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Shih-Kai Huang

Jacksonville State University, shuang@jsu.edu

Michael K. Lindell

University of Washington - Seattle Campus

Carla S. Prater

University of Washington - Seattle Campus

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**Toward a Multi-stage Model of Hurricane Evacuation Decision:
An Empirical Study of Hurricanes Katrina and Rita**

Shih-Kai Huang^{1*}

Michael K. Lindell²

Carla S. Prater³

¹ Assistant Professor, Jacksonville State University, Aniston AL.

² Affiliate Professor, University of Washington, Seattle WA.

³ Affiliate Instructor, University of Washington, Seattle WA.

*Corresponding author, Tel +1 979 308 9985

Email: skysnow0080@gmail.com, mlindell@uw.edu, csprater@uw.edu

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Abstract

24

25 *This study extended previous research by testing the Protective Action Decision Model (PADM)*
26 *on hurricane evacuation decisions during Hurricanes Katrina and Rita. An examination of this*
27 *mediation model shows that a household's evacuation decision, as predicted, is determined most*
28 *directly by expected wind impacts and expected evacuation impediments. In turn, expected wind*
29 *impacts and expected hydrological impacts are primarily determined by expected storm threat and*
30 *expected rapid onset. Finally, expected storm threat, expected rapid onset, and expected*
31 *evacuation impediments are determined by households' personal characteristics, their reception*
32 *of hurricane information, and their observations of social and environmental cues. These results*
33 *are generally consistent with the PADM and reinforce the importance of testing multi-stage multi-*
34 *equation models of hurricane evacuation.*

35

36 **Keywords:** Evacuation, Protective Action Decision Model, Hurricane Katrina, Hurricane Rita,
37 Mediation effects

Introduction

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Evacuation is an effective response to hurricanes (Baker 1991; Dash and Gladwin 2007) and other emergencies (Sorensen 2000; Sorensen and Sorensen 2007), but most evacuation research has focused on direct effects of each predictor on evacuation decisions and overlooked the possibility that the effects of some variables are mediated by others (Lindell 2012). Huang et al. (2012) did conduct such a mediation analysis of the response to Hurricane Ike, but there is some indication that many risk area residents failed to take this hurricane seriously (Morss and Hayden 2010; Wei, Lindell and Prater 2014). The storm only had Category 2 wind speed so people seem to have underestimated the hydrological threats from coastal surge and inland flooding (De Young et al. in press). Consequently, the Huang et al (2012) model might not generalize to major hurricanes of Category 3-5.

To better understand household evacuation decisions, the present study tested a comprehensive model of the causal relationships among variables that have only been studied previously as predictors. A mediation analysis can examine whether one variable— independent variable, X — influences evacuation decision— dependent variable, Z — via another variable— mediator variable, Y (Huang et al. in press; Lindell 2008; MacKinnon, Fairchild and Fritz 2007). The present study begins by summarizing the most important predictors of evacuation decisions and then tests the Huang et al. (2012) formulation of the Protective Action Decision Model (PADM—Lindell and Perry 1992, 2004, 2012) using data from Hurricanes Katrina and Rita, which both reached Category 5 intensity and had late-changing tracks.

Predictors for Household Evacuation Decisions

61

62 Baker (1991) found that the best predictors of household evacuation are hurricane risk area
63 (e.g., barrier islands and other low-lying sites on the open coast or inland tidewater), official
64 notices, mobile home residence, personal risk perceptions, and storm severity. Demographic
65 characteristics and previous experiences had weak and inconsistent effects. More recently, Huang
66 et al. (in press) conducted a statistical meta-analysis (SMA) that confirmed and extended Baker's
67 findings. Hurricane risk area, official warnings, expected personal impacts, and observed
68 environmental/social cues were consistent across studies in producing effect sizes ranging $.15 \leq r$
69 $\leq .35$. Demographic characteristics and previous experiences had weaker effect sizes ($.01 \leq r \leq .09$
70 in absolute value), perhaps because their influence on evacuation decisions was mediated by those
71 variables that had stronger effects.

72 Thus, research to date indicates that receiving an official warning, seeing peers evacuating, and
73 expecting personal casualties affect evacuation decisions directly. The finding that reliance on
74 peers, gender, homeownership, risk area, environmental cues, seeing businesses closing, intensity,
75 nearby landfall, perceiving flood risk, perceiving surge risk, perceiving wind risk, and perceiving
76 service disruptions have weaker but significant effect sizes suggests that these variables' effects
77 on evacuation decisions are mediated by other variables. In addition, other variables—reliance on
78 authorities, reliance on news media, age, previous experiences, unnecessary evacuation
79 experiences, rapid onset, perceiving job disruptions, looting concerns, and evacuation expense—
80 had nonsignificant effect sizes but those results were inconsistent among studies. Thus, more tests
81 are required to examine whether their effects are mediated. Moreover, the nonsignificant effect
82 sizes of property protection concerns and traffic jam concerns were based on only a few studies,
83 so further examination of their effects is also required. Finally, the consistently nonsignificant

84 correlations of race, marital status, household size, education, and income need further
85 examination to see if these variables are significantly related to variables that are directly related
86 to evacuation decisions. Thus, this study tests the model depicted in Figure 1 that is defined by 16
87 hypotheses, and also addresses two research questions, using data from the Hurricane Katrina and
88 Rita evacuations.

89

90 H1: Age will be negatively related to perceived storm characteristics (expected hurricane
91 intensity, expected nearby landfall, and expected rapid onset).

92 H2: Female gender will be positively related to perceived storm characteristics.

93 H3: Homeownership will be negatively related to perceived storm characteristics.

94 H4: Information sources will be positively related to perceived storm characteristics.

95 H5: Official warnings (hearing a hurricane watch or warning and receiving an official
96 evacuation order) will be positively related to perceived storm characteristics.

97 H6: Previous hurricane experience will be positively related to perceived storm characteristics.

98 H7: Coastal proximity will be positively related to perceived storm characteristics.

99 H8: Observation of environmental and social cues (observations of environmental cues,
100 businesses closing, and peers evacuating) will be positively related to perceived storm
101 characteristics.

102 H9: “Unnecessary” evacuation experience will be positively related to perceived evacuation
103 impediments (concern about protecting property from looters, concern about protecting
104 property from the storm, concern about evacuation expenses, and concern about traffic
105 jams).

- 106 H10: Perceived storm characteristics will be positively related to expected personal impacts
107 (expected surge damage, flood damage, wind damage, household casualties, job disruption,
108 and service disruption).
- 109 H11: Expected personal impacts will be positively related to evacuation decisions.
- 110 H12: Perceived evacuation impediments will be negatively related to evacuation decisions.
- 111 H13: When other variables are controlled in the prediction of evacuation decisions, only
112 expected personal impacts and expected evacuation impediments will have statistically
113 significant regression coefficients.
- 114 H14: When other variables are controlled in the prediction of expected personal impacts, only
115 perceived storm characteristics will have statistically significant regression coefficients.
- 116 H15: When other variables are controlled in the prediction of perceived storm characteristics,
117 age, female gender, homeownership, reliance on information sources, official warning,
118 experience, risk area, and environmental and social cues will all have statistically
119 significant regression coefficients.
- 120 H16: When other variables are controlled in the model of expected evacuation impediments,
121 only “unnecessary” evacuation experience will have a statistically significant regression
122 coefficient.
- 123 RQ1: Did households have similar evacuation rates in Katrina and Rita?
- 124 RQ2: What is the distribution of evacuation departure times in Katrina and Rita?

