

**Qn 1**

- (a) Define a Sigma-algebra (2 points)
- (b) Let  $\mathcal{A}$  and  $\mathcal{B}$  belong to a sigma-algebra  $\mathcal{F}$ . Use the definition above to show that  $A \cap B^c \in \mathcal{F}$ . (3 points)
- (c) Let  $\mathcal{F}$  be a sigma-algebra of subsets of  $\Omega$  and suppose  $B \in \mathcal{F}$ . Show that  $G = \{A \cap B : A \in \mathcal{F}\}$  is a sigma-algebra of subsets of  $\Omega$ . (5 points)

**Solution.**

- (a) A sigma-algebra, denoted by  $\mathcal{F}$ , on a set  $\Omega$  is a collection of subsets of  $\Omega$  that satisfies the following three properties:
- The empty set  $\emptyset$  is in  $\mathcal{F}$ .
  - If  $A$  is in  $\mathcal{F}$ , then its complement  $A^c$  is also in  $\mathcal{F}$ .
  - If  $A_1, A_2, A_3, \dots$  is a countable sequence of sets in  $\mathcal{F}$ , then their union  $\bigcup_{i=1}^{\infty} A_i$  is also in  $\mathcal{F}$ .
- (b) We need to prove  $A \cap B^c \in \mathcal{F}$
- As  $A \in \mathcal{F}$ , then by (property 2),  $A^c$  is also in  $\mathcal{F}$ .  $B \in \mathcal{F}$  implies,  $A^c \cup B \in \mathcal{F}$  by (property 3).
  - As  $A^c \cup B \in \mathcal{F}$ , then  $(A^c \cup B)^c \in \mathcal{F}$  by (property 2).
  - By DeMorgan's laws,  $(A^c \cup B)^c = A \cap B^c$ .
  - $(A \cap B^c) \in \mathcal{F}$

- (c) Let  $\mathcal{F}$  be a sigma-algebra of subsets of  $\Omega$ , and suppose  $B \in \mathcal{F}$ . We want to show that  $G = \{A \cap B : A \in \mathcal{F}\}$  is a sigma-algebra of subsets of  $\Omega$ .

1. **Contains the Empty Set:**

$$\emptyset = \emptyset \cap B$$

Since  $\emptyset \in \mathcal{F}$  (property 1 of sigma-algebra),  $\emptyset \cap B$  is in  $G$ .

2. **Closed under Complement:** Let  $E$  be in  $G$ , so  $E = A \cap B$  for some  $A$  in  $\mathcal{F}$ . Then, the complement of  $E$  is given by:

$$E^c = (A \cap B)^c = A^c \cup B^c$$

Since  $A^c \in \mathcal{F}$  (property 2 of sigma-algebra) and  $B^c \in \mathcal{F}$  (as  $B \in \mathcal{F}$ ),  $E^c$  is in  $G$ .

3. **Closed under Countable Unions:** Let  $E_1, E_2, \dots$  be a countable sequence in  $G$ . So,  $E_i = A_i \cap B$  for each  $i$ , where  $A_i \in \mathcal{F}$ .

$$\bigcup_{i=1}^{\infty} E_i = \bigcup_{i=1}^{\infty} (A_i \cap B) = \left( \bigcup_{i=1}^{\infty} A_i \right) \cap B$$

Since  $\bigcup_{i=1}^{\infty} A_i \in \mathcal{F}$  (property 3 of sigma-algebra),  $\bigcup_{i=1}^{\infty} E_i$  is in  $G$ .

Therefore,  $G$  is a sigma-algebra of subsets of  $\Omega$ .

**2(a):** Suppose,  $i < j$  and  $m < n$ . If  $j < m$ , then  $A_{ij}$  and  $A_{mn}$  are determined by distinct independent rolls, and are therefore independent. For the case  $j = m$  we have that,

$$\begin{aligned}\mathbb{P}(A_{ij} \cap A_{jn}) &= \mathbb{P}(i^{th}, j^{th} \text{ and } n^{th} \text{ rolls show same number}), \\ &= \sum_{r=1}^6 \frac{1}{6} \mathbb{P}(j^{th} \text{ and } n^{th} \text{ show } r | i^{th} \text{ shows } r) = \frac{1}{36} = \mathbb{P}(A_{ij})\mathbb{P}(A_{jn}),\end{aligned}$$

