

Extreme Values on Spatial Fields

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Background

- 5th year graduate student
- 2nd year with GSP
- Anticipated graduation date: Fall 2005

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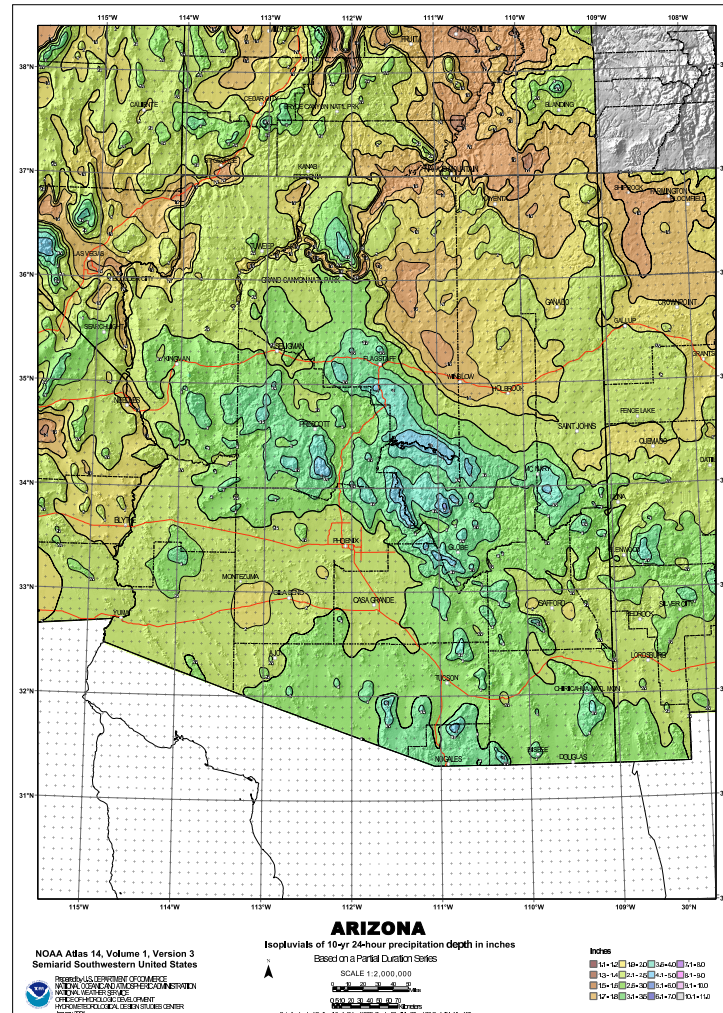
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Projects

- Extreme value model for Lichenometry
- Spatial dependence estimation and prediction for annual maxima
- Colorado Front Range Precipitation

Colorado Precipitation Project

Goal: Create a map of precipitation return levels for Colorado's Front Range.



