

- 1(a) Check if: $f_x(n) \cdot f_y(y) \cdot f_z(z) = f_{x,y,z}(n,y,z)$ ①
 implies that the three X and Y are independent.
 i.e. check if ① implies $f_{x,y}(n,y) = f_x(n) \cdot f_y(y)$?

Answer: Integrating both sides of ① w.r.t. Z :

$$\int f_x(n) \cdot f_y(y) \cdot f_z(z) dz = \int f_{x,y,z}(n,y,z) dz$$

on $\forall z$: since $f_z(z) = 1$ (constant) \Rightarrow must integrate w.r.t. z

$$\text{or } f_x(n) \cdot f_y(y) \int f_z(z) dz = f_{x,y}(n,y) \quad (\text{marginal pdf of } x,y)$$

$$\text{or } f_x(n) \cdot f_y(y) = 1 \quad (\text{if } f_{x,y}(n,y) = 1 \text{ then})$$

$$\text{or } [f_x(n) \cdot f_y(y) = f_{x,y}(n,y)] \text{ which is } ② \text{ which is what we set to check for!}$$

$$\therefore \boxed{f_x(n) \cdot f_y(y) \cdot f_z(z) = f_{x,y,z}(n,y,z)} \Rightarrow X \text{ and } Y \text{ are independent} \quad \text{Ans}$$

$$\therefore \boxed{P(X)P(Y)P(Z) = P(X,Y,Z)} \quad ③$$

implies that X and Y are independent i.e. $P(X) \cdot P(Y) = P(X,Y)$?
 where X, Y, Z are events of a probabilistic experiment.

No.

Counterexample:

Set experiment

In an experiment, let we are picking a number from 1 to 8 randomly: Each number has equal probability of being picked. This experiment is like throwing a fair 8-sided die.

Now let us define three events:

$$X = \{1, 2, 3, 4\} \Rightarrow P(X) = \frac{1}{2}$$
$$Y = \{1, 3, 4, 5\} \Rightarrow P(Y) = \frac{1}{2}$$
$$Z = \{1, 6, 7, 8\} \Rightarrow P(Z) = \frac{1}{2}$$

~~$$X \cap Y \cap Z = \{1\}$$~~
$$\Rightarrow P(X \cap Y \cap Z) = \frac{1}{8}$$

So $P(X) \neq P(\bar{X})$

So equation ① is followed by events X, Y, Z , as

$$P(X) \cdot P(Y) \cdot P(Z) = \left(\frac{1}{2}\right)^3 = \frac{1}{8} = P(X, Y, Z)$$

But $P(X) \cdot P(Y) = \left(\frac{1}{2}\right)^2 = \frac{1}{4}$

$$P(X \cap Y) = P(\{1, 3, 4\}) = \frac{3}{8} \quad \boxed{\text{X and Y are NOT independent.}}$$

In fact, $P(Y) \cdot P(Z) = \frac{1}{4}$

$$P(Y \cap Z) = P(\{1\}) = \frac{1}{8} \quad \boxed{\text{Y and Z are NOT independent either}}$$

And $P(X) \cdot P(Z) = \frac{1}{4}$

$$P(X \cap Z) = P(\{1\}) = \frac{1}{8} \quad \boxed{\text{X and Z are NOT independent either.}}$$

∴ The analogous event $P(X) \cdot P(Y) \cdot P(Z) = P(X, Y, Z)$ does NOT imply independence of X and Y. Ans

∴ $P(X) \cdot P(Y) \cdot P(Z) = P(X, Y, Z)$ does NOT imply independence of X and Y. Ans

2.1

2. $x, y \sim \mathcal{N}(\mu_x=0, \mu_y=0, \sigma_x^2=1, \sigma_y^2=4, \rho_{xy}=0.5)$

$$f_{x,y}(n, y) = \frac{1}{\sqrt{1-\rho^2} \sigma_x \sqrt{2\pi} \sigma_y \sqrt{2\pi}}$$

$$= \frac{1}{\sqrt{1-(0.5)^2}} \left\{ \frac{(n-\mu_x)^2 + (\frac{y-\mu_y}{2})^2}{\sigma_x^2} - \frac{2(n-\mu_x)(y-\mu_y)}{\sigma_x \sigma_y} \right\}$$

only component (say $g(n, y)$) dependent on n and y .

Taking out the component $g(n, y)$ and putting the given values of $\mu_x, \mu_y, \sigma_x^2, \sigma_y^2$ and ρ_{xy} :

$$g(n, y) = \left(\frac{n}{1}\right)^2 + \left(\frac{y}{2}\right)^2 - \frac{2(n)(y)}{(1)(2)}$$

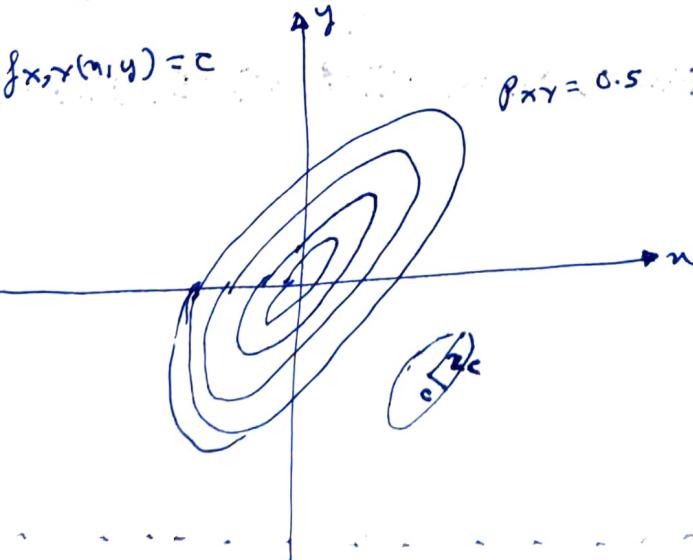
$\Rightarrow f_{x,y}(n, y) \rightarrow c$ for $g(n, y) \rightarrow c$
i.e. is constant is constant

\Rightarrow Contours of $f_{x,y}(n, y)$ is

$$g(n, y) = \left(\frac{n}{1}\right)^2 + \left(\frac{y}{2}\right)^2 - ny = c$$

which is the equation for an ellipse, tilted and with positive slope.

2(a)

Ans

2(b)

$$f_X(n) = \mathcal{N}(x, 0, 1^2) = \frac{1}{1 \cdot \sqrt{2\pi}} e^{-\frac{1}{2} \left\{ \left(\frac{n}{1}\right)^2 \right\}} \quad \forall n$$

$$f_Y(y) = \mathcal{N}(y, 0, 2^2) = \frac{1}{2 \cdot \sqrt{2\pi}} e^{-\frac{1}{2} \left\{ \left(\frac{y}{2}\right)^2 \right\}} \quad \forall y$$

2(c)

$$f_{Y|X}(y|x=x) = \frac{f_{X,Y}(x=n, y)}{f_X(x=n)} \quad \forall n, y$$

$$\text{or } f_{Y|X}(y|x=x) = \frac{1}{\sqrt{(1-\frac{1}{2})} \cdot 1 \cdot \sqrt{2n} \cdot 2 \cdot \sqrt{2\pi}} e^{-\frac{1}{2} \cdot \frac{1}{1-\frac{1}{2}} \cdot \left\{ \frac{n}{2} \left(\frac{y}{1}\right)^2 + \left(\frac{y}{2}\right)^2 \right\}} - 2 \times \frac{1}{2} \pi \frac{n}{1} \times \frac{y}{2}$$

Am

$$\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} \left\{ \left(\frac{n}{1}\right)^2 \right\}} \quad \forall (n, y)$$

$$\text{or } f_{Y|X}(y|x=x) = \frac{1}{\sqrt{3} \cdot \sqrt{2\pi}} e^{-\frac{1}{2} \cdot \left\{ \left(\frac{n}{\sqrt{3}}\right)^2 + \left(\frac{y}{\sqrt{3}}\right)^2 - 2 \cdot 1 \cdot \frac{n}{\sqrt{3}} \cdot \frac{y}{\sqrt{3}} \right\}}$$

Am

2(c)

$$\text{or } f_{Y|X}(y|x=x) = \sqrt{2\pi} \mathcal{N}(\mu_x=0, \mu_y=0, \sigma_x^2=3, \sigma_y^2=3, \rho_{xy}=1) \quad \text{Am } (m, y)$$

$$\text{or } f_{Y|X}(y|x=x) = \mathcal{N}(y=\cancel{\text{m}}, \sigma_y^2=3) \quad \forall y$$

$$\text{or } f_{Y|X}(y|x=x) = \frac{1}{\sqrt{3}} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} \left\{ \left(\frac{y-n}{\sqrt{3}}\right)^2 \right\}} \quad \forall y$$

2(c)

$$\text{or } f_{Y|X}(y|x=x) = \mathcal{N}(y=\cancel{n}, \mu_y=n, \sigma_y^2=3) \quad \text{Am}$$

PBO.

2(d) Find n s.t. $E[Y|X=n] = -2$.

From 2(c), we know that $Y|X=n$ is fully correlated to $X=n$. ($P_{Y|X=n}=1$).

