

## Solutions for Work Sheet 8

### Problem 1 (Constructing a 'germ-grain model' which is not closed)

Find a dimension  $d \in \mathbb{N}$ , a probability measure  $\mathbb{Q}$  on  $\mathcal{C}^d$ , and a point process  $\Phi$  in  $\mathbb{R}^d$  such that the germ-grain model corresponding to the independent  $\mathbb{Q}$ -marking of  $\Phi$  is almost surely not a closed set.

**Proposed solution:** We choose  $d = 2$ , and let  $\Phi = \sum_{j=1}^{\infty} \delta_{(j, j^{-1})}$  be deterministic. Let  $X$  be a discrete random variable with

$$\mathbb{P}(X = k) = \frac{1}{k} - \frac{1}{k+1}, \quad k \in \mathbb{N}.$$

We choose as  $\mathbb{Q}$  the distribution of the random compact set  $Z_0 := [-X, 0] \times \{0\}$ . The corresponding germ-grain model  $Z$  almost surely has an accumulation point in the origin, but the origin is almost surely not contained in  $Z$ . Indeed, notice that the independent  $\mathbb{Q}$ -marking of  $\Phi$  is given through

$$\Psi = \sum_{j=1}^{\infty} \delta_{((j, j^{-1}), [-X_j, 0] \times \{0\})},$$

where  $X_1, X_2, \dots$  are i.i.d. copies of  $X$ , and that

$$\sum_{k=1}^{\infty} \mathbb{P}(X_k \geq k) = \sum_{k=1}^{\infty} \frac{1}{k} = \infty,$$

so the Borel-Cantelli lemma yields that (almost surely) infinitely many of the events

$$\left\{ (0, \frac{1}{k}) \in (k, \frac{1}{k}) + ([-X_k, 0] \times \{0\}) \right\}$$

occur and thus  $Z$  has an accumulation point in the origin  $(0, 0)$ . However,

$$\mathbb{P}((0, 0) \in Z) = \mathbb{P}\left((0, 0) \in \bigcup_{k=1}^{\infty} \left( (k, \frac{1}{k}) + ([-X_k, 0] \times \{0\}) \right)\right) = 0,$$

that is to say,  $Z$  is almost surely not closed.

### Problem 2 (The mean covariogram of a Boolean model)

Let  $Z = \bigcup_{k=1}^{\infty} (Z_k + \xi_k)$  be a stationary Boolean model in  $\mathbb{R}^d$  with parameters  $\gamma$  and  $\mathbb{Q}$ . Let  $Z_0 \sim \mathbb{Q}$  be the typical grain and  $p_Z := \mathbb{E}[\lambda^d(Z \cap [0, 1]^d)] = \mathbb{P}(0 \in Z)$  the volume fraction of  $Z$  (see Theorem 1.31). Further, let  $\Phi = \sum_{j=1}^{\infty} \delta_{\xi_j}$  be the (stationary) Poisson process of the germs.

a) Prove that, for any  $C \in \mathcal{C}^d$  and  $x, y \in \mathbb{R}^d$ ,

$$(C + y) \cap \{x, 0\} \neq \emptyset \iff y \in C^* \cup (x + C^*).$$

b) The mean covariogram  $C_0$  of  $Z$  is given through

$$C_0(x) := \mathbb{E}[\lambda^d(Z_0 \cap (Z_0 + x))], \quad x \in \mathbb{R}^d.$$

Prove that, for each  $x \in \mathbb{R}^d$ ,

$$\mathbb{E}[\lambda^d(Z_0 \cup (Z_0 - x))] = 2 \cdot \mathbb{E}[\lambda^d(Z_0)] - C_0(x).$$

c) Use parts a) and b) to show that

$$\mathbb{P}(0 \in Z, x \in Z) = p_Z^2 + (1 - p_Z)^2(e^{\gamma \cdot C_0(x)} - 1).$$

