

Solutions for Work Sheet 6

Problem 1 (Some advanced properties of the Poisson process)

Let (\mathbb{X}, ρ) be a separable metric space, and let $(\Omega, \mathcal{A}, \mathbb{P})$ be the probability space which underlies the random quantities that appear in the following.

- a) Let Φ be a Poisson process in \mathbb{X} with intensity measure Θ . For $A \in \mathcal{X}$ such that $0 < \Theta(A) < \infty$, and $k \in \mathbb{N}$, we have

$$\mathbb{P}(\Phi_A \in \cdot \mid \Phi(A) = k) = \mathbb{P}\left(\sum_{j=1}^k \delta_{X_j} \in \cdot\right)$$

with independent $X_1, \dots, X_k \sim \frac{\Theta_A}{\Theta(A)}$.

- b) Let Ψ be a point process in \mathbb{X} . Then Ψ is a Poisson process with intensity measure $\lambda \in \mathcal{M}(\mathbb{X})$ if, and only if,

$$\mathbb{E}[F(\Psi_A)] = e^{-\lambda(A)} \left(F(\mathbf{0}) + \sum_{k=1}^{\infty} \frac{1}{k!} \int_{A^k} F\left(\sum_{j=1}^k \delta_{y_j}\right) d\lambda^k(y_1, \dots, y_k) \right)$$

for all $A \in \mathcal{X}$ with $\lambda(A) < \infty$, and all measurable $F : \mathcal{N}(\mathbb{X}) \rightarrow [0, \infty]$.

Proposed solution:

- a) By Problem 2 b) on Sheet 5, Φ_A is a Poisson process with intensity measure Θ_A . By Theorem 2.30 of the lecture notes, we have

$$\Phi_A \stackrel{d}{=} \sum_{j=1}^{\tau} \delta_{X_j},$$

where $X_j \sim \frac{\Theta_A}{\Theta(A)}$, $\tau \sim \text{Po}(\Theta(A))$, and where $(X_j)_{j \in \mathbb{N}}$ and τ are independent. In particular, $\tau \stackrel{d}{=} \Phi(A)$, and we conclude that, for $k \in \mathbb{N}$,

$$\begin{aligned} \mathbb{P}(\Phi_A \in \cdot \mid \Phi(A) = k) &= \mathbb{P}\left(\sum_{j=1}^{\tau} \delta_{X_j} \in \cdot, \tau = k\right) / \mathbb{P}(\tau = k) = \mathbb{P}\left(\sum_{j=1}^k \delta_{X_j} \in \cdot, \tau = k\right) / \mathbb{P}(\tau = k) \\ &= \mathbb{P}\left(\sum_{j=1}^k \delta_{X_j} \in \cdot\right). \end{aligned}$$

- b) We show the two implications separately:

' \implies ' Assume that Ψ is a Poisson process with intensity measure λ , and let $A \in \mathcal{X}$ with $\lambda(A) < \infty$ and $F : \mathcal{N}(\mathbb{X}) \rightarrow [0, \infty]$ a measurable function. If $\lambda(A) = 0$, the statement is trivial, so we assume

$\lambda(A) \in (0, \infty)$. Then, using part a) of the exercise, we have

$$\begin{aligned}\mathbb{E}[F(\Psi_A)] &= \sum_{k=0}^{\infty} \mathbb{P}(\Psi(A) = k) \cdot \mathbb{E}[F(\Psi_A) \mid \Psi(A) = k] \\ &= e^{-\lambda(A)} F(\mathbf{0}) + e^{-\lambda(A)} \sum_{k=1}^{\infty} \frac{\lambda(A)^k}{k!} \int_{A^k} F\left(\sum_{j=1}^k \delta_{y_j}\right) d\left(\frac{\lambda}{\lambda(A)}\right)^k(y_1, \dots, y_k) \\ &= e^{-\lambda(A)} \left(F(\mathbf{0}) + \sum_{k=1}^{\infty} \frac{1}{k!} \int_{A^k} F\left(\sum_{j=1}^k \delta_{y_j}\right) d\lambda^k(y_1, \dots, y_k) \right).\end{aligned}$$

