

## Method 2.

$$M_x(t) = E(e^{tx})$$

$$= E\left(1 + \frac{tx}{1!} + \frac{t^2 x^2}{2!} + \frac{t^3 x^3}{3!} + \dots + \frac{t^r x^r}{r!} + \dots\right)$$

$$= 1 + \frac{t}{1!} E(x) + \frac{t^2}{2!} E(x^2) + \frac{t^3}{3!} E(x^3) + \dots + \frac{t^r}{r!} (E x^r)$$

the coefficient of  $\frac{t^r}{r!}$  in the expansion of mgf is the  $r^{\text{th}}$  moment.

$E(x^r) \Rightarrow r^{\text{th}}$  moment

$E(x - \mu)^r \Rightarrow r^{\text{th}}$  central moment

$$\text{mgf } M_x(t) = E(e^{tx})$$

Find the mgf of Binomial distribution with parameter  $n$  & hence find the mean & variance.

Proof of Binomial distribution:  $f(x) = {}^n C_x p^x q^{n-x}$   $x=0, \dots, n$

$$\text{mgf } \underline{M_x(t)} = E(e^{tx})$$

$$= \sum e^{tx} {}^n C_x p^x q^{n-x}, \quad x=0, \dots, n$$

$$= \sum {}^n C_x (pe^t)^x q^{n-x}$$

$$\stackrel{\text{WKT}}{=} \sum {}^n C_x p^x q^{n-x} = (pq)^n$$

$$\boxed{M_x(t) = (pe^t + q)^n}$$

$$E(x) = M'_x(t) \Big|_{t=0} = n(pe^t + q)^{n-1} \cdot pe^t \Big|_{t=0}$$

$$\boxed{M_x(t) \Big|_{t=0} = np}$$



$$\begin{aligned}
 E(x^2) &= M''_2(t) \Big|_{t=0} = \frac{(np)}{e^t} \Big|_{t=0} \\
 &= np \left[ (pe^t + q)^{n-1} e^t + e^t (n-1)(pe^t + q)^{n-2} pe^t \right] \\
 &= np \left[ 1 + (n-1)p \right] \\
 &= np \left[ 1 + np - p \right] \\
 &= np \left[ q + np \right] \\
 &= npq + n^2 p^2
 \end{aligned}$$

$$\begin{aligned}
 \text{Variance} &= E(x^2) - E(x)^2 \\
 &= npq + n^2 p^2 - n^2 p^2 \\
 &= npq
 \end{aligned}$$

$$\begin{aligned}
 \text{Mean} &= E(x) \\
 &= np
 \end{aligned}$$

Find the mgf of exponential distribution with parameter  $\lambda$

$$\begin{aligned}
 M_x(t) &= E(e^{tx}) \\
 &= \int_0^{\infty} e^{tx} \lambda e^{-\lambda x} dx \\
 &= \frac{\lambda}{\lambda - t} \\
 E(x) &= M'_x(t) \Big|_{t=0} \\
 &= \lambda \frac{-1}{(\lambda - t)^2} (-1) \Big|_{t=0} \\
 &= \frac{1}{\lambda}
 \end{aligned}$$



$$E(x^2) = \lambda \frac{(-2)}{(\lambda - t)^3} (-1) \Big|_{t=0}$$

$$= \frac{2}{\lambda^3}$$

$$\text{Variance}(x) = E(x^2) - E(x)^2$$

$$= \frac{2}{\lambda^2} - \frac{1}{\lambda^2}$$

$$= \frac{1}{\lambda^2}$$

$$\text{Mean} = E(x) = \frac{1}{\lambda}$$

$$\text{Variance} = E(x^2) - E(x)^2 = \frac{1}{\lambda^2}$$

$$M_x(t) = E(e^{tx})$$

$$= \sum e^{tx} \frac{e^{-\lambda} \lambda^x}{x!} \quad x=0, 1, 2$$

$$= e^{\lambda(e^t - 1)}$$

$$= e^{-\lambda} \sum \frac{(e^t \lambda)^x}{x!}$$

$$= e^{-\lambda} e^{e^t \lambda} \Big|_0^2$$

$$= e^{\lambda(e^t - 1)}$$

$$E(x) = M_x'(t) \Big|_{t=0}$$

$$= e^{\lambda(e^t - 1)} \lambda e^t \Big|_{t=0}$$

$$= \lambda$$



Properties of mgf:

1) Two r.v  $x$  &  $y$  with same mgf have same probability distribution

$\mu$  = mgf of a r.v is unique

$$\text{If } M_x(t) = M_y(t)$$

$$2) M_x(0) = 1$$

3) mgf of  $cx = M_{cx}(t)$  In mgf of  $x$ , replace  $t$  by  $ct$

$$E(x) = \mu$$

$$E(cx) = cE(x)$$

$$\mu_x(t) = E(e^{tx})$$

$$\mu_{c+x}(t) = E(e^{t(c+x)})$$

$$= E[e^{ct} \cdot e^{xt}]$$

$$= e^{ct} \mu_x(t)$$

$$\mu_{c+x}(t) = e^{ct} \mu_x(t)$$

Q) mgf of the sum of 2 independent r.v.s.

If  $x$  &  $y$  are independent  $\Rightarrow \mu_{x+y}(t) = \mu_x(t) \cdot \mu_y(t)$

$$\mu_{x+y}(t) = E(e^{t(x+y)})$$

$$= E[e^{tx} e^{ty}]$$

$$= E(e^{tx}) E(e^{ty}) \dots x \text{ \& } y \text{ are independent}$$

$$= \mu_x(t) \mu_y(t)$$



Q) using mgf of S ND find mgf of ND with mean  $\mu$   
 & SD  $\sigma$  mgf of S ND  $M_X(t) = e^{t^2/2}$  &  $Z = \frac{X - \mu}{\sigma}$   
 $X = \mu + Z\sigma$

$$\begin{aligned} \text{mgf of ND } M_X(t) &= M_{\mu + \sigma Z}(t) \\ &= e^{\mu t} M_{\sigma Z}(t) \text{ - using prop. } M_{aX+b}(t) = e^{at} M_X(t) \\ &= e^{\mu t} M_Z(\sigma t) \text{ using } M_{cX}(t) = M_X(ct) \\ &= e^{\mu t} e^{(\sigma t)^2/2} \end{aligned}$$

$$M_X(t) = e^{\mu t + \sigma^2 t^2/2}$$

$$\begin{aligned} F(x) &= M_X(t)_{t=0} = e^{\mu t + \sigma^2 t^2/2} \\ &= \mu \quad \leftarrow (\mu + \sigma^2 t)_{t=0} \end{aligned}$$

Q) let  $X$  be a r.v with mgf  $M_X(t) = e^{2t^2}$  with mgf of  $N(\mu, \sigma)$

$$\begin{aligned} \mu &= 0 \text{ \& } \frac{\sigma^2}{2} \Rightarrow \sigma^2 = 4 \\ \sigma &= 2 \end{aligned}$$

$$X \sim N(\mu=0, \sigma=2) \Rightarrow Z = \frac{X - \mu}{\sigma} = \frac{X - 0}{2}$$

$$\begin{aligned} P[0 < X < 1] &= P[0 < Z < 1/2] = \Phi(0.5) - \Phi(0) \\ &= 0.6915 - 0.5 = 0.1915 \end{aligned}$$