125

126

Insert Figure 1 about here

127

128

Methods

129

130 **The Two Hurricanes**

131

132 Hurricane Katrina was one of the most powerful hurricanes in the history of the Atlantic basin
133 (NHC 2005a). The storm rapidly intensified to Category 5 on August 28, 2005, when it entered
134 the Gulf of Mexico and gradually curved from southwest to north over the next three days. The
135 National Hurricane Center (NHC) issued a hurricane watch at 10:00 am Central Daylight Time
136 (CDT) on Saturday, August 27th and a hurricane warning at 11:00 pm CDT on Saturday, August
137 27th when the storm was a Category 3 and intensifying. Katrina made landfall close to Buras,
138 Louisiana at 6:10 am on Monday, August 29th as a Category 3 hurricane with a storm surge of
139 2.4-6.7m (8-22 ft).

140 One month later, Hurricane Rita threatened the Texas coast. The NHC issued a hurricane watch
141 at 4:00 pm CDT on Wednesday, September 21st and a hurricane warning at 10:00 am on Thursday,
142 September 22nd (NHC 2005b) when the storm reached Category 5. Rita's track initially threatened
143 Corpus Christi but gradually curved east toward Galveston and made landfall between Sabine Pass,
144 Texas, and Johnson Bayou, Louisiana, at 02:38 am CDT on September 24th as a Category 3
145 hurricane with a storm surge of 1.2-2.1m (4-7 ft).

146

147 **Data Collection Procedures**

148

149 After receiving approval from the Texas A&M University Institutional Review Board, the
150 Hazard Reduction & Recovery Center conducted two mail surveys beginning four months after
151 Hurricane Katrina. The Katrina survey comprised two Louisiana parishes (Jefferson and St.

152 Charles) whereas the Rita survey included seven Texas counties. The Texas sample comprised two
153 coastal counties from the Sabine Study Area (SSA—Orange and Jefferson), three inland SSA
154 counties (Newton, Jasper, and Hardin), one coastal county from the Houston/Galveston Study Area
155 (GSA—Galveston), and one inland GSA county (Harris).

156 Respondents were selected with a disproportionate stratified sampling procedure designed to
157 yield 200 households in each parish or county, so the questionnaire was mailed to randomly
158 selected households—800 in Louisiana and 2,800 in Texas. The survey generally followed
159 Dillman’s (1999) procedure; selected households were sent a packet containing a cover letter, a
160 questionnaire, and a stamped, self-addressed reply envelope. A reminder post card was sent to
161 those who did not return a completed questionnaire within two weeks. Replacement packets were
162 sent at two week intervals thereafter until respondents either returned a questionnaire or received
163 as many as one reminder post card and three questionnaire packets.

164 Of the 800 households in the Katrina survey, 123 had either incorrect addresses or could not
165 be forwarded and 270 returned valid questionnaires for a response rate of 39.9% (37% in Jefferson
166 Parish and 43% in St. Charles Parish). Moreover, 392 of 2,800 households in the Rita survey had
167 an incorrect address or could not be forwarded but 1,007 returned a questionnaire for an overall
168 response rate of 41.8%, which was relatively similar across counties.

169

170 **Survey Instrument**

171

172 The evacuation zones differed between the two Texas areas (SSA and GSA) and their
173 evacuation zones differed from Louisiana’s so comparable codes for all three coastal areas were
174 created by coding respondents’ geographic locations into five risk areas. A household was coded

175 0 for barrier islands, 1 for locations exposed to Category 1 or 2 hurricanes (Risk Area 1 and 2 for
176 SSA, Zip-Zone A for GSA, and Phase I for Louisiana State), 2 for locations exposed to Category
177 3 hurricanes (Risk Area 3 for SSA, Zip-Zone B for GSA, and Phase II for Louisiana State), 3 for
178 locations exposed to Category 4 or 5 hurricanes (Risk Area 4 and 5 for SSA, Zip-Zone C for GSA,
179 and Phase III for Louisiana), and 4 for locations farther inland (Texas Department of Public Safety
180 2014a, b; Governor’s Office of Homeland Security and Emergency Preparedness, State of
181 Louisiana 2014).

182 As indicated in the Appendix, the first items in the questionnaire asked, on average, how many
183 times per day the respondent consulted four different sources—*local authorities*, *local news media*,
184 *national news media*, and *peers* on a five category scale of 0 times (= 1), 1-2 times (= 2), 3-4 times
185 (= 3), 5-6 times (= 4), and 7 or more times (= 5). The second set of items asked participants to rate
186 the extent to which they thought the storm would have three *observed storm characteristics*—
187 nearby landfall, major intensity, and rapid onset—and six *expected personal impacts*—surge
188 damage, inland flood damage, storm wind damage, personal casualties, job disruption, and basic
189 services disruption. Each item was measured on a scale from not at all likely (= 1) to almost certain
190 (= 5). The next set of items assessed the extent to which respondents considered 11 different issues
191 when deciding to evacuate. There were three items about *environmental and social cues*, two items
192 about *official warnings*, and four items about *perceived evacuation impediments*. There was one
193 additional item measuring participants’ *previous personal experience with hurricane storm*
194 *conditions* and another item measuring respondents’ *previous experience with “unnecessary”*
195 *evacuation*. Each item was rated on a scale from not at all (= 1) to very great extent (= 5). Finally,
196 respondents were asked about their *evacuation decision* using a dichotomous variable—evacuated
197 (= 1) or not (= 0). Evacuation date was recorded from Friday August 26 to Monday August 29

198 (*Friday = 1, Saturday = 2, Sunday = 3, Monday = 4*) for Hurricane Katrina and Wednesday
199 September 21 to Saturday September 24 (*Wednesday = 1, Thursday = 2, Friday = 3 or Saturday =*
200 *4*) for Hurricane Rita. The time of evacuation was measured by a 24-hour timeline beginning at
201 1am and ending at midnight.

202

203 **Preliminary Tests**

204

205 *Treatment of Missing Data*

206

207 Among 1,277 respondents, 719 (57.3%) completed all questions. The majority of variables had
208 missing data rates lower than 5%, but the aggregate level of missing data across all variables might
209 have significant impacts on the results. The results of Little's (1998, see Howell 2013) missing
210 completely at random (MCAR) test and the Potthoff, Tudor, Pieper, and Hasselblad's (2006)
211 missing at random plus (MAR+) test revealed that data were at least missing at random.
212 Consequently, missing data were replaced using the Expectation-Maximization (EM) algorithm
213 (Howell 2013).

214

215 *Homogeneity of Correlations Between Hurricanes*

216

217 As in Huang et al. (2012), a test of the homogeneity of the covariance matrices was conducted
218 to justify pooling two datasets from different locations. Box's M ($= 1019.13, F_{630, 203367} = 1.46$)
219 was statistically significant but the extremely large number of degrees of freedom gives this test
220 the statistical power to detect trivial levels of heterogeneity. Gnanadesikan's (1977) graphical

221 homogeneity test shows the obtained value of each correlation for respondents from the Katrina
222 dataset plotted against the corresponding value of that correlation for respondents from the Rita
223 dataset (Figure 2). The cross-plot of $595 - k(k-1)/2 = 35(34)/2$ —interitem correlations from the
224 two samples is approximately linear ($r = .85$) and has no obvious outliers. Consequently, a pooled
225 correlation matrix was used in subsequent analyses.

226

227 ***Insert Figure 2 about here***

228

229 *Interrater (Respondent) Agreement and Factor Analysis*

230

231 Table 1 indicates that none of the variables had a level of interrater agreement $r_{wg} > .50$ —the
232 mid-point between a uniform distribution ($r_{wg} = 0$) and concentration of all values at a single point
233 ($r_{wg} = 1$). This result indicates that respondents varied substantially in their perceptions of these
234 variables. Consequently, the intercorrelations among variables and evacuation decisions are not
235 artifactually depressed by variance restriction (Nunnally and Bernstein 1994).

236

237 ***Insert Table 1 about here***

238

239 Next, a principal factor analysis with equamax rotation yielded a seven-factor solution in which
240 items with factor loadings $\lambda > .40$ defined eight scales. Among perceived storm characteristics
241 variables, nearby landfall and major intensity formed expected storm threat (ExStmThreat).
242 Consequently, perceived storm characteristics was replaced by expected storm threat
243 (ExStmThreat) and expected rapid onset (ExRapOnset) in H1-H8, H14, and H15.

