

Let S be a sample space, a map $X: S \rightarrow \mathbb{R}^n$,
 $x(w) = \{x_1(w), x_2(w), \dots, x_n(w)\}$ where
 w is an outcome called n -dimensional
random - vector or Random variable if
each x_i is a random variable on S .

Two-dimensional RV :-

Let X_1 and X_2 be two random variable
then a function $z = (X_1, X_2)$ that assigns a
point in \mathbb{R}^2 is called two dimensional
Random Variable.

$$\leftarrow z(w) = [x_1(w), x_2(w)]$$

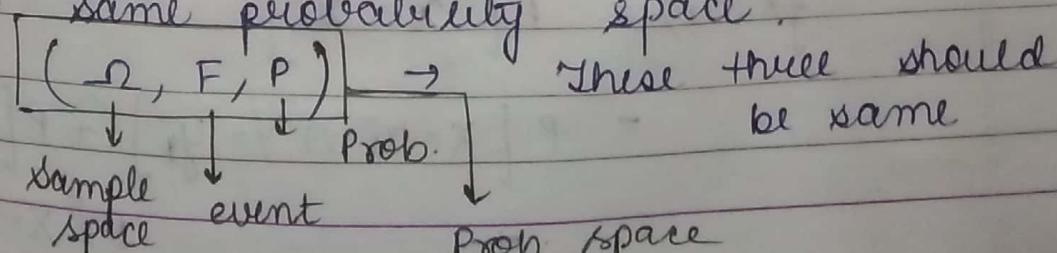
$[x_1 \leq a, x_2 \leq b]$ denotes the event of all
elements of S . s.t. $x_1(w) \leq a, x_2(w) \leq b$

The probability of event $[x_1 \leq a, x_2 \leq b]$
is $= P[x_1 \leq a, x_2 \leq b]$.

Two dimensional RV X and Y is called
discrete if Both X and Y are individually
Random Variable.

called continuous if X and Y are continuous R.V.

Two dimensional RV, X and Y are called
jointly distributed if they are defined
on same probability space.



Joint Probability Mass Function :- (Joint PMF)
when PMF = RV (discrete)

→ Let $Z = (X, Y)$ be a discrete RV then we define a mapping $f_{XY} : R^2 \rightarrow R$

$$P(X, Y) = f_{XY}(x, y) = P[X=x, Y=y] \quad \forall x, y \in R$$

This f_{XY} is called joint Prob. Mass function of $Z = (X, Y)$.

- For one D RV we have $f_X(x) = P[X=x]$

→ f_{XY} satisfies following Properties :-

(i) $f_{XY}(x, y) \geq 0, \quad \forall (x, y) \in R^2$

(ii) set $\{(x, y) \in R^2 : f_{XY}(x, y) \neq 0\}$

contains at most countably infinite point in R^2 .

(iii) $\sum_x \sum_y f_{XY}(x, y) = 1$

Any Real valued functions defined on R^2 having above properties is called joint PMF of RV (X and Y).

- Marginal Probability Mass Function :-

When we calculate f_X and f_Y from the joint PMF f_{XY} . (Then are Marginal PMF)

Let (X, Y) be a discrete RV
 $f_X(x_i) = P[X=x_i]$

$$= P[X=x_i \cap Y=y_1] + P[X=x_i \cap Y=y_2] \\ + \dots + P[X=x_i \cap Y=y_m]$$

we are taking intersection of x_i with all y , then at some pt. we will get x_i because all are mutually exclusive events they can be added up.

$$f_X(x_i) = \sum_{j=1}^m P[x=x_i \cap Y=y_j] \\ = \sum_{j=1}^m f_{XY}(x_i, y_j)$$

similarly

$$f_Y(y_i) = \sum_{i=1}^m f_{XY}(x_i, y_i)$$

$f_X(x_i)$, $f_Y(y_i)$ are called marginal pmf of $Z = (X, Y)$.

- (a) Suppose 3 balls are randomly selected from a bag containing 3R, 4W, 5 Blue Balls. If we let X, Y denotes respectively the no. of Red Balls and no. of white Balls. Find the joint PMF and Marginal PMF.

$$X \rightarrow 0, 1, 2, 3 \quad (\text{no. of Red Balls}) \\ Y \rightarrow 0, 1, 2, 3, \cancel{X} \quad (\text{no. of white Balls})$$

$$p(x, y) = f_{XY}(x, y) = p(0, 0) = \frac{5C_3}{12C_3}$$

$$p(0, 1) = \frac{4C_1 \cdot 5C_2}{12C_3}$$

$$p(0, 2) = \frac{4C_2 \cdot 5C_1}{12C_3}$$

$$p(0, 3) = \frac{4C_3}{12C_3}$$

$$p(1,0) = \frac{3c_1 \cdot 5c_2}{12c_3}$$

$$p(1,1) = \frac{3c_1 \cdot 4c_1 \cdot 5c_1}{12c_3}$$

$$p(1,2) = \frac{3c_1 \cdot 4c_2}{12c_3}$$

$$p(2,0) = \frac{3c_2 \cdot 5c_1}{12c_3}$$

$$p(2,1) = \frac{3c_2 \cdot 4c_1}{12c_3}$$

$$p(3,0) = \frac{3c_3}{12c_3}$$

Joint PMF

Marginal PMF :-

$x=i$	0	1	2	3	4	
0	$\frac{10}{220} = f_{x(0)}$	$\frac{40}{220}$	$\frac{30}{220}$	$\frac{4}{220}$	0	$\frac{84}{220}$
1	$\frac{30}{220}$	$\frac{60}{220} = f_{x(1)}$	$\frac{18}{220}$	0	0	$\frac{108}{220}$
2	$\frac{15}{220}$	$\frac{12}{220}$	0	0	0	$\frac{27}{220} = f_{x(2)}$
3	$\frac{1}{220}$	0	0	0	0	$\frac{1}{220} = f_{x(3)}$

On adding Rows we get Marginal PMF
of 1st RV

$$f_{x(0)} = \frac{56}{220} \quad \frac{112}{220} \quad \frac{48}{220} \quad \frac{4}{220} \quad 0 \rightarrow f_{x(4)}$$

$$f_{x(1)} \quad -f_{x(2)} \quad f_{x(3)}$$

on adding columns Marginal
PMF of
2nd RV.

Joint distribution function :-

Let (X, Y) be a random variable. Joint distribution function,

$$F_{XY}(x, y) = P \{ X \leq a, Y \leq b \} \quad -\infty < a, b < \infty$$

$$\begin{aligned} F_X(a) &= P \{ X \leq a \} = P \{ X \leq a, Y < \infty \} \\ &= P \left[\lim_{b \rightarrow \infty} \{ X \leq a, Y \leq b \} \right] \\ &= \lim_{b \rightarrow \infty} P \{ X \leq a, Y \leq b \} \\ F_X(a) &= \lim_{b \rightarrow \infty} F_{XY}(a, b) = F_{XY}(a, \infty) \end{aligned}$$

Similarly
 $F_Y(b) = \lim_{a \rightarrow -\infty} F_{XY}(a, b) = F_{XY}(-\infty, b)$

F_X, F_Y are Marginal distribution function
 of X and Y w.r.t (X, Y)

$$\rightarrow P \{ a_1 \leq X \leq a_2, b_1 \leq Y \leq b_2 \} = F_{XY}(a_2, b_2) + F(a_1, b_1) - F(a_1, b_2) - F(a_2, b_1)$$

$$\begin{aligned} F(-\infty, y) &= 0 = F(x, -\infty) \\ F(-\infty, +\infty) &= 1 \end{aligned}$$

$$\begin{aligned} P \{ X > a, Y > b \} &= 1 - P \{ X \leq a, Y \leq b \}^c \\ &= 1 - P \{ X > a \}^c \cup \{ Y > b \}^c \\ &= 1 - P \{ X \leq a \} \cup \{ Y \leq b \} \end{aligned}$$

ginal
of
N.

$$= 1 - P\{X \leq a\} - P\{X \leq b\} + P\{[X \leq a], [Y \leq b]\}$$

$$P\{X > a, Y > b\} = 1 - F_X(a) - F_Y(b) + F_{XY}[a, b]$$

#

$(X, Y) \rightarrow$ discrete random variable

→

$$F_X(x) = \sum_y P\{X \leq x, Y = y\}$$

$$F_Y(y) = \sum_x P\{X = x, Y \leq y\}$$

marginal distribution f"

→ summation taking over x

#

if $(X, Y) \rightarrow$ continuous random variable

$$F_X(x) = \int_{-\infty}^x \left(\int_{-\infty}^{\infty} f_{XY}(x, y) dy \right) dx$$

$$F_Y(y) = \int_{-\infty}^y \left(\int_{-\infty}^{\infty} f_{XY}(x, y) dx \right) dy$$

distribution f" in terms of joint distribution f°.

→

Joint Density function :-

let $f_{XY}(x, y)$ be a joint distribution function of continuous RV (x, y) .

$$\rightarrow f_{XY}(x, y) = \frac{\partial^2 F(x, y)}{\partial x \partial y}$$

→

Marginal Probability function of X and Y :-

$$f_X(x) = \begin{cases} \sum_y f_{XY}(x, y) & \text{discrete} \\ \int_{-\infty}^{\infty} f_{XY}(x, y) dy & \text{continuous} \end{cases}$$

discrete \rightarrow PMF
continuous \rightarrow PDF

joint PMF

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$$f_{Y|X}(y|x) = \begin{cases} \sum_x f_{XY}(x,y) & \text{discrete} \\ \int_{-\infty}^{\infty} f_{XY}(x,y) dx & \text{continuous} \end{cases}$$

joint probability density function

Result: Let (X, Y) be a random variable with pdf
if for any set $A \subseteq \mathbb{R}^2$

$$P\{(X, Y) \in A\} = \iint_A f(x, y) dx dy$$

integration over A

Question: joint pdf of X and Y is

$$f(x, y) = \begin{cases} 2 & \text{if } x > 0, y > 0, 0 < x+y < 1 \\ 0 & \text{otherwise} \end{cases}$$

Find $P\left\{X+Y < \frac{1}{2}\right\}$

$$A = \left\{ (x, y) \in \mathbb{R}^2 ; x + y < \frac{1}{2} \right\} = \iint_A f(x, y) dx dy$$

$$= \int_{-\infty}^{1/2} \int_{-\infty}^{1/2-x} f(x, y) dx dy$$

$x(0, 1/2)$

$y(1/2, 0)$

$\boxed{1/4}$

$\boxed{1/4}$

Question Joint pdf $f(x, y) = \begin{cases} 2e^{-x} e^{-2y} & 0 < x, y < \infty \\ 0 & \text{otherwise} \end{cases}$

