

## Probability, homework 6, due October 25.

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

**Exercise 2.** Let  $(X_i)_{i \geq 1}$  be i.i.d. Gaussian with mean 1 and variance 3. Show that

$$\lim_{n \rightarrow \infty} \frac{X_1 + \dots + X_n}{X_1^2 + \dots + X_n^2} = \frac{1}{4} \text{ a.s.}$$

**Exercise 3.** Let  $f$  be a continuous function on  $[0, 1]$ . Calculate the asymptotics, as  $n \rightarrow \infty$ , of

$$\int_{[0,1]^n} f\left(\frac{x_1 + \dots + x_n}{n}\right) dx_1 \dots dx_n.$$

**Exercise 4.** The goal of this exercise is to prove that any function, continuous on an interval of  $\mathbb{R}$ , can be approximated by polynomials, arbitrarily close for the  $L^\infty$  norm (this is the Bernstein-Weierstrass theorem). Let  $f$  be a continuous function on  $[0, 1]$ . The  $n$ -th Bernstein polynomial is

$$B_n(x) = \sum_{k=0}^n \binom{n}{k} x^k (1-x)^{n-k} f\left(\frac{k}{n}\right).$$

a) Let  $S_n(x) = B^{(n,x)}/n$ , where  $B^{(n,x)}$  is a binomial random variable with parameters  $n$  and  $x$ :  $B^{(n,x)} = \sum_{\ell=1}^n X_\ell$  where the  $X_i$ 's are independent and  $\mathbb{P}(X_i = 1) = x$ ,  $\mathbb{P}(X_i = 0) = 1 - x$ . Prove that  $B_n(x) = \mathbb{E}(f(S_n(x)))$ .

b) Prove that  $\|B_n - f\|_{L^\infty([0,1])} \rightarrow 0$  as  $n \rightarrow \infty$ .

**Exercise 5.** Calculate

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

**Exercise 6.** Let  $\alpha > 0$  and, given  $(\Omega, \mathcal{A}, \mathbb{P})$ , let  $(X_n, n \geq 1)$  be a sequence of independent real random variables with law  $\mathbb{P}(X_n = 1) = \frac{1}{n^\alpha}$  and  $\mathbb{P}(X_n = 0) = 1 - \frac{1}{n^\alpha}$ . Prove that  $X_n \rightarrow 0$  in  $\mathcal{L}^1$ , but that almost surely

$$\limsup_{n \rightarrow \infty} X_n = \begin{cases} 1 & \text{if } \alpha \leq 1 \\ 0 & \text{if } \alpha > 1 \end{cases}.$$

**Exercise 7.** A sequence of random variables  $(X_i)_{i \geq 1}$  is said to be completely convergent to  $X$  if for any  $\varepsilon > 0$ , we have  $\sum_{i \geq 1} \mathbb{P}(|X_i - X| > \varepsilon) < \infty$ . Prove that complete convergence implies almost sure convergence.

**Exercise 8.** Let  $(X_n)_{n \geq 1}$  be a sequence of random variables, on the same probability space, with  $\mathbb{E}(X_\ell) = \mu$  for any  $\ell$ , and a weak correlation in the following sense:  $\text{Cov}(X_k, X_\ell) \leq f(|k - \ell|)$  for all indexes  $k, \ell$ , where the sequence  $(f(m))_{m \geq 0}$  converges to 0 as  $m \rightarrow \infty$ . Prove that  $(n^{-1} \sum_{k=1}^n X_k)_{n \geq 1}$  converges to  $\mu$  in  $L^2$ .

**Long problem.** The goal is to prove the Erdős-Kac theorem: if  $w(m)$  denotes the number of distinct prime factors of  $m$  and  $k$  is a random variable uniformly distributed on  $[[1, n]]$ , then the following convergence in distribution holds:

$$\frac{w(k) - \log \log n}{\sqrt{\log \log n}} \xrightarrow[n \rightarrow \infty]{} \mathcal{N}(0, 1).$$

(i) Prove that if  $(X_n)_{n \geq 1}$  converges in distribution to  $\mathcal{N}(0, 1)$  and  $\sup_{n \geq 1} \mathbb{E}[X_n^{2k}] < \infty$  for any  $k \in \mathbb{N}$ , then  $\lim_{n \rightarrow \infty} \mathbb{E}[X_n^k] = \mathbb{E}[\mathcal{N}(0, 1)^k]$  for any  $k \in \mathbb{N}$ .

(ii) Prove that for any  $x \in \mathbb{R}$  and  $d \geq 1$  we have

$$\left| e^{ix} - \sum_{\ell=0}^d \frac{(ix)^\ell}{\ell!} \right| \leq \frac{|x|^{d+1}}{(d+1)!}.$$

(iii) Assume that  $\lim_{n \rightarrow \infty} \mathbb{E}[X_n^k] = \mathbb{E}[\mathcal{N}(0,1)^k]$  for any  $k \in \mathbb{N}$ . Prove that  $X_n$  converges in distribution to  $X$ .

(iv) Let  $w_y(m)$  be the number of prime factors of  $m$  which are smaller than  $y$ . Let  $(B_p)_{p \text{ prime}}$  be independent random variables such that  $\mathbb{P}(B_p = 1) = 1 - \mathbb{P}(B_p = 0) = \frac{1}{p}$ ,  $W_y = \sum_{p \leq y} B_p$ ,  $\mu_y = \sum_{p \leq y} \frac{1}{p}$ ,  $\sigma_y^2 = \sum_{p \leq y} (\frac{1}{p} - \frac{1}{p^2})$ . Prove that if  $y = n^{o(1)}$ , then for any  $d \in \mathbb{N}$  we have

$$\mathbb{E} \left[ \left( \frac{w_y(k) - \mu_y}{\sigma_y} \right)^d \right] - \mathbb{E} \left[ \left( \frac{W_y - \mu_y}{\sigma_y} \right)^d \right] \rightarrow 0$$

as  $n \rightarrow \infty$ .

(v) Conclude.