244 The items in expected personal impacts loaded on two different factors, so ExHydroImpacts
245 (surge risk and flood risk) and ExWindImpact (the remaining items) replaced expected personal
246 impacts in H10, H11, H13, and H14. The remaining items loaded on their expected factors. Among
247 these scales, six (ExStmThreat, ExHydroImp, ExWindImp, SocialCues, OffWarn, and
248 ExEvacImp) reached a conventionally acceptable level of reliability ($\alpha \geq .70$, although see
249 Schmitt, 1996, for a discussion of conventional levels of coefficient α) but NewsMedia and Exper
250 have lower levels of reliability ($\alpha = .65$ and $.57$, respectively). Contact with peers and
251 environmental cues have factor loadings $\lambda > .40$ on NewsMedia and SocialCues, respectively, but
252 these items have distinctly different content from other items in these scales. In addition, the
253 reliabilities of NewsMedia and SocialCues would increase if contact with peers and environmental
254 cues were deleted from their respective scales so previous hurricane experience (HurrExper),
255 unnecessary evacuation experience (UnnecEvac), contact with peers (Peers), and environmental
256 cues (EnvironCues) were retained as separate variables, as were contacts with local authorities
257 (LocAuth) and expected rapid onset (ExRapOnset).

258 The redefinition of some variables requires a corresponding revision of some hypotheses. H1-
259 8 substitute expected storm threat and expected rapid onset for perceived storm characteristics and
260 H10-12 substitute expected wind impacts and expected hydrological impacts for expected personal
261 impacts. Moreover, H14a has expected wind impacts as the dependent variable and H14b has
262 expected hydrological impacts as the dependent variable. Finally, H15a has expected storm threat
263 as the dependent variable and H15b has expected rapid onset as the dependent variable.

264

265 *Statistical Significance Levels*

266 The correlation and regression analyses described in the next two sections involve a
267 substantial number of statistical tests, so a Bonferroni correction was used to control the
268 experiment-wise error rate. Selecting $p < .001$ level for each of 50 individual tests yields an
269 experiment-wise error rate of $p < .05$. According to this criterion, only correlations of $r \geq .10$ are
270 statistically significant.

271

272 **Results**

273

274 **Respondent Characteristics**

275

276 The respondents' demographic characteristics were generally consistent with the
277 corresponding Census data (see Table 2), but the Katrina sample had a slightly higher proportion
278 of White (77%), middle-aged (arithmetic mean, $M = 50.2$ years), married (64%), homeowners
279 (81%) with slightly higher education ($M = 14.2$ years), lower annual income (\$39,332) and larger
280 household size ($M = 3.1$). Similarly, the Rita sample had a higher proportion White (77%), middle-
281 aged ($M = 54.4$ years), married (71%), homeowners (89%) with higher education ($M = 14.0$ years),
282 and lower annual income (\$37,445).

283

284 ***Insert Table 2 about here***

285

286 **Hypothesis Tests: Correlation Analysis**

287

288 The correlation analysis in Table 3 yielded partial support for H1 (age will be negatively related
289 to perceived storm characteristics) and H2 (female gender will be positively related to perceived
290 storm characteristics) because age was only significantly correlated with expected rapid onset ($r =$
291 $-.10$) and female gender was not significantly correlated with either variable. Contrary to H3
292 (homeownership will be negatively related to perceived storm characteristics), homeownership
293 was not significantly correlated with either expect rapid onset or expected storm threat.

294 H4 (information sources will be positively related to perceived storm characteristics), H5
295 (official warnings will be positively related to perceived storm characteristics), and H8
296 (observation of environmental and social cues will be positively related to perceived storm
297 characteristics) were mostly supported by significant correlations of expected rapid onset and
298 expected storm threat with contacts with local authorities ($r = .12$ and $.12$), peers ($r = .14$ and $.10$),
299 and news media ($r = .05$ and $.18$); official warnings ($r = .10$ and $.29$); social cues ($r = .22$ and $.21$);
300 and environmental cues ($r = .13$ and $.28$), respectively. However, H6 (previous hurricane
301 experience will be positively related to perceived storm characteristics) and H7 (coastal proximity
302 will be positively related to perceived storm characteristics) were only partially supported by
303 correlations of risk area ($r = -.21$) and previous hurricane experience ($r = .25$) with expected storm
304 threat.

305 H9 (“unnecessary” evacuation experience will be positively related to perceived evacuation
306 impediments) was supported by $r = .34$. Moreover, as predicted by H10 (perceived storm
307 characteristics will be positively related to expected personal impacts), expected rapid onset was
308 correlated with expected hydrological impacts and expected wind impacts ($r = .28$ and $.22$,
309 respectively) and expected storm threat was correlated with expected hydrological impacts and
310 expected wind impacts ($r = .38$ and $.53$, respectively). Meanwhile, H11 (expected personal impacts

311 will be positively related to evacuation decisions) was supported by correlations of expected
312 hydrological impacts and expected wind impacts with evacuation decisions ($r = .24$ and $.31$,
313 respectively). However, H12 (perceived evacuation impediments will be negatively related to
314 evacuation decisions) was contradicted by a nonsignificant correlation ($r = .02$).

315

316 ***Insert Table 3 about here***

317

318 **Hypothesis Tests: Regression Analyses**

319

320 H13 (predicting evacuation decision) was partially supported by Table 4 Model 1 (all displayed
321 coefficients were re-estimated after deleting nonsignificant predictors). Specifically, expected
322 wind impacts ($b = .72$) and expected evacuation impediments ($b = -.34$) had their hypothesized
323 effects but, surprisingly, expected hydrological impacts, official warnings ($b = .85$), risk area ($b =$
324 $-.77$), and expected rapid onset ($b = -.33$) also had significant coefficients.

325

326 ***Insert Table 4 about here***

327

328 H14a (predicting expected wind impacts) was partially supported by Model 2, which indicates
329 expected storm threat had a significant coefficient ($\beta = .42$) but expected rapid onset did not.
330 Surprisingly, age ($\beta = -.08$), official warning ($\beta = .18$), environmental cues ($\beta = .13$), and
331 expected evacuation impediments ($\beta = .13$) also had significant coefficients. H14b was partially
332 supported by Model 3, which indicates expected rapid onset ($\beta = .22$) and expected storm threat (

333 $\beta = .22$) had significant coefficients but there was also significant coefficients for risk area ($\beta = -$
334 $.36$) and official warning ($\beta = .13$).

335 H15a was partially supported by Model 4, which shows risk area ($\beta = -.17$), hurricane
336 experience ($\beta = .17$), “unnecessary” evacuation experience ($\beta = -.11$), news media ($\beta = .11$),
337 official warning ($\beta = .15$), and environmental cues ($\beta = .17$) had significant coefficients but age,
338 gender, homeownership, and social cues did not. H15b was weakly supported by Model 5, which
339 indicates social cues ($\beta = .20$) had the only predicted effect and White ethnicity ($\beta = -.11$) had an
340 unpredicted effect.

341 H16 was partially supported by Model 6, which indicates “unnecessary” evacuation experience
342 had a significant coefficient ($\beta = .26$), but education ($\beta = -.11$), income ($\beta = -.16$), social cues
343 ($\beta = .24$), and environmental cues ($\beta = .14$) also had significant coefficients.

344 The regression analysis results are summarized in Figure 3, which show the two predicted
345 paths affecting evacuation decisions. In the first path, the effects of the antecedent variables on
346 evacuation decisions are mediated by expected storm threat and expected rapid onset, expected
347 personal hydrological and wind impacts. In the second path, the effects of the antecedent variables
348 on evacuation are mediated by expected evacuation impediments—although expected evacuation
349 impediments affect evacuation decisions indirectly (via expected wind impacts) as well as directly.

