



# Spatial Statistics in Practice: When and Why does *Where* matter?

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# Outline

- Foundations
- Tools of the Trade
- Research example
- Lessons and Resources



# Foundations

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# Spatial Thinking

- By studying space, we study:
  - the shape, size, orientation, location, direction or trajectory of objects, processes or phenomena
  - the relative spatial positions of multiple objects, process or phenomena
- Location is a first-level descriptor
  - **Distributions** with patterns, density, and other sample properties
  - **Regions** with similar or different properties
  - **Hierarchies** with multiple nested levels
  - **Networks** of linked features, spatial or otherwise
  - **Surfaces** (3D)

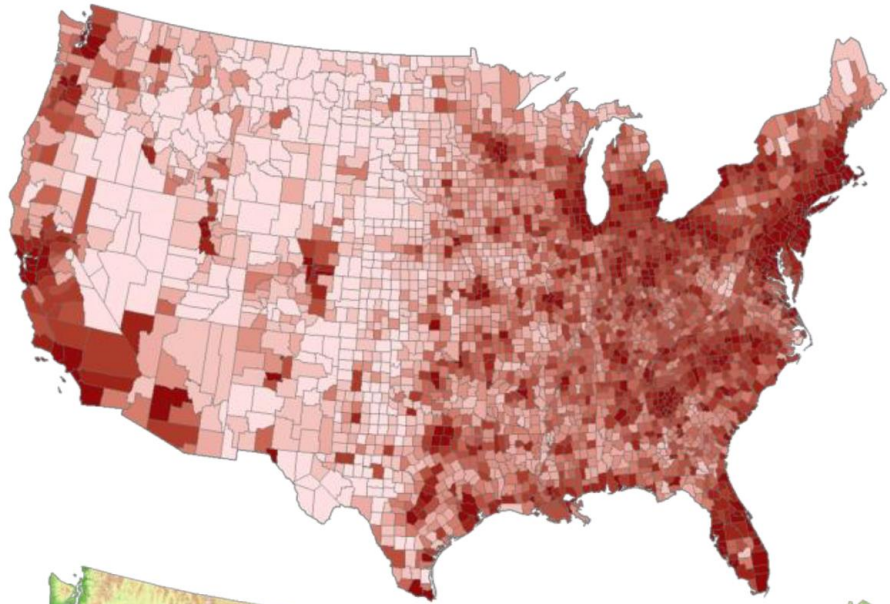
*“Everything is related to everything else.  
But near things are more related than  
distant things”*

*- Waldo R. Tobler, 1970*

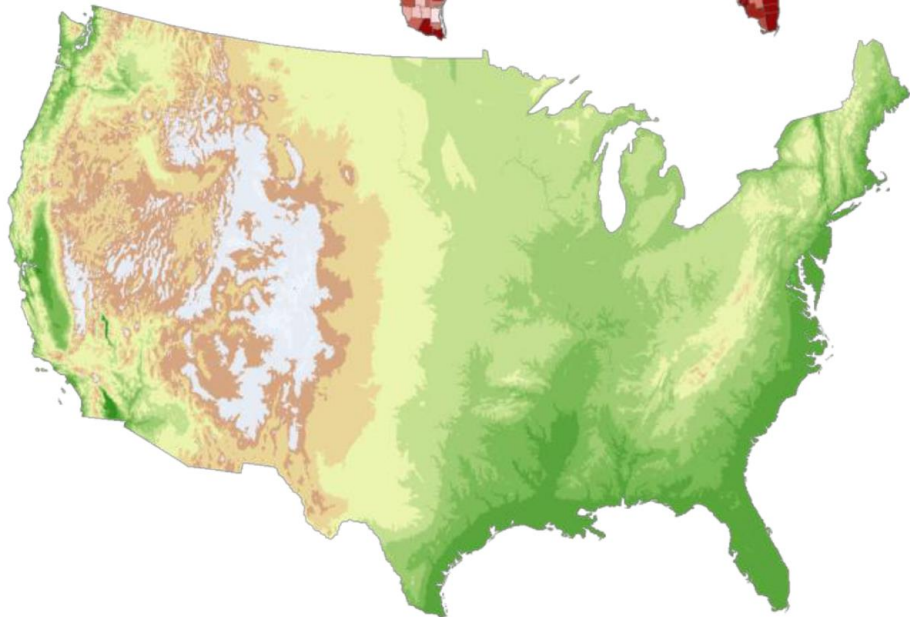
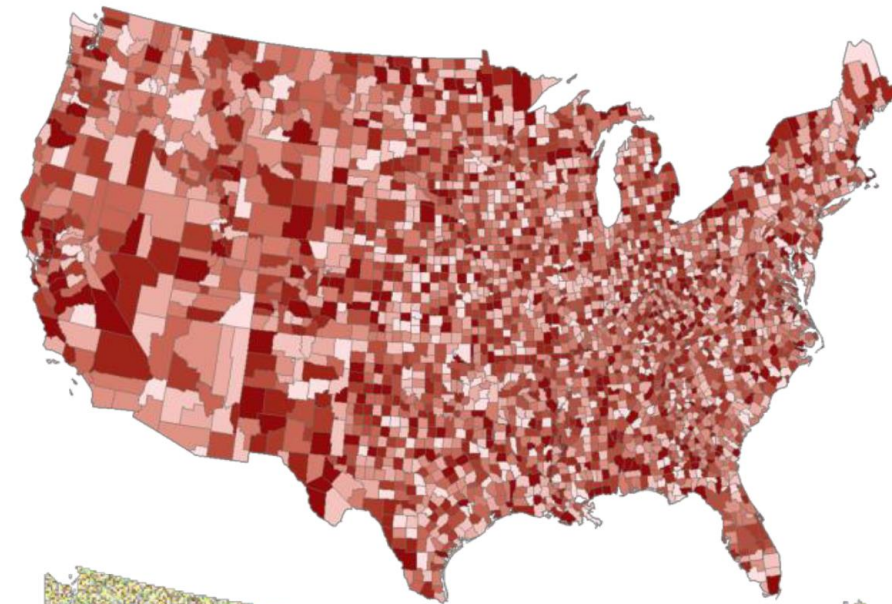
*Geographic variables exhibit  
uncontrolled variance*

*- Michael F. Goodchild, 2004*

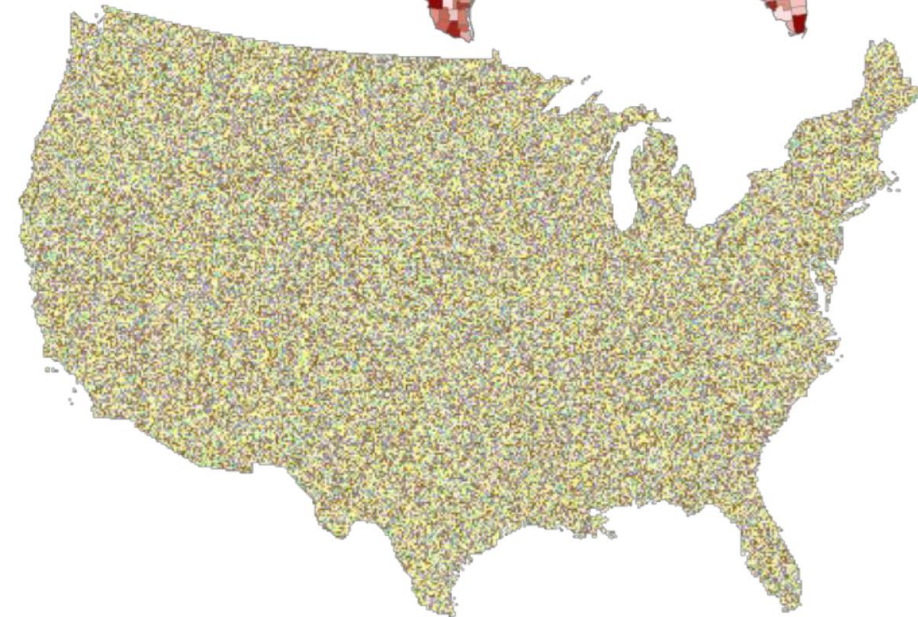
If features were randomly distributed ...



