,517	$70,593,232,417
250	0.40%	$33,498,855	$63,014,590,531
100	1.00%	$26,507,007	$51,220,150,596
50	2.00%	$21,362,926	$41,661,348,371
20	5.00%	$14,137,594	$28,316,328,171
10	10.00%	$8,525,494	$18,053,148,377
5	20.00%	$3,067,170	$7,132,720,438

Part B

Mean (Total Average Annual Loss)	$2,418,765	$4,928,676,541
Median	$0	$1,664
Standard Deviation	$5,549,370	$10,862,493,939
Interquartile Range	$1,625,907	$3,451,849,482
Sample Size	57000	57000

Part A

Return Period (Years)	Probability of Exceedance	Estimated Loss Notional Risk Data Set	Estimated Personal and Commercial Residential Loss FHCF Data Set
Top Event	NA	$74,944,743	$149,236,026,414.33
10000	0.01%	$62,108,674	$110,363,047,526.39
5000	0.02%	$56,415,817	$103,653,454,759.56
2000	0.05%	$48,481,152	$89,171,511,950.63
1000	0.10%	$43,545,403	$82,077,582,104.74
500	0.20%	$38,311,517	$70,668,053,017.36
250	0.40%	$33,498,855	$63,092,975,549.47
100	1.00%	$26,507,007	$51,219,389,336.64
50	2.00%	$21,362,926	$41,624,547,108.07
20	5.00%	$14,137,594	$28,276,495,761.84
10	10.00%	$8,525,494	$18,005,689,210.46
5	20.00%	$3,067,170	$7,110,085,466.45

Part B

Mean (Total Average Annual Loss)	$2,418,765	$4921290491
Median	$0	$1,661
Standard Deviation	$5,549,370	$10860375041
Interquartile Range	$1,625,907	$3440709519
Sample Size	57000	57000

Appendix Q – Form S-3: Distributions of Stochastic Hurricane Parameters

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
November 1, 2014

Stochastic Hurricane Parameter (Function or Variable)	Functional Form of Distribution	Data Source	Year Range Used	Justification for Functional Form
Holland B Error term	Normal	Willoughby and Rahn (2004)	1977-2000	The Gaussian Distribution provided a good fit for the error term. See Standard S-1, Disclosure 1.
Rmax	Gamma	Ho et al. (1987) , supplemented by the extended best track data of DeMaria (Penington 2000), NOAA HRD research flight data, and NOAA-HRD H*Wind analyses (Powell et al. 1996, 1998).	1901-2012	Rmax is skewed, nonnegative and does not have a long tail. So the gamma distribution was tried and found to be a good fit. We limit the range of Rmax to the interval (4, 120). See Standard S-1, Disclosure 1.
Pressure decay Term	Normal	Vickery (2005)	1979-1996	From Vickery (2005)
Storm initial location perturbation	Uniform	N/A	N/A	Plausible variations in initial storm locations are assumed to be uniform
Storm initial motion perturbation	Uniform	N/A	N/A	Plausible variations in initial storm motion are assumed to be uniform
Storm change in motion and intensity distributions	Empirical	HURDAT	1900-2013	Sampling from historical data See Standard G-1, Disclosure 2 for details

Appendix R – Form S-4: Validation Comparisons

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
~~November 1, 2014~~March 15, 2015

Personal Residential:

Comparison #1: Hurricane Charley and Company O by Coverage

	Company Actual	Modeled	Difference
Coverage	Loss/Exposure	Loss/Exposure	
Building	0.00764	0.00927	-0.00163
Contents	0.00007	0.00247	-0.00241
Appurtenants	0.00107	0.01045	-0.00938
ALE	0.00025	0.00175	-0.00150
Total	0.00424	0.00651	-0.00226

Comparison #2: Different Companies by Different Hurricanes

		Company Actual	Modeled	Difference
Company	Event	Loss/Exposure	Loss/Exposure	
K	Jeanne	0.01370	0.01476	-0.00106
Q	Wilma	0.01481	0.00916	0.00565
A	Charley	0.01885	0.01364	0.00521
O	Frances	0.00245	0.00451	-0.00206
O	Charley	0.00424	0.00651	-0.00226

Comparison #3: Company O by Hurricane Frances, Charley, Jeanne

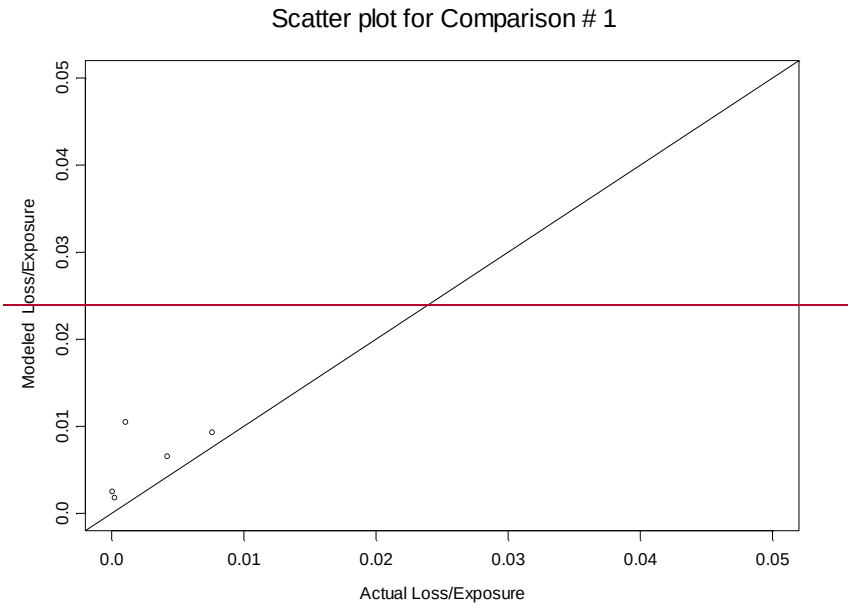
		Company Actual	Modeled	Difference
Company	Event	Loss/Exposure	Loss/Exposure	
O	Frances	0.00245	0.00451	-0.00206
O	Charley	0.00424	0.00651	-0.00226
O	Jeanne	0.00143	0.00434	-0.00291