350 However, some of the predicted regression coefficients were nonsignificant. Specifically, the
351 coefficients for female gender and peers failed to achieve statistical significance, so these variable
352 have been eliminated from Figure 3. Conversely, there are some significant unpredicted regression
353 coefficients—showing that some variables whose effects were predicted to be completely
354 mediated were only partially mediated.

355

356

Insert Figure 3 about here

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Insert Table 10 about here

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Regarding RQ1, Table 5 shows hurricane evacuation rates decreased with distance from the coast. The difference was small in Louisiana (from 87.4% to 85.1%), but significantly larger in SSA (100% to 77.8%) and GSA (93.3% to 35.7%). In addition, evacuation rates in SSA (where Rita made landfall) were higher than in GSA—especially Risk Area 3 and the inland area. Even though there was a sharper decline between the risk areas and the inland area, the inland evacuation rates of 77.8% in SSA and 35.7% in GSA indicate high rates of *evacuation shadow*—evacuation in areas that were not advised to do so (Zeigler, Brunn and Johnson 1981). There were no significant difference in evacuation rates between Risk Area 2 and 3 ($\chi^2_{(1)} = .58, p > .05$) for Katrina, but significant differences in evacuation rates between the risk areas and the inland area ($\chi^2_{(3)} = 48.81, p < .001$) in SSA for Rita. Within GSA for Rita, there were significant differences in evacuation rates between the risk areas and the inland area ($\chi^2_{(4)} = 99.95, p < .001$) but no significant differences among the evacuation rates within the risk areas. SSA had significantly higher evacuation rates than GSA in Risk Area 3 ($\chi^2_{(1)} = 6.73, p < .05$) and the inland area ($\chi^2_{(1)} = 66.33, p < .001$).

Figures 4a and 4b show respondents' evacuation departure time distributions for Katrina and Rita, respectively. Departure timing was significantly higher in daytime hours (71% left their homes between 6 am and 6 pm in Katrina and 73% in Rita) than during evening and nighttime

378 hours (6 pm and 6 am). This effect was particularly pronounced two days (August 27) and one day
379 (August 28) before Katrina made landfall. Similarly, the peak for Rita occurred during the daytime
380 hours two days before landfall (September 22). Moreover, 46.3% of respondents in Katrina and
381 45.8% in Rita left *before* the NHC warning, 31.9% in Katrina and 31.3% in Rita left in the first 12
382 hours after the NHC warning, 19.2% in Katrina and 12.5% in Rita left in the second 12 hours after
383 the NHC warning, and only 2.6% in Katrina and 10.0% in Rita began their evacuation during the
384 storm or after it.

385

386 ***Insert Figure 4a and 4b about here***

387

388

Discussion

389

390 The Katrina and Rita data completely support H4, H5, H8, H9, H10, and H11; partially support
391 H1, H2, H6, H7, H12, H13, H14a, H14b, H15a, H15b, and H16; and disconfirm H3. The partial
392 support for H1 (age negatively related to perceived storm characteristics) indicates older people
393 were less likely to expect rapid onset but people of all ages responded similarly to expected storm
394 threat. These findings are generally consistent with studies showing that older respondents tend to
395 experience more hurricane evacuations but have lower risk perceptions (Huang et al. 2012;
396 Trumbo, Lueck, Marlatt and Peek 2011). The first result is unsurprising because those who are
397 older have had more hurricane exposure, but it is not obvious why older people were less likely to
398 expect rapid onset.

399 The findings for H2 (female gender positively related to perceived storm characteristics) are
400 consistent with coastal residents' responses in Hurricane Ike (Huang et al., 2012) because,

401 although female gender failed to have a significant correlation with expected rapid onset, it did
402 correlate with other risk perception measures—expected wind impacts ($r = .11$) and expected
403 evacuation impediments ($r = .10$). These results are consistent with a wide range of previous
404 studies finding that women are generally more sensitive to environmental threats than are men
405 (Bateman and Edwards 2002; Davidson and Freudenberg 1996; DeYoung et al. in press). However,
406 the mediation effects of female gender through the risk perception variables on evacuation
407 decisions are extremely small. Specifically, the MacKinnon et al. (2007) product of coefficients
408 procedure shows the path through expected wind impact is trivial ($.11 \times .30 = .03$). Thus, the
409 significant correlation of female gender with evacuation decision—which is consistent with the
410 findings of the Huang et al. (2012) SMA—suggests that there are paths other than risk perception
411 that need to be identified.

412 Contrary to H3, homeownership had nonsignificant correlations with expected rapid onset and
413 expected storm threat. This finding is theoretically significant because many evacuation studies
414 have reported that homeowners are less concerned about warnings and possible storm impacts,
415 and less likely to evacuate (Huang et al. in press). Future studies need to determine if homeowners
416 have accurate perceptions of the ability of their homes to withstand wind and water impacts and,
417 thus, if more specific warnings about structural vulnerability are needed (cf. Morss and Hayden
418 2010; Wei et al. 2014).

419 The support for H4 (information sources positively related to perceived storm characteristics)
420 and H5 (official warnings positively related to perceived storm characteristics) confirms that
421 information about storm conditions was communicated through multiple information channels
422 (Lindell et al. 2007), especially the news media ($M = 3.73$) and this information affected people's
423 threat perceptions and evacuation decisions. Interestingly, as previously reported by Lindell et al.

424 (2005), there were positive correlations among the information sources, so people who accessed
425 one source tended to access other sources as well. This is consistent with previous findings that
426 people who receive an initial warning seek to confirm it through other sources (Drabek 1986; Perry
427 and Greene 1982) but the present study extends those results by showing that this resulted in an
428 increase in hurricane threat perceptions and expected evacuation impediments. However, these
429 results raise questions about the potential for conflict between information from the NHC and other
430 sources that could cause households to delay or forego evacuation (Gruntfest, Downing and White
431 1978). Consequently, further research should assess the perceived characteristics (e.g., expertise,
432 trustworthiness, and protection responsibility; Arlikatti et al. 2007) of different information
433 sources so emergency personnel can understand how credible they are compared to other sources
434 (Lamb et al. 2012).

435 The significant effects of previous hurricane experience (H6—previous hurricane experience
436 positively related to perceived storm characteristics), risk area (H7—coastal proximity positively
437 related to perceived storm characteristics), and environmental/social cues (H8—observation of
438 environmental/social cues positively related to perceived storm characteristics) are also
439 noteworthy because they confirm that people do not rely exclusively on official warnings and
440 hurricane information transmitted through the news media. Past experience might increase
441 people’s awareness that hurricanes can rapidly intensify and change course—concerns that might
442 be accentuated by observation of social/environmental cues. However, the effect of coastal
443 proximity is puzzling, so further research is needed to replicate and explain these findings.

444 The support for H9 (“unnecessary” evacuation experience positively related to perceived
445 evacuation impediments) suggests that memories of false alarms will discourage people from
446 evacuating (Baker 1991; Dow and Cutter 1998). Significantly, the data indicate that “unnecessary”

447 evacuation experience not only has the strongest effect in predicting expected evacuation
448 impediments ($\beta = .26$), but also has a negative effect on expected storm threat ($\beta = -.11$). In
449 addition, there are unpredicted effects of other variables on expected evacuation impediments. The
450 negative effects of income and education suggest that households with higher socioeconomic
451 status are less concerned about the adverse consequences of evacuation—probably because they
452 usually have multiple transportation options, can afford to stay in hotels, and are more likely to
453 have home insurance (Gladwin and Peacock 1997). The effects of environmental and social cues
454 are also easy to understand; observations of neighbors evacuating and businesses closing suggest
455 that evacuation routes will be crowded and it will be a long distance to vacant accommodations in
456 hotels and motels.

