



A spatial propinquity extreme-value model for assessing monsoon-type precipitation extremes in future climates

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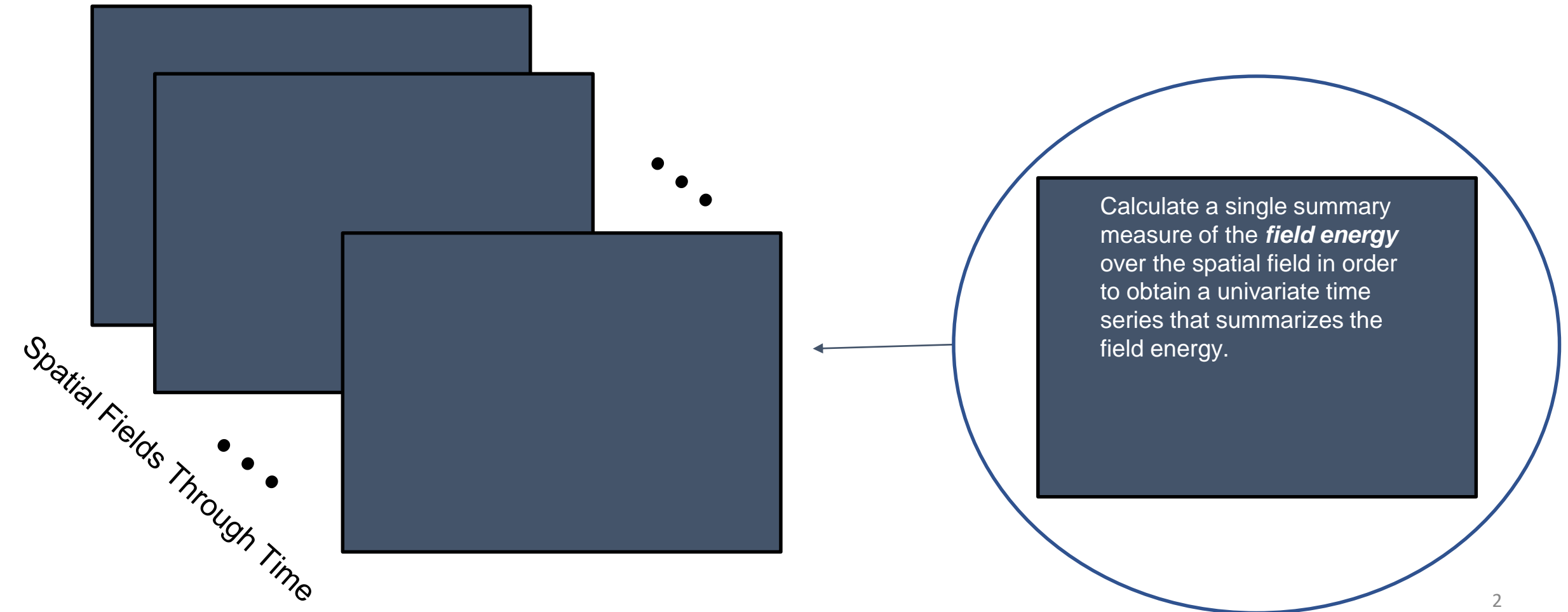
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University of Maryland

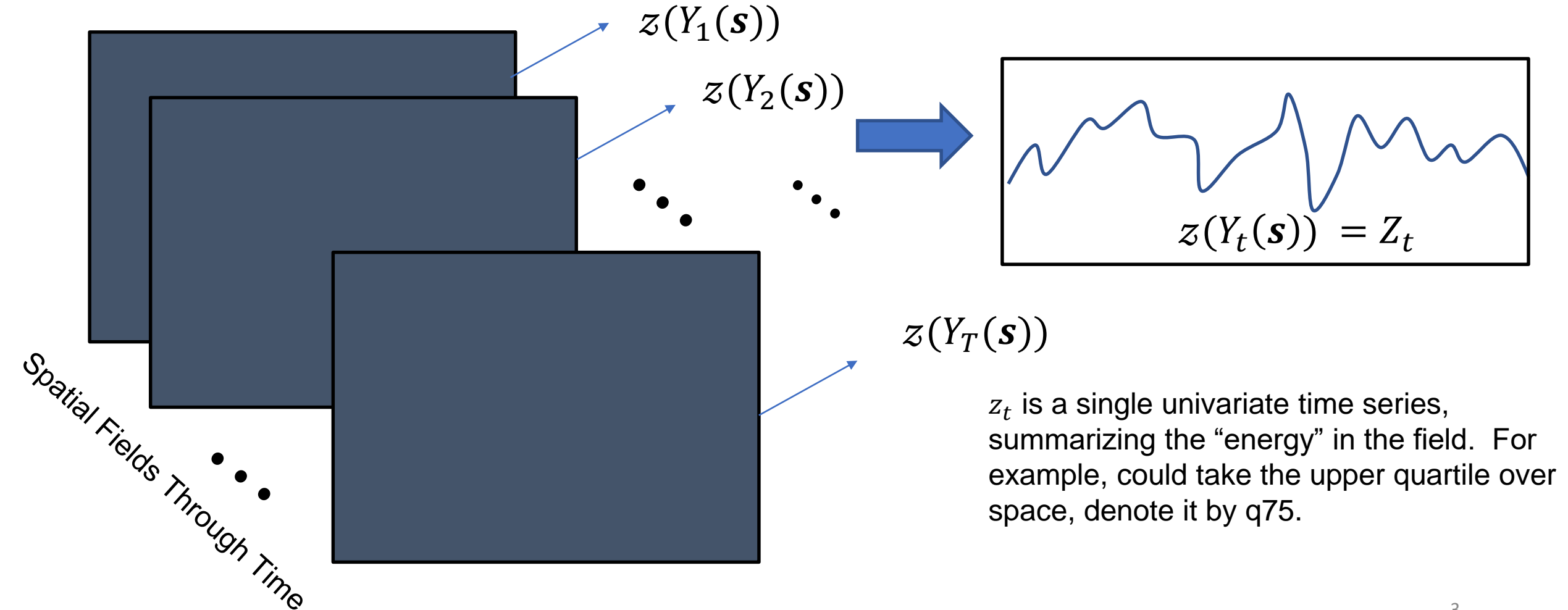
WET Weather 2019: Gregynog, Wales, U.K., 29 April to 1 May 2019

