Spatio-temporal statistics (MATH4341)

Michaelmas term, 2023

Homework 3: Geostatistics (Block Kriging)

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Exercise 1. (*) Suppose a large volume V is partitioned into n smaller units v of equal size. Show that the dispersion variance $\sigma^2(u|V) = \frac{1}{n} \sum_{j=1}^n \sigma_E^2(v_j, V)$ can be written in term of variogram integrals

$$\bar{\gamma}(v, V) = \frac{1}{|v||V|} \int_{s \in v} \int_{s \in V} \gamma(s - s') \, \mathrm{d}s \mathrm{d}s'$$

as

$$\sigma^{2}(v|V) = \bar{\gamma}(V,V) - \bar{\gamma}(v,v)$$

Exercise 2. (*) Consider a statistical model which is a stochastic process $(Z_s)_{s\in\mathbb{R}}$ (so s has dimension 1), where $Z(\cdot) \sim \operatorname{GP}(\mu(\cdot), c(\cdot, \cdot))$ with mean function $\mu(s) = 1$ and covariance function $c(s,t) = (s-t)^2$. Assume there is available a dataset $\{(Z_i, s_i)\}_{i=1}^n$ where $Z_i = Z(s_i)$ and $s_i \in \mathbb{R}$ are point sites.

- 1. Compute the length |v| of the block $v = [a, b] \subset \mathbb{R}$.
- 2. Compute the block mean $\mu(v)$ for some block $v = [a, b] \subset \mathbb{R}$ and point $s \in \mathbb{R}$.
- 3. Compute the block covariance function c(v,s) for some block $v=[a,b]\subset\mathbb{R}$.
- 4. Compute the block covariance function c(v, v') for some blocks $v = [a, b] \subset \mathbb{R}$ and $v' = [a', b'] \subset \mathbb{R}$.
- 5. Denote $Z = (Z_1, ..., Z_n)^{\top}$, and $S = \{s_1, ..., s_n\}$. Let $v = [a, b] \subset \mathbb{R}$ and $v' = [a', b'] \subset \mathbb{R}$ be two intervals. Compute the joint distribution of $(Z(v), Z(v'), Z)^{\top}$ as a function of $c(\cdot, \cdot)$, S, v, v', Z, and $\mu(\cdot)$. What is the name of the distribution and what are the parameter functions defining it?
- 6. (Bayesian Kriging) Compute the predictive stochastic process [Z(v)|Z] at blocks $v = [a, b] \subset \mathbb{R}$ with |v| > 0.

Hint: Let $x_1 \in \mathbb{R}^{d_1}$, and $x_2 \in \mathbb{R}^{d_2}$. If

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \sim \mathcal{N}_{d_1 + d_2} \left(\begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix}, \begin{bmatrix} \Sigma_1 & \Sigma_{21}^\top \\ \Sigma_{21} & \Sigma_2 \end{bmatrix} \right)$$

then it is

$$x_2|x_1 \sim N_{d_2}(\mu_{2|1}, \Sigma_{2|1})$$

where

$$\mu_{2|1} = \mu_2 + \Sigma_{21} \Sigma_1^{-1} (x_1 - \mu_1) \text{ and } \Sigma_{2|1} = \Sigma_2 - \Sigma_{21} \Sigma_1^{-1} \Sigma_{21}^{\top}$$