$\therefore Y$ and X ~~have~~ have the same support/domain of $n \in \mathbb{R}$ and $y \in \mathbb{R}$,

$$\therefore E_Y[Y|X=n] = -2 \equiv E_X[X=n] = -2$$

$$\text{But } E_X[X=n] = n$$

$$\therefore n = -2 \quad \boxed{n = -2} \text{ Ans}$$

2(d) Find n s.t. $E[Y|X=n] = -2$

From 2(c), we know that $Y|X=n$ is a gaussian

with mean n .

$$\therefore E[Y|X=n] = -2 \Rightarrow \boxed{n = -2} \text{ Ans}$$

$$2(e) \quad Z = X + Y - 1$$

Z is also a Gaussian, we need to only compute μ_Z and σ_Z^2

To find $f_Z(z)$: Expectation and Variance

$$E(Z) = E(X) + E(Y) - E(1)$$

$$\text{or } \mu_Z = \mu_X + \mu_Y - 1$$

$$\text{or } \mu_Z = 0 + 0 - 1$$

$$\text{or } \boxed{\mu_Z = -1}$$

$$\sigma_Z^2 = E[(X + Y - 1) - E[X + Y - 1]]^2$$

$$\text{or } \sigma_Z^2 = E[(X - \mu_X + Y - \mu_Y + (-1 - E(-1)))^2]$$

$$\text{or } \sigma_Z^2 = E[(X - \mu_X)^2 + (Y - \mu_Y)^2]$$

$$\text{or } \sigma_Z^2 = E[(X - \mu_X)^2] + E[(Y - \mu_Y)^2] + \cancel{2\text{cov}}[X, Y]$$

$$\text{or } \sigma_Z^2 = \sigma_X^2 + \sigma_Y^2 + 2\rho_{XY}\sigma_X\sigma_Y$$

$$\text{or } \sigma_Z^2 = 1^2 + 2^2 + 2 \times 0.5 \times 1 \times 2$$

$$\text{or } \boxed{\sigma_Z^2 = 7}$$

$$\begin{aligned} &2(e) \\ \therefore &Z \sim \mathcal{N}(3, \mu_Z = -1, \sigma_Z^2 = 7) \end{aligned}$$

2(f)

$$Z = 2X + 3Y$$

2.5

$$W = X - Y$$

$$E(Z) = 2\mu_X + 3\mu_Y$$

$$E(W) = \mu_X - \mu_Y$$

or $\boxed{2(f)(ii)} \boxed{\mu_Z = 0} \text{ Ans}$

or $\boxed{2(f)(ii)} \boxed{\mu_W = 0} \text{ Ans}$

$$\sigma_Z^2 = E[(Z - \mu_Z)^2] = E[(2X + 3Y - (2\mu_X + 3\mu_Y))^2]$$

$$\sigma_W^2 = E[(W - \mu_W)^2] = E[(X - Y - (\mu_X - \mu_Y))^2]$$

$$\text{or } \sigma_Z^2 = E[2(X - \mu_X)^2 + 3(Y - \mu_Y)^2 + 2 \cdot 3 \cdot (X - \mu_X)(Y - \mu_Y)]$$

$$\text{or } \sigma_W^2 = E[(X - \mu_X)^2 + (Y - \mu_Y)^2 - 2 \cdot 1 \cdot (X - \mu_X)(Y - \mu_Y)]$$

$$\text{or } \sigma_Z^2 = 4\sigma_X^2 + 9\sigma_Y^2 + 2 \cdot 2 \cdot 12\rho_{XY}\sigma_X\sigma_Y$$

$$\sigma_W^2 = \sigma_X^2 + \sigma_Y^2 - 2\rho_{XY}\sigma_X\sigma_Y$$

$$\text{or } \sigma_Z^2 = 4 \cdot 1^2 + 9 \cdot 2^2 + 12 \cdot (0.5) \cdot 1 \cdot 2$$

$$\sigma_W^2 = 1^2 + 2^2 - 2 \cdot (0.5) \cdot 1 \cdot 2$$

$$\text{or } \sigma_Z^2 = 4 + 36 + 12$$

$$\sigma_W^2 = 1 + 4 - 2$$

or $\boxed{2(f)(ii)} \boxed{\sigma_Z^2 = 52} \text{ Ans}$

or $\boxed{2(f)(iv)} \boxed{\sigma_W^2 = 3} \text{ Ans}$

or $\text{Cor}(Z, W) = E[(Z - \mu_Z)(W - \mu_W)]$

$$\text{or } \text{Cor}(Z, W) = E[(2X + 3Y - 2\mu_X - 3\mu_Y)(X - Y - \mu_X - \mu_Y)]$$

$$\text{or } \text{Cor}(Z, W) = E[2(X - \mu_X) + 3(Y - \mu_Y) \{ 2(X - \mu_X) - 1(Y - \mu_Y) \}]$$

$$\text{or } \text{Cor}(Z, W) = E[2(X - \mu_X)^2 - 3(Y - \mu_Y)^2 + 2(X - \mu_X)(Y - \mu_Y)]$$

$$\text{or } \text{Cor}(Z, W) = 2\sigma_X^2 - 3\sigma_Y^2 + \rho_{XY}\sigma_X\sigma_Y$$

$$\text{or } \text{Cor}(Z, W) = 2 \cdot 1^2 - 3 \cdot 2^2 + (0.5) \cdot 1 \cdot 2$$

$$\text{or } \text{Cor}(Z, W) = 2 - 12 + 1$$

2(f)(v)

$$\text{con}(z, w) = -9 \quad \text{Ans}$$

$$\rho_{z,w} = \frac{\text{con}(z, w)}{\sigma_z \sigma_w} = \frac{-9}{\sqrt{52} \cdot \sqrt{3}} \approx -0.7206$$

2(f)(vi)

$$f_{z,w}(z, w) = N(\mu_z=0, \mu_w=0, \sigma_z^2=52, \sigma_w^2=3, \rho_{zw}=-0.7206) \quad \text{Ans}$$

$$2(g) R = ax + by \quad \text{and} \quad \sigma_R^2 = a^2 \cdot \sigma_x^2 + b^2 \cdot \sigma_y^2 \\ \text{or} \sigma_R^2 = a^2 \cdot 1^2 + b^2 \cdot 2^2 \\ \sigma_R^2 = a^2 + 4b^2$$

$\text{con}(R, Y) = 0$ is a sufficient condition for R and Y to be independent, as both are gaussian.

$$\text{con}(R, Y) = E[(ax + by) - a\mu_x - b\mu_y](Y - \mu_y)$$

$$\text{or } \text{con}(R, Y) = E[a(X - \mu_x)(Y - \mu_y) + b(Y - \mu_y)^2]$$

$$\text{or } \text{con}(R, Y) = a \rho_{x,y} \sigma_x \sigma_y + b \sigma_y^2$$

$$\text{or } \text{con}(R, Y) = a(0.5) \cdot 1 \cdot 2 + b \cdot 2^2$$

$$\text{or } \text{con}(R, Y) = a + 4b$$

2(g)

For R, Y to be independent distributions, $a + 4b = 0$

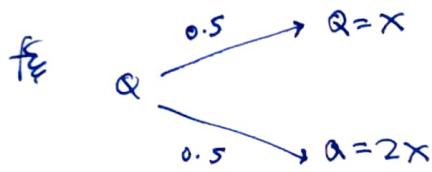
we can use any set of real numbers to do so;

$$\text{say } a = -4, b = 1$$

$$2(b) \quad Q = aX$$

given $P(a=2) = 0.5 \quad f(Q|a=2) = \cancel{2} \cdot \cancel{f_X}(x=2n)$

$$P(a=1) = 0.5 \quad f(Q|a=1) = f_X(x=n)$$



$$f_Q(q) = f_{Q|a}(Q|a=1) \cdot P(a=1) + f_{Q|a}(Q|a=2) \cdot P(a=2) \quad (\text{LTP})$$

$$\text{or } f_Q(q) = 0.5 N(q, \mu_q = \mu_n, \sigma_q^2 = \sigma_x^2) + 0.5 N(q, \mu_q = 2\mu_n, \sigma_q^2 = 2^2 \sigma_x^2)$$

2(b)

$$\text{or } f_Q(q) = 0.5 N(q, \mu_q = 0, \sigma_q^2 = 1) + 0.5 N(q, \mu_q = 0, \sigma_q^2 = 4) \quad \text{Ans}$$

→ X → X → X → X → X

3

 X, Y independent. $Z = X + Y$

31

$$F_Z(z) = P(Z \leq z)$$

$$\text{or } F_Z(z) = P(X+Y \leq z)$$

$$\text{or } F_Z(z) = P(Y \leq z - X)$$

$$\text{or } F_Z(z) = \int_{n=-\infty}^{n=\infty} \int_{y=-\infty}^{y=z-n} f_{X,Y}(n,y) dy dn$$

$$\text{or } F_Z(z) = \int_{n=-\infty}^{n=\infty} \left(\int_{y=-\infty}^{y=z-n} f_X(n) f_Y(y) dy \right) dn$$

$$\text{or } \frac{d}{dz} F_Z(z) = \frac{d}{dz} \left[\int_{n=-\infty}^{n=\infty} f_X(n) \left(\int_{y=-\infty}^{y=z} f_Y(y) dy \right) dn \right] \Big|_{z=3}$$

