

**INDIAN STATISTICAL INSTITUTE**

**Class Test II**

M Tech (CS), 2023 (Semester – I)

*Probability and Stochastic Processes*

Date: 18.10.2023

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  $X$  and  $Y$  be independent Bernoulli random variables with parameter  $\frac{1}{2}$ . Show that  $X + Y$  and  $|X - Y|$  are dependent though uncorrelated. [5]

(Ans:) Let us look at the covariance of the random variables  $X + Y$ , and  $|X - Y|$ .

$$\text{cov}(X + Y, |X - Y|) = E[(X + Y)(|X - Y|)] - E[X + Y] \cdot E[|X - Y|]$$

The values of the random variables and their corresponding probabilities are as follows:

$(X + Y) \quad  X - Y $	0	1	2
Probability	$\frac{5}{8}$	$\frac{1}{4}$	$\frac{1}{8}$

$(X + Y)$	0	1	2
Probability	$\frac{1}{4}$	$\frac{1}{2}$	$\frac{1}{4}$

$(X - Y)$	0	1
Probability	$\frac{1}{2}$	$\frac{1}{2}$

$$\begin{aligned} \text{cov}(X + Y, |X - Y|) &= E[(X + Y)(|X - Y|)] - E[X + Y] \cdot E[|X - Y|] \\ &= \frac{1}{4} + \frac{1}{4} - 1 \cdot \frac{1}{2} \\ &= 0 \end{aligned}$$

So,  $X + Y$  and  $|X - Y|$  are uncorrelated.

Looking at the probabilities  $\Pr(X + Y = 0, |X - Y| = 0)$ ,  $\Pr(X + Y = 0)$  and  $\Pr(|X - Y| = 0)$ , we see that  $\Pr(X + Y = 0, |X - Y| = 0) \neq \Pr(X + Y = 0) \cdot \Pr(|X - Y| = 0)$ . So, they are dependent. ◀

(Q2) A *tournament* on a set  $V$  of  $n$  players is an orientation  $T = (V, E)$  of the edges of the complete graph on the set of vertices  $V$ . Thus for every two distinct elements  $x$  and  $y$  of  $V$ , either  $(x, y)$  or  $(y, x)$  belongs to  $E$ , but not both. A simple interpretation of *tournament* is in terms of games where each distinct pair  $x, y$  of players,  $x, y \in V$ , play a single match; the outcome of the games are either win or loss.  $(x, y)$  is in the *tournament* if and only if  $x$  beats  $y$ .

$T$  has the property  $S_k$  if for every set of  $k$  players there is one who beats them all.

Show that if  $\binom{n}{k}(1 - 2^{-k})^{n-k} < 1$ , then there is a tournament on  $n$  vertices that has the property  $S_k$ . [10]

[(Ans:)] Consider a fixed subset  $K$  ( $K \subset V$ ) of size  $k$  from a random tournament  $V = \{1, 2, \dots, n\}$ .

Notice that for each vertex  $v \in V \setminus K$ , the probability that  $v$  does not beat all the members of  $K$  is  $1 - 2^{-k}$  and all of these  $n - k$  events for a choice of  $v$  are independent.

Let  $\mathcal{E}_K$  be the event that there is no player (vertex) who beats all the members of  $K$ . So,  $\Pr(\mathcal{E}_K) = (1 - 2^{-k})^{n-k}$ . We next use the union bound and the condition given in the question to deduce the result, as follows:

$$\Pr\left(\bigcup_{K \subset V, |K|=k} \mathcal{E}_K\right) = \sum_{K \subset V, |K|=k} \Pr(\mathcal{E}_K) = \binom{n}{k} (1 - 2^{-k})^{n-k} < 1.$$



(Q3) We toss a biased coin  $n$  times. The probability of heads for the biased coin is given by  $q$ , the value of a random variable  $Q$  having mean  $\mu$  and positive variance  $\sigma^2$ . Let  $X_i$ 's be Bernoulli random variables modelling the outcome of the  $i$ -th toss;  $X_i = 1$  if and only if the  $i$ -th toss is a head. Let the  $X_i$ 's ( $i = 1, \dots, n$ ) be conditionally independent, given  $Q = q$ . Let  $X$  be the number of heads obtained in the  $n$  tosses. Find (i)  $E[X_i]$  and  $E[X]$ ; (ii) Find  $\text{cov}(X_i, X_j)$ . Are  $X_1, \dots, X_n$  independent?; (iii) Find  $\text{var}(X)$ . [(2+2)+(4+1)+6=15]

[(Ans (i):] As each  $X_i$  is a 0-1 random variable with the probability of head following a random variable  $Q$ , we have  $E[X_i | Q] = \Pr(X_i = 1) = Q$ . So,  $E[X_i] = E[E[X_i | Q]] = E[Q] = \mu$ . Since  $X = \sum_{i=1}^n X_i$ ,  $E[X] = n\mu$ .