Q) let  $X$  be a r.v with pmf  $p(x) = 2(1/3)^x$ ,  $x = 1, 2, \dots$  find mgf of  $X$  &  $E(X)$

$$\begin{aligned} M_X(t) &= E(e^{tx}) \\ &= \sum_{x=1}^{\infty} e^{tx} 2(1/3)^x, \quad x = 1, 2, \dots \\ &= 2 \sum_{x=1}^{\infty} (e^t/3)^x \\ &= 2 \left\{ \frac{e^t}{3} + \left(\frac{e^t}{3}\right)^2 + \dots \right\} \\ &= 2 \left\{ \frac{1}{1 - e^{t/3}} - 1 \right\} \\ &= 2 \left( \frac{3}{3 - e^t} - 1 \right) \end{aligned}$$



## Sum of independent random variables

If  $x_1, \dots, x_n$  are independent rvs with mgfs  $M_{x_1}(t), \dots, M_{x_n}(t)$ , then the mgf of  $X = x_1 + \dots + x_n$  is

$$M_X(t) = M_{x_1}(t) + \dots + M_{x_n}(t)$$

1) Sum of 2 independent ~~random~~ normal random variable is a normal random variable

Proof

$$\text{let } x_1 \sim N(\mu_1, \sigma_1)$$

$$x_2 \sim N(\mu_2, \sigma_2)$$

$$\text{let } X = x_1 + x_2$$

$$M_X(t) = M_{x_1}(t) \cdot M_{x_2}(t)$$

$$= e^{\mu_1 t + \frac{\sigma_1^2 t^2}{2}} \cdot e^{\mu_2 t + \frac{\sigma_2^2 t^2}{2}}$$

$$= e^{(\mu_1 + \mu_2)t + \frac{(\sigma_1^2 + \sigma_2^2)t^2}{2}}$$

normal of mgf

$$M_X(t) = e^{\mu t + \frac{\sigma^2 t^2}{2}}$$

This is in the form of Normal of mgf

$X$  is normal with mean  $\mu_1 + \mu_2$

& variance  $\sigma_1^2 + \sigma_2^2$



2) If  $x_1, \dots, x_n$  are independent binomial random variables with parameters  $(n_1, p), \dots, (n_k, p)$

then  $x_1 + \dots + x_n$  is binomial with parameters  $(n_1 + \dots + n_k, p)$

(Be cos  $p$  is a common parameter for all  $x_i$ )

Proof:

$$\text{mgf of } x = x_1 + \dots + x_n$$

$$\boxed{\text{mgf of Binomial } (q + pe^t)^n}$$

$$M_x(t) = (q + pe^t)^{n_1} \times \dots \times (q + pe^t)^{n_k}$$

$$= (q + pe^t)^{n_1 + \dots + n_k}$$

3) If  $x_1, \dots, x_k$  are independent poisson random variables with parameters  $\lambda_1, \dots, \lambda_k$  then  $x_1 + \dots + x_k$  is a poisson random variable with mean  $\lambda_1 + \dots + \lambda_k$

Proof:

$$M_x(t) = e^{\lambda_1(e^t - 1)} \times \dots \times e^{\lambda_k(e^t - 1)}$$

$$= e^{\underbrace{(\lambda_1 + \dots + \lambda_k)}_{\lambda}(e^t - 1)}$$

$$\boxed{\text{mgf of poisson } e^{\lambda(e^t - 1)}}$$

**ERLONG / GAMMA distribution.**

ERLONG distribution

Generalization of expo distribution

exponential distribution describes the time b/w successive events

While expo distribution describes the time b/w successive events, the  $k$  Erlong random variable describes the time b/w any and the  $k$ th occurrence of the event.

exponential  $x$  : Time for the next arrival  
(Time b/w one arrival & next arrival)



Erlang  $y$ : Time for the  $K^{\text{th}}$  arrival

eg:

time for 4<sup>th</sup> arrival.

$y$ : 4 Erlang.

Pdf

$$f(x) = \frac{\lambda e^{-\lambda x} (\lambda x)^{K-1}}{\Gamma(K)}, \quad x \geq 0$$

parameters  $\lambda$  &  $K$

When  $K=1 \rightarrow$  Erlang distribution reduces to exponential distribution with parameter  $\lambda$

Bees,

$y = 1$  Erlang = exponential distribution

Mean & variance.

Mean  $E(x) = \frac{K}{\lambda}$

$\text{var}(x) = K/\lambda^2$

Erlang is for  $K^{\text{th}}$  arrival.

expo is for 1<sup>st</sup> arrival.

For expo distribution

$$E(x) = 1/\lambda$$

$$\text{var}(x) = 1/\lambda^2$$

The generalized form of Erlang dis. is called gamma distribution



For any real number  $r$ , a continuous r.v.,  $x$  follows gamma ~~distribution~~ distribution is its pdf is given by

pdf

$$f(x) = \frac{\lambda e^{-\lambda x} (\lambda x)^{r-1}}{\Gamma(r)}, \quad x \geq 0$$

$x \sim \text{Gamma}(r, \lambda)$

Mean  $E(x) = r/\lambda$

variance  $= r/\lambda^2$

sum of ~~all~~ exponential r.v is Erlang r.v

mgf of Exponential r.v  $= \frac{\lambda}{\lambda - t}$

mgf of  $r$ -Erlang  $= \left( \frac{\lambda}{\lambda - t} \right)^r$

To find mgf of  $r$ -erlang distribution.

pdf of  $r$ -Erlang

$$f(x) = \frac{\lambda e^{-\lambda x} (\lambda x)^{r-1}}{\Gamma(r)}, \quad x \geq 0$$

mgf  $M_x(t) = E(e^{tx})$

$$= \int_0^{\infty} e^{tx} \cdot \frac{\lambda e^{-\lambda x} (\lambda x)^{r-1}}{\Gamma(r)} dx$$



$$= \frac{\lambda^r}{\Gamma(r)} \int_0^{\infty} e^{-(\lambda-t)x} x^{r-1} dx$$

$$= \frac{\lambda^r}{\Gamma(r)} \cdot \frac{\Gamma(r)}{(\lambda-t)^r}$$

WKT

$$\int_0^{\infty} e^{-ax} x^{n-1} dx = \frac{\Gamma(n)}{a^n}$$

$$= \left( \frac{\lambda}{\lambda-t} \right)^r$$

Sum of  $r$  exponential r.v with parameter  $\lambda$   
 is  $r$ -Erlang r.v with parameter  $r$  and  $\lambda$

$$X = x_1 + \dots + x_r$$

Proof:

$$M_X(t) = M_{x_1}(t) + \dots + M_{x_r}(t)$$

$$= \left( \frac{\lambda}{\lambda-t} \right) \times \dots \times \left( \frac{\lambda}{\lambda-t} \right)$$

$$= \left( \frac{\lambda}{\lambda-t} \right)^r$$

$$\text{Mgf of } r \text{ Erlang} = \left( \frac{\lambda}{\lambda-t} \right)^r$$



## Joint probability distribution

When 2 or more random variable defined on the joint same sample space.

consider,

$x \rightarrow$  height of a student

$y \rightarrow$  weight of the same student.