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Model Approach

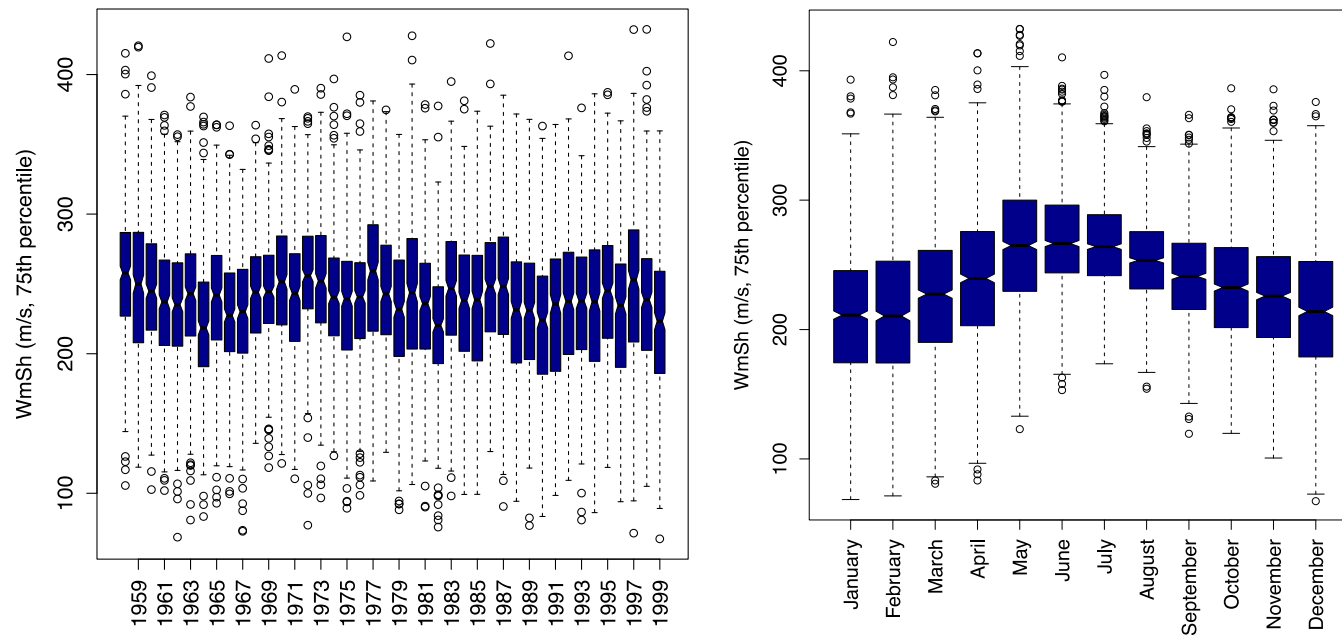


Model Approach



Example: WmSh (a large-scale severe storm indicator variable)

The upper quartile over space, $q75$, is a univariate time series that gives a measure of high WmSh over, possibly a small area of, space.



G. et al 2013. Spatial extreme value analysis to project extremes of large-scale indicators for severe weather. *Environmetrics*, **24** (6), 418 - 432, DOI: 10.1002/env.2234

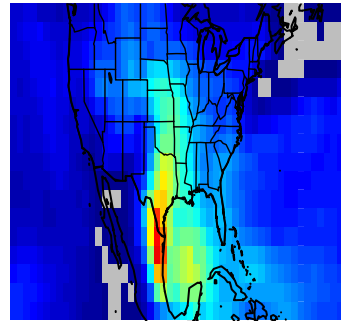
Example: WmSh (a large-scale severe storm indicator variable)

$$[WmSh_1, \dots, WmSh_n | q75 > u]$$

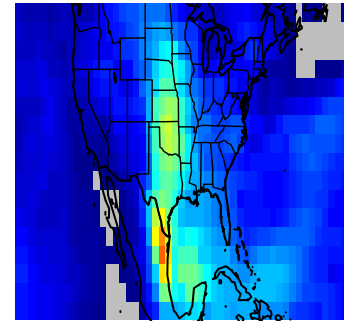


The mean of WmSh at each grid point conditioned on high q75. Using the Heffernan and Tawn (2004, <https://doi.org/10.1111/j.1467-9868.2004.02050.x>) conditional extreme value model. Henceforth HT2004.

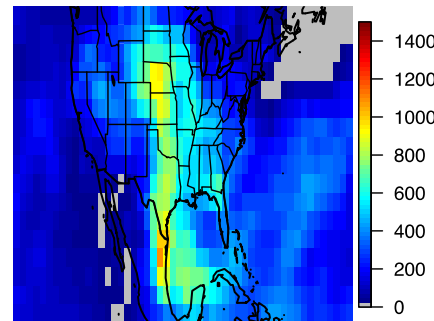
1958 – 1978



1979 – 1992



1993 – 1999



Spring
WmSh (m^2s^{-2})

G. et al 2013. Spatial extreme value analysis to project extremes of large-scale indicators for severe weather. *Environmetrics*, **24** (6), 418 - 432, DOI: 10.1002/env.2234

The field-energy stairway



Photo of Endless stairway at KPMG, [Munich, Germany](#) by Oliver Raupach, [Creative Commons Attribution-Share Alike 2.5 Generic](#)

The propinquity Model

From Wikipedia: In [social psychology](#), **propinquity** ([/prəˈpɪŋkwɪti/](#); from [Latin](#) *propinquitās*, "nearness") is one of the main factors leading to [interpersonal attraction](#).