' \Leftarrow ' Now, assume that the stated equation holds for all sets $A \in \mathcal{X}$ with $\lambda(A) < \infty$ and all measurable $F : N(\mathbb{X}) \rightarrow [0, \infty]$. Thus, for arbitrary $B \in \mathcal{X}$ with $\lambda(B) < \infty$ and $k \in \mathbb{N}_0$, we define the function $F(\mu) := \mathbb{1}\{\mu(B) = k\}$ to conclude that

$$\begin{aligned}\mathbb{P}(\Psi(B) = k) &= \mathbb{E}[F(\Psi_B)] = e^{-\lambda(B)} \left(F(\mathbf{0}) + \sum_{\ell=1}^{\infty} \frac{1}{\ell!} \int_{B^\ell} F\left(\sum_{j=1}^{\ell} \delta_{y_j}\right) d\lambda^\ell(y_1, \dots, y_\ell) \right) \\ &= e^{-\lambda(B)} \frac{\lambda(B)^k}{k!}.\end{aligned}$$

For $B \in \mathcal{X}$ with $\lambda(B) = \infty$, we choose $C_i \in \mathcal{X}$ with $\lambda(C_i) < \infty$ and $C_i \nearrow \mathbb{X}$, and observe that, for $k \in \mathbb{N}_0$,

$$\mathbb{P}(\Psi(B) \leq k) \leq \liminf_{i \rightarrow \infty} \mathbb{P}(\Psi(B \cap C_i) \leq k) = \liminf_{i \rightarrow \infty} \sum_{j=0}^k e^{-\lambda(B \cap C_i)} \frac{\lambda(B \cap C_i)^j}{j!} = 0,$$

so $\mathbb{P}(\Psi(B) = \infty) = 1$. Now, let $B_1, \dots, B_k \in \mathcal{X}$ be pairwise disjoint with $\lambda(B_j) < \infty$, $j = 1, \dots, k$, and let $n_1, \dots, n_k \in \mathbb{N}_0$, $n := \sum_{j=1}^k n_j$. If $n = 0$, i.e., if $n_j = 0$ for $j = 1, \dots, k$, then

$$\mathbb{P}(\Psi(B_1) = 0, \dots, \Psi(B_k) = 0) = \exp\left(-\lambda\left(\bigcup_{j=1}^k B_j\right)\right) = \prod_{j=1}^k e^{-\lambda(B_j)} = \prod_{j=1}^k \mathbb{P}(\Psi(B_j) = 0).$$

If $n \geq 1$, we define $\tilde{F}(\mu) := \mathbb{1}\{\mu(B_1) = n_1, \dots, \mu(B_k) = n_k\}$ and $B := \bigcup_{j=1}^k B_j$ (note: $\lambda(B) < \infty$), and conclude that

$$\begin{aligned}\mathbb{P}(\Psi(B_1) = n_1, \dots, \Psi(B_k) = n_k) &= \mathbb{E}[\tilde{F}(\Psi_B)] = \frac{e^{-\lambda(B)}}{n!} \int_{B^n} \tilde{F}\left(\sum_{j=1}^n \delta_{y_j}\right) d\lambda^n(y_1, \dots, y_n) \\ &= \frac{e^{-\lambda(B)}}{n!} \binom{n}{n_1, \dots, n_k} \prod_{j=1}^k \lambda(B_j)^{n_j} \\ &= \prod_{j=1}^k \mathbb{P}(\Psi(B_j) = n_j).\end{aligned}$$

Thus follows the independence of $\Psi(B_1), \dots, \Psi(B_k)$. For general pairwise disjoint $B_1, \dots, B_k \in \mathcal{X}$, choose a sequence $A_i \in \mathcal{X}$ of pairwise disjoint sets with $\lambda(A_i) < \infty$ and $\bigcup_{i \in \mathbb{N}} A_i = \mathbb{X}$. Then, $\Psi(B_j \cap A_i)$, $j = 1, \dots, k$ and $i \in \mathbb{N}$, are independent and, therefore, so are $\Psi(B_j) = \sum_{i=1}^{\infty} \Psi(B_j \cap A_i)$, for $j = 1, \dots, k$.