... population density map of the US would look like this

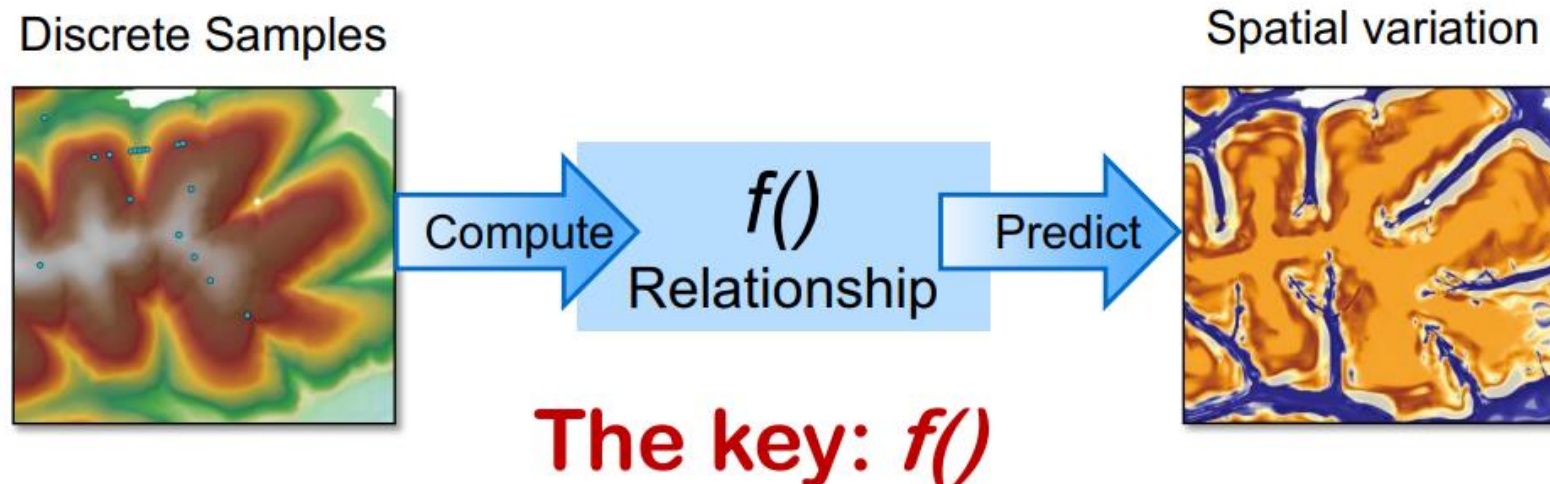


... elevation map of the US would look like this



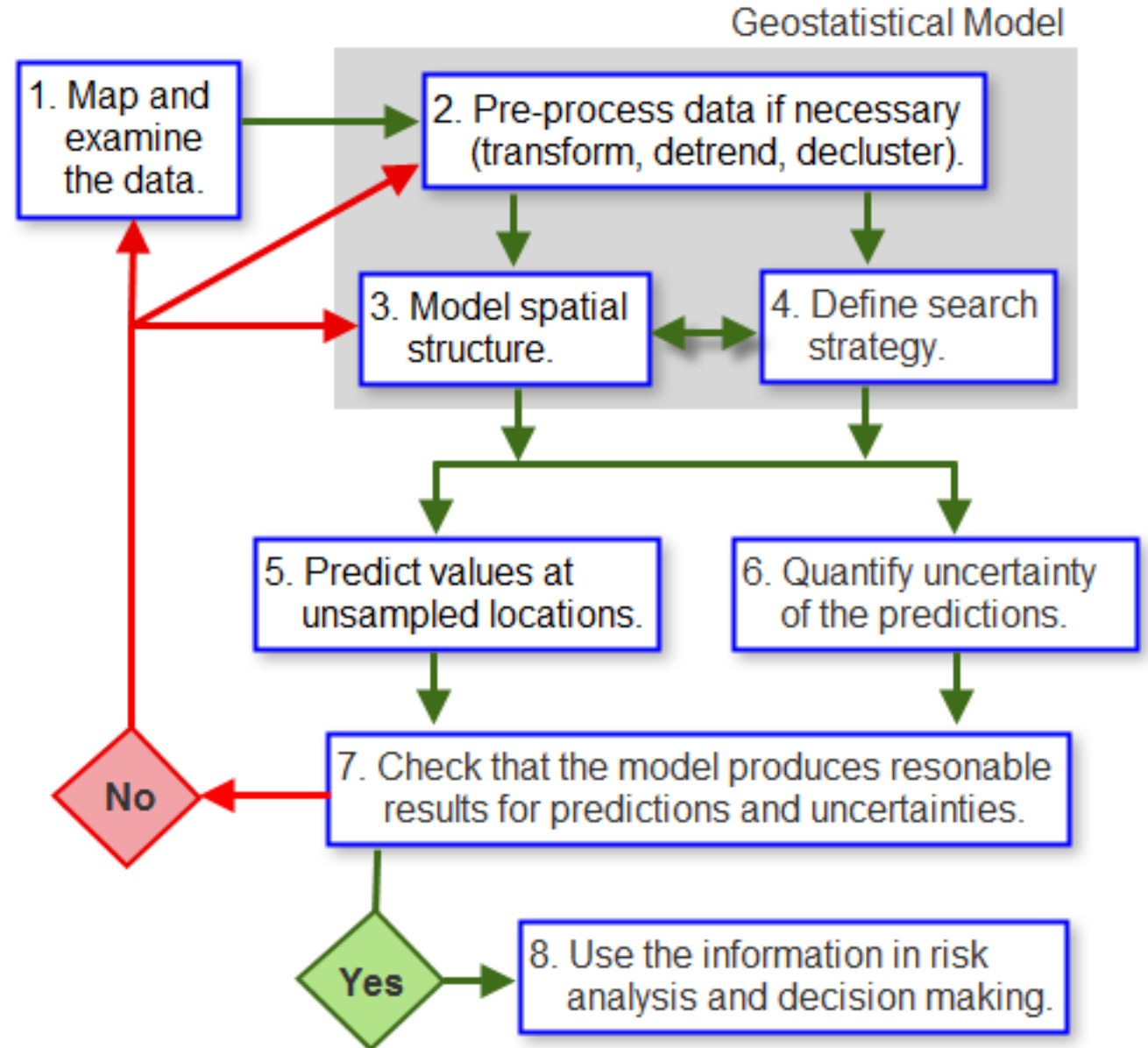
# Kinds of Analyses

- Descriptive : What happened?
- Diagnostic : Why did it happen?
- Predictive : What will happen?
- Prescriptive : What should happen?