**Proposed solution:**

a) For  $C \in \mathcal{C}^d$  and  $x, y \in \mathbb{R}^d$  we have

$$\begin{aligned} (C + y) \cap \{x, 0\} \neq \emptyset &\iff \text{there exists } z \in C \text{ such that } z + y = x \text{ or } z + y = 0 \\ &\iff \text{there exists } z \in C \text{ such that } y = x - z \text{ or } y = -z \\ &\iff y \in (x + C^*) \cup C^*. \end{aligned}$$

b) For  $x \in \mathbb{R}^d$  we have

$$\begin{aligned} \mathbb{E}[\lambda^d(Z_0 \cup (Z_0 - x))] &= \mathbb{E}[\lambda^d(Z_0) + \lambda^d(Z_0 - x) - \lambda^d(Z_0 \cap (Z_0 - x))] \\ &= \mathbb{E}[2 \cdot \lambda^d(Z_0) - \lambda^d((Z_0 + x) \cap Z_0)] \\ &= 2 \cdot \mathbb{E}[\lambda^d(Z_0)] - C_0(x). \end{aligned}$$

c) From Equation (3.11) of the lecture notes we know that

$$p_Z = 1 - e^{-\gamma \cdot C_0(0)}.$$

Therefore,

$$\begin{aligned} p_Z^2 + (1 - p_Z)^2(e^{\gamma \cdot C_0(0)} - 1) &= p_Z^2 + (1 - p_Z)^2 \left( \frac{1}{-(1 - e^{-\gamma \cdot C_0(0)}) + 1} - 1 \right) \\ &= p_Z^2 + (1 - p_Z)^2 \left( \frac{1}{1 - p_Z} - 1 \right) \\ &= p_Z^2 + 1 - p_Z - (1 - p_Z)^2 \\ &= p_Z \\ &= \mathbb{P}(0 \in Z), \end{aligned}$$

which is the claim for  $x = 0$ . Now, let  $x \neq 0$ . Then, (using Theorem 3.16 a) and Theorem 3.6)

$$\begin{aligned} \mathbb{P}(0 \in Z, x \in Z) &= 1 - \mathbb{P}(\{0 \notin Z\} \cup \{x \notin Z\}) \\ &= 1 - \mathbb{P}(0 \notin Z) - \mathbb{P}(x \notin Z) + \mathbb{P}(0 \notin Z, x \notin Z) \\ &= 1 - \mathbb{P}(0 \notin Z) - \mathbb{P}(x \notin Z) + \mathbb{P}(Z \cap \{0, x\} = \emptyset) \\ &= 1 - \mathbb{P}(0 \notin Z) - \mathbb{P}(x \notin Z) + (1 - T_Z(\{0, x\})) \\ &= 2p_Z - 1 + \exp \left( -\gamma \int_{\mathbb{R}^d} \int_{\mathcal{C}^d} \mathbb{1}\{(K + y) \cap \{0, x\} \neq \emptyset\} dQ(K) d\lambda^d(y) \right) \\ &= 2p_Z - 1 + \exp \left( -\gamma \int_{\mathbb{R}^d} \mathbb{E}[\mathbb{1}\{(Z_0 + y) \cap \{0, x\} \neq \emptyset\}] d\lambda^d(y) \right) \end{aligned} \quad (1)$$

Part a) implies

$$\begin{aligned} \int_{\mathbb{R}^d} \mathbb{1}\{(Z_0 + y) \cap \{0, x\} \neq \emptyset\} d\lambda^d(y) &= \int_{\mathbb{R}^d} \mathbb{1}\{y \in (x + Z_0^*) \cup Z_0^*\} d\lambda^d(y) \\ &= \lambda^d((x + Z_0^*) \cup Z_0^*) \\ &= \lambda^d((Z_0 - x) \cup Z_0). \end{aligned}$$

Therefore, Equation (1) simplifies to

$$\mathbb{P}(0 \in Z, x \in Z) = 2p_Z - 1 + \exp\left(-\gamma \cdot \mathbb{E}\left[\lambda^d(Z_0 \cup (Z_0 - x))\right]\right).$$

