Homework 3: Geostatistics (Change of support)

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Exercise 1. (\star) Assume we wish to estimate the average value in a domain V

$$Z_V = \frac{1}{|V|} \int_V Z(s) \, \mathrm{d}s$$

with the average of n sample points $\{s_i; i = 1, ..., n\}$.

$$\hat{Z} = \frac{1}{n} \sum_{i=1}^{n} Z(s_i)$$

Show that the estimation variance (or else extension variance)

$$\operatorname{Var}\left(\hat{Z} - Z_{V}\right) = -\frac{1}{n^{2}} \sum_{i=1}^{n} \sum_{j=1}^{n} \gamma\left(s_{i} - s_{j}\right) + \frac{1}{n |V|} \sum_{i=1}^{n} \int_{V} \gamma\left(s_{i} - x\right) dx - \frac{1}{|V|^{2}} \int_{x \in B} \int_{y \in B} \gamma\left(x - y\right) dx dy$$

Hint:: Consider as known that

$$Cov(Z(t) - Z(s), Z(v) - Z(u)) = \gamma(t - u) + \gamma(s - v) - \gamma(s - u) - \gamma(t - v)$$

Solution. It is

$$\begin{aligned} \operatorname{Var}\left(\hat{Z} - Z_{V}\right) &= \operatorname{Cov}\left(\hat{Z} - Z_{V}, \hat{Z} - Z_{V}\right) \\ &= \frac{1}{n^{2}} \sum_{i=1}^{n} \sum_{j=1}^{n} \operatorname{Cov}\left(Z_{i} - Z_{V}, Z_{j} - Z_{V}\right) \\ &= \frac{1}{n^{2}} \frac{1}{|V|^{2}} \sum_{i=1}^{n} \sum_{j=1}^{n} \int_{x \in V} \int_{y \in V} \operatorname{Cov}\left(Z_{i} - Z\left(x\right), Z_{j} - Z\left(y\right)\right) dxdy \\ &= \frac{1}{n^{2}} \frac{1}{|V|^{2}} \sum_{i=1}^{n} \sum_{j=1}^{n} \int_{x \in V} \int_{y \in V} \left[\gamma\left(s_{i} - y\right) + \gamma\left(x - s_{j}\right) - \gamma\left(x - y\right) - \gamma\left(s_{i} - s_{j}\right)\right] dxdy \\ &= -\frac{1}{n^{2}} \sum_{i=1}^{n} \sum_{j=1}^{n} \gamma\left(s_{i} - s_{j}\right) + \frac{1}{n |V|} \sum_{i=1}^{n} \int_{V} \gamma\left(s_{i} - x\right) dx - \frac{1}{|V|^{2}} \int_{x \in B} \int_{y \in B} \gamma\left(x - y\right) dxdy \end{aligned}$$

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 $c(s,t) = \exp\left(-(s-t)^2\right)$ for any $s \in \mathbb{R}$ and $t \in \mathbb{R}$. 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 $v=[a,b]\subset\mathbb{R}$ and point $s\in\mathbb{R}$.
- (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-1:: Let $x_1 \in \mathbb{R}^{d_1}$, and $x_2 \in \mathbb{R}^{d_2}$. If

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

then it is

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

where

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

Hint-2: You can use that $\int \operatorname{erf}(x) dx = x \operatorname{erf}(x) + \frac{\exp(-x^2)}{\sqrt{\pi}} + \operatorname{const}$, when $\operatorname{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x \exp(-t^2) dt$

Solution.

- (1) It is |v| = b a
 - (a) It is

$$\mu(v) = \mu([a, b]) = \frac{1}{|v|} \int_{v} \mu(s) ds = \frac{1}{|v|} \int_{v} 1 ds = \frac{1}{|v|} |v| = 1$$

(b) It is

$$\begin{split} c\left(v,s\right) &= \frac{1}{|v|} \int_{v} c\left(t,s\right) \mathrm{d}t = \frac{1}{b-a} \int_{a}^{b} \exp\left(-\left(t-s\right)^{2}\right) \mathrm{d}t \\ &= \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \int_{a}^{b} \frac{2}{\sqrt{\pi}} \exp\left(-\left(t-s\right)^{2}\right) \mathrm{d}t = \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \int_{a-s}^{b-s} \frac{2}{\sqrt{\pi}} \exp\left(-\xi^{2}\right) \mathrm{d}\xi \\ &= \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \int_{0}^{b-s} \frac{2}{\sqrt{\pi}} \exp\left(-\xi^{2}\right) \mathrm{d}\xi - \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \int_{0}^{a-s} \frac{2}{\sqrt{\pi}} \exp\left(-\xi^{2}\right) \mathrm{d}\xi \\ &= \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \mathrm{erf}\left(b-s\right) - \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \mathrm{erf}\left(a-s\right) \end{split}$$