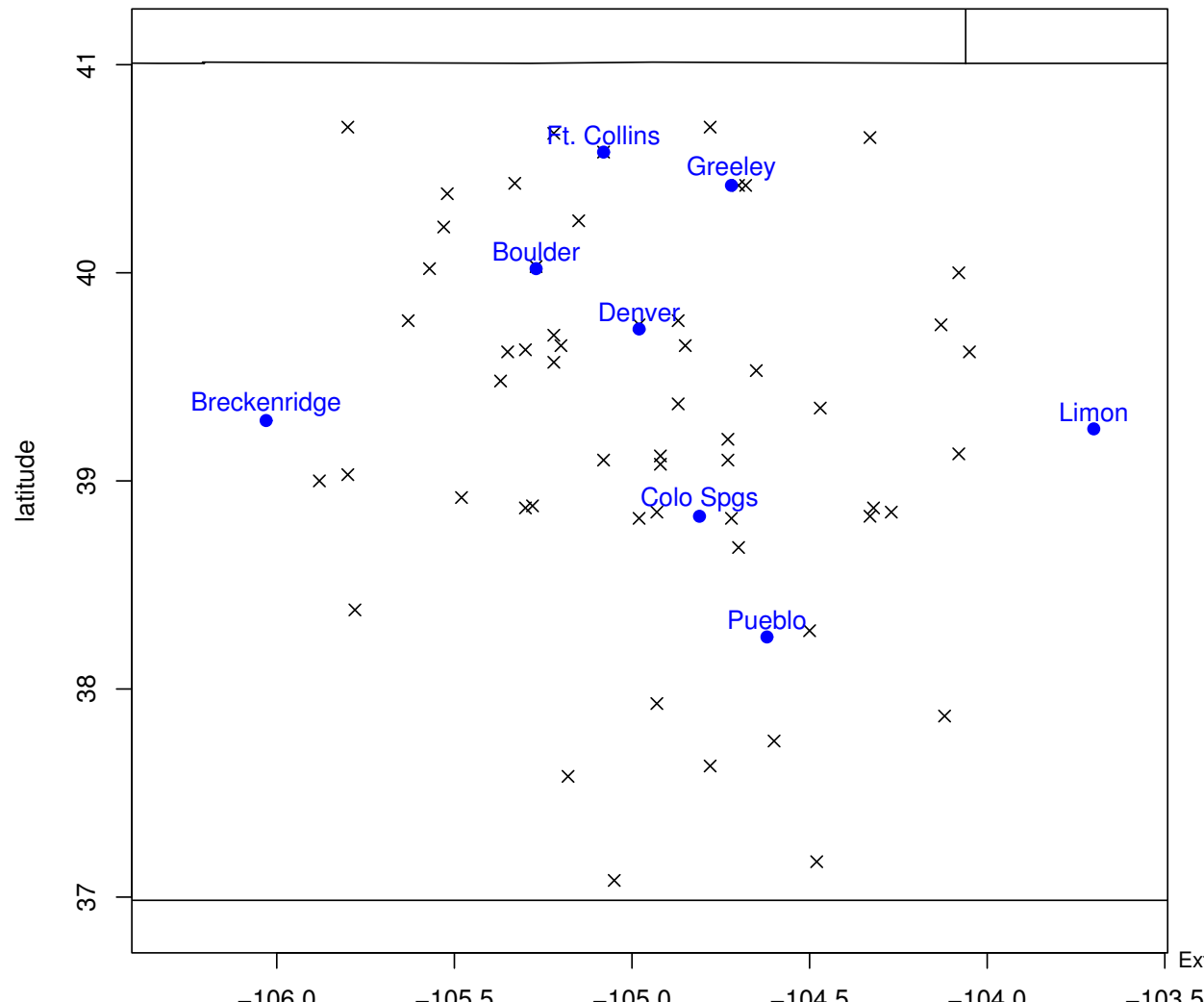
Colorado Precipitation Project

Goal: Create a map of precipitation return levels for Colorado's Front Range.

- project originated from ISSE, interested in flooding
- NWS has done maps for Southwest US (AZ, NM, UT, NV) and mid-Atlantic (OH, PA, MD, NJ, NC, WV)
- plans to do entire US (contingent on funding)
- handles spatial dependence and prediction differently
- present NWS with our method and results

Front Range Data

Data: hourly precipitation from 56 weather stations, 12-60 years of data, Apr1 - Oct 31.



Univariate Extreme Values

GEV: Used to model block (annual) maxima

$$G(z; \mu, \sigma, \xi) = \exp \left[- \left[1 + \xi \left(\frac{z - \mu}{\sigma} \right) \right]^{-\frac{1}{\xi}} \right]$$

- wasteful of data
- used by NWS to produce their maps

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GPD: Models exceedences above a threshold

$$P\{X - u < y | X > u\} \approx 1 - \left(1 + \frac{\xi y}{\tilde{\sigma}} \right)^{-1/\xi}$$

$$\text{where } \tilde{\sigma} = \sigma + \xi(u - \mu)$$

- Less wasteful of data
- Must choose threshold u
- How to go spatial?