(a) $P\{X > 1, Y < 1\}$

(b) $P\{X < Y\}$

(c) $P\{X < a\}$

Solution :-

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} 2e^{-x} e^{-2y} dy dx$$

$$[2y]_{-\infty}^{\infty}$$

 \rightarrow

(b)

$$P(X < y) = \int_{y=-\infty}^{\infty} \int_{x=-\infty}^{y} f(x,y) dx dy$$

not restriction on y

$$= \int_{y=0}^{\infty} \int_{x=0}^{y} 2e^{-x} e^{-2y} dx dy$$

$$\rightarrow -2 \int_{y=0}^{\infty} [e^{-x}]_0^{y-2y} dy$$

$$\rightarrow -2 \int_0^{\infty} (e^{-y} - 1) e^{-2y} dy$$

$$\rightarrow -\frac{2}{3} [e^{-3y}]_0^{\infty}$$

$$\rightarrow \frac{2}{3} [0 - 1] = -\frac{2}{3}$$

$$\rightarrow \frac{2}{3} \text{ Ans}$$

(c)

$$P\{X < a\} = \int_{y=-\infty}^{\infty} \int_{x=-\infty}^a f(x,y) dx dy$$

$$\rightarrow \int_0^{\infty} \int_0^a 2e^{-x} e^{-2y} dx dy$$

$$\rightarrow -2 \int_0^{\infty} [e^{-x}]_0^a e^{-2y} dy$$

$$\rightarrow -2 \int_0^{\infty} [e^{-a} - 1] e^{-2y} dy$$

$$\rightarrow -2 \int_0^{\infty} \left(e^{-a} - e^{-2y-a} \right) dy \rightarrow \boxed{1 - e^{-a}}$$

→ Independent Random Variable :-

Two RV X and Y are called Independent if for any $A \subseteq R$, $B \subseteq R$.

$$P[X \in A, Y \in B] = P[X \in A] P[Y \in B]$$

or

$$P[X \leq a, Y \leq b] = P[X \leq a] P[Y \leq b]$$

for all $a, b \in R$.

$$F_{XY}(a, b) = F_X(a) F_Y(b)$$

Now, $(X, Y) \rightarrow$ discrete RV, Independent if
 $P_{XY}(x, y) = p_X(x) p_Y(y)$
 joint probability mass function

$(X, Y) \rightarrow$ continuous RV,

$$f_{XY}(a, b) = f_X(a) f_Y(b)$$

for all points at which f_X, f_Y, f_{XY}
 are continuous.

$X=1$; heads

$X=0$; tails

$Y \rightarrow 1, 2, 3, 4, 5, 6$

$X, Y \rightarrow$ for any two $A, B \subseteq R$

$$E_A = [X \in A]$$

$$E_B = [X \in B]$$

event corresponding to this interval

Then X & Y are called Independent if E_A and E_B are Independent for any 2 subsets
 A and B.

ex

$x \backslash y$	-1	0	1
-1	0	$\frac{1}{4}$	0
0	$\frac{1}{4}$	0	$\frac{1}{4}$
1	0	$\frac{1}{4}$	0

Joint Probability Mass function

Are X and Y independent?

$$P[X=0] = \frac{1}{2}$$

Sol.

$$P[X = -1, Y = -1] = 0$$

$$P[X = -1] \cdot P[Y = -1] = \frac{1}{4} \cdot \frac{1}{4} = \frac{1}{16}$$

dependentex →Joint PDF of (X, Y)

$$f(x, y) = \begin{cases} 6(1-x), & 0 < y < x \quad 0 < x < 1 \\ 0, & \text{otherwise} \end{cases}$$

Are X and Y independent?Sol

$$\begin{aligned} f_X(x) &= \int_{y=-\infty}^{\infty} f(x, y) dy \\ &= \int_0^1 6(1-x) f(x, y) dy \\ &= \int_{y=0}^x 6(1-x) + \int_x^1 0 dy \end{aligned}$$

$$f_X(x) = \begin{cases} 6x(1-x); & 0 < x < 1 \\ 0; & \text{otherwise} \end{cases}$$

$$\begin{aligned} f_Y(y) &= \int_{x=-\infty}^{\infty} f(x, y) dx = \int_{x=0}^1 f(x, y) dx \\ &= \int_{\substack{y \\ x=0}}^y + \int_{\substack{y \\ x>y}}^1 6(1-x) dx \end{aligned}$$

$$f_{XY}(a, b) = f_X(a) \cdot f_Y(b)$$

for all points at which

$$f_Y(y) = \left[-\frac{6(1-y)^2}{2} \right]_y + 3(1-y)^2$$

$$f_Y(y) = \begin{cases} 3(y-1)^2 & 0 < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\left(\frac{1}{2}, \frac{1}{2}\right) \quad f_{XY} = 6 \cdot \frac{1}{2} = 3$$

$$f_X f_Y = 6 \cdot \frac{1}{2} \cdot \frac{1}{2} \cdot 3 \left(\frac{1}{2}\right)^2$$

Dependent

→ Necessary and sufficient condition for the RV X and Y to be independent :-

Two RV X and Y are independent iff joint pdf (pmf) can be expressed as →

$$f(x, y) = h(x) g(y) \quad -\infty < x < \infty$$

$$-\infty < y < \infty$$

Joint pdf (pmf) can be factored into 2 terms, one depending only on x and other depending only on y.

Let x and y be independent variables →

$$f_{XY}(x, y) = f_X(x) \cdot f_Y(y) \quad -\infty < x < \infty$$

$$-\infty < y < \infty$$

$$= h(x) g(y)$$

Conversely, let $f(x, y) = h(x) g(y)$

$$I = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y) dx dy$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(x) g(y) dx dy$$

$$I = \int_{-\infty}^{\infty} h(x) dx \int_{-\infty}^{\infty} g(y) dy$$

$$\boxed{I = C_1 C_2}$$

where $C_1 = \int_{-\infty}^{\infty} h(x) dx$

$$C_2 = \int_{-\infty}^{\infty} g(y) dy$$

$$f_X(x) = \int_{-\infty}^{\infty} f(x,y) dy = \int_{-\infty}^{\infty} h(x) g(y) dy$$

$$\boxed{f_X(x) = h(x) C_2} \quad - (2)$$

$$f_Y(y) = \int_{-\infty}^{\infty} f(x,y) dx$$

$$f_X(x) = \int_{-\infty}^{\infty} h(x) g(y) dx \quad - (3)$$

$$\boxed{f_Y(y) = C_1 g(y)}$$

$$\begin{aligned} f_X(x) f_Y(y) &= C_1 C_2 h(x) g(y) \\ &= h(x) g(y) \\ &= f_{XY}(x,y) \end{aligned}$$

Hence They are independent.

Joint pdf $f(x,y) = \begin{cases} 6e^{-2x}e^{-2y} & 0 < x < \infty \\ & 0 < y < \infty \\ 0 & \text{otherwise} \end{cases}$

Independent

2) $f(x,y) = \begin{cases} 24xy & 0 < x < 1, 0 < y < 1 \\ & 0 < x+y < 1 \\ 0 & \text{otherwise} \end{cases}$

Dependent

$$\rightarrow f_X(x) = \int_{y=-\infty}^{\infty} f(x,y) dy$$

• X_1, X_2, \dots, X_n be a RV

for any $A_1, A_2, \dots, A_n \subseteq R$

$$P[X_1 \in A_1, X_2 \in A_2, \dots, X_n \in A_n] = \prod_{i=1}^n P[X_i \in A_i]$$

$$P[X_1 \leq a_1, X_2 \leq a_2, \dots, X_n \leq a_n] = \prod_{i=1}^n P[X_i \leq a_i]$$

Infinite collection of RV's are independent if every finite sub collection is independent.

Sum of Independent RV's :-

Suppose X and Y are two independent continuous RV with joint pdf $f(x,y)$.

$$F_{X+Y}(a) = P\{X+Y \leq a\}$$

$$\begin{aligned} &= \iint_{\substack{x+y \leq a \\ y=-\infty, x=-\infty}} f(x,y) dx dy \\ &= \int_{-\infty}^a \int_{-\infty}^{a-y} f_X(x) f_Y(y) dx dy \end{aligned}$$

$$F_{x+y}(a) = \int_{y=-\infty}^{\infty} \left[\int_{x=-\infty}^{a-y} f_x(x) dx \right] f_y(y) dy$$

distribution function $F_{x+y}(a) = \int_{y=-\infty}^{\infty} F_x(a-y) f_y(y) dy$ This is

also called convolution distribution of X and Y .

$$* \boxed{f_{x+y}(a) = \frac{d}{da} [F_{x+y}(a)]} *$$

$$= \int_{y=-\infty}^{\infty} \left[\frac{d}{da} F_x(a-y) \right] f_y(y) dy$$

density function $f_{x+y}(a) = \int_{y=-\infty}^{\infty} f_x(a-y) f_y(y) dy$

ex

if X, Y are two independent continuous RV distributed uniformly $(0,1)$

$$f_x(a) = f_y(a) = \begin{cases} 1 & \text{if } 0 < a < 1 \\ 0 & ; \text{others} \end{cases}$$

$$F_x(a) = \begin{cases} 0 & \text{if } a < 0 \\ a & 0 \leq a < 1 \\ 1 & a > 1 \end{cases}$$

$$f_{x+y}(a) = \int_{y=-\infty}^{\infty} f_x(a-y) f_y(y) dy$$

limits only 0 to 1

$$\int_0^1$$

$$= \int_{y=0}^1 f_x(a-y) dy$$

Now if $0 < a < 1$

f_{x+y}

$$= \int_{y=0}^a 1 dy + \int_{y=a}^{\infty} 0 dy$$

$y \geq 0$
 $y < a$
 $a-y > 0$

$$f_{X+Y}(a) = a$$

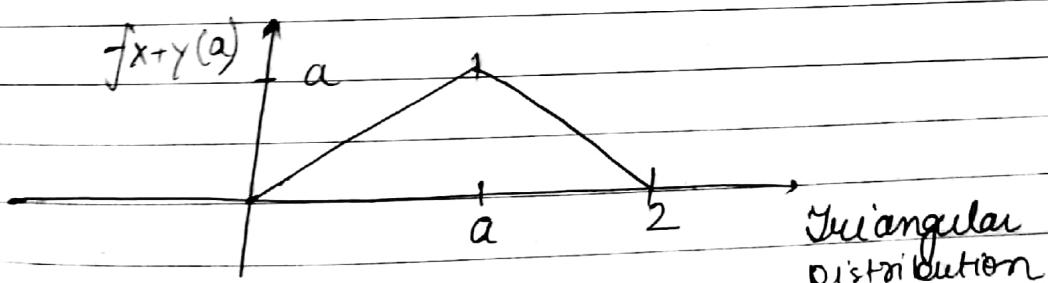
if $1 < a < 2$

$$f_{X+Y}(a) = \int_{y=0}^{a-1} 0 dy + \int_{y=a-1}^a 1 dy$$

$y < a-1$
 $a-y > 1$

$$\Rightarrow 1-a+1 = \frac{2-a}{2}$$

$$f_{X+Y}(a) = \begin{cases} a & 0 < a < 1 \\ 2-a & 1 \leq a < 2 \\ 0 & a \geq 2 \end{cases}$$



