Analyzing the Extreme Behavior of ...

Large-Scale Meteorlogical Variables Found To Have Influence on
Severe Storms and Tornadic Events Using Global Reanalysis Data

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Outline

- Motivation and Goals
- Reanalysis Data
- Preliminary Analysis
- Preliminary Results
- Future and Ongoing Work

Motivation and Goals

Primary Goal: Project Frequency and Intensity of Severe Weather







Motivation and Goals

Challenges

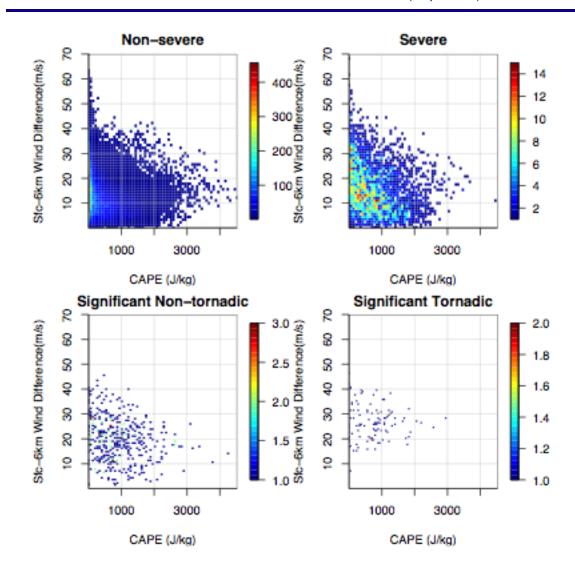
- Severe convective storms not resolved by Global Models or Datasets
- Historical records limited
- Weak relationship to larger-scale phenomena

Motivation and Goals

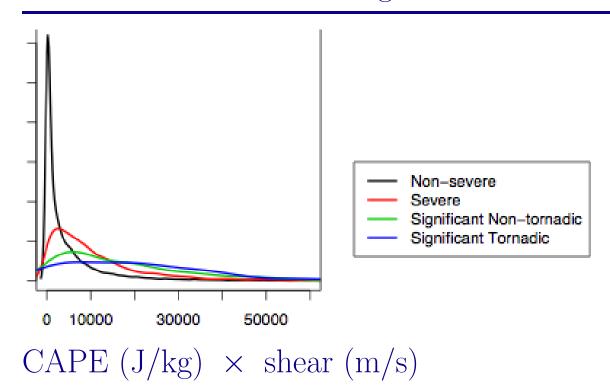
Intermediate Goals: Large-scale indicators for severe weather

- Characteristics
- Verification
- Past and Future trends

Large-scale indicators: CAPE (J/kg) and 6-km shear (m/s)



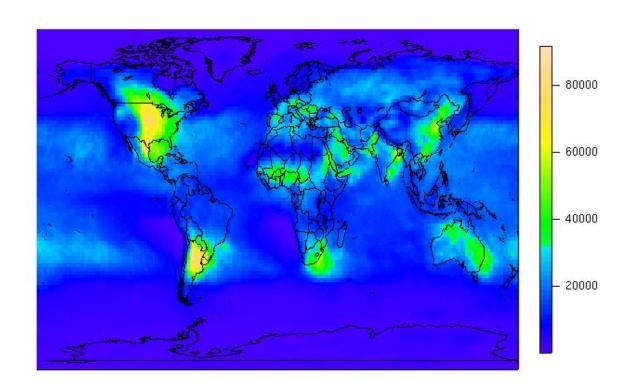
Motivation and Goals: Large-scale indicators



Reanalysis Data

- Resolution $\approx 1.875^{\circ}$ longitude by 1.915° latitude
- 17 856 grid point locations (192 \times 94 grid)
- Temporal spacing every 6 hours
- 1958 through 1999 (42 years)
- Convective available potential energy (CAPE, J/kg)
- Magnitude of vector difference between surface and 6-km wind (shear, m/s)
- Both CAPE and shear ≥ 0 (Lots of zeros!)

Reanalysis Data



Upper quartile of annual maximum CAPE × shear

- Initially looked at trends in mean CAPE, shear and CAPE \times shear
- Trends in counts of CAPE \times shear > 10000
- Trends in counts of CAPE \times shear > 20~000
- Annual maximum CAPE × shear fit to GEV (ignoring spatial dependence!)

Initially modeling counts with a generalized linear model (GLM) with negative binomial family.

CAPE \times shear seldom exceeds 10 000 over most of the globe. Therefore, only areas where an exceedance occurred at least once per year are examined.

