

Motivation Agreement between exposure metrics Software Conclusions

Assessing county-level exposure to hurricanes and other tropical storms in the United States for epidemiological research

CHE: 20 Pioneers Under 40 in Environmental Public Health

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Health risks associated with Hurricane Sandy (2012)



Source: NOAA / NASA GOES Project

Health risks in storm-affected areas

- Change in patterns of emergency department visits (Kim et al. 2016)
- Increased outpatient cases of food and waterborne disease among elderly (Bloom et al. 2016)
- Increased rate of myocardial infarctions (Swerdel et al. 2014)
- Increased hospitalizations for dehydration (Lee et al. 2016)
- Difficulty obtaining medical care, medications, and medical equipment (Davidow et al. 2016)

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Hazard-specific tropical storm metrics

Tropical storm hazard metrics

- Distance from the storm
- High winds
- Rainfall
- Storm surge
- Flood events
- Tornado events



Image sources: Los Angeles Times, NBC

3 / 16

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Assessing tropical storm exposure

Challenge for epidemiological research

How should we determine whether a county was exposed to a tropical storm for epidemiological research?



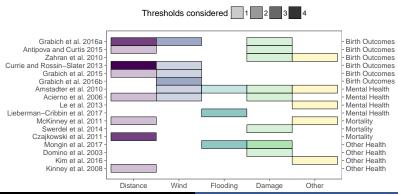
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Challenge for epidemiological research

How should we determine whether a county was exposed to a tropical storm for epidemiological research?



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Project aims

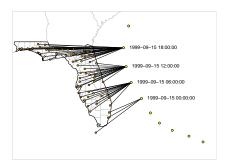
Project aims

- Develop exposure classifications of all U.S. Atlantic basin tropical storms, 1996–2011, based on reasonable measurements of tropical storm hazards
- Assess agreement between hazard-based county-specific exposure classifications
- Make exposure assessments accessible to other researchers for epidemiological and other impact studies

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Assessing tropical storm exposure



Example of "Best Tracks" data

Distance metric

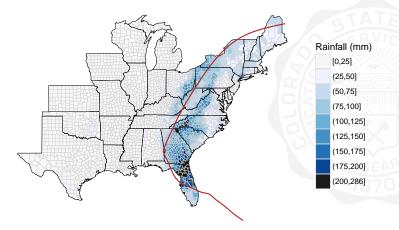
- **Distance:** National Hurricane Center Best Tracks data
- Wind: Wind model based on Willoughby et al. (2006)
- Rain: Re-analysis rain data (NLDAS-2)
- Flood and tornado events: NOAA Storm Events database

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Rain exposure

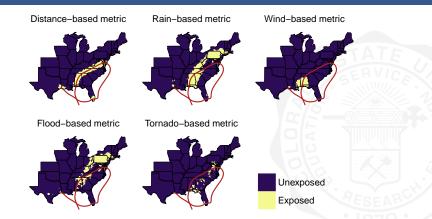
Rainfall during Frances, 2004



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County-level exposure to Hurricane Ivan (2004)



Criteria for exposure classifications: Distance: Within 100 kms of storm track. Rain: \geq 75 mm of rain total for two days before to one day after storm. Wind: Modeled wind of \geq 15 m/s. Flood, Tornado: Listed event in NOAA Storm Events database.

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County-level agreement in storm exposure

Assessing agreement in county classifications

For each storm and each pair of metrics, we measured the *Jaccard index* as a measure of county-level agreement in exposure classification for a storm:

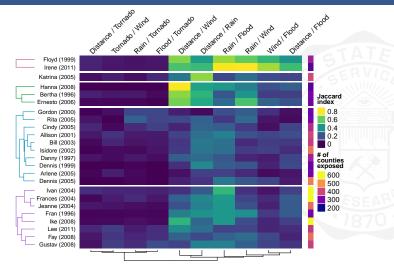
$$J = \frac{X_1 \cap X_2}{X_1 \cup X_2}$$

where X_1 is the set of counties exposed to a storm based on the first metric and X_2 is the set of counties exposed to the storm based on the second metric.

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County-level agreement in storm exposure



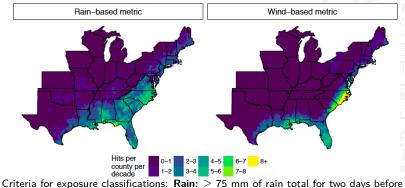
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11 / 16

Tropical storm exposure in U.S. counties

Storm hits per county per decade based on rain (left) and wind (right) exposure metrics.



Criteria for exposure classifications: Rain: \geq 75 mm of rain total for two days before to one day after storm. Wind: Modeled wind of \geq 15 m/s.

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Project software

'hurricaneexposure'

Create county-level exposure time series for tropical storms in U.S. counties. Exposure can be determined based on several hazards (e.g., distance, wind, rain), with user-specified thresholds. On CRAN.

##	#	A tibble: 4 3	ς 5			
##		storm_id	fips	closest_date	storm_dist	tot_precip
##		<chr></chr>	< chr >	<chr></chr>	<dbl></dbl>	<dbl></dbl>
##	1	Bill-2003	22071	2003-06-30	38.78412	141.1
##	2	Charley-2004	51700	2004-08-14	43.01152	136.2
##	3	Cindy-2005	22071	2005-07-06	32.21758	113.2
##	4	Floyd-1999	51700	1999-09-16	46.50729	207.5

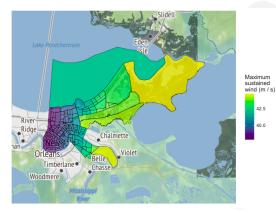
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Project software

'stormwindmodel'

Model storm winds from Best Tracks data at U.S. locations. Includes modeling sustained and gust winds, as well as duration of sustained and gust winds above a specified threshold. On CRAN.



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Project software

'countyweather', 'countyfloods'

Download weather monitor data through NOAA and USGS APIs by U.S. county. Includes functions to map available monitors / gages for each county. On CRAN.

'noaastormevents'

Download and explore listings from the NOAA Storm Events database. Includes the ability to pull events based on a tropical storm, using events listed close in time and distance to the storm's tracks. On CRAN.

'countytimezones'

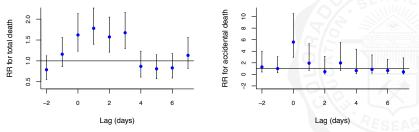
Convert time-stamps from UTC to local time zones for U.S. counties based on county FIPs. Facilitates merging weather observations with locally measured data, including health outcomes. On CRAN.

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Continuing work

Relative risk for all-cause (left) and accidental (right) mortality in Miami, FL, at lags from the Hurricane Andrew storm day (lag 0) compared to non-storm days.



Estimates were obtained by comparing storm days to matched non-storm days in the same time of year and day of week in other years. Matched days were picked to exclude days near other storms. Lag 0 represents the storm day. Negative lags represent days before the storm and positive lags represent days after the storm. Vertical lines give 95% confidence intervals.

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Collaborators

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