Problem 2

Let $m \in \mathbb{N}$ and $\gamma_1, \dots, \gamma_m > 0$ such that $\gamma_j \neq \gamma_k$ for all $j, k \in \{1, \dots, m\}$, $j \neq k$. Further, let $p_1, \dots, p_m \in (0, 1]$ with $\sum_{j=1}^m p_j = 1$. Let X be a discrete random variable taking values in $\{\gamma_1, \dots, \gamma_m\}$ such that

$$\mathbb{P}(X = \gamma_j) = p_j, \quad j = 1, \dots, m.$$

Consider a point process Φ in \mathbb{R}^d whose distribution is specified by the conditional distributions

$$\mathbb{P}(\Phi(B) = n \mid X = \gamma_j) = \frac{(\gamma_j \cdot \lambda^d(B))^n}{n!} \cdot e^{-\gamma_j \cdot \lambda^d(B)},$$

where $B \in \mathcal{B}(\mathbb{R}^d)$, $n \in \mathbb{N}_0$, and $j = 1, \dots, m$. Prove the following assertions.

- The process Φ is stationary.
- The process Φ is not a Poisson process for $m \geq 2$.

Proposed solution:

- First of all, notice that the process Φ is simple. Indeed, for any $x \in \mathbb{R}^d$, we have

$$\mathbb{P}(\Phi(\{x\}) \geq 1) = \sum_{n=1}^{\infty} \sum_{j=1}^m p_j \cdot \mathbb{P}(\Phi(\{x\}) = n \mid X = \gamma_j) = \sum_{j=1}^m p_j \sum_{n=1}^{\infty} \frac{(\gamma_j \cdot \lambda^d(\{x\}))^n}{n!} \cdot e^{-\gamma_j \cdot \lambda^d(\{x\})} = 0,$$

so $\mathbb{P}^\Phi(N_s(\mathbb{R}^d)) = \mathbb{P}(\Phi \in N_s(\mathbb{R}^d)) = 1$. In order to prove that Φ is stationary, let $B \in \mathcal{B}(\mathbb{R}^d)$ and $t \in \mathbb{R}^d$. We have

$$\begin{aligned} \mathbb{P}((\Phi + t)(B) = 0) &= \mathbb{P}(\Phi(B - t) = 0) = \sum_{j=1}^m p_j \cdot \mathbb{P}(\Phi(B - t) = 0 \mid X = \gamma_j) \\ &= \sum_{j=1}^m p_j \cdot \frac{(\gamma_j \cdot \lambda^d(B - t))^0}{0!} \cdot e^{-\gamma_j \cdot \lambda^d(B - t)} \\ &= \sum_{j=1}^m p_j \cdot \frac{(\gamma_j \cdot \lambda^d(B))^0}{0!} \cdot e^{-\gamma_j \cdot \lambda^d(B)} \\ &= \mathbb{P}(\Phi(B) = 0). \end{aligned}$$

Hence (by the hint), the measures $\mathbb{P}^{\Phi+t}$ and \mathbb{P}^Φ agree on a generating π -system of $N_s(\mathbb{R}^d)$ and are thus identical, that is, $\Phi + t \stackrel{d}{=} \Phi$.