Part b) and Equation (3.11) of the lecture notes yield

$$\begin{aligned}\mathbb{P}(0 \in Z, x \in Z) &= 2p_Z - 1 + \exp\left(-\gamma \cdot (2\mathbb{E}[\lambda^d(Z_0)] - C_0(x))\right) \\ &= 2p_Z - 1 + \exp\left(-2\gamma \cdot \mathbb{E}[\lambda^d(Z_0)]\right) \exp(\gamma \cdot C_0(x)) \\ &= 2p_Z - 1 + (1 - p_Z)^2 \cdot \exp(\gamma \cdot C_0(x)) \\ &= p_Z^2 + (1 - p_Z)^2 (e^{\gamma \cdot C_0(x)} - 1).\end{aligned}$$

### Problem 3 (The concept of visibility)

Let  $Z = \bigcup_{k=1}^{\infty} (Z_k + \xi_k)$  be a stationary Boolean model in  $\mathbb{R}^d$  with intensity  $\gamma > 0$  and shape distribution  $\mathcal{Q}$  which is concentrated on those sets in  $\mathcal{C}^d$  that contain  $0 \in \mathbb{R}^d$ . Further, let  $v_d := \int_{\mathcal{C}^d} \lambda^d(C) d\mathcal{Q}(C) < \infty$ . A point  $z \in Z$  is called visible if

$$|\{k \in \mathbb{N} : z \in Z_k + \xi_k\}| = 1,$$

that is, if  $z$  is contained in precisely one grain of the Boolean model. Define

$$\Phi := \sum_{j=1}^{\infty} \delta_{\xi_j} \cdot \mathbb{1}\{\xi_j \text{ is visible}\}.$$

Prove that  $\Phi$  is a stationary point process in  $\mathbb{R}^d$  with intensity  $\gamma_{\mathcal{Q}} := \gamma \cdot e^{-\gamma \cdot \mathbb{E}[\lambda^d(Z_1)]}$ .

**Hint:** For the calculation of the intensity you may use, without proof, Mecke's formula, which states that a Poisson process  $\eta$  with intensity measure  $\Lambda$  satisfies

$$\mathbb{E}\left[\int_{\mathbb{R}^d} f(x, \eta) d\eta(x)\right] = \mathbb{E}\left[\int_{\mathbb{R}^d} f(x, \eta + \delta_x) d\Lambda(x)\right]$$

for every measurable map  $f : \mathbb{R}^d \times \mathcal{N}(\mathbb{R}^d) \rightarrow [0, \infty)$ .

**Proposed solution:** Let  $\Psi := \sum_{j=1}^{\infty} \delta_{(\xi_j, Z_j)}$  be the stationary, independently marked Poisson process which underlies the Boolean model at hand. Put

$$\mu_{\Psi}(x) := \sum_{j=1}^{\infty} \mathbb{1}_{Z_j + \xi_j}(x), \quad x \in \mathbb{R}^d.$$

We obtain that a point  $x \in \mathbb{R}^d$  is visible if, and only if,  $\mu_{\Psi}(x) = 1$ . We can thus write

$$\Phi(A) = \sum_{j=1}^{\infty} \delta_{\xi_j}(A) \cdot \mathbb{1}\{\mu_{\Psi}(\xi_j) = 1\} = \int_{\mathbb{R}^d \times \mathcal{C}^d} \mathbb{1}\{x \in A\} \cdot \mathbb{1}\{\mu_{\Psi}(x) = 1\} d\Psi(x, K), \quad A \in \mathcal{B}(\mathbb{R}^d),$$

which shows that  $\Phi$  is a point process in  $\mathbb{R}^d$ . The stationarity of  $\Phi$  follows from

$$\begin{aligned}(\Phi + t)(A) &= \Phi(A - t) = \int_{\mathbb{R}^d \times \mathcal{C}^d} \mathbb{1}\{x \in A - t\} \cdot \mathbb{1}\{\mu_{\Psi}(x) = 1\} d\Psi(x, K) \\ &\stackrel{d}{=} \int_{\mathbb{R}^d \times \mathcal{C}^d} \mathbb{1}\{x \in A - t\} \cdot \mathbb{1}\{\mu_{T_{-t}(\Psi)}(x) = 1\} d\Psi(x, K) \\ &\stackrel{d}{=} \int_{\mathbb{R}^d \times \mathcal{C}^d} \mathbb{1}\{x - t \in A - t\} \cdot \mathbb{1}\{\mu_{T_{-t}(\Psi)}(x - t) = 1\} d\Psi(x, K) \\ &= \int_{\mathbb{R}^d \times \mathcal{C}^d} \mathbb{1}\{x \in A\} \cdot \mathbb{1}\{\mu_{\Psi}(x) = 1\} d\Psi(x, K) \\ &= \Phi(A),\end{aligned}$$