Bayesian Hierarchical Model

Let $X_{i,j}$ be observation j from station i .

$$X_{i,j} \sim GPD(\tilde{\sigma}_i, \xi_i)$$

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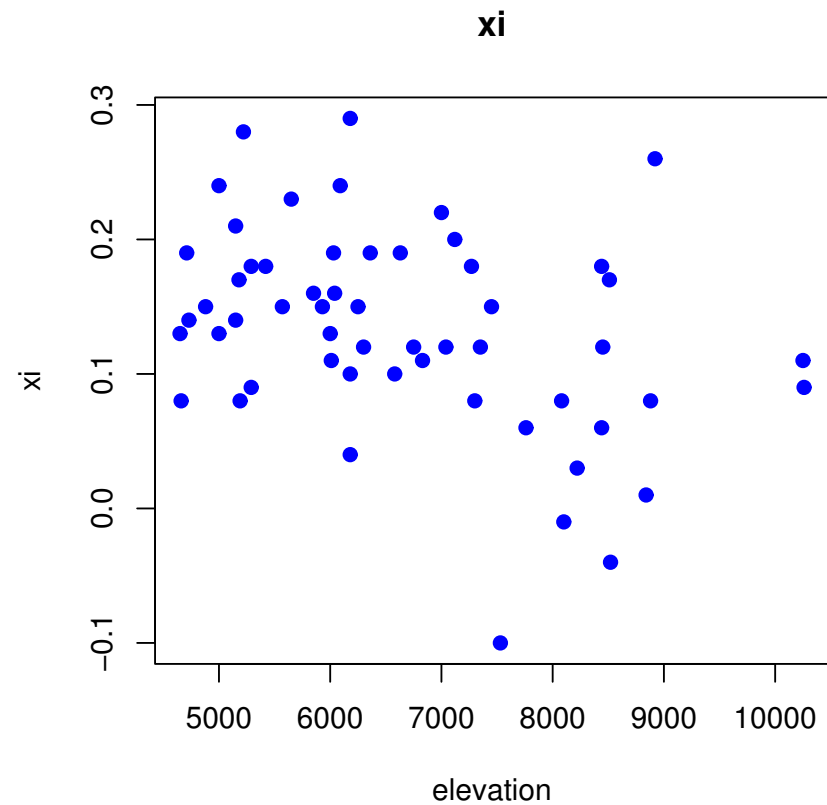
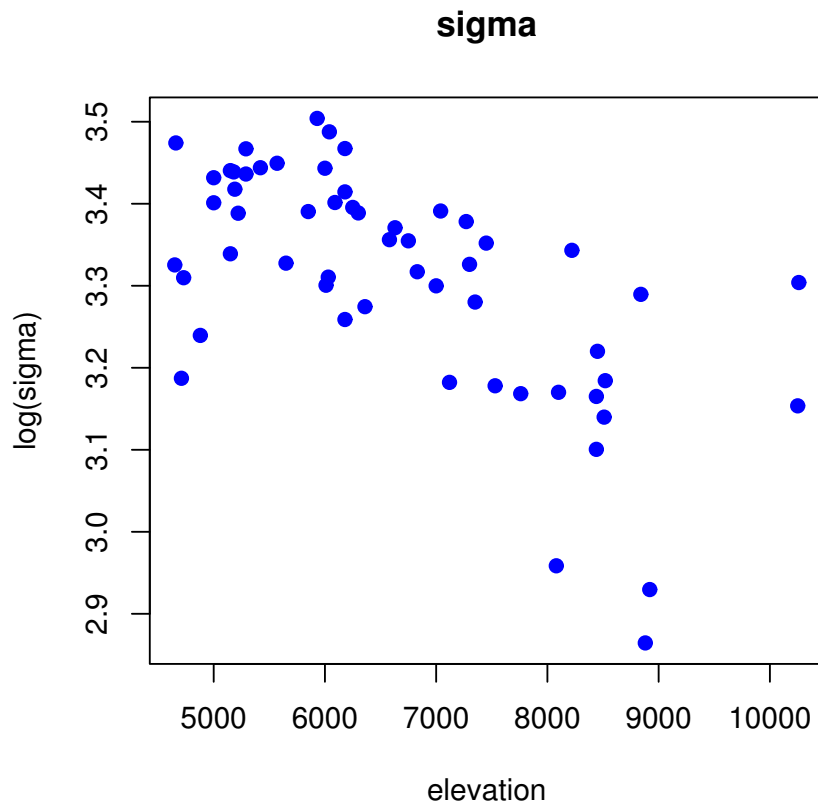
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Plan: Use existing spatial techniques on the parameters, then convert to the desired return levels.

Covariates

To do spatial prediction, any covariate must be available for every location in the region.



