Statistical Modeling and Inference for Climate Applications

Jon Hobbs

Jet Propulsion Laboratory, California Institute of Technology Jonathan.M.Hobbs@jpl.nasa.gov

1. Bayesian Inference

• Hierarchical Model

$$\begin{aligned} \text{Data} &= Y(\mathbf{s}_i, t) \\ \text{Process} &= \alpha(\mathbf{s}_i, t) \\ \text{Parameters} &= \boldsymbol{\theta} \end{aligned}$$

$$[Data, Process, Parameters] = [Data|Process, Parameters] \\ \times [Process|Parameters] \\ \times [Parameters]$$

• Posterior Distribution

$$[\boldsymbol{\theta}|\mathbf{Y}] \propto \int [\mathbf{Y}|\boldsymbol{\alpha}, \boldsymbol{\theta}] [\boldsymbol{\alpha}|\boldsymbol{\theta}] [\boldsymbol{\theta}] d\boldsymbol{\alpha}$$

2. Spatial Models

Model the joint distribution for the spatial field \mathbf{Y}_t

- Covariance model Model $Cov(Y(\mathbf{s}_i,t),Y(\mathbf{s}_j,t))$ as a function of spatial distance
- Spatial random effects Model $Y(\mathbf{s}_i, t)$ as a linear combination of a low-dimensional random vector and spatial basis functions.
- Conditionally-specified model Specify conditional distributions given neighboring values

$$Y(\mathbf{s}_i,t)|\mathbf{y}_t(N_i)$$

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