where  $A \in \mathcal{B}(\mathbb{R}^d)$  and  $t \in \mathbb{R}^d$  are arbitrary. To calculate the intensity of  $\Phi$ , note that Mecke's formula and the stationarity of  $Z$  imply

$$\begin{aligned}
\gamma_Q &= \mathbb{E} \left[ \int_{\mathbb{R}^d} \mathbb{1}\{x \in [0, 1]^d\} d\Phi(x) \right] \\
&= \mathbb{E} \left[ \int_{\mathbb{R}^d \times \mathcal{C}^d} \mathbb{1}\{x \in [0, 1]^d\} \cdot \mathbb{1}\{\mu_\Psi(x) = 1\} d\Psi(x, K) \right] \\
&= \gamma \int_{\mathbb{R}^d} \int_{\mathcal{C}^d} \mathbb{E} \left[ \mathbb{1}\{x \in [0, 1]^d\} \cdot \mathbb{1}\{\mu_{\Psi+\delta_{(x,K)}}(x) = 1\} \right] dQ(K) d\lambda^d(x) \\
&= \gamma \int_{\mathbb{R}^d} \mathbb{E} \left[ \mathbb{1}\{x \in [0, 1]^d\} \cdot \mathbb{1}\{\mu_\Psi(x) = 0\} \right] d\lambda^d(x) \\
&= \gamma \int_{\mathbb{R}^d} \mathbb{1}\{x \in [0, 1]^d\} \cdot \mathbb{E} \left[ \mathbb{1}\{x \notin Z\} \right] d\lambda^d(x) \\
&= \gamma \cdot \mathbb{P}(0 \notin Z) \\
&= \gamma \cdot (1 - p_Z) \\
&= \gamma \cdot e^{-\gamma \cdot \mathbb{E}[\lambda^d(Z_1)]}.
\end{aligned}$$

Note that in the fourth equality we have used that  $Q$  is concentrated on sets that contain  $0 \in \mathbb{R}^d$ , and in the last equality Equation (3.11) of the lecture notes comes into play.

#### Problem 4 (Spherical contact distribution functions)

Let  $Z$  be the stationary Boolean model in  $\mathbb{R}^d$  with intensity  $\gamma > 0$  and typical grain  $Z_0$  given through a  $d$ -dimensional ball of random radius  $R_0 \sim \text{Exp}(\lambda)$  ( $\lambda > 0$ ) around  $0 \in \mathbb{R}^d$ .

- Calculate the spherical contact distribution function  $H_{B(0,1)}(r)$ ,  $r > 0$ .
- Now, let  $d = 2$  and  $r_1, r_2 > 0$  such that  $r_1 \neq r_2$ . Assume that  $H_{B(0,1)}(r_1)$  and  $H_{B(0,1)}(r_2)$  are known. Determine the intensity  $\gamma$  as well as the parameter  $\lambda$  in dependence of  $H_{B(0,1)}(r_1)$  and  $H_{B(0,1)}(r_2)$ .

#### Proposed solution:

- Theorem 3.27 of the lecture notes, and the fact that  $\mathbb{E}[R_0^k] = \frac{k!}{\lambda^k}$ , imply

