

## Exercises for the lecture „Probability Theory I“

### Sheet 8

**Submission deadline:** Thursday, 26.06.2025, until 10:15 o'clock in the mailbox in the math institute

(You may deliver the exercise solutions in pairs.)

#### Exercise 1

(4 points)

Let  $(\Omega, \mathcal{A}, \mathbb{P})$  be a probability space and  $(\mathcal{A}_n)_{n \in \mathbb{N}}$  a sequence of independent  $\sigma$ -algebras.

- (a) Let  $\{I_k \mid k \in K\}$  be a partition of  $\mathbb{N}$ , i.e.  $\bigcup_{k \in K} I_k = \mathbb{N}$  and  $I_k \cap I_j = \emptyset$  for  $j \neq k$ . Prove that  $(\sigma(\mathcal{A}_j : j \in I_k))_{k \in K}$  is an independent family.

HINT: Consider the sets  $\mathcal{C}_k := \left\{ \bigcap_{j \in J_k} \mathcal{A}_j \mid J_k \subset I_k \text{ finite, } \mathcal{A}_j \in \mathcal{A}_j \right\}$ .

- (b) We define the *terminal  $\sigma$ -algebra*  $\mathcal{T}$  as

$$\mathcal{T} = \mathcal{T}(\mathcal{A}_1, \mathcal{A}_2, \dots) := \bigcap_{n \geq 1} \sigma \left( \bigcup_{m \geq n} \mathcal{A}_m \right).$$

Prove the 0-1-law of Kolmogorov: For every  $A \in \mathcal{T}$  we have  $\mathbb{P}(A) \in \{0, 1\}$ .

HINT: Prove that  $\mathcal{T}$  is independent of itself.

#### Exercise 2

(4 points)

Let  $X_1, X_2, \dots$  be a sequence of real valued random variables defined on some probability space  $(\Omega, \mathcal{A}, \mathbb{P})$ ,  $\mathcal{A}_n := \sigma(X_n)$  and  $S_n := \sum_{k=1}^n X_k$ . Prove or disprove with a counterexample whether or not the following events are part of the terminal  $\sigma$ -algebra  $\mathcal{T}(\mathcal{A}_1, \mathcal{A}_2, \dots)$  (defined in Exercise 1):

- (a)  $\{\omega \in \Omega \mid X_n(\omega) = 0\}$  for fixed  $n \in \mathbb{N}$ ,
- (b)  $\{\omega \in \Omega \mid X_n(\omega) = 0 \text{ for some } n \in \mathbb{N}\}$ ,
- (c)  $\{\omega \in \Omega \mid X_n(\omega) = 0 \text{ for finitely many } n \in \mathbb{N}\}$ ,
- (d)  $\{\omega \in \Omega \mid S_n(\omega) = 0 \text{ finitely many } n \in \mathbb{N}\}$ ,
- (e)  $\left\{ \omega \in \Omega \mid \limsup_{n \rightarrow \infty} S_n(\omega) > \liminf_{n \rightarrow \infty} S_n(\omega) \right\}$ .

(please turn over)

### Exercise 3

(4+2(bonus) points)

Let  $(\Omega, \mathcal{A}, \mathbb{P})$  a probability space and  $\mathcal{F} \subset \mathcal{A}$  a sub- $\sigma$ -algebra. Prove the following results about conditional expectation (Proposition 4.12):

- (a) For all  $\varepsilon > 0$ , we have  $\mathbb{P}(|X| \geq \varepsilon | \mathcal{F}) \leq \varepsilon^{-2} \mathbb{E}[X^2 | \mathcal{F}]$   $\mathbb{P}$ -a.s.
- (b) If  $\mathbb{E}[X^2] < \infty$  and  $\mathbb{E}[Y^2] < \infty$ , we have

$$\mathbb{E}[XY | \mathcal{F}]^2 \leq \mathbb{E}[X^2 | \mathcal{F}] \mathbb{E}[Y^2 | \mathcal{F}] \quad \mathbb{P}\text{-a.s.}$$

- (c) Let  $\varphi : \mathbb{R}^k \rightarrow \mathbb{R}$  be convex with  $\mathbb{E}[|\varphi(X)|] < \infty$ . Then Jensen's inequality holds for conditional expectation, i.e.

$$\varphi(\mathbb{E}[X | \mathcal{F}]) \leq \mathbb{E}[\varphi(X) | \mathcal{F}] \quad \mathbb{P}\text{-a.s.}$$

HINT: We say that a function  $\varphi : \mathbb{R}^k \rightarrow \mathbb{R}$  is convex if the *epigraph*  $\{(x, y) \in \mathbb{R}^k \times \mathbb{R} : \varphi(x) \leq y\}$  is a convex set, where a set  $C \subset \mathbb{R}^n$  is said to be convex if for all  $x, y \in C$  and  $0 \leq \lambda \leq 1$  we also have  $\lambda x + (1 - \lambda)y \in C$ . You can use the following fact without proof: For any convex set  $C \subset \mathbb{R}^n$  and  $p \in \partial C$ , there exists  $v_p \in \mathbb{R}^n$  such that  $\langle v_p, p - x \rangle \leq 0$  for all  $x \in C$ . If you work out a solution for this, you get two additional points.

- (d) Let  $X_n \geq 0$ ,  $X_n \nearrow X$  and  $\mathbb{E}[X] < \infty$ . Then,  $\mathbb{E}[X_n | \mathcal{F}] \nearrow \mathbb{E}[X | \mathcal{F}]$   $\mathbb{P}$ -a.s.

HINT: First, prove the convergence in  $L^1(\mathbb{P})$ .

### Exercise 4

(4 points)

Let  $(X_n)_{n \in \mathbb{N}}$  be a martingale and  $p > 1$ . Prove that

$$\mathbb{E} \left[ \sup_{k \leq n} |X_k|^p \right] \leq \left( \frac{p}{p-1} \right)^p \mathbb{E}[|X_n|^p].$$

HINT: Use Doob's maximal inequality and Hölder's inequality (Analysis III, also true for the special case of probability measures) to prove

$$\mathbb{E}[(|X_n|^* \wedge K)^p] = \mathbb{E} \left[ \int_0^{|X_n|^* \wedge K} p \lambda^{p-1} d\lambda \right] \leq \frac{p}{p-1} \mathbb{E}[(|X_n|^* \wedge K)^p]^{\frac{p-1}{p}} \mathbb{E}[|X_n|^p]^{\frac{1}{p}}.$$

Then consider  $K \rightarrow \infty$ .

### Exercises for self-monitoring

- (1) Recall *Doob's maximal inequality*.
- (2) Recall *Kolmogorov's inequality* and prove it.
- (3) Recall the *upcrossing inequality*.
- (4) Show  $\mathbb{E}[Y_{\tau-1}] = -\infty$  for  $Y$  and  $\tau$  given in Example 5.14 (doubling strategy).
- (5) Let  $(S_n)_{n \in \mathbb{N}}$  be the simple random walk given in Exercise 4 on Sheet 7. Prove that

$$\mathbb{P} \left( \sup_{k \leq n} |S_k| \geq \sqrt{nt} \right) \leq \frac{1}{t^2}.$$