Comparison #4: Construction Type for Hurricane Charley

Construction	Company	Company Actual	Modeled	Difference
		Loss/Exposure	Loss/Exposure	
Frame	B	0.01363	0.01694	-0.00331
Masonry	B	0.01584	0.01685	-0.00101
Manufactured	O	0.05476	0.03724	0.01752
Other	A	0.01803	0.01448	0.00355

Comparison #5: County wise for Company A and Hurricane Frances

	Company Actual	Modeled	Difference
Lee	0.000019	0.000025	-0.000007
Sarasota	0.000122	0.000260	-0.000138
Collier	0.000031	0.000081	-0.000051
Madison	0.000865	0.000931	-0.000066
Manatee	0.000257	0.000456	-0.000199



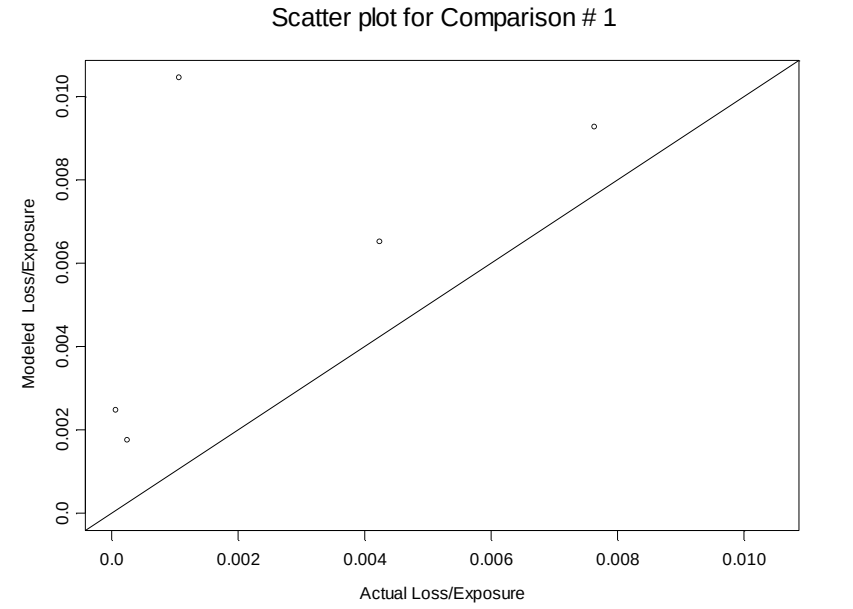
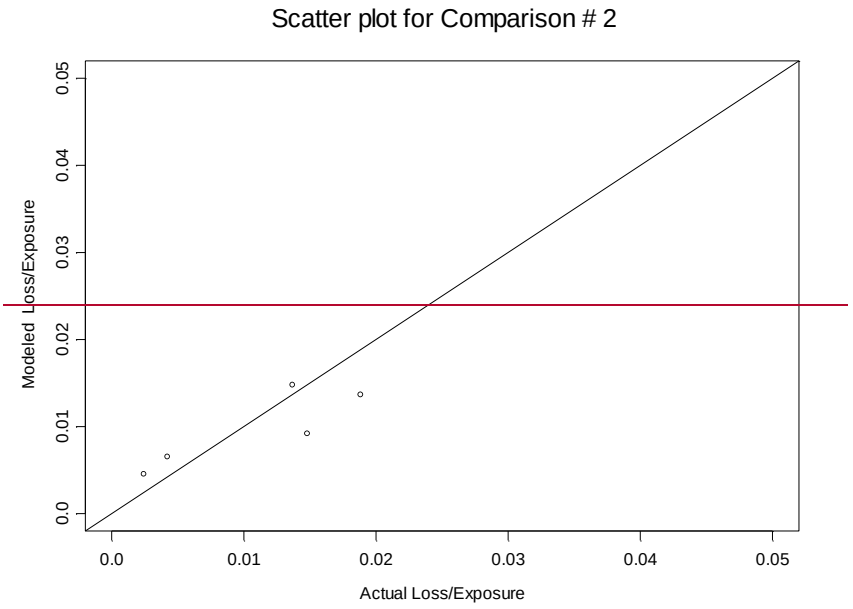


Figure 98. Scatter plot for comparison # 1.



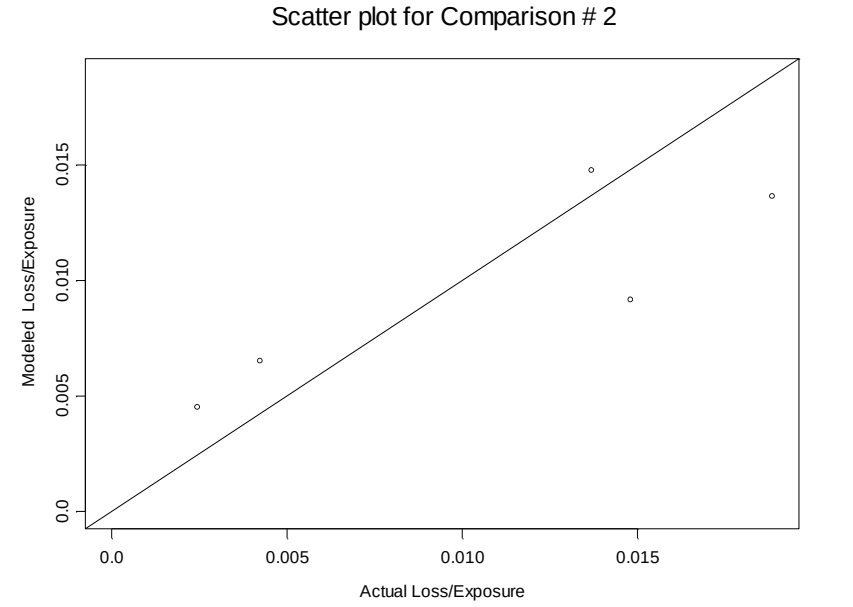


Figure 99. Scatter plot for comparison # 2.

