

Preliminary Exam: Probability, August 2025.

Modality: In-person.

Time: 10:00am - 3:00pm, Friday, August 22, 2025.

Place: A102 Wells Hall.

Your goal should be to demonstrate mastery of probability theory and maturity of thought. Your arguments should be clear, careful and complete.

The exam consists of 6 main problems, each with several steps designed to help you with the overall solution.

**Important: If you cannot solve a certain part of a problem, you still may use its conclusion in a later part!**

Please make sure to apply the following guidelines:

1. On each page you turn in, write your assigned code number. Don't write your name on any page.
2. Start each problem on a new page.

**Problem 1.**

Let  $N \sim \text{Poisson}(\lambda), \lambda > 0$ .



- a. Show the steps to calculate  $\varphi_N(t) = E(e^{itN}), t \in \mathcal{R}$ .
- b. (i) Let  $\{W, W_k\}_{k=1,2,\dots}$  be i.i.d. and assume that  $\{W_k\}$  and  $N$  are independent. Let  $X = \sum_{k=0}^N W_k, X = 0$  if  $N = 0$ . Calculate  $E(e^{itX} | \sigma\{N\})$ .  
(ii) Calculate  $\varphi_X(t) = E(e^{itX}), t \in \mathcal{R}$ .
- c. For each  $n = 1, 2, \dots : N_n \sim \text{Poisson}(n)$ ,  $\{W_{n,k}\}_{k \geq 1}$  are i.i.d, and assume that  $N_n$  and  $\{W_{n,k}\}_{k \geq 1}$  are independent. The distribution of  $W_{n,1}$  is given by  $P(W_{n,1} = \frac{1}{\sqrt{n}}) = \frac{1}{2} = P(W_{n,1} = -\frac{1}{\sqrt{n}})$ . Finally, let  $X_n = \sum_{k=0}^{N_n} W_{n,k}$ .  
(i) Calculate  $\varphi_{X_n}(t) = E(e^{itX_n}), t \in \mathcal{R}$ .  
(ii) Prove that  $X_n$  converge in distribution as  $n \rightarrow \infty$ , and identify the limit distribution.

**Problem 2.**

- a. (i) Let  $X$  be a symmetric random variable (i.e.  $X = -X$  in distribution) with  $P(X = 0) = 0$ .  
 Prove that  $X = \epsilon \cdot |X|$ , a.s., where  $|X|, \epsilon$  are independent, and  $P(\epsilon = 1) = \frac{1}{2} = P(\epsilon = -1)$ .
- (ii) Let  $X, Y$  be symmetric, independent, and squared integrable random variables with  $P(X = 0) = P(Y = 0) = 0$ . Let  $\mathcal{H}$  be the  $\sigma$ -algebra  $\mathcal{H} = \sigma\{|X|, |Y|\} = \sigma\{X^2, Y^2\}$ . Prove that  $E_{\mathcal{H}}(XY) = 0$ . Also, calculate  $E_{\mathcal{H}}[(X + Y)^2]$ .
- b. Let  $\{B_t: t \geq 0\}$  is standard Brownian motion. Let  $Q_n = \sum_{k=1}^n D_{k,n}^2$ , where  $D_{k,n} = B_{t_{k,n}} - B_{t_{k-1,n}}$ ,  $\Delta_{n,k} = t_{k,n} - t_{k-1,n}$ ,  $1 \leq k \leq n$ , and  $0 = t_{0,n} < t_{1,n} < \dots < t_{n,n} = 1$ ,  $n = 1, 2, \dots$
- (i) Calculate:  $E(Q_n)$  and  $Var(Q_n)$ .
- (ii) Denote  $\Delta_n = \max_{1 \leq k \leq n} \{\Delta_{k,n}\}$ . Prove that:  
 If  $\Delta_n \xrightarrow{n \rightarrow \infty} 0$  then  $Q_n \xrightarrow{n \rightarrow \infty} 1$  in probability.

- c. We continue with the setup of part b. Assume that  $\{t_{k,n}\}_{0 \leq k \leq n} \subset \{t_{k,n+1}\}_{0 \leq k \leq n+1}$ ,  $n = 1, 2, \dots$ , where  $\{t_{k,n+1}\}_{0 \leq k \leq n+1}$  is a refinement of  $\{t_{k,n}\}_{0 \leq k \leq n}$  by an addition of one point.

