



1
2 **Rapid Assessment of Damaged Homes in the Florida Keys after Hurricane**
3 **Irma**

4 Siyuan Xian^{1,*}, Kairui Feng¹, Ning Lin¹, Reza Marsooli¹, Dan Chavas², Jie Chen², Adam
5 Hatzikyriakou¹

6
7 ¹Department of Civil and Environmental Engineering, Princeton University

8 ²Department of Earth, Atmospheric, and Planetary Sciences, Purdue University

9
10
11 **Abstract**

12 On September 10, 2017, Hurricane Irma made landfall in the Florida Keys and caused significant
13 damage. Informed by hydrodynamic storm surge and wave modeling and post-storm satellite
14 imagery, a rapid damage survey was soon conducted for 1600+ residential buildings in Big Pine
15 Key and Marathon. Damage categorizations and statistical analysis reveal distinct factors
16 governing damage at these two locations. The distance from the coast is significant for the damage
17 in Big Pine Key, as severely damaged buildings were located near narrow waterways connected
18 to the ocean. Building type and size are critical in Marathon, highlighted by the near-complete
19 destruction of trailer communities there. These observations raise issues of affordability and equity
20 that need consideration in damage recovery and rebuilding for resilience.

21
22 **Introduction**

23 Hurricane Irma made landfall near Cudjoe Key (lower Florida Keys) on September 10, 2017, as a
24 Category 3 storm. Irma caused widespread damage to the Florida Keys due to storm surge and
25 waves. Informed by hydrodynamic modeling and post-storm satellite imagery, we carried out a
26 field survey soon after (September 21-24) the event to investigate the damage to the Keys,
27 particularly the Big Pine Key and Marathon areas.

28 Post-hurricane damage studies have improved our understanding of coastal vulnerability (e.g. Xian
29 et al., 2015 and Hatzikyriakou et al., 2015 for Hurricane Sandy; Eamon et al., 2007 and van de



30 Lindt et al., 2007 for Hurricane Katrina). Here, we conduct a rapid damage survey and assessment
31 for Hurricane Irma, and we use a statistical regression approach to quantify the contribution of
32 specific vulnerability factors to the damage. Such rapid post-event assessments can provide crucial
33 information for implementing post-storm response measures (Lin et al., 2014; Horner et al., 2011;
34 AL-Kanj et al., 2016). The raw and analyzed data from this study appear on DesignSafe¹, a web-
35 based research platform of the National Science Foundation's (NSF) Natural Hazards Engineering
36 Research Infrastructure (NHERI).

37

38 **Storm Surge and Wave Simulation**

39 To understand the hazard and inform the field survey, we first use the coupled hydrodynamic and
40 wave model ADCIRC+SWAN (Dietrich et al. 2012, Marsooli and Lin 2017) to simulate the storm
41 tide (i.e., water level) and wave height for Hurricane Irma. To simulate Irma's storm tide and wave
42 (Figure 1), we apply the surface wind (at 10-m) and sea-level pressure fields from National Center
43 for Environmental Prediction Final (NCEP FNL) operational global analysis data (0.25° x 0.25° x
44 6 hours). The model results, e.g., time series in Figure 1, indicate that the model satisfactorily
45 captures the temporal evolution and the peak values of the water levels and wave heights induced
46 by Hurricane Irma. The model results show that the highest water levels, between 2 and 2.5 m,
47 occurred in South/Southwest Florida. However, coastal zones in this region are predominantly
48 uninhabited and covered by wetlands, so little loss of life or property is expected. High water levels
49 are also estimated for the Florida Keys, especially islands located on the right side of the storm
50 track. For example, the peak storm tide in Big Pine Key and Marathon reaches up to 2 m. The
51 model results also show that large waves with a significant wave height of about 14 m reached a

¹ <https://www.designsafe-ci.org/#research>



52 few kilometers off the Florida Keys. In contrast, wave heights off the southern and southwestern
53 coasts of Florida were small (< 2 m).