$f_{xy}(x, y) \rightarrow$  joint pdf

WKT

$$F(x) = P[X \leq x]$$

$P_{xy}(x, y) = P[X = x, Y = y] \rightarrow$  joint pmf

$F_{xy}(x, y) = P[X \leq x, Y \leq y] \rightarrow$  joint cdf.

The individual random variable.

The individual prob distri of  $x$  &  $y$  are called marginal distributions denoted by  $f_x(x), f_y(y)$

Marginal distributions of  $x$  and  $y$

Let  $P_{xy}(x, y)$  be the joint prob of  $x$  and  $y$ , then marginal distribution of  $x$

$$P_x(x) = \sum_{y \in \mathcal{Y}} P_{xy}(x, y)$$

$$P_y(y) = \sum_{x \in \mathcal{X}} P_{xy}(x, y)$$



Eg:

2 dice are rolled. let  $x$  denotes max of the 2 throws,  $y$  denotes the 'no. of times an even no appears'. Find the joint pmf of  $x$  and  $y$ . Also find the marginal distribution of  $x$  &  $y$ .

$y = \{0, 1, 2\}$  num. of even no. appears  
 $x = \{1, 2, 3, 4, 5, 6\}$

$x \backslash y$		0	1	2	Pdf	
(1,1)	1	$\frac{1}{36}$	0	0	$\frac{1}{36}$	$P_x(1)$
(1,2) (2,1) (2,2)	2	0	$\frac{2}{36}$	$\frac{1}{36}$	$\frac{3}{36}$	$P_x(2)$
(3,1) (3,2) (1,3) (2,3) (2,4)	3	$\frac{3}{36}$	$\frac{2}{36}$	0	$\frac{5}{36}$	$P_x(3)$
(4,1) (4,2) (4,3) (4,4) (3,4) (2,4) (1,4)	4	0	$\frac{4}{36}$	$\frac{3}{36}$	$\frac{7}{36}$	$P_x(4)$
	5	$\frac{5}{36}$	$\frac{4}{36}$	0	$\frac{9}{36}$	$P_x(5)$
	6	0	$\frac{6}{36}$	$\frac{5}{36}$	$\frac{11}{36}$	$P_x(6)$
		$\frac{9}{36}$	$\frac{13}{36}$	$\frac{9}{36}$	<b>1</b>	
		$P_y(0)$	$P_y(1)$	$P_y(2)$		

This is the joint pmf of  $x, y$ .

$P_x$  here is the marginal probability distribution of  $x$

$P_y(0), P_y(1), P_y(2)$  is the marginal distribution of  $y$



Marginal distribution of  $x$

$x$	1	2	3	4	5	6
$P_x(x)$	$\frac{1}{36}$	$\frac{3}{36}$	$\frac{5}{36}$	$\frac{7}{36}$	$\frac{9}{36}$	$\frac{11}{36}$

Marginal distribution of  $y$

$y$	0	1	2
$P_y(y)$	$\frac{9}{36}$	$\frac{18}{36}$	$\frac{9}{36}$

Properties of joint pmf/pdf.

Pmf

$$\rightarrow P_{xy}(x, y) \geq 0$$

$$\rightarrow \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} P(x, y) = 1$$

pdf

$$\rightarrow f_{xy}(x, y) \geq 0$$

$$\rightarrow \iint f_{xy}(x, y) dy dx = 1$$

Q1  $f_{xy}(x, y)$  is the joint pdf of  $x$  &  $y$  then

$$\text{marginal distribution of } x : f_x(x) = \int f(x, y) dy$$

$$\text{marginal distribution of } y : f_y(y) = \int f(x, y) dx$$

Note

Q1  $x$  &  $y$  are independent rv

$$f(x, y) = f_x(x) f_y(y)$$



Conditional probability .

conditional distribution of  $Y$  given  $X = x$

For 2 events  $A, B$

$$P(A|B) = \frac{P(A \cap B)}{P(B)}, \quad P(B) \neq 0$$

For 2 random variable  $X$  &  $Y$  with joint pmf/pdf  $f(x, y)$ , the conditional distribution of  $Y$  given  $X = x$  is given by

$$f_{Y|X=x}(y|x) = \frac{f(x, y)}{f_X(x)}, \quad f_X(x) \neq 0$$

function of  $y$

Conditional expectation & variance

The conditional expectation of  $Y$  given  $X = x$  is given by

if  $X$  &  $Y$  are discrete

$$E[Y|X=x] = \sum_{y \in Y} y \cdot f_{Y|X}(y|x)$$

if  $X$  &  $Y$  are continuous

$$E[Y|X=x] = \int y f_{Y|X}(y|x) dy$$

value  
sum of the  
random variable  
and conditional  
probability



$$E[Y^2 | X=x] = \sum y^2 f_{y|x}(y|x)$$

Variance

$$\text{Var}[Y | X=x] = E[Y^2 | X=x] - [E[Y | X=x]]^2$$

Result

For 2 r.v.  $X$  &  $Y$  with joint pmf/pdf  $F(x, y)$

$$E[E(X|Y)] = E(X)$$

Proof:

LHS

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

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x \boxed{f_{X|Y}(x|y) f_Y(y)} dy$$

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

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

$$= \int_{-\infty}^{\infty} f_X(x) x dx$$

$$= E(X)$$

WKT

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



the joint pdf of r.v's  $x$  and  $y$  is

$$f(x, y) = \begin{cases} kxy^2 & 0 \leq x \leq y \leq 1 \\ 0 & \text{o/e} \end{cases}$$

Find i) the value of  $k$

ii) Marginal distribution of  $x$  and  $y$

iii)  $E(x)$  and  $E(y)$

iv) if  $x$  &  $y$  are independent

$$f(x, y) = f_x(x) f_y(y)$$

$$E(xy) = E(x) E(y)$$

$$\text{Var}(x+y) = \text{Var}(x) + \text{Var}(y) + 2 \cdot$$

i)