Let  $\{\mathcal{H}_n\}_{n=1,2,\dots}$  be a filtration defined by  $\mathcal{H}_n = \sigma(\bigcup_{m=n}^{\infty} \{D_{k,m}^2, 1 \leq k \leq m\})$ . Observe that the sequence of  $\sigma$ -algebras  $\{\mathcal{H}_n\}_{n=1,2,\dots}$  is decreasing in  $n$ .

- (i) Find  $E_{\mathcal{H}_{n+1}}(Q_n)$ . Which type of process is  $\{Q_n, \mathcal{H}_n\}, n \geq 1$ ?

**Hint:** The assumption  $\{t_{k,n}\}_{0 \leq k \leq n} \subset \{t_{k,n+1}\}_{0 \leq k \leq n+1}$  implies that all the intervals

$\{(t_{k-1,n}, t_{k,n}): 1 \leq k \leq n\}$  are contained in  $\{(t_{k-1,n+1}, t_{k,n+1}): 1 \leq k \leq n+1\}$  except one denoted by  $(t_{k_*,n}, t_{k_*,n+1}], 0 \leq k_* \leq n-1$  that satisfies

$$(t_{k_*,n}, t_{k_*,n+1}] = (t_{k_*,n+1}, t_{k_*,n+1,n+1}] \cup (t_{k_*,n+1,n+1}, t_{k_*,n+2,n+1}].$$

- (ii) Prove that if  $\Delta_n \xrightarrow{n \rightarrow \infty} 0$  then  $Q_n \xrightarrow{n \rightarrow \infty} 1$ , almost surely.

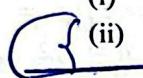
### Problem 3

Let  $(X, Y) \in \mathbb{R}^2$  be a random vector. The characteristic function (c.f.) of  $(X, Y)$  is defined as

$\varphi_{(X,Y)}(s, t) = E(e^{i(sX+tY)})$ ,  $(s, t) \in \mathbb{R}^2$ . It is known that if the characteristics functions of the two random vectors  $(X, Y)$  and  $(U, V)$  are identical then  $(X, Y) = (U, V)$  in distribution, where  $(U, V) \in \mathbb{R}^2$

- a. Let  $\varphi_X(s), \varphi_Y(t), s, t \in \mathbb{R}$  be the cf. of  $X, Y$ , respectively. Prove that

(i) If  $X, Y$  are independent, then  $\varphi_{(X,Y)}(s, t) = \varphi_X(s) \cdot \varphi_Y(t), s, t \in \mathbb{R}$ .



(ii) If  $\varphi_{(X,Y)}(s, t) = \varphi_X(s) \cdot \varphi_Y(t), s, t \in \mathbb{R}$  then  $X, Y$  are independent.

- b. Let  $X \in \mathbb{R}^3$  satisfy  $X = \sum_{m=1}^n X_m$ , where  $\{X_m \in \mathbb{R}^3 : m = 1, \dots, n\}$  are i.i.d. The distribution of  $X_1 \in \mathbb{R}^3$  is given by  $P(X_1 = \varepsilon_k) = p_k, k = 1, 2, 3$ , where  $\sum_{k=1}^3 p_k = 1, p_k \geq 0, \varepsilon_1 = (1, 0, 0), \varepsilon_2 = (0, 1, 0)$ , and  $\varepsilon_3 = (0, 0, 0)$ .

(i) Find the c.f. of  $X_1$ , namely  $\varphi_{X_1}(t) = E(e^{it \cdot X_1})$ ,  $t = (t_1, t_2, t_3) \in \mathbb{R}^3$ ,

$X_1 = (X_{1,1}, X_{1,2}, X_{1,3})$  and  $t \cdot X_1 = \sum_{k=1}^3 t_k X_{1,k}$ ,

where  $X_{1,k} \in \mathbb{R}, k = 1, 2, 3$ .

Also, how is  $X_{1,3}$  distributed?

(ii) Find the c.f. of  $X$ , namely  $\varphi_X(t) = E(e^{it \cdot X})$ ,  $t \in \mathbb{R}^d$ .