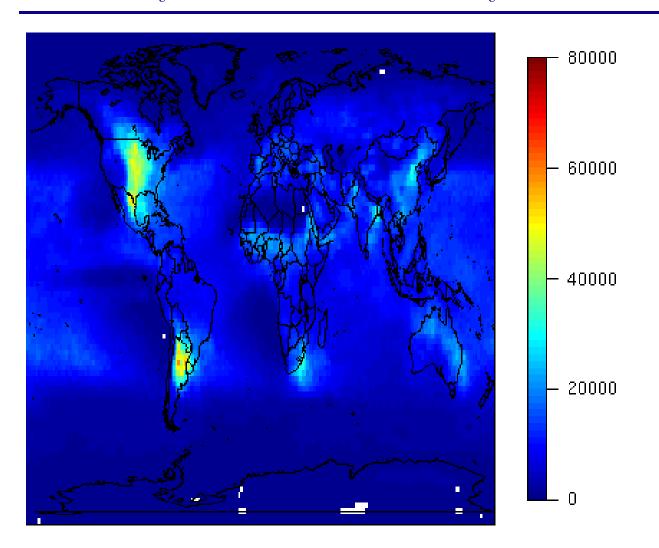
Although no spatial analysis performed (yet), multiple comparisons and spatial dependence are accounted for by way of protecting against the false discovery rate (FDR) using the methods of Ventura $et\ al.$ (2004).

Extreme-value theory

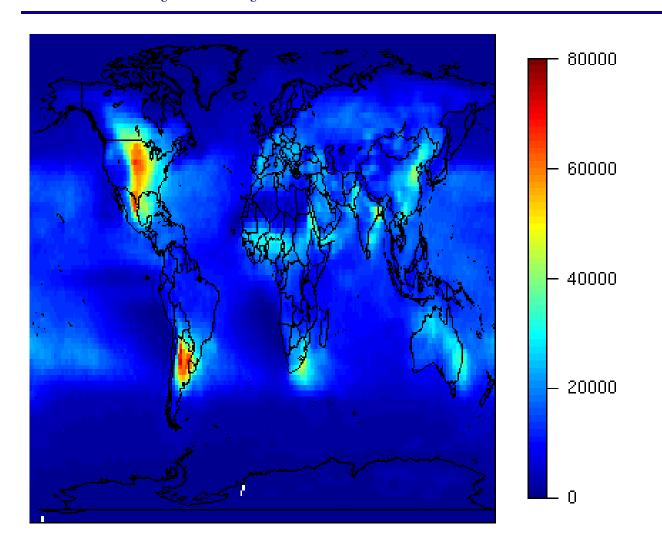
- Bivariate extremes difficult because CAPE and shear tend not to be large together
- Nonstationary spatial structure

Initial analysis: just fit GEV to individual grid points without worrying about spatial structure.

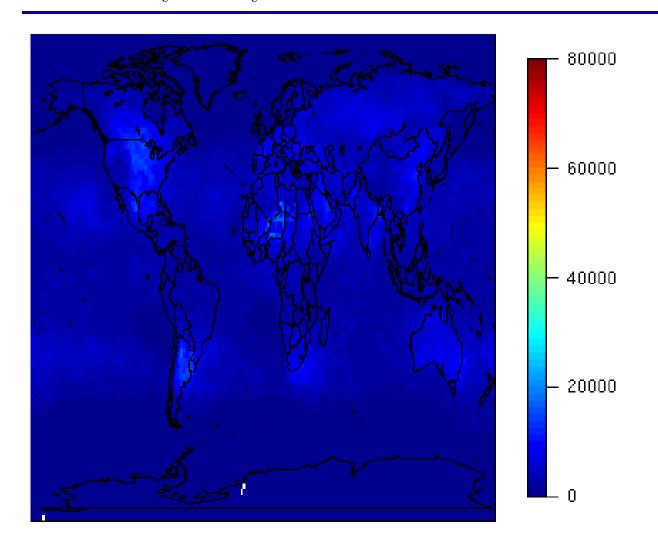
Preliminary Results: Estimated 20-year Return Level