[2 Marks]

as required. However, if  $i \neq j \neq k$ ,

$$\mathbb{P}(A_{ij} \cap A_{jk} \cap A_{ik}) = \frac{1}{36} \neq \frac{1}{216} = \mathbb{P}(A_{ij})\mathbb{P}(A_{jk})\mathbb{P}(A_{ik}).$$

[3 Marks]

**2(b):**

$$\mathbb{P}(1^{\text{st}} \text{ shows } r \text{ and sum is } 7) = \frac{1}{36} = \mathbb{P}(1^{\text{st}} \text{ shows } r)\mathbb{P}(\text{sum is } 7).$$

[5 Marks]

**Q3.** A dart is thrown by a player at a circular target of a radius of three units. Assume that the dart is guaranteed to hit the circular target. Further, assume that the probability that the dart lands in a region of the target is proportional to the area of that region.

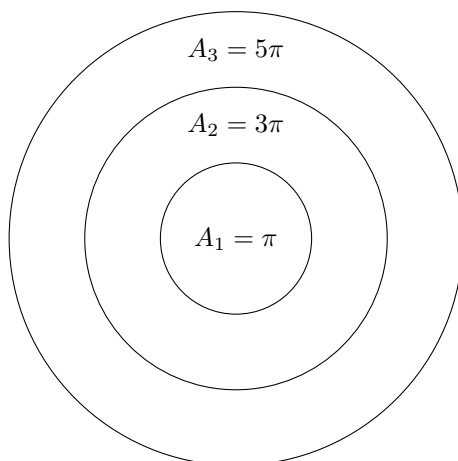
- (a) The target is partitioned into three concentric annuli  $A_1$ ,  $A_2$ , and  $A_3$ , where

$$A_k = \{(x, y) : k - 1 \leq \sqrt{x^2 + y^2} < k\}$$

The player scores an amount  $k$  if and only if the dart hits  $A_k$ . Let  $X$  be the random variable denoting the score. Characterize and plot the CDF of  $X$ .

- (b) Let us revise the scoring method so that the player gets an amount equal to the distance between the hitting point and the center of the target. Let  $Y$  be the random variable denoting the score. Characterize and plot the CDF of  $Y$ .

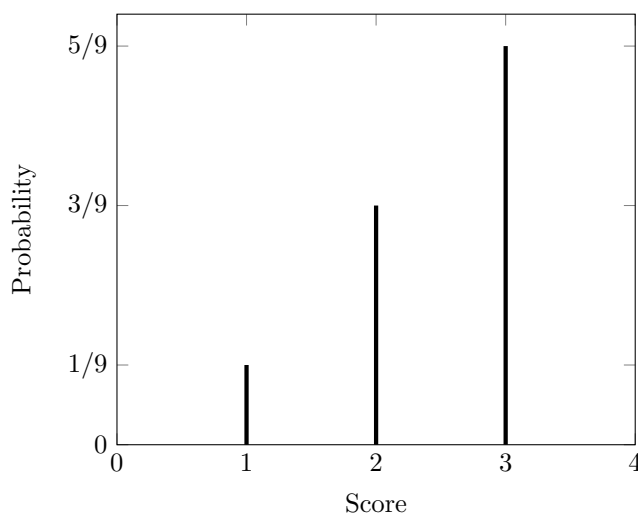
**Ans a)**



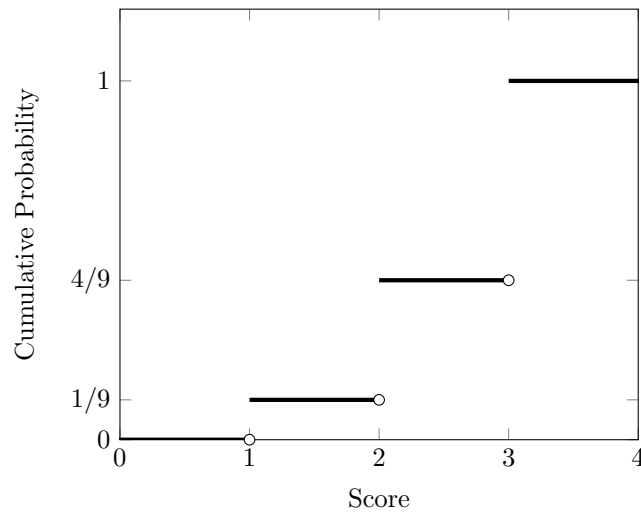
Since the probability of the dart landing in a region is proportional to the target, we have

$$\begin{aligned} P(\text{Dart in } A_k) &\propto \text{Area of } A_k \\ \implies P(\text{Dart in } A_1) &\propto \text{Area of } A_1 = \pi * 1^2 = \pi \\ P(\text{Dart in } A_2) &\propto \text{Area of } A_2 = \pi * 2^2 - \pi * 1^2 = 3 * \pi \\ P(\text{Dart in } A_3) &\propto \text{Area of } A_3 = \pi * 3^2 - \pi * 2^2 = 5 * \pi \end{aligned}$$