- c. Let  $X_n \in \mathbb{R}^3$  satisfy  $X_n = \sum_{m=1}^n X_{n,m}$ ,  $n = 1, 2, \dots$  where  $\{X_{n,m} \in \mathbb{R}^3 : m = 1, \dots, n\}$  are i.i.d. The distribution of  $X_{n,1} \in \mathbb{R}^3$  is given by  $P(X_{n,1} = \varepsilon_k) = p_{n,k}, k = 1, 2, 3$ , where  $\sum_{k=1}^3 p_{n,k} = 1, p_{n,k} \geq 0, \{\varepsilon_k, k = 1, 2, 3\}$  are as in part b.

Assumption:  $n \cdot p_{n,k} \xrightarrow{n \rightarrow \infty} \lambda_k$  and  $0 < \lambda_k < \infty, k = 1, 2$ .

**Hint:** The assumption does not hold for  $k = 3$ .

(i) Prove that  $\varphi_{X_n}(t) \xrightarrow{n \rightarrow \infty} \varphi_{X_\infty}(t), t \in \mathbb{R}^3$ , where  $X_\infty \in \mathbb{R}^3$  is a random vector.

Hint.  $\sum_{k=1}^3 a_k p_{n,k} = 1 + \sum_{k=1}^2 (a_k - 1)p_{n,k}$  if  $a_3 = 1$ .

(ii) What is the relationship between the first 2 coordinates of  $X_\infty$ ? What is the distribution of each of the 3 coordinates of  $X_\infty$ ? Justify your answer.

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### Problem 4

Let  $X$  be a random variable that satisfy  $\sum_{k=1}^{\infty} P(|X| \geq a_k) < \infty$ , where  $\{a_k\}_{k \geq 1}$  is a non-negative and non-decreasing sequence of real numbers that satisfy

- (i)  $a_k \xrightarrow{k \rightarrow \infty} \infty$ , and
- (ii) there exists  $C < \infty$  so that for each  $n \geq 1$ ,  $\sum_{k=n}^{\infty} a_k^{-2} \leq C n a_n^{-2}$ .

Let  $\{X, X_k\}_{k \geq 1}$  be an i.i.d. sequence of random variables, and denote

$Y_k = X_k \cdot 1_{\{|X_k| \leq a_k\}}$ ,  $k \geq 1$ . Prove the following:

- a. (i)  $\frac{\sum_{k=1}^n X_k \cdot 1_{\{|X_k| > a_k\}}}{a_n} \xrightarrow{n \rightarrow \infty} 0$ , a.s.
- (ii) If  $\frac{\sum_{k=1}^n Y_k - E(Y_k)}{a_n} \xrightarrow{n \rightarrow \infty} 0$ , a.s., then  $\frac{\sum_{k=1}^n X_k - E(Y_k)}{a_n} \xrightarrow{n \rightarrow \infty} 0$ , a.s.
- b. (i) Let  $a_0 = 0$ . Prove:  $\sum_{n=1}^{\infty} a_n^{-2} E(Y_n^2) = \sum_{k=1}^{\infty} E(X^2 \cdot 1_{\{a_{k-1} < |X| \leq a_k\}}) \cdot (\sum_{n=k}^{\infty} a_n^{-2})$ .  
Hint:  $(0, a_n] = \bigcup_{k=1}^n (a_{k-1}, a_k]$
- (ii) Use the result of part (i) to prove that  $\sum_{n=1}^{\infty} a_n^{-2} E(Y_n^2) \leq C \sum_{k=1}^{\infty} k P(a_{k-1} < |X| \leq a_k)$
- c. (i)  $\sum_{k=1}^{\infty} \frac{Y_k - E(Y_k)}{a_k}$  converge a.s.

