Spatial Grids for Hurricane Climate Research A Meteorologist-turned-Geographer's Perspective

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Take Home Points

- Here I introduce a new spatial framework for studying hurricane climate change.
- I show hexagons are efficient at covering hurricane tracks.
- They provide a scaffolding for combining hurricane attribute data and spatial climate data.
- Hurricane seasons having similar tracks can be assessed and grouped.
- Regional cyclone frequency and intensity variations can be mapped.

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Take Home Points (continued)

- Geographically-weighted regression of cyclone intensity on SST is performed using the hexagons.
- Results confirm the importance of warm oceans to cyclone intensity, especially over regions where the heat content is largest.
- Greatest mismatch between observed and predicted intensity occurs near land.
- The framework is ideally suited for comparing tropical cyclones generated from climate models.

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Problem: Storm tracks don't match climate data



Solution: Tracks as grids; New problem: Efficiency



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Hexagons: More efficient at covering tracks



Hexagons: More efficient at covering tracks



Analogue seasons



How well do they match?

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How well do they match?

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Hurricane frequency and intensity



Sea surface & upper air temperatures



Grouped by SST and Intensity



Low SST, Low Intensity
Low SST, High Intensity
High SST, Low Intensity
High SST, High Intensity

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Export as KML layer



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$$I = \frac{n}{s} \frac{y^T W y}{y^T y}$$

• *n* is the number of hexagons

- y is the vector of cyclone intensities one for each hexagon [deviations from the overall mean]
- *W* is a weights matrix indicating which hexagons are neighbors
- s is the sum over all the weights, and
- subscript T is the vector transpose operator

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Spatial autocorrelation: Statistical significance



Ordinary least-squares regression

$$y = X\beta + \varepsilon$$

- β is a vector of regression coefficients
- $\varepsilon \sim N(0, \sigma^2)$ is a vector of independent and identically distributed residuals with variance σ^2

Maximum likelihood estimate for β

$$\hat{\beta} = (X^T X)^{-1} X^T y$$

Ordinary vs geographically-weighted regression

Geographically-weighted regression

$$y = X\beta(g) + \varepsilon$$

where g is a vector of hexagon locations

Maximum likelihood estimate for $\beta(g)$

$$\hat{\beta}(g) = (X^T W X)^{-1} X^T W y$$

where W is a weights matrix given by

Weights matrix

$$W = \exp(-D^2/h^2)$$

where D is a matrix of pairwise distances between the hexagons and h is the bandwidth.

Geographically-weighted regression: Results



Geographically-weighted regression: Does scale matter?



Geographically-weighted regression: Residuals







Local Poisson regression: Trends in hurricane frequencies



Observed vs modeled (CMIP3) tropical cyclones



Hurricanes since 1943



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- It is ideally suited for quantitative comparisons of tropical cyclones generated from Global Climate Models (GCMs).

Thank you for your attention. Questions?

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