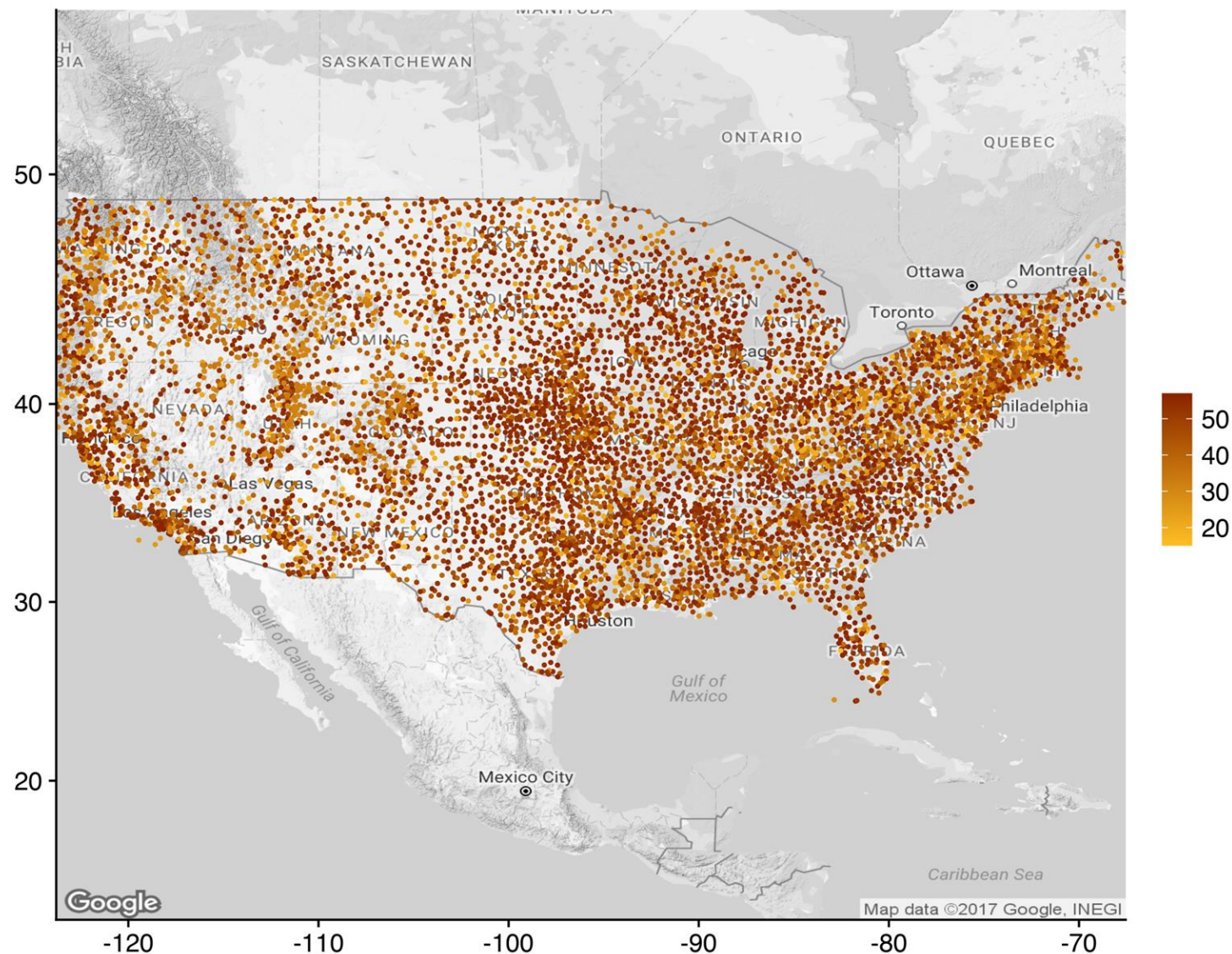
People who live on the same floor of a building have a higher propinquity than those who live on a different floor, unless they live near the staircase.

Using the term propinquity for the model from 2013 (also G. et al. 2016, DOI: [10.5194/ascmo-2-137-2016](#)). Though it is a new term not used in those papers.

Data

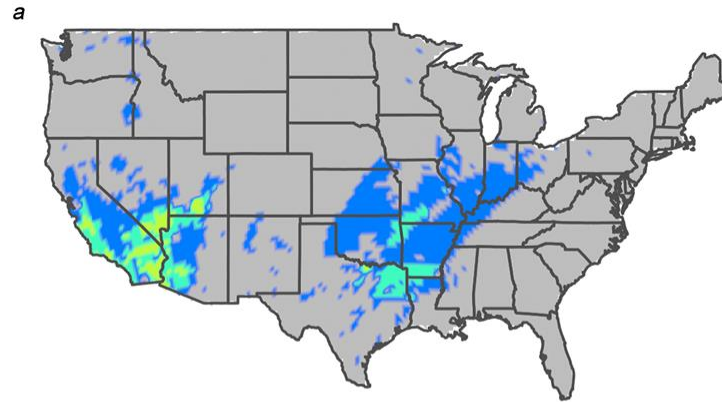
GHCN-Daily precipitation dataset

- Number of stations = 8516
- Stations are interpolated to a 30x30km grid.
- The color shade indicates the length of the observation period
- Generally, 56 years (1961 – 2016) of daily data in mm/day

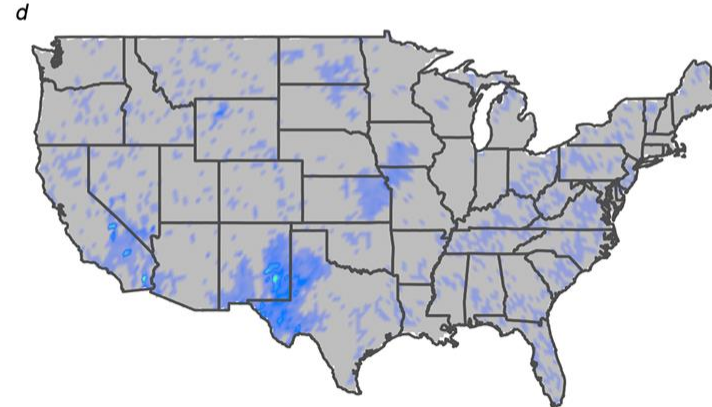
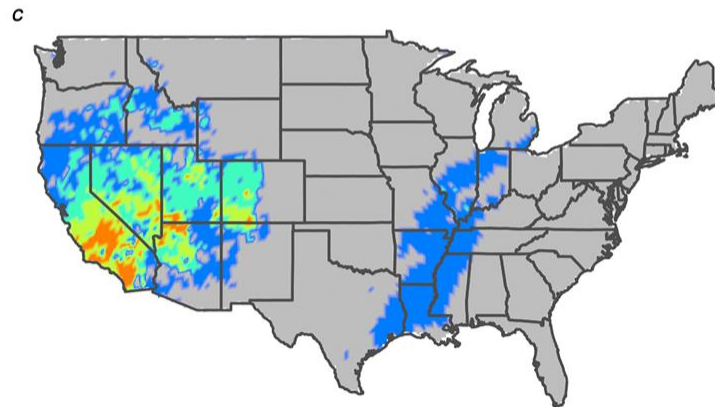
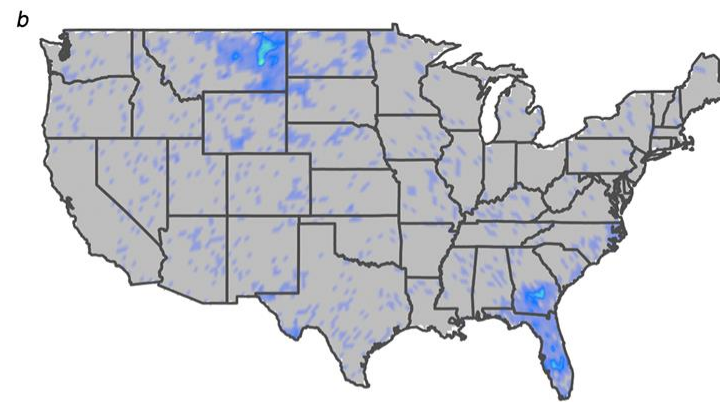