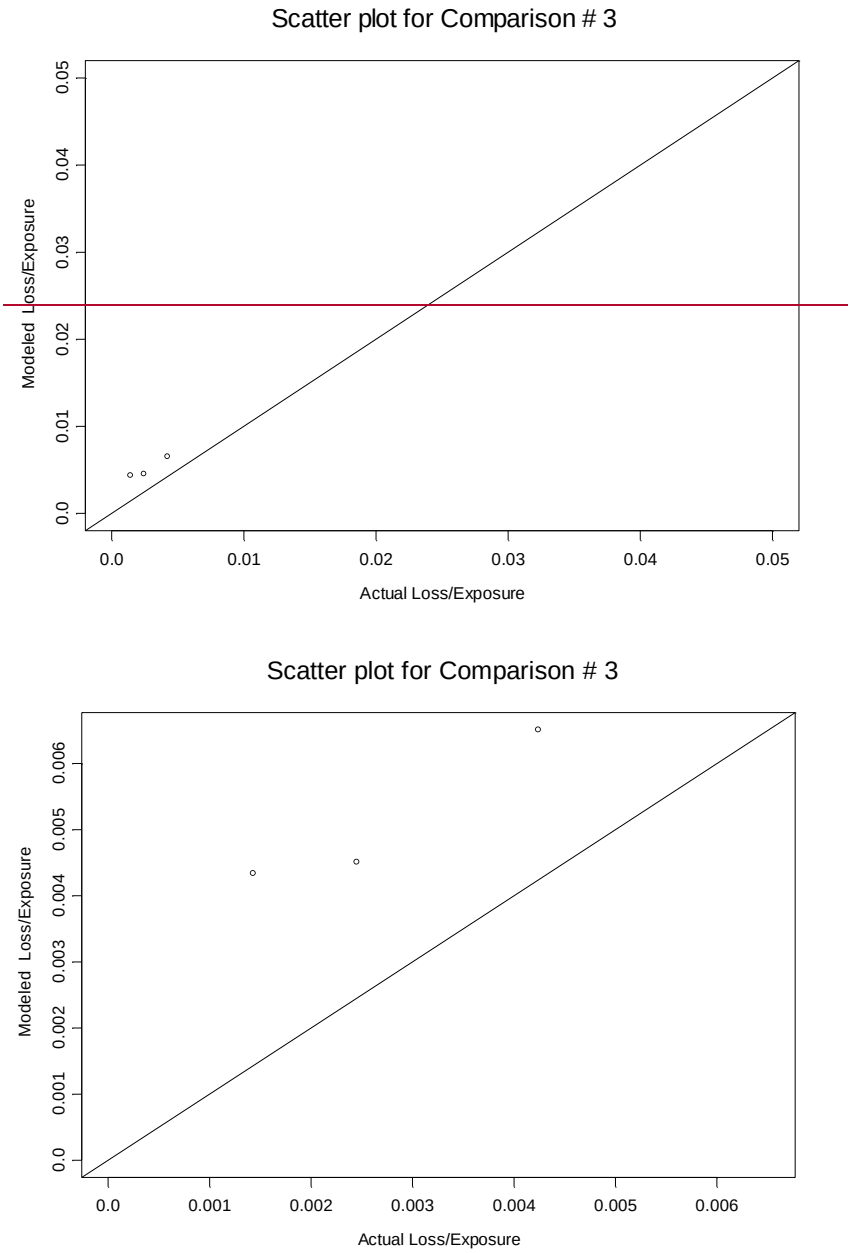
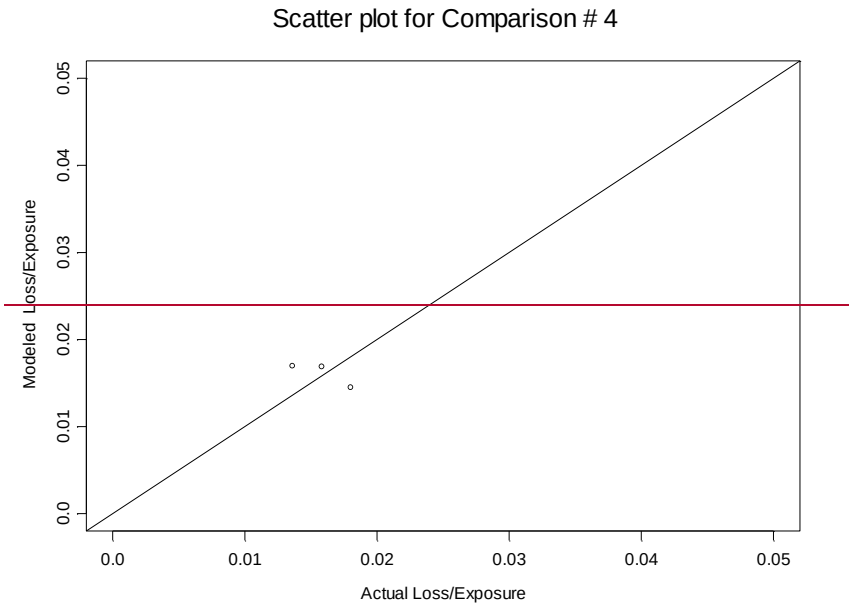


Figure 100. Scatter plot for comparison # 3.