X_1, X_2, \dots, X_n , independent RV distributed uniformly on $(0,1)$

$$F_n(x) = P[X_1 + X_2 + \dots + X_n \leq x]$$

if $x \leq 1$

$$F_n\left(\frac{x}{n}\right) = \frac{x^n}{n!}$$

Proving it by Mathematical induction:
for $n=1$ $F_X(x) = \frac{x^1}{1!} = x$

This result is true for $n=1$; suppose true
for $n-1$

Suppose $f_{n-1}(x) = \frac{x^{n-1}}{(n-1)!}$

$$\sum_{i=1}^n X_i = \sum_{i=1}^{n-1} X_i + X_n \quad \begin{matrix} \text{Independent Random Variable} \\ \text{Random Variable} \end{matrix}$$

$$F_n(x) = \int_{-\infty}^{\infty} f_{n-1}(x-y) f_{X_n}(y) dy$$

$$\rightarrow F_n(x) = \int_0^x \frac{(x-y)^{n-1}}{(n-1)!} dy = \frac{1}{(n-1)!} \int_0^x (x-y)^{n-1} dy$$

Sum of discrete Independent Random Variable :-

(i) X, Y are two independent Poisson RV with parameters $\lambda_1 > 0$ and $\lambda_2 > 0$ respectively.

We are finding $\{X+Y=n\} = \text{union of disjoint event } \{X=k; Y=n-k\} \quad k=0, 1, 2, \dots$

$$f_{X+Y}(n) = P[X+Y=n] = \sum_{k=0}^n P[X=k, Y=n-k]$$

$$= \sum_{k=0}^n P[X=k] \cdot P[Y=n-k]$$

$$P[X+Y=n] = \sum_{k=0}^n \frac{e^{-\lambda_1} \lambda_1^k}{(k)!} \frac{e^{-\lambda_2} \lambda_2^{n-k}}{(n-k)!}$$

$$\Rightarrow \frac{e^{-\lambda_1} e^{-\lambda_2}}{(n!)^2} \sum_{k=0}^n \frac{\lambda_1^k \lambda_2^{n-k}}{(k)!(n-k)!} (n!)^2$$

$$P[X+Y=n] \Rightarrow \frac{e^{-(\lambda_1+\lambda_2)}}{n!} (\lambda_1 + \lambda_2)^n \quad \begin{matrix} \text{Poission} \\ \text{distribution} \\ \text{of } n \text{ with Par. } \lambda_1 + \lambda_2 \end{matrix}$$

→ distribution of $X+Y$ is also poision with parameter $\lambda_1 + \lambda_2$.

$$\begin{array}{l} \text{independent } X \rightarrow (\mu_1, \sigma_1^2) \\ \text{independent } Y \rightarrow (\mu_2, \sigma_2^2) \end{array}$$

$X+Y$ is also Normal Random variable with $(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$.

3(b) sum of Independent Random Variable

i) suppose X and Y are independent Binomial RV with parameters (n, p) and (m, p)

$$\begin{aligned} P\{X+Y=R\} &= \sum_{i=0}^n P\{X=i, Y=R-i\} \\ &= \sum_{i=0}^n P\{X=i\} P\{Y=R-i\} \\ &= \sum_{i=0}^n {}^n C_i (p)^i (q)^{n-i} {}^m C_{R-i} (p)^{R-i} (q)^{m-R+i} \end{aligned}$$

$$\sum_{i=0}^n (p)^i (q)^{m+n-i} \underbrace{{}^n C_i {}^m C_{R-i}},$$

$$\sum_{i=0}^n {}^{m+n} C_R (p)^i (q)^{m+n-i}$$

$$\sum_{i=0}^n {}^n C_i {}^m C_{R-i} = {}^n C_0 {}^m C_R + \frac{{}^n C_1 {}^m C_{R-1}}{n(n-1)} + \dots + \frac{{}^n C_n {}^m C_{R-n}}{n(n-1)\dots(2)(1)}$$

$$(x+y)^n = {}^n C_0 x^n + {}^n C_1 x^{n-1} y + \dots + {}^n C_n y^n$$

$$(x+y)^m = {}^m C_0 x^m + {}^m C_1 x^{m-1} y + \dots + {}^m C_m y^m$$

$$\quad \quad \quad {}^m C_{R-1} x^{m-R+1} y^{R-1} + {}^m C_R x^{m-R} y^R$$

$$\cdot {}^{nC_0} m_{CR} = \text{coeff of } x^n x^{m-k} y^R \\ \Rightarrow x^{m+m-k} y^R$$

$$\cdot {}^{nC_1} m_{CR-1} = \text{coeff of } yx^{m-1} y^{k-1} x^{m-k+1} \\ \Rightarrow \underline{x^{m+n-k} y^R}$$

from RHS we have:-

$$\underline{{}^{m+n} C_k x^{m+n-k} y^k} \Rightarrow \underline{\text{RHS}}$$

Theorem :- Let X and Y be two independent random variables & $g \circ h$ be two functions defined on \mathbb{R} . Then $g(X)$ and $h(Y)$ are also independent random variables.

X & Y independent \rightarrow

- 1) $|X|$ and $|Y|$ are also independent
- 2) X and $\sin Y$ are also independent
- 3) X^2 and Y^2 are also independent

Ques

Let X and Y are two poission identically distributed random RV. Find PMF of minimum of X and Y (Proof)

Sol

PMF of $\min [X, Y]$.

Suppose $Z = \min [X, Y]$

$X, Y = 0, 1, 2$

$Z = 0, 1, 2$

For PMF we are interested in calculating probability at any value R .

$$\{Z=R\} = \{X=R, Y \geq R\} \cup \{X \geq R, Y=R\}$$

$$P\{X=k\} = P\{(X=k, Y \geq k) \cup (X \geq k, Y=k)\}$$

$$= P\{(X=k, Y \geq k)\} + P\{(X \geq k, Y=k)\} -$$

$$P\left(\{X \geq k, Y \geq k\} \cap \{X \geq k, Y=k\}\right)$$

$$\Rightarrow P\{(X=k, Y \geq k)\} + P\{(X \geq k, Y=k)\} - P\{X=k, Y \geq k\}$$

$$\Rightarrow P\{(X=k, Y \geq k)\} + \frac{P\{X \geq k, Y=k\}}{P\{X=k\} P\{Y=k\}}$$

Because X and Y are independent
if X and Y are identically distributed
then $P\{(X \geq k, Y \geq k)\}$ and $P\{X \geq k, Y=k\}$
will be same.

When identically distributed we have

$$\boxed{2P\{X=k, Y \geq k\} - P\{X=k\}^2}$$

X, Y are independent RV uniformly distributed over $(0, 1)$. find the pdf of $\frac{Y}{X}$.

* At we find the distribution f^n of this.
 $Z = Y/X$.

if $z < 0$

$$f_Z(z) = 0 \quad (\text{defined only on } (0, 1))$$

if $z > 0$

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

$$\Rightarrow P\left[\frac{Y}{X} \leq z\right]$$

$$\Rightarrow \iint_A f(x, y) dx dy$$

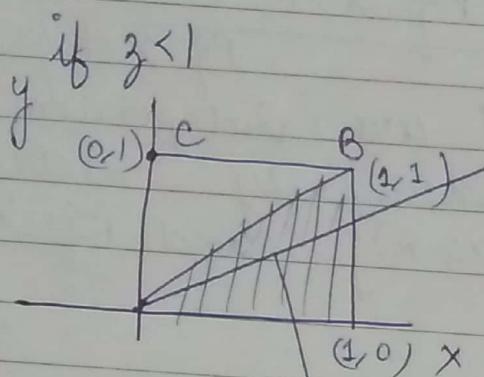
Over the Region A, we Integrate

$$A = \left[(x, y) \in (0, 1) \times \mathbb{R}^2 \mid \frac{y}{x} \leq z \right]$$

$$f(x, y) = \begin{cases} 1, & 0 < x < 1 \quad 0 < y < 1 \\ 0, & \text{otherwise} \end{cases}$$

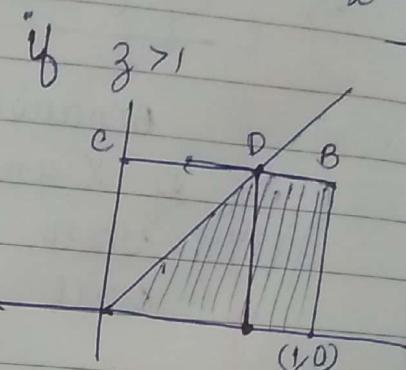
$f(x, y) = \iint_S dx dy =$

$$S = \left[(x, y) \in (0, 1) \times (0, 1) \mid \frac{y}{x} \leq z \right]$$

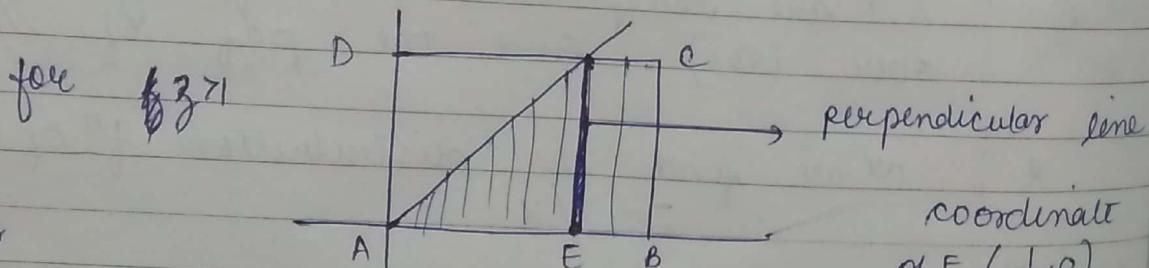


$$\begin{aligned} y &\leq xz \\ \text{for } z=1 \\ y &= x \end{aligned}$$

for $z \geq 1$
this line leaves
the region from
corner



for $z \geq 1$
This side shift



coordinate
of E $(\frac{1}{2}, 0)$

$$F_Z(z) = \frac{1}{2} \times z \times 1 = \frac{z}{2} \quad \text{for } 0 < z \leq 1$$

$$F_Z(z) = \frac{1}{2} \times \frac{1}{2} \times 1 + \left(1 - \frac{1}{2}\right) \times 1$$

$$\Rightarrow \frac{1}{2} - \frac{1}{2} + 1 = 1 - \frac{1}{2} \quad z \geq 1$$

$$F_Z(z) = \begin{cases} \frac{z}{2} & ; \quad 0 \leq z \leq 1 \\ 1 - \frac{1}{2z} & ; \quad z > 1 \\ 0 & ; \quad \text{Otherwise} \end{cases}$$

$$f_Z(z) = \begin{cases} \frac{1}{2} & ; \quad 0 < z < 1 \\ \cdot \frac{1}{2z^2} & ; \quad z > 1 \\ 0 & ; \quad \text{otherwise} \end{cases}$$