Examples

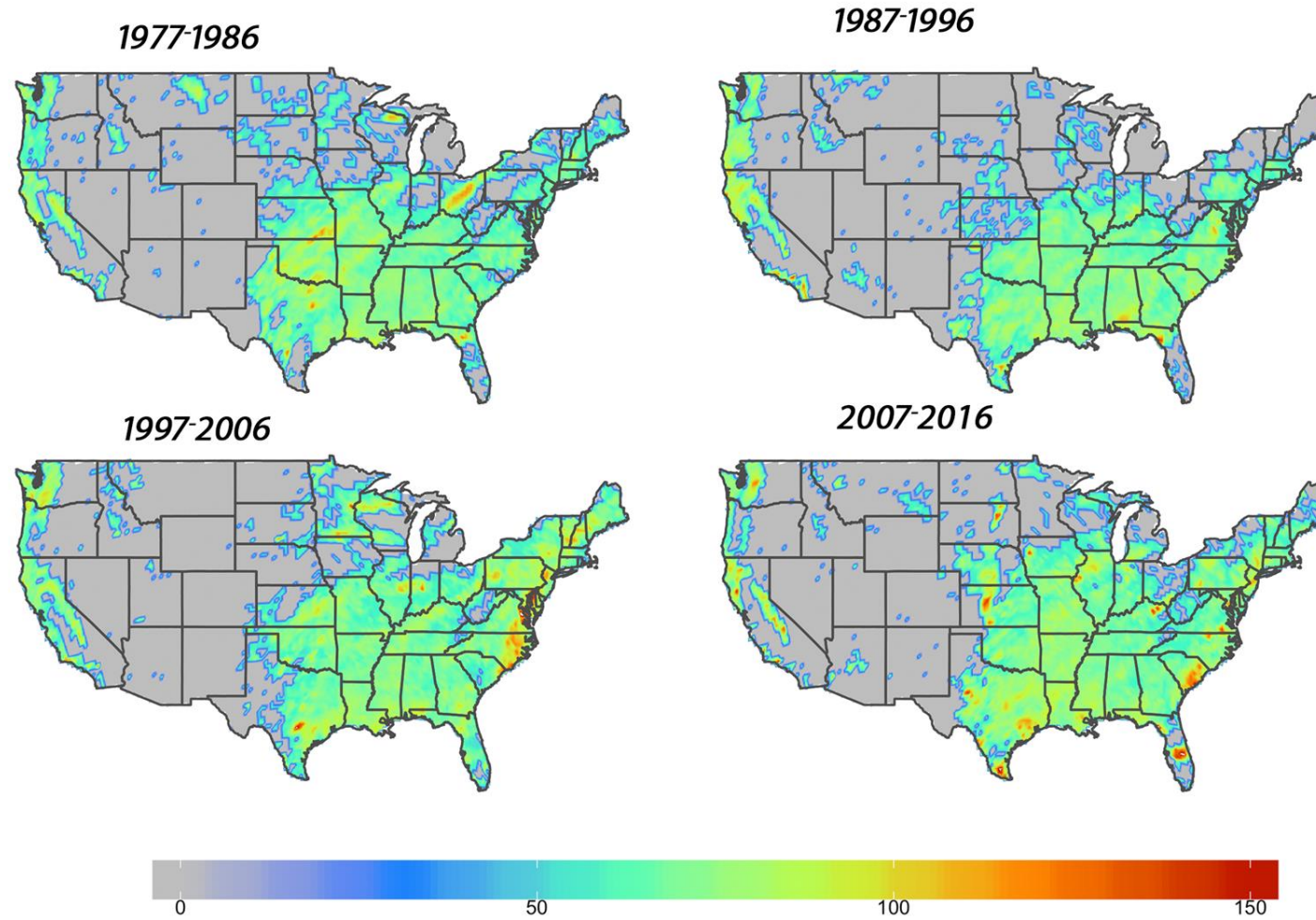
Propinquity Model



Traditional (individual grid points)