54 **Damage Survey and Analysis**

55 NOAA's post-storm satellite imagery² provides an overview of Irma's impact. The two selected
56 survey areas in Florida Keys, the Big Pine Key and Marathon, suffered the most severe damage,
57 according to the satellite imagery, and experienced high water levels and wave heights, indicated
58 by hydrodynamic modeling.

59

60 Field surveys can provide detailed information for analyzing damage mechanisms. However,
61 traditional on-site surveys require a significant time and effort, as surveyors must walk through
62 affected areas and photograph damaged properties. Thus, we applied a rapid survey method.
63 Rather than walking, we drove at a speed of 10 mph throughout the affected areas, taking GPS-
64 informed pictures from the rare side windows. Over two days, the team took 3700+ pictures for
65 1600+ residential buildings comprised of single family and mobile homes (e.g., trailers).

66 Using the collected photos and the satellite images, we categorized the damage state for each
67 surveyed residential house. Satellite images were primarily used to assess roof damage. More
68 detailed damage mechanisms were further evaluated from the photos. We adopted FEMA's
69 damage state criteria used in the damage assessment study for Hurricane Sandy³. The categories
70 include: No/very limited damage; Minor damage; Major damage; and Destroyed.

² <https://storms.ngs.noaa.gov/storms/irma/index.html#6/28.139/-81.547>

³ <https://www.arcgis.com/home/item.html?id=307dd522499d4a44a33d7296a5da5ea0>



71 We found that the destroyed and severely damaged buildings were caused largely by
72 hydrodynamic forces induced by storm surge/waves. For example, Fig. 2a shows that storm
73 surge/waves completely crashed the lower part of a building in Big Pine Key. Fig. 2b shows debris
74 from damaged trailers floating in the water in a trailer community in Marathon. The observed
75 storm surge damage is consistent with the high surge and wave heights estimated for the two sites
76 (Figs. 3a and 3b). The assessed damage state for each house appears in Figs. 3c and 3d. The slightly
77 and moderately damaged buildings are 72.7% and 75% of the total surveyed building for the
78 assessed areas in Big Pine Key and Marathon, respectively. The percentages of the destroyed
79 buildings are 13.9% and 16.9%, respectively. In both areas, the destroyed buildings are clustered.
80 The destroyed buildings in Big Pine Key are near the coastline and narrow waterways, a strong
81 indication that the damage was caused mainly by hydrodynamic forces. The completely destroyed
82 buildings in Marathon cluster in the north and middle parts of the study area. The majority of those
83 buildings are mobile homes.

84 Statistical analysis confirms these general observations. We use an ordered logistic regression
85 model to correlate the damage state with the following factors: distance from the coastline (m),
86 building type, and building size (m²). Our analysis for Big Pine Key shows that the distance from
87 the coastline is the single significant predictor of damage state (p-value < 0.001; Table 1a), as the
88 damage is dominated by buildings located near narrow waterways connected to the ocean. For
89 Marathon, although many damaged houses are near the coast, house type and house size are the
90 two significant predictors (p-value < 0.001; Table 1b), highlighting the near-complete destruction
91 of trailers (which are often small).

92 Possible measures to reduce flood vulnerability in the study areas include elevating and
93 strengthening the buildings (especially mobile homes) and relocating homeowners living near the



94 coastline (and narrow waterways) further inland. However, potential financial challenges exist,
95 especially for Marathon, where the median annual income is \$50,976 vs. \$63,716 for Big Pine
96 Key. Some local homeowners in a destroyed trailer community in Marathon (indicated by the red
97 rectangle in Fig. 3d) with whom we talked had lived in trailers as their primary homes for decades
98 without flood insurance. Financial constraints may hinder their rebuilding or relocating to
99 somewhere safer. As low-income people living in mobile homes suffered most, natural hazards
100 worsen economic inequality in this case. In contrast, discussion with local residents in Big Pine
101 Key indicated that many structures there were second homes and, furthermore, were designed to
102 withstand hurricane hazards (e.g., key assets were raised above the ground floor). These
103 observations raise again issues of affordability and equity (Montgomery and Chakraborty, 2015).
104 Policies relevant to hurricane damage recovery and rebuilding must address these issues.

