

PROBABILITY THEORY II - EXERCISE SET VI

**Exercise 1.** Let  $X_n$  be a simple random walk on  $\mathbb{Z}$ , and define

$$\tau := \min\{n : X_n = -1\}.$$

- (1) Prove that  $e^{\theta X_n} (\cosh \theta)^{-n}$  is a martingale.
- (2) Deduce that, for every  $0 \leq \alpha < 1$ , we have  $\mathbb{E}\alpha^\tau = \frac{1 - \sqrt{1 - \alpha^2}}{\alpha}$ , that is,

$$\mathbb{P}(\tau = 2k - 1) = (-1)^{k+1} \cdot \frac{\frac{1}{2}(\frac{1}{2} - 1) \dots (\frac{1}{2} - k + 1)}{k!}.$$

**Exercise 2.** Let  $X_n$  be a simple random walk on  $\mathbb{Z}$ , and  $a, b \in \mathbb{N}$ , and denote  $\tau := \min\{n : X_n = -a \text{ or } X_n = b\}$ ,  $Y_n := X_{n \wedge \tau}$ . Prove that, conditionally on the event  $A = \{X_\tau = a\}$ ,  $Y_n$  is a Markov chain, that is,

$$\mathbb{P}(Y_n = x_n | Y_{n-1} = x_{n-1}, \dots, Y_1 = x_1, A) = \mathbb{P}(Y_n = x_n | Y_{n-1} = x_{n-1}, A).$$

Compute the transition probabilities of that chain.

**Exercise 3.** Let  $X_1, X_2, \dots$  be a sequence of i. i. d., random variables, such that each  $X_i$  is uniformly chosen from the 26 letters of the English alphabet. Define  $\tau$  to be the minimal  $n$  such that

$$(X_n, \dots, X_{n+10}) = (A, B, R, A, C, A, D, A, B, R, A).$$

- (1) At each time  $n$ , a new gambler arrives and bets 1 EUR on the event that  $X_n = A$ . If he loses, he leaves; if he wins, he receives 26 EUR, all of which he bets on the event that  $X_{n+1} = B$ . If he loses, he leaves; if he wins, he receives  $26^2$  EUR, all of which he bets on  $X_{n+2} = R$ , and so on. Prove that if  $Y_m$  is the total gain of all the gamblers by time  $m$ , then  $Y_m$  is a martingale.
- (2) Prove that

$$Y_{\tau+10} = 26^{11} + 26^4 + 26 - \tau - 10.$$

Deduce that

$$\mathbb{E}(\tau) = 26^{11} + 26^4 + 26 - 10.$$

**Exercise 4.** Alice and Bob are playing the following game: Bob names a string of three letters (e. g.,  $AAB$ ), after which Alice names another string of three letters (e. g.  $BAB$ ). Then, they sample independent random variables  $X_1, X_2, \dots$  with  $\mathbb{P}(X_i = A) = \mathbb{P}(X_i = B) = \frac{1}{2}$ ; Alice wins if her string occurs in the sequence  $X_1, X_2, \dots$  before Bob's. We say that Alice's string beats Bob's if the probability of Alice to win is greater than  $\frac{1}{2}$ .

- (1) Assume that the strings chosen by Alice and Bob are given. Explain how to compute  $\mathbb{P}(\text{Alice wins})$  in terms of  $\mathbb{P}(\tau_x < \tau_y)$ , where  $\tau_a := \min\{n : Y_n = a\}$  and  $Y_0, Y_1, \dots$  is a certain Markov chain with six states.
- (2) Prove that  $AAB$  beats  $ABB$ .
- (3) Prove that whatever Bob does, Alice can guarantee a win with probability strictly greater than  $\frac{1}{2}$ .

Let  $\mathcal{D}_k := \{[l2^{-k}, (l+1)2^{-k}), l = 0, \dots, 2^k - 1\}$  denote the set of dyadic intervals of rank  $k$  in  $[0; 1)$ . For  $x \in [0; 1)$ , let  $I_k(x)$  denote  $I \in \mathcal{D}_k$  such that  $x \in I$ . Given an integrable function  $f : [0; 1) \mapsto \mathbb{R}$ , denote by  $M_k f$  its dyadic averages:

$$(M_k(f))(x) = |I_k(x)|^{-1} \int_{I_k(x)} f(y) dy.$$

**Exercise 5.** Check that if an integrable function  $f$  is viewed as a random variable on the probability space  $[0; 1)$ , equipped with the Borel  $\sigma$ -algebra and the Lebesgue measure, then

$$M_k(f) = \mathbb{E}(f | \sigma(\mathcal{D}_k)).$$

Before doing the next exercises, consult Definition 4.8.9 and the statement of Theorem 4.8.10 in the lecture notes (to be covered in the next lecture).

**Exercise 6.** Define the *dyadic Hardy-Littlewood maximal operator*  $Af := \sup_k M_k(|f|)$ . Prove that there exists a constant  $C > 0$  such that

$$\int_0^1 (Af)^p \leq C \int_0^1 |f|^p.$$

**Exercise 7.** Prove the weak-type Hardy-Littlewood bound

$$a\lambda(\{\omega \in [0; 1] : (Af)(\omega) > a\}) \leq C \int_0^1 |f|,$$

where  $f : [0; 1] \mapsto \mathbb{R}$  is an integrable function.

**Exercise 8.** Prove the *dyadic Lebesgue differentiation theorem*: if  $f : [0; 1) \rightarrow \mathbb{R}$  is integrable, then

$$\lim_{k \rightarrow \infty} (M_k f)(x) \rightarrow f(x)$$

for almost every  $x$ .