[(Ans (ii):)] Using the assumption on conditional independence, we have for  $i \neq j$ ,

$$E[X_i X_j | Q] = E[X_i | Q] \cdot E[X_j | Q] = Q^2.$$

Also,  $E[X_i X_j] = E[E[X_i X_j | Q]] = E[Q^2]$ . So,

$$\text{cov}(X_i, X_j) = E[X_i X_j] - E[X_i] \cdot E[X_j] = E[Q^2] - \mu^2 = \sigma^2 > 0.$$

Since,  $\text{cov}(X_i, X_j) > 0$ ,  $X_1, \dots, X_n$  are not independent.

[(Ans (iii):)] As  $X_i$  is a 0-1 random variable, we know that  $X_i^2 = X_i$ . Also, for  $i = j$ ,

$$\text{var}(X_i) = E[X_i^2] - (E[X_i])^2 = E[X_i] - (E[X_i])^2 = \mu - \mu^2.$$

Now, use the law of total variance, to get

$$\begin{aligned} \text{var}(X) &= E[\text{var}(X | Q)] + \text{var}(E[X | Q]) \\ &= E[\text{var}(X_1 + \dots + X_n | Q)] + \text{var}(E[X_1 + \dots + X_n | Q]) \\ &= E[nQ(1 - Q)] + \text{var}(nQ) \\ &= nE[Q - Q^2] + n^2 \text{var}(Q) \\ &= n(\mu - \mu^2 - \sigma^2) + n^2 \sigma^2 \\ &= n(\mu - \mu^2) + n(n - 1)\sigma^2 \end{aligned}$$

- (Q4) (i) Let  $X$  be a random variable that assumes only nonnegative values. Show that for all  $a > 0$ ,  $\Pr(X \geq a) \leq \frac{E[X^m]}{a^m}$ , where  $m$  is a positive integer  $\geq 2$ .
- (ii) Let  $X_1, \dots, X_n$  be a sequence of i.i.d. random variables with mean  $\mu$  and variance  $\sigma^2$ . Let the sample mean  $M_n = \frac{X_1 + \dots + X_n}{n}$ . Show that  $\Pr(|M_n - \mu| \geq \varepsilon) \rightarrow 0$ , as  $n \rightarrow \infty$ . [5+5=10]

[Ans (i):] For  $a > 0$ , let

$$I = \begin{cases} 1 & \text{if } X^m \geq a^m; \\ 0 & \text{otherwise.} \end{cases}$$

Since  $X \geq 0$ , we have  $I \leq 1 \leq \frac{X^m}{a^m}$ . As  $I$  is a 0-1 r.v.,  $E[I] = \Pr(I = 1) = \Pr(X^m \geq a^m)$ . Take expectations on both sides to get

$$\Pr(X^m \geq a^m) = E[I] \leq E\left[\frac{X^m}{a^m}\right] = \frac{E[X^m]}{a^m}.$$

[Ans (ii):] Let  $X_1, \dots, X_n$  be a sequence of i.i.d. random variables with mean  $\mu$  and variance  $\sigma^2$ . The sample mean  $M_n = \frac{X_1 + \dots + X_n}{n}$ . The mean and variance of  $M_n$  are  $\mu$  and  $\frac{\sigma^2}{n}$ . Now, applying Chebyshev inequality with  $\varepsilon > 0$ , we get

$$\Pr(|M_n - \mu| \geq \varepsilon) \leq \frac{\sigma^2}{n\varepsilon^2}.$$

Now for every  $\varepsilon > 0$  as  $n \rightarrow \infty$ , we have

$$\Pr(|M_n - \mu| \geq \varepsilon) = \Pr\left(\left|\frac{X_1 + \dots + X_n}{n} - \mu\right| \geq \varepsilon\right) \leq \frac{\sigma^2}{n\varepsilon^2} \rightarrow 0.$$