3(a)

$$\text{or } f_Z(z) = \int_{n=-\infty}^{n=\infty} f_X(n) f_Y(z-n) dn$$

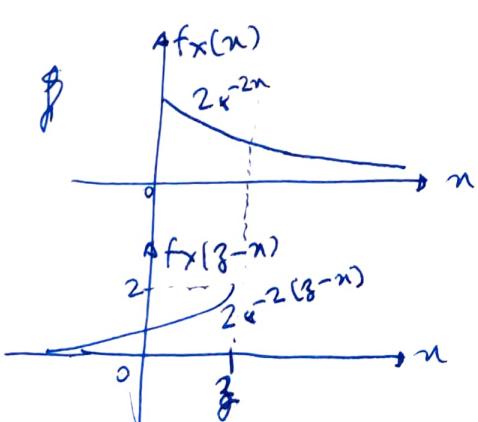
Am
Rig Leibniz rule
Kenne Produkt \odot

3(b)

$$f_Z(z) = \int_{y=-\infty}^{y=\infty} f_Y(y) f_X(z-y) dy$$

Am
Kenne Produkt \odot

3(c) $f_X(n) = 2e^{-2n} \quad f_X(y) = 2e^{-2y}$
 $n \geq 0, \quad y \geq 0$



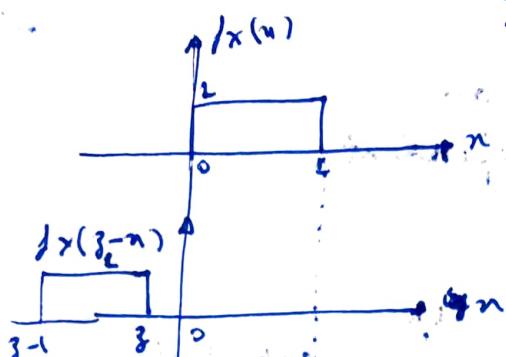
$$f_Z(z) = \int_{n=0}^{n=3} 2e^{-2n} \cdot 2e^{-2(z-n)} dn$$

$$\text{or } f_Z(z) = \int_{n=0}^{n=3} 4e^{-2z} dn$$

3(c)

$$\text{or } f_Z(z) = 4e^{-2z} \cdot 3 \quad z \in (0, \infty)$$

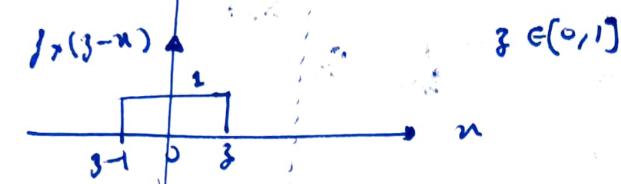
3(d)



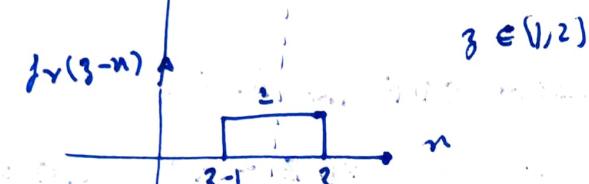
$$X \sim \text{unif}(0,1)$$

$$Y \sim \text{unif}[0,2]$$

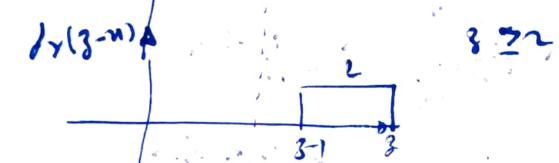
3.2



$$z \in (0,1)$$



$$z \in (1,2)$$



$$z \geq 2$$

$$z \leq 0$$

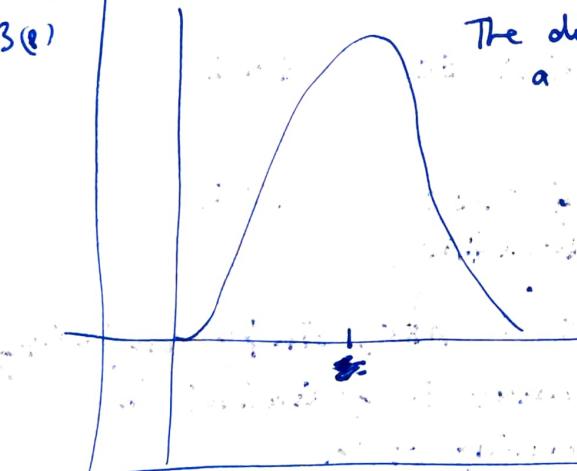
$$z \in (0,1)$$

$$2-z \quad z \in (1,2)$$

$$z \geq 2$$

Ans

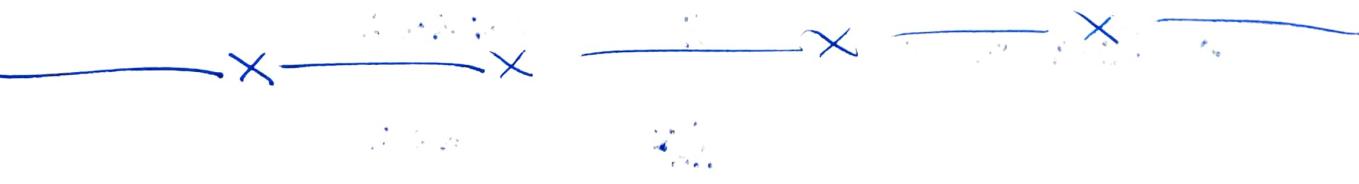
3(e)



The distribution resembles a Gaussian distribution.

$$x_i \in \text{unif}[0,2]$$

x_1, x_2 are independent for $i \neq j$

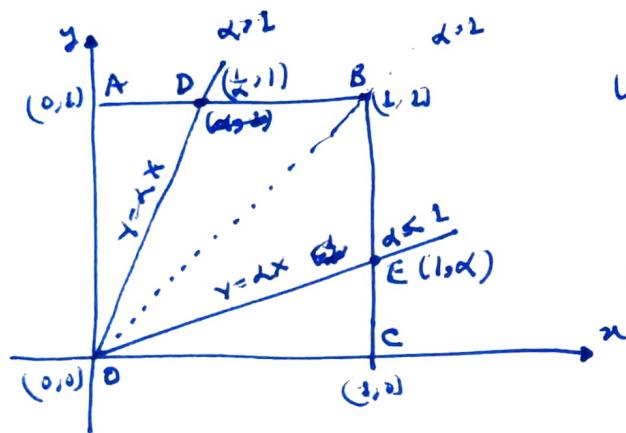
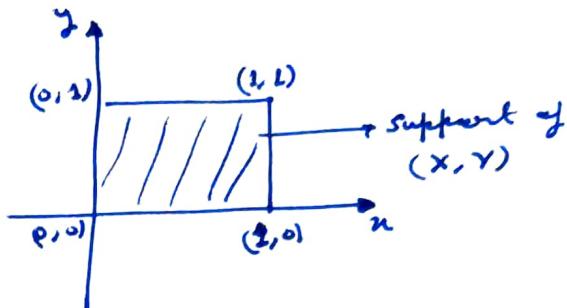
Ans

4. $X \sim \text{unif}[0, 1]$
 $Y \sim \text{unif}[0, 1]$

$$Z = \frac{Y}{X}$$

Ques $\rightarrow f_Z(z) = ?$

X, Y are independent



Using CDF method:

$$F_Z(z) = P(Z \leq z) \quad z \in (0, \infty)$$

$$\text{or } F_Z(z) = P\left(\frac{Y}{X} \leq z\right)$$

$$\text{or } F_Z(z) = P(Y \leq zx)$$

or $F_Z(z) = \begin{cases} \text{Area of } \triangle OEC & z \leq 1 \\ \text{Area of } ODBC \\ = 1 - \text{Area } (AOD) & z \geq 1 \end{cases}$

Note: $\because f_{X,Y}(x,y) = 1 \quad \forall (x,y) \in OABC$, simply computing the area under the curves is equivalent to computing the corresponding CDFs.

or $F_Z(z) = \begin{cases} \frac{1}{2} & 0 < z \leq 1 \\ \frac{1}{2z^2} & z \geq 1 \\ 0 & z < 0 \end{cases}$

4(a)

$$\Rightarrow f_Z(z) = \frac{d}{dz} F_Z(z=d) \Big|_{d=z} = \begin{cases} 0 & z < 0 \\ \frac{1}{2} & z \in (0, 1) \\ \frac{1}{2z^2} & z \geq 1 \end{cases}$$

4.2

Ans

4(b) $E[X^2 + Y^2] = E[X^2] + E[Y^2]$

or $E[X^2 + Y^2] = \sum_{n=0}^{n=1} n^2 \cdot 2 \cdot d_n + \int_{y=0}^{y=1} y^2 \cdot 1 \cdot dy$

or $E[X^2 + Y^2] = \frac{1}{3} + \frac{1}{3}$

4(b) or $E[X^2 + Y^2] = \frac{2}{3}$ Ans

4(c) $f_{X,Z}(n, z) = \begin{cases} f_Z|_{X=n}(z) \cdot f_X(x=n) & n \in (0, 1) \\ 0 & z \in (0, \frac{1}{n}) \end{cases}$

else.

or $f_{X,Z}(n, z) = \begin{cases} \text{pdf of } \text{pdf of } \left(\frac{1}{n} \cdot Y \right) \cdot 1 & n \in (0, 1) \\ 0 & z \in (0, \frac{1}{n}) \end{cases}$

else

But $Y \sim \text{unif}(0, 1)$

$$Z|_{X=n} \sim \frac{1}{n} \cdot \text{unif}(0, 1)$$

$$\therefore Z|_{X=n} \sim \text{unif}\left(0, \frac{1}{n}\right)$$

$\Rightarrow \text{pdf}(Z|_{X=n}) = \begin{cases} n & z \in (0, \frac{1}{n}) \\ 0 & \text{else} \end{cases}$

4.3

4(c) f_{x,z} $f_{x,z}(n, z) = \begin{cases} n & z \in (0, \frac{1}{n}] \\ 0 & \text{else} \end{cases}$ Ans