Also, total probability should be 1, which gives  $P(\text{Dart in } A_1) = \frac{1}{9}$ ,  $P(\text{Dart in } A_2) = \frac{3}{9}$ ,  $P(\text{Dart in } A_3) = \frac{5}{9}$ . Score is equal to the annuli that the dart landed on, hence  $P(X = 1) = \frac{1}{9}$ ,  $P(X = 2) = \frac{3}{9}$ ,  $P(X = 3) = \frac{5}{9}$ . PMF of the above function is:



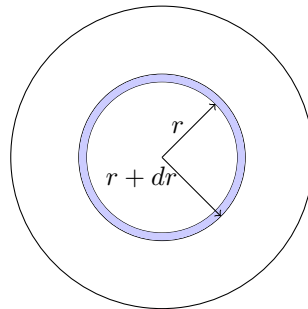
The CDF would be



Marking Scheme:

- 3 points for PDF calculation
  - 3 points for correct calculation of probabilities/PDF/CDF
  - 2 points if there is minor calculation mistake or in defining probability
  - 2 points if proportionality constant not calculated
  - 0 points otherwise
- 2 points for CDF plot
  - 2 points for correct CDF plot
  - 1 point if PDF is plot instead
  - 1 point if Score > 3 is missing
  - 1 point if labels on the CDF graph is incorrect
  - 1 point if lines/shape is incorrect

**Ans b)** Now, the score is defined to be the distance between the centre and the dart. The probability to receive a score  $r$  is the same as landing in a thin circle of radius ' $r$ ', as shown in the shaded region.



This implies  $P(Y = r) \propto 2\pi r = 2\pi r k$ . Since the score is now continuous, we have

$$\int_0^3 P(Y = r) dr = 1 \implies \pi(3^2)k = 1 \implies k = \frac{1}{9\pi}$$

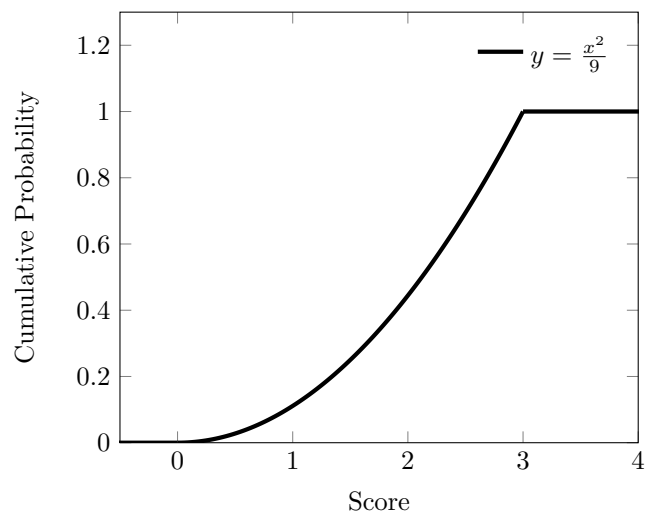
The PDF of the score is  $P(Y = r) = \frac{2r}{9}$ . The CDF of the score would be

$$P(Y \leq r) = \int_0^r (P(Y = x) dx) = \frac{r^2}{9} \text{ for } 0 \leq r \leq 3$$

$$P(Y \leq r) = 1 \text{ for } r \geq 3$$

$$P(Y \leq r) = 0 \text{ otherwise}$$

One could directly argue this CDF from the area of the circle of radius  $r$  and the area of the circle of radius 3. The plot of the Cdf is:



Marking Scheme:

- 3 points for PDF calculation
  - 3 points for correct calculation of probabilities/PDF/CDF
  - 2 points if there is minor calculation mistake but  $\text{CDF} \propto r^2$
  - 2 points if proportionality constant not calculated
  - 0 points otherwise
- 2 points for CDF plot
  - 2 points for correct CDF plot
  - 2 points for the CDF plotted of incorrect PDF calculated ( $\frac{x^3}{27}$  or linear)
  - 0 points for plotting CDF/PDF of previous question plotted
  - 1.5 point if shape is incorrect
  - 1 point is Score  $> 3$  is missing
  - 1 point if labels on the CDF graph is incorrect