# Workflow

- 0. Collect and extract the **right** data
  - From reputable data sources
  - For the correct spatial and temporal extent







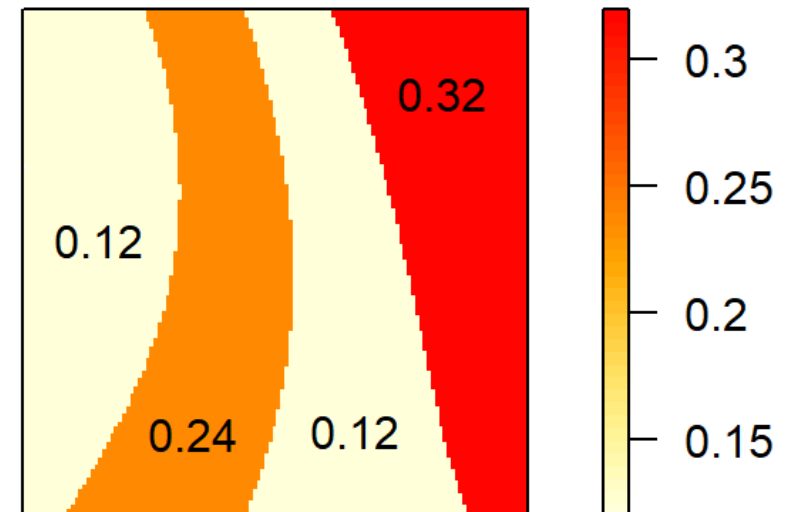
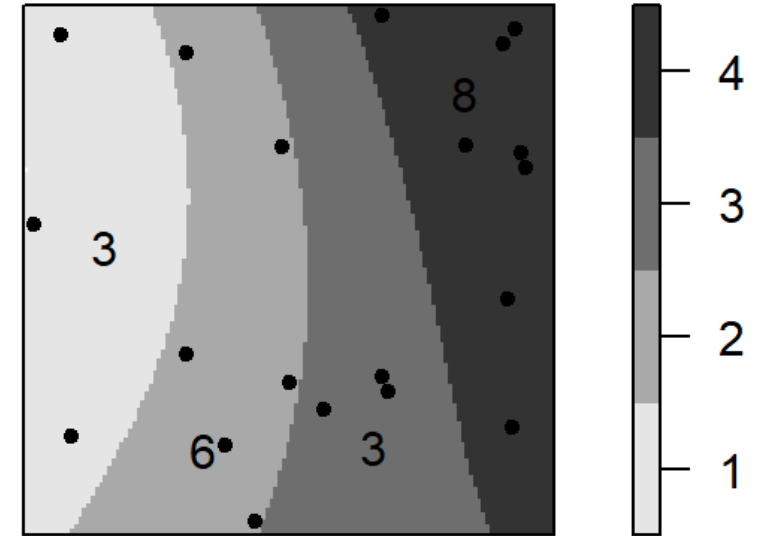
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# Key Techniques

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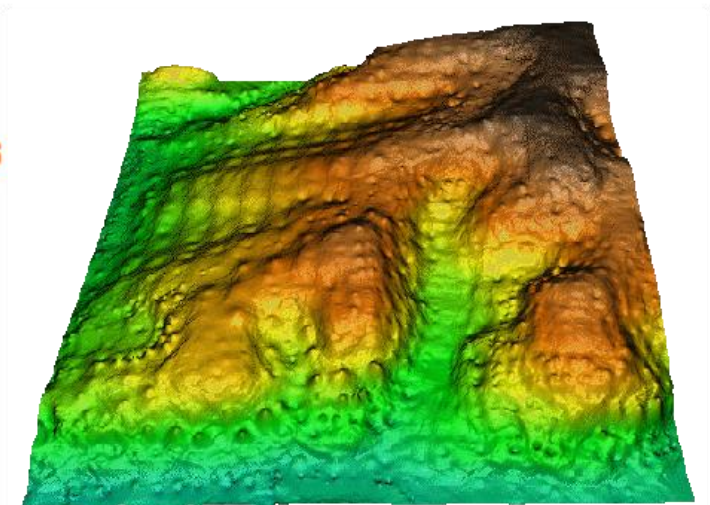
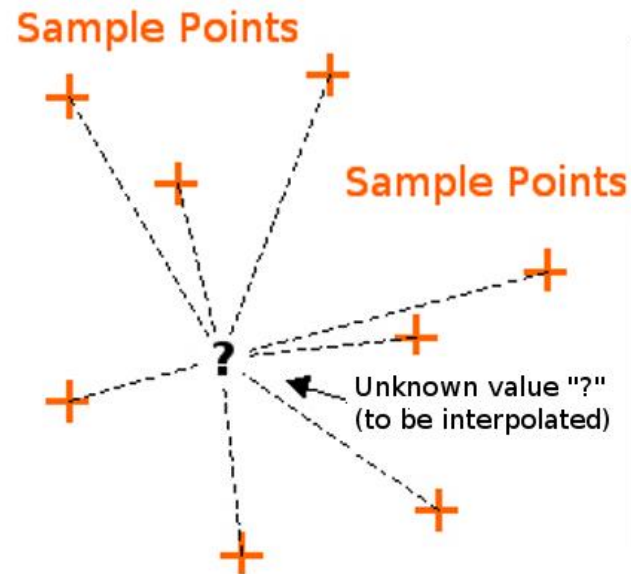
# Point Pattern Analysis

- Point Pattern: set of events/points in a study region
  - Pattern is realized by an underlying process
  - Useful for points with only categorical descriptors: e.g. trees
- First-order properties: variation over a surface
  - Density, intensity (points/area)
- Second-order properties: observations' influence on one another
  - Average nearest neighbor
  - K, L, g functions



# Interpolation

- “Filling in the blanks” based on sampled locations
- Deterministic (no randomness)
  - Thiessen polygons
  - Density estimation
  - Inverse distance weighted
  - Splines
- Stochastic (some randomness)
  - Kriging