- Choose any set $B \in \mathcal{B}(\mathbb{R}^d)$ with $0 < \lambda^d(B) < \infty$. The intensity measure of Φ evaluated in B satisfies

$$\begin{aligned} \mathbb{E}[\Phi(B)] &= \sum_{n=0}^{\infty} n \cdot \mathbb{P}(\Phi(B) = n) = \sum_{n=0}^{\infty} n \sum_{j=1}^m p_j \cdot \mathbb{P}(\Phi(B) = n \mid X = \gamma_j) \\ &= \sum_{n=0}^{\infty} n \sum_{j=1}^m p_j \cdot \frac{(\gamma_j \cdot \lambda^d(B))^n}{n!} \cdot e^{-\gamma_j \cdot \lambda^d(B)} \\ &= \sum_{j=1}^m p_j \cdot e^{-\gamma_j \cdot \lambda^d(B)} \sum_{n=1}^{\infty} \frac{(\gamma_j \cdot \lambda^d(B))^n}{(n-1)!} \\ &= \sum_{j=1}^m p_j \cdot e^{-\gamma_j \cdot \lambda^d(B)} \cdot \gamma_j \cdot \lambda^d(B) \cdot e^{\gamma_j \cdot \lambda^d(B)} \\ &= \lambda^d(B) \cdot \mathbb{E}[X], \end{aligned}$$

hence if Φ was a Poisson process, then we ought to have

$$\mathbb{P}(\Phi(B) = 0) = e^{-\lambda^d(B) \cdot \mathbb{E}[X]}.$$

By definition of Φ , however, we have

$$\mathbb{P}(\Phi(B) = 0) = \sum_{j=1}^m p_j \cdot \mathbb{P}(\Phi(B) = 0 \mid X = \gamma_j) = \sum_{j=1}^m p_j \cdot e^{-\gamma_j \cdot \lambda^d(B)} = \mathbb{E} \left[e^{-\lambda^d(B) \cdot X} \right].$$

Now, notice that since the function $[0, \infty) \ni x \mapsto \varphi(x) := e^{-\lambda^d(B) \cdot x}$ is convex, and since neither φ is linear nor are $\gamma_1 = \dots = \gamma_m$, Jensen's inequality yields

$$\mathbb{P}(\Phi(B) = 0) = \mathbb{E} \left[e^{-\lambda^d(B) \cdot X} \right] > e^{-\lambda^d(B) \cdot \mathbb{E}[X]},$$

so Φ cannot be a Poisson process.

Problem 3 (Some properties of p -thinnings)

Let (\mathbb{X}, ρ) be a separable metric space. Let $p : \mathbb{X} \rightarrow [0, 1]$ be measurable, and let Φ be a point process in \mathbb{X} and Φ_p the p -thinning of Φ , both defined on some probability space $(\Omega, \mathcal{A}, \mathbb{P})$.

a) Prove that, for any measurable $g : \mathbb{X} \rightarrow [0, \infty]$, the Laplace functional of Φ_p is given through

$$L_{\Phi_p}(g) := \mathbb{E} \left[\exp \left(- \int_{\mathbb{X}} g(x) d\Phi_p(x) \right) \right] = \mathbb{E} \left[\exp \left(\int_{\mathbb{X}} \log \left(1 - p(x) [1 - e^{-g(x)}] \right) d\Phi(x) \right) \right].$$

b) Let $B \in \mathbb{X}$. Can we interpret Φ_B as a p -thinning of Φ ? If so, how do we have to choose p ?

c) Prove that, for any $n \in \mathbb{N}$, Φ can be written as a superposition of identically distributed point processes Φ_1, \dots, Φ_n such that $\mathbb{P}(\Phi(\mathbb{X}) \geq 1) > 0$ implies

$$\mathbb{P}(\Phi_k(\mathbb{X}) \geq 1) > 0, \quad k = 1, \dots, n.$$

d) Prove that c) is not necessarily true if we require Φ_1, \dots, Φ_n to be independent.

e) Prove that if Φ is a Poisson process with intensity measure $\Theta \in \mathcal{M}(\mathbb{X})$ then Φ is infinitely divisible, that is, for each $n \in \mathbb{N}$ we find i.i.d. point processes Φ_1, \dots, Φ_n such that $\Phi \stackrel{d}{=} \Phi_1 + \dots + \Phi_n$.