4. Let  $\Omega = \{\omega_1, \omega_2, \omega_3\}$ , with  $P(\{\omega_1\}) = P(\{\omega_2\}) = P(\{\omega_3\}) = 1/3$ . Define random variables  $X, Y$ , and  $Z$  on  $\Omega$  as follows:

$$\begin{aligned} X(\omega_1) &= 1, X(\omega_2) = 2, X(\omega_3) = 3, \\ Y(\omega_1) &= 2, Y(\omega_2) = 3, Y(\omega_3) = 1, \\ Z(\omega_1) &= 2, Z(\omega_2) = 2, Z(\omega_3) = 1. \end{aligned}$$

- (a) Show that  $X$  and  $Y$  have the same probability mass functions. (2 points)
- (b) Find the probability mass function of  $U = X \times Y$ . (2 points)
- (c) Find  $p_{Y|Z}$ . (3 points)
- (d) Find  $p_{Z|Y}$ . (3 points)

## Solution

- (a) Clearly  $p_X(i) = p_Y(i) = \frac{1}{3}$  for  $i = 1, 2, 3$ , and other values are 0.
- (b)  $(XY)(\omega_1) = 2, (XY)(\omega_2) = 6, (XY)(\omega_3) = 3$ , and therefore  $p_{XY}(i) = \frac{1}{3}$  for  $i = 2, 3, 6$ , and other values are 0.
- (c)

$$p_{Y|Z}(2 | 2) = \frac{P(Y = 2, Z = 2)}{P(Z = 2)} = \frac{P(\omega_1)}{P(\omega_1 \cup \omega_2)} = \frac{1}{2},$$

and similarly  $p_{Y|Z}(3 | 2) = \frac{1}{2}, p_{Y|Z}(1 | 1) = 1$ , and other values are 0 .

- (d) Likewise  $p_{Z|Y}(2 | 2) = p_{Z|Y}(2 | 3) = p_{Z|Y}(1 | 1) = 1$ , and other values are 0 .

## Marking scheme:

2 points each for reaching the final answer in parts (a) and (b). In part B, no partial marks awarded if the sample space of  $U$  is misinterpreted. For parts (c) and (d) partial marks are awarded only if steps are detailed out correctly.

# EE325 Endsem Solutions

Autumn 2023

## 1 Question 5

### 1.1 Part a

Given  $X \geq_{st} Y$  and Theorem 1, we always have random variables  $X'$  and  $Y'$  satisfying  $F_X(x) = F_{X'}(x)$  and  $F_Y(y) = F_{Y'}(y)$  **on the same sample space** such that  $\mathbb{P}(X' \geq Y') = 1$ . (1 mark)

Also  $\mathbb{E}[X] = \mathbb{E}[X']$  and  $\mathbb{E}[Y] = \mathbb{E}[Y']$ . (1 mark)

The random variable  $X' - Y'$  takes non-negative values with probability 1. Hence  $\mathbb{E}[X' - Y'] \geq 0$  or  $\mathbb{E}[X] = \mathbb{E}[X'] \geq \mathbb{E}[Y'] = \mathbb{E}[Y]$  (3 marks)

Alternatively

One can write  $\mathbb{E}[X] = \int_0^\infty (1 - F_X(a))da - \int_{-\infty}^0 F_X(a)da$ , also  $\mathbb{E}[Y] = \int_0^\infty (1 - F_Y(a))da - \int_{-\infty}^0 F_Y(a)da$  if the expectations exist. (4 marks)

Given  $X \geq_{st} Y$  we have  $F_X(a) \leq F_Y(a) \forall a$ . Combining the two we have  $\mathbb{E}[X] \geq \mathbb{E}[Y]$ . (1 mark)

Using integral of CCDF is the expected value is only true for positive random variables - partial marks (3 marks)

### 1.2 Part b

Let  $P(X = a) = \frac{e^{-\lambda}\lambda^a}{a!}$  be denoted by  $f_X^\lambda(a)$  and  $F_X^\lambda(a) = \sum_{i=0}^a f_X^\lambda(i) = \sum_{i=0}^a \frac{e^{-\lambda}\lambda^i}{i!}$

Given  $\lambda \geq \mu$ , one can show  $\exists i \in \mathbb{N} \cup \{0\}$  such that  $f_X^\lambda(k) \leq f_Y^\mu(k) \forall k \leq i$  and  $f_X^\lambda(k) > f_Y^\mu(k) \forall k > i$ . (2 marks)