457 The significant coefficients for expected rapid onset and expected storm threat on expected
458 hydrological impacts and the significant coefficient for expected storm threat on expected wind
459 impacts support H10 (perceived storm characteristics positively related to expected personal
460 impacts) and the significant regression coefficient for expected wind impacts on evacuation
461 decisions supports H11 (expected personal impacts positively related to evacuation decisions).
462 These results suggest that NHC information about storm conditions does have its intended effect
463 on people's expected personal impacts and, in turn, their evacuation decisions. Indeed, these results
464 confirm the findings of the Huang et al. (in press) SMA that expected personal impact is one of
465 the strongest and most consistent predictors of households' evacuation decisions (see also
466 Fitzpatrick and Mileti 1993; Sorensen 2000; Sorensen and Sorensen 2007; DeYoung et al. in press).
467 Moreover, these findings support the PADM's proposition that, once people receive a warning,
468 they will try to determine if there is a real threat they need to pay attention to, followed by assessing
469 their risk of personal impacts, and deciding whether to adopt a protective action (Lindell and Perry

470 1992, 2004, 2012). The lower level of expected hydrological impacts— $M = 2.14$, compared to M
471 $= 3.43$ for expected wind impacts—and the smaller correlation of expected hydrological impacts
472 with evacuation decisions are consistent with the finding that respondents were more concerned
473 about wind threat than inland flood or storm surge threat (DeYoung et al. in press). However, this
474 might be a result of residents believing that expected casualties, job disruption and service
475 disruption would only be caused by wind impacts (see Table 1, Factor 3). Unfortunately, as
476 Hurricane Ike revealed, focusing on wind impacts to the exclusion of hydrological impacts can be
477 a major mistake (Morss and Hayden 2010; Wei et al. 2014).

478 The support for H12 (Perceived evacuation impediments negatively related to evacuation
479 decisions) is consistent with Huang et al. (2012), which found expected evacuation impediments
480 had a nonsignificant correlation coefficient but a significant regression coefficient. However,
481 unlike the Hurricane Ike data, the Katrina/Rita data showed that perceived evacuation impediments
482 had both direct and indirect effects on evacuation decision rather than just a direct effect. These
483 studies appear to conflict with previous research on hurricane evacuation concluding that false
484 alarms and other impediments have little or no effect on evacuation decisions (Baker 1991; Dow
485 and Cutter 1998). The conflict can be resolved by noting that the Katrina/Rita and Ike studies, like
486 the previous ones, reported nonsignificant bivariate correlations. However, the Katrina/Rita and
487 Ike studies found significant negative coefficients in regression analyses, which were not
488 conducted in the previous studies. Thus, further study is needed to examine the effect of
489 “unnecessary” evacuations and other impediments (e.g., Dow and Cutter 1998; see also Dillon and
490 Tinsley 2008; Dillon, Tinsley and Cronin 2011; Tinsley, Dillon and Cronin 2012).

491 There was only partial support for H13 (only expected personal impacts and expected
492 evacuation impediments will affect evacuation decision). Although expected wind impacts and

493 expected evacuation impediments did have the largest regression coefficients, expected
494 hydrological impacts was not significant—perhaps because storm surge is usually limited to the
495 immediate coast—and seems to be underestimated even there. Surprisingly, however, other
496 variables (official warnings, risk area, and expected rapid onset) also had significant coefficients.
497 The incomplete mediation of official warnings through wind impacts might be due to people
498 deciding to evacuate simply because officials told them to do so, even if they did not expect
499 significant personal impacts (e.g., Baker 1991; Gladwin, Gladwin and Peacock 2001). In addition,
500 the incomplete mediation of risk area through wind impacts might be due to people’s reaction to
501 their neighborhoods being labeled as a “risk area”, also independent of any expectation of
502 significant personal impacts. Finally, the incomplete mediation of expected storm threat through
503 wind impacts might be due to people’s reaction to the storm’s category and impact proximity,
504 again independent of any expectation of significant personal impacts. These speculations need to
505 be tested in future research.

506 In addition, expected rapid onset, which usually only has slight effect on evacuation decisions
507 (Riad, Norris and Ruback 1999; Smith and McCarty 2009), had a direct effect as well. One
508 plausible explanation is that both Hurricanes Katrina and Rita were Category 5 storms with late-
509 changing tracks. Thus, residents had limited time to consider whether they needed to evacuate and,
510 at the same time, had to determine whether they could reach safety before the hurricane arrived.
511 Consequently, some people might not have wanted to evacuate from a rapidly approaching storm
512 because of the risk of being caught on the road, especially if there were traffic jams. Unfortunately,
513 too few previous studies have examined the effect of expected rapid onset, so further research is
514 needed.

515 The data also partially supported H14 (only perceived storm characteristics will predict
516 expected personal impacts), which was revised to substitute expected storm threat (H14a) and
517 expected rapid onset (H14b) for perceived storm characteristics. Although expected storm threat
518 did have a direct effect on expected wind impacts and both expected storm threat and expected
519 rapid onset had direct effects on expected hydrological impacts, other variables (official warning,
520 risk area, age, environmental cues, and expected evacuation impediments) also had direct effects
521 on expected hydrological impact, or expected wind impact, or both.

522 The effects of official warning and risk area on expected hydrological and wind impacts can
523 be explained in much the same way as the effects of these variables on evacuation decisions; there
524 might be *labeling effects* (influence by a term that is readily accessible in memory—Cialdini 2001)
525 that operate directly on expected hydrological and wind impacts but are independent of expected
526 storm threat and rapid onset. Similarly, age might be associated with memories of past wind
527 impacts that are unrelated to current descriptions of a hurricane’s expected intensity, nearby
528 landfall, and speed of onset. This explanation does not seem highly plausible because, if it were
529 the case, one would expect an effect for past hurricane experience as well. The unpredicted effect
530 of expected evacuation impediments on expected wind impacts might be the outcome of a
531 correlation whose causality runs in the reverse direction; that is, increased expectations of wind
532 impacts cause increased expectations of evacuation impediments because of respondents’
533 assumption that a more severe storm will increase the number of evacuees and, in turn, the severity
534 of evacuation impediments. These speculations also need to be tested in future research.

535 The test of H15 (age, female gender, homeownership, reliance on information sources, official
536 warning, hurricane experience, risk area, and environmental and social cues will predict perceived
537 storm characteristics) was partially supported. Specifically, there was evidence of direct effects of

538 risk area, hurricane experience, “unnecessary” evacuation experience, news media, official
539 warnings, and environmental cues on expected storm threat (H15a). As noted earlier, the effect of
540 hurricane experience is relatively easy to understand (see the discussion of H6), as are the effects
541 of information sources (H4), official warning (H5) and environmental cues (H8). However, as also
542 noted earlier, the effects of risk area (H7) and “unnecessary” evacuation experience (H9) are not.
543 Moreover, there were nonsignificant effects of age, female gender, peers, local authorities, and
544 social cues. The nonsignificance of age is consistent with the findings of Huang et al. (2012) but
545 the nonsignificance of female gender and social cues is not; Huang et al. (2012) did not report the
546 effects of local authorities, peers, and news media on perceived storm characteristics so no
547 comparisons can be made for these variables. Even more puzzling are the results for the test of
548 H15b because Table 8 revealed effects only for White ethnicity and social cues. In addition, the
549 R^2 (= .06) for this equation was quite small. Thus, further research is needed to better understand
550 people’s perceptions of expected rapid onset.

551 As predicted by H16 (only “unnecessary” evacuation experiences will predict expected
552 evacuation impediments), this variable has a direct effect on expected evacuation impediments.
553 However, income, education, risk area, and social and environmental cues have unpredicted direct
554 effects. The effects of income, and education indicate that people’s personal characteristics are
555 related to expected evacuation impediments—a finding that is consistent with the significant
556 correlations between personal characteristics and evacuation concerns in Hurricane Ike (Siebeneck
557 et al. 2013). It is less obvious why risk area, and social and environmental cues are directly related
558 to expected evacuation impediments.