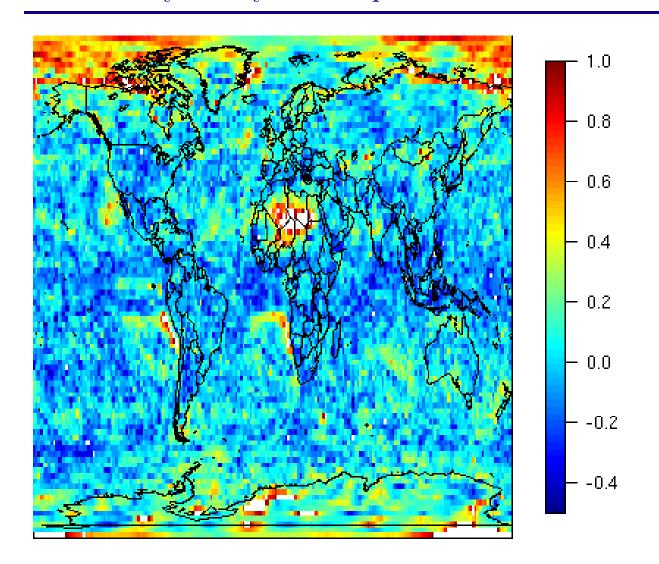
Preliminary Analysis: Location Parameter



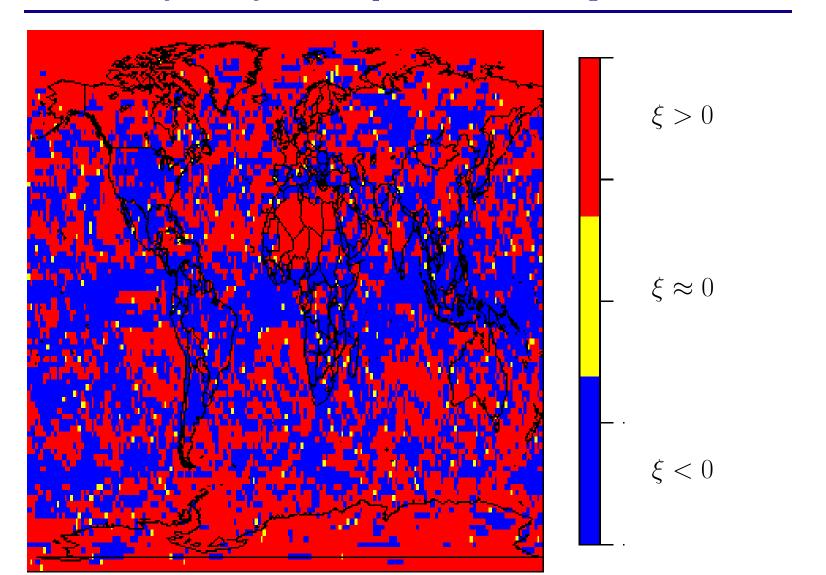
Preliminary Analysis: Scale Parameter



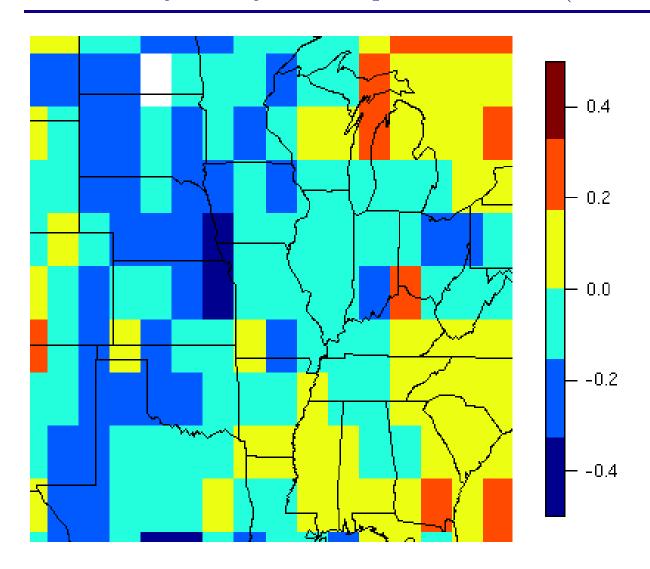
Preliminary Analysis: Shape Parameter



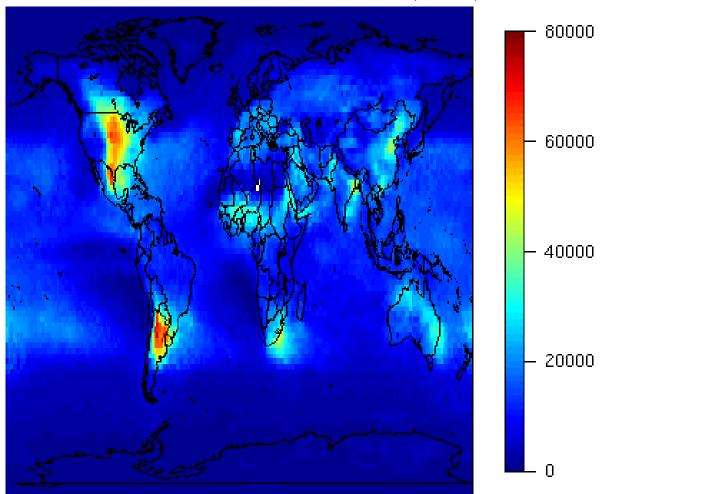
Preliminary Analysis: Shape Parameter Sign