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conditional distribution :-

E, F are events

$$P[E|F] = \frac{P[E \cap F]}{P[F]}, \text{ provided } P(F) > 0$$

Discrete: Suppose X and Y are two discrete random variable with pmf $p(x,y)$ = conditional prob. Mass function of X given that $Y=y$.

$$p_{X/Y}(x/y) = \frac{P[X=x | Y=y]}{P[Y=y]}$$

$$p_{X/Y}(x/y) = \frac{p(x,y)}{p(y)}$$

cumulative conditional distribution of X given that $Y=y$.

$$F_{X/Y}(x/y) = \sum_{a \leq x} p_{X/Y}\left(\frac{a}{y}\right) = P[X \leq x | Y \leq y]$$

$$X = a_1, a_2, a_3, \dots, a_n$$

$$F_{X/Y}(a_i) = P[X \leq a_i | Y=y]$$

$$F_{X/Y}(a_i) = P[X = a_1 | Y=y] + P[X = a_2 | Y=y] - \dots - P[X = a_{i-1} | Y=y] + P[X = a_i | Y=y]$$

When X, Y are independent

$$P_{X/Y}(x/y) = \frac{P[X=x]}{P[Y=y]} = P[X=x]$$

Ques pmf is given

$$p(0,0) = 0.4$$

$$p(0,1) = 0.2$$

$$p(1,0) = 0.1$$

$$p(1,1) = 0.3$$

$$P_{X/Y}\left(\frac{x}{y=1}\right) = \frac{p(x,1)}{p(1)}$$

for all values of x

$$P_{X/Y}(0/1) = \frac{p(0,1)}{p(1)} = 0.2 / 0.5$$

$$P_{X/Y}(1/1) \Rightarrow \frac{p(1,1)}{p(1)} \Rightarrow \frac{0.3}{0.5}$$

Ans
column sum

Ques If X and Y are independent poission RV with parameters λ_1 and λ_2 . Find the conditional probability of X given that $X+Y=n$.

$$\text{Sol :- } P[X=k | Y=n] = \frac{P[X=k, X+Y=n]}{P[X+Y=n]}$$

$$\bullet \frac{P[X=R, Y=n-R]}{P[X+Y=n]}$$

* $\frac{P[X=R]}{P[X+Y=n]} \cdot \frac{P[Y=n-R]}{P[Y=n]}$ [X, Y are independent]

$$\rightarrow \frac{e^{-\lambda_1} \lambda_1^R}{(R)!} \cdot \frac{e^{-\lambda_2} (\lambda_2)^{n-R}}{(n-R)!}$$

$$\frac{e^{-(\lambda_1 + \lambda_2)}}{(n)!} \frac{(\lambda_1 + \lambda_2)^n}{(\lambda_1 + \lambda_2)^n}$$

$$\rightarrow \frac{(n)!}{(R)! (n-R)!} \left[\frac{\lambda_1^R \cdot \lambda_2^{n-R}}{(\lambda_1 + \lambda_2)^n} \right]$$

$$\rightarrow {}^n C_R \left(\frac{\lambda_1}{\lambda_1 + \lambda_2} \right)^R \left(\frac{\lambda_2}{\lambda_1 + \lambda_2} \right)^{n-R}$$

All X, Y independent, identically distributed Geometric random variable. Find the conditional PMF

$$P\left[\frac{Y=y}{X+Y=n}\right]$$

$$\rightarrow P\left[\frac{Y=y}{P[X+Y=n]}\right] = P\left[\frac{Y=y, X=n-y}{P[X+Y=n]}\right]$$

$$\rightarrow \frac{P[Y=y] P[X=n-y]}{P[X+Y=n]} \xrightarrow{n-1} \sum_{k=1}^{n-1} P[X=R, Y=n-k]$$

$$\rightarrow \frac{(1-p)^{y-1} p}{(1-p) \sum_{k=1}^{n-1} P[X=k] \cdot P[Y=n-k]}$$

$$\rightarrow \frac{(1-p)^{n-2} p^2}{\sum_{k=1}^{n-1} p (1-p)^{k-1} p (1-p)^{n-k-1}}$$

$$\rightarrow \frac{\sum_{n=1}^{\infty} p^2 (1-p)^{n-2}}{p^2 (1-p)^{n-2}} = \frac{1}{n-1}$$

→ continuous Random Variable :-

Suppose $f(x,y)$ is a joint pdf of two RV X and Y
conditional prob of X given that $X=Y$.

$$f_{X/Y}\left(\frac{x}{y}\right) = \frac{f(x,y)}{f(y)}$$

$$F_{X/Y}\left(\frac{a}{y}\right) = \int_{-\infty}^a f_{X/Y}\left(\frac{x}{y}\right) dx$$

$$P[X \leq a | Y=y]$$

Ques Joint density of X and Y :-

$$f(x,y) = \begin{cases} \frac{12}{5} x(2-x-y) & ; 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0 & ; \text{otherwise} \end{cases}$$

$$f_{X/Y}\left(\frac{x}{y}\right) = \frac{f(x,y)}{f_Y(y)} \quad 0 < y < 1$$

$$\rightarrow f_Y(y) = \int_{-\infty}^{\infty} f(x,y) dx$$

$$= \int_0^1 \frac{12}{5} (2x - x^2 - xy) dx$$

$$= \frac{12}{5} \left[2xy - \frac{x^2}{2} - \frac{xy^2}{2} \right]_0^1$$

$$\Rightarrow \frac{12}{5} \left[2 - 1 - \frac{1}{2} \right] = \frac{12}{5} \times \frac{1}{2} = \frac{6}{5}$$

$$\rightarrow \frac{12}{5} \left[x^2 - \frac{x^3}{3} - \frac{x^2}{2} y \right]_0^1$$

$$\rightarrow \frac{12}{5} \left[1 - \frac{1}{3} - \frac{1}{2} \right] = \frac{12}{5} \left[1 - \frac{1}{6} \right]$$

$$\frac{f(x,y)}{f_Y(y)} = \frac{6x(2-x-y)}{4-3y} \stackrel{?}{=} 2$$

$f(x,y) = \begin{cases} \frac{e^{-x/y} e^{-y}}{y} & 0 < x < \infty \\ 0 & \text{otherwise} \end{cases}$

$$P[x > 1 | Y=y]$$

or $1 - P[x \leq 1 | Y=y] = 1 - \int_0^1 f_X\left(\frac{x}{y}\right) dx$

conditional density function of X

$$1 - \int_0^1 \left(\frac{1}{y} e^{-x/y} \right) dx$$

$$1 - \left[\frac{1}{y} e^{-x/y} x \right]_0^1$$

$$1 + \left[e^{-1/y} - 1 \right] \Rightarrow \underline{e^{-1/y}}$$

Q Why we are taking y as constant?

Because Random variable is taking values on a point

20/3/18

Covariance :- The covariance of two random variables X and Y is denoted by $\text{Cov}(X,Y)$ and defined by

$$\text{Cov}(X,Y) = E \left[\underbrace{(X - E(X))(Y - E(Y))}_{\text{deviation of } X \text{ about its Mean}} \right]$$

deviation of X about its Mean

covariance measures the behaviour of two

Random variable ; how they behave together

If $\text{cov}(X, Y)$ is positive \rightarrow Both X and Y are simultaneously increasing or decreasing

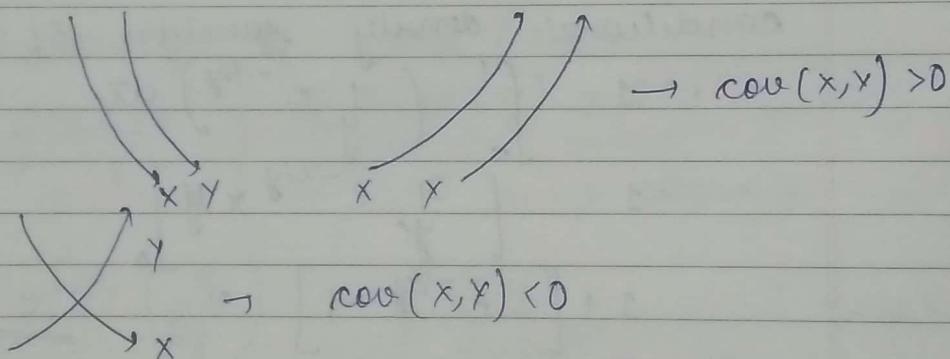
\rightarrow If $\text{cov}(X, Y) < 0 \Rightarrow$ one RV increase and another one decrease

$\rightarrow \text{cov}(X, Y) = 0 \Rightarrow X$ and Y are uncorrelated

$$\text{cov}(X, Y) = E \left[XY - \underbrace{Y E(X)}_{\text{constant}} - \underbrace{X E(Y)}_{\text{constant}} + E(X) E(Y) \right]$$

$$\rightarrow E(XY) - E(Y) E(X) - E(X) E(Y) + E(X) E(Y)$$

$$\rightarrow \boxed{\text{cov}(X, Y) = E(XY) - E(X) E(Y)}$$



ex

\rightarrow If there is no correlation b/w ~~2 independent~~ random variables then it is not necessary that they will be ~~independent~~

Result

\rightarrow If two RV X and Y are Independent then that implies that X and Y are uncorrelated i.e. $\text{cov}(X, Y) = 0$

\rightarrow To Prove when $X \& Y \rightarrow$ Independent $\text{cov}(X, Y) = 0$

$$\begin{aligned} \text{cov}(X, Y) &= -E(X)E(Y) + E(XY) \\ &\quad - E(X)E(Y) + E(X)E(Y) \quad] ? \\ &\quad \approx 0 \end{aligned}$$

To prove the last result :

To prove this let $f_{XY}(x,y)$ is joint pdf of $X \& Y$. Then,

$$E(XY) = \iint xy f_{XY}(x,y) dx dy$$

$$\cdot \{ E[g(X,Y)] = \iint g(x,y) f_{XY}(x,y) dx dy \}$$

$$E(XY) = \iint xy f_X(x) f_Y(y) dx dy$$

$$E(XY) = \left[\underbrace{\int x f_X(x) dx}_{E(X)} \right] \left[\underbrace{\int y f_Y(y) dy}_{E(Y)} \right]$$

(because of independency of $X \& Y$)

Range of X does not depend on Range of Y .

$$E(XY) \rightarrow E(X) E(Y)$$

$$\Rightarrow \text{cov}(X,Y) = 0$$

$\rightarrow X \& Y$ are uncorrelated.