# Spatial Weights Matrix

- Binary (0/1) representation of how features are related to each other

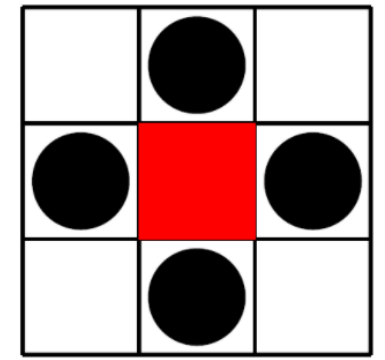
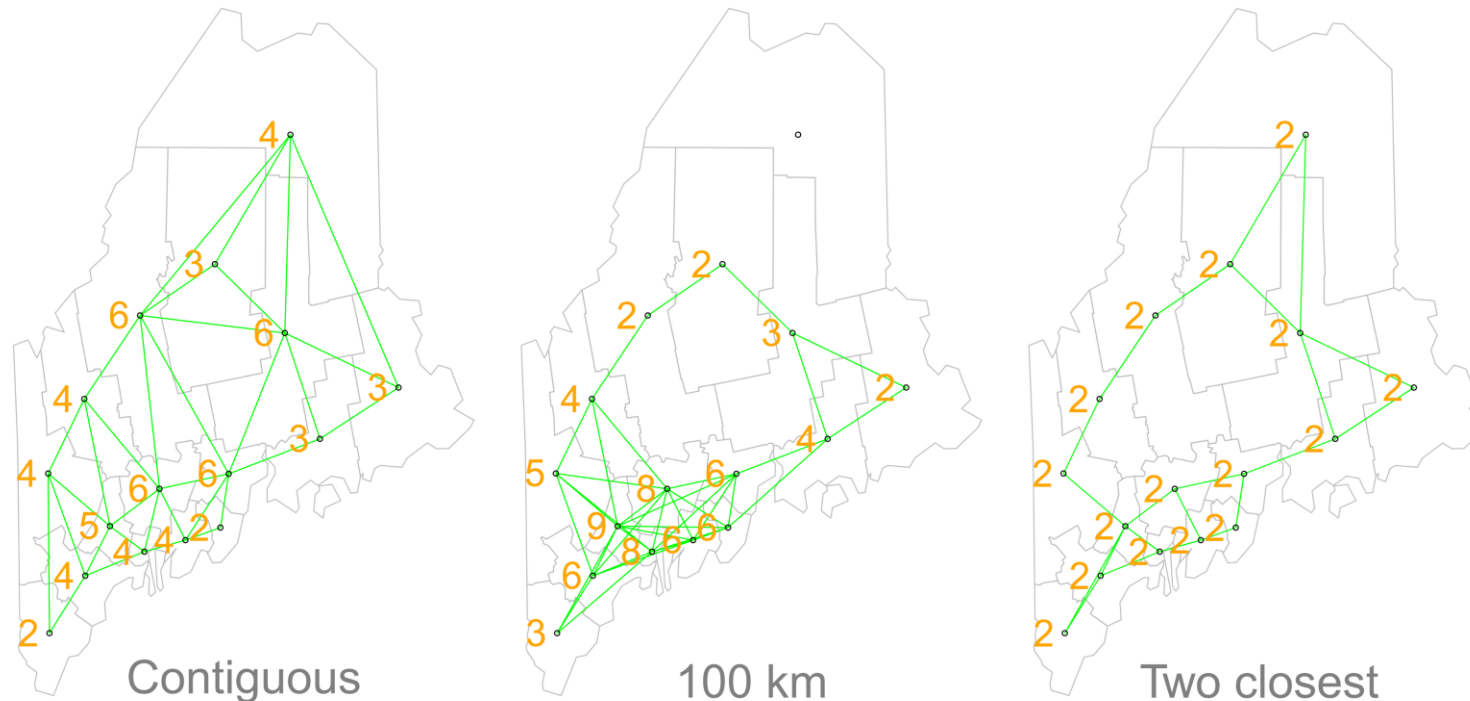


Figure 1: Rook's

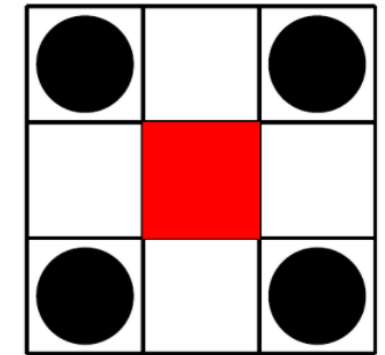


Figure 2: Bishop's

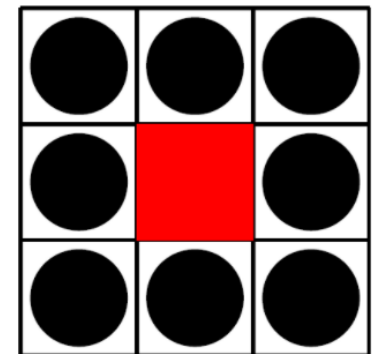


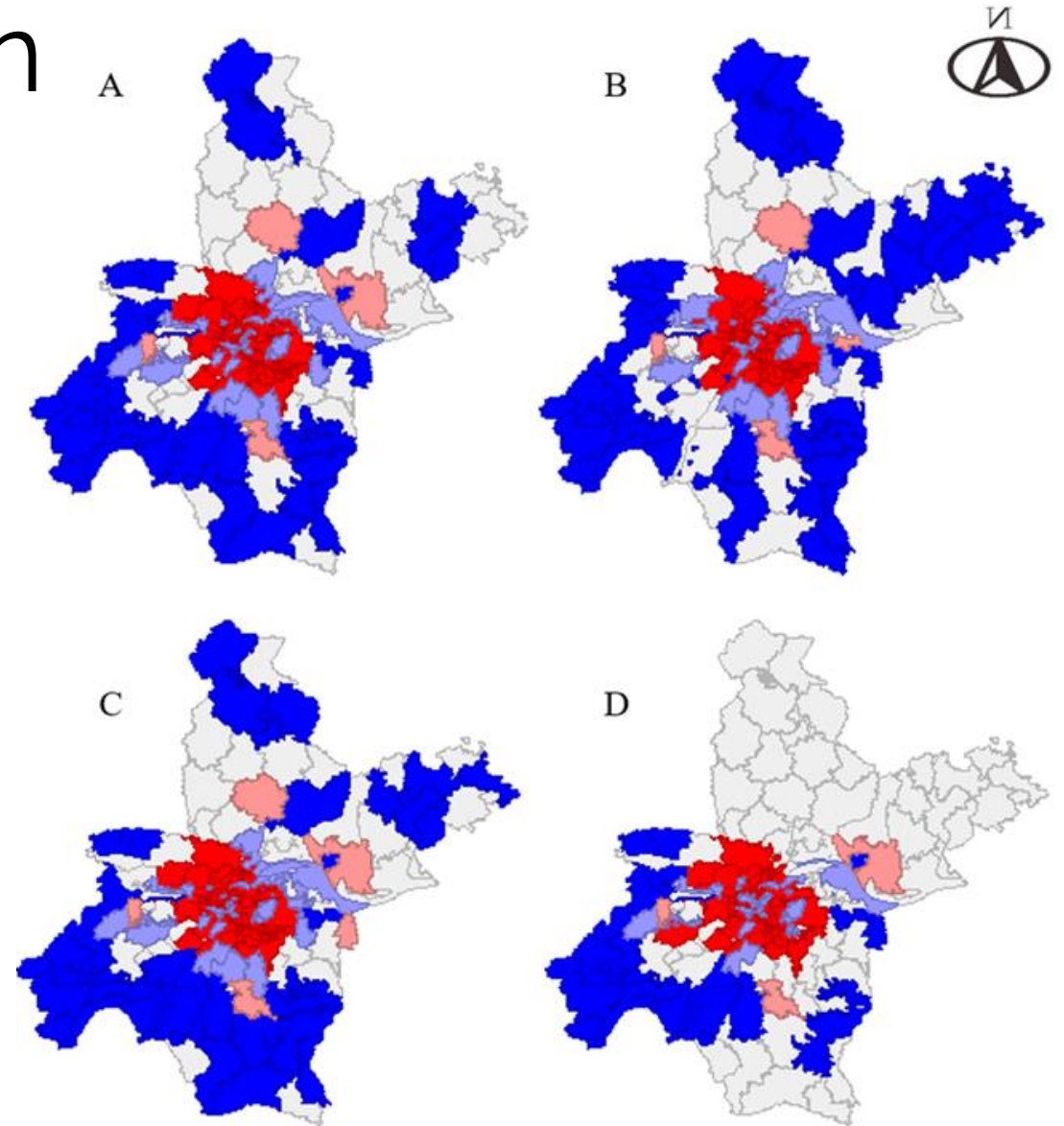
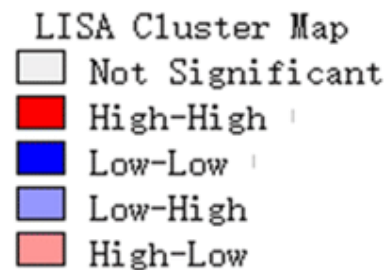
Figure 3: Queen's

Source: Gimond, M. Chapter 13: Spatial Autocorrelation. In: Intro to GIS and Spatial Analysis. (2022).

Canche, M.S.G. Matrices of Influence Lecture Notes. (2020).

