## Probability Comprehensive Exam Spring 2024

Student Number:	
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*Instructions:* Complete 5 of the 8 problems, and **circle** their numbers below – the uncircled problems will **not** be graded.

 $1 \quad 2 \quad 3 \quad 4 \quad 5 \quad 6 \quad 7 \quad 8$ 

Write **only on the front side** of the solution pages. A **complete solution** of a problem is preferable to partial progress on several problems.

1. Let  $X = (X_1, \ldots, X_4)$  be a mean zero normal random vector in  $\mathbb{R}^4$  with covariance matrix  $\Sigma = (\sigma_{ij} : 1 \le i, j \le 4)$ . Show that

$$\mathbb{E}X_1 X_2 X_3 X_4 = \sigma_{12} \sigma_{34} + \sigma_{13} \sigma_{24} + \sigma_{14} \sigma_{23}.$$

2. Let  $X_1, \ldots, X_n, \ldots$  be i.i.d. exponential random variables with expectation 1 and let  $Y_1, \ldots, Y_n, \ldots$  be i.i.d. standard normal random variables. Show that

$$\max(X_1,\ldots,X_n) - \max(Y_1,\ldots,Y_n) \to \infty \text{ as } n \to \infty \text{ a.s.}$$

3. Suppose  $X_0 = 0$  and

$$X_n = aX_{n-1} + \xi_n, n \ge 1,$$

where  $a \in \mathbb{R}$  and  $\xi_1, \xi_2, \ldots$  are independent mean zero random variables. Let  $\mathcal{F}_n := \sigma(X_1, \ldots, X_n)$  be the  $\sigma$ -field generated by  $X_1, \ldots, X_n$ . Show that

$$\mathbb{E}(X_{n+1}|\mathcal{F}_n) = aX_n$$

4. Let  $f:[0,1] \mapsto \mathbb{R}$  be a twice continuously differentiable function. Show that

$$\lim_{n \to \infty} n \int_0^1 \cdots \int_0^1 \left[ f\left(\frac{x_1 + \dots + x_n}{n}\right) - f(1/2) \right] dx_1 \dots dx_n = \frac{f''(1/2)}{24}$$

5. (i) Let X be a non-zero random variable and let  $b_1, b_2 \in (0, +\infty)$  be such that  $b_1 X \stackrel{d}{=} b_2 X$  where d indicates equality in distribution. Show that  $b_1 = b_2$ .

(ii) Let X be a non-constant random variable and let  $b_1, b_2 \in (0, +\infty)$  and  $c_1, c_2 \in \mathbb{R}$  be such that  $b_1X + c_1 \stackrel{d}{=} b_2X + c_2$  where again d indicates equality in distribution. Show that  $b_1 = b_2$  and  $c_1 = c_2$ .

6. Let  $(\Omega, \mathcal{F}, \mathbb{P})$  be a probability space and let  $\mathcal{G}$  be a sub- $\sigma$ -field of  $\mathcal{F}$ . Two random variables X et Y are said to be independent conditionally to  $\mathcal{G}$  if for any non-negative measurable functions, f and g, one has:

$$\mathbb{E}[f(X)g(Y) | \mathcal{G}] = \mathbb{E}[f(X) | \mathcal{G}] \mathbb{E}[g(Y) | \mathcal{G}].$$
 (C<sub>1</sub>)

(i) What is the meaning of  $(C_1)$  if  $\mathcal{G} = \{\emptyset, \Omega\}$ ? And if  $\mathcal{G} = \mathcal{F}$ ?

(ii) Show that the definition  $(C_1)$  is equivalent to the fact that for any non-negative random variable Z which is  $\mathcal{G}$ -mesurable, and for any non-negative measurable functions f and g one has

$$\mathbb{E}[f(X)g(Y)Z] = \mathbb{E}[f(X)Z\mathbb{E}[g(Y)|\mathcal{G}]].$$
(C<sub>2</sub>)

7. Let D be a domain of  $\mathbb{R}^d$  and let f be a measurable real-valued function defined on  $\mathbb{R}^d$ such that  $\mathbf{1}_D f$  is Lebesgue integrable. Let  $(U_n)_{n\geq 1}$  be a sequence of iid (independent and identically distributed) random variables uniformly distributed on (0, 1). Then, for each n = 1, 2, 3, ..., let  $\mathbf{V}_n$  be the random vector defined by  $\mathbf{V}_n = (U_{nd+1}, U_{nd+2}, \ldots, U_{(n+1)d})$ , and let  $X_n$  be the random variable defined by  $X_n = (\mathbf{1}_D f)(\mathbf{V}_n)$ .

(i) Show that the sequence  $(S_n)_{n\geq 1}$  given by  $S_n = \sum_{k=1}^n X_k/n$  converges almost surely towards a limit L that you will identify.

(ii) Let now f be bounded by M > 0, show that for any  $\lambda > 0$ ,

$$\mathbb{P}(|S_n - L| \ge \lambda) \le \frac{M^2}{n\lambda^2}.$$

8. Let  $(a_n)_{n\geq 1}$  and  $(b_n)_{n\geq 1}$  be two sequences of reals such that  $\lim_{n\to+\infty} a_n = a$  and  $\lim_{n\to+\infty} b_n = b$ , with a < b. Let  $(X_n)_{n\geq 1}$  be a sequence of random variables converging in distribution to a random variable  $X_{\infty}$  having a continuous distribution function  $F_{\infty}$ . Prove that

$$\lim_{n \to +\infty} \mathbb{P}(X_n \in [a_n, b_n]) = \mathbb{P}(X_\infty \in [a, b]).$$