

Applied Stochastic Processes

Solution sheet 12

Solution 12.1

- (a) First, we note that the product measure of two σ -finite measures taking values in $\mathbb{N} \cup \{\infty\}$ is itself a σ -finite measure taking values in $\mathbb{N} \cup \{\infty\}$. By Exercise 9.2 (a), showing that N is a point process on $(E \times F, \mathcal{E} \otimes \mathcal{F})$ is equivalent to showing that for all $C \in \mathcal{E} \otimes \mathcal{F}$, $N(C)$ is a random variable. By applying Dynkin's lemma, it actually suffices to show that $N(C)$ is a random variable for sets of the form $C = A \times B$ with $A \in \mathcal{E}$ and $B \in \mathcal{F}$ (which form a π -system). But in this case, we have by the definition of the product measure $N(A \times B) = N(A) \cdot N'(B)$, which is a product of two random variables and thus a random variable.
- (b) No, N is not a Poisson point process. To illustrate this, we consider the following example: Let $E = F = \mathbb{R}$ and $\mu = \nu = \text{Leb}$. Then

$$\{N([1, 2]^2) = 1, N([1, 2] \times [3, 4]) = 0\} = \{N([1, 2]) = 1, N'([1, 2]) = 1, N'([3, 4]) = 0\},$$

and so

$$\mathbb{P}[N([3, 4]^2) = 0 | N([1, 2]^2) = 1, N([1, 2] \times [3, 4]) = 0] = 1 \neq \mathbb{P}[N([3, 4]^2) = 0],$$

which shows that N cannot be a Poisson point process as it contradicts the independence property on disjoint sets.

- (c) Yes, this is exactly the marking theorem from Section 4.8.
- (d) Yes, we can define a probability measure $\bar{\nu}$ by $\bar{\nu}(B) = \frac{\nu(B)}{\nu(F)}$ for $B \in \mathcal{F}$ and consider a Poisson point process M on E with intensity measure $\nu(F) \cdot \mu$. Then by the marking theorem, the marked process \bar{M} is a Poisson point process on $E \times F$ with intensity measure $(\nu(F) \cdot \mu) \otimes \bar{\nu} = \mu \otimes \nu$. The equality of the two measures can be obtained by first noticing that they agree on sets of the form $C = A \times B$ for $A \in \mathcal{E}$ and $B \in \mathcal{F}$ and then applying Dynkin's lemma.

Solution 12.2

- (a) Let us consider the marked process $\bar{M} = \sum_i \delta_{(X_i, R_i)}$. By the marking theorem, this is a Poisson point process on $\mathbb{R}^d \times \mathbb{R}^+$ with intensity measure $\mu \otimes \rho$. In this process, each point (x, r) of the space corresponds to the ball $B(x, r)$. Note that the number of balls that intersect the origin is given by the number of points in the set $A = \{(x, r) \in \mathbb{R}^d \times \mathbb{R}^+ : |x| < r\}$. In other words $N_0 = \bar{M}(A)$. To see that A is measurable, we note that $A := f^{-1}((0, \infty))$ for the measurable function $f((x, r)) = r - |x|$. This implies that N_0 is a well defined random variable and that $N_0 \sim \text{Poisson}((\mu \otimes \rho)(A))$. We know by Fubini's Theorem that

$$(\mu \otimes \rho)(A) = \int_{\mathbb{R}^d \times \mathbb{R}^+} 1_A(y) (\mu \otimes \rho)(dy) = \int_{\mathbb{R}^d} \int_{|x|}^{\infty} \rho(dr) \mu(dx),$$

which shows what we wanted.

(b) First, note that $\{\mathcal{O} = \mathbb{R}^d\} = \bigcap_{N \geq 1} \{\mathcal{O} \supset \overline{B(0, N)}\}$. By compactness of $\overline{B(0, N)}$,

$$\{\mathcal{O} \supset \overline{B(0, N)}\} = \bigcup_{I \geq 1} \left\{ \overline{B(0, N)} \subset \bigcup_{1 \leq i \leq I} B(X_i, R_i) \right\},$$

and for every $I \in \mathbb{N}$,

$$\left\{ \overline{B(0, N)} \subset \bigcup_{1 \leq i \leq I} B(X_i, R_i) \right\} = \bigcup_{n \geq 1} \left\{ \overline{B(0, N)} \subset \bigcup_{1 \leq i \leq I} B(X_i, R_i) \text{ and } R_i \geq \frac{1}{n}, \forall i \leq I \right\}.$$