For joint pdf

with respect to  $y$

$$\int \int f(x, y) dy dx = 1$$

$$= \int_0^1 \int_x^1 kxy^2 dy dx = 1$$

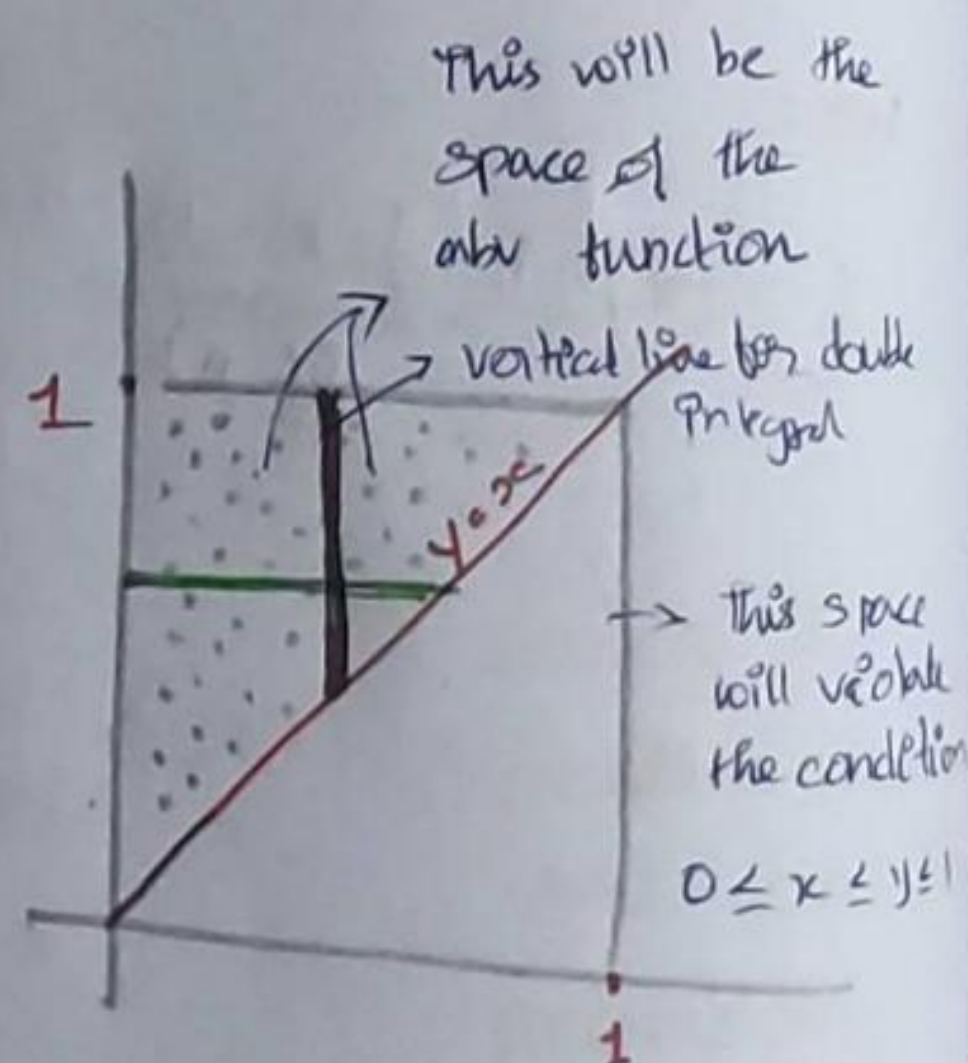
$$\int_0^1 kx \left[ \frac{y^3}{3} \right]_x^1 dx = 1$$

$$\frac{k}{3} \int_0^1 (x - x^4) dx = 1$$

$$\frac{k}{3} \left[ \frac{x^2}{2} - \frac{x^5}{5} \right]_0^1 = 1$$

$$\frac{k}{3} \left[ \left( \frac{1}{2} - \frac{1}{5} \right) - 0 \right] = 1$$

$$k = 10$$





$$\boxed{\iint f(x, y) dx dy = 1}$$

with respect to ...

$$\int_0^1 \int_0^y k x y^2 dx dy = 1$$

$$\int_0^1 k y^2 \left[ \frac{x^2}{2} \right]_0^y dy = 1$$

$$\frac{k}{2} \int_0^1 y^4 dy = 1$$

$$\frac{k}{2} \left[ \frac{y^5}{5} \right]_0^1 = 1$$

$$\frac{k}{2} \left[ \frac{1}{5} \right] = 1$$

$$\boxed{k = 10}$$

Marginal distribution of  $x$  with respect to  $y$

$$f_x(x) = \int_x^1 f(x, y) dy$$

Marginal distribution of  $y$  with respect to  $x$

$$f_y(y) = \int_0^y f(x, y) dx$$

$$f_x(x) = \int f(x, y) dy$$

$$= \int_x^1 10 x y^2 dy$$

$$= 10 x \left[ \frac{y^3}{3} \right]_x^1$$

$$= \frac{10}{3} x (1 - x^3)$$

$$f_x(x) = \frac{10}{3} x (1 - x^3) \quad 0 \leq x \leq 1$$



$$\begin{aligned}
 f_y(y) &= \int f(x, y) dx \\
 &= \int_0^y 10xy^2 dx \\
 &= 5y^4
 \end{aligned}$$

$$\begin{aligned}
 \text{iii]} \quad E(x) &= \int x f_x(x) dx \\
 &= \int_0^1 x \left( \frac{10}{3} x (1-x^3) \right) dx \\
 &= \frac{10}{3} \int_0^1 (x^2 - x^5) dx \\
 &= \frac{10}{3} \left[ \frac{x^3}{3} - \frac{x^6}{6} \right]_0^1 \\
 &= \frac{10}{3} \left[ \frac{1}{3} - \frac{1}{6} - 0 \right] \\
 &= \frac{10}{3} \times \frac{1}{6} = \frac{5}{9}
 \end{aligned}$$

$$\begin{aligned}
 E(y) &= \int y f_y(y) dy \\
 &= \int_0^1 y \cdot 5y^4 dy \\
 &= \frac{5}{6}
 \end{aligned}$$



