Probability, homework 6.

Exercise 1. Let $(\Omega, \mathcal{A}, \mathbb{P})$ be such that Ω is countable and $\mathcal{A} = 2^{\Omega}$. Prove that almost sure convergence and convergence in probability are the same on this probability space.

Exercise 2. Let $(X_n)_{n\geq 1}$ be a sequence of random variables, with respective distributions being Gaussian, with respective mean $\mu_n \in \mathbb{R}$ and variance $\sigma_n^2 > 0$. Prove that if X_n converges in distribution, then μ_n and σ_n^2 need to converge, and identify the limiting random variable.

Exercise 3. Let $(X_n)_{n\geq 1}$ be a sequence of i.i.d. real random variables, with $\mathbb{E}(X_1) = 0$, $\operatorname{var}(X_1) = 1$. Let $S_n = X_1 + \cdots + X_n$.

- a) Prove that for any A > 0, $\mathbb{P}\left(\limsup_{n \to \infty} \frac{S_n}{\sqrt{n}} > A\right) > 0$.
- b) Prove that $\{\limsup_{n\to\infty} \frac{S_n}{\sqrt{n}} > A\} \in \bigcap_{n\geq 1} \sigma(X_i, i\geq n).$
- c) Deduce that $\mathbb{P}\left(\limsup_{n\to\infty}\frac{S_n}{\sqrt{n}}=+\infty\right)=1.$
- d) Prove that for any subsequence $(n_k)_{k\geq 1}$, $\mathbb{P}\left(\limsup_{k\to\infty}\frac{S_{n_k}}{\sqrt{n_k}}=+\infty\right)=1.$
- e) Prove that $(\frac{S_n}{\sqrt{n}})_{n\geq 1}$ does not converge in probability.

Exercise 4 Central value of the partial exponential function. Calculate

$$\lim_{n \to \infty} e^{-n} \sum_{k=0}^{n} \frac{n^k}{k!}$$

Hint: this is a probability related to a sum of independent Poisson random variables with parameter 1.

Exercise 5 The number of buses stopping till time t. Let $(X_n)_{n\geq 1}$ be i.i.d random variables on $(\Omega, \mathcal{A}, \mathbb{P})$, X_1 being an exponential random variable with parameter 1. Define $T_0 = 0$, $T_n = X_1 + \cdots + X_n$, and for any t > 0,

$$N_t = \max\{n \ge 0 \mid T_n \le t\}$$

a) For any $n \ge 1$, calculate the joint distribution of (T_1, \ldots, T_n) .

b) Deduce the distribution of N_t , for arbitrary t.

Exercise 6 Large deviations. Let $(X_n)_{n\geq 1}$ be i.i.d random variables on $(\Omega, \mathcal{A}, \mathbb{P})$, X_1 with mean μ , and

$$L(\lambda) = \begin{cases} \log \mathbb{E}(e^{\lambda X_1}) & \text{if } \mathbb{E}(e^{\lambda X_1}) < \infty, \\ +\infty & \text{otherwise,} \end{cases}$$

and $L^*(x) = \sup(x\lambda - L(\lambda) \mid \lambda \in \mathbb{R}).$

a) Check that for any $\lambda \in \mathbb{R}, L(\lambda) \ge \lambda \mu$.

b) Prove that for any $\alpha > 0$ and $n \ge 1$,

$$\mathbb{P}\left(\frac{X_1 + \dots + X_n}{n} - \mu \ge \alpha\right) \le e^{-nL^*(\mu + \alpha)}.$$

c) Prove that for any $\alpha > 0$ and $n \ge 1$,

$$\mathbb{P}\left(\left|\frac{X_1 + \dots + X_n}{n} - \mu\right| \ge \alpha\right) \le e^{-nL^*(\mu + \alpha)} + e^{-nL^*(\mu - \alpha)}.$$

d) Deduce the most general law of large numbers you can from the previous inequality. As a first step, you could for example calculate, L, L^* for ± 1 or Cauchy random variables and see what happens in the previous inequality.