First Model

$$X_{i,j} \sim GPD(\tilde{\sigma}_i, \xi_i)$$

$$\log \tilde{\sigma} \sim MVN(\underline{a}, B)$$

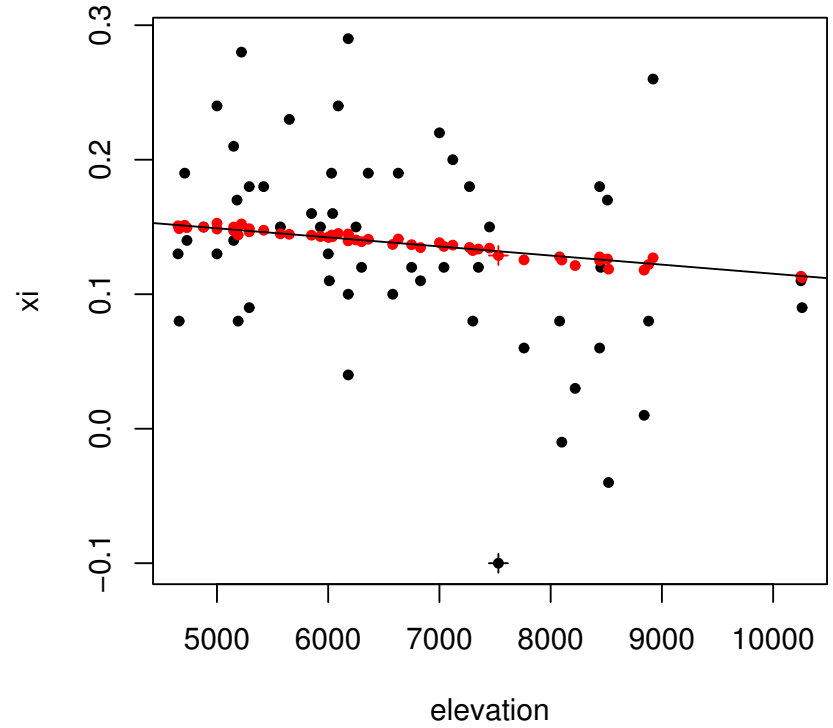
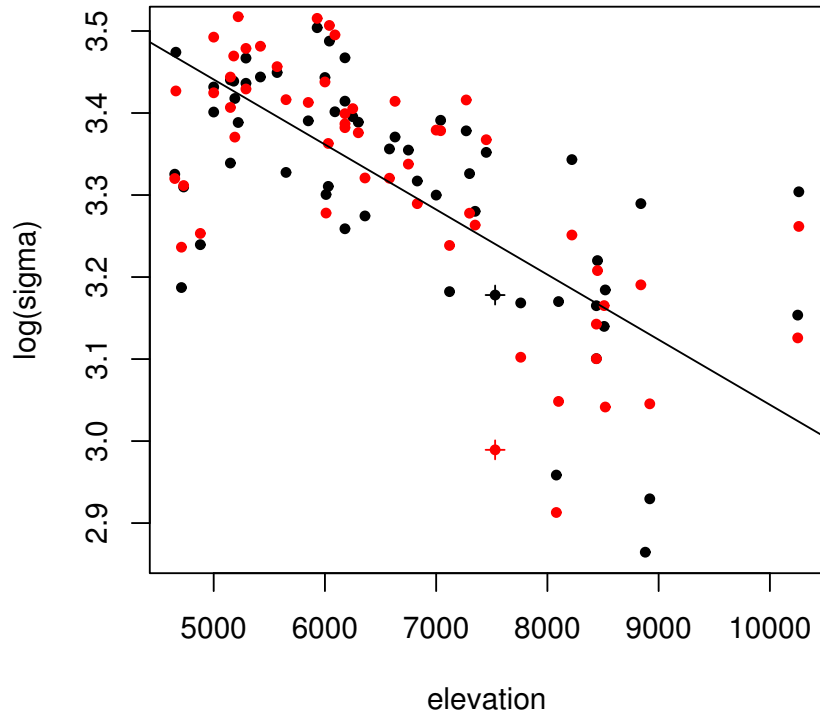
$$\underline{\xi} \sim MVN(\underline{c}, D)$$

$$\underline{a} = \alpha_{\sigma 1} + \alpha_{\sigma 2}(\text{elevation}) \quad B = \beta_{\sigma 1} \exp[-\beta_{\sigma 2}(\text{distance})]$$

$$\underline{c} = \alpha_{\xi 1} + \alpha_{\xi 2}(\text{elevation}) \quad D = \beta_{\xi 1} \exp[-\beta_{\xi 2}(\text{distance})]$$