• converse of this is not true.

ex	det	X (RV)	-1	0	1
	$P(X=x)$		$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$

$$\text{define } Y = \begin{cases} 0 & \text{if } X=0 \\ 1 & \text{if } X \neq 0 \end{cases}$$

Uncorrelatedness does not define the independence of two Random Variables.

$$XY = 0 \Rightarrow E(XY) = 0$$

constant

$$E(X) = -\frac{1}{3} + 0 + \frac{1}{3} = 0 \quad \sum x_i f(x_i)$$

$$\text{cov}(X,Y) = 0 \rightarrow X \text{ and } Y \text{ are uncorrelated variables}$$

• definition of Y
depends on definition of X

Ex → consider RV X & Y where the joint pmf is given by

$X \setminus Y$	-1	0	1	
-1	0	$\frac{1}{4}$	0	→
0	$\frac{1}{4}$	0	$\frac{1}{4}$	→ X
1	0	$\frac{1}{4}$	0	→

$P(X=-1, Y=-1) = 0 \rightarrow$ RV X and Y are uncorrelated but dependent.

$$P(X=-1) = \frac{1}{4}$$

$$P(X=0) = \frac{1}{2}$$

$$P(Y=-1) = \frac{1}{4}$$

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

$$P(Y=0) = \frac{1}{2}$$

$$P(Y=1) = \frac{1}{4}$$

Both X and Y are having same distribution because their Marginal probabilities are same

$$\begin{aligned} E(X) &= \sum x_i p(x_i) \\ &= -1 \times \frac{1}{4} + 0 \times \frac{1}{2} + 1 \times \frac{1}{4} = 0 \end{aligned}$$

$$E(Y) = 0$$

$$E(XY) =$$

$$\begin{array}{ccccccc} XY : & -1 & & 0 & & +1 & \\ \hline P(XY=xy) & \underline{0+0} & & \frac{1}{4} + 0 + \frac{1}{4} & + \frac{1}{4} + \frac{1}{4} & 0+0 & \\ & & & - & \frac{1}{2} + \frac{1}{2} = 1 & & \end{array}$$

$\text{cov}(X, Y) \Rightarrow 0 \Rightarrow X \& Y \text{ are uncorrelated}$

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$$P(X=1, Y=-1) = 0 \neq P(X=-1) P(Y=-1)$$

→ X and Y are dependent

$$\rightarrow E(X, Y) = \iint xy f_{XY}(x, y) dx dy$$

ans consider two RV X and Y jointly uniformly distributed over triangle with vertices $(-1, 0)$, $(1, 0)$, $(0, 1)$ with joint pdf

$$f_{XY}(X, Y) = \begin{cases} 1 & \text{if } (x, y) \in T \\ 0 & \text{otherwise} \end{cases}$$

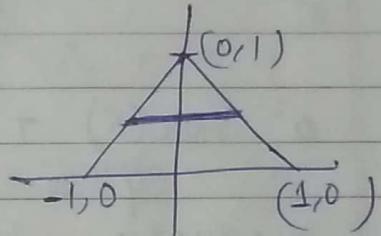
show RV X and Y are uncorrelated but dependent.

line eqn $\rightarrow y = mx + c$

$$m = \frac{1}{1}$$

$$1 = c$$

$$\boxed{y = x + 1}$$



$$E(X) = \iint_{y=1}^{x-y} x dx dy = 0$$

$$E(XY) = E(XY) = \iint_{y=1}^{x-y} xy dx dy = 0$$

$$\text{cov}(X, Y) = 0$$

→ X & Y are independent

2/3/18

Correlation coefficient for two X and Y correlation coefficient is denoted by $P(X, Y)$ and defined as →

$$P(X, Y) = \frac{\text{cov}(X, Y)}{\sqrt{\text{Var}(X) \text{Var}(Y)}} = \frac{\text{cov}(X, Y)}{\sigma_X \sigma_Y}$$

Property of correlation function :-

$$1 * \rho(x, y) \leq 1$$

consider

$$E\left\{ a(x - E(x)) + (y - E(y))^2 \right\} \geq 0$$

$$\rightarrow E \left[a^2 (x - E(x))^2 + (y - E(y))^2 + 2a (x - E(x))(y - E(y)) \right] \geq 0$$

LHS

$$\rightarrow a^2 E[(x - E(x))^2] + E[(y - E(y))^2] + 2a E[(x - E(x))(y - E(y))] \geq 0$$

$$\rightarrow a^2 \text{Var}(x) + \text{Var}(y) + 2a \text{cov}(x, y) \geq 0$$

Eq^n is quadratic in a
quadratic is always greater than zero.
so $D < 0$

$$b^2 - 4ac < 0$$

$$(2 \text{cov}(x, y))^2 - 4 \text{Var}(x) \text{Var}(y) < 0$$

$$(\text{cov}(x, y))^2 - \text{Var}(x) \text{Var}(y) < 0$$

$$\frac{(\text{cov}(x, y))^2}{\text{Var}(x) \text{Var}(y)} < 1$$

Taking square Root

$$\left| \frac{\text{cov}(x, y)}{\sqrt{\text{Var}(x) \text{Var}(y)}} \right| < 1$$

Bounded by 1

↓
This value is in between -1 to 1.

Properties \rightarrow

$$\text{cov}(X, Y) = \text{cov}(Y, X)$$

$$\text{cov}(X, X) = \text{Var}(X)$$

$$\text{cov}(aX, Y) = a \text{cov}(X, Y)$$

$$\text{cov}\left(\sum_{i=1}^n X_i, \sum_{j=1}^m Y_j\right) = \sum_{i=1}^n \sum_{j=1}^m \text{cov}(X_i, Y_j)$$

Prove this

LHS

$$\text{cov}\left(\sum_{i=1}^n X_i, \sum_{j=1}^m Y_j\right)$$

$$\rightarrow E\left[\left(\sum_{i=1}^n X_i - \sum_{i=1}^n E(X_i)\right) \left(\sum_{j=1}^m Y_j - \sum_{j=1}^m E(Y_j)\right)\right]$$

$$\rightarrow E\left[\sum_{i=1}^n (X_i - E(X_i)) \left(\sum_{j=1}^m (Y_j - E(Y_j))\right)\right]$$

$$\rightarrow \sum_{i=1}^n \sum_{j=1}^m E[(X_i - E(X_i))(Y_j - E(Y_j))]$$

$$\rightarrow \sum_{i=1}^n \sum_{j=1}^m \text{cov}(X_i, Y_j) \quad \underline{\text{RHS}}$$

Let (S, F, P) be a prob model and $A \in F$,

$B \in F$

$$\det X = \begin{cases} 1 & \text{if } A \text{ occurs} \\ 0 & \text{if } \bar{A} \text{ occurs} \end{cases}$$

$$Y = \begin{cases} 1 & \text{if } B \text{ occurs} \\ 0 & \text{if } \bar{B} \text{ occurs} \end{cases}$$

$\text{cov}(X, X) > 0$ Then what about $A \& B$?

$\text{cov}(X, Y), E(XY), E(X), E(Y)$

$$\boxed{E(X) = P(A)}$$

$$\boxed{E(Y) = P(B)}$$

$$\cdot \quad \boxed{E(XY) = P(AB)}$$

$$XY = \begin{cases} 1 & \text{if A and B both occur} \\ 0 & \text{if A or B doesn't occur} \end{cases}$$

$$E(XY) = P(AB)$$

$$\begin{aligned} \rightarrow \text{cov}(X, Y) &> 0 \\ \rightarrow P(AB) &> P(A)P(B) \\ \rightarrow \frac{P(AB)}{P(B)} &> P(A) \\ \rightarrow P(A/B) &> P(A) \end{aligned}$$

\rightarrow If $\text{cov}(X, Y)$ is positive then conditional prob of A and B is greater than prob of A.

$$\begin{aligned} \text{If } \text{cov}(X, Y) < 0 \\ P(A/B) &< P(A) \end{aligned}$$

$$\text{If } \text{cov}(X, Y) = 0$$

$$\rightarrow \boxed{\text{Unconditional Prob} = \text{Conditional Prob}}$$

23/3/18

conditional Distribution :-

(continuous Random V) $P(A/B) = \frac{P(AB)}{P(B)}$ provided $P(B) > 0$

In the similar manner, conditional distribution $F_{x/M}(x/M)$ of a random variable X, assuming event M has occurred, is defined as

$$\rightarrow \boxed{F_{x/M}(x/M) = \frac{P(X \leq x, M)}{P(M)}}$$

$F_x(x) = P(X \leq x)$

↓
when X is continuous

→ conditional distribution : Discrete Random V :-
 If $M = \{Y=y\}$

If X and Y are two Random Var. and
 $M = \{Y=y\}$ then

$$F_{X/Y}(x/y) = \frac{P(X \leq x, Y=y)}{P(Y=y)}$$

If X and Y are discrete Random V. associated to some experiment and $M = \{Y=y\}$ then conditional PMF of X given than $\{Y=y\}$ is

$$p_{X/Y}(x/y) = \frac{P(X=x, Y=y)}{P(Y=y)}$$

$$p_{X/Y}(x/y) = \frac{p(x,y)}{p(y)} \rightarrow \begin{array}{l} \text{Joint PMF} \\ \text{Marginal PMF} \end{array}$$

3/3/18

$$P(E/M) = \frac{P(EM)}{P(M)} \quad P(M) > 0$$

If X is a random variable then conditional probability distribution of X given that M has occurred can be thought as - the conditional Probability as →

$$F_{X/M}(x/M) = \frac{P(X \leq x, M)}{P(M)} \quad \text{provided } P(M) > 0$$

conditional distribution ; discrete case !

