

Indian Institute of Information Technology Allahabad
Probability and Statistics (PAS)
C2 Review Test Tentative Marking Scheme

1. Let X_1 and X_2 have the joint probability density function [11]

$$f_{X_1, X_2}(x_1, x_2) = \begin{cases} 10x_1x_2^2, & 0 < x_1 < x_2 < 1 \\ 0, & \text{otherwise.} \end{cases}$$

- (a) Find the probability density function of $Y = \frac{X_1}{X_2}$ by using transformation of variables technique.
 (b) Find $Cov(X_1, X_2)$.
 (c) Are Y and X_2 uncorrelated? Justify your answer.

Solution:

- (a) Let $Z = X_2$. Then $x_1 = yz$ and $x_2 = z$. [1/2]

Hence

$$J = \begin{vmatrix} \frac{\partial x_1}{\partial y} & \frac{\partial x_1}{\partial z} \\ \frac{\partial x_2}{\partial y} & \frac{\partial x_2}{\partial z} \end{vmatrix} = \begin{vmatrix} z & y \\ 0 & 1 \end{vmatrix} = z \quad [1/2]$$

Hence, the joint pdf of Y and Z is

$$\begin{aligned} f_{Y,Z}(y, z) &= f_{X_1, X_2}(yz, z)|J| \\ &= \begin{cases} 10yz^4, & 0 < yz < z < 1 \\ 0, & \text{otherwise} \end{cases} \\ &= \begin{cases} 10yz^4, & 0 < y < 1, 0 < z < 1 \\ 0, & \text{otherwise} \end{cases} \end{aligned} \quad [1+1]$$

Hence the pdf of Y is

$$\begin{aligned} f_Y(y) &= \int_{-\infty}^{\infty} f_{Y,Z}(y, z) dz \\ &= \begin{cases} \int_0^1 10yz^4 dz, & 0 < y < 1 \\ 0, & \text{otherwise} \end{cases} \\ &= \begin{cases} 2y, & 0 < y < 1 \\ 0, & \text{otherwise} \end{cases} \end{aligned} \quad [1+1]$$

$$(b) \ E(X_1X_2) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x_1x_2f_{X_1, X_2}(x_1, x_2)dx_1dx_2 = \int_0^1 \int_0^{x_2} 10x_1^2x_2^3dx_1dx_2 = \frac{10}{21} \quad [1]$$

$$E(X_1) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x_1f_{X_1, X_2}(x_1, x_2)dx_1dx_2 = \int_0^1 \int_0^{x_2} 10x_1^2x_2^2dx_1dx_2 = \frac{5}{9} \quad [1]$$

$$E(X_2) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x_2f_{X_1, X_2}(x_1, x_2)dx_1dx_2 = \int_0^1 \int_0^{x_2} 10x_1x_2^3dx_1dx_2 = \frac{5}{6} \quad [1]$$

$$Cov(X_1, X_2) = E(X_1X_2) - E(X_1)E(X_2) = \frac{10}{21} - \frac{25}{54} = \frac{5}{378} \quad [1]$$

$$(c) E(Y) = \int_{-\infty}^{\infty} y f_Y(y) dy = \int_0^1 2y^2 dy = \frac{2}{3} \quad [1]$$

Now, $Cov(Y, X_2) = E(YX_2) - E(Y)E(X_2) = E(X_1) - E(Y)E(X_2) = \frac{5}{9} - \frac{10}{18} = 0$.
Thus Y and X_2 are uncorrelated. [1]

2. Let $(X_n)_{n \geq 1}$ be a sequence of random variables with corresponding distribution functions given by [7]

$$F_n(x) = \begin{cases} 0, & x < -n \\ \frac{x+n}{2n}, & -n \leq x < n \\ 1, & x \geq n. \end{cases}$$

Does F_n converge to a distribution function. Justify your answer.

Solution: For $x \in \mathbb{R}$ with $x < 0$, we have a $N \in \mathbb{N}$ such that $-N \leq x < -(N-1)$.
Then

$$F_n(x) = \begin{cases} 0, & n < N \\ \frac{x+n}{2n}, & n \geq N \end{cases} \quad [2]$$

Similarly, for $x \in \mathbb{R}$ with $x > 0$, we have a $M \in \mathbb{N}$ such that $M-1 \leq x < M$. Then

$$F_n(x) = \begin{cases} 1, & n < M \\ \frac{x+n}{2n}, & n \geq M \end{cases} \quad [2]$$

Also $F_n(0) = \frac{1}{2}$, for all $n \in \mathbb{N}$. [1]

Hence $\lim_{n \rightarrow \infty} F_n(x) = \frac{1}{2}$, for all $x \in \mathbb{R}$. That is. F_n converges to the function $f(x) = \frac{1}{2}$,
for all $x \in \mathbb{R}$, which is not a distribution function. [1+1]

3. Let $(X_n)_{n \geq 1}$ be a sequence of identically and independently distributed random variables with common probability density function [7]

$$f(x) = \begin{cases} e^{-x+\theta}, & x \geq \theta \\ 0, & x < \theta. \end{cases}$$

Let $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$. Show that \bar{X}_n converges to $1 + \theta$ in probability.

$$\textbf{Solution: } E(X_i) = \int_{-\infty}^{\infty} x f(x) dx = \int_{\theta}^{\infty} x e^{-x+\theta} dx = 1 + \theta, \forall i \in \mathbb{N} \quad [1]$$

$$E(X_i^2) = \int_{-\infty}^{\infty} x^2 f(x) dx = \int_{\theta}^{\infty} x^2 e^{-x+\theta} dx = \theta^2 + 2\theta + 2, \forall i \in \mathbb{N} \quad [1]$$

$$Var(X_i) = E(X_i^2) - (E(X_i))^2 = 1, \forall i \in \mathbb{N}. \quad [1]$$

$$E(\bar{X}_n) = \frac{1}{n} \sum_{i=1}^n E(X_i) = 1 + \theta \quad [1]$$

$$Var(\bar{X}_n) = \frac{1}{n^2} \sum_{i=1}^n Var(X_i) = \frac{1}{n} \quad [1]$$

$$\text{By Chebyshev's Inequality, for every } \epsilon > 0, \text{ we have } P(\{|\bar{X}_n - (1 + \theta)| \geq \epsilon\}) \leq \frac{1}{n\epsilon^2} \quad [1]$$

$$\text{Since } \lim_{n \rightarrow \infty} \frac{1}{n\epsilon^2} = 0, \lim_{n \rightarrow \infty} P(\{|\bar{X}_n - (1 + \theta)| \geq \epsilon\}) = 0. \quad [1]$$