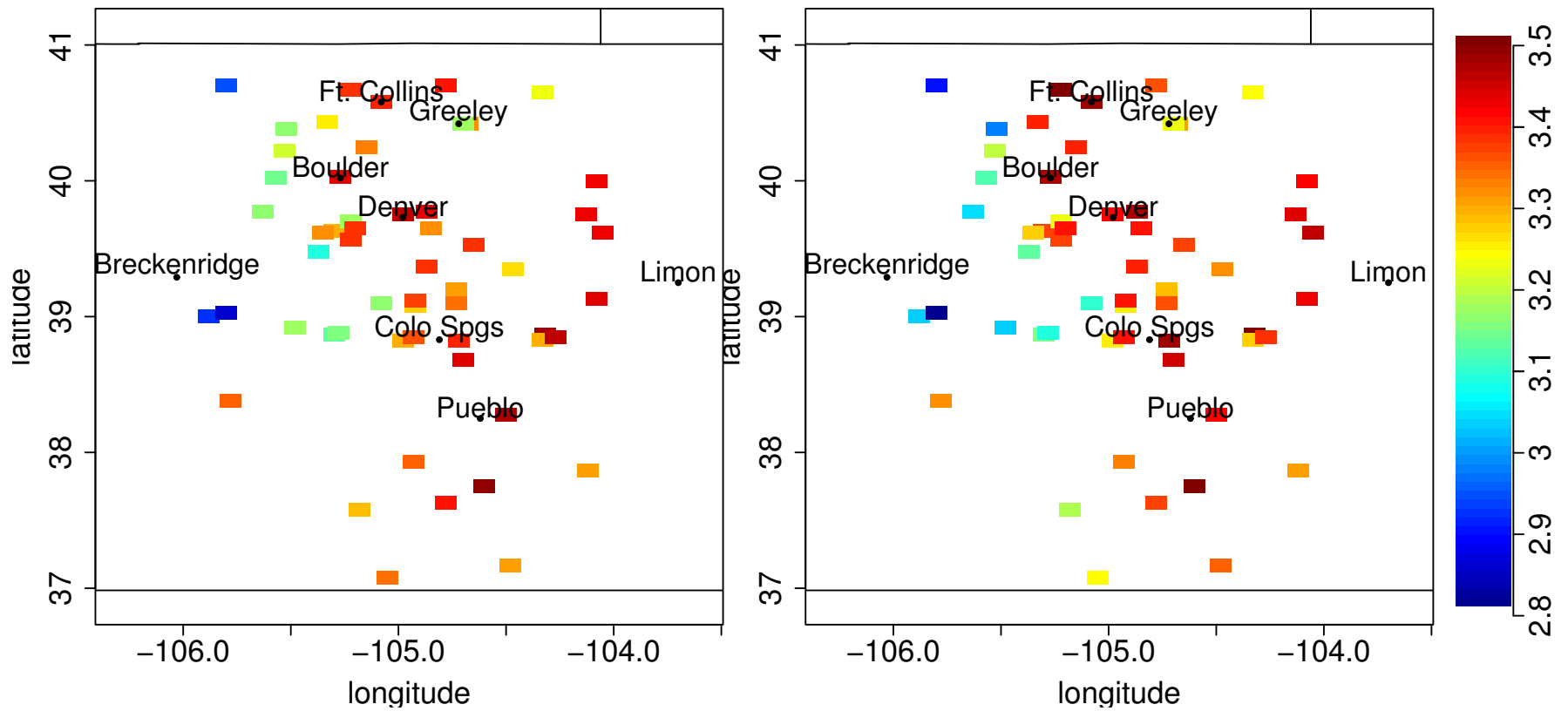
$$\alpha_{..} \sim Unif(-\infty, \infty) \quad \beta_{..} \sim Unif(-\infty, \infty)$$