We clearly have  $F_X^\lambda(a) \leq F_Y^\mu(a) \forall a \leq i$ . Now assume  $F_X^\lambda(b) > F_Y^\mu(b)$  for some  $b > i$ . Then we have

$$1 = \sum_{i=0}^{\infty} f_X^\lambda(i) = F_X^\lambda(b) + \sum_{i=b+1}^{\infty} f_X^\lambda(i) > F_Y^\mu(b) + \sum_{i=b+1}^{\infty} f_Y^\mu(i) = \sum_{i=0}^{\infty} f_Y^\mu(i) = 1$$

which is a contradiction, hence  $F_X^\lambda(a) \leq F_Y^\mu(a) \forall a$  and thus  $X \geq_{st} Y$  (3 marks)

Any valid way that shows  $F_X^\lambda(a) \leq F_Y^\mu(a) \forall a$  will be awarded full marks (5 marks)

**6. Let  $X_1, X_2, X_3, \dots$  be a sequence of independent and identically distributed random variables with a moment generating function  $M_X(s)$ . Let  $N$  be a random variable independent of  $X_1, X_2, \dots$  with a moment generating function  $M_N(s)$ . Let  $S = X_1 + X_2 + \dots + X_N$ . Compute the moment generating function of  $S$  in terms of  $M_X(s)$  and  $M_N(s)$ .**

**Solution:** By the definition of a moment generating function we have

$$M_X(s) = \mathbb{E}(e^{X \cdot s})$$

$$M_N(s) = \mathbb{E}(e^{N \cdot s})$$

$$M_S(s) = \mathbb{E}(e^{S \cdot s}) \quad (2 \text{ marks})$$

$S$  depends on  $N$  conditionally which gives

$$\mathbb{E}(e^{S \cdot s}) = \mathbb{E}(\mathbb{E}(e^{S \cdot s} | N)) = \sum_n \mathbb{E}(e^{S \cdot s} | N = n) \mathbb{P}(N = n) \quad (2 \text{ marks})$$

This gives us

$$\begin{aligned} M_S(s) &= \sum_n \mathbb{E}(e^{S \cdot s} | N = n) \mathbb{P}(N = n) \\ &= \sum_n \mathbb{E}(e^{(X_1 + X_2 + \dots + X_n) \cdot s}) \mathbb{P}(N = n) \\ &= \sum_n \mathbb{E}(e^{X_1 \cdot s}) \dots \mathbb{E}(e^{X_n \cdot s}) \mathbb{P}(N = n) \\ &= \sum_n (M_X(s))^n \mathbb{P}(N = n) \\ &= \mathbb{E}(M_X(s)^N) \end{aligned} \quad (2 \text{ marks})$$

Now  $M_X(s)^N = e^{N \cdot \ln M_X(s)}$  which gives

$$\mathbb{E}(M_X(s)^N) = \mathbb{E}(e^{N \cdot \ln M_X(s)}) \quad (2 \text{ marks})$$

$$= M_N(\ln M_X(s)) \quad (2 \text{ marks})$$



**Q7**

- **Part a)** To show

$$|\mathbb{E}[XY]| \leq \sqrt{\mathbb{E}[X^2]\mathbb{E}[Y^2]}$$

Consider the linear combination  $Z = aX + bY$  where  $a, b \in \mathbb{R}$ . The second moment of  $Z$  is

$$\begin{aligned}\mathbb{E}[Z^2] &= \mathbb{E}[(aX + bY)^2] \\ &= \mathbb{E}[a^2X^2 + b^2Y^2 + 2abXY] \\ &= a^2\mathbb{E}[X^2] + b^2\mathbb{E}[Y^2] + 2ab\mathbb{E}[XY]\end{aligned}$$

By definition  $f(a, b) = \mathbb{E}[Z^2] \geq 0 \forall a, b \in \mathbb{R}$ . Thus we have

$$a^2\mathbb{E}[X^2] + b^2\mathbb{E}[Y^2] + 2ab\mathbb{E}[XY] \geq 0. \quad (2 \text{ marks})$$

Substituting  $\frac{b}{a} = -\frac{\mathbb{E}[XY]}{\mathbb{E}[Y^2]}$  we have

$$\begin{aligned}\mathbb{E}[X^2] + \left(\frac{\mathbb{E}[XY]}{\mathbb{E}[Y^2]}\right)^2 \mathbb{E}[Y^2] - 2\frac{\mathbb{E}[XY]}{\mathbb{E}[Y^2]}\mathbb{E}[XY] &\geq 0 \\ \implies \mathbb{E}[X^2] - \frac{\mathbb{E}[XY]^2}{\mathbb{E}[Y^2]} &\geq 0\end{aligned}$$

