

**Homework 1: Point referenced data (building concepts)**

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**Exercise 1.** (★) Let  $Z = (Z(s) : s \in \mathbb{R}^d)$  be an intrinsic random field with  $E(Z(s) - Z(t)) = 0$  and let  $\gamma : \mathbb{R}^d \rightarrow \mathbb{R}$  be its semivariogram.

(1) Let  $a \in \mathbb{R}^n$  be a vector of constants. Consider sites  $\{s_1, \dots, s_n \subseteq \mathbb{R}^d\}$  Show that

$$\text{Var} \left( \sum_{i=1}^n a_i Z(s_i) \right) = \sum_{i=1}^n \sum_{j=1}^n a_i a_j c_I(s_i, s_j)$$

where  $c_I(s, t) = \gamma(s - s_0) + \gamma(t - s_0) - \gamma(s - t)$  at some additional  $s_0 \in \mathbb{R}^d$ .

(2) Show that for all  $n \in \mathbb{N}$ ,  $(a_1, \dots, a_n) \subseteq \mathbb{R}^n$  s.t.  $\sum_{i=1}^n a_i = 0$ , and for all  $(s_1, \dots, s_n) \subseteq S^n$ , it is

$$\sum_{i=1}^n \sum_{j=1}^n a_i a_j \gamma(s_i - s_j) \leq 0$$

**Solution.** Assume origin  $s_0 \in \mathbb{R}^d$  with random  $Z(s_0)$ .

(1) I use  $Z(s_0)$  at some location let's say  $s_0$ . It is

$$\begin{aligned} \text{Var} \left( \sum_{i=1}^n a_i Z(s_i) \right) &= \text{Var} \left( \sum_{i=1}^n a_i Z(s_i) - \overbrace{\sum_{i=1}^n a_i}^{0=} Z(s_0) \right) = \text{Var} \left( \sum_{i=1}^n a_i (Z(s_i) - Z(s_0)) \right) \\ &= \sum_{i=1}^n \sum_{j=1}^n a_i a_j E((Z(s_i) - Z(s_0))(Z(s_j) - Z(s_0))) \end{aligned}$$

Let  $c_I(s, t) = E((Z(s_i) - Z(s_0))(Z(s_j) - Z(s_0)))$ .

(2) It is

$$\begin{aligned} \gamma(s - t) &= \frac{1}{2} E(Z(s) - Z(s_0) + Z(t) - Z(s_0))^2 \\ &= \frac{1}{2} (2\gamma(s - s_0) + 2\gamma(t - s_0) - 2c_I(s, t)) \\ \implies c_I(s, t) &= \gamma(s - s_0) + \gamma(t - s_0) - \gamma(s - t) \end{aligned}$$

It is

$$\begin{aligned}
0 \leq \text{Var} \left( \sum_{i=1}^n a_i Z(s_i) \right) &= \sum_{i=1}^n \sum_{j=1}^n a_i a_j c_I(s_i, s_j) \\
&= \sum_{i=1}^n \sum_{j=1}^n a_i a_j (\gamma(s_i) + \gamma(s_j) - \gamma(s_i - s_j)) \\
&= \sum_{i=1}^n a_i \gamma(s_i) \sum_{j=1}^n a_j + \sum_{j=1}^n a_j \gamma(s_j) \sum_{i=1}^n a_i - \sum_{i=1}^n \sum_{j=1}^n a_i a_j \gamma(s_i - s_j)
\end{aligned}$$

hence

$$\sum_{i=1}^n \sum_{j=1}^n a_i a_j \gamma(s_i - s_j) \leq 0$$


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**Exercise 2.** (★) Consider the zero-mean random field  $Z = (Z(s) : s \in \mathbb{R}^d)$  with covariogram function given by

$$c(h) = \begin{cases} \xi^2 (1 + \rho \|h\|) \exp(-\rho \|h\|), & h > 0 \\ \nu^2 + \xi^2, & h = 0 \end{cases}$$

- (1) Compute the semivariogram for the random field  $(Z(s) : s \in \mathbb{R}^d)$
- (2) What are the nugget, sill and partial sill for this covariance model? Justify your answer.
- (3) Would the slightly altered covariance function defined below be a good model for spatial data for  $\phi > 0$ ? Justify your answer.

$$c(h) = \begin{cases} \xi^2 (1 + \rho \|h\|) \exp(-\rho \|h\|) + \phi, & h > 0 \\ \nu^2 + \xi^2 + \phi, & h = 0 \end{cases}$$

**Solution.**

- (1) For all  $h \neq 0$ , it is

$$\begin{aligned}
\gamma(h) &= c(0) - c(h), \\
&= \nu^2 + \xi^2 - \xi^2 (1 + \rho \|h\|) \exp(-\rho \|h\|) \\
&= \nu^2 + \xi^2 (1 - (1 + \rho \|h\|) \exp(-\rho \|h\|))
\end{aligned}$$

then

$$\gamma(h) = \begin{cases} \nu^2 + \xi^2 (1 - (1 + \rho \|h\|) \exp(-\rho \|h\|)) & h > 0 \\ 0 & h = 0 \end{cases}$$

- (2)

- The sill is the covariogram function at distance 0, that is  $c(0) = \nu^2 + \xi^2$ . Or since analogously, it is  $\lim_{\|h\| \rightarrow \infty} \gamma(h)$ . So,

$$\begin{aligned} \lim_{\|h\| \rightarrow \infty} (\|h\| \exp(-\rho \|h\|)) &= \lim_{\|h\| \rightarrow \infty} (\|h\| / \exp(\rho \|h\|)) \\ &= \lim_{\|h\| \rightarrow \infty} (\|h\| / \exp(\rho \|h\|)) = \lim_{\|h\| \rightarrow \infty} (\exp(-\rho \|h\|)) = 0 \end{aligned}$$

then

$$\lim_{\|h\| \rightarrow \infty} \gamma(h) = \nu^2 + \xi^2$$

- The nugget effect is the limiting value of the semicovariogram as  $h \rightarrow 0$  from above, hence it is  $\gamma(h) \rightarrow \nu^2$  as  $h \rightarrow 0^+$ .
  - The partial sill is the sill minus the nugget and is hence  $\xi^2$ .
- (3) No, it would be unrealistic because if  $\phi > 0$  then the covariance is always positive for infinitely large distances  $h$ . In practical terms this means that two points will always be correlated however far apart they are, it would be unrealistic.