559 The answer to RQ1 (similarity of evacuation rates in Katrina and Rita) is that the evacuation
560 patterns were similar in both hurricanes. Not only did evacuation rates decline with distance from

561 the coast for both hurricanes, there were nonsignificant differences between GSA (which was
562 farther from Rita's eventual landfall) and SSA (which was closer to the point of landfall). The
563 latter finding appears to conflict with previous reports that evacuation rates decline with distance
564 along the coast from the point of landfall and distance inland from the coast (e.g., Baker 1991;
565 Zhang et al. 2004; Cutter et al. 2011). However, the similar evacuation rates in all risk areas and
566 the high level of shadow evacuation in the inland area might have arisen from two important events.
567 First, all risk areas in the SSA coastal counties received an evacuation order. Thus, it is
568 unsurprising that Louisiana and SSA had high evacuation rates in all risk areas and SSA had a high
569 rate in the inland area because both locations were close to the point of hurricane landfall. Second,
570 GSA's high evacuation rates might have occurred because it had earlier been the expected point
571 of landfall before Rita gradually shifted its track eastward toward SSA. Moreover, the Houston
572 mayor warned everyone to evacuate who had ever previously experienced flooding, which was
573 widespread during Tropical Storm Allison. Thus, the inland evacuation rate of 35.7% in Harris
574 County was much higher than would otherwise be expected.

575 The data related to RQ2 (distribution of evacuation departure times in Katrina and Rita) show
576 that the Katrina and Rita departure time distributions were somewhat different from Ike's.
577 Although respondents consistently started their trips during daylight hours (72.3% in Katrina/Rita
578 and 80% in Ike), about 24.7% delayed their departures until shortly before landfall in Ike (Huang
579 et al. 2012) whereas less than 10% delayed that long before Katrina and Rita. A reasonable
580 explanation is that Ike was a Category 2 hurricane when NHC issued a hurricane warning whereas
581 Katrina and Rita were Category 5 storms when NHC warnings were issued. Not only are
582 respondents are more likely to imagine a Category 5 hurricane as a deadly hazard and are less

583 likely to wait for further information (Czajkowski 2011), they are also more likely to expect that
584 evacuation routes will be clogged because of the number of vehicles in the larger evacuation zones.

585 This study has some important practical implications. As mentioned earlier, respondents'
586 misconception of wind damage as more dangerous than inland flood and coastal surge is a
587 potentially fatal mistake because the majority of deaths in Ike were due to storm surge (Morss and
588 Hayden 2010; Wei et al. 2014). Thus, emergency managers should increase their jurisdictions'
589 awareness of the severity of hydrological impacts.

590 It is also noteworthy that these data suggest respondents made their evacuation decision by the
591 multi-source, multi-stage risk assessment process summarized in the PADM. Local authorities
592 who understand this process are more likely to be successful in obtaining compliance with their
593 evacuation recommendations by communicating timely and accurate information about expected
594 personal impacts. The negative effect of evacuation impediments on evacuation decision is also
595 noteworthy. Local governments should anticipate concerns about those perceived evacuation
596 impediments and establish appropriate strategies for communicating the their evacuation plan
597 provisions for minimizing those impediments. Finally, the findings regarding household
598 evacuations being directly affected by official warnings and risk area are important because they
599 imply that local officials can decrease shadow evacuation if they provide information that allows
600 people to more accurately recognize their risk areas (Arlkatti et al. 2006; Zhang et al. 2004).

601 Of course, this study has some limitations. First, the response rate was only 41.4% (39.9% for
602 Katrina and 41.8% for Rita). Although this is generally consistent with other HRRC mail
603 surveys—ranging from 24.6% (Lindell et al. 2001) to 50.7% (Lindell et al. 2005)—the sample
604 may fail to represent some specific demographic categories. However, the sample's demographic
605 characteristics were generally consistent with the average of the values for 2000 and 2010 census

606 data. Moreover, overrepresentation of specific demographic categories will produce bias in other
607 variables only to the degree that demographic variables are correlated with those variables. Indeed,
608 Table 3 shows the correlations of demographic variables with other variables are small in this
609 sample, as well as more generally (Huang et al. in press; Lindell 2013; Lindell and Perry 2000).

610 Another limitation is the disconfirmation of the MAR assumption underlying the EM algorithm
611 for estimating missing data (recall the discussion in *Treatment of Missing Data*). This could
612 produce bias in variables with a substantial amount of missing data, the Katrina/Rita data have a
613 relatively low level of missing data so the bias is likely to be small (Howell 2013).

614 In addition, the regression models only accounted for a modest percentage of variance ($.06 \leq$
615 $R^2 \leq .38$), which might result from low reliabilities of single-item variables (e.g., expected rapid
616 onset) or unmeasured causes of the dependent variables. Some potentially relevant unmeasured
617 variables are the presence of livestock or pets (DeYoung et al. in press) and job demands (Baker
618 1991; Dow and Cutter 1998, 2000) but these tend to be mentioned by few respondents. Another
619 potentially relevant variable is disagreement among household members about the need for
620 evacuating, which might delay them to the point that it is no longer safe to leave.

621 The major limitation of post-storm surveys is that they are necessarily based on a
622 nonexperimental design because it is not possible to randomly assign respondents to hurricane
623 conditions or personal circumstances. This could bias regression coefficients (Lindell 2008). In
624 addition, cross-sectional studies such as this cannot verify the temporal ordering of the variables
625 in each correlation. Although it is reasonable to assume that antecedent variables such as age, risk
626 area, and experience preceded the risk perception variables, it is not possible to definitively
627 determine the temporal ordering among the latter.

628 Finally, the analyses are based on respondents' self-reports of their personal experiences and
629 perceptions that were collected months after the evacuation. Thus, there is no direct evidence
630 whether the recall of these perceptions might have changed during the months between the
631 evacuation and the time the respondents completed the survey. However, Lindell et al. (2016)
632 report evidence from other research that people's memories for events that occurred during
633 disasters are reasonably accurate.

634

635 **Conclusions**

636

637 Respondents generally considered wind impacts to be more dangerous than hydrological
638 impacts (storm surge and inland flooding)—which might be an unintentional outcome of the
639 Saffir-Simpson hurricane intensity scale being defined in terms of wind speed. The results of the
640 correlation and regression analyses were generally consistent with, but extended, the results of
641 most previous hurricane evacuation studies (Baker 1991; Huang et al. in press). In addition, the
642 results supported the Huang et al. (2012) abbreviated PADM but are somewhat more complex than
643 it proposed. In addition, the results also suggest that environmental cues, risk area, and hurricane
644 experience affect people's expectations of storm threat, wind impacts, and hydrological impacts
645 as strongly as information transmitted by the NHC through the news media or official warnings.
646 Nonetheless, some of the results conflict with the model presented by Huang et al. (2012), so
647 further research is needed to determine whether the conflicting results can be replicated and,
648 consequently, require revision of the model.

649 Overall, this study's findings confirm the need to study households' evacuation decisions as a
650 multi-stage process tested in earlier studies (Gladwin et al. 2001; Huang et al. 2012; Mileti & Beck,

651 1975; Perry, Lindell and Greene 1981). In turn, this underscores a need for future studies to report
652 not only the final regression models or even the correlation and regression coefficients of predictor
653 variables with evacuation decisions but the entire matrix of correlations among all variables.