Thus we can conclude from above that

$$\begin{aligned}\mathbb{E}[X^2]\mathbb{E}[Y^2] &\geq (\mathbb{E}[XY])^2 \\ \implies \sqrt{\mathbb{E}[X^2]\mathbb{E}[Y^2]} &\geq |\mathbb{E}[XY]| \quad (3 \text{ marks})\end{aligned}$$

- **Part b)** To show

$$|\rho(X, Y)| \leq 1$$

By definition we have

$$\rho(X, Y) = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y}$$

where  $\text{Cov}(X, Y)$  is the covariance given as  $\text{Cov}(X, Y) = \mathbb{E}[(X - \mathbb{E}[X])(Y - \mathbb{E}[Y])]$  and  $\sigma_X, \sigma_Y$  are the standard deviation of  $X$  and  $Y$  resp. given as  $\sqrt{\mathbb{E}[(X - \mathbb{E}[X])^2]}$  and  $\sqrt{\mathbb{E}[(Y - \mathbb{E}[Y])^2]}$ . Consider random variables  $\bar{X} = X - \mathbb{E}[X]$  and  $\bar{Y} = Y - \mathbb{E}[Y]$  then we have

$$\rho(X, Y) = \frac{\mathbb{E}[\bar{X}\bar{Y}]}{\sqrt{\mathbb{E}[\bar{X}^2]\mathbb{E}[\bar{Y}^2]}}. \quad (2 \text{ marks})$$

From Part 1, we have that

$$|\mathbb{E}[\bar{X}\bar{Y}]| \leq \sqrt{\mathbb{E}[\bar{X}^2]\mathbb{E}[\bar{Y}^2]}. \quad (1)$$

Using the definition and (1) we get

$$|\rho(X, Y)| \leq 1 \quad (3 \text{ marks})$$

8. Let  $U_1, U_2, \dots$  be a sequence of independent random variables, with each being uniformly distributed over the interval  $[0, 1]$ , and let  $X_n = \min\{U_1, \dots, U_n\}$  for  $n \geq 1$ .

- (a) Does  $\{X_n\}$  converge in distribution? If yes, to what? (2 marks)
- (b) Does  $\{X_n\}$  converge in probability? If yes, to what? (2 marks)
- (c) Does  $\{X_n\}$  converge in the mean-squared sense? If yes, to what? (3 marks)
- (d) Does  $\{X_n\}$  converge almost surely? If yes, to what? (3 marks)

## Solution

(a)  $X_n \xrightarrow{d} 0$ :

Note that

$$F_{U_n}(u) = \begin{cases} 0 & \text{if } u < 0, \\ 1 & \text{if } u \geq 1, \\ 1 - u & \text{if } 0 \leq u \leq 1. \end{cases}$$

Also,

$$\begin{aligned} F_{X_n}(x) &= P(X_n \leq x) = 1 - P(X_n > x) = 1 - P(U_1 > x, U_2 > x, \dots, U_n > x) \\ &= 1 - P(U_1 > x)P(U_2 > x) \dots P(U_n > x) \quad (\text{since } U_i \text{'s are independent}) \\ &= 1 - (1 - F_{U_1}(x))(1 - F_{U_2}(x)) \dots (1 - F_{U_n}(x)) \\ &= 1 - (1 - x)^n. \end{aligned}$$

Therefore, we conclude

$$\lim_{n \rightarrow \infty} F_{X_n}(x) = \begin{cases} 0 & \text{if } x \leq 0, \\ 1 & \text{if } x > 0. \end{cases}$$

Therefore,  $X_n \xrightarrow{d} 0$ .