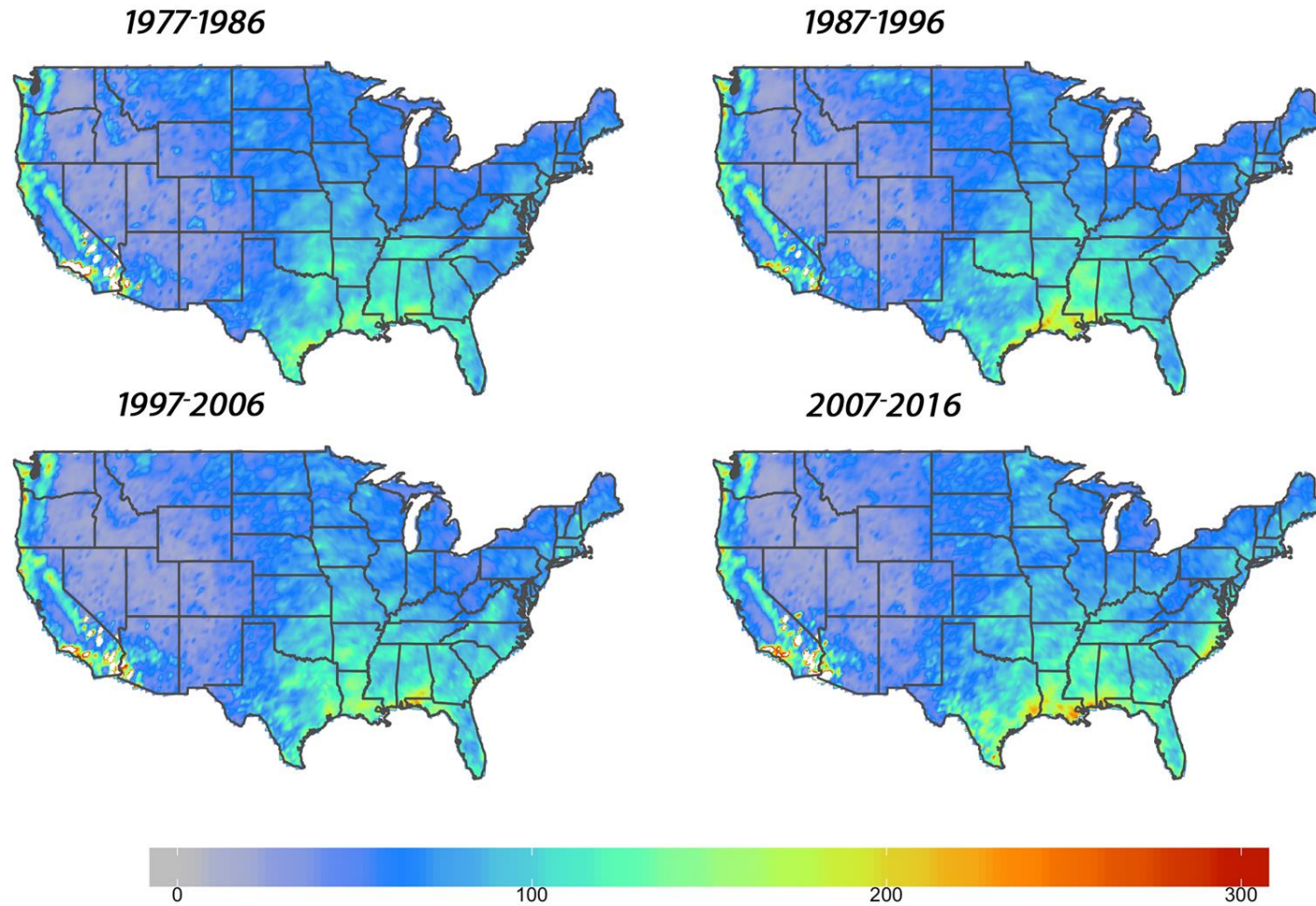


Empirical Propinquity Model



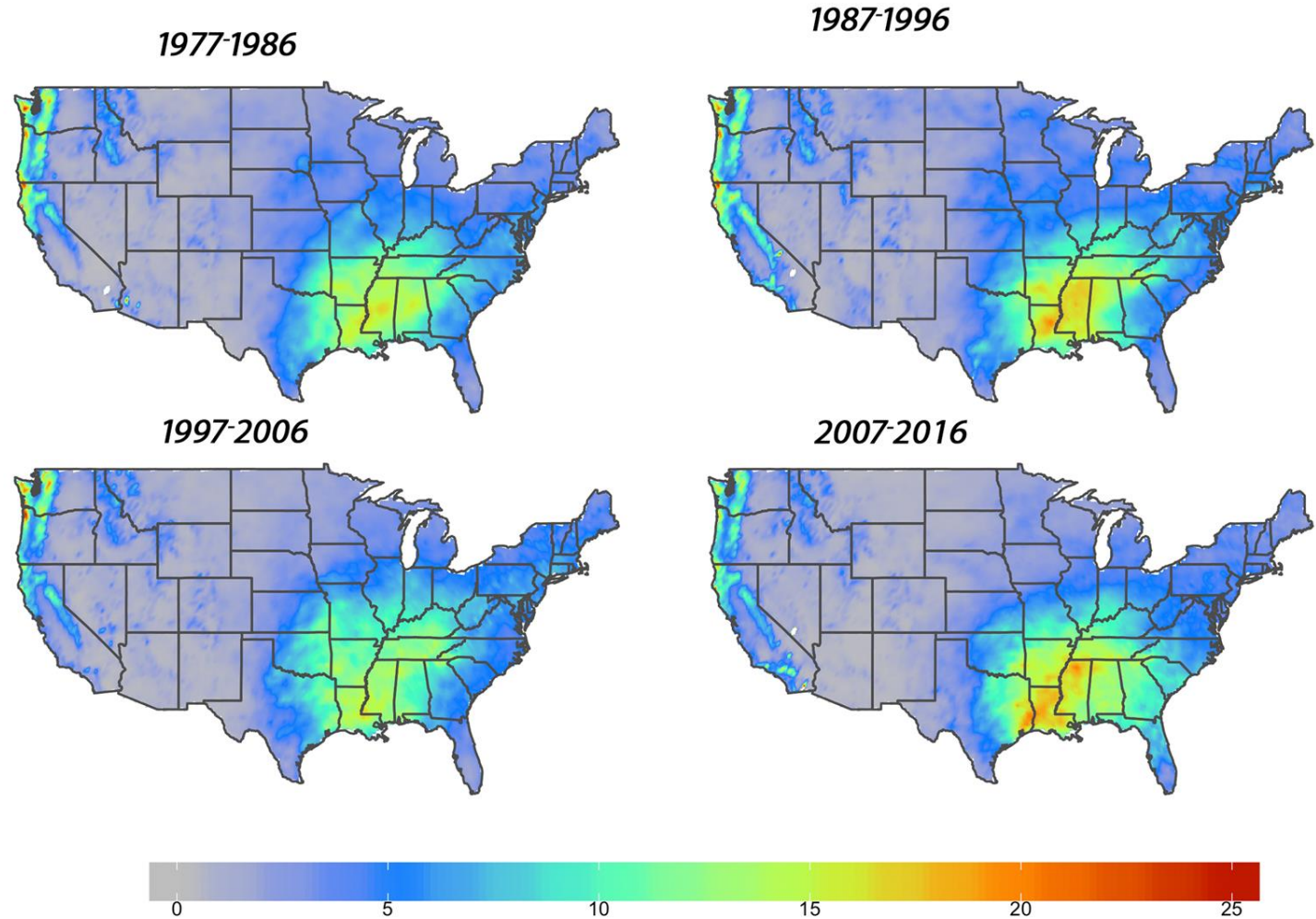
Marginal GPD's fit to data at each grid point

Threshold is set at 90th percentile.
10-year return values.



Propinquity Model with HT2004

Mean of simulated values
conditioned on q_{75} = 10-year return
level (based on its marginal GPD fit)



Summary

- Propinquity Model is a spatial model where the dependence results from the degree of **propinquity** of the random variables at each grid point to an overall summary of spatial field energy.
- Results in spatially cohesive fields of statistics that represent large-scale, process-oriented behaviors rather than localized phenomena.
- Spatial verification measures can be employed to analyze differences, e.g., between time periods, which can help to determine if significant changes result from changes in intensity and/or large-scale processes that have “moved” over time.
- **Field energy** might be defined through any statistic of interest calculated across space, e.g., the mean, upper quartile, median, sum, sum of squares, etc.
- Can be employed empirically or with a model, e.g., HT2004 is a natural choice if interest is in projecting extreme values.