$$\begin{aligned}
H_{B(0,1)}(r) &= 1 - \exp \left( -\gamma \cdot \mathbb{E} \left[ \lambda^d ((Z_0 + r \cdot B(0,1)^*) \setminus Z_0) \right] \right) \\
&= 1 - \exp \left( -\gamma \cdot \mathbb{E} \left[ \lambda^d ((R_0 + r) \cdot B(0,1) \setminus R_0 \cdot B(0,1)) \right] \right) \\
&= 1 - \exp \left( -\gamma \kappa_d \cdot \mathbb{E} \left[ (R_0 + r)^d - R_0^d \right] \right) \\
&= 1 - \exp \left( -\gamma \kappa_d \cdot \mathbb{E} \left[ \sum_{j=0}^{d-1} \binom{d}{j} r^{d-j} R_0^j \right] \right) \\
&= 1 - \exp \left( -\gamma \kappa_d \sum_{j=0}^{d-1} \binom{d}{j} r^{d-j} \mathbb{E}[R_0^j] \right) \\
&= 1 - \exp \left( -\gamma \kappa_d \sum_{j=0}^{d-1} \binom{d}{j} \frac{j!}{\lambda^j} r^{d-j} \right) \\
&= 1 - \exp \left( -\gamma \kappa_d \cdot \frac{d!}{\lambda^d} \sum_{j=0}^{d-1} \frac{(r \cdot \lambda)^{d-j}}{(d-j)!} \right) \\
&= 1 - \exp \left( -\gamma \kappa_d \cdot \frac{d!}{\lambda^d} \sum_{k=1}^d \frac{(r \cdot \lambda)^k}{k!} \right), \quad r > 0.
\end{aligned}$$

- Define

$$f(r) := -\log(1 - H_{B(0,1)}(r)), \quad r > 0.$$

Part a) yields

$$f(r) = \gamma \pi \left( r^2 + \frac{2}{\lambda} r \right), \quad r > 0,$$

so, in particular,

$$\frac{f(r_1)}{r_1} = \gamma \pi \left( r_1 + \frac{2}{\lambda} \right) \quad \text{and} \quad \frac{f(r_2)}{r_2} = \gamma \pi \left( r_2 + \frac{2}{\lambda} \right),$$

which is equivalent to  $Ax = b$  with

$$A := \pi \begin{pmatrix} r_1 & 2 \\ r_2 & 2 \end{pmatrix}, \quad x := \begin{pmatrix} \gamma \\ \gamma/\lambda \end{pmatrix}, \quad b := \begin{pmatrix} f(r_1)/r_1 \\ f(r_2)/r_2 \end{pmatrix}.$$

Notice that

$$A^{-1} = \frac{1}{2\pi(r_1 - r_2)} \begin{pmatrix} 2 & -2 \\ -r_2 & r_1 \end{pmatrix},$$

hence

$$\gamma = \frac{1}{\pi(r_1 - r_2)} \left( \frac{f(r_1)}{r_1} - \frac{f(r_2)}{r_2} \right),$$

and

$$\frac{\gamma}{\lambda} = \frac{1}{2\pi(r_1 - r_2)} \left( -\frac{r_2 \cdot f(r_1)}{r_1} + \frac{r_1 \cdot f(r_2)}{r_2} \right) = \frac{r_1 \cdot r_2}{2\pi(r_1 - r_2)} \left( -\frac{f(r_1)}{r_1^2} + \frac{f(r_2)}{r_2^2} \right).$$

As the function  $r \mapsto \frac{f(r)}{r^2}$  is strictly decreasing, the right hand side of the previous equation is  $\neq 0$  (recall that  $r_1 \neq r_2$ ). We conclude that

$$\begin{aligned} \lambda &= \frac{1}{\pi(r_1 - r_2)} \left( \frac{f(r_1)}{r_1} - \frac{f(r_2)}{r_2} \right) \bigg/ \frac{r_1 \cdot r_2}{2\pi(r_1 - r_2)} \left( -\frac{f(r_1)}{r_1^2} + \frac{f(r_2)}{r_2^2} \right) \\ &= \frac{2}{r_1 r_2} \cdot \frac{r_1 r_2^2 f(r_1) - r_1^2 r_2 f(r_2)}{r_1^2 r_2^2} \cdot \frac{r_1^2 r_2^2}{-r_2^2 f(r_1) + r_1^2 f(r_2)} \\ &= 2 \cdot \frac{f(r_1) \cdot r_2 - f(r_2) \cdot r_1}{r_1^2 \cdot f(r_2) - r_2^2 \cdot f(r_1)}. \end{aligned}$$