654

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656

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Table 1

Interrater Agreement (r_{WG}), Principal Axis Factors, and Scale Reliabilities

Variable	r_{WG}	Factors							Scale α
		1	2	3	4	5	6	7	
1.LocAuth	.49*	.08	.15	.03	.02	.14	.31	.06	
2.Peers	.12	.13	.29	-.03	.11	.07	.47	.01	
3.LocNews	.18*	-.02	-.01	.13	.08	.01	.71	.10	
4.NatNews	.11	.07	.03	.06	.09	.02	.63	.06	
NewsMedia (Item 3-4)									.65
5.ExRapOnset	.45*	.17	.16	.11	-.09	.39	.07	-.01	
6.NearbyLand	.09	-.02	.13	.60	.10	.25	.08	.15	
7.Intensity	.23*	-.03	.08	.54	.14	.19	.11	.12	
ExStmThreat (Item 6-7)									.70
8.SurgeRisk	.18*	-.04	.01	.18	.22	.64	.01	.20	
9.FloodRisk	.23*	.01	.07	.17	.18	.65	.08	.11	
ExHydroImp (Item 8-9)									.70
10.WindRisk	.22*	.22	.05	.53	.39	.35	.10	.02	
11.Casualties	-.02	.14	.08	.41	.52	.29	.11	.10	
12.JobDisrupt	-.15	.12	.12	.41	.11	.22	.09	-.01	
13.ServDisrupt	.47*	.08	.03	.58	.09	.10	.02	.08	
ExWindImp (Item 10-13)									.73
14.EnvironCues	-.10	.17	.41	.25	.18	.11	.11	.17	
15.BusClos	.00	.23	.71	.06	.10	.15	.13	.14	
16.PeerEvac	.07	.18	.75	.02	.23	.14	.14	.10	
SocialCues (Item 15-16)									.81
17.HearWarn	.20*	.09	.55	.15	.45	.03	.13	.14	
18.OffOrder	.29**	.08	.39	.08	.57	.04	.17	.12	
OffWarn (Item 17-18)									.75
19.HurrExper	-.06	.01	.05	.16	.13	.07	.05	.64	
20.UnnecEvac	.03	.25	.07	-.09	.04	.06	.06	.57	
Exper (Item 19-20)									.57
21.ProtLoot	-.13	.62	.12	.06	-.06	.07	.06	.29	
22.ProtStm	-.04	.61	.20	.14	-.02	.05	.12	.24	
23.EvacExp	-.23	.73	.11	.01	.15	.08	.10	.02	
24.TrafRisk	-.17	.64	.18	.03	.04	.06	.06	.11	
ExEvacImped (Item 21-24)									.79
EvaDec (Corr)		.36	-.21	-.24	-.14	.15	-.16	.27	
(Rotated)		-.07	.09	.06	.58	.10	.06	.11	

Note: Bold entries have factor loadings $\lambda > .40$ and are included in the scales listed following the group of items loading on the corresponding factor. Entries with an asterisk (*) have a significant interrater agreement (r_{WG}) at $p < .001$.

LocAuth = contacts with local authorities; Peers = contacts with peers; LocNews = contacts with local news media; NatNews = contacts with national news media; NewsMedia = contacts with news media; ExRapOnset = expected rapid onset; NearbyLand = nearby landfall location; Intensity = expected hurricane intensity; ExStmThreat = expected storm threat; SurgeRisk = expected surge damages; FloodRisk = expected flood damages; ExHydroImp = expected hydrological impacts; WindRisk = expected wind damages; Casualties = perceived risk on being killed or injured; JobDisrupt = perceived risk on job disruptions; ServDisrupt = perceived risk on service disruptions; ExWindImp = expected wind impacts; EnvironCues = environmental cues; BusClos = seeing business closing; PeerEvac = seeing peers evacuating; SocialCues = social cues; HearWarn = hearing warning; OffOrder = receiving an official order; OfficialWarn = official warnings; HurrExper = hurricane experience; UnnecEvac = previous unnecessary evacuation; Exper = previous experience; ProtLoot = concern about protecting property from looters; ProtStm = concern about protecting property from the storm; EvacExp = concern about evacuation expenses; TrafRisk = concern about traffic jams; ExEvacImped = expected evacuation impediments; EvaDec = evacuation decision.

Table 2. *Descriptive Statistics for the Hurricane Katrina and Rita Samples*

Hurricane	Location	Source	Variables								
			Population / N	Age*	Female	White	Married*	HHSIZE	Education	Income	HomeOwner
Katrina	Overall	Census 2000	503.5K	43.41	0.52	0.70	0.52	2.59	13.12	49,666	0.66
		Census 2010	485.0K	45.33	0.51	0.65	0.49	2.59	13.29	65,663	0.65
		Survey	270	50.15	0.52	0.77	0.64	3.05	14.15	39,332	0.81
	Jefferson Parish, LA	Census 2000	455.5K	43.58	0.52	0.70	0.51	2.56	13.14	49,200	0.64
		Census 2010	432.4K	45.52	0.51	0.64	0.48	2.57	13.28	64,754	0.63
		Survey	129	49.44	0.56	0.71	0.59	3.15	14.09	38,319	0.71
	St. Charles Parish, LA	Census 2000	48.1K	41.84	0.51	0.72	0.57	2.90	12.92	54,086	0.81
		Census 2010	52.7K	43.8	0.51	0.72	0.53	2.80	13.36	73,120	0.83
		Survey	141	50.79	0.48	0.82	0.7	2.97	14.20	40,273	0.90
Rita	Overall	Census 2000	4,086.5K	40.85	0.50	0.61	0.54	2.76	13.18	55,226	0.57
		Census 2010	4,839.5K	42.00	0.50	0.65	0.51	2.86	13.31	74,014	0.59
		Survey	1008	54.43	0.51	0.77	0.71	0.74	13.96	37,445	0.89
	Orange County, TX	Census 2000	85.0K	44.13	0.51	0.88	0.60	2.65	12.67	46,875	0.77
		Census 2010	82.0K	45.93	0.50	0.88	0.56	2.62	12.86	59,878	0.77
		Survey	160	55.49	0.50	0.85	0.73	2.63	13.82	37,014	0.89
	Jefferson County, TX	Census 2000	252.1K	43.85	0.50	0.57	0.54	2.55	12.89	45,698	0.66
		Census 2010	252.2K	44.31	0.49	0.56	0.47	2.55	13.06	58,464	0.63
		Survey	146	55.17	0.53	0.71	0.71	2.55	14.32	37,336	0.83
	Newton County, TX	Census 2000	15.1K	44.85	0.49	0.76	0.60	2.59	11.99	35,401	0.85
		Census 2010	14.4K	46.44	0.48	0.76	0.58	2.68	12.42	47,659	0.83
		Survey	125	55.98	0.53	0.70	0.70	2.75	12.94	31,250	0.90
	Jasper County, TX	Census 2000	35.6K	45.72	0.51	0.78	0.61	2.58	12.40	38,994	0.81
		Census 2010	35.8K	47.46	0.51	0.81	0.59	2.51	12.78	50,724	0.79
		Survey	149	56.03	0.51	0.79	0.68	2.93	13.26	34,366	0.92
	Hardin County, TX	Census 2000	48.1K	43.92	0.51	0.91	0.63	2.68	12.70	46,642	0.83
		Census 2010	54.8K	45.91	0.51	0.91	0.62	2.62	13.04	65,668	0.79
		Survey	160	54.13	0.53	0.85	0.66	2.63	13.48	36,799	0.93
	Galveston County, TX	Census 2000	250.2K	43.42	0.51	0.73	0.56	2.60	13.32	54,730	0.66
		Census 2010	292.0K	44.54	0.51	0.80	0.55	2.65	13.69	78,803	0.69
		Survey	141	53.26	0.43	0.75	0.77	2.86	14.56	42,460	0.86
Harris County, TX	Census 2000	3,400.6K	40.25	0.50	0.59	0.54	2.79	13.22	56,557	0.55	
	Census 2010	4,108.4K	41.48	0.50	0.64	0.51	2.90	13.32	75,317	0.57	
	Survey	127	50.47	0.54	0.72	0.70	2.84	15.41	42,991	0.88	

Population = number of population; N = number of respondents; Age = respondent's age; Female = respondent's gender; White = respondent's ethnicity; Married = respondent's marriage status; HHSIZE = household size; Education = respondent's education; Income = respondent's household income; HomeOwner = home ownership;

* Age and marital status for census data were evaluated for people who were older than 15 yrs.