Since \mathbb{Q}^d is dense in \mathbb{R}^d , we have

$$\left\{ \overline{B(0, N)} \subset \bigcup_{1 \leq i \leq I} B(X_i, R_i) \text{ and } R_i \geq \frac{1}{n}, \forall i \leq I \right\} = \bigcap_{q \in \overline{B(0, N)} \cap \mathbb{Q}^d} \bigcup_{1 \leq i \leq I} \underbrace{\{q \in B(X_i, R_i)\}}_{=\{|X_i - q| < R_i\}}.$$

In summary, we have shown that $\{\mathcal{O} = \mathbb{R}^d\}$ can be written in terms of countable unions and intersections of measurable sets, and is therefore measurable. Second, we know by Fubini's theorem that

$$\int_{\mathbb{R}^d} \int_{|x|}^{\infty} \rho(dr) \mu(dx) = \int_0^{\infty} \int_{B(0, r)} \mu(dx) \rho(dr) = \pi_d \int_0^{\infty} r^d \rho(dr)$$

Hence,

$$\mathbb{P}[0 \notin \mathcal{O}] = \mathbb{P}[N_0 = 0] = \exp\left(-\pi_d \int_0^{\infty} r^d \rho(dr)\right).$$

Suppose that $\mathbb{P}[\mathcal{O} = \mathbb{R}^d] = 1$. Then $\mathbb{P}[0 \in \mathcal{O}] = 1$, and we deduce from the last expression that $\int_0^{\infty} r^d \rho(dr) = \infty$. To prove the converse, assume that $\int_0^{\infty} r^d \rho(dr) < \infty$. As a preliminary result we first show for any $n \in \mathbb{N}$ that

$$(\mu \otimes \rho)\left(\{(x, r) \in \mathbb{R}^d \times \mathbb{R}^+ : B(0, n) \subset B(x, r)\}\right) < \infty. \quad (1)$$

Since $B(0, n) \subset B(x, r)$ if and only if $r \geq |x| + n$, the left-hand side of equation (1) equals

$$\int_0^{\infty} \int_{\mathbb{R}^d} 1_{\{r \geq |x| + n\}} \mu(dx) \rho(dr) = \pi_d \int_n^{\infty} (r - n)^d \rho(dr).$$

This is bounded below by

$$\pi_d \int_{2n}^{\infty} \left(\frac{r}{2}\right)^d \rho(dr) = \pi_d 2^{-d} \int_0^{\infty} 1_{\{r \geq 2n\}} r^d \rho(dr),$$

proving (1). Since \overline{M} is a Poisson point process with intensity $\mu \otimes \rho$, the ball $B(0, n)$ is almost surely covered even by one of the balls $B(X_i, R_i)$. Since n is arbitrary, it follows that $\mathbb{P}[\mathcal{O} = \mathbb{R}^d] = 1$.

Solution 12.3

(a) Recall by definition that $\mu(B) = \mathbb{E}[M(B)]$. We have

$$\mathcal{L}_M(t1_B) = \mathbb{E} \left[\exp \left(-t \int_E 1_B M(dx) \right) \right] = \mathbb{E}[\exp(-tM(B))].$$

Since $M(B) \geq 0$, the exponential above is bounded by 1. Besides, $M(B) \in L^1(\mathbb{P})$, so we can exchange the derivative and the expectation in the Laplace functional, therefore

$$-\frac{d}{dt} \mathcal{L}_M(t1_B) = \mathbb{E}[M(B) \exp(-tM(B))].$$

It suffices now to take $t = 0$ to conclude.

(b) For all $t > 0$, we have $L_M(t1_B) = \mathbb{E}[\exp(-tM(B))] = \mathbb{E}[1_{\{M(B)=0\}} + e^{-tM(B)} 1_{\{M(B) \geq 1\}}]$. By dominated convergence, we get

$$\lim_{t \rightarrow \infty} \mathcal{L}_M(t1_B) = \mathbb{E}[1_{\{M(B)=0\}}] + 0 = \mathbb{P}[M(B) = 0].$$