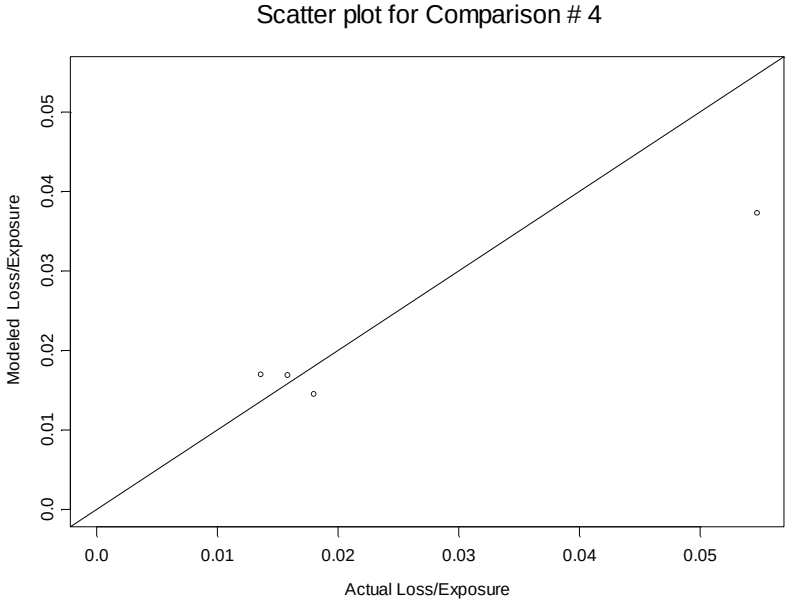
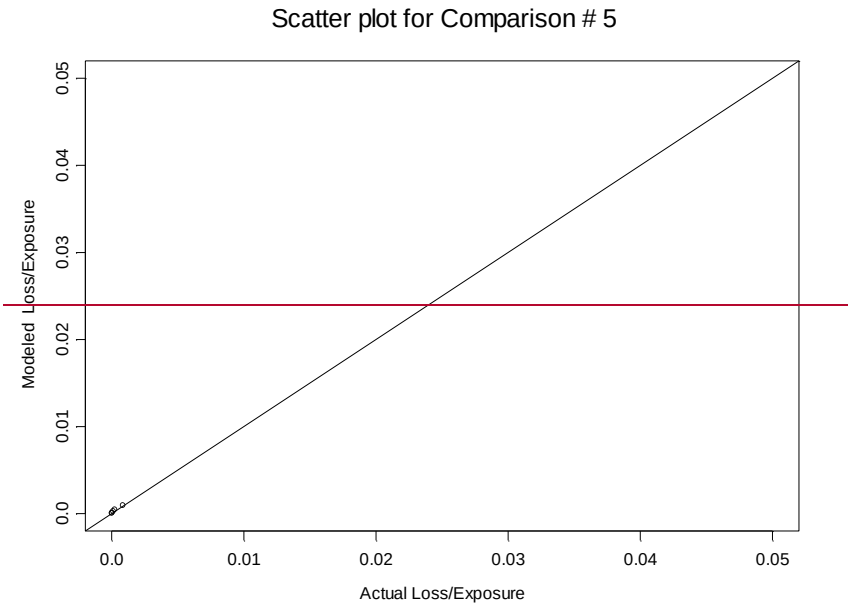


Figure 101. Scatter plot for comparison # 4.



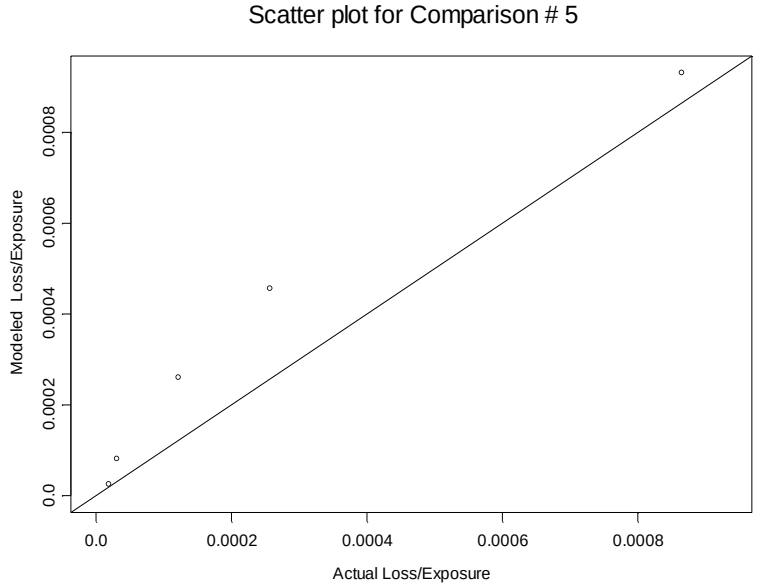


Figure 102. Scatter plot for comparison # 5.