(c) It is

$$\begin{split} c\left(v,v'\right) &= \frac{1}{|v'|} \frac{1}{|v|} \int_{a'}^{b'} \int_{a}^{b} c\left(t,s\right) \mathrm{d}t \mathrm{d}s = \frac{1}{|v'|} \frac{1}{|v|} \int_{a'}^{b'} \left[\int_{a}^{b} c\left(t,s\right) \mathrm{d}t \right] \mathrm{d}s = \frac{1}{b'-a'} \int_{a'}^{b'} c\left(v,s\right) \mathrm{d}s \\ &= \frac{1}{b'-a'} \int_{a'}^{b'} \left(\frac{1}{b-a} \frac{\sqrt{\pi}}{2} \mathrm{erf}\left(b-s\right) - \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \mathrm{erf}\left(a-s\right) \right) \mathrm{d}s \\ &= \frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \int_{a'}^{b'} \mathrm{erf}\left(b-s\right) \mathrm{d}s - \frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \int_{a'}^{b'} \mathrm{erf}\left(a-s\right) \mathrm{d}s \\ &= \frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \left(-1\right) \int_{b-a'}^{b-b'} \mathrm{erf}\left(\xi\right) \mathrm{d}\xi - \frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \left(-1\right) \int_{a-a'}^{a-b'} \mathrm{erf}\left(\xi\right) \mathrm{d}\xi \\ &= \frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \left(-1\right) \left[\xi \mathrm{erf}\left(\xi\right) + \frac{\exp\left(-\xi^2\right)}{\sqrt{\pi}} \right]_{b-a'}^{b-b'} - \frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \left(-1\right) \left[\xi \mathrm{erf}\left(\xi\right) + \frac{\exp\left(-\xi^2\right)}{\sqrt{\pi}} \right]_{a-a'}^{a-b'} \\ &= -\frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \left(b-b'\right) \mathrm{erf}\left(b-b'\right) - \frac{1}{b'-a'} \frac{1}{b-a} \frac{1}{2} \left(b-b'\right) \exp\left(-\left(b-b'\right)^2\right) \\ &+ \frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \left(b-a'\right) \mathrm{erf}\left(b-a'\right) + \frac{1}{b'-a'} \frac{1}{b-a} \frac{1}{2} \left(b-a'\right) \exp\left(-\left(b-a'\right)^2\right) \\ &+ \frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \left(a-b'\right) \mathrm{erf}\left(a-b'\right) + \frac{1}{b'-a'} \frac{1}{b-a} \frac{1}{2} \left(a-b'\right) \exp\left(-\left(a-b'\right)^2\right) \\ &- \frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \left(a-a'\right) \mathrm{erf}\left(a-a'\right) - \frac{1}{b'-a'} \frac{1}{b-a} \frac{1}{2} \left(a-a'\right) \exp\left(-\left(a-a'\right)^2\right) \\ &+ \frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \left(a-a'\right) \mathrm{erf}\left(a-a'\right) - \frac{1}{b'-a'} \frac{1}{b-a} \frac{1}{2} \left(a-a'\right) \exp\left(-\left(a-a'\right)^2\right) \\ \end{array}$$

(d) It is

$$\begin{bmatrix} Z\left(v\right) \\ Z\left(v'\right) \\ Z \end{bmatrix} \sim \mathcal{N} \left(\begin{bmatrix} \mu\left(v\right) \\ \mu\left(v'\right) \\ \mu\left(S\right) \end{bmatrix}, \begin{bmatrix} c\left(v,v\right) & c\left(v,v'\right) & c\left(v,S\right) \\ c\left(v',v\right) & c\left(v',v'\right) & c\left(v',S\right) \\ c\left(S,v\right) & c\left(S,v'\right) & c\left(S,S\right) \end{bmatrix} \right)$$

(e) Taking a better look at part 1, I can see

$$\begin{bmatrix}
\begin{bmatrix} Z(v_1) \\ Z(v_2) \end{bmatrix} \\
\begin{bmatrix} Z(v_1) \\ Z(v_2) \end{bmatrix}
\end{bmatrix} \sim \mathcal{N} \left(\begin{bmatrix} \mu(v) \\ \mu(v_2) \\ \mu(S) \end{bmatrix} \right), \begin{bmatrix} c(v_1, v_1) & c(v_1, v_2) \\ c(v_2, v_1) & c(v_2, v_2) \end{bmatrix} & \begin{bmatrix} c(v_1, S) \\ c(v_2, S) \end{bmatrix} \\
\begin{bmatrix} c(S, v_1) & c(S, v_2) \end{bmatrix} & \begin{bmatrix} c(S, S) \end{bmatrix} \end{bmatrix}$$

From the hint I can see, I can see that

$$\begin{bmatrix} Z(v_1) \\ Z(v_2) \end{bmatrix} | Z \sim \mathcal{N}\left(\mu^{\dagger}, C^{\dagger}\right)$$

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m with}$

$$C^{\dagger} = \begin{bmatrix} C_{11}^{\dagger} & C_{12}^{\dagger} \\ C_{21}^{\dagger} & C_{22}^{\dagger} \end{bmatrix} = \begin{bmatrix} c\left(v_{1}, v_{1}\right) & c\left(v_{1}, v_{2}\right) \\ c\left(v_{2}, v_{1}\right) & c\left(v_{2}, v_{2}\right) \end{bmatrix} - \begin{bmatrix} c\left(v_{1}, S\right) \\ c\left(v_{2}, S\right) \end{bmatrix} \left[c\left(S, S\right) \right]^{-1} \left[c\left(S, v_{1}\right) & c\left(S, v_{2}\right) \right]$$
and
$$\mu^{\dagger} = \mu\left(v_{1}\right) + c\left(v_{1}, S\right) \left[c\left(S, S\right) \right]^{-1} \left(Z - \mu\left(S\right)\right)$$

As this is consistent for any vector of blocks with any size, not only $V = \{v_1, v_2\}$, but also $V = \{v_1, v_2, ..., v_q\}$ then the predictive stochastic process is a Gaussian Process

$$Z(\cdot)|Z \sim \operatorname{GP}(\mu^*(\cdot), c^*(\cdot, \cdot))$$

with mean function at block v

$$\mu^*(v) = \mu(v) + c(v, S) [c(S, S)]^{-1} (Z - \mu(S))$$

by looking at μ^{\dagger} , and with covariance function at any pair of blocks v and v'

$$c^*(v, v') = c(v, v') - c(v, S) [c(S, S)]^{-1} c(S, v')$$

looking at any off-diagonal element of C^{\dagger} e.g. the (1,2) element marked in red.