4(d) $E[XZ] = E[Y] = \frac{1}{2}$

~~Berechnung von $E[XZ]$ mit $\int f_{x,z}(x, z) x z dx dz$~~

4(d) ii $E[XZ] = \frac{1}{2}$ Ans

$\text{Cov}(X, Z) = E[XZ] - \mu_X \mu_Z$

or $\text{Cov}(X, Z) = E[Y] - \frac{1}{2} \mu_Z$

or $\text{Cov}(X, Z) = \frac{1}{2} - \frac{1}{2} \mu_Z$

$\mu_Z = \int f_Z(z) \cdot z \cdot dz$

or $\mu_Z = \int_{z=0}^{z=1} \left(\frac{1}{z}\right) \cdot z \cdot dz + \int_{z=1}^{z=\infty} \left(\frac{1}{2z^2}\right) z \cdot dz$

or $\mu_Z = \frac{1}{4} z^2 \Big|_{z=0}^{z=1} + \frac{1}{2} \ln z \Big|_{z=1}^{z=\infty}$

or $\mu_Z \rightarrow \infty$

4(d) iii $\text{Cov}(X, Z) \rightarrow -\infty$ Ans

$$P_{X,Z}(n, z) = \frac{\text{Cov}(X, Z)}{\sigma_X \sigma_Z}$$

$$\sigma_X = \sqrt{\frac{1}{12}}$$

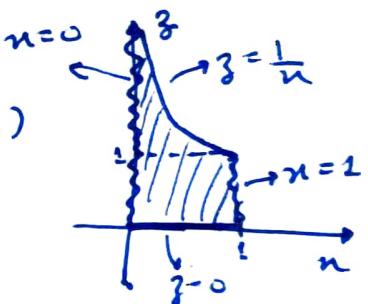
$$\sigma_Z = \int (z - \mu_Z)^2 \cdot f_Z(z) dz \rightarrow \infty$$

\downarrow
 $\mu_Z \rightarrow \infty$

So $P_{X,Z}(n, z) \approx \frac{-\infty}{\sqrt{n} \cdot \infty} \approx -1$

or 4(d) iii) P_{X,Z}(n, z) = -1 Ans

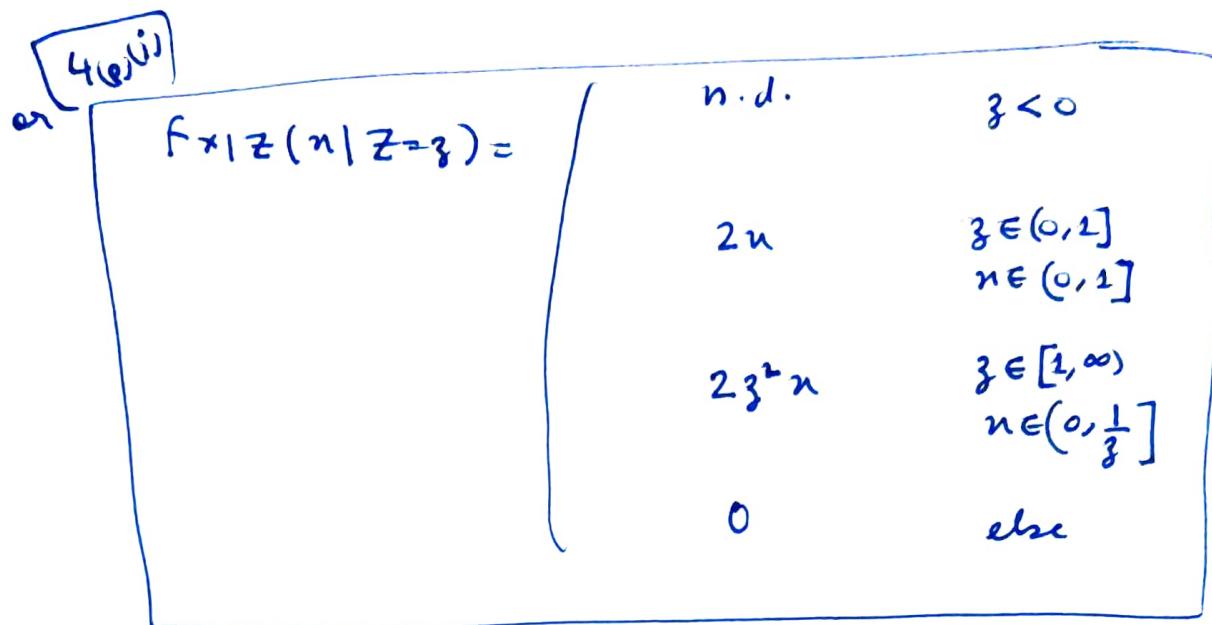
support of (X, Z) :



4(e) $f_{X|Z}(n | Z=z) = \frac{f_{Z|X}(z | X=n) \cdot f_X(X=n)}{f_Z(z)}$

or $f_{X|Z}(n | Z=z) = \begin{cases} n.d. & z < 0 \text{ or } z \geq \frac{1}{n} \\ \frac{n+1}{2} & z \in (0, 1] \\ \frac{n+1}{2z^2} & n \in (0, \frac{1}{z}] \\ 0 & \text{else} \end{cases}$

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$$\mathbb{E}[X|Z=g] = \int_{n=0}^{n=2} n \cdot f_{x|Z}(x|Z=g) dn$$

or $\mathbb{E}[X|Z=g] = \begin{cases} \int_{n=0}^{n=2} n \cdot 2n dn & g \in (0, 1] \\ \int_{n=0}^{n=\frac{1}{g}} n \cdot 2g^2 n dn & g \in [1, \infty) \end{cases}$

4(iv)

or $\mathbb{E}[X|Z=g] = \begin{cases} \frac{2}{3} & g \in (0, 1] \\ \frac{2}{3g} & g \in [1, \infty) \end{cases}$

Ans

$$4(f) \quad E_Z [E_X [x | Z=3]]$$

$$= \int_{-\infty}^{\infty} E_X [x | Z=3] \cdot f_Z(z) dz$$

$$\text{or } E_Z [E_X [x | Z=3]] = \int_{z=2}^{z=1} \frac{2}{3} \cdot \frac{1}{2} dz + \int_{z=1}^{z=\infty} \frac{2}{3z} \cdot \frac{1}{2z^2} dz$$

$$\text{or } E_Z [E_X [x | Z=3]] = \int_{z=0}^{z=1} \frac{1}{3} dz + \int_{z=1}^{z=\infty} \frac{1}{3} z^{-3} dz$$

$$\text{or } E_Z [E_X [x | Z=3]] = \frac{1}{3} + \int_{z=\infty}^{z=1} \frac{1}{6} z^{-2} dz$$

$$\text{or } E_Z [E_X [x | Z=3]] = \frac{1}{3} + \frac{1}{6}$$

4(f)

$$\text{or } E_Z [E_X [x | Z=3]] = \frac{1}{2} = E[X]$$

Hence Verified!



Ans

$$4(g) \quad E[xz | Z=3] = E[zx | Z=3]$$

$$\text{or } E[xz | Z=3] = 3 E[x | Z=3]$$

4(g)

$$E[xz | Z=3] = \begin{cases} \frac{2}{3} z & z \in [0, 1] \\ \frac{2}{3} & z \in [1, \infty) \end{cases}$$

Ans

Veify (optional) by $E_Z [E_Y [Y | Z=3]] = E_Y [Y]$

$$E_Z [E_Y [Y | Z=3]] = \int_{z=0}^{z=2} \frac{2}{3} z \cdot \frac{1}{2} dz + \int_{z=2}^{z=\infty} \frac{2}{3} \cdot \frac{1}{2z^2} dz$$

$$\text{or } E_Z [E_Y [Y | Z=3]] = \left. \frac{1}{3} \frac{z^2}{2} \right|_0^1 + \left. \frac{1}{3z} \right|_2^\infty$$

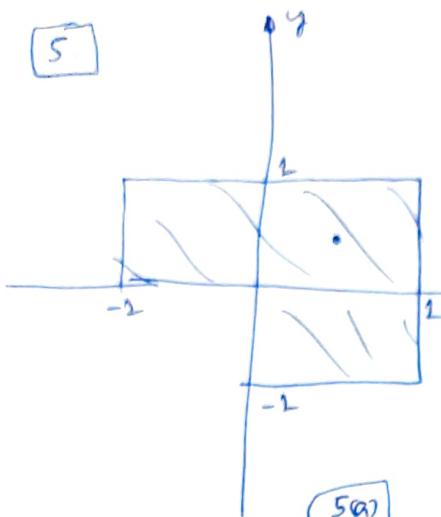
$$\text{or } E_Z [E_Y [Y | Z=3]] = \frac{1}{6} + \frac{1}{3} = \frac{1}{2} = E[Y]$$

Hence Verified!

X X X X X ☺

5

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* X, Y uniform in shaded region.

$$f_{X,Y}(x,y) = \frac{1}{3} \text{ in shaded region}$$

$$\mu_x = \int_{n=-1}^{n=0} \int_{y=0}^{y=1} \frac{1}{3} n dy dn + \int_{n=0}^{n=1} \int_{y=-1}^{y=1} \frac{1}{3} n dy dn$$

5(a)
or $\mu_x = \frac{1}{6}$

5(b)
 $\Rightarrow \mu_y = \frac{1}{6}$ (by symmetry)

$$\mu_{xy} = \int_{-1}^0 \int_{y=0}^1 \frac{1}{3} ny dy dn + \int_0^1 \int_{y=-1}^1 \frac{1}{3} ny dy dn$$

5(c)
or $\mu_{xy} = -\frac{1}{12}$

5(d)
 $\sigma_x^2 = \mu_{x^2} - (\mu_x)^2$

$$\mu_{x^2} = \int \int \frac{1}{3} n^2 dy dn = \frac{1}{3}$$

$$\hat{\sigma}_x^2 = \frac{1}{3} - \left(\frac{1}{6}\right)^2$$

5(e)
or $\hat{\sigma}_x^2 = \frac{11}{36}$

5(f)
 $\Rightarrow \hat{\sigma}_x^2 = \frac{11}{36}$ (by symmetry)