X and Y two discrete RV with joint PMF

$p_{XY}(x,y)$ Then conditional pmf is given by

$$p_{X/Y}(x/y) = P(X=x / Y=y)$$

$$p_{x/y}(x/y) = \frac{P(X=x, Y=y)}{P(Y=y)} = \frac{p_{xy}(x,y)}{p_y(y)}$$

$$\left[p_{x/y}(x/y) = \frac{p_{xy}(x,y)}{p_y(y)} \right] \quad p_y(y) > 0$$

$$\sum p_x(x) = 1$$

$$\sum_x p_{x/y}(x/y) = \sum_x \frac{p_{xy}(x,y)}{p_y(y)}$$

$$= \frac{1}{p_y(y)} \sum_x p_{xy}(x,y)$$

$$\left[\sum_x p_{xy}(x/y) = \frac{1}{p_y(y)} \cdot p_y(y) = 1 \right]$$

* If X and Y are independent :-

$$p_{x/y}(x/y) = p_x(x)$$

Ques Let the joint pmf of X and Y is given by

X \ Y	-1	0	1
-1	0	$\frac{1}{4}$	0
0	$\frac{1}{4}$	0	$\frac{1}{4}$
1	0	$\frac{1}{4}$	0

or ampl

compute pmf of X given that $Y=0$, also
compute conditional distribution f_x^y of X

$F_{x/y}(x/y)$ given $y=0$?

$p_{x/y}(x/y) \ y=0$?

Ans

$$P(Y=0) = 1/2$$

$$p_{X/Y}(x_0) = \frac{p(x_0)}{p_{Y=0}} = \frac{P(X=x_0, Y=0)}{P(Y=0)}$$

$$p_{X/Y}(x_0) = \begin{cases} \frac{1}{4}/\frac{1}{2} = 1/2 & x=-1 \\ 0 & x=0 \\ \frac{1}{4}/\frac{1}{2} = \frac{1}{2} & x=1 \end{cases}$$

y=0
always

$$\rightarrow f_{X/Y}(x/y) = P(X \leq x / Y=y)$$

$$[F_{X/Y}(x/y) = \sum_{a \leq x} p_{X/Y}(a/y)]$$

→ In last example :

$$F_{X/Y}(x_0) = \begin{cases} 0 & x < -1 \\ \frac{1}{2} & -1 \leq x < 0 \\ \frac{1}{2} & 0 \leq x < 1 \\ 1 & x > 1 \end{cases}$$

Poisson

Example Let X and Y be two independent Random V. with parameters λ_1 & λ_2 respectively. Then find out the conditional distribution of X given that $X+Y=n$.

Sol

$$P(X=k / X+Y=n) = \frac{P(X=k, X+Y=n)}{P(X+Y=n)}$$

$$\frac{P(X=k, Y=n-k)}{P(X+Y=n)} = \frac{P(X=k) P(Y=n-k)}{P(X+Y=n)}$$

Since X and Y are independent

$$\frac{e^{-\lambda_1} \lambda_1^k}{k!} \frac{e^{-\lambda_2} \lambda_2^{n-k}}{(n-k)!} \left[\frac{n!}{e^{-(\lambda_1+\lambda_2)} (\lambda_1+\lambda_2)^n} (\lambda_1+\lambda_2)^n \right]$$

$$\rightarrow \frac{n!}{k! (n-k)!} \frac{\lambda_1^k}{(\lambda_1+\lambda_2)^n} = {}^n C_R \left(\frac{\lambda_1}{\lambda_1+\lambda_2} \right)^k \left(\frac{\lambda_1}{\lambda_1+\lambda_2} \right)^{n-k}$$

$$\begin{aligned} \rightarrow p_X(x) &= \sum_y p_{XY}(x/y) \\ &= \sum_y p_{XY}(x/y) f_Y(y) \end{aligned}$$

Conditional distribution : Continuous case :

Let X and Y be two jointly continuous RV's with joint pmf $f_{XY}(x/y)$ then conditional pdf of X given that $\{X=y\}$ is defined

as

$$f_{X/Y}(x/y) = \frac{f_{XY}(x,y)}{f_X(y)} \quad \text{provided } f_X(y) > 0$$

example

$$\int_{-\infty}^{\infty} f_{X/Y}(x/y) dx = 1$$

$$\int_{-\infty}^{\infty} \frac{f_{XY}(x,y)}{f_X(y)} dx = \frac{1}{f_X(y)} \int_{-\infty}^{\infty} f_{XY}(x,y) dx$$

$$= \frac{f_X(y)}{f_X(y)} = 1 \quad \text{Hence proved}$$

→ If X and Y are jointly continuous then for any $x \in A$ (A is any set on the Real line)

$$P(X \in A | Y=y) = \int_{x \in A} f_{X/Y}(x/y) dx$$

→ we are just defining this f^n in a new real dimension by limiting the value of Y .

$$A = (-\infty, a)$$

$$f_{X/Y}(a/y) = P(X \leq a / Y=y) \\ = \int_{-\infty}^a f_{X/Y}(x/y) dx$$

It helps us to find out prob of those events when one random variable is already assigned with some value.

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Ans

Let X & Y be two RV's having the joint pdf

$$f_{XY}(x,y) = \begin{cases} 2 & \text{when } 0 < x < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$P(X \leq \frac{2}{3} | Y = \frac{3}{4}) = ?$$

$$P(X \leq a | Y=y) = \int_{x \leq a} f_{X/Y}(x/y) dx$$

conditional density f^n .

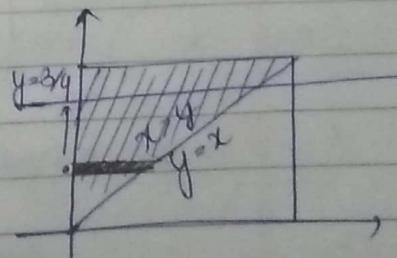
We need to find out $f_{X/Y}(x/y)$

$$f_{X/Y}(x/y) = \frac{\int_{x \leq a} f_{XY}(x,y) dx}{f_Y(y)}$$

$$f_Y(\frac{3}{4}) = \int_{-\infty}^{\frac{3}{4}} f_{XY}(x, \frac{3}{4}) dx \\ = \int_0^{\frac{3}{4}} 2 dx$$

$\rightarrow x$ ranges from 0 to $\frac{3}{4}$

$$\boxed{\text{Ans} = \frac{3}{2}}$$



conditional density f^n

from given
function ↑

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$$f_{X/Y}(x/y) = \begin{cases} \frac{2}{3}/\frac{1}{2} & 0 < x < \frac{2}{3}, \frac{3}{4} \\ 0 & \text{otherwise} \end{cases}$$
$$= \begin{cases} \frac{4}{3} & 0 < x < \frac{3}{4} \\ 0 & \text{otherwise} \end{cases}$$

$$\rightarrow P(X \leq \frac{2}{3} / Y = \frac{3}{4}) = \frac{2}{3} \int_0^{\frac{2}{3}} \frac{4}{3} dx$$

$$\Rightarrow \frac{4}{3} \times \frac{2}{3} \Rightarrow \frac{8}{9}$$

$$\rightarrow \text{when } P(X \leq 1 / Y = \frac{3}{4}) = \frac{3}{4} \int_0^1 f_{X/Y}(x/y) dx +$$

We need to add because upto

$$Y = \frac{3}{4} f_{X/Y}(x/y)$$

is different after
that it is
different

$$\rightarrow \frac{3}{4} \int_0^{\frac{3}{4}} \frac{4}{3} dx + \int_{\frac{3}{4}}^1 0 dx$$
$$\rightarrow \frac{4}{3} \times \frac{3}{4} = 1$$

ex

su
dist
fin
x

Ex

Let X & Y be independent continuous RV with pdf f_X and f_Y respectively. Let $Z = X+Y$. Determine conditional density of Z given $X = x$.

$$F_{Z/X}(z/x) = P(Z \leq z | X=x) = ?$$

$$P(Z \leq z | X=x) = \int_{-\infty}^z f_{Z/X}(t/x) dt \quad \text{--- ①}$$

$$P(Z \leq z | X=x) = P(X+Y \leq z | X=x)$$

$$= P(Y \leq z-x | X=x)$$

As X and Y are independent RV.

$$= P(Y \leq z-x)$$

→

Y

$$\begin{aligned} &= \int_{-\infty}^{z-x} f_Y(y) dy \\ &= \int_{-\infty}^z f_Y(t+x) dt \quad \text{say } y = t + x \end{aligned}$$

comparing ① and ②

$$\begin{aligned} F_{Z/X}(z/x) &= \int_{-\infty}^z f_{Z/X}(t/x) dt \\ &= \int_{-\infty}^z f_Y(t-x) dt \end{aligned}$$

$$\boxed{f_{Z/X}(z/x) = f_Y(z-x)}$$

ex Suppose X and Y are independent & identically distributed, geometric RV's with parameter p find the conditional pmf of X given $X+Y=n$. where $n \geq 2$.

- X and Y can take values from natural numbers $R_X = 1, 2, 3, \dots$ Because they are geometric random variables.

$$X+Y = 2, 3, 4, \dots$$

$$P(Y=y \mid X+Y=n) = ?$$

→ Y takes $n, n+1, n+2, \dots$

$$P(Y=y \mid X+Y=n) = 0$$

for $y = 1, 2, 3, \dots, n-1$

$$P(Y=y \mid X+Y=n) = \frac{P(Y=y, X+Y=n)}{P(X+Y=n)}$$

$$\rightarrow \frac{P(Y=y, X=n-y)}{P(X+Y=n)} = \frac{P(Y=y)}{P(X+Y=n)} \frac{P(X=n-y)}{P(X+Y=n)} \quad \text{--- ①}$$

$$P(X+Y=n) = \sum_{k=1}^{n-1} P(X=k, Y=n-k)$$

$$= \sum_{k=1}^{n-1} P(X=k) P(Y=n-k)$$

$$= \sum_{k=1}^{n-1} p \cdot (1-p)^{k-1} p(1-p)^{n-k-1}$$

$$\Rightarrow \sum_{k=1}^{n-1} (1-p)^{n-2} (p)^2$$

Thus from eq ①

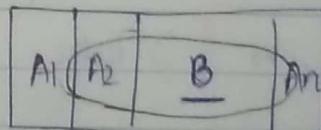
$$P(Y=y | X+Y=n) = \frac{p^2 (1-p)^{n-2}}{\sum_{k=1}^{n-1} (p)^2 (1-p)^{n-2}}$$

$$= \frac{1}{n-1}$$

$$P(Y=y | X+Y=n) = \begin{cases} 0 & ; y = n, n+1, n+2 \\ \frac{1}{n-1} & ; y = 1, 2, 3, \dots, n-1 \end{cases}$$

27/3/18 If set of events $\{A_1, A_2, A_3, \dots, A_n\}$ with $P(A_i) > 0$ partitioned the sample space S and if B is an arbitrary event then

$$P(B) = \sum_{i=1}^n P(B|A_i) P(A_i)$$



• Total Probability Law (Discrete case) :-

Theorem: Let X and Y be two discrete random variables

Then for any event B

$$P(B) = \sum_{y \in \text{Range}(Y)} P(B|Y=y) p_Y(y)$$

Example

solution

Proof :

If Y is a discrete RV with pmf $f_Y(y) = P(Y=y)$, then the set $\{[Y=y]\}_{y \in \text{RY}}$ partitioned the sample space.