Hint:  $\sum_{k=1}^{\infty} k P(a_{k-1} < |X| \leq a_k) = \sum_{k=0}^{\infty} P(|X| > a_k)$  by summation by parts.

$$(ii) \quad \frac{\sum_{k=1}^n X_k - E(Y_k)}{a_n} \xrightarrow{n \rightarrow \infty} 0, \text{ a.s.}$$

$\Rightarrow \sum_{k=1}^n X_k - E(Y_k) \xrightarrow{n \rightarrow \infty} 0$   
 $\Rightarrow \sum_{k=1}^n X_k - \sum_{k=1}^n E(Y_k) \xrightarrow{n \rightarrow \infty} 0$   
 $\Rightarrow \sum_{k=1}^n X_k - \sum_{k=1}^n E(X_k \cdot 1_{\{|X_k| \leq a_k\}}) \xrightarrow{n \rightarrow \infty} 0$   
 $\Rightarrow \sum_{k=1}^n X_k - \sum_{k=1}^n E(X_k) + \sum_{k=1}^n E(X_k \cdot 1_{\{|X_k| > a_k\}}) \xrightarrow{n \rightarrow \infty} 0$   
 $\Rightarrow \sum_{k=1}^n E(X_k \cdot 1_{\{|X_k| > a_k\}}) \xrightarrow{n \rightarrow \infty} 0$   
 $\Rightarrow \sum_{k=1}^{\infty} E(X_k \cdot 1_{\{|X_k| > a_k\}}) < \infty$

Since  $E(X_k \cdot 1_{\{|X_k| > a_k\}}) = P(|X_k| > a_k) E(X_k)$  and  $P(|X_k| > a_k) \leq 1/a_k$ , we have

$$E(X_k \cdot 1_{\{|X_k| > a_k\}}) = P(|X_k| > a_k) E(X_k) \leq P(|X_k| > a_k) \cdot a_k = P(|X_k| > a_k) a_k$$

Since  $\sum_{k=1}^{\infty} a_k^{-2} E(Y_k^2) = \sum_{k=1}^{\infty} E(X_k^2 \cdot 1_{\{a_{k-1} < |X_k| \leq a_k\}}) \cdot (\sum_{n=k}^{\infty} a_n^{-2})$ , we have

$$E(X_k^2 \cdot 1_{\{a_{k-1} < |X_k| \leq a_k\}}) = P(|X_k| > a_k) E(X_k^2)$$

$$P(|X_k| > a_k) E(X_k^2) = P(|X_k| > a_k) E(X_k^2) + P(|X_k| > a_k) a_k^2$$

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Problem 5.

Let  $\{X_k, \mathcal{F}_k\}_{k=0,1,\dots}$  be a supermartingale sequence of random variables where  $X_0 = 1$ . Next, we define a sequence of stopping times:

$$T_{-1} = T_0 = 0, T_1 = \inf\{m > 0 : X_m \leq 0\},$$

$$T_{2k} = \inf\{m > T_{2k-1} : X_m \geq 1\}, k \geq 1, \text{ and}$$

$$T_{2k+1} = \inf\{m > T_{2k} : X_m \leq 0\}, k \geq 1.$$

Each interval  $[T_{2k}, T_{2k+1}], k \geq 0$  represents a down-crossing of the interval  $[0, 1]$  by  $\{X_k\}_{k=0,1,\dots}$  as  $X_{T_{2k}} \geq 1, X_m > 0$  if  $T_{2k} < m < T_{2k+1}$ , and  $X_{T_{2k+1}} \leq 0$ .

For each  $n \geq 0$  let  $D_n(X)$  denotes the number of down-crossings of the interval  $[0, 1]$  by  $\{X_k\}, 0 \leq k \leq n$ . A formal definition of  $D_n(X)$  is

$$D_n(X) = \max\{k \geq 1 : T_{2k-1} \leq n\}, (D_n(X) = 0 \text{ if no such } k \text{ exists.})$$

Prove:

- a. (i) Let  $Y_k = \min\{X_k, 1\}, k \geq 0$ . Then  $\{Y_k, \mathcal{F}_k\}_{k=0,1,\dots}$  is a supermartingale, and  $D_n(X) = D_n(Y)$ , a.s. where  $D_n(Y)$  denotes the number of down-crossings of the interval  $[0, 1]$  by  $\{Y_k\}, 0 \leq k \leq n$ .

