

# INDIAN STATISTICAL INSTITUTE

## Class Test III

M Tech (CS), 2023 (Semester – I)

Probability and Stochastic Processes

Date: 10.11.2023

Maximum Marks : 30

Duration : 1.0 hours

**Note:** The question paper is of 40 marks. Answer as much as you can, but the maximum you can score is 30. Answer a question within its allotted box.

**Course:**(MTech/JRF/PLP) \_\_\_\_\_ **Name:** \_\_\_\_\_ **Roll Number:** \_\_\_\_\_

(Q1) Let  $X_1, \dots, X_n$  be independent uniform random variables over  $[0, 1]$ . Let  $Y = \min(X_1, \dots, X_n)$ . Show that  $E[Y] = \frac{1}{n+1}$ . [10]

**(Ans 1:)** We know for a continuous random variable  $X$ , that takes only nonnegative values, and has a density function  $f(x)$ ,

$$\begin{aligned} \int_{x=0}^{\infty} \Pr(X \geq x) dx &= \int_{x=0}^{\infty} \int_{y=x}^{\infty} f(y) dy dx \\ &= \int_{y=0}^{\infty} \int_{x=0}^y f(y) dx dy \\ &= \int_{y=0}^{\infty} y f(y) dy \\ &= E[X]. \end{aligned}$$

Now, as per the question,  $Y_1 = \min(X_1, \dots, X_n)$  and  $Y_1$  takes value in the range  $[0, 1]$ . We want to find  $E[Y_1]$  which is  $\int_{y=0}^1 \Pr(Y_1 \geq y) dy$ . So, we need to figure out the probability of  $\Pr(Y_1 \geq y)$ .

$$\begin{aligned} \Pr(Y_1 \geq y) &= \Pr(\min(X_1, \dots, X_n) \geq y) \\ &= \Pr((X_1 \geq y) \cap (X_2 \geq y) \cap \dots \cap (X_n \geq y)) \\ &= \prod_{i=1}^n \Pr(X_i \geq y) \\ &= (1 - y)^n \end{aligned}$$

Therefore,  $E[Y_1] = \int_{y=0}^1 (1 - y)^n dy = \frac{1}{n+1}$ . ◀

- (Q2) (i) Let  $f$  and  $g$  be two PDFs corresponding to two random variables  $X$  and  $Y$ , respectively. Show that the function  $h = \alpha f + (1 - \alpha)g$  is also a PDF, where  $0 \leq \alpha \leq 1$ .
- (ii) Compute the moment generating function of a normal random variable  $X \sim N(\mu, \sigma^2)$ . [5+5=10]

**(Ans 2(i))** As  $0 \leq \alpha \leq 1$ , and  $f, g \geq 0$ , so  $h \geq 0$ . For  $h$  to be a pdf, we need to show  $\int_{\mathbb{R}} h dx = 1$ . Indeed it is, because  $\int_{\mathbb{R}} h dx = \int_{\mathbb{R}} \alpha f dx + \int_{\mathbb{R}} (1 - \alpha) g dx = \alpha \int_{\mathbb{R}} f dx + (1 - \alpha) \int_{\mathbb{R}} g dx = \alpha \cdot 1 + (1 - \alpha) \cdot 1 = 1$ . ◀

**(Ans 2(ii))** Done in the class.

(Q3) Let  $N, X_1, X_2, \dots$  be independent random variables, where  $X_i$ 's are Bernoulli with parameter  $p$ , and  $N$  takes nonnegative integral values. Let  $Y = X_1 + \dots + X_N$  and  $Y = 0$  when  $N = 0$ . Show that if  $N$  is binomial with parameters  $m$  and  $q$ , then  $Y$  is binomial with parameters  $m$  and  $pq$ . [10]

**(Ans 3:)** The moment generating function of  $Y$ , i.e.,  $M_Y(t) = E[e^{tY}]$ .

$$\begin{aligned}
 E[e^{tY}] &= E[E[e^{tY} | N]] \\
 &= E[M_{X_1}(t)^N] \\
 &= E[(pe^t + (1-p))^N] \\
 &= \sum_{i=0}^m \Pr(N = i) \cdot (pe^t + (1-p))^i \\
 &= \sum_{i=0}^m \binom{m}{i} q^i (1-q)^{m-i} \cdot (pe^t + (1-p))^i \\
 &= \sum_{i=0}^m \binom{m}{i} (pqe^t + (1-p)q)^i (1-q)^{m-i} \\
 &= (pqe^t + (1-pq))^m
 \end{aligned}$$

This implies  $Y$  follows the binomial distribution with parameters  $m$  and  $pq$ . ◀

- (Q4) (i) A sequence  $X_n$  of random variables is said to converge to a number  $c$  in the mean square, if  $\lim_{n \rightarrow \infty} E[(X_n - c)^2] = 0$ . Show that convergence in the mean square implies convergence in probability.
- (ii) Jobs are processed, one at a time. The processing times of jobs are independent random variables, uniformly distributed in  $[1, 10]$ . Find or approximate the probability that the number of jobs processed within 200 time units is at least 75. You can use standard normal distribution table. [5+5=10]

**(Ans 4(i))** Let  $X_n$  converges to  $c$  in the mean square. Using the Markov inequality, we get

$$\Pr(|X_n - c| \geq \epsilon) \leq \frac{E[(X_n - c)^2]}{\epsilon^2}.$$

This implies  $\lim_{n \rightarrow \infty} \Pr(|X_n - c| \geq \epsilon) = 0$ , which is the desired result. ◀

**(Ans 4(ii))** Let  $X_i$  be the random variable that denotes the processing time of the  $i^{\text{th}}$  job,  $i \in \mathbb{N}$ . Note that  $X_i \in [1, 10]$  and, is uniformly distributed.  $Z = \frac{X_1 + \dots + X_{75} - 75 \cdot \mu}{\sqrt{75} \sigma}$ , where  $\mu$  and  $\sigma$  are the mean and standard deviation of each  $X_i$ . So,  $\mu = 11/2$  and  $\sigma \approx 2.59$ .

Observe that we have to compute the following.

$$\begin{aligned} & \Pr(X_1 + \dots + X_{75} \leq 200) \\ & \approx \Pr(Z \leq -9.47) \\ & \approx \Phi(-9.47) \text{ (by CLT)} \end{aligned}$$