105 Acknowledgments

106 This study is supported by NSF grant CMMI-1652448.

107

108

109

110 References:

111 Al-Kanj, L., Powell, W. B., & Bouzaiene-Ayari, B. (2016). The Information-Collecting Vehicle Routing
112 Problem: Stochastic Optimization for Emergency Storm Response. *arXiv preprint arXiv:1605.05711*.

113

114 Dietrich, J C, S Tanaka, J J Westerink, C N Dawson, R A Luettich Jr., M Zijlema, L H Holthuijsen, J M
115 Smith, L G Westerink, and H J Westerink. 2012. "Performance of the Unstructured-Mesh,
116 SWAN+ADCIRC Model in Computing Hurricane Waves and Surge." *Journal of Scientific Computing* 52
117 (2). Springer US: 468–97. doi:10.1007/s10915-011-9555-6.

118

119 Hatzikyriakou, A., Lin, N., Gong, J., Xian, S., Hu, X., & Kennedy, A. (2015). Component-based
120 vulnerability analysis for residential structures subjected to storm surge impact from Hurricane Sandy.
121 *Natural Hazards Review*, 17(1), 05015005.

122

123 Horner, M. W., & Widener, M. J. (2011). The effects of transportation network failure on people's
124 accessibility to hurricane disaster relief goods: A modeling approach and application to a Florida case
125 study. *Natural hazards*, 59(3), 1619-1634.

126

127 Lin, C. C., Siebeneck, L. K., Lindell, M. K., Prater, C. S., Wu, H. C., & Huang, S. K. (2014). Evacuees'
128 information sources and reentry decision making in the aftermath of Hurricane Ike. *Natural*
129 *Hazards*, 70(1), 865-882.

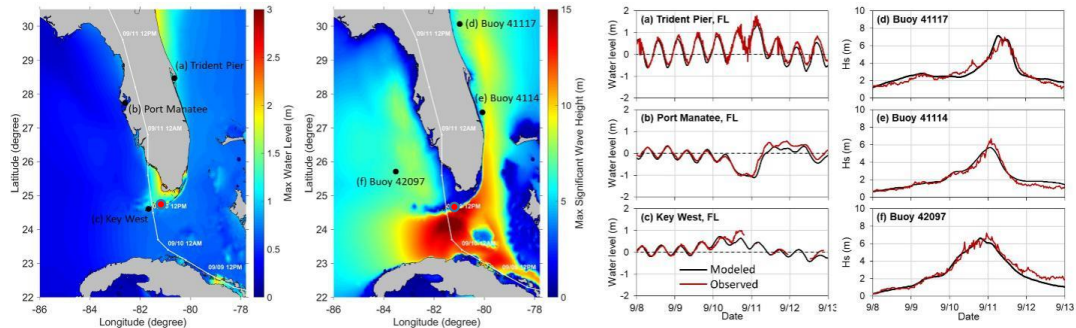
130



- 131 Marsooli, R., and N. Lin (2017). Numerical Modeling of Storm Tides and Waves Induced by Historical
132 Tropical Cyclones along the East and Gulf Coasts of the United States. *Journal of Geophysical Research:*
133 *Oceans*, Submitted.
134
135 Montgomery, M. C., & Chakraborty, J. (2015). Assessing the environmental justice consequences of
136 flood risk: a case study in Miami, Florida. *Environmental Research Letters*, 10(9), 095010.
137
138 Xian, S., Lin, N., & Hatzikyriakou, A. (2015). Storm surge damage to residential areas: a quantitative
139 analysis for Hurricane Sandy in comparison with FEMA flood map. *Natural Hazards*, 79(3), 1867-1888.
140
141 Xian, S., Lin, N., & Kunreuther, H. (2017). Optimal house elevation for reducing flood-related
142 losses. *Journal of Hydrology*, 548, 63-74.
143