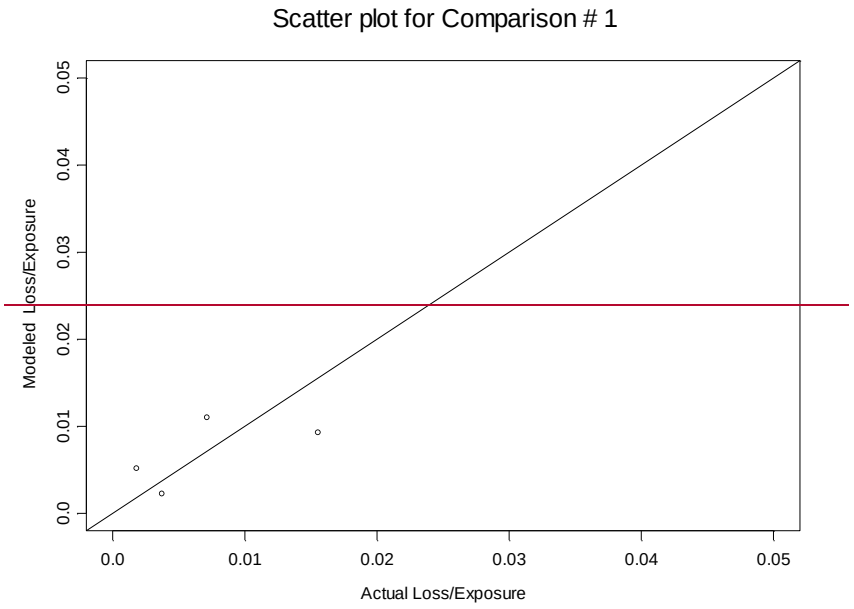
Commercial Residential:

~~Comparison # 1: Company A and Q by Hurricane Jeanne, Katrina, and Wilma~~

Company	Event	Company Actual Loss/Exposure	Modeled Loss/Exposure	Difference
A	Jeanne	0.00716	0.01099	0.00384
A	Katrina	0.00183	0.00514	0.00332
A	Wilma	0.01555	0.00927	-0.00628
Q	Wilma	0.00375	0.00223	-0.00151

Comparison # 1: Company D and Q by Hurricane Jeanne, Katrina, and Wilma

Company	Event	Company Actual Loss/Exposure	Modeled Loss/Exposure	Difference
D	Jeanne	0.007157	0.010808	0.003651
D	Katrina	0.001825	0.004821	0.002996
D	Wilma	0.015553	0.008776	-0.00678
Q	Wilma	0.025790	0.007518	-0.01827



Scatter plot comparison # 1

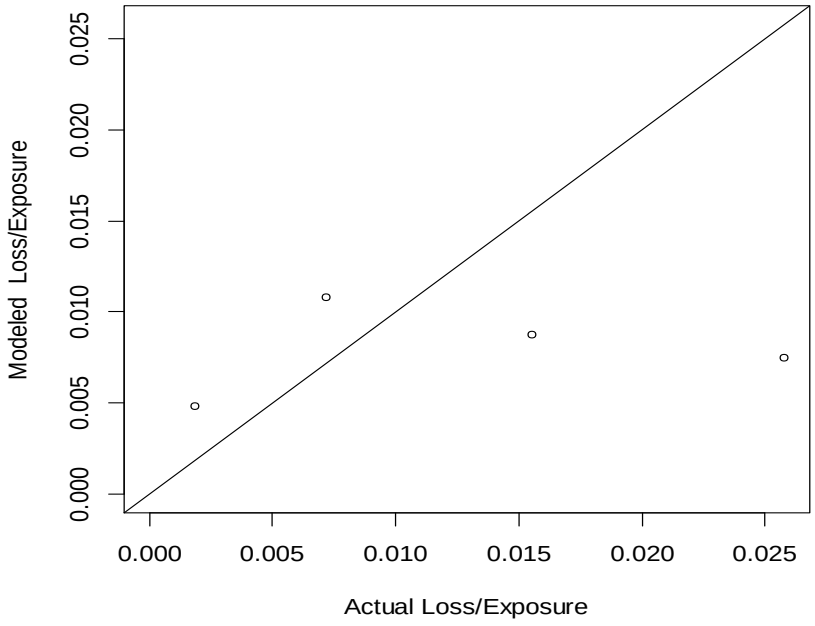


Figure 103. Scatter plot for comparison # 1.

Appendix S – Form S-5: Average Annual Zero Deductible Statewide Loss Costs – Historical versus Modeled

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
~~November 1, 2014~~March 15, 2015

Part A

Average Annual Zero Deductible Statewide Personal Residential Loss Costs
(in millions of dollars)

Time Period – 2007 FHCF Exposure Data	Historical Hurricanes	Produced by Model
Current Submission	$6,005.69	$5,215.98
Previously Accepted Submission	$5,277.44	$5,407.99
Percentage Change Current Submission/Previously Accepted Submission	13.80%	-3.55%
Second Previously Accepted Submission	$5,938.63	$5,896.72
Percentage Change Current Submission/Second Previously Accepted Submission	1.13%	-11.54%

Part B

Average Annual Zero Deductible Statewide Personal and Commercial Residential Loss Costs (in millions of dollars)

Time Period – 2012 FHCF Exposure Data	Historical Hurricanes	Produced by Model
Current Submission	$5,693.07	$4,928.66

Part A

Average Annual Zero Deductible Statewide Personal Residential Loss Costs
(in millions of dollars)

Time Period – 2007 FHCF Exposure Data	Historical Hurricanes	Produced by Model
Current Submission	$5,998.34	$5,210.65
Previously Accepted Submission	$5,277.44	$5,407.99
Percentage Change Current Submission/Previously Accepted Submission	13.66%	-3.65%
Second Previously Accepted Submission	$5,938.63	$5,896.72
Percentage Change Current Submission/Second Previously Accepted Submission	1.01%	-11.63%

Part B

Average Annual Zero Deductible Statewide Personal and Commercial Residential Loss Costs (million)

Time Period – 2012 FHCF Exposure Data	Historical Hurricanes	Produced by Model
Current Submission	$5,683.10	$4,921.29

Appendix T – Form V-1: One Hypothetical Event

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
November 1, 2014

Part A

All reference structures combined.

