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Name: SR No.: Dept.:

Maximum Points: 15 E2-243: Quiz 4 **Duration**: 40 minutes

- Tick the correct answer (1+1+1+1) points. -0.5 point for each negative answer
  - 1. Variance of any random variable is
    - a) Always positive
    - b) Always zero
    - c) Could be positive or negative
    - d) Always non negative

**Answer:** d. Since it is expectation of non-negative variable  $(X - \mu_x)^2$ 

- 2. X and Y are random variables defined on the same probability space. It is given that Var(X + Y) = Var(X) + Var(Y). Then
  - a) X and Y are always independent
  - b) Cov(X,Y) = 0
  - c) Both (a) and (b) are true
  - d) E(X) = E(Y)

**Answer:** b. Cov(X,Y)=0

- 3. Let W be a non empty subset of  $\mathbb{R}^3$ .
  - a) If null vector  $\theta_3 \in W$  then W is always a subspace of  $\mathbb{R}^3$
  - b) If  $x \in W$ ,  $\alpha \in R \Rightarrow \alpha * x \in W$  then W is always a subspace of  $R^3$
  - c) If W is a subspace of  $R^3$  then null vector  $\theta_3 \in W$  is always true
  - d) Both (a) and (c) are always true

**Answer:** c. null vector  $\theta_3 \in W$  and  $\alpha * x \in W$  are necessary conditions but not sufficient condition for W to be a subspace of  $R^3$ 

- 4. Let W be a subspace in  $\mathbb{R}^3$ . It is given that the vector  $\alpha_1 u_1 + \alpha_2 u_2 + \alpha_3 u_3 \in W \forall \alpha_1, \alpha_2, \alpha_3 \in \mathbb{R}$ . Then
  - a) Only one of  $u_1, u_2, u_3 \in W$
  - b) Exactly two amongst  $u_1, u_2, u_3 \in W$
  - c) All  $3 u_1, u_2, u_3 \in W$
  - d) None of  $u_1, u_2, u_3 \in W$

**Answer:** c. Could be proven by making any two of the vectors  $\alpha_1$ ,  $\alpha_2$ ,  $\alpha_3 = 0$ 

- Random variables X and Y are defined on the same probability space as follows:
  - X = 1 and -1 with probability 0.05 each
  - X = 2 and -2 with probability 0.1 each
  - X = 3 and -3 with probability 0.15 each
  - X = 4 and -4 with probability 0.2 each
  - Y = |X|

Answer the following questions (6 points.)

1. Plot the CDF of Y.

(1 point)

Ans:

$$F_Y(x) = \begin{cases} 0, & \text{if } x \in (-\infty, 1) \\ 0.1, & \text{if } x \in [1, 2) \\ 0.3, & \text{if } x \in [2, 3) \\ 0.6, & \text{if } x \in [3, 4) \\ 1, & \text{if } x \in [4, \infty) \end{cases}$$

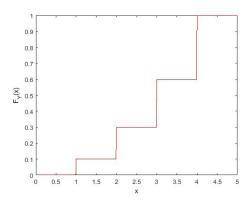


Figure 1: Optimal and approximate policy for  $p = 0.7, \psi = 10, d = 21$ .

2. Find the expectation of Y.

(1 point)

**Ans:** 3. (1\*0.1)+(2\*0.2)+(3\*0.3)+(4\*0.4)

- 3. Prove or disprove. X and Y are independent. (1.5 points) Ans: No. P(X=1,Y=2) = 0. P(X=1).P(Y=2)=0.05\*0.2. Since both are not equal, they are not independent.
- 4. Prove or disprove. X and Y are uncorrelated. (1.5 points) Ans: Yes. E(XY)=0=E(X)E(Y).
- 5. Can we apply Markov's inequality to find  $P(X \ge 0)$ . State your reason. (1 point) **Ans:** No. Since Markov's inequality could be applied only to non-negative random variables.
- State whether the following are true or false (1+1+1 points. -0.5 point for each negative answer)
  - 1. A sequence,  $\{X_n\}_{n\in\mathbb{N}}$ , of random variables on a Probability space  $(\Omega, \mathbb{B}, \mathbb{P})$  is said to converge "in probability" to a random variable X on  $(\Omega, \mathbb{B}, \mathbb{P})$  if

$$\lim_{n \to \infty} P(w \in \Omega : |X_n(w) - X(w)| < \epsilon) = 1$$

for every  $\epsilon > 0$ 

**Ans:** True. As the definition says,

$$\lim_{n \to \infty} P(w \in \Omega : |X_n(w) - X(w)| \ge \epsilon) = 0$$

for every  $\epsilon > 0$ , it implies

$$\lim_{n \to \infty} P(w \in \Omega : |X_n(w) - X(w)| < \epsilon) = 1$$

for every  $\epsilon > 0$ 

2. If x > y, then  $\forall a \in \mathbb{R}$ , ax > ay.

**Ans:** False. If a > 0, then ax > ay. Note that 3 > 2 but  $-1 \times 3 < -1 \times 2$ .

3.  $\lim_{n\to\infty} \mathbb{E}(|X_n - X|^k) = 0 \iff X_n \xrightarrow{p} X$ , where k is any positive integer.

**Ans:** False. As  $\lim_{n\to\infty} \mathbb{E}(|X_n-X|^k) = 0 \Rightarrow X_n \xrightarrow{p} X$ , where k is any positive integer. Find the following example

$$X_n = \begin{cases} n, & \text{w.p. } \frac{1}{n} \\ 0, & \text{w.p. } 1 - \frac{1}{n} \end{cases}$$

$$P(X_n > \epsilon) = \begin{cases} \frac{1}{n}, & \text{if } \epsilon \in [0, n) \\ 0, & \text{otherwise} \end{cases}$$

Therefore,  $\lim_{n\to\infty} P(|X_n-0|>\epsilon)=0$ . Hence it converges in probability. Let us look at  $k^{th}$  mean convergence

$$\lim_{n \to \infty} E[|X_n - 0|^k] = \lim_{n \to \infty} n^k * \frac{1}{n} + 0^2 * (1 - \frac{1}{2}) \neq 0$$

Therefore, convergence in probability does not imply convergence in  $k^{th}$  mean.

• Give an example where Markov inequality provides a tightest bound. . (2 points) Ans: Consider the following example where X, a non negative random variable that can take either 0 or k > 0.

$$P(X = 0) = 1 - \frac{1}{k^2}, P(X = k) = \frac{1}{k^2}$$

Let us evaluate Markov inequality as follows

$$P(X \ge k) \le \frac{E(X)}{k} = \frac{1}{k^2}$$

Also, note that  $P(X \ge k) = \frac{1}{k^2}$ . Thus Markov inequality is tight for this example.