Table 3

Means (*M*), *SD*, and Intercorrelations (r_{ij}) among Variables

Variable	<i>M</i>	<i>SD</i>	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	
1.Age	53.5	15.1																							
2.Female	.51	.50	-.12																						
3.White	.77	.42	.05	-.10																					
4.Married	.69	.46	-.09	-.20	.11																				
5.HHSize	2.84	1.57	-.39	-.01	-.07	.32																			
6.Education	14.0	2.42	-.15	-.04	.11	.09	.02																		
7.Income ^a	38.1	12.7	-.29	-.15	.24	.43	.15	.44																	
8.HomeOwner	.87	.33	.16	-.07	.09	.25	.02	.02	.17																
9.RiskArea	2.91	1.10	.00	.05	.01	-.04	.00	-.07	-.11	.06															
10.HurrExper	2.94	1.45	.15	-.03	.07	.01	-.03	.08	.04	.05	-.21														
11.UnnecEvac	2.34	1.40	.01	.01	.04	-.02	-.01	-.02	-.02	-.01	-.14	.40													
12.LocAuth	1.65	1.01	-.03	.09	-.10	.01	-.01	-.01	-.04	-.05	-.05	.11	.07												
13.Peers	3.24	1.33	-.14	.18	-.13	-.08	.07	-.11	-.08	-.03	.05	.07	.12	.27											
14.NewsMedia	3.73	1.13	-.01	.08	.01	.08	.08	.00	.05	.04	-.01	.12	.10	.24	.37										
15.OfficialWarn	3.79	1.10	-.02	.19	-.11	-.03	.02	-.08	-.09	-.05	-.10	.23	.13	.16	.30	.22									
16.SocialCues	3.06	1.27	-.04	.16	-.14	-.08	.02	-.14	-.12	-.07	-.03	.16	.23	.21	.35	.17	.57								
17. EnvironCues	3.57	1.48	-.10	.11	-.03	.01	.02	-.10	.00	-.01	-.04	.22	.16	.14	.19	.17	.40	.49							
18.ExRapOnset	1.69	1.05	-.10	.02	-.14	-.02	.07	-.06	-.03	-.08	-.02	.04	.08	.12	.14	.05	.10	.22	.13						
19.ExStmThreat	3.34	1.14	-.03	.07	.01	-.01	.00	.03	.03	-.07	-.21	.25	.04	.12	.10	.18	.29	.21	.28	.21					
20.ExHydroImp	2.14	1.11	-.10	.07	-.07	.02	.02	.04	.06	-.05	-.43	.22	.12	.14	.08	.12	.26	.23	.23	.28	.38				
21.ExWindImp	3.43	.98	-.11	.11	-.02	.03	.06	-.05	.00	-.02	-.17	.23	.09	.13	.17	.22	.39	.31	.37	.22	.53	.47			
22.ExEvacImped	2.97	1.18	-.01	.10	-.11	-.06	.00	-.23	-.24	-.02	.03	.17	.34	.16	.21	.14	.28	.40	.30	.22	.13	.13	.28		
23.EvacDec	.83	.38	-.06	.11	-.02	-.01	.04	-.01	-.02	-.05	-.26	.14	.10	.06	.13	.11	.38	.21	.20	-.04	.20	.24	.31	.02	

Note: Italicized correlations are significant at $p < .001$.

Bold correlations test Hypotheses 1-12

Age = respondent's age; Female = respondent's gender; White = respondent's ethnicity; Married = respondent's marriage status; HHSize = household size; Education = respondent's education; Income = respondent's household income; HomeOwner = home ownership; RiskArea = respondent's risk area; HurrExper = hurricane experience; UnnecEvac = previous unnecessary evacuation; LocAuth = contacts with local authorities; Peers = contacts with peers; NewsMedia = contacts with news media; OfficialWarn = official warnings; SocialCues = social cues; EnvironCues = environmental cues; ExRapOnset = expected rapid onset; ExStmThreat = expected storm threat; ExHydroImp = expected hydrological impacts; ExWindImp = expected wind impacts; ExEvacImped = expected evacuation impediments; EvacDec = evacuation decision.

^a \$1,000 USD.

Table 4

Prediction of Evacuation Decision, Expected Wind Impacts, Expected Hydrological Impacts, Expected Storm Threat, Expected Rapid Onset, and Expected Evacuation Impediments

Dependent Variable	Model 1			Model 2			Model 3			Model 4			Model 5			Model 6		
	Evacuation Decision			Expected Wind Impacts			Expected Hydrological Impacts			Expected Storm Threat			Expected Rapid Onset			Expected Evacuation Impediments		
Predictors	B	Stand ard error (B)	Exp(B)	B	Stand ard error (B)	β	B	Stand ard error (B)	β	B	Stand ard error (B)	β	B	Stand ard error (B)	β	B	Stand ard error (B)	β
Age	-	-	-	-.01*	.00	-.08	-	-	-	-	-	-	-	-	-	-	-	-
White	-	-	-	-	-	-	-	-	-	-	-	-	-.29*	.07	-.11	-	-	-
Education	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-.05*	.01	-.11
Income	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-.00*	.00	-.16
RiskArea	-.77*	.10	.46	-	-	-	-.36*	.02	-.36	-.17*	.03	-.17	-	-	-	-	-	-
HurrExper	-	-	-	-	-	-	-	-	-	.14*	.02	.17	-	-	-	-	-	-
UnnecEvac	-	-	-	-	-	-	-	-	-	-.09*	.02	-.11	-	-	-	.22*	.02	.26
NewsMedia	-	-	-	-	-	-	-	-	-	.11*	.03	.11	-	-	-	-	-	-
OfficialWarn	.85*	.09	2.35	.16*	.02	.18	.13*	.02	.13	.16*	.03	.15	-	-	-	-	-	-
SocialCues	-	-	-	-	-	-	-	-	-	-	-	-	.16*	.02	.20	.22*	.03	.24
EnvironCues	-	-	-	.09*	.02	.13	-	-	-	.13*	.02	.17	-	-	-	.10*	.02	.14
ExRapOnset	-.33*	.08	.72	-	-	-	.23*	.03	.22	-	-	-	-	-	-	-	-	-
ExStmThreat	-	-	-	.36*	.02	.42	.22*	.02	.22	-	-	-	-	-	-	-	-	-
ExHydroImp	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-
ExWindImp	.72*	.10	2.06	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-
ExEvaclmped	-.34*	.09	.71	.11*	.02	.13	-	-	-	-	-	-	-	-	-	-	-	-
Constant	.42	.49	1.52	1.27*	.13		1.59*	.14		2.17*	.16		1.41*	.14		2.71*	.20	
χ^2 / F		327.38*			159.50*			131.16*			51.25*			40.10*			103.76*	
Degrees of freedom		5			(5,1271)			(4,1272)			(6,1270)			(2,1274)			(5,1271)	
% Correct		86.3																
Cox & Snell R ²		.23																
Nagelkerke R ² / Adj R ²		.38			.38			.33		.18			.06			.29		

Table 5

Evacuation Rates by Risk Area

Risk Area	Hurricane Katrina			Hurricane Rita					
	Louisiana			SSA			GSA		
	No	Yes	N	No	Yes	N	No	Yes	N
Barrier Island							6.7%	93.3%	30
1				0.0%	100.0%	26	5.3%	94.7%	95
2	12.6%	87.4%	135	2.2%	97.8%	137	7.4%	92.6%	27
3	14.9%	85.1%	134	5.6%	94.4%	125	23.5%	76.5%	17
Inland Area				22.2%	77.8%	410	64.3%	35.7%	98
Total	37	232	269	101	597	698	76	191	267
	13.8%	86.2%	100.0%	14.5%	85.5%	100.0%	28.5%	71.5%	100.0%