Proposed solution: Recall that by Theorem 2.8 there exist \mathbb{X} -valued random elements ξ_1, ξ_2, \dots as well as an $\mathbb{N}_0 \cup \{\infty\}$ -valued random variable τ such that $\Phi = \sum_{j=1}^{\tau} \delta_{\xi_j}$. Let K be the stochastic kernel from \mathbb{X} to $\{0, 1\}$ given through

$$K(x, \cdot) := (1 - p(x)) \cdot \delta_0 + p(x) \cdot \delta_1, \quad x \in \mathbb{X}.$$

Further, let Ψ be a K -marking of Φ , that is, $\Psi = \sum_{j=1}^{\tau} \delta_{(\xi_j, Y_j)}$, where, conditional on $\tau, \xi_1, \xi_2, \dots$, the random variables Y_1, Y_2, \dots are independent with $Y_j \sim K(\xi_j, \cdot)$, $j \in \mathbb{N}$. By Definition 2.39, the p -thinning Φ_p of Φ is the point process $\Psi(\cdot \times \{1\})$. Observe that

$$\Phi_p(\cdot) = \Psi(\cdot \times \{1\}) = \sum_{j=1}^{\tau} \delta_{(\xi_j, Y_j)}(\cdot \times \{1\}) = \sum_{j=1}^{\tau} \delta_{\xi_j}(\cdot) \mathbb{1}\{Y_j = 1\}.$$

a) For measurable $f : \mathbb{X} \times \{0, 1\} \rightarrow [0, \infty]$ put $f^*(x) := -\log \left(\int_{\mathbb{X}} \exp(-f(x, y)) K(x, dy) \right)$. In the proof of Theorem 2.38 it was shown that

$$\mathbb{E} \left[\exp \left(- \int_{\mathbb{X}} f(x, y) d\Psi(x, y) \right) \right] = \mathbb{E} \left[\exp \left(- \int_{\mathbb{X}} f^*(x) d\Phi(x) \right) \right].$$

We define

$$f(x, y) = \begin{cases} g(x), & \text{if } y = 1, \\ 0, & \text{else.} \end{cases}$$

It follows that $f^*(x) = -\log [p(x) \cdot \exp(-g(x)) + (1 - p(x))]$, and, by choice of f ,

$$\mathbb{E} \left[\exp \left(- \int_{\mathbb{X}} f(x, y) d\Psi(x, y) \right) \right] = \mathbb{E} \left[\exp \left(- \int_{\mathbb{X}} g(x) d\Phi_p(x) \right) \right].$$

We conclude that

$$\begin{aligned}\mathbb{E}\left[\exp\left(-\int_{\mathbb{X}} g(x) d\Phi_p(x)\right)\right] &= \mathbb{E}\left[\exp\left(-\int_{\mathbb{X}} -\log(p(x) \cdot \exp(-g(x)) + (1-p(x))) d\Phi(x)\right)\right] \\ &= \mathbb{E}\left[\exp\left(\int_{\mathbb{X}} \log(1-p(x)[1-e^{-g(x)}]) d\Phi(x)\right)\right].\end{aligned}$$

- b) Yes, such an interpretation is possible. Indeed, choose $p(x) = \mathbb{1}_B(x)$ and $Y_j = \mathbb{1}\{\xi_j \in B\}$, for each $j \in \mathbb{N}$. Then,

$$\Phi_p = \sum_{j=1}^{\tau} \delta_{\xi_j} \mathbb{1}\{Y_j = 1\} = \sum_{j=1}^{\tau} \delta_{\xi_j} \mathbb{1}_B(\xi_j) = \Phi_B.$$