(2 marks)

(b)  $X_n \xrightarrow{p} 0$ :

Note that as we found in part (a)

$$F_{X_n}(x) = \begin{cases} 0 & \text{if } x < 0, \\ 1 - (1 - x)^n & \text{if } 0 \leq x \leq 1, \\ 1 & \text{if } x > 1. \end{cases}$$

In particular, note that  $X_n$  is a continuous random variable. To show  $X_n \xrightarrow{p} 0$ , we need to show that

$$\lim_{n \rightarrow \infty} P(|X_n| \geq \epsilon) = 0, \quad \text{for all } \epsilon > 0.$$

Since  $X_n \geq 0$ , it suffices to show that

$$\lim_{n \rightarrow \infty} P(X_n \geq \epsilon) = 0, \quad \text{for all } \epsilon > 0.$$

For  $\epsilon \in (0, 1)$ , we have

$$\begin{aligned} P(X_n \geq \epsilon) &= 1 - P(X_n < \epsilon) = 1 - P(X_n \leq \epsilon) \quad (\text{since } X_n \text{ is a continuous random variable}) \\ &= 1 - F_{X_n}(\epsilon) = (1 - \epsilon)^n. \end{aligned}$$

Therefore,

$$\lim_{n \rightarrow \infty} P(|X_n| \geq \epsilon) = \lim_{n \rightarrow \infty} (1 - \epsilon)^n = 0, \quad \text{for all } \epsilon \in (0, 1].$$

**(2 marks)**

**(c)**  $X_n \xrightarrow{L^2} 0$ :

By differentiating  $F_{X_n}(x)$ , we obtain

$$f_{X_n}(x) = \begin{cases} n(1-x)^{n-1} & \text{if } 0 \leq x \leq 1, \\ 0 & \text{otherwise.} \end{cases}$$

$$\begin{aligned} E[|X_n|^2] &= \int_0^1 nx^2(1-x)^{n-1} dx \leq \int_0^1 nx(1-x)^{n-1} dx \\ &= -x(1-x)^n \Big|_0^1 + \int_0^1 (1-x)^n dy \quad (\text{integration by parts}) \\ &= \frac{1}{n+1}. \end{aligned}$$

Therefore,

$$\lim_{n \rightarrow \infty} E[|X_n|^2] = 0.$$

**(3 marks)**

**(d)**  $X_n \xrightarrow{a.s.} 0$ :

We will prove

$$\sum_{n=1}^{\infty} P(|X_n| > \epsilon) < \infty,$$

which implies  $X_n \xrightarrow{a.s.} 0$ . By our discussion in part (b),

$$\sum_{n=1}^{\infty} P(|X_n| > \epsilon) = \sum_{n=1}^{\infty} (1 - \epsilon)^n = \frac{1 - \epsilon}{\epsilon} < \infty \quad (\text{geometric series}).$$

**(3 marks)**

### Marking scheme:

Partial marks ( $\approx 1$  mark) are awarded if the final answer or the definitions of convergence are correct.

**Problem 9.** Let  $\{X_i\}$  and  $\{Y_i\}$  be independent Bernoulli processes with parameters  $p$  and  $q$  respectively.

(a) Consider the process  $Z_i = \min\{X_i, Y_i\}$ . What kind of process is  $\{Z_i\}$  ? (3 points)

*Solution.*

Let's consider the probability distribution of  $Z_i$ .

Given that  $\{X_i\}$  and  $\{Y_i\}$  are independent processes:

$$\begin{aligned}\mathbb{P}(Z_i = 1) &= \mathbb{P}(\min\{X_i, Y_i\} = 1) \\ &= \mathbb{P}(X_i = 1 \text{ and } Y_i = 1) \\ &= \mathbb{P}(X_i = 1) \times \mathbb{P}(Y_i = 1) \\ &= p \times q\end{aligned}$$

[+1]

$$\begin{aligned}\mathbb{P}(Z_i = 0) &= \mathbb{P}(\min\{X_i, Y_i\} = 0) \\ &= \mathbb{P}(X_i = 0 \text{ or } Y_i = 0) \\ &= 1 - \mathbb{P}(Z_i = 1) \\ &= 1 - p \times q\end{aligned}$$

[+1]

Therefore, the probability distribution of the process  $\{Z_i\}$  is:

$$\begin{aligned}\mathbb{P}(Z_i = 1) &= pq \\ \mathbb{P}(Z_i = 0) &= 1 - pq\end{aligned}$$

$\{Z_i\}$  is also a Bernoulli process but with a parameter equal to the product of the component parameters  $p$  and  $q$ . [+1]

□

(b) Is  $\{Z_i\}$  strict sense stationary (SSS)? (3 points)

*Solution.* Consider time instants  $i$  and  $j$ , such that  $j = i + n$ ,  $\forall n \leq N$ :

$$\begin{aligned}Z_i &= \min\{X_i, Y_i\} \\ Z_j &= \min\{X_j, Y_j\}\end{aligned}$$

[+1]