Wind Speed (mph) 1 min sustained Wind	Estimated Damage/ Subject Exposure
41-50	0.00%
51-60	0.05%
61-70	0.38%
71-80	1.11%
81-90	3.30%
91-100	7.32%
101-110	10.79%
111-120	15.80%
121-130	21.77%
131-140	23.57%
141-150	28.22%
151-160	29.62%
161-170	31.60%

Only personal residential reference structures combined (Timber + Masonry + MH).

Wind Speed (mph) 1 min sustained Wind	Estimated Damage/ Subject Exposure
41-50	0.00%
51-60	0.87%
61-70	2.57%
71-80	3.84%
81-90	6.16%
91-100	12.30%
101-110	17.34%
111-120	25.34%
121-130	40.90%
131-140	43.82%
141-150	54.41%
151-160	57.48%
161-170	65.64%

Only commercial residential reference structures (Concrete).

Wind Speed (mph) 1 min sustained Wind	Estimated Damage/ Subject Exposure
41-50	0.00%
51-60	0.04%
61-70	0.33%
71-80	1.06%
81-90	3.25%
91-100	7.22%
101-110	10.66%
111-120	15.61%
121-130	21.39%
131-140	23.16%
141-150	27.70%
151-160	29.06%
161-170	30.92%

Part B

Construction Type	Estimated Damage/ Subject Exposure
Wood Frame	14.75%
Masonry	12.88%
Mobile Home	36.57%
Concrete	10.74%

The structures used in completing the form are identical to those in the table provided.

Part C

All reference structures combined.

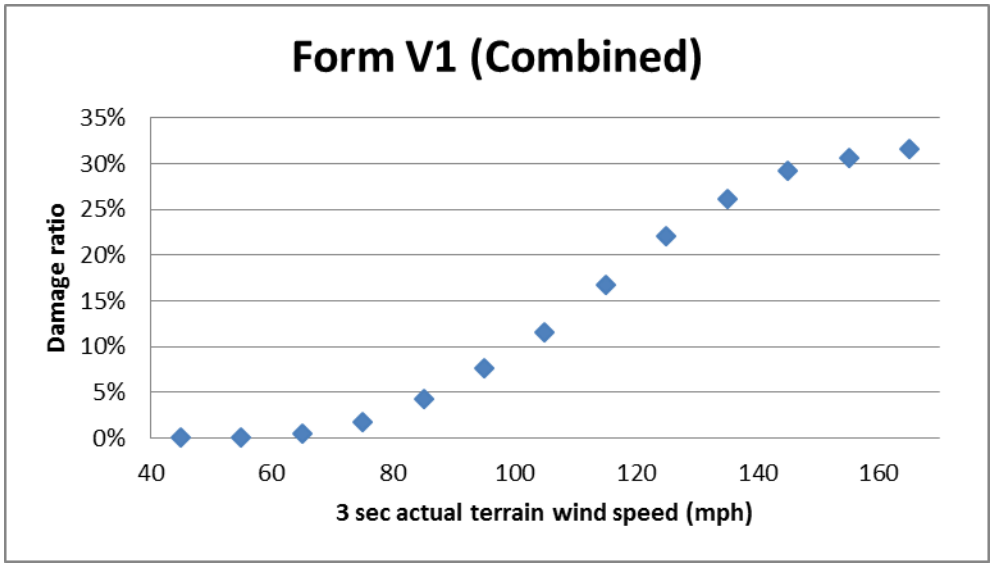


Figure 104. Structure damage vs. 3 sec actual terrain wind speed.

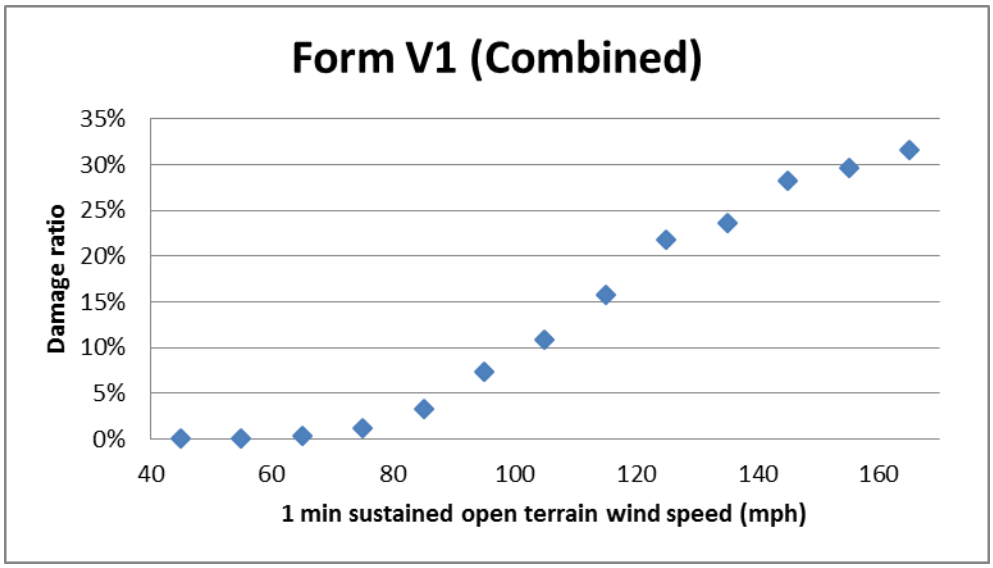


Figure 105. Structure damage vs. 1 minute sustained wind speed.

Only personal residential reference structures combined (Timber + Masonry + MH).

