Final Project

Temperature Volatility Aggregation Sandbox

Presentation by: Cole Davis

Brown University

April 22, 2024

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 - Specifically, how should we aggregate temperature volatility?



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 - Should we compute temperature volatility at a weather station level and aggregate that?
 - How do we account for correlation between weather stations?
 - At what stage should we demean the data?
 - Can we introduce a custom weighting scheme in the aggregation process?
 - How do we measure temperature volatility?



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 - 1 Create a general region and split it into states
 - 2 Create weather stations in each state
 - 3 Generate temperature data over a given period of time

- Region size is based on the contiguous region formed by Iowa, Illinois, Nebraska, and Minnesota
 - Can be modified

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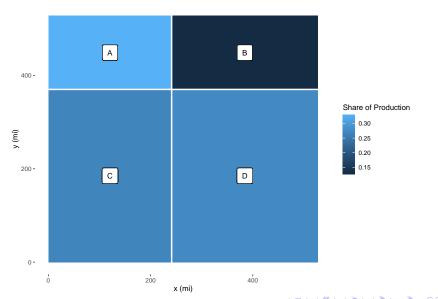
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- (x, y) coordinates of state corners and shares are exported to a csv

Region Map



Weather Stations

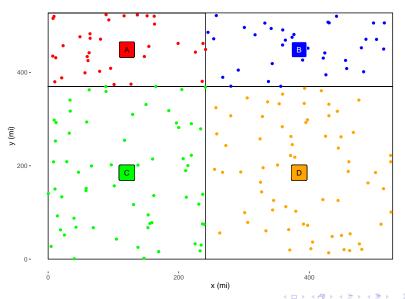
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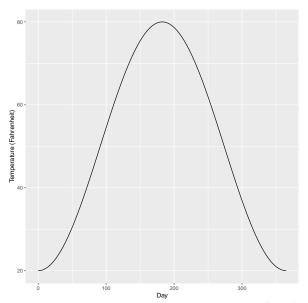
Weather Stations

- 200 weather stations are placed randomly throughout the region
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- Weather stations are then sorted into their appropriate state
- (x, y) coordinates of weather stations in a state are exported to a csv

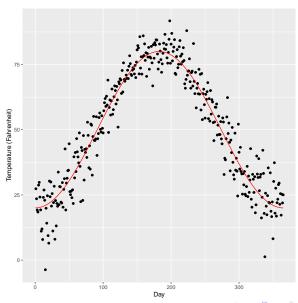
Weather Station Map



Temperature Data - Base



Temperature Data - Noisy



Temperature Data - Generation

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Temperature Data - Generation

- 4 Random points are chosen to be centroids of a cluster
- Centroids are given a year of randomly generated temperature data
- At time t, a non centroid weather station x has temperature distributed as

$$\mathcal{N}\left(x^*(t), 2.5d\left(\frac{x}{\|x\|}, \frac{x^*}{\|x^*\|}\right)\right)$$

in the winter and

$$\mathcal{N}\left(x^*(t), 2d\left(\frac{x}{\|x\|}, \frac{x^*}{\|x^*\|}\right)\right)$$

in the summer



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Thank you!

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