Then by total Probability theorem

$$P(B) = \sum_{y \in \text{RY}} P(B|Y=y) P(Y=y)$$

$$\boxed{P(B) = \sum_{y \in \text{RY}} P(B|Y=y) f_Y(y)}$$

This result is just simple extension of the
total Probability Theorem.

Total Probability law : continuous case :-

Let X be a continuous RV. Then for any event B we have :-

$$\boxed{P(B) = \int_{-\infty}^{\infty} P(B|X=x) f_X(x) dx}$$

Example: Let X and Y be two independent RV on $(0, 1)$.
Then find $P(X^3 + Y > 1)$?

Solution

$$f_X(x) = \begin{cases} 1 & 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$f_Y(y) = \begin{cases} 1 & 0 < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$f_{XY}(xy) = \begin{cases} 1 & 0 < x < 1, 0 < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

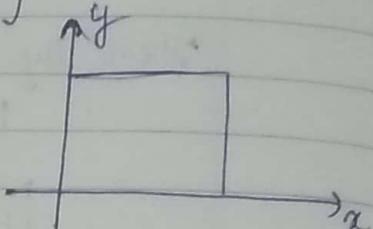
It is multiplication of $f_X(x)$ & $f_Y(y)$ because X & Y are independent Random Var.

Method 1 :-

$$P(x^3 + y > 1) = \iint_A f_{xy}(x, y) dx dy$$

$$A = \{(x, y) \in \mathbb{R}^2 \mid x^3 + y > 1\}$$

To find out set A in the given region we need to draw a curve $f(x) = 1 - x^3$



$$\lim_{x \rightarrow \infty} f(x) = -\infty \quad \lim_{x \rightarrow -\infty} f(x) = +\infty$$

It implies range of this f lies on whole Real line. Range $(-\infty, \infty)$

$$f'(x) = -3x^2 < 0 \quad \forall x \in (-\infty, \infty)$$

$f(x)$ is always decreasing

$$f'(x) = 0 \Rightarrow x=0 \text{ critical point}$$

It can be point of local Maxima or local Minima. $x=0$ is neither a point of local Maxima or Minima as in the nbd of $x=0$ $f(x)$ is always decreasing.

$$\rightarrow f''(x) = -6x < 0 \quad \forall x \in (0, \infty)$$

so on the positive axis function is concave down type and on negative axis function is concave up i.e convex

at $x=0$ $f''(x)=0 \Rightarrow x$ is a point of Inflection
(change its concavity)

→ ~~that~~ Region in ~~the~~ square for which $x^3 + y > 1$ should get satisfied.

concave up

(0, 1)

(1, 0)

$$y > 1 - x^3$$

$$1 - x^3 = y$$

$$\rightarrow \int_{x=0}^1 \int_{y=1-x^3}^1 f(x,y) dx dy$$

$$y \in (1-x^3, 1)$$

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

$$\rightarrow \int_0^1 [y]_{1-x^3}^1 dx$$

$$\Rightarrow \int_0^1 [1 - x + x^3] dx \Rightarrow \left[\frac{x^4}{4} \right]_0^1 = \frac{1}{4} \text{ Ans}$$

Method 2 :-

$$P(X^3 + Y > 1) = \int_{-\infty}^{\infty} P(X^3 + Y > 1 / x) f(x) dx$$

$$= \int_{-\infty}^{\infty} \int_0^1 P(X^3 + Y > 1 / x) dx$$

$$= \int_0^1 P(Y > 1 - x^3 / x) dx$$

$$= \int_0^1 P(Y > 1 - x^3) dx$$

$$= \int_0^1 \left[\int_{1-x^3}^1 f(y) dy \right] dx$$

$$\Rightarrow \int_0^1 [1 - x + x^3] dy$$

$$\Rightarrow \left[\frac{x^4}{4} \right]_0^1 \Rightarrow \frac{1}{4} \text{ Ans}$$

Conditional Expectation :

Discrete case -

Let X and Y be discrete random variable with conditional pmf $P_{X|Y}(x|y)$ of X given $Y=y$

Then conditional expectation

$$E[X|Y=y] = \sum_x x P_{X|Y}(x|y)$$

$$X, p(x) . \quad E(X) = \sum_x x p(x)$$

continuous case: Let X and Y be continuous RV with conditional pdf $f_{X|Y}(x|y)$ of X given $Y=y$.

$$E[X|Y=y] = \sum_x x P_{X|Y}(x|y) \rightarrow \text{discrete case}$$

$$E[X|Y=y] = \int_{-\infty}^{\infty} x f_{X|Y}(x|y) dx \quad \text{continuous case.}$$

Ex

Let X and Y be independent RV with geometric distribution of parameter p ; $p \in (0, 1)$. Calculate

$$E[Y|X+Y=n] \text{ where } n \geq 2$$

$$\sum_y y P_{X|Y}(x|y) \quad \text{conditional pmf we need to find out}$$

$$P_{X|Y}(y/n) = \begin{cases} 0 ; & y \geq n \\ \frac{1}{n-1} ; & y = 1, 2, \dots, n-1 \end{cases}$$

$$\begin{aligned} \sum_y P_{X|Y}(y/n) &= \sum_{y=1}^{n-1} y \cdot \frac{1}{n-1} = \frac{1}{n-1} \sum_{y=1}^{n-1} (y-1) \\ &= \frac{1}{(n-1)} \frac{(n-1)(n)}{2} \\ &= \frac{n}{2} \end{aligned}$$

Theorem

Proof

Theorem: Let X, Y be two discrete RV's with joint pmf $p_{XY}(x, y)$ Marginal pmf $p_X(x), p_Y(y)$ respectively Then

$$\rightarrow E(Y) = \sum_x E\left(\frac{Y}{X=x}\right) p_X(x)$$

unconditional expectation of Y .

Proof:

Take R.H.S.

$$\begin{aligned} \sum_x E\left[\frac{Y}{X=x}\right] p_X(x) &= \sum_x \sum_y y \frac{p_{Y|X}(y|x)}{p_X(x)} \\ p_{Y|X}(y|x) &= \frac{p_{XY}(x,y)}{p_X(x)} \\ - \sum_x \sum_y y p_{XY}(x,y) &= \sum_y y \sum_x p_{XY}(xy) \\ &= \sum_y y p_Y(y) \\ &= \underline{E(Y)} \end{aligned}$$

\rightarrow for X, Y continuous RV; then with joint pdf $f_{XY}(xy)$ Marginal pdf $f_X(x) f_Y(y)$

respectively Then

$$E(Y) = \int_{-\infty}^{\infty} E\left[\frac{Y}{X=x}\right] f_X(x) dx$$

Total expectation theorem.

Ex:Let X and Y be continuous RV with joint pdf

$$f_{XY}(x,y) = \begin{cases} 6(y-x) & 0 \leq x \leq y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

Find $E\left[\frac{Y}{X=x}\right]$ and then $E(Y)$?

sol

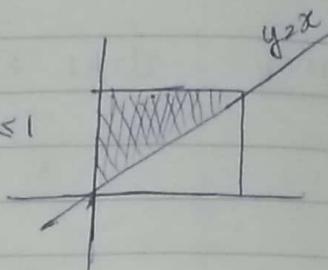
$$E[Y|X=x] = \int_{-\infty}^{\infty} y f_{Y|X}(y|x) dy$$

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

Marginal density f_x of $X \rightarrow$

$$f_X(x) = \int_{-\infty}^{\infty} f_{XY}(x,y) dy$$

$$= \begin{cases} \int_x^1 6(y-x) dy & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$



$$= \begin{cases} 3(x-1)^2 & 0 \leq x < 1 \\ 0 & \text{otherwise} \end{cases}$$

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

$$= \begin{cases} \frac{2(y-x)}{(x-1)^2} & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\Rightarrow E[Y|X=x] = \int_x^1 \frac{2y(y-x)}{(x-1)^2} dy \quad 0 \leq x \leq 1$$

$$= \frac{2}{(x-1)^2} \int_x^1 y(y-x) dy = \boxed{\frac{x^2+x+2}{3(x-1)} \quad 0 \leq x \leq 1}$$

→ expectation of Y

$$E(Y) = \int_0^1 E[Y|X=x] f_X(x) dx$$

$$= \int_{-\infty}^{\infty} \frac{x^2+x+2}{3(x-1)} \cdot 3(x-1)^2 dx$$

$$0 \int (x^2 + x + 2)(x - 1) dx$$

$$= \frac{3}{4}$$

→ complex valued Random Variables :-

$$x: S \rightarrow \mathbb{R}$$

$$z: x + iy$$

where x and y both are real.

$$z: S \rightarrow \mathbb{C} \quad (\text{complex plane})$$

$$E(z) = E(x + iy) = E(x) + iE(y)$$

$$E(az_1 + bz_2) = aE(z_1) + b(E(z_2))$$

→ Characteristic Function: let X be a RV then characteristic f^n of X defined by

$$\rightarrow \boxed{\phi_X(t) = E[e^{itX}] ; t \in \mathbb{R}}$$

characteristic f^n exist for all Random Vari.

$$\phi_X(t) = E[\cos(tx) + i\sin(tx)]$$

$\cos(tx)$ & $\sin(tx)$ both are f^n of

Random variable and bounded by 1

so their expectation will also be finite

charac. f^n exist for all R.V.

$\phi_X(t)$ exists for all RV. X and $t \in \mathbb{R}$

$$e^{itX} = \underbrace{\cos(tx)}_{g_1(x)} + i\underbrace{\sin(tx)}_{g_2(x)}$$

$$|g_1(x)| \leq 1 \quad \& \quad |g_2(x)| \leq 1$$

$$\boxed{E[e^{itX}] = E[\cos(tx) + i\sin(tx)]}$$

ensure the existence of RV in f^n for all R.V.