# Spatial Autocorrelation

- Degree of correlation between neighboring values
- Metrics: *Moran's I*, *Geary's C*, *Getis-Ord G*
  - **Global**: overall extent of clustering
  - **Local**: clusters and outliers of high and low values
- Local Indicators of Spatial Association (LISA)

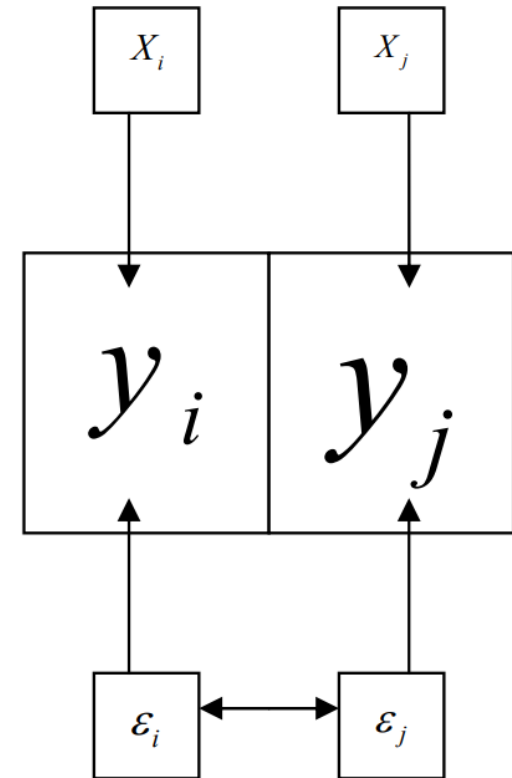
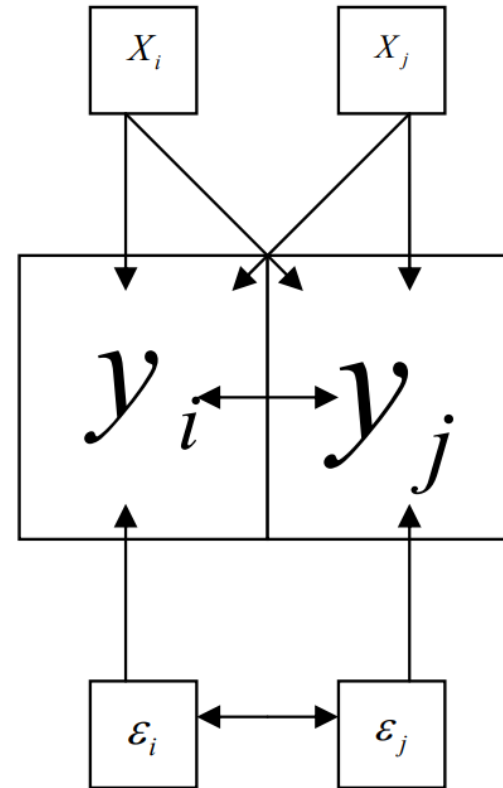


Source: Liu, W., Wang, D., Hua, S., Xie, C., Wang, B., Qiu, W., & Chen, W. (2021).

Spatiotemporal analysis of COVID-19 outbreaks in Wuhan, China. *Scientific reports*, 11(1), 1-9.

# Spatial Regressions

- Spatial data often violates normality assumptions for OLS
- **Spatial Lag:** outcome and predictors are correlated in space
- **Spatial Error:** only error terms across different spatial units are correlated
- **Geographically Weighted Regression:** one regression equation for each feature, accounting for its closest neighbors