5(f) $\text{Corr}(X, Y) = E[XY] - E[X] \cdot E[Y]$

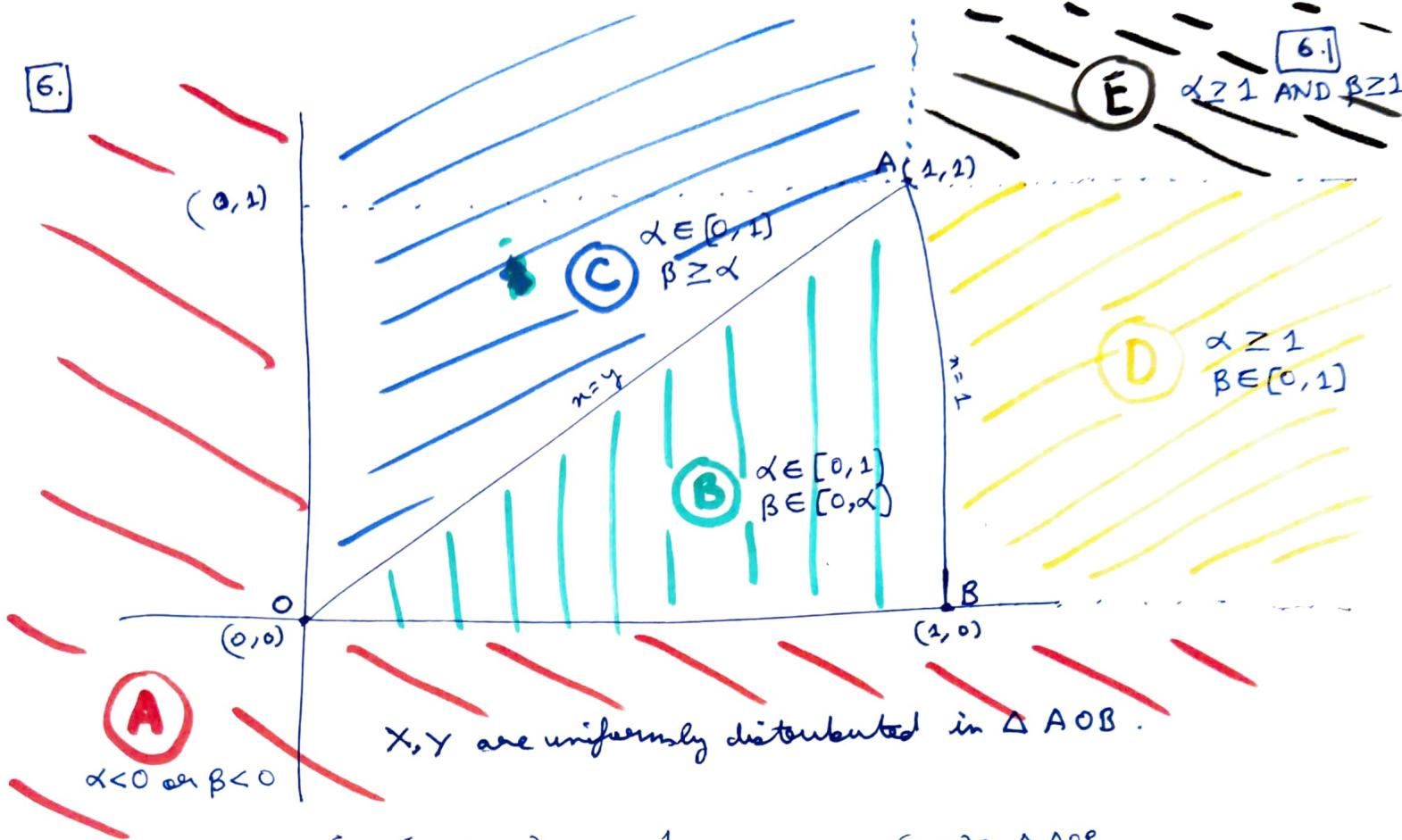
$$\text{or } \text{Corr}(X, Y) = -\frac{1}{12} - \frac{1}{36} \left(\frac{1}{6}\right)^2$$

5(g)
or $\text{Corr}(X, Y) = -\frac{1}{9}$

$$\Rightarrow \rho_{XX} = \frac{\text{Corr}(X, X)}{\hat{\sigma}_x \hat{\sigma}_x} = -\frac{4}{11}$$

or $\rho_{XX} \approx -0.3636$

6.



$$\text{So } f_{x,y}(n, y) = \frac{1}{\text{Area of } \triangle AOB} \quad (n, y) \in \triangle AOB$$

or $f_{x,y}(n, y) = \frac{1}{\frac{1}{2} \times 1 \times 1}$

~~$y \in [0,1]$~~
 $y \in [0,1]$
 $n \in [y,1]$

6(a)

$f_{x,y}(n, y) = \begin{cases} 2 & \text{if } y \in [0,1] \\ n \in [y,1] \end{cases}$
$= 0 \quad \text{else}$

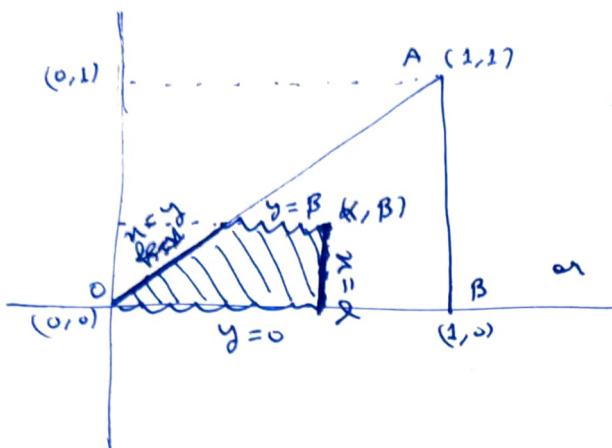
Ans

6(b) (Calculations on next pages.)

$F_{x,y}(n=d, y=\beta) =$	$\left\{ \begin{array}{ll} 0 & \alpha < 0 \text{ and } \beta < 0 \\ 2\alpha\beta - \beta^2 & \alpha \in [0,1], \beta \in [0, \alpha] \\ \alpha^2 & \alpha \in [0,1], \beta = \alpha \\ 2\beta - \beta^2 & \alpha \geq 1, \beta \in [0, \alpha] \\ 1 & \alpha \geq 1, \beta \geq 2 \end{array} \right.$	A
		C
		D
		E

For (α, β) in **B**:

i.e. $\alpha \in [0, 1], \beta \in [0, \alpha]$



~~$F_{X,Y}(x,y)$~~

$$F_{X,Y}(n=\alpha, y=\beta) = \iint_{\text{shaded region}} 2 dndy \quad (\alpha, \beta) \in \textcircled{B}$$

$$\text{or } F_{X,Y}(n=\alpha, y=\beta) = \iint_{y=0}^{y=\beta} \iint_{n=\alpha}^{n=1} 2 dn dy \quad (\alpha, \beta) \in \textcircled{B}$$

$$\text{or } F_{X,Y}(\alpha, \beta) = \int_{y=0}^{y=\beta} 2 \alpha \Big|_y^1 dy \quad (\alpha, \beta) \in \textcircled{B}$$

$$\text{or } F_{X,Y}(\alpha, \beta) = \int_{y=\alpha}^{y=\beta} 2(\alpha - y) dy \quad (\alpha, \beta) \in \textcircled{B}$$

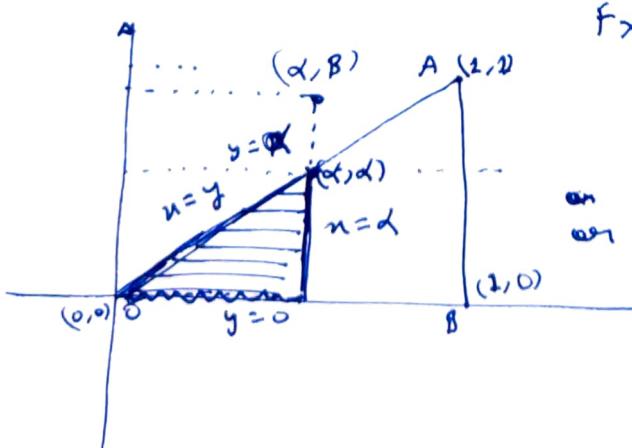
$$\text{or } F_{X,Y}(\alpha, \beta) = 2\alpha y - y^2 \Big|_0^\beta \quad (\alpha, \beta) \in \textcircled{B}$$

$$\text{or } \boxed{F_{X,Y}(\alpha, \beta) = 2\alpha\beta - \beta^2 \quad (\alpha, \beta) \in \textcircled{B}}$$

or $\alpha \in [0, 1]$
 $\beta \in [0, \alpha]$

For (α, β) in **C**:

i.e. $\alpha \in [0, 1], \beta \geq \alpha$



$$f_{X,Y}(n=\alpha, y=\beta) = \iint_{\text{shaded region}} 2 dndy \quad (\alpha, \beta) \in \textcircled{C}$$

$$\text{or } F_{X,Y}(n=\alpha, y=\beta) = \iint_{y=0}^{y=\alpha} \iint_{n=\alpha}^{n=\beta} 2 dn dy \quad (\alpha, \beta) \in \textcircled{C}$$

or $F_{X,Y}(x=\alpha, y=\beta) = \int_{y=0}^{y=\alpha} 2(\alpha-y) dy \quad (\alpha, \beta) \in \textcircled{C}$

or $F_{X,Y}(x=\alpha, y=\beta) = 2\alpha^2 - \alpha^2 \quad (\alpha, \beta) \in \textcircled{C}$

