Spatio-temporal statistics (MATH4341)

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## Homework 3: Geostatistics (Change of support)

Lecturer: Georgios P. Karagiannis

georgios.karagiannis@durham.ac.uk

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

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

Solution.

$$\sigma^{2}(v|V) = \frac{1}{n} \sum_{j=1}^{n} \sigma_{E}^{2}(v_{j}, V)$$

$$= \frac{1}{n} \sum_{j=1}^{n} \left[ 2\bar{\gamma}(v_{j}, V) - \bar{\gamma}(v_{j}, v_{j}) - \bar{\gamma}(V, V) \right]$$

$$= \frac{2}{n} \sum_{j=1}^{n} \bar{\gamma}(v_{j}, V) - \frac{1}{n} \sum_{j=1}^{n} \bar{\gamma}(v_{j}, v_{j}) - \frac{1}{n} \sum_{j=1}^{n} \bar{\gamma}(V, V)$$

$$= \frac{2}{n} \sum_{j=1}^{n} \frac{1}{|v_{j}| |V|} \int_{s \in v_{j}} \int_{s \in V} \gamma(s - s') \, ds ds'$$

$$- \frac{1}{n} \sum_{j=1}^{n} \bar{\gamma}(v_{j}, v_{j}) - \frac{1}{n} \sum_{j=1}^{n} \bar{\gamma}(V, V) \quad \text{(but } |v_{j}| = |v|)$$

$$= 2 \frac{1}{n |v| |V|} \sum_{j=1}^{n} \int_{s \in v_{j}} \int_{s \in V} \gamma(s - s') \, ds ds' - \bar{\gamma}(v, v) - \bar{\gamma}(V, V)$$

$$= 2 \frac{1}{n |v| |V|} \sum_{j=1}^{n} \int_{s \in V_{j}} \int_{s \in V} \gamma(s - s') \, ds ds' - \bar{\gamma}(v, v) - \bar{\gamma}(V, V)$$

$$= 2 \frac{1}{n} \sum_{j=1}^{n} \bar{\gamma}(V, V) - \bar{\gamma}(v, v) - \bar{\gamma}(V, V) = \bar{\gamma}(V, V) - \bar{\gamma}(v, v)$$

**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

 $c(s,t) = \exp\left(-(s-t)^2\right)$  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

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

then it is

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

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
- (2) It is

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

(3) It is

$$c(v,s) = \frac{1}{|v|} \int_{v} c(t,s) dt = \frac{1}{b-a} \int_{a}^{b} \exp\left(-(t-s)^{2}\right) dt$$

$$= \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \int_{a}^{b} \frac{2}{\sqrt{\pi}} \exp\left(-(t-s)^{2}\right) dt = \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \int_{a-s}^{b-s} \frac{2}{\sqrt{\pi}} \exp\left(-\xi^{2}\right) d\xi$$

$$= \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \int_{0}^{b-s} \frac{2}{\sqrt{\pi}} \exp\left(-\xi^{2}\right) d\xi - \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \int_{0}^{a-s} \frac{2}{\sqrt{\pi}} \exp\left(-\xi^{2}\right) d\xi$$

$$= \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \operatorname{erf}(b-s) - \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \operatorname{erf}(a-s)$$

(4) 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} \operatorname{erf}\left(b-s\right) - \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \operatorname{erf}\left(a-s\right) \right) \mathrm{d}s \\ &= \frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \int_{a'}^{b'} \operatorname{erf}\left(b-s\right) \mathrm{d}s - \frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \int_{a'}^{b'} \operatorname{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'} \operatorname{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'} \operatorname{erf}\left(\xi\right) \mathrm{d}\xi \\ &= \frac{1}{b'-a'} \frac{1}{b-a} \frac{\sqrt{\pi}}{2} \left(-1\right) \left[ \xi \operatorname{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 \operatorname{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) \operatorname{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) \operatorname{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) \operatorname{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) \operatorname{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{aligned}$$

(5) It is

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

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

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

From the hint I can see, I can see that

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

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

$$\mu^{\dagger} = \mu(v_1) + c(v_1, S) [c(S, S)]^{-1} (Z - \mu(S))$$

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  $\mu^{\dagger}$ , 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.