$X_i, Y_i, X_j$  and  $Y_j$  are independent of each other, hence,  $Z_i$  and  $Z_j$  are independent of each other. Also,  $X_i$  and  $X_j$  share the parameter  $p$  and,  $Y_i$  and  $Y_j$  share the parameter  $q$ . [+1]

Therefore,

$$\text{PMF}(Z_i) \equiv \text{PMF}(Z_j)$$

Thus,  $\{Z_i\}$  is  $N^{\text{th}}$  order stationary. Since the above argument holds for all  $N \in \mathbb{N}$ ,  $\{Z_i\}$  is strict sense stationary. [+1]

□

(c) For the process  $\{X_i\}$ , what is the probability that the  $k^{\text{th}}$  success occurs before the  $j^{\text{th}}$  failure? (4 points)

*Solution.*

We want at least  $k$  successes to have occurred before the  $j^{\text{th}}$  failure occurs, which implies at most  $(j - 1)$  failures can occur before the  $k^{\text{th}}$  success.

Consider the case where the  $k^{\text{th}}$  success occurs after exactly  $n$  failures. The corresponding probability  $\mathbb{P}_{n,k}$  is obtained by multiplying the probability of observing exactly  $k - 1$  successes and  $n$  failures by the probability of getting the  $k^{\text{th}}$  success immediately after:

$$\begin{aligned}\mathbb{P}_{n,k} &= \left( \binom{k-1+n}{k-1} \cdot p^{k-1} \cdot (1-p)^n \right) \times p \\ &= \binom{k-1+n}{k-1} \cdot p^k \cdot (1-p)^n\end{aligned}$$

The required quantity is then observed by summing  $\mathbb{P}_{n,k}$  for the failures  $n$  varying from 0 to  $(j - 1)$ : [+2]

$$\sum_{n=0}^{j-1} \mathbb{P}_{n,k} = \sum_{n=0}^{j-1} \binom{k-1+n}{k-1} \cdot p^k \cdot (1-p)^n$$

[+2]

□

# EE 325

## Probability and Random Processes

Endsem Q.10

- A random walk is a discrete-time random process with  $X_0 = 0$ , and the update rule,

$$X_{n+1} = X_n + U_n, \quad n \in \mathbb{N} \cup \{0\},$$

where the  $U_n$ 's are independent and identically distributed random variables with mean  $\mu$  and variance  $\sigma^2$ .

1. Prove that  $X_t - X_s = \sum_{i=s}^t U_i$ . (3 Points)

**Solution:** The random walk has the “independent increment property”. To see this, note that for  $s, t \in \mathbb{N} \cup \{0\}$ ,

$$\begin{aligned} X_t &= X_t - X_0 \\ X_t &= (X_t - X_s) + (X_s - X_0) \\ \implies X_t - X_s &= (X_t - X_0) - (X_s - X_0) \\ X_t - X_s &= \sum_{i=s}^t U_i \end{aligned}$$

which is a function of  $U_s, \dots, U_{t-1}$ , while  $X_s - X_0$  is a function of  $U_0, \dots, U_{s-1}$ . Since  $U$  is an i.i.d. RV, the independent increment property holds.

2. Compute  $\mu_X(t)$ . (2 Points)

**Solution:**  $\mathbb{E}[U_n] = \mu$ . Hence,  $\mu_X(t) = \mathbb{E}[X_t] = t\mu$ .

3. Compute  $R_X(t, s)$ . (3 Points)

**Solution:**

$$\begin{aligned} R_X(t, s) &= \mathbb{E}[X_t - X_s] \quad (\text{assume WLOG that } t > s) \\ &= \mathbb{E}[(X_t - X_0) - (X_s - X_0)] \\ &= \mathbb{E}[(X_s - X_0)(X_t - X_s + X_s - X_0)] \\ &= \mathbb{E}[(X_s - X_0)(X_t - X_s)] + \mathbb{E}[(X_s - X_0)^2] \\ &= \mathbb{E}[X_s - X_0] \mathbb{E}[X_t - X_s] + \mathbb{E}[(X_s - X_0)^2] \\ &= s\mu^2 + s\sigma^2. \end{aligned}$$

4. Is this process WSS?

(2 Points)

**Solution:** For WSS,

(a)  $\mu_X(t) = \mu_X, \forall t \in \mathbb{R}$

(b)  $R_X(t, s) = R_X(t - s), \forall t, s \in \mathbb{R}.$

It is immediate that if  $\mu \neq 0$  or  $\sigma \neq 0$ , then the random walk above is not stationary.