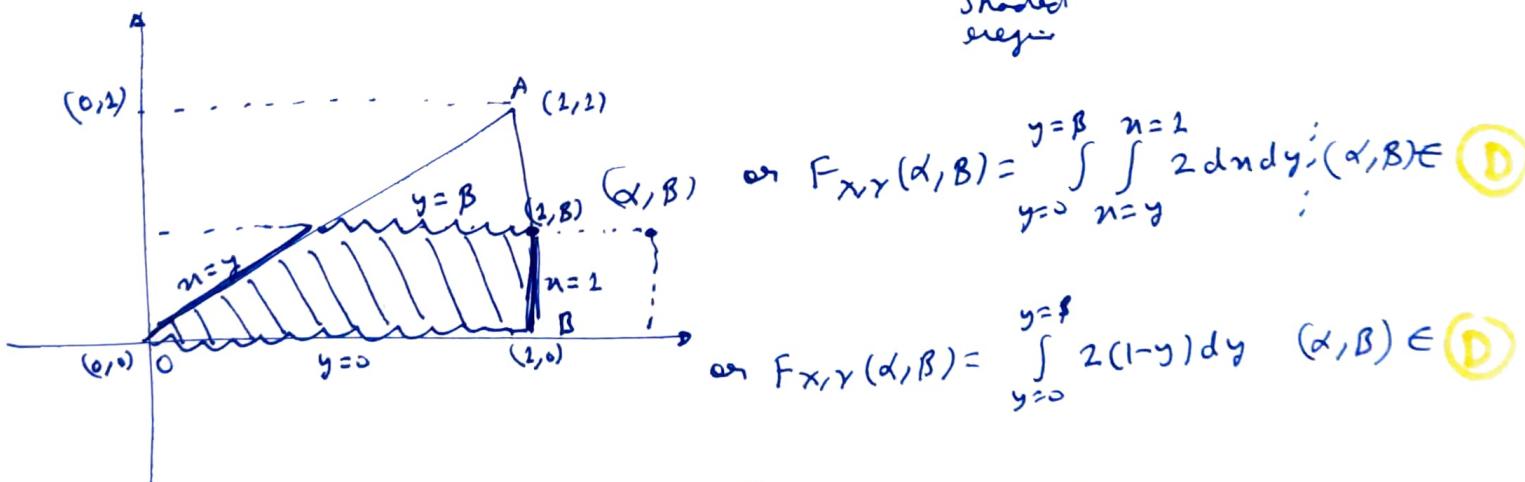
or $F_{X,Y}(x=\alpha, y=\beta) = \alpha^2 \quad (\alpha, \beta) \in \textcircled{C}$
 or $\alpha \in [0, 1], \beta \leq \alpha$

Area of the shaded Δ .
 * $f_{X,Y}(x,y)$ in the region.

For $(\alpha, \beta) \in \textcircled{D}$

i.e. $\alpha \geq 1, \beta \in [0, 1]$

$F_{X,Y}(x=\alpha, y=\beta) = \iint_{\text{shaded reg}} 2 dx dy \quad (\alpha, \beta) \in \textcircled{D}$



or $F_{X,Y}(\alpha, \beta) = 2y - y^2 \Big|_0^\beta \quad (\alpha, \beta) \in \textcircled{D}$

or $F_{X,Y}(\alpha, \beta) = 2\beta - \beta^2 \quad (\alpha, \beta) \in \textcircled{D}$
 or $\alpha \geq 1, \beta \in [0, 1]$

$$f_{X,Y}(n) = \int_{-\infty}^{\infty} f_{X,Y}(n,y) dy$$

$$\text{or } f_{X,Y}(n) = \begin{cases} 0 & n < 0 \\ \int_{y=0}^{y=n} 2 dy & n \in [0, 2] \\ 0 & n > 2 \end{cases}$$

6(c)

$$\text{or } f_X(n) = \begin{cases} 0 & n < 0 \\ 2n & n \in [0, 2] \\ 0 & n > 2 \end{cases} \quad \underline{\text{Ans}}$$

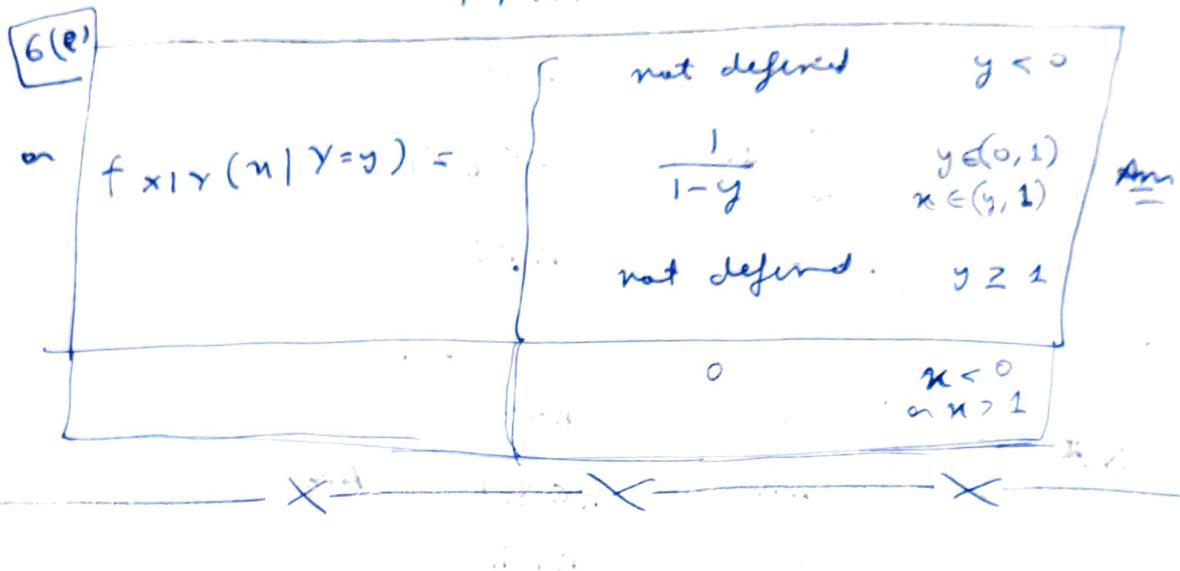
$$f_{Y|X}(y) = \int_{-\infty}^{\infty} f_{X,Y}(n,y) dn$$

$$\text{or } f_Y(y) = \begin{cases} 0 & y < 0 \\ \int_{n=y}^{n=1} 2 dn & y \in [0, 2] \\ 0 & y > 2 \end{cases}$$

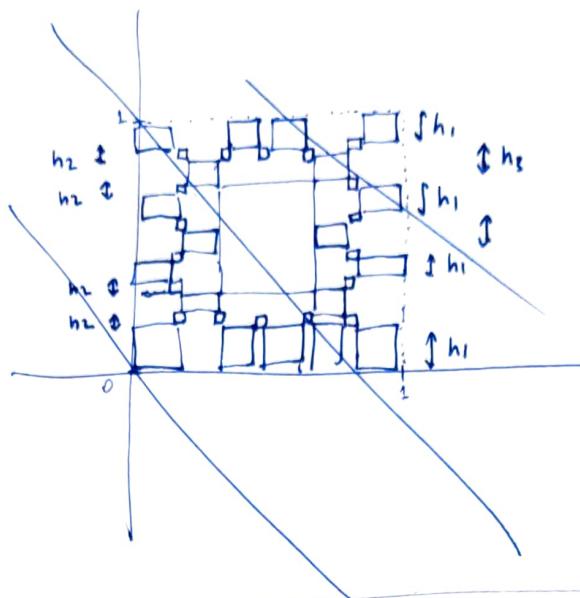
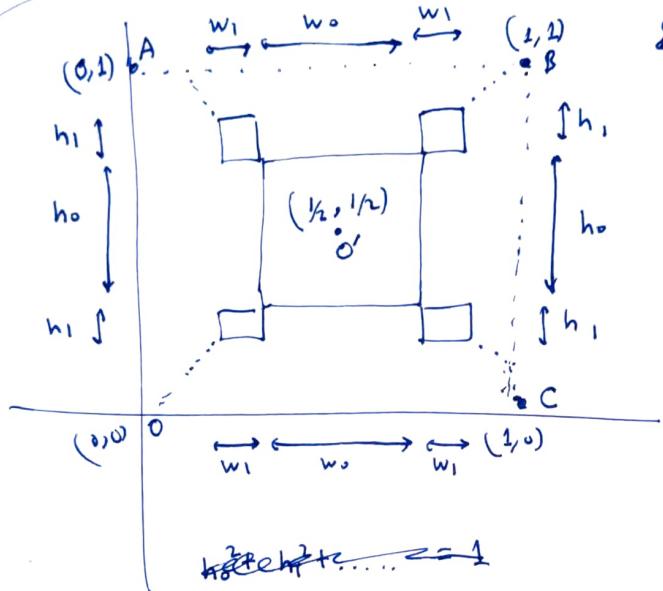
6(d)

$$\text{or } f_Y(y) = \begin{cases} 0 & y < 0 \\ 2(1-y) & y \in [0, 2] \\ 0 & y > 2 \end{cases} \quad \underline{\text{Ans}}$$

$$f_{X|Y}(x|y) = \frac{f_{XY}(x,y)}{f_Y(y)}$$



7.