**For the rest of the problem, we assume that  $X_k = Y_k$ , namely  $X_k \leq 1, k \geq 0$ .**

- (ii) Is  $T_{2D_n-1}$  a stopping time? explain.
- b. For  $a, b$  integers denote  $[a, b] = \{a, a+1, \dots, b-1\}, a < b$ . For each  $m \geq 1, \omega \in \Omega$  we define the random variable  $H_m$  by  

$$\begin{cases} H_m(\omega) = 1, & \text{if there is } k \geq 0 \text{ so that } m-1 \in [T_{2k}(\omega), T_{2k+1}(\omega)). \\ H_m(\omega) = 0, & \text{otherwise.} \end{cases}$$

Prove the following:

- (i)  $\{H_k, \mathcal{F}_k\}_{k \geq 1}$  is predictable, i.e.  $H_m \in \mathcal{F}_{m-1}, m \geq 1$ .

Hint: Prove first that the event  $\{m-1 \in [T_{2k}, T_{2k+1}]\}$  belongs to  $\mathcal{F}_{m-1}$  for each  $k \geq 0$ .

- (ii) The gambling systems  $(H \cdot X)_{n \geq 0}$  and  $((1-H) \cdot X)_{n \geq 0}$  are both supermartingales. Recall that a gambling system is defined by  $(H \cdot X)_n = \sum_{m=1}^n H_m (X_m - X_{m-1}), n \geq 1$ , and  $(H \cdot X)_0 = 0$ .

- c. (i)  $-D_n \geq (H \cdot X)_{T_{2D_n-1}}$ , a.s. Also,  $(H \cdot X)_{T_{2D_n-1}} \geq (H \cdot X)_n$ , a.s.

- (ii)  $E(D_n) \leq E(X_0 - X_n) = 1 - E(X_n)$ .

Hint. Find the smallest upper bound that you can for  $E((1-H) \cdot X)_n$ . Also, look at  $E[((1-H) \cdot X)_n] + E[(H \cdot X)_n]$ .

**Problem 6.**

Let  $\{X_k\}_{k=1,2,\dots}$  be uncorrelated random variables. Also, there exists a constant

$0 < M < \infty$  so that  $0 < X_k < M$  a.s.,  $k \geq 1$ . Assume that  $\sum_{k=1}^{\infty} E(X_k) = \infty$ .

Denote  $S_n = \sum_{k=1}^n X_k$ ,  $n \geq 1$ . The goal of the problem is to prove that

$\frac{S_n}{E(S_n)} \xrightarrow{n \rightarrow \infty} 1$ , a.s. by proving the following steps:

- a. (i) If the result holds for  $M = 1$ , then the result holds for any  $0 < M < \infty$ .

**For the rest of the problem, we assume without loss of generality that  $M = 1$ .**

(ii)  $Var(S_n) \leq E(S_n)$ ,  $n \geq 1$ . Use this to show that  $\frac{S_n}{E(S_n)} \xrightarrow{n \rightarrow \infty} 1$  in probability.

- b. Let  $\{a_k\}_{k=1,2,\dots}$  be a strictly increasing sequence of positive integers so that  $a_k \xrightarrow{k \rightarrow \infty} \infty$ , and

$\frac{a_{k+1}}{a_k} \xrightarrow{k \rightarrow \infty} 1$ . Let  $\{n_k\}_{k=1,2,\dots}$  be a strictly increasing sequence of positive integers so that

$$a_k \leq E(S_{n_k}) \leq 1 + a_k, k \geq 1.$$

If  $\frac{S_{n_k}}{E(S_{n_k})} \xrightarrow{k \rightarrow \infty} 1$ , a.s. then  $\frac{S_n}{E(S_n)} \xrightarrow{n \rightarrow \infty} 1$ , a.s.

Hint: show first that  $\frac{E(S_{n_{(k+1)}})}{E(S_{n_k})} \xrightarrow{k \rightarrow \infty} 1$ . Then work with  $n_k \leq n < n_{k+1}$ .

- c. (i) We continue with the notations of part b. Let  $\delta > 0$ . Then for each  $k \geq 1$

$$P(|S_{n_k} - E(S_{n_k})| > \delta(1 + a_k)) \leq \frac{1}{\delta^2(1+a_k)}$$

- (ii) Let  $a_k = k^3$ ,  $k \geq 1$ . Prove that  $\frac{S_n}{E(S_n)} \xrightarrow{n \rightarrow \infty} 1$ , a.s.