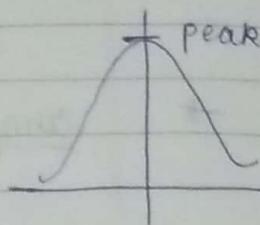
It is finite ; exists for all RV. $t \in \mathbb{R}$

skewness, contours?
(Peak)

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Ex $X \sim$ Bernoulli RV with parameter p ;
 $\phi_X(t) = ?$

→ Random variable is skewed
characteristic function
(P), ($1-P$)



$$E[g(x)] = \sum_x g(x) p_x(x)$$

$$E[e^{itx}] = \sum_x e^{itx} p_x(0)$$

$$\begin{aligned} &= e^{ito} p(x=0) + e^{its} p(x=1) \\ &= (1-p) + e^{it}(p) \end{aligned}$$

$$\boxed{\phi_X(t) = (1-p) + e^{it}(p)}$$

Ex $\phi_X(t)$ for $X \sim N(0,1)$ standard Normal Random.
 $\phi_X(t) = e^{-t^2/2}$ Prove it

Q/H/17 If for a RV, X we have distribution function and density function then we have its characteristic function also

Properties of characteristic function :-

Ex Let X be a RV and a and b are real countable

then :

$$\boxed{\phi_{a+bx}(t) = e^{iat} \phi_X(bt)}$$

Proof : consider LHS

$$\phi_{a+bx}(t) = E[e^{it(a+bx)}]$$

$$\begin{aligned} &= E[e^{iat}, e^{ibtx}] = e^{iat} \cdot E[e^{ibtx}] \\ &= e^{iat} \phi_X(bt) \end{aligned}$$

ex

$$X \sim N(\mu, \sigma^2)$$

$$Y = \frac{X-\mu}{\sigma} \sim N(0, 1)$$

Mean = 0

Variance = 1

$$\phi_Y(t) = e^{-t^2/2}$$

what is $\phi_X(t)$?

$$X = \mu + \sigma(Y)$$

$$\begin{aligned}\phi_X(t) &= \phi_{\mu+\sigma Y}(t) = e^{i\mu t} \phi_Y(\sigma t) \\ &= e^{i\mu t} e^{-\sigma^2 t^2/2}\end{aligned}$$

②

Let X and Y be two Independent Random Var.

then $\boxed{\phi_{x+y}(t) = \phi_x(t) \phi_y(t)}$

Proof LHS $\phi_{x+y}(t) = E[e^{it(x+y)}]$

$$\rightarrow E[e^{ixt} \cdot e^{iyt}]$$

$$\rightarrow E[e^{ixt}] \cdot E[e^{iyt}]$$

Because X and Y are Independent RV

$$\rightarrow \boxed{\phi_x(t) \phi_y(t)}$$

Corollary :- Let $x_1, x_2, x_3, \dots, x_n$ be n independent Random variables

$$\phi_{x_1+x_2+x_3+\dots+x_n}(t) = \phi_{x_1}(t) \phi_{x_2}(t) \dots \phi_{x_n}(t)$$

Ex → find out the ch. function of a Random variable $X \sim B(n, p)$.

If X is a Binomial random variable then it can be written as $\sum_{n \text{ Independent}}$

Binomial Trials
Bernoulli

$$X = \underbrace{Y_1 + Y_2 + Y_3 + \dots + Y_n}_{\text{Bernoulli Trials}}$$

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$$X = X_1 + X_2 + X_3 + \dots + X_n$$

where X_i are n independent Bernoulli RV with parameter p .

$$\begin{aligned}\phi_X(t) &= \phi_{X_1 + X_2 + X_3 + \dots + X_n}(t) \\ &= \phi_{X_1}(t) \cdot \phi_{X_2}(t) \cdot \phi_{X_3}(t) \cdots \phi_{X_n}(t) \\ &\rightarrow [p e^{it} + (1-p)]^n\end{aligned}$$

Theorem :- Uniqueness Theorem :- Let X and Y be random variables s.t. $\phi_X = \phi_Y$. Then $X = Y$ i.e. X and Y will have same distribution.

$$\begin{aligned}\text{Proof :- } \phi_X(t) &= \phi_X(t) \\ E[e^{itX}] &= E[e^{itY}] \\ e^{itX} &= e^{itY} \\ X &= Y \quad \text{Hence proved.}\end{aligned}$$

Ex : Let $X \sim B(n_1, p)$ & $Y \sim B(n_2, p)$ with X and Y are independent.

$$\begin{aligned}X+Y &\sim B(n_1+n_2, p) \\ \text{Binomially distributed}\end{aligned}$$

$$\begin{aligned}\phi_{X+Y}(t) &= \phi_X(t) \cdot \phi_Y(t) \\ &\rightarrow [p e^{it} + (1-p)]^{n_1} [p e^{it} + (1-p)]^{n_2} \\ &\rightarrow [p e^{it} + (1-p)]^{n_1+n_2}\end{aligned}$$

RHS is the ch function of a Binomial RV say Z with parameter (n_1+n_2, p) while LHS is the ch function of $X+Y$.

By uniqueness theorem we can say
 $x+y = z$
 $x+y \sim B(n_1+n_2, p)$

Ex $X \sim N(\mu_1, \sigma_1^2)$ & $Y \sim N(\mu_2, \sigma_2^2)$
 Both are independent then show that
 $\rightarrow X+Y \sim N(\mu_1+\mu_2, \sigma_1^2 + \sigma_2^2)$.

Inequalities :-

1) Jensen's Inequality :- Let X be a RV and $f: I \rightarrow R$ be a convex function where $I \subseteq R$ in an interval. Then

$$E[f(x)] \geq f[E(x)]$$

provided expectations exist.

If f is concave then :-

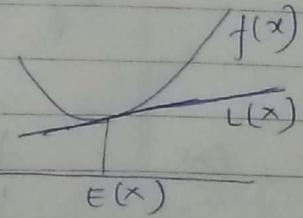
$$E[f(x)] \leq f[E(x)]$$

Proof of this inequality \rightarrow

Assumption :- function f is differentiable

Proof :- let $f: I \rightarrow R$ be differentiable

Let $E(x) = a+bx$ be a line which is tangent to $f(x)$ at $E(x)$.



Due to differentiability $L(x)$ always lie below the curve

$$\cdot E[f(x)] \geq E[L(x)]$$

because this line always lie below the curve

$$E[a+bx] = a+bE(x)$$

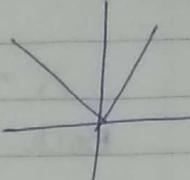
$$= L(E(x)) = f(E(x))$$

$$E(f(x)) \geq f(E(x))$$

Ex: $f(x) = |x|$

function is defined on $I \rightarrow R$

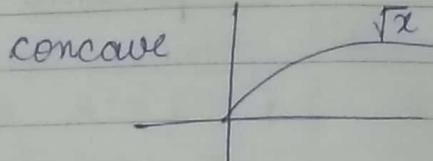
→ By the graph it is clear that graph is ~~convex~~ concave.



$$E[|x|] \geq |E[x]| \quad \text{for all Real Values.}$$

Ex: $f(x) = \sqrt{x} \quad ; \quad x > 0$

$$E[\sqrt{x}] \leq \sqrt{E[x]}$$



Ex: $|E(x)| \leq E[|x|] \leq \sqrt{E(x^2)}$

x^2 is always positive; from previous example we have $E[\sqrt{x}] \leq \sqrt{E[x]}$

$$\begin{matrix} \rightarrow & E[\sqrt{x^2}] \leq \sqrt{E(x^2)} \\ x^2 > 0 \quad \forall x \in R & \downarrow \end{matrix} \quad \text{True for all } x \in R$$

→ $E[|x|] \leq \sqrt{E(x^2)}$

by above example

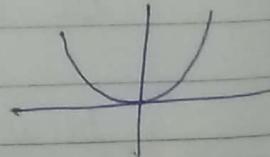
$$|E(x)| \leq E[|x|] \leq \sqrt{E(x^2)}$$

Ex: $\text{Var}(x) \geq 0$

$$\text{Var}(x) = E(x^2) - [E(x)]^2$$

→ consider $f(x) = x^2 \quad \forall x \in R$

convex



→ $E(x^2) \geq (E(x))^2$

→ $E(x^2) - [E(x)]^2 \geq 0$

$$\boxed{\text{Var}(x) \geq 0}$$

→ Moments of a Random Variable :-

Let μ be a positive real number and X be a RV. Then n -th moment of X about a point a is defined by : $E[(X-a)^n]$

- If we set $a=0$; then $E[X^n]$ is the n th moment of RV X about the origin.
- If we set $a=\mu$ (expectation of X)
Then

$E[(X - E(X))^n]$ gives us n th moment of X about its mean.

- Skewness is obtained with the help of moment
- Variance (spread), skewness, corssis.

- We know that $E[X^n]$ exists and is finite if $E[|X^n|]$ exists and is less than ∞

$$E[X^n] = \sum_x x^n p_x(x)$$

$$\sum_x x^n p_x(x) \leq \left| \sum_x x^n p_x(x) \right| \\ \leq \sum_x |x^n| p_x(x)$$

PMF always take positive value

$$\rightarrow = \sum_x |x^n| / p_x(x) = E[|X|^n]$$

$$\sum_x x^n p_x(x) \leq \sum_x |x^n| / p_x(x)$$

Ex If the moment (about origin) of order $q > 0$ exists for a random variable x , then show that moments of order p exists where $0 < p < q$

Sol: $f: (0, \infty) \rightarrow \mathbb{R}$ $f(x) = x^{\mu}$ where $\mu > 1$

$$f''(x) = \mu(\mu-1)x^{\mu-2}$$

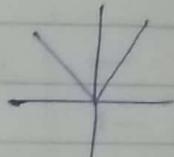
convex $\forall x \in (0, \infty)$

$$\Rightarrow E[x^{\mu}] \geq [E(x)]^{\mu}$$

$$[E(x)]^{\mu} \leq E[x^{\mu}] \leq E[|x|^{\mu}]$$

$$\Rightarrow [E(x)] \leq (E[|x|^{\mu}])^{1/\mu} \quad \text{--- (1)}$$

$0 < p < q$; then $q/p > 1$



Replace x by q/p in (1) we get

$$[E(x)] \leq (E[|x|^{q/p}])^{p/q} \quad \text{--- (2)}$$

set x by x^p we get

$$E[x^p] \leq (E[x]^q)^{p/q}$$

p^{th} Moment exist if q^{th} Moment exist
if high moment exist low moment also exists

Ex: Let x be a RV with $E(x) = 10$ show that
 $E[\ln \sqrt{x}] \leq \frac{1}{2} \ln 10$

Sol $f(x) = \ln \sqrt{x} = \frac{1}{2} \ln x$

$$f'(x) = \frac{1}{2x} \quad f''(x) = -\frac{1}{2x^2} < 0 \quad \text{concave}$$

$$E[\ln \sqrt{x}] \leq \ln \sqrt{10} = \frac{1}{2} \ln 10$$