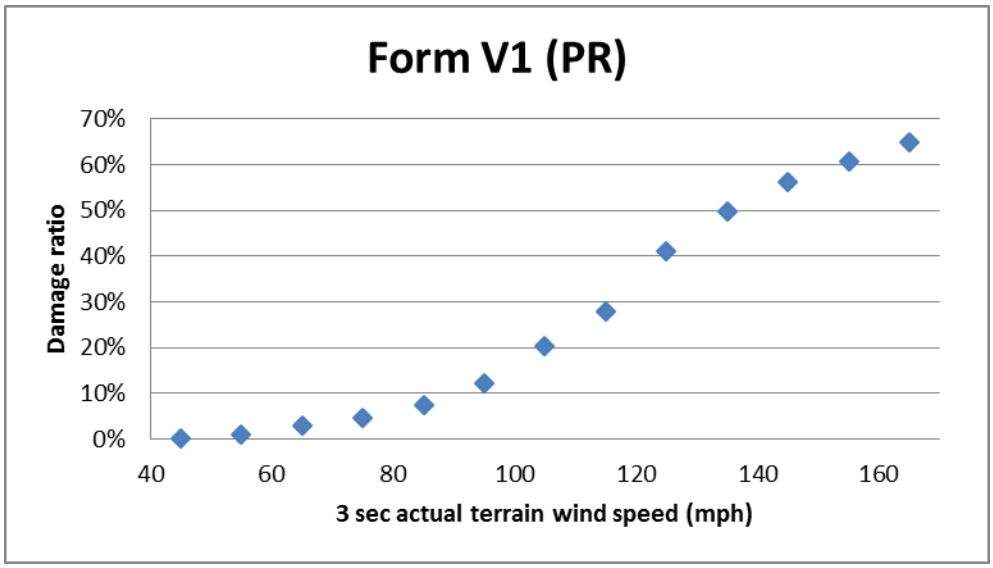


Figure 106. Structure damage vs. 3 sec actual terrain wind speed.

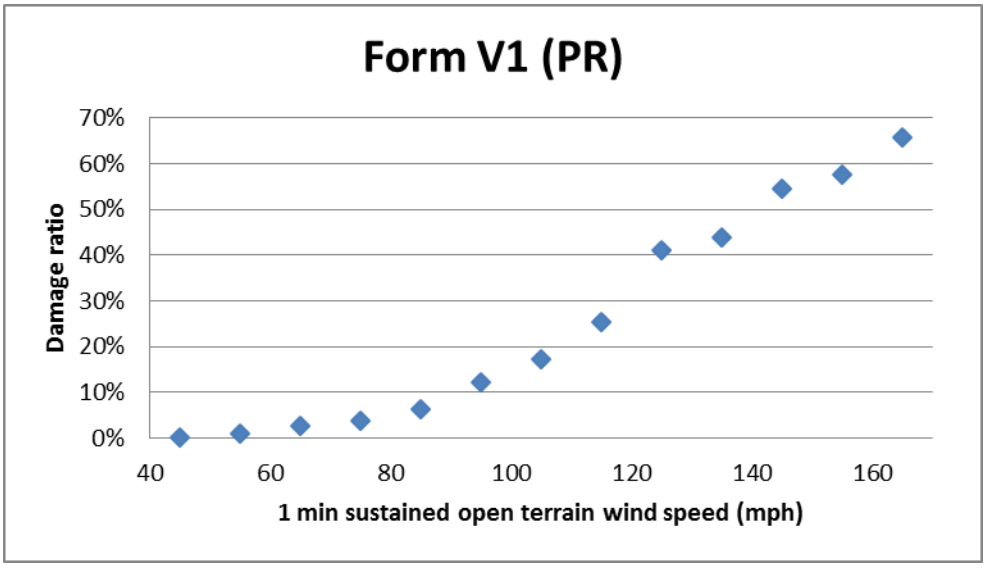


Figure 107. Structure damage vs. 1 minute sustained wind speed.

Only commercial residential reference structures (Concrete).

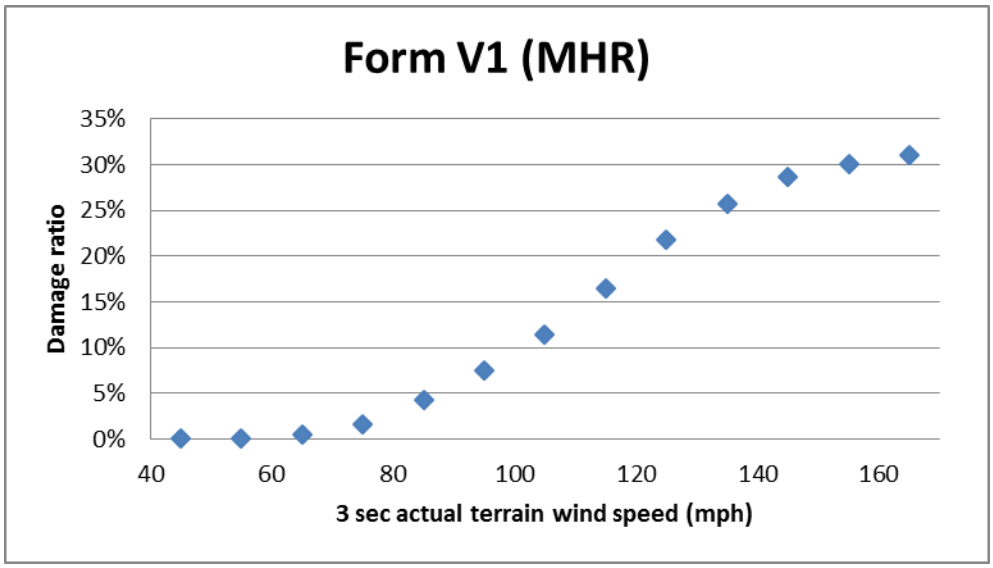


Figure 108. Structure damage vs. 3 sec actual terrain wind speed.