144 Figures & Tables:

145
146



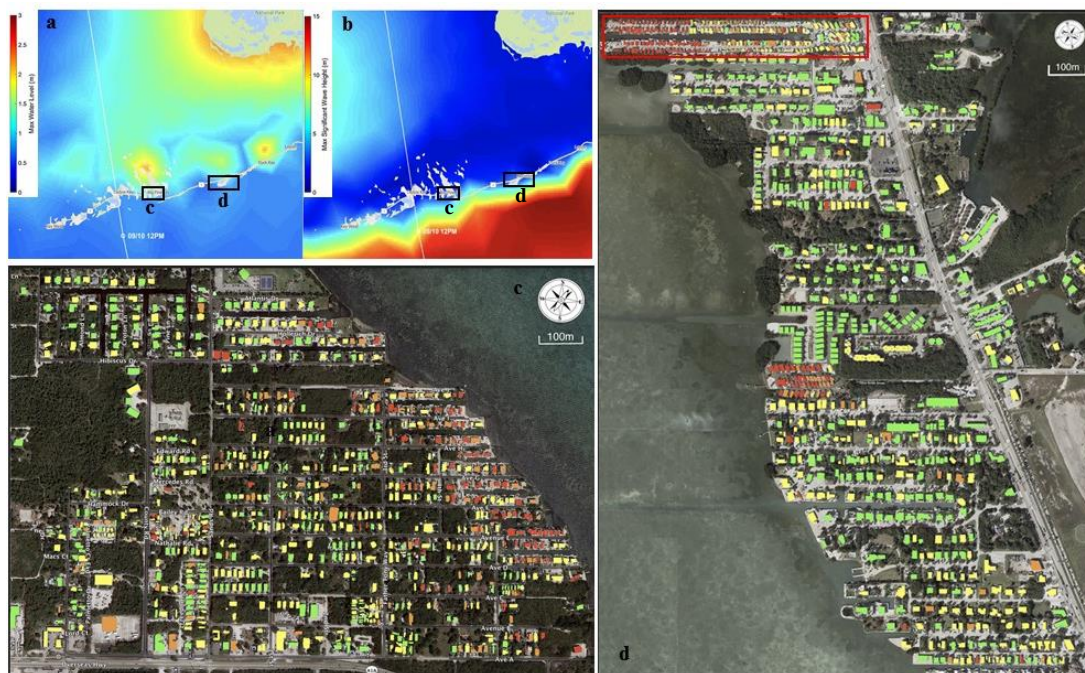
147
148 Figure 1. Hydrodynamic modeling of water level and wave height for Hurricane Irma. Left two
149 panels show spatial distribution of modeled maximum water level and significant wave height,
150 respectively. White curve represents storm track. Black points show locations of available tidal
151 gauge and buoy stations. Red point indicates approximate location of study area. Right two
152 panels compare observed and modeled time series of water level and significant wave height
153 (H_s), respectively.
154
155
156
157

158
159





160 Figure 2. Photos of damage in (a) Big Pine Key: storm surge damage besides waterway (left side
 161 of building) and (b) Marathon: trailer community with house debris filling waterway
 162



163 Figure 3. Spatial distribution of estimated hazards and damage states in study areas. (a) and (b)
 164 show simulated maximum total water level and significant wave height, respectively; (c) and (d)
 165 show assessed damage state (none: green; minor: yellow; major: orange; destroyed: red) for
 166 residential buildings in Big Pine Key and Marathon, respectively.
 167
 168
 169

170 Table 1 Ordered logistic regression models that correlate damage state with vulnerability factors
 171 (a) for 846 assessed buildings in Big Pine Key; (b) for 811 buildings in Marathon.

(a) Factors in damage state	Coef.	Std. Err.	z	p-value	95% conf. interval
House Type	0.0233	1.987	0.12	0.906	(-0.366 0.413)
House Size (square meters)	-0.00081	0.00059	-1.36	0.174	(-0.0198 0.000358)
Distance to Coast (meters)	0.00718	0.00069	10.42	0.000	(0.00583 0.00853)

172

(b) Factors in damage state	Coef.	Std. Err.	z	p-value	95% conf. interval
House Type	-1.64	0.207	-7.92	0.000	(-2.05 1.236)
House Size (square meters)	-0.04961	0.001	-4.88	0.000	(-0.069 0.0029)
Distance to Coast (meters)	-0.0002145	0.00058	-0.37	0.713	(-0.0136 0.00093)