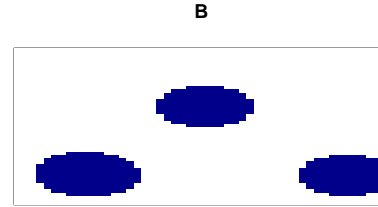
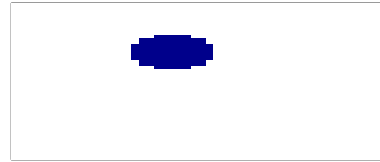
Thank you

Questions?

Note: The R package **texmex** was used for fitting the HT2004 model in these slides.

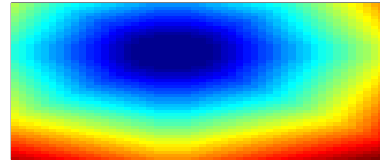
Harry Southworth, Janet E. Heffernan and Paul D. Metcalfe (2018). **texmex**: Statistical modelling of extreme values. R package version 2.4.2.

Mean Error Distance

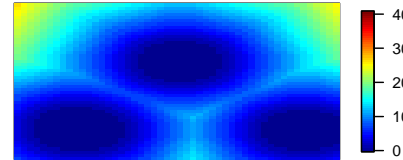


Distance map of A

$d(x, A)$

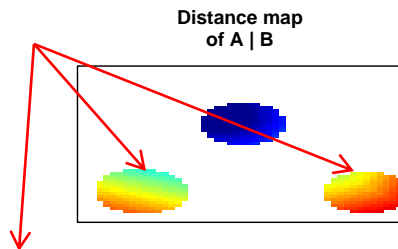


Distance map of B



$d(x, B)$

Very large because
much of B is very far
from A

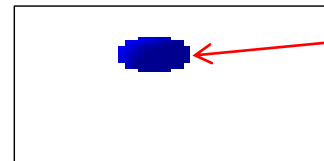


$$\text{MED}(A, B) = \sum_{x \text{ in } B} d(x, A) / N$$

N is the size of the domain

Average distance from B to A.

Distance map
of B | A



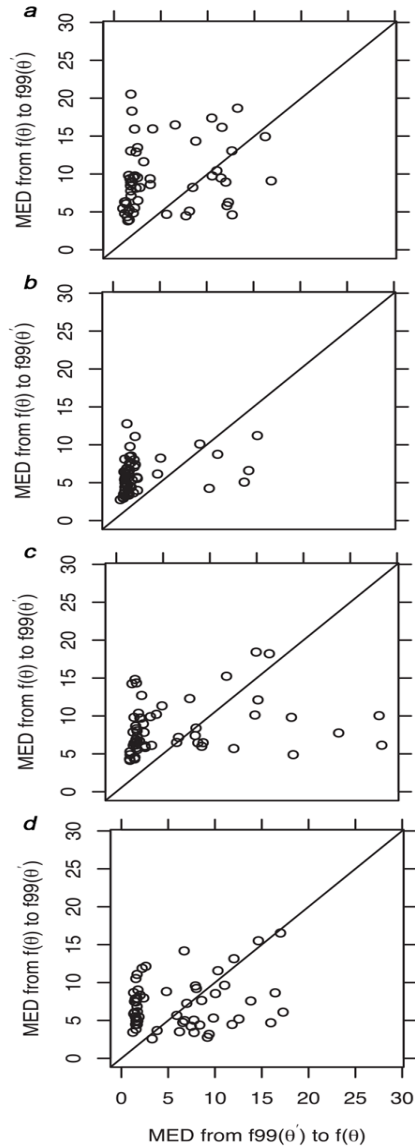
Very small
because all of A
is very close to
B in space

$$\text{MED}(B, A) = \sum_{x \text{ in } A} d(x, B) / N$$

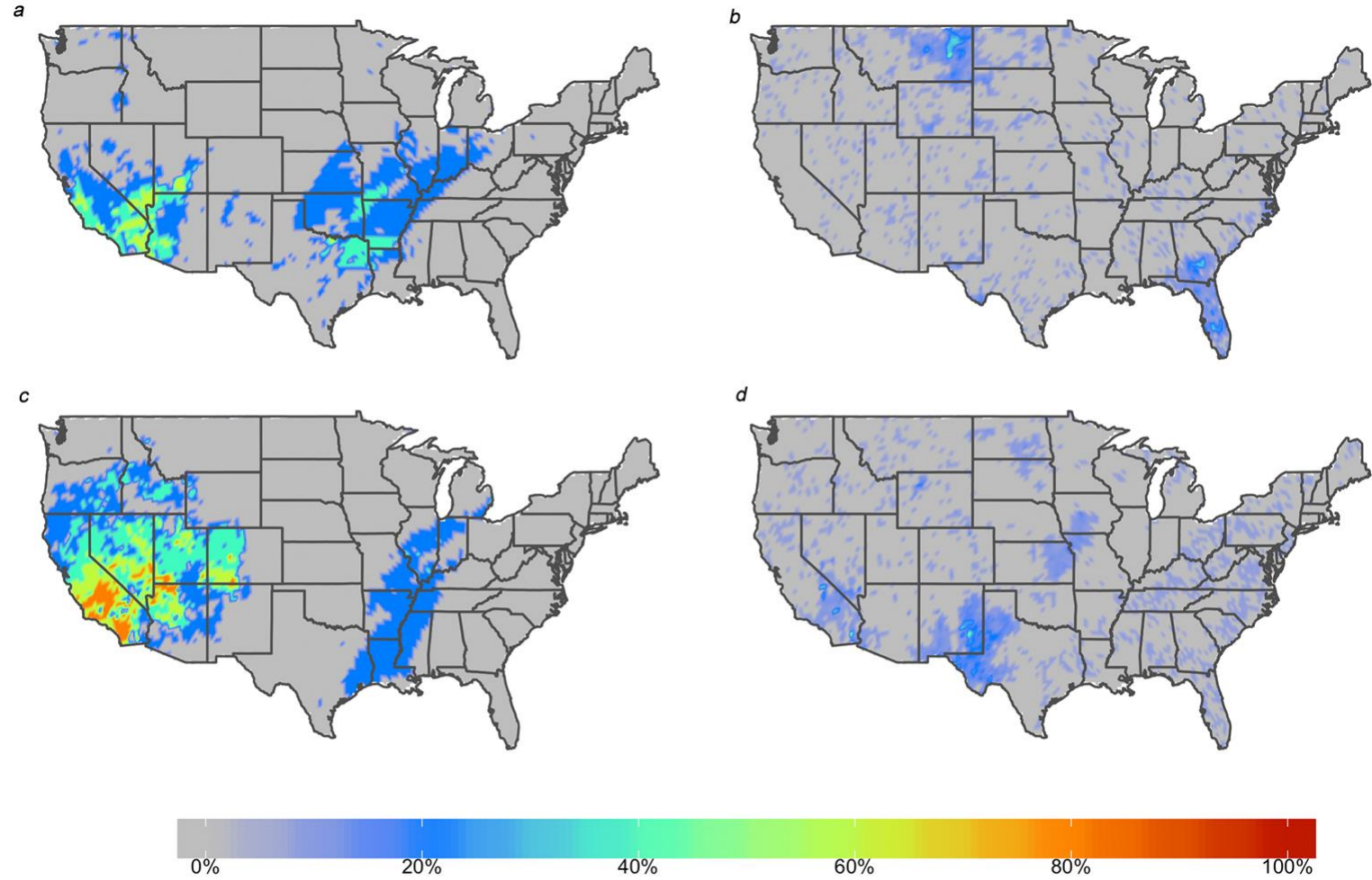
Average distance from A to B.