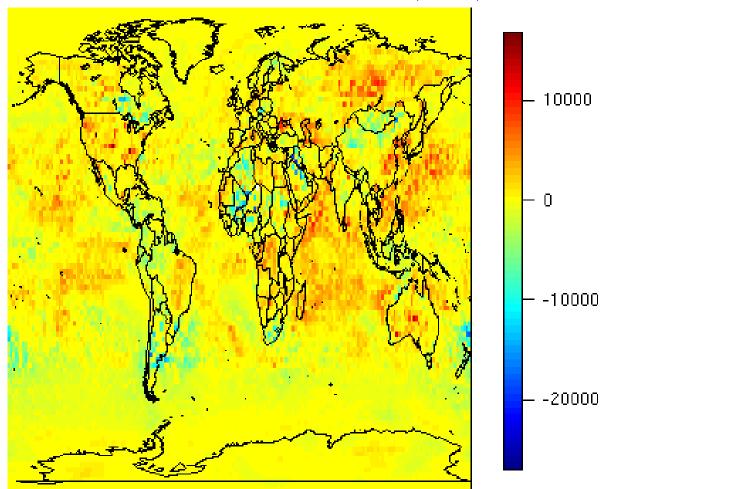
Preliminary Analysis: Shape Parameter (smaller region)



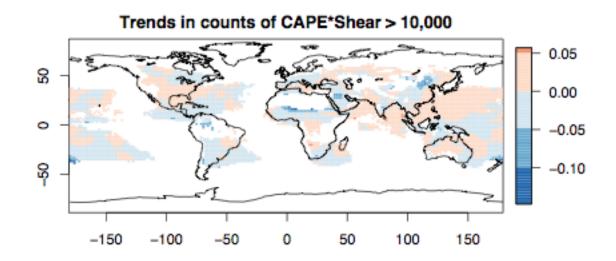
Trend in location parameter: $\mu(year) = \mu_0 + \mu_1 \cdot year$

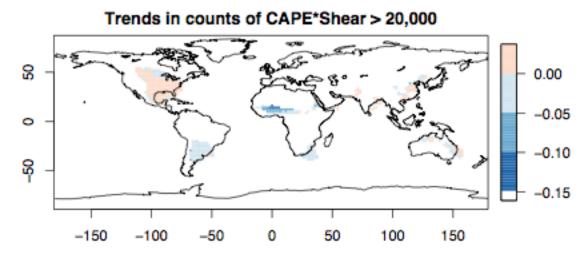


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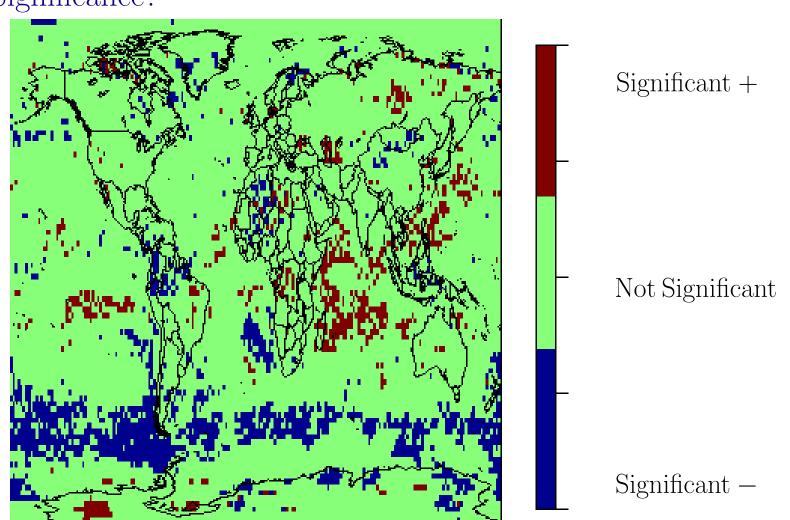
Preliminary Results





Preliminary Results

Significance?



Future and Ongoing Work

- Incorporate spatial dependence into the parameter estimates
- Incorporate spatial structure into the random processes
- Model counts of high CAPE × shear spatially
- Work out hypothesis testing for trends in location parameter
- Apply techniques to Global climate model data

References

Ventura, V., C.J. Paciorek, and J.S. Risbey. 2004. Controlling the proportion of falsely-rejected hypotheses when conducting multiple tests with climatological data. *J. Climate* **17**:4343-4356.