iv] If  $x$  &  $y$  are independent

WKT  $f(x, y) = f_x(x) f_y(y)$

But here

$$f_x(x) f_y(y) = \frac{10}{3} x(1-x^3) 5y^4$$

$$\neq f(x, y)$$

So they are not dependent

$$E(xy) = \iint xy f(x, y) dy dx$$

$$= \int_0^1 \int_0^1 xy \cdot 10x y^3 dy dx$$

$$= 10 \int_0^1 x^2 \left[ \frac{y^4}{4} \right]_0^1 dx$$

$$= 10 \int_0^1 x^2 \left[ \frac{1}{4} - \frac{x^4}{4} \right] dx$$

$$= \frac{10}{4} \int_0^1 x^2 (1 - x^4) dx$$

$$= \frac{5}{2} \left[ \frac{x^3}{3} - \frac{x^7}{7} \right]_0^1$$

$$= \frac{5}{2} \times \frac{4}{21}$$

$$= \frac{10}{21}$$

$$\text{Then } E(x) \cdot E(y) = \frac{5}{9} \cdot \frac{5}{6}$$

$$= \frac{25}{54} \neq E(xy)$$

Not independent



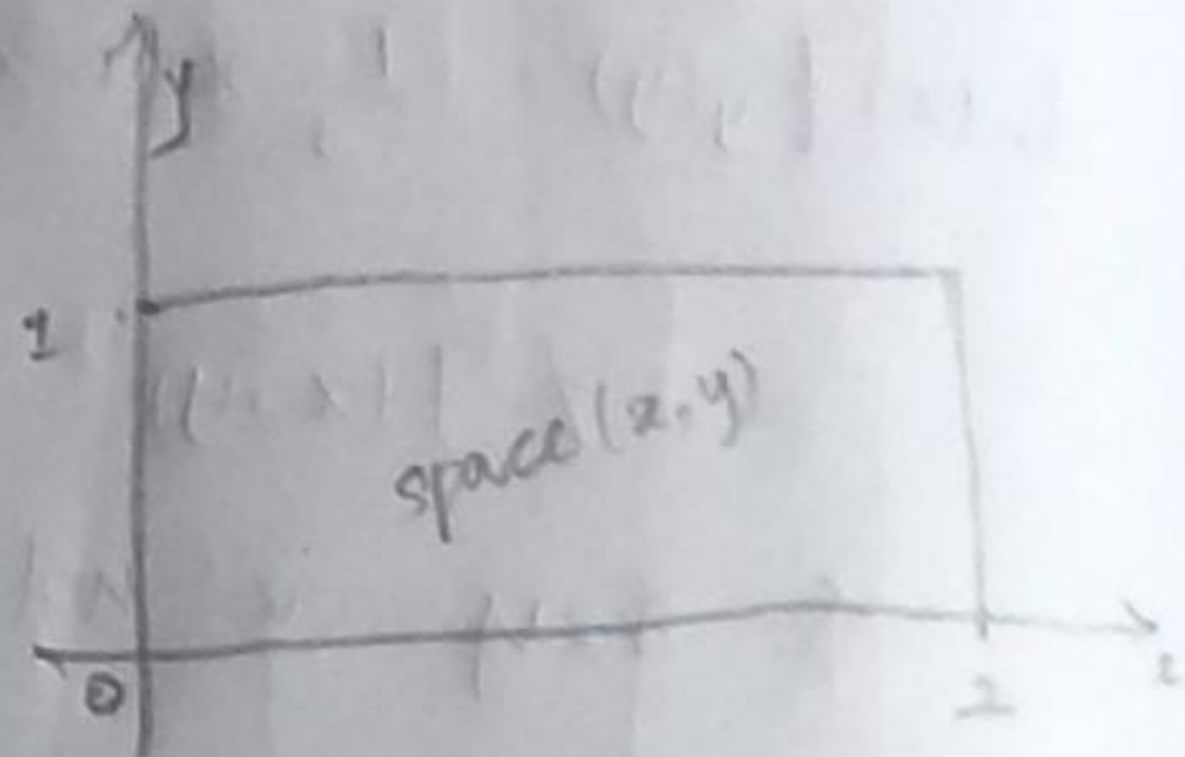
$$f(x, y) = \begin{cases} xy^2 + \frac{x^2}{8} \\ 0 \end{cases}$$

$$0 \leq x \leq 2, 0 \leq y \leq 1$$

O/w

Find

- (i)  $P[x > 1]$
- (ii)  $P[y < 1/2]$
- (iii)  $P[x > 1, y < 1/2]$
- (iv)  $P[x < y]$
- (v)  $P[y < 1/2 | x > 1]$
- (vi)  $P[x + y \leq 1]$



$$\begin{aligned} f_x(x) &= \int f(x, y) dy \\ &= \int_0^1 \left( xy^2 + \frac{x^2}{8} \right) dy \\ &= \left[ \frac{xy^3}{3} + \frac{x^2 y}{8} \right]_0^1 \end{aligned}$$

$$= \frac{x}{3} + \frac{x^2}{8}, 0 \leq x \leq 2.$$

$$= \frac{8x + 3x^2}{24}$$

$$f_y(y) = \int f(x, y) dx$$

$$= \int \left( 2y^2 + \frac{1}{3} \right) dx$$



$$(i) P[X > 1]$$

$$= \int_1^2 f_X(x) dx$$

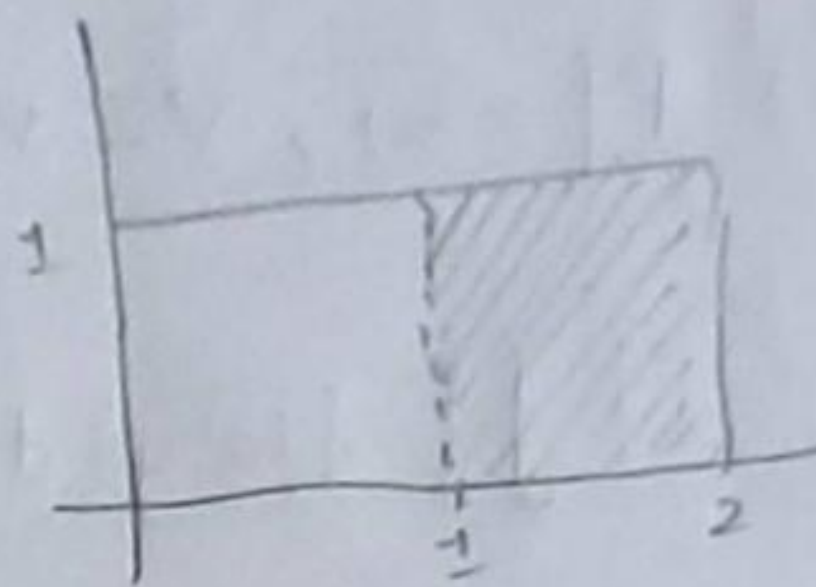
$$= \int_1^2 \left( \frac{x^2}{3} + \frac{x^2}{8} \right) dx$$

$$= \frac{1}{3} \left[ \frac{x^3}{3} \right]_1^2 + \frac{1}{8} \left[ \frac{x^3}{3} \right]_1^2$$

$$= \frac{1}{6} (4-1) + \frac{1}{24} (8-1)$$

$$= \frac{1}{6} (3) + \frac{1}{24} (7)$$

$$= \frac{12+7}{24} = \frac{19}{24}$$



$$(ii) P[X < 1/2]$$

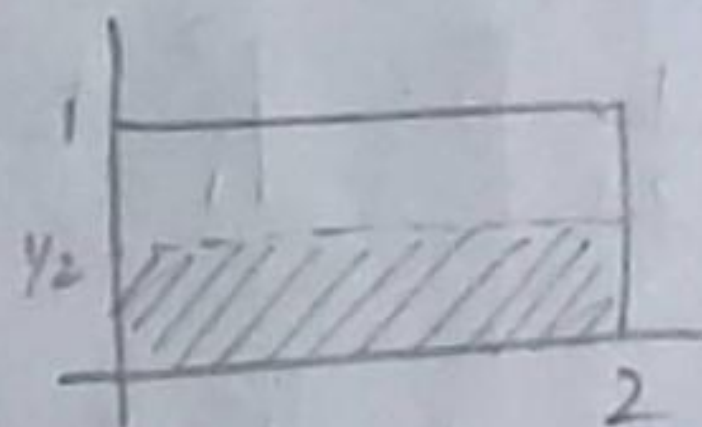
$$= \int_0^{1/2} f_Y(y) dy$$

$$= \int_0^{1/2} \left( 2y^2 + \frac{1}{3} \right) dy$$

$$= \left[ \frac{2y^3}{3} + \frac{y}{3} \right]_0^{1/2}$$

$$= \frac{1}{12} + \frac{1}{6} - 0$$

$$= \frac{3}{12} = \frac{1}{4}$$





$$(ii) P[x > 1, Y < Y_2]$$

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

$$= \int_1^2 \int_0^{1/2} (xy^2 + \frac{x^2}{8}) dy dx$$

$$= \int_1^2 x \left[ \frac{y^3}{3} \right]_0^{1/2} + \left[ \frac{x^2 y}{8} \right]_0^{1/2} dx$$

$$= \int_1^2 \frac{x}{3} \left[ \frac{1}{8} \right] + \frac{x^2}{8} \left[ \frac{1}{2} \right] dx$$