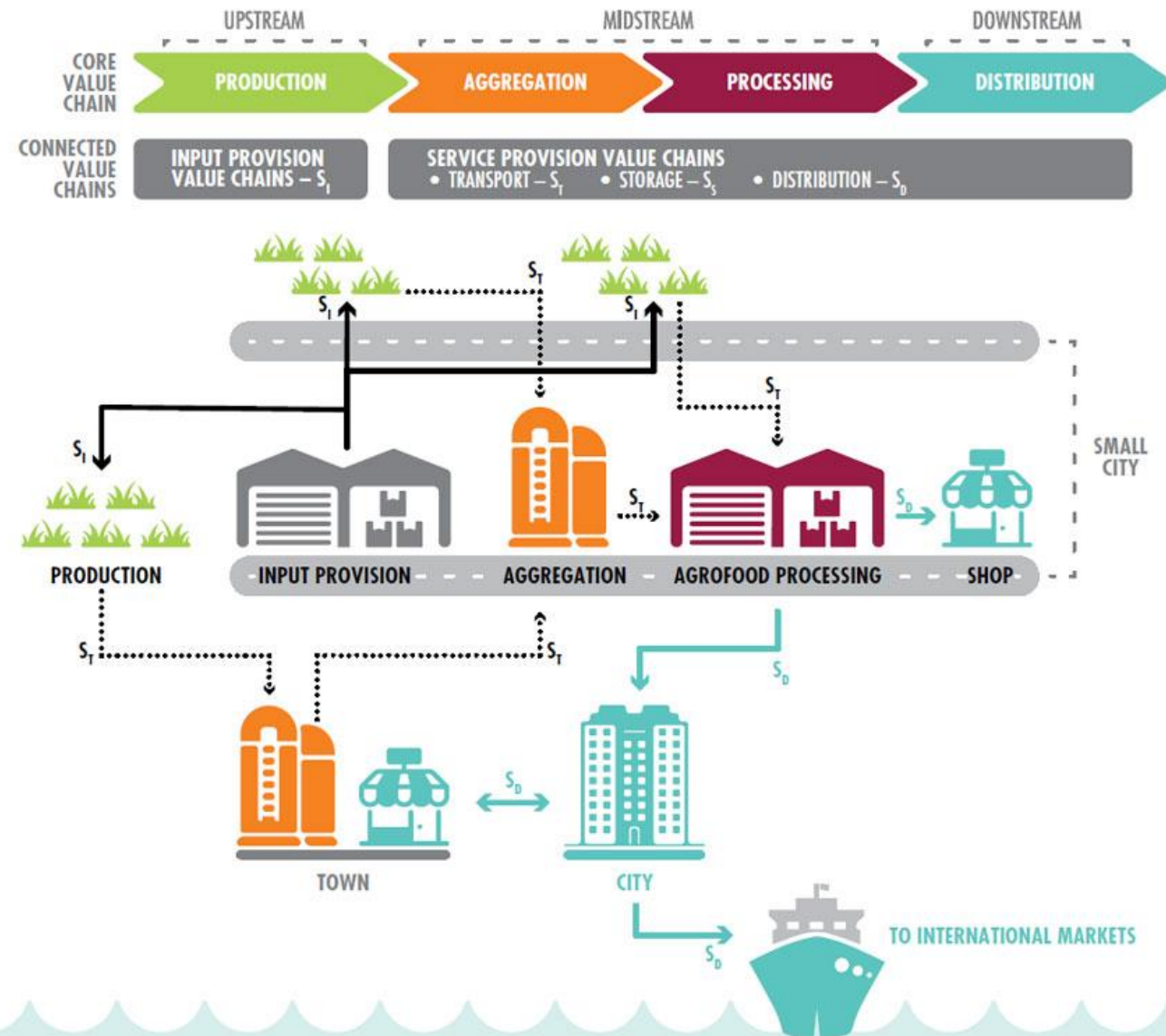


# Research Example

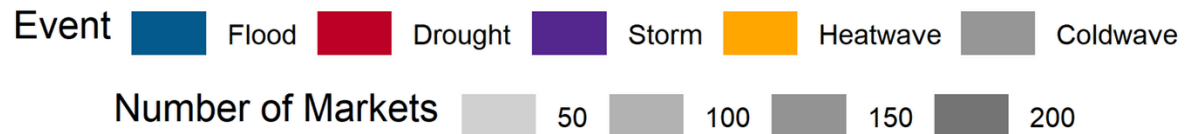
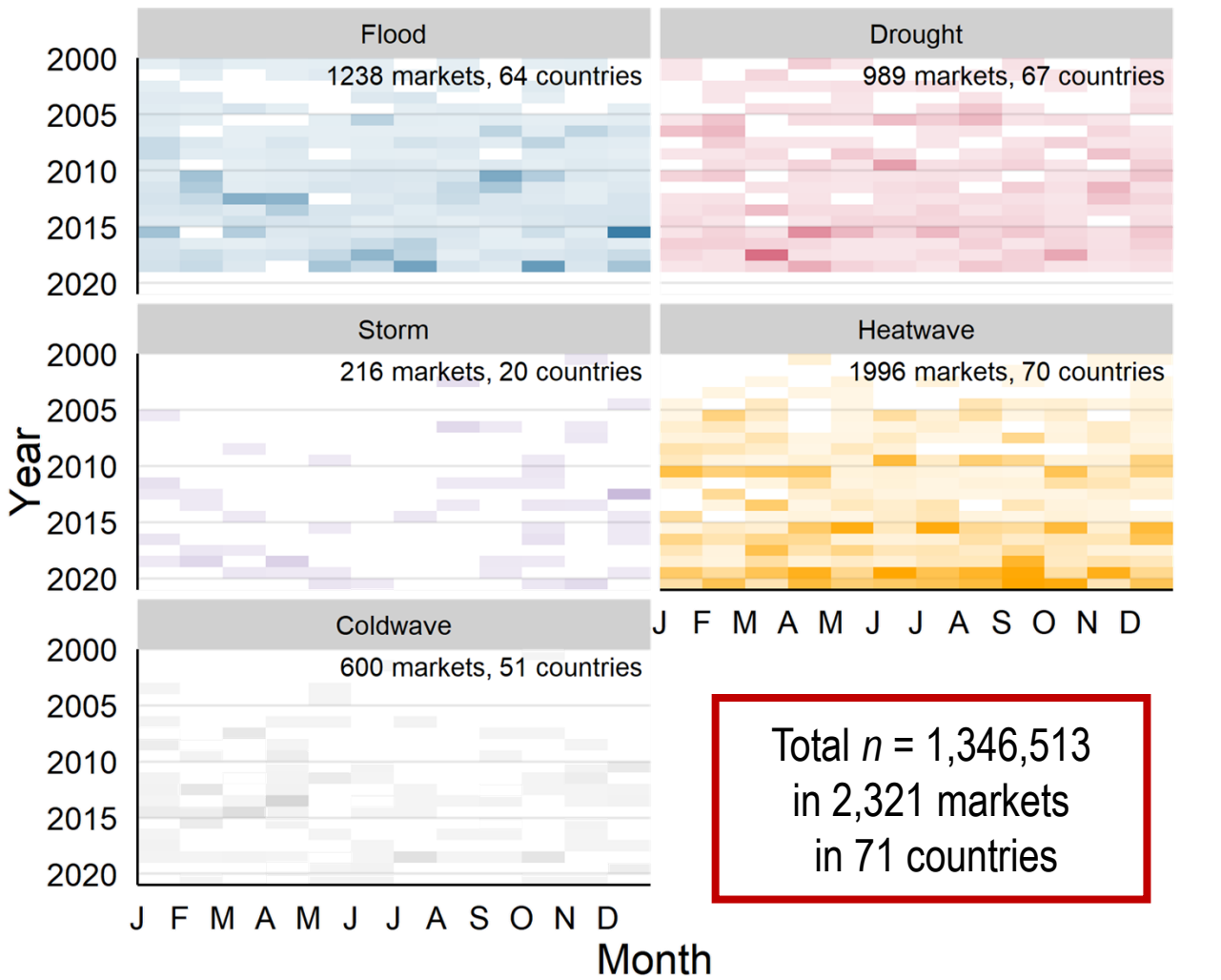
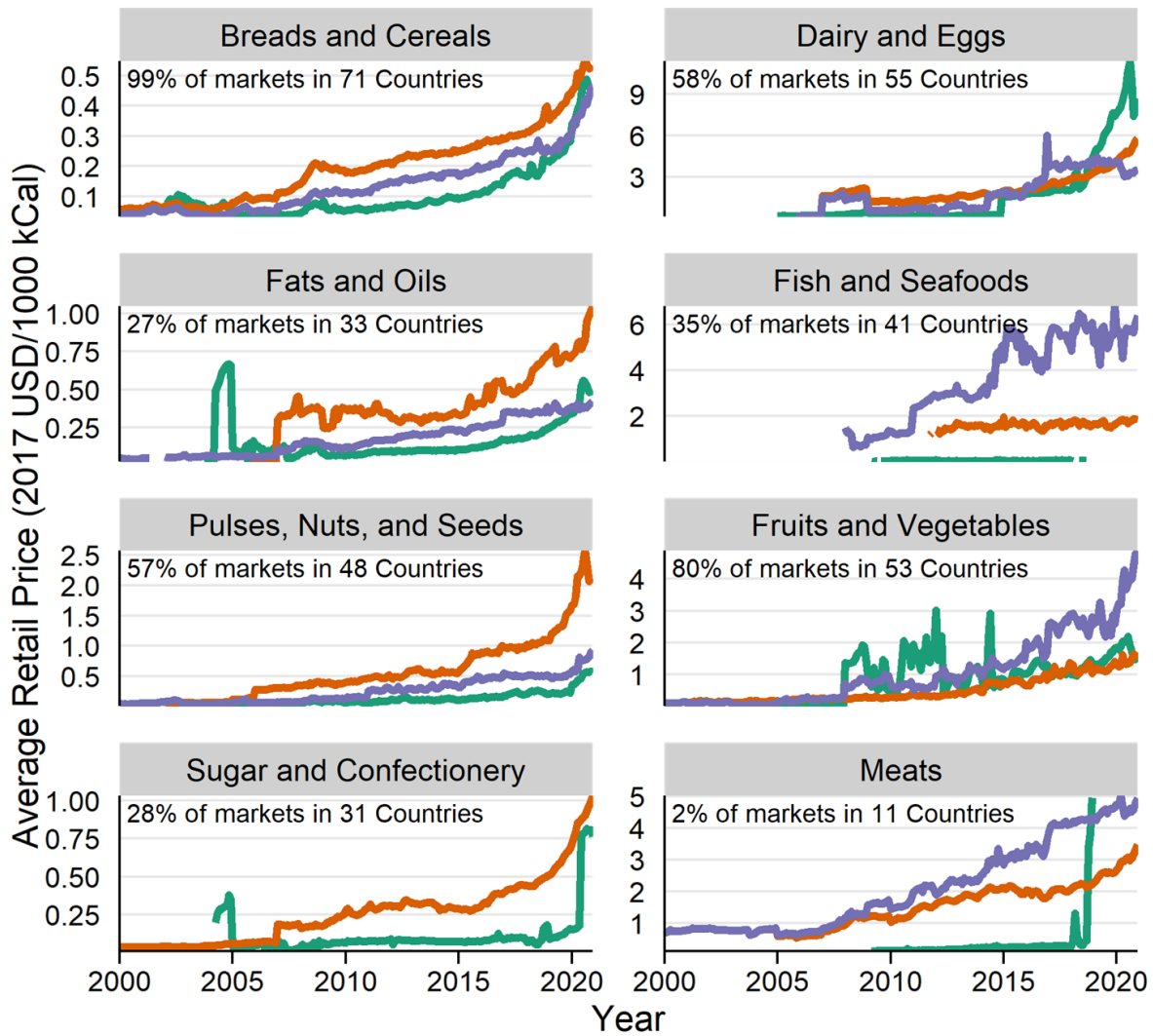
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# Research Questions