Model Results

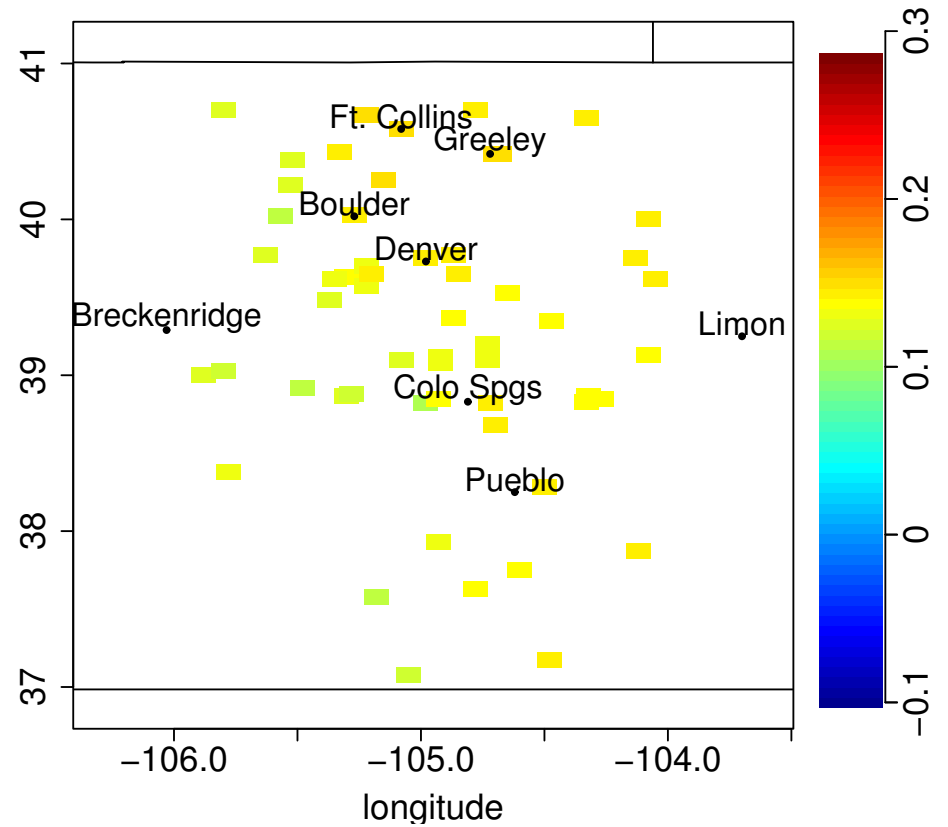
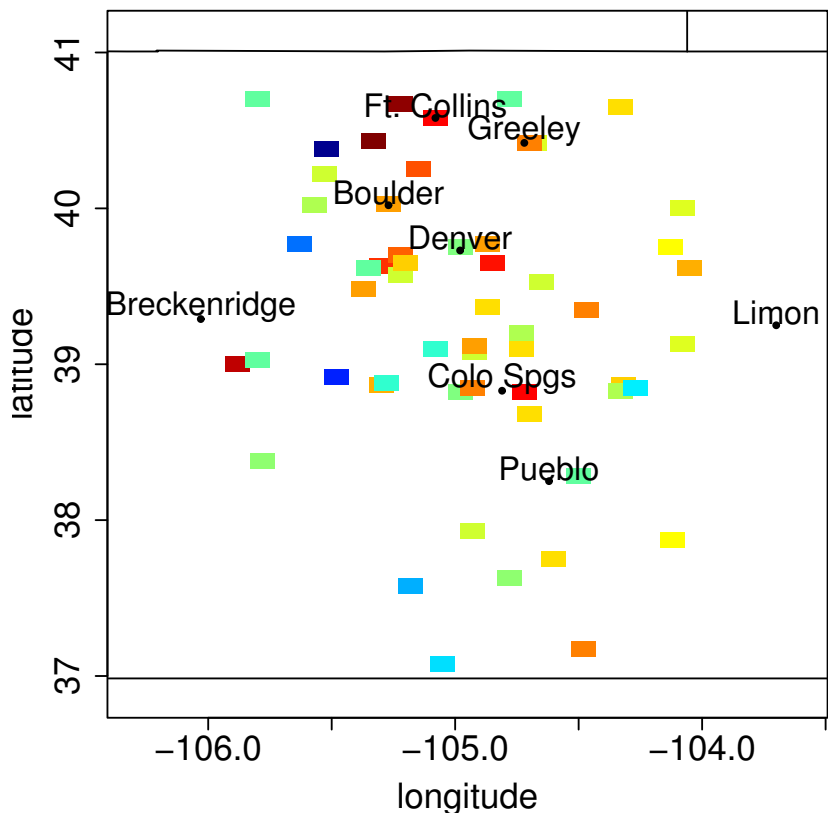