Examples

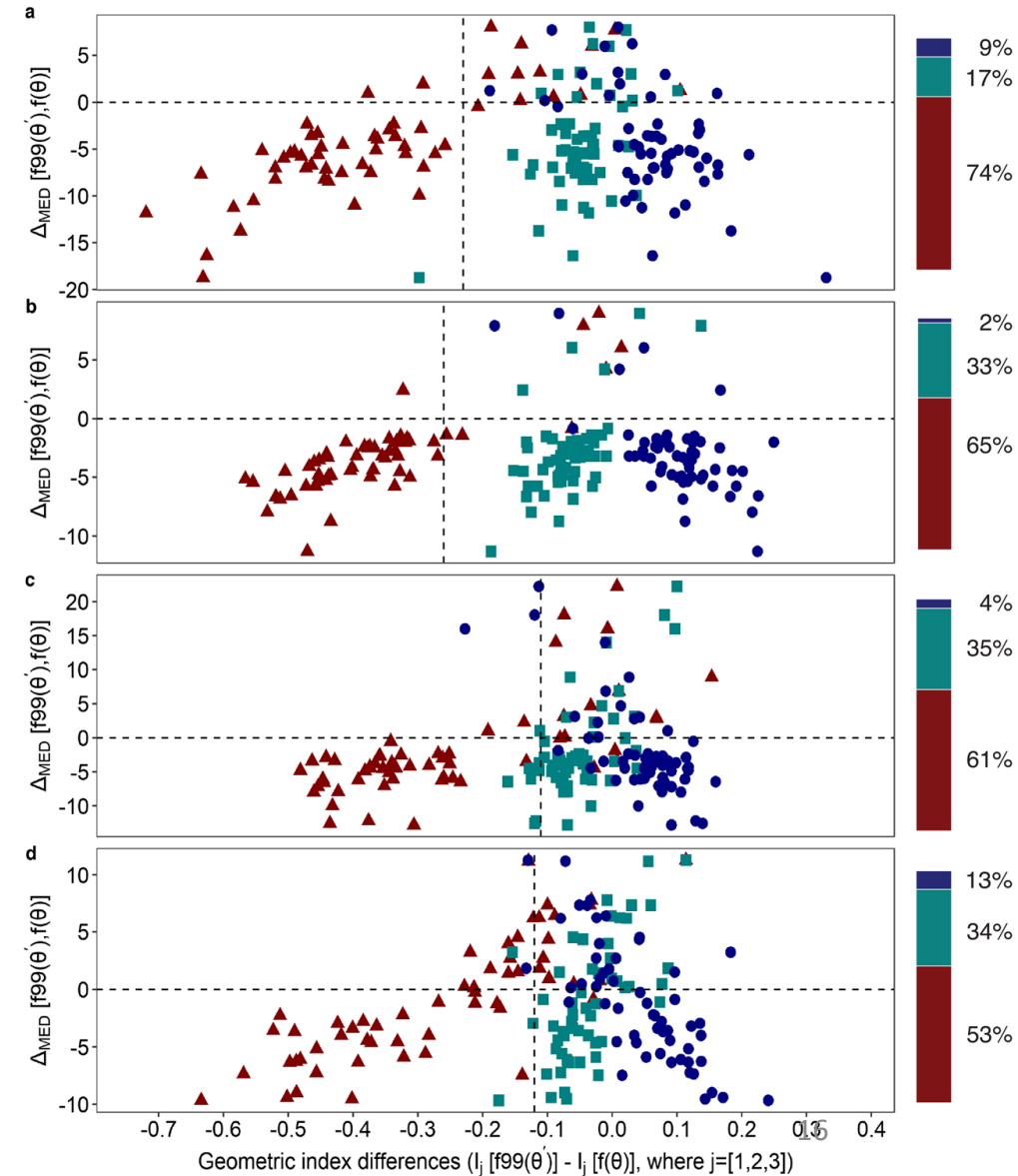
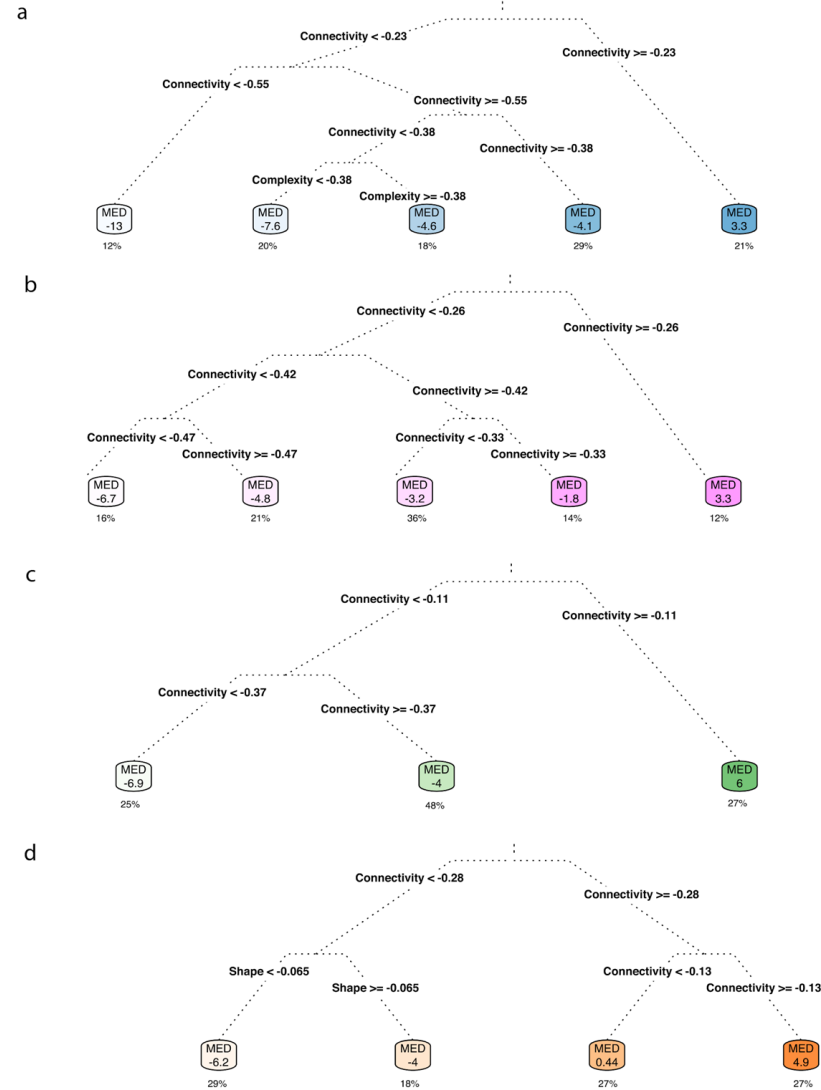
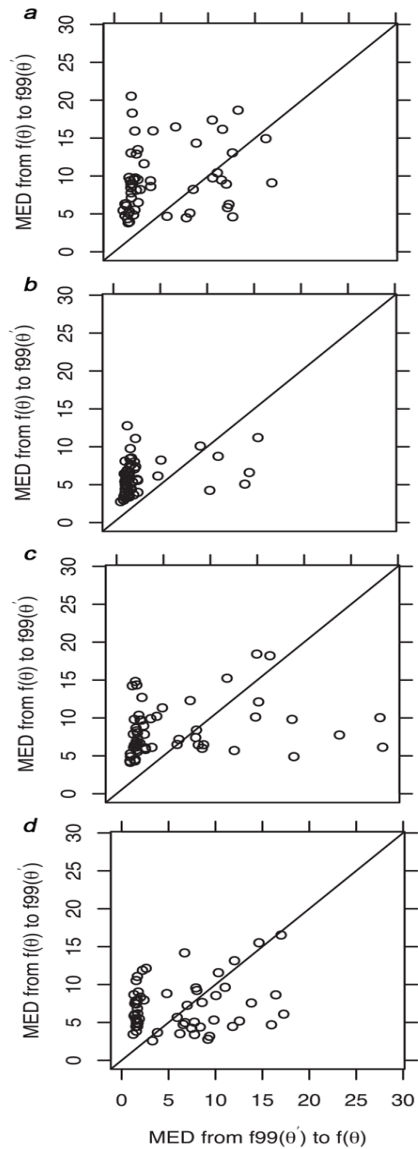
From univariate GP to (spatial) propinquity