$$= \int_1^2 \left[ \frac{x}{24} + \frac{x^2}{16} \right] dx$$

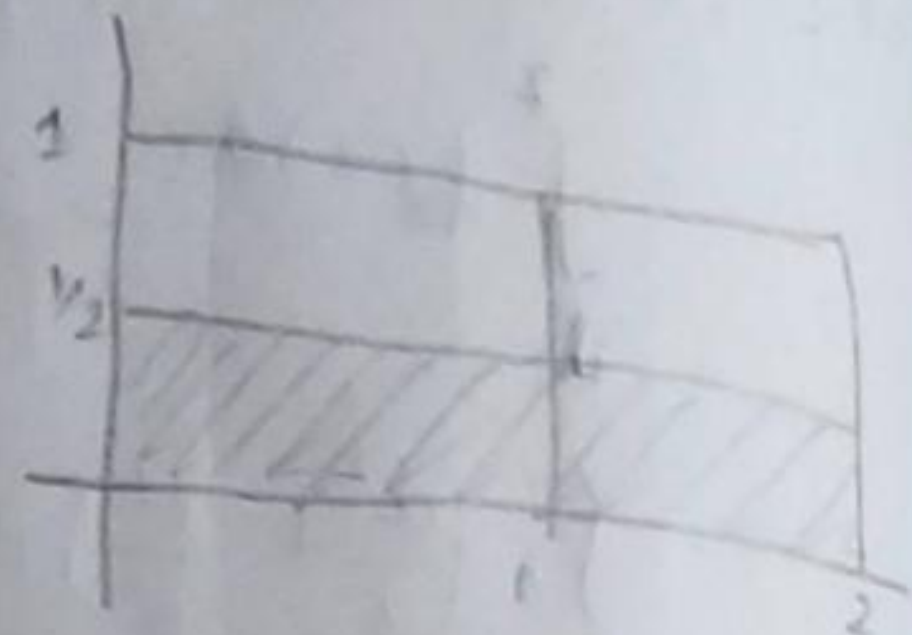
$$= \frac{1}{24} \left[ \frac{x^2}{2} \right]_1^2 + \frac{1}{16} \left[ \frac{x^3}{3} \right]_1^2$$

$$= \frac{1}{24 \cdot 2} [4] - [1] + \frac{1}{16 \times 3} [8 - 1]$$

$$= \frac{1}{24 \cdot 2} (3) + \frac{1}{16 \times 3} (7)$$

$$= \frac{3}{48} + \frac{7}{48}$$

$$= \frac{10}{48} = \frac{5}{24}$$





$$ii) P[X < Y]$$

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

$$= \int_0^1 \int_x^1 (xy^2 + \frac{x^2}{8}) dy dx$$

$$= \int_0^1 \left[ \frac{xy^3}{3} + \frac{x^2 y}{8} \right]_x^1 dx$$

$$= \int_0^1 \left[ \frac{x}{3} + \frac{x^2}{8} \right] - \left[ \frac{x^4}{3} + \frac{x^3}{8} \right] dx$$

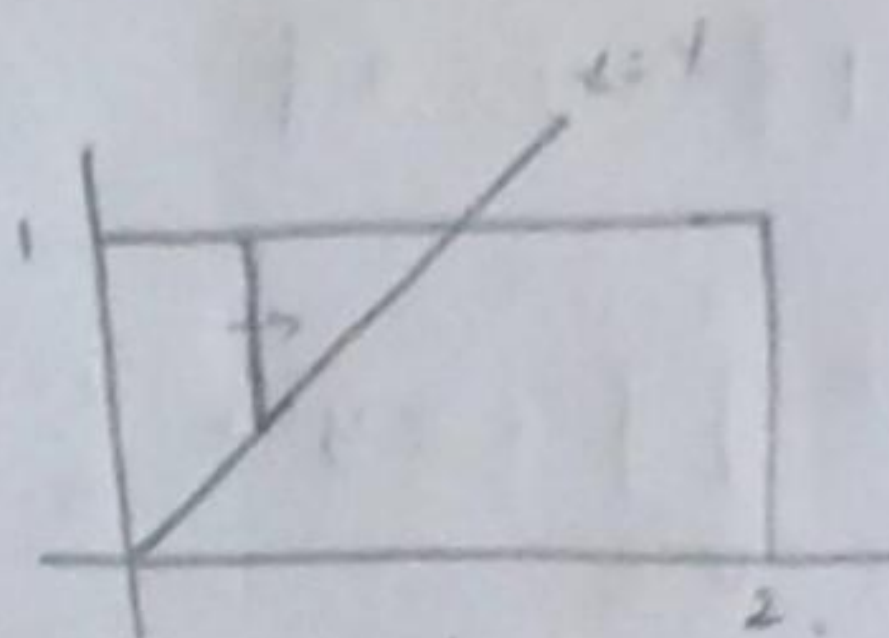
$$= \int_0^1 \left[ \frac{x}{3} + \frac{x^2}{8} - \frac{x^4}{3} - \frac{x^3}{8} \right] dx$$

$$= \int_0^1 \left[ \frac{8x + 3x^2 - 8x^4 - 3x^3}{24} \right] dx$$

$$= \int_0^1 \left[ \frac{x^2}{6} + \frac{x^3}{24} - \frac{x^5}{15} - \frac{x^4}{32} \right] dx$$

$$= \left[ \frac{1}{6} + \frac{1}{24} - \frac{1}{15} - \frac{1}{32} - 0 \right]$$

$$= \frac{80 + 20 - 32 - 15}{480} = \frac{53}{480}$$



$$ii) P[Y < 1/2 | X > 1]$$

$$= \frac{P[X > 1, Y < 1/2]}{P[X > 1]}$$

$$P[X > 1]$$

$$= \frac{\frac{5}{24}}{\frac{19}{24}} = \frac{5}{19}$$



$$vi) P[x+y \leq 1]$$

$$= \int_0^1 \int_0^{1-x} \left( xy^2 + \frac{x^2}{8} \right) dy dx$$

$$= \int_0^1 \left[ \frac{xy^3}{3} + \frac{x^2y}{8} \right]_0^{1-x} dx$$

$$= \int_0^1 \left[ \frac{x(1-x)^3}{3} + \frac{x^2(1-x)}{8} \right] dx$$

$$= \int_0^1 \left[ \frac{x(1-3x+3x^2-x^3)}{3} \right] dx + \int_0^1 \frac{x^2-x^3}{8} dx$$

$$= \frac{1}{3} \int_0^1 (x-3x^2+3x^3-x^4) dx + \frac{1}{8} \int_0^1 (x^2-x^3) dx$$