Ans

~~h_0^2 + h_1^2 + h_2^2 + ... = 1~~

but ~~h_0 + h_1 + h_2 + ... = 1~~

where $h_0 + h_1 + h_2 + \dots = 1$

$$(uniform pdf) \times \{h_0 w_0 + h_1 w_1 + h_2 w_2 + \dots\} = 1$$

In each square OABC of size 1×1 unit squares, any pdf distribution which is

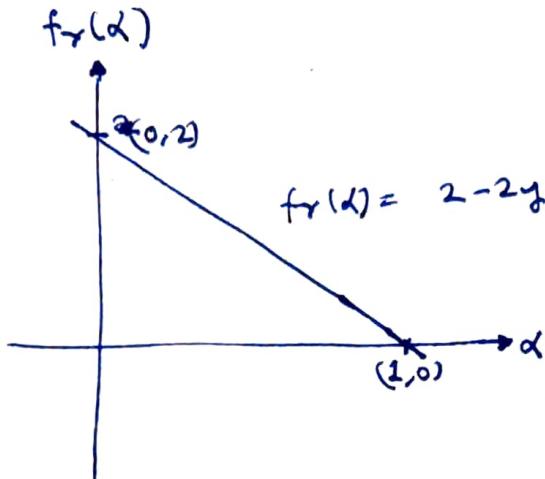
- 1) symmetric about $(\frac{1}{2}, \frac{1}{2})$.
- 2) NOT a single rectangle.
- 3) ~~filled with-~~ a set of more than one (possibly infinite) rectangles which are touching together (to maintain continuity in both x and y) as well as are symmetrically placed around $(\frac{1}{2}, \frac{1}{2})$ (to maintain zero covariance), and

can qualify for a distribution rel in x, y s.t.

- 1) X and Y are uncorrelated.
- 2) X and Y are NOT independent.
- 3) $X \sim \text{unif}(0,1)$ and $Y \sim \text{unif}(0,1)$



8.



$$8(a) \hat{Y}_{MP} = \underset{x}{\operatorname{arg\,min}} (f_Y(x)) \Big|_{x \in [0, 1]}$$

$$\text{or } \hat{Y}_{MP} = \underset{x}{\operatorname{arg\,min}} (2 - 2x) \Big|_{x \in [0, 1]}$$

$$\boxed{\begin{array}{l} 8(a) \\ \text{or } \hat{Y}_{MP} = 0 \end{array}} \quad \text{Ans for which } f_Y(x=0) = 2$$

Ans

$$8(b) \hat{Y}_{MMSE} = \underset{x}{\operatorname{arg\,min}} \cdot [E[(Y-x)^2]]$$

$$\text{We know that } \hat{Y}_{MMSE} = \mu_Y = E(Y)$$

$$\text{or } \hat{Y}_{MMSE} = \int_{y=0}^{y=1} y \cdot f_Y(y) dy$$

$$\text{or } \hat{Y}_{MMSE} = \int_{y=0}^{y=1} y \cdot (2 - 2y) dy$$

$$\text{or } \hat{Y}_{MMSE} = \int_{y=0}^{y=1} 2y - 2y^2 dy$$

$$\text{or } \hat{Y}_{MMSE} = \left[y^2 - \frac{2}{3}y^3 \right]_{y=0}^{y=1}$$

$$\boxed{\begin{array}{l} 8(b) \\ \text{or } \hat{Y}_{MMSE} = \frac{1}{3} \end{array}} \quad \text{Ans}$$

None the less, let us derive \hat{y}_{MMSE} using only first principles.

$$\hat{y}_{MMSE} = \underset{\alpha}{\operatorname{argmin}} \left[E[(y-\alpha)^2] \right]$$

$$\text{or } \hat{y}_{MMSE} = \underset{\alpha}{\operatorname{argmin}} \left[\int_{y=0}^{y=1} (y-\alpha)^2 \cdot (2-2y) dy \right]$$

$$\text{or } \hat{y}_{MMSE} = \underset{\alpha}{\operatorname{argmin}} \left[2 \int_{y=0}^{y=1} \left\{ y^2 - 2\alpha y + \alpha^2 \right\} \{1-y\} dy \right]$$

$$\text{or } \hat{y}_{MMSE} = \underset{\alpha}{\operatorname{argmin}} \left[2 \int_{y=0}^{y=1} \left\{ -y^3 + (1+2\alpha)y^2 - (\alpha^2 + 2\alpha)y + \alpha^2 \right\} dy \right]$$

$$\text{or } \hat{y}_{MMSE} = \underset{\alpha}{\operatorname{argmin}} \left[2 \left[-\frac{y^4}{4} + \frac{(1+2\alpha)}{3} y^3 - \frac{(\alpha^2+2\alpha)}{2} y^2 + \alpha^2 y \right] \Big|_{y=0}^{y=1} \right]$$

$$\text{or } \hat{y}_{MMSE} = \underset{\alpha}{\operatorname{argmin}} \left[2 \left[-\frac{1}{4} + \frac{(1+2\alpha)}{3} - \frac{(\alpha^2+2\alpha)}{2} + \alpha^2 \right] \right]$$

$$\text{or } \hat{y}_{MMSE} = \underset{\alpha}{\operatorname{argmin}} \left[2 \left[\frac{\alpha^2}{2} - \frac{1}{3}\alpha + \frac{1}{12} \right] \right]$$

$$\text{or } \hat{y}_{MMSE} = \underset{\alpha}{\operatorname{argmin}} \left[\left(\alpha - \frac{1}{3} \right)^2 + \frac{1}{18} \right]$$

8(b)

$\hat{y}_{MMSE} = \frac{1}{3}$	Ans
with error = $\frac{1}{18}$	

$$\hat{y}_{\text{MMAE}} = \underset{\alpha}{\operatorname{arg\,min}}(8(c)) = \underset{\alpha}{\operatorname{arg\,min}} \left(\int_{y=0}^{y=\alpha} (\alpha - y)(2 - 2y) dy + \int_{y=\alpha}^{y=1} (y - \alpha)(2 - 2y) dy \right)$$

$$\text{or } \hat{y}_{\text{MMAE}} = \underset{\alpha}{\operatorname{arg\,min}} \left(2 \int_{y=0}^{y=\alpha} (\alpha - y)(y - 1) dy + 2 \int_{y=1}^{y=\alpha} (y - \alpha)(y - 1) dy \right)$$

$$\text{or } \hat{y}_{\text{MMAE}} = \underset{\alpha}{\operatorname{arg\,min}} \left(2 \int_{y=0}^{y=\alpha} \{y^2 - (\alpha+1)y + \alpha\} dy + 2 \int_{y=1}^{y=\alpha} \{y^2 - (\alpha+1)y + \alpha\} dy \right)$$

$$\text{or } \hat{y}_{\text{MMAE}} = \underset{\alpha}{\operatorname{arg\,min}} \left(4 \left[\frac{y^3}{3} - \frac{(\alpha+1)y^2 + \alpha y}{2} \right] \Big|_{y=\alpha} - 2 \left[\frac{y^3}{3} - \frac{(\alpha+1)y^2 + \alpha y}{2} \right] \Big|_{y=1} - 2 \left[\frac{y^3}{3} - \frac{(\alpha+1)y^2 + \alpha y}{2} \right] \Big|_{y=0} \right)$$

$$\text{or } \hat{y}_{\text{MMAE}} = \underset{\alpha}{\operatorname{arg\,min}} \left(4 \left[\frac{\alpha^3}{3} - \frac{(\alpha+1)\alpha^2 + \alpha^2}{2} \right] - 2 \left[\frac{1}{3} - \frac{(\alpha+1)}{2} + \alpha \right] \right)$$

$$\text{or } \hat{y}_{\text{MMAE}} = \underset{\alpha}{\operatorname{arg\,min}} \left(4 \left[-\frac{\alpha^3}{6} + \frac{\alpha^2}{2} \right] - 2 \left[\frac{\alpha}{2} - \frac{1}{6} \right] \right)$$

$$\text{or } \hat{y}_{\text{MMAE}} = \underset{\alpha}{\operatorname{arg\,min}} \left(-\frac{4}{6} \alpha^3 + 2\alpha^2 - \alpha + \frac{1}{6} \right)$$

$$\text{or } \hat{y}_{\text{MMAE}} = \underset{\alpha}{\operatorname{arg}} \left(\frac{d}{d\alpha} \left[-\frac{2}{3} \alpha^3 + 2\alpha^2 - \alpha + \frac{1}{3} \right] = 0 \right)$$

$$\text{or } \hat{y}_{\text{MMAE}} = \underset{\alpha}{\operatorname{arg}} \left(-2\alpha^2 + 4\alpha - 1 = 0 \right)$$

$$\text{or } \hat{y}_{\text{MMAE}} = \underset{\alpha}{\operatorname{arg}} \left(-2(\alpha - 1)^2 + 1 = 0 \right)$$

$$\text{or } \hat{y}_{\text{MMAE}} = \underset{\alpha}{\operatorname{arg}} \left(\alpha(\alpha - 1)^2 = \frac{1}{2} \right)$$

$\therefore \hat{y}_{\text{MMAE}} = 1 - \frac{1}{\sqrt{2}}$

$\text{or } \boxed{\hat{y}_{\text{MMAE}} \approx 0.293}$ Ans



From BC formulas, we require:

9.1

$\boxed{9.1} \quad Y|X \sim \text{exp}\left(\frac{1}{n^2}\right) \Rightarrow f_{Y|X}(y|x=n) = \frac{1}{n^2} e^{-\frac{1}{n^2}y} \quad y \geq 0$

$X \sim \text{unif}(0,1) \Rightarrow f_X(n) = 1 \quad n \in [0,1]$

$\boxed{g(a)} \quad \hat{Y}|X=n \underset{\text{MMSE}}{=} M_{Y|X} = n^2 \quad \text{Ans}$

$\boxed{g(b)} \quad \text{var. } \hat{Y}|X = \sigma^2_{Y|X} = n^4$

or std. dev. $\hat{Y}|X = \sigma_{Y|X} = n^2$

overall var. = $\int_{n=0}^{\infty} \sigma^2_{Y|X} d\pi =$

or overall var. = $\int_{n=0}^{\infty} n^4 \cdot 1 \cdot d\pi$

$\boxed{g(c)}$

or overall var. = $\frac{1}{5} \quad \text{Ans}$

$\hat{Y}^2|X \underset{\text{MMSE}}{=} E(Y^2|X=n)$

or $\hat{Y}^2|X \underset{\text{MMSE}}{=} \int_{y=0}^{y=\infty} \frac{1}{n^2} e^{-\frac{1}{n^2}y} \cdot y^2 dy$

or $\hat{Y}^2|X \underset{\text{MMSE}}{=} (y^3 + 2n^2y + 2n^4) e^{-\frac{1}{n^2}y} \Big|_{y=\infty}^{y=0}$