- c) Let Z_1, Z_2, \dots be independent random variables which are independent of Φ and which are uniformly distributed over $\{1, \dots, n\}$. Then,

$$\Phi = \sum_{j=1}^{\tau} \delta_{\xi_j} = \sum_{k=1}^n \underbrace{\sum_{j=1}^{\tau} \mathbb{1}\{Z_j = k\}}_{=: \Phi_k} \cdot \delta_{\xi_j} = \sum_{k=1}^n \Phi_k.$$

As the sequence $(Z_j)_{j \in \mathbb{N}}$ is independent of Φ and we have

$$\{\Phi_k(\mathbb{X}) \geq 1\} \subset \{\Phi(\mathbb{X}) \geq 1\} = \{\tau \geq 1\},$$

it follows that

$$\mathbb{P}(\Phi_k(\mathbb{X}) \geq 1) = \mathbb{P}(\Phi_k(\mathbb{X}) \geq 1, \tau \geq 1) \geq \mathbb{P}(Z_1 = k, \tau \geq 1) = \frac{1}{n} \cdot \mathbb{P}(\tau \geq 1) > 0.$$

- d) Consider $\Phi := \delta_x$ for some fixed $x \in \mathbb{X}$. Suppose there exist $n \in \mathbb{N}$, $n \geq 2$, and i.i.d. point processes Φ_1, \dots, Φ_n such that $\delta_x = \Phi_1 + \dots + \Phi_n$ almost surely. Then, for $k \in \{1, \dots, n\}$, we have $\Phi_k(\mathbb{X} \setminus \{x\}) = 0$ as well as $\Phi_k(\{x\}) \in \{0, 1\}$ almost surely. Therefore, $\Phi_k(\{x\}) \sim \text{Bern}(p_k)$ for some $p_k \in [0, 1]$. As Φ_1, \dots, Φ_n have the same distribution, $p_1 = \dots = p_n =: p$. Moreover, we have

$$0 = \mathbb{P}(\delta_x(\{x\}) = 2) = \mathbb{P}((\Phi_1 + \dots + \Phi_n)(\{x\}) = 2) = \binom{n}{2} p^2 (1-p)^{n-2},$$

hence $p \in \{0, 1\}$. However, neither for $p = 1$ nor for $p = 0$ do we have $\Phi_1 + \dots + \Phi_n = \delta_x$.

- e) Denote by \mathcal{U} the uniform distribution over $\{1, \dots, n\}$, and let Ψ be an independent \mathcal{U} -marking of Φ . By Corollary 2.37 and Theorem 2.38, Ψ is a Poisson process with intensity measure $\Theta \otimes \mathcal{U}$. Thus,

$$\Phi_1 := \Psi(\cdot \times \{1\}), \dots, \Phi_n := \Psi(\cdot \times \{n\})$$

are also Poisson processes, each of them having intensity measure $\frac{1}{n} \cdot \Theta$, as

$$(\Theta \otimes \mathcal{U})(B \times \{k\}) = \frac{1}{n} \cdot \Theta(B), \quad B \in \mathcal{X}, \quad k \in \{1, \dots, n\}.$$

To prove that Φ_1, \dots, Φ_n are independent, we write

$$\Phi_k = \Psi(\cdot \times \{k\}) = \Psi_{\mathbb{X} \times \{k\}}(\cdot \times \{k\}).$$

As the sets $\mathbb{X} \times \{k\}$, $k = 1, \dots, n$, are disjoint, the processes $\Psi_{\mathbb{X} \times \{1\}}, \dots, \Psi_{\mathbb{X} \times \{n\}}$ (and hence Φ_1, \dots, Φ_n) are independent by Problem 2 c) of Sheet 5. By Problem 4 b) on Sheet 5, $\Phi_1 + \dots + \Phi_n$ is a Poisson process with intensity measure $\sum_{k=1}^n \frac{1}{n} \cdot \Theta = \Theta$, and the claim follows.