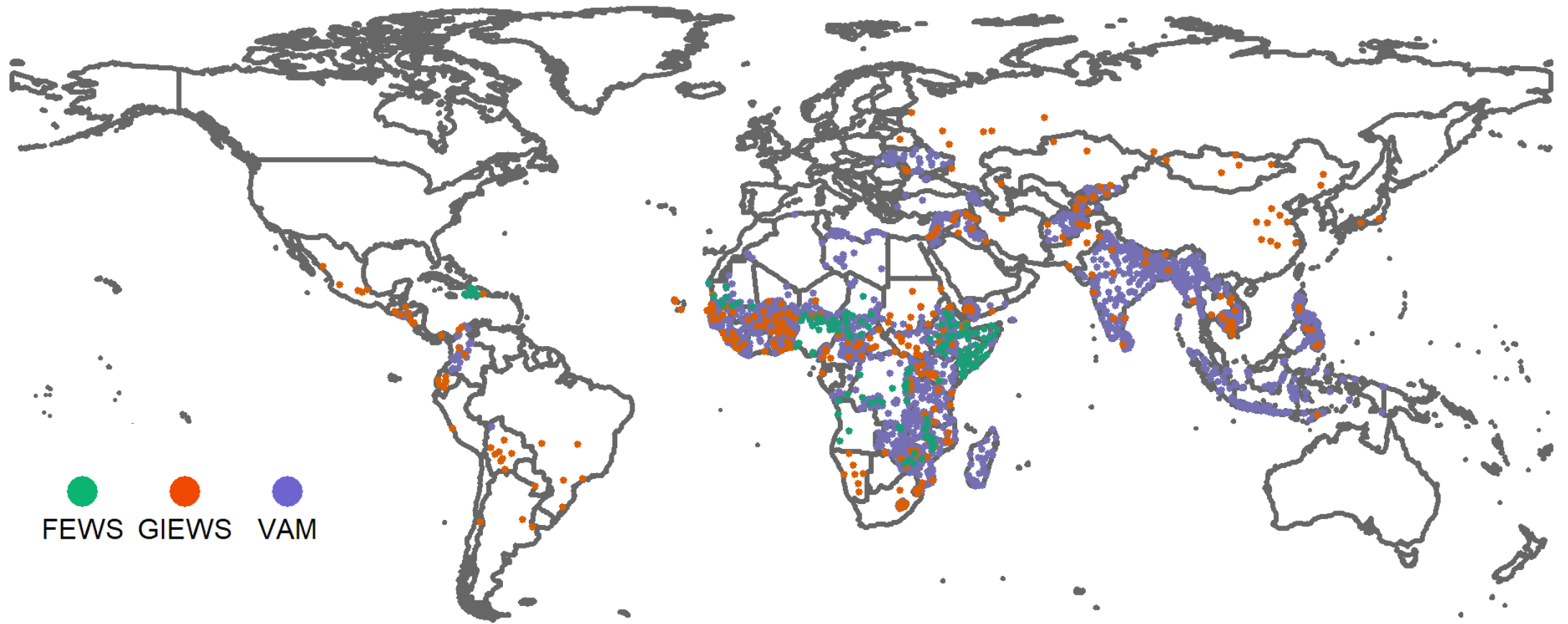
- Big question: do extreme weather events affect retail food prices?
- Spatial questions
  - How integrated are markets?
  - To what degree is price in one market affected by price in:
    - nearest city
    - nearest port city
    - rest of world
  - Does observed effect persist after accounting for spatial lag?





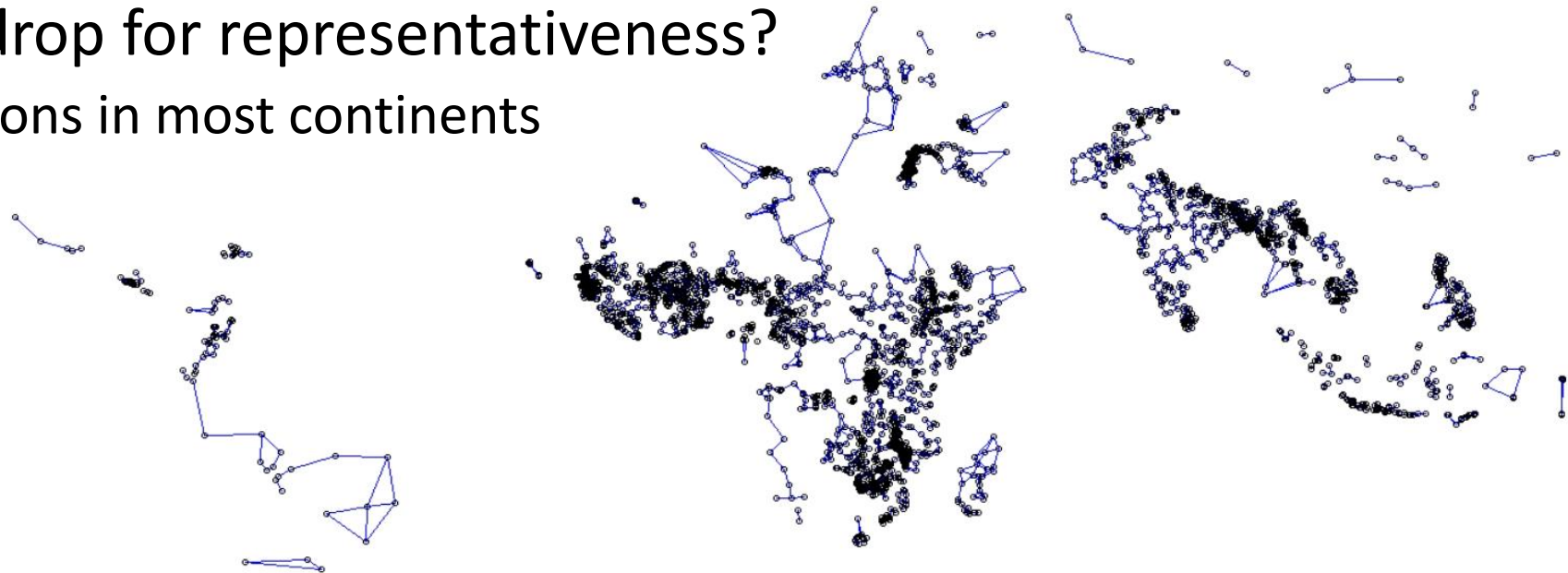


# Markets



# Spatial Decisions

- Is there an underlying process?
  - Yes! Retail price is observed at some markets in this dataset but similar price dynamics hold true in unobserved markets
- How many neighbors to consider, at what distance?
- Which markets to drop for representativeness?
  - Very few observations in most continents



# Next steps

- Wholesale – Retail market pairs to study markups
  - Distance-based thresholds—how far can a food item realistically be transported?
  - Travel time
- Changing combination of market/item/price for each time point
- Temporal lags between price at different markets