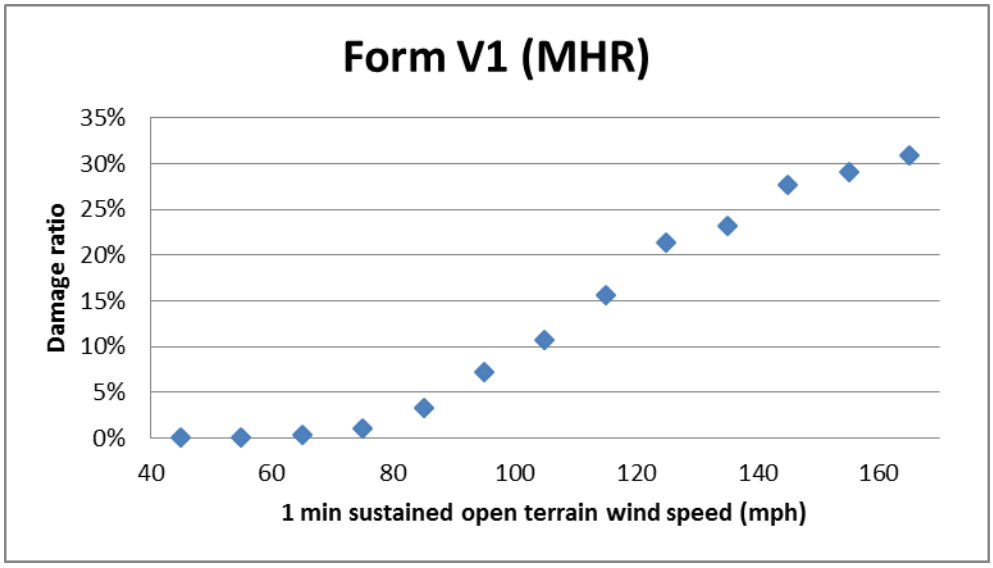


Figure 109. Structure damage vs. 1 minute sustained wind speed.

Appendix U – Form V-2: Mitigation Measures – Range of Changes in Damage

Florida International University
Florida Public Hurricane Loss Model ~~6.0~~6.1
November 1, 2014

Form V-2: Mitigation Measures – Range of Changes in Damage (1 min)

INDIVIDUAL MITIGATION MEASURES			PERCENTAGE CHANGES IN DAMAGE (REFERENCE DAMAGE RATE - MITIGATED DAMAGE RATE)/(REFERENCE DAMAGE RATE)*100									
			FRAME BUILDING					MASONRY BUILDING				
			WIND SPEED (MPH)					WIND SPEED (MPH)				
			60	85	110	135	160	60	85	110	135	160
	REFERENCE BUILDING		-	-	-	-	-	-	-	-	-	-
ROOF STRENGTH												
	BRACED GABLE ENDS		0%	0%	0%	0%	1%	0%	0%	0%	0%	1%
	HIP ROOF		1%	6%	5%	11%	4%	1%	6%	1%	7%	5%
ROOF COVERING												
	METAL		0%	1%	0%	0%	0%	0%	1%	0%	0%	0%
	RATED SHINGLES (110 MPH)		0%	1%	0%	0%	0%	0%	1%	0%	0%	0%
	MEMBRANE		0%	0%	0%	0%	0%	0%	0%	0%	0%	0%
	NAILING OF DECK 8d		2%	37%	0%	-7%	-1%	3%	38%	13%	-4%	-1%
ROOF-WALL STRENGTH												
	CLIPS		0%	0%	5%	14%	11%	0%	-1%	0%	7%	12%
	STRAPS		0%	0%	5%	18%	23%	0%	-1%	0%	8%	15%
WALL-FLOOR STRENGTH												
	TIES OR CLIPS		0%	0%	4%	6%	4%	-	-	-	-	-
	STRAPS		0%	0%	4%	6%	4%	-	-	-	-	-
WALL FOUNDATION STRENGTH												
	LARGER ANCHORS OR CLOSER SPACING		-	-	-	-	-	-	-	-	-	-
	STRAPS		-	-	-	-	-	-	-	-	-	-
	VERTICAL REINFORCING		-	-	-	-	-	0%	-1%	0%	10%	22%
OPENING PROTECTION												
	WINDOW SHUTTERS	PLYWOOD	0%	2%	6%	2%	0%	0%	2%	7%	3%	0%
		STEEL	0%	4%	10%	4%	1%	0%	3%	11%	6%	1%
		ENGINEERED	0%	4%	13%	8%	3%	0%	4%	15%	9%	3%
	DOOR AND SKYLIGHT COVERS		0%	0%	1%	1%	0%	0%	0%	1%	1%	1%
WINDOW DOOR, SKYLIGHT STRENGTH												
	WINDOWS	LAMINATED	0%	3%	11%	7%	2%	0%	3%	12%	8%	3%
		IMPACT GLASS	0%	4%	13%	10%	5%	0%	3%	14%	12%	6%
	ENTRY DOORS	HIGH STRENGTH	0%	0%	0%	1%	1%	0%	0%	0%	1%	1%
	GARAGE DOORS	HIGH STRENGTH	0%	16%	3%	1%	0%	0%	16%	5%	1%	0%
	SLIDING GLASS DOORS	HIGH STRENGTH	0%	0%	1%	1%	1%	0%	0%	1%	1%	1%

MITIGATION MEASURES IN COMBINATION		PERCENTAGE CHANGES IN DAMAGE (REFERENCE DAMAGE RATE - MITIGATED DAMAGE RATE)/(REFERENCE DAMAGE RATE)*100									
		FRAME BUILDING					MASONRY BUILDING				
		WIND SPEED (MPH)					WIND SPEED (MPH)				
		60	85	110	135	160	60	85	110	135	160
BUILDING											
	MITIGATED BUILDING	2%	40%	27%	26%	25%	3%	40%	24%	16%	16%