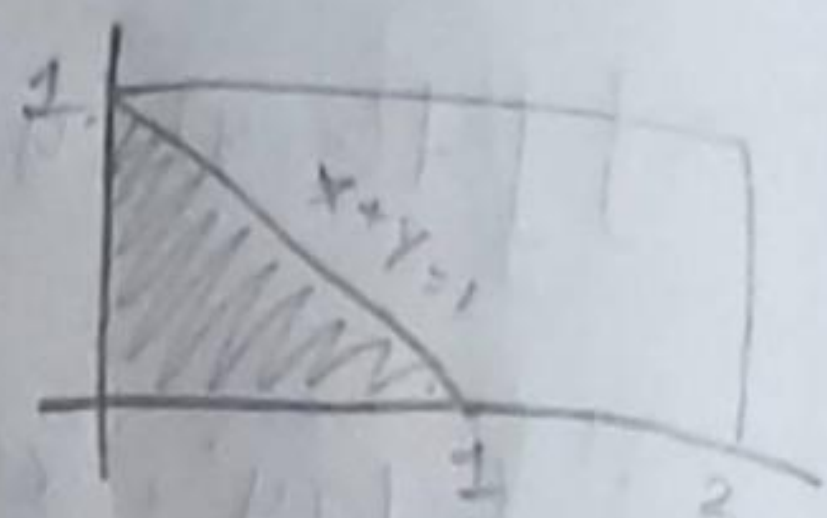
$$= \frac{1}{3} \left[ \frac{x^2}{2} - \frac{3x^3}{3} + \frac{3x^4}{4} - \frac{x^5}{5} \right]_0^1 + \frac{1}{8} \left[ \frac{x^3}{3} - \frac{x^4}{4} \right]_0^1$$

$$= \frac{1}{3} \left[ \frac{1}{2} - 1 + \frac{3}{4} - \frac{1}{5} \right] + \frac{1}{8} \left[ \frac{1}{3} - \frac{1}{4} \right]$$

$$= \frac{1}{3} \left[ \frac{10-20+15-4}{20} \right] + \frac{1}{8} \left[ \frac{1}{3} - \frac{1}{4} \right]$$

$$= \frac{1}{3} \left[ \frac{1}{20} \right] + \frac{1}{8} \left[ \frac{1}{12} \right] = \frac{1}{3} \cdot \frac{1}{20} + \frac{1}{96}$$

$$= \frac{96+60}{96 \times 60} = \frac{13}{480}$$





$$f(x, y) = \frac{6}{7} \left( x^2 + \frac{xy}{2} \right), \quad 0 < x < 1, \quad 0 < y < 2$$

Find (i)  $P(x > y)$

(ii)  $P(y > 1/2, x < 1/2)$

$$(i) \quad P[x > y] = \int_0^1 \int_0^x \frac{6}{7} \left( x^2 + \frac{xy}{2} \right) dy dx$$

$$= \int_0^1 \frac{6}{7} \left[ x^2 y + \frac{xy^2}{2} \right]_0^x dx$$

$$= \int_0^1 \frac{6}{7} \left[ x^3 + \frac{x^3}{4} \right] dx$$

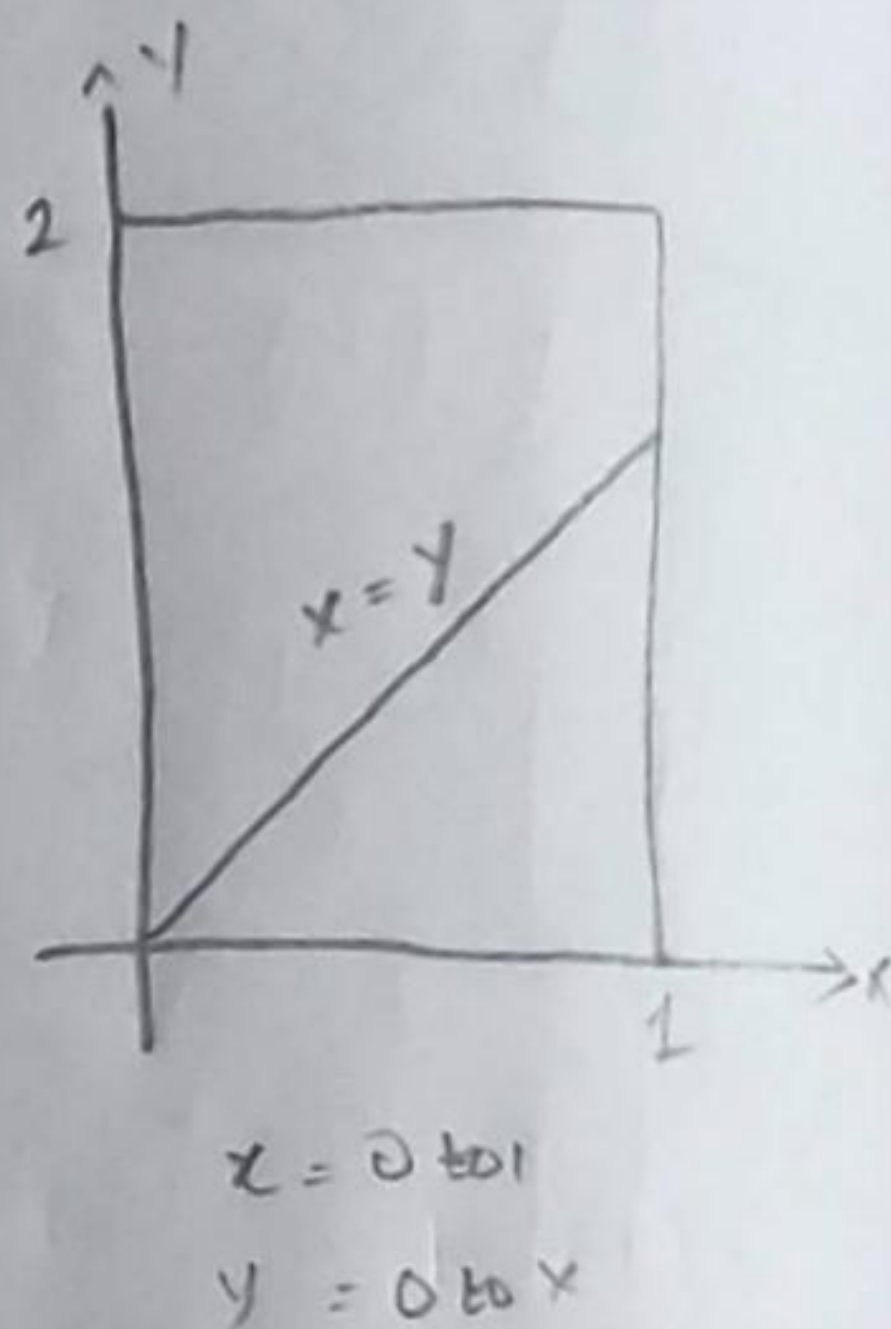
$$= \frac{6}{7} \int_0^1 \frac{5x^3}{4} dx$$