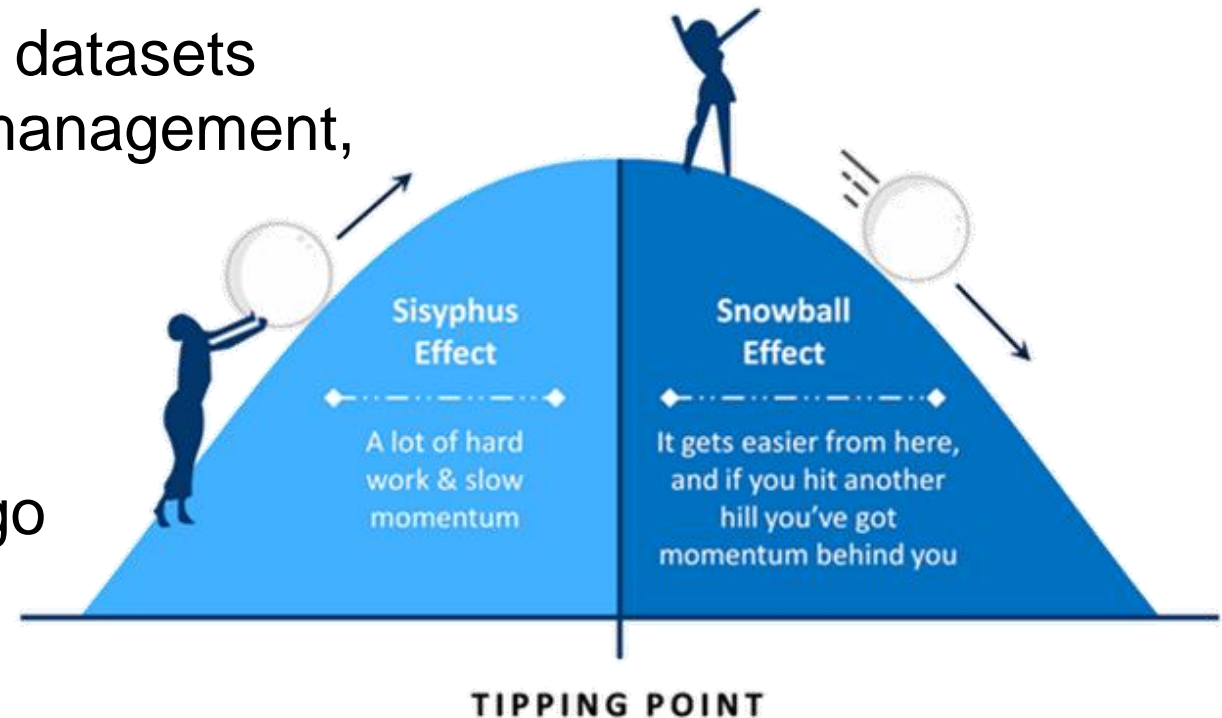


# Lessons and Resources

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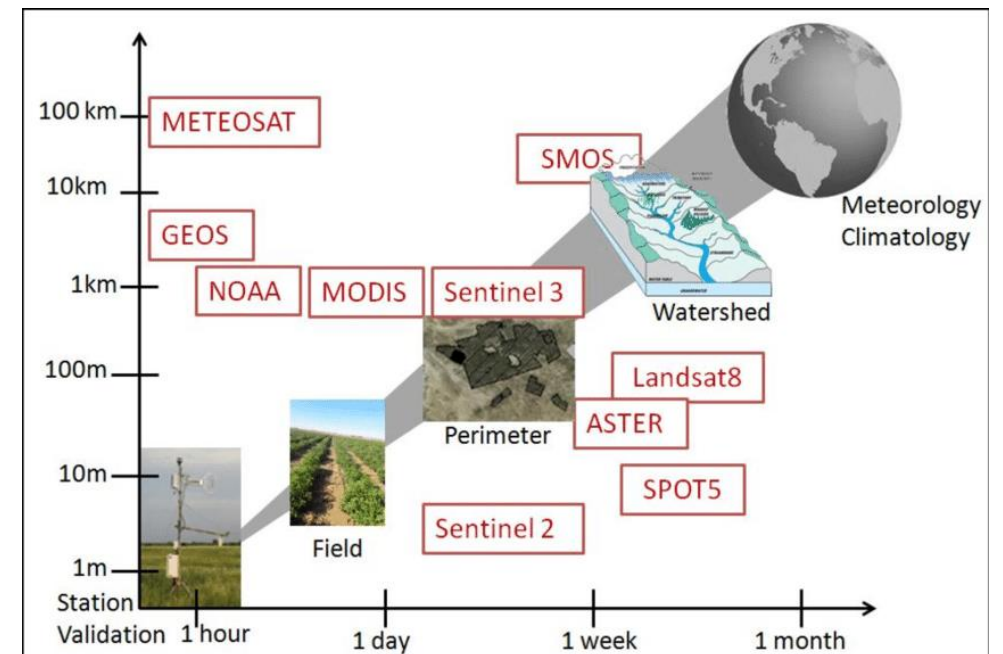
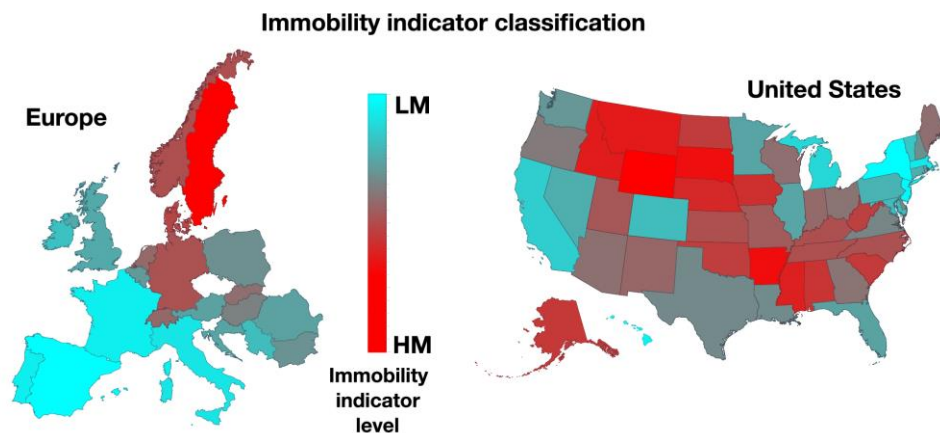
# Lessons

- Identify the universe of predictors early
  - Everything is related, but not everything is relevant/directly measurable
- Research techniques
  - Explore and bookmark interesting datasets
  - Related skills: data mining, data management, independent learning
  - Learn by doing, snowball effect!
- Logistics
  - Workflow matters!
  - Document data decisions on the go
  - Keep reproducibility in mind



# New Data & New Demands

- Spatial data is becoming ubiquitous
  - At multiple spatial scales (local, regional, national, global)
  - At high resolution
- User-generated data
  - Auxiliary, unstructured, individual



Sources: Malbêteau, Y. (2016). *Suivi des ressources en eau par une approche combinant la télédétection multi-capteur et la modélisation phénoménologique* (Doctoral dissertation, Université Paul Sabatier-Toulouse III).

Cot, C., Cacciapaglia, G., & Sannino, F. (2021). *Mining Google and Apple mobility data: temporal anatomy for COVID-19 social distancing*. *Scientific reports*, 11(1), 1-8.

# Resources

- R
  - [RSpatial](#)
  - [Spatial Regression Analysis in R: A Workbook](#)
  - [Geocomputation with R](#)
- Python
  - [Geographic Data Science with Python](#)
  - [PySAL](#)
  - [Spatial Modelling for Data Scientists](#)
- Stata
  - [Reference Manual: Spatial Autoregressive Models](#)
  - Lecture Notes: [Raschky](#), [Drukker](#), [Pisati](#)



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Thank you!  
Questions?

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