Model Results

$$\log(\sigma)$$

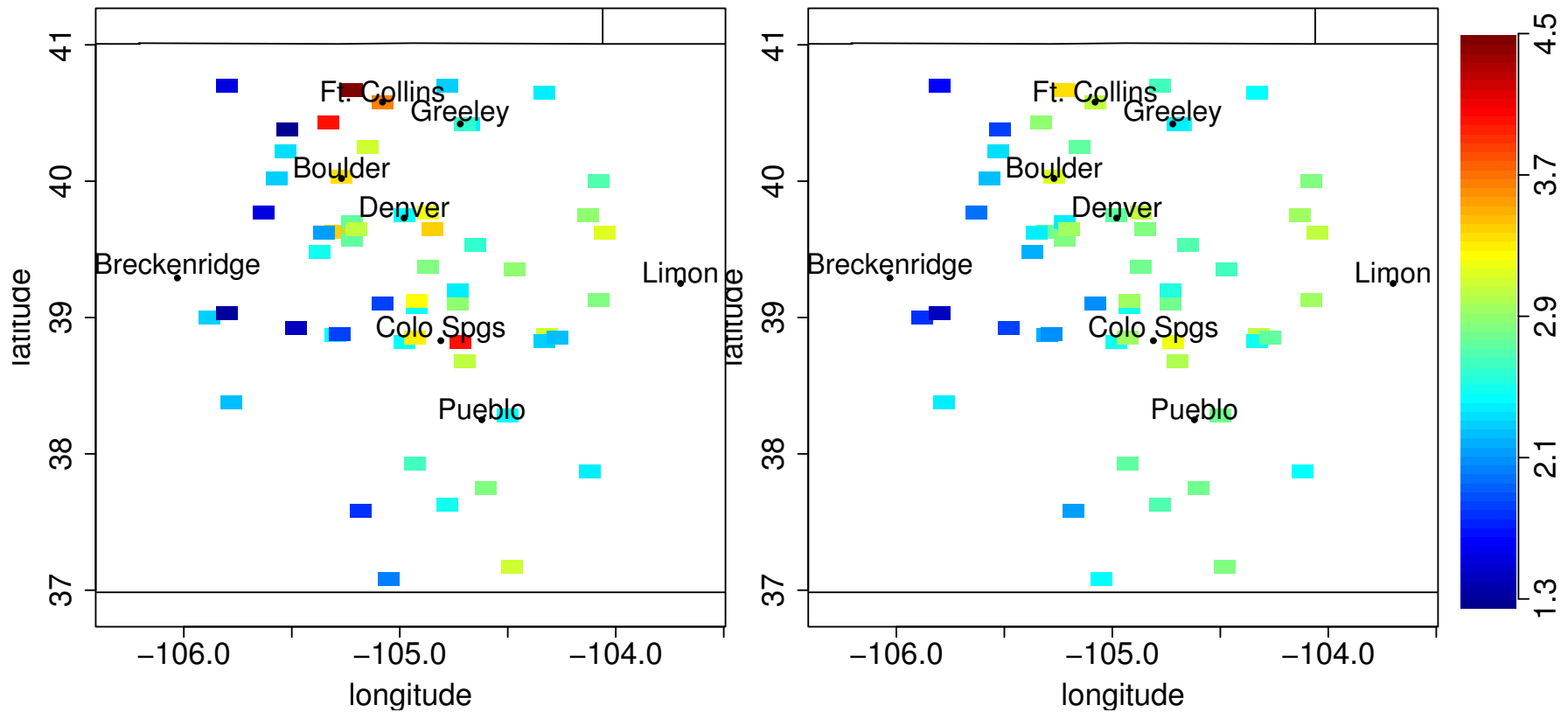


Model Results



Model Results

20-year Return Levels



Future Work

- Extend spatially
- Search for covariates
- Speed up MCMC method
- Examine covariance structure
- Test other models
- Model comparison (DIC)
- ???