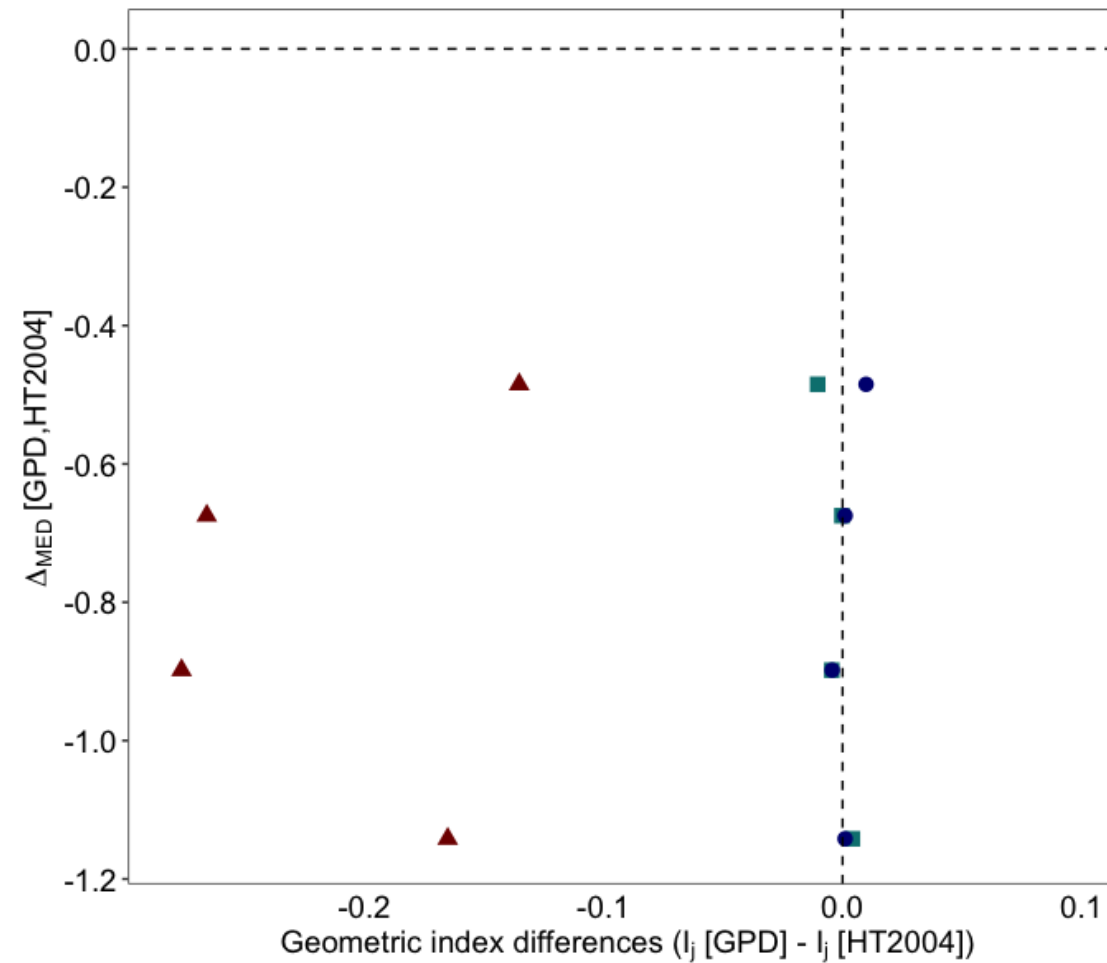
From propinquity to univariate GPD



Comparison (obs.)



Comparison (GPD vs. HT2004)



Threshold Selection Algorithm for Extremes

1. Essential Field Quantity (EFQ) is closely matched to the timeframe of interest
2. Dimension reduction based on statistical quantiles
3. Spatial domain mapping represented by geometric indices
4. Time series clustering and threshold selection

“Extremeness” is determined by an underlying process in space and time!