$\boxed{g(d)}$

or $\hat{Y}^2|X \underset{\text{MMSE}}{=} 2n^4 \quad \text{Ans} \neq (\hat{Y}|X \underset{\text{MMSE}}{=})^2$

Mean of squares \neq Square of mean $\quad \text{Ans}$

Ans

$$\begin{aligned} & \boxed{g(e)} \quad A\hat{y}^2 e^{-Ay} \\ &= \frac{d}{dy} [(C\hat{y}^2 + Dy + E)e^{-Ay}] \end{aligned}$$

$$\Rightarrow -CA = A$$

$$2C - AD = 0$$

$$D - AE = 0$$

$$\Rightarrow C = -1$$

$$D = -2n^2$$

$$E = 2n^4$$

$$\begin{aligned} & \Rightarrow \int A\hat{y}^2 e^{-Ay} \\ &= (y^3 + 2n^2y + 2n^4) e^{-\frac{1}{n^2}y} \end{aligned}$$

g(c)

$$f_{X|Y=y}(x|Y=y) = \frac{f_{Y|X}(y|X=x) f_X(x)}{f_Y(y)}$$

or $f_{X|Y}(x|Y=y) = \left(\frac{1}{n^2} e^{-\frac{1}{n^2} y} \right) \cdot (1)$

$\int_{y=0}^{y=\infty} \frac{1}{n^2} e^{-\frac{1}{n^2} y} dy$

 $y \in [0, \infty)$
 $n \in (0, \infty)$

g(c) $\mu_{X|Y} = \sum_{n=0}^{n=2} n \cdot f_{X|Y}(x|Y=y) dn$ $\hat{\quad}$

g(d) $\hat{y}|x^m$ LMMSE = $a^* n + b^*$ where,

or $a^* = \frac{\text{Cov}(X, Y)}{\sigma_x^2}$

$$\text{or } a^* = \frac{\mu_{XY} - \mu_X \cdot \mu_Y}{\sigma_x^2}$$

$$\text{or } a^* = \frac{\sum_{n=0}^{n=1} \int_{y=0}^{y=\infty} ny \cdot \frac{1}{n^2} e^{-\frac{1}{n^2} y} dy dn - \frac{1}{2} \sum_{n=0}^{n=1} \int_{y=0}^{y=\infty} y \cdot \frac{1}{n^2} e^{-\frac{1}{n^2} y} dy dn}{(\frac{1}{12})}$$

or ~~$a^* = \frac{1}{n^2} \int_{y=0}^{y=\infty} y \cdot \frac{1}{n^2} e^{-\frac{1}{n^2} y} dy$~~

$$\text{or } a^* = \frac{\sum_{n=0}^{n=1} \frac{1}{n} \left(\int_{y=0}^{y=\infty} y \cdot e^{-\frac{1}{n^2} y} dy \right) dn - \frac{1}{2} \sum_{n=0}^{n=1} \frac{1}{n^2} \left(\int_{y=0}^{y=\infty} e^{-\frac{1}{n^2} y} dy \right) dn}{1/12}$$

$$\text{or } a^* = \frac{\sum_{n=0}^{n=1} \frac{1}{n} \cdot \left(\left. \left(ny + \frac{n^2}{2} y^2 \right) \right|_{y=0}^{y=\infty} \right) dn - \frac{1}{2} \sum_{n=0}^{n=1} \frac{1}{n^2} \left(\left. \left(ny + \frac{n^2}{2} y^2 \right) e^{-\frac{1}{n^2} y} \right|_{y=0}^{y=\infty} \right) dn}{1/12}$$

Hilf Rechenregel

$$\text{or } a^* = \frac{\int_{n=2}^{n=1} \frac{1}{n} \cdot (n^4) dn - \frac{1}{2} \int_{n=0}^{n=1} \frac{1}{n^2} \cdot (n^4) dn}{1/n}$$

$$\text{or } a^* = \frac{\frac{1}{4} - \frac{1}{2} \cdot \frac{1}{3}}{\frac{1}{n}}$$

$$\text{or } \boxed{a^* = 2}$$

$$b^* = \mu_x - a^* m_x$$

$$\text{or } b^* = \frac{1}{3} - 2 \cdot \frac{1}{2}$$

$$\text{or } \boxed{b^* = -\frac{1}{6}}$$

$g(d)$

$$\hat{x}|x_{\text{LMMSE}} = 1 \cdot n - \frac{1}{6} \quad \text{Ans}$$

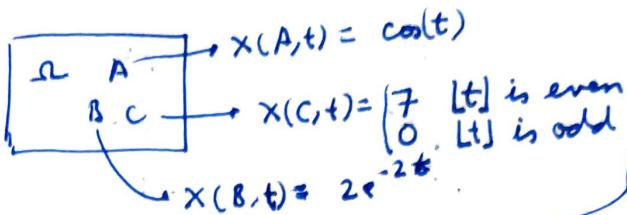
— x — x — x — x —

20.

10(a)

10-1

Ques Mapping of outcomes of an experiment to signals.

Ans

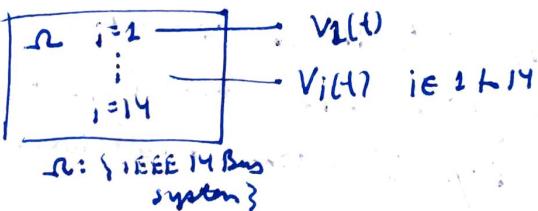
where,

$[t]$: Floor function or greatest integer less than or equal to the t .

10(b)

Virtually any physical phenomenon or simulated phenomenon can be represented as a Random Process.

Eg 1: Bus voltages $V_i(t)$ are time varying signals time varying signals, and there are multiple buses in a power system.



Being able to estimate / predict these processes is in most cases beneficial to humans.

Eg 2: if tomorrow's weather could be predicted using local temperatures, humidity and others, with reasonable accuracy, it would help us all in determining the optimal clothes to wear, time to travel, mode of transport (bus if snowing, car on a sunny day, etc.).

Ans

10(c)

Random Process:

US Daily Power generation MWh:

Percentage values of biggest 5 contributors:

	12 Nov 2022	13 Nov 2022	14 Nov 2022
Natural gas	40	38	37
Coal	19	22	20
Nuclear	16	14	16
Wind	14	12	16
Hydro	6	7	6

Ans

10(d)

Some advantages of modelling $n_i(t)$ instead of $n_i(t=T)$ separately for several T :

- > Our interest is in determining how a particular process varies with time. ~~It makes sense for the sampling to be done in a single experiment with identical conditions.~~

$$n_i(t) \uparrow \text{as } t \uparrow \text{as}$$

Eg... Theory of Critical Slowing Down says that the autocorrelation and variance of the state variables of a dynamical system \uparrow as the dynamical system approaches a catastrophic point w.r.t. its stability.

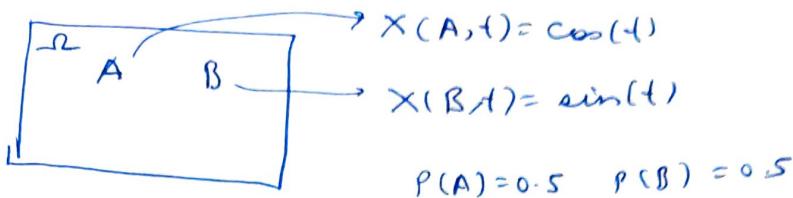
- > Similarly, we may be interested in cross correlation b/w different sysys w.r.t. time as a parameter

Eg. Electricity prices vs Food demand, which themselves are functions of time.

Ans

[11]

[11.1]



$$11.1) \quad f_{X(t)}^{(n)} = \underset{\cancel{X \in A}}{\delta(n - X(A,t))} P(A) + \underset{\cancel{X \in B}}{\delta(n - X(B,t))} P(B)$$

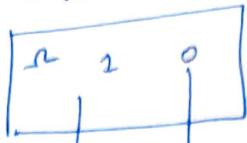
$$+ P(A) \cdot \delta(n - \cos(t)) + P(B) \cdot \delta(n - \sin(t))$$

or

$$11.1) \quad f_{X(t)}^{(n)} = 0.5 \delta(n - \cos(t)) + 0.5 \delta(n - \sin(t)) \quad \text{Ans}$$

— x — x — x —

12.

 $x(k) =$ 

$$P(x(k)=0) = 0.6$$

$$P(x(k)=1) = 0.4$$

$$\text{Def } f_{x(k)}(n, k) = \sum_{n=0}^k 0.6^k \cdot 0.4^{n-k}$$

$$12(c) \quad Y(k) = \sum_{i=0}^{k-1} x(i)$$

$$Y(k) = \begin{cases} 0 & \text{w.p. } \\ 1 & \text{w.p. } \\ \vdots & \\ k & \text{w.p.} \end{cases}$$

$$\begin{aligned} & k \in \{0, 1, 2, \dots\} \\ & k \in \{0, 1, 2, \dots\} \\ & k \in \{0, 1, 2, \dots\} \end{aligned}$$

$$\text{z.B. } x(k) = i \quad \text{w.p. } k \in \{0, 1, \dots, k\}$$

$$12(c) \quad \Rightarrow f_{Y(k)}(y, k) = \sum_{i=0}^k k \cdot (0.6)^{k-i} (0.4)^i \delta(y-i)$$

— X — X — X — X —