$$= \frac{6}{7} \times \frac{5}{4} \left[ \frac{x^4}{4} \right]_0^1$$

$$= \frac{30}{28} \left[ \frac{1}{4} \right]$$

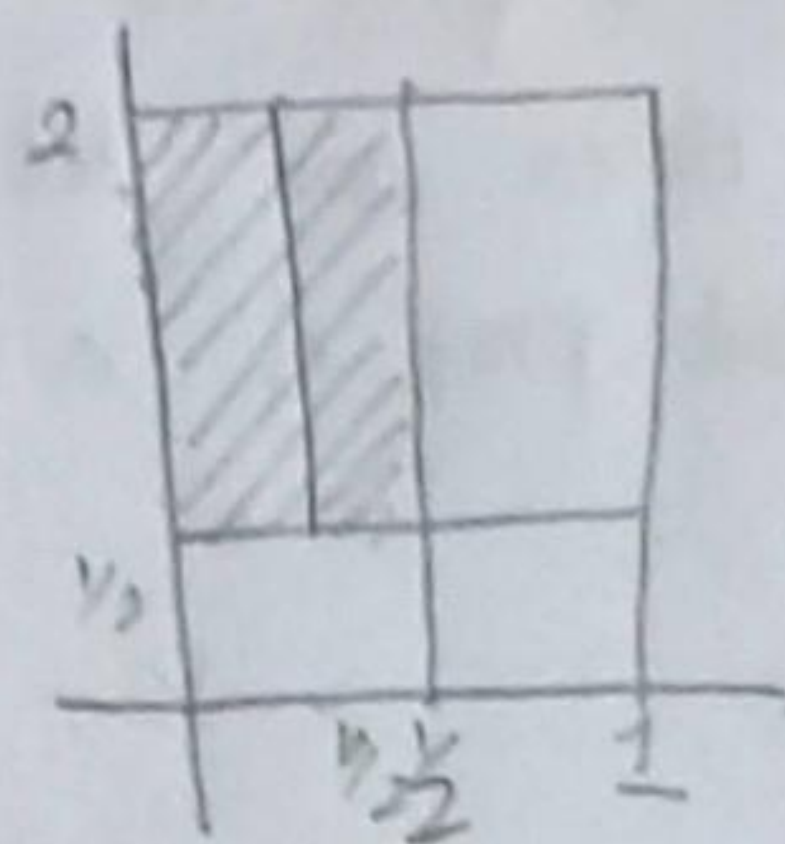
$$= \frac{15}{56}$$

$$P[x > y] = 15/56$$





$$c) P[Y > 1/2, X < 1/2]$$



$$= \int_0^{1/2} \int_{1/2}^2 \frac{6}{7} \left( x^2 + \frac{xy}{2} \right) dy dx$$

$$= \int_0^{1/2} \frac{6}{7} \left[ x^2 y + \frac{xy^2}{2} \right]_{1/2}^2 dx$$

$$= \int_0^{1/2} \frac{6}{7} \left[ 2x^2 + \frac{4x}{2} - \left( \frac{x^2}{2} + \frac{x}{4} \right) \right] dx$$

$$= \frac{6}{7} \int_0^{1/2} \left[ 2x^2 + x - \frac{x^2}{2} - \frac{x}{4} \right] dx$$

$$= \frac{6}{7} \int_0^{1/2} \left[ \frac{4x^2 - x^2}{2} + \frac{16x - x}{16} \right] dx$$

$$= \frac{6}{7} \int_0^{1/2} \left[ \frac{3x^2}{2} + \frac{15x}{16} \right] dx$$

$$= \frac{6}{7} \left[ \frac{3x^3}{3 \cdot 2} + \frac{15x^2}{16 \cdot 2} \right]_0^{1/2}$$

$$= \frac{6}{7} \left[ \frac{1}{16} + \frac{15}{32 \cdot 4} \right]$$

$$= \frac{6}{7} \left[ \frac{8 + 15}{128} \right] = \frac{6}{7} \left[ \frac{23}{128} \right] = \frac{69}{448}$$



Roll a fair die let the o/c be  $x$ , they toss a fair coin  $x$  times and let  $y$  denotes the no. of tails. Find the joint pmf of  $x \times y$  also find the marginal pmf of  $x$ .

$$X = \{1, 2, 3, 4, 5, 6\}$$

$$Y = \{0, 1, 2, 3, 4, 5, 6\}$$

$x \backslash y$	0	1	2	3	4	5	6	$P_x(x)$
1	$\frac{1}{6} \cdot \frac{1}{2}$	$\frac{1}{6} \cdot \frac{1}{2}$	0	0	0	0	0	$\frac{1}{6}$
2	$\frac{1}{6} \cdot \frac{1}{4}$	$\frac{1}{6} \cdot \frac{2}{4}$	$\frac{1}{6} \cdot \frac{1}{4}$	0	0	0	0	$\frac{1}{6}$
3	$\frac{1}{6} \cdot \frac{1}{8}$	$\frac{1}{6} \cdot \frac{3}{8}$	$\frac{1}{6} \cdot \frac{3}{8}$	$\frac{1}{6} \cdot \frac{1}{8}$	0	0	0	$\frac{1}{6}$
4	$\frac{1}{6} \cdot \frac{1}{16}$	$\frac{1}{6} \cdot \frac{4}{16}$	$\frac{1}{6} \cdot \frac{6}{16}$	$\frac{1}{6} \cdot \frac{4}{16}$	$\frac{1}{6} \cdot \frac{1}{16}$	0	0	$\frac{1}{6}$
5	$\frac{1}{6} \cdot \frac{1}{32}$	$\frac{1}{6} \cdot \frac{5}{32}$	$\frac{1}{6} \cdot \frac{10}{32}$	$\frac{1}{6} \cdot \frac{10}{32}$	$\frac{1}{6} \cdot \frac{5}{32}$	$\frac{1}{6} \cdot \frac{1}{32}$	0	$\frac{1}{6}$
6	$\frac{1}{6} \cdot \frac{1}{64}$	$\frac{1}{6} \cdot \frac{6}{64}$	$\frac{1}{6} \cdot \frac{15}{64}$	$\frac{1}{6} \cdot \frac{20}{64}$	$\frac{1}{6} \cdot \frac{15}{64}$	$\frac{1}{6} \cdot \frac{6}{64}$	$\frac{1}{6} \cdot \frac{1}{64}$	$\frac{1}{6}$
$P(x, y)$	$\frac{63}{384}$	$\frac{120}{384}$	$\frac{90}{384}$	$\frac{64}{384}$	$\frac{29}{384}$	$\frac{8}{384}$	$\frac{1}{384}$	

$$P(x, y) = \frac{1}{6} p^y q^{x-y}$$

Marginal pmf of  $x$

$x$	1	2	3	4	5	6
$P_x(x)$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$

Marginal pmf of  $y$

$y$	0	1	2	3	4	5	6
$P_y(y)$	$\frac{63}{384}$	$\frac{120}{384}$	$\frac{99}{384}$	$\frac{64}{384}$	$\frac{29}{384}$	$\frac{8}{384}$	$\frac{1}{384}$



Suppose three cards are drawn from a deck of 52 cards.  
 $x$  &  $y$  denote the no. of diamonds & spades respectively.

$$x = \{0, 1, 2, 3\}$$

$$y = \{0, 1, 2, 3\}$$

$y \backslash x$	0	1	2	3
0	$\frac{{}^{26}C_3}{{}^{52}C_3}$	$\frac{{}^{13}C_1 {}^{26}C_2}{{}^{52}C_3}$	$\frac{{}^{13}C_2 {}^{26}C_1}{{}^{52}C_3}$	$\frac{{}^{13}C_3}{{}^{52}C_3}$
1	$\frac{{}^{13}C_1 {}^{26}C_2}{{}^{52}C_3}$	$\frac{{}^{13}C_1 {}^{13}C_1 {}^{26}C_1}{{}^{52}C_3}$	$\frac{{}^{13}C_2 {}^{13}C_1}{{}^{52}C_3}$	0
2	$\frac{{}^{13}C_2 {}^{26}C_1}{{}^{52}C_3}$	$\frac{{}^{13}C_2 {}^{13}C_1}{{}^{52}C_3}$	0	0
3	$\frac{{}^{13}C_3}{{}^{52}C_3}$	0	0	0