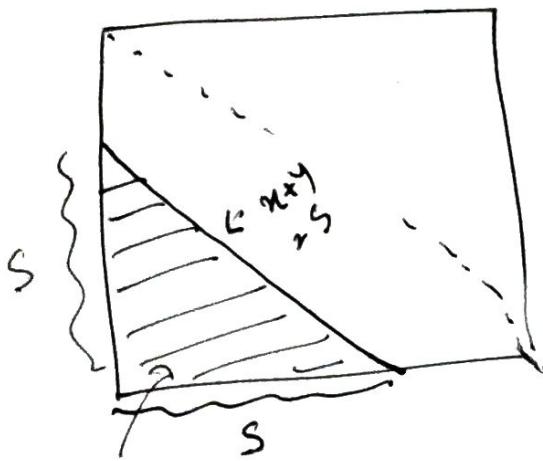


$$\textcircled{1} \quad U_1, U_2 \sim U[0,1]$$

$$S = U_1 + U_2$$

$$F(s) P(S \leq s) = P(U_1 + U_2 \leq s)$$

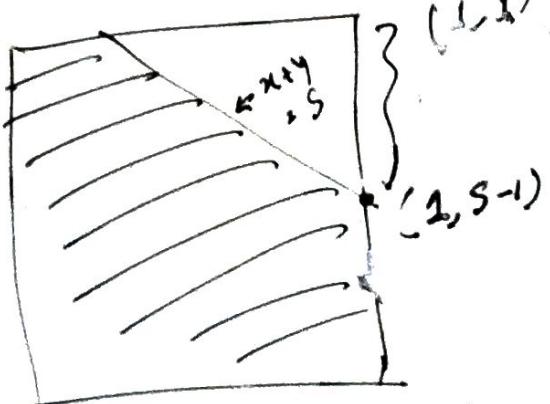
$$\text{Fix } s \in [0,1] \quad = \frac{1}{2} s^2$$



$$\Rightarrow \frac{1}{2}s^2 \quad F(s) =$$

$$P(S \leq s) = P(U_1 + U_2 \leq s)$$

$$(1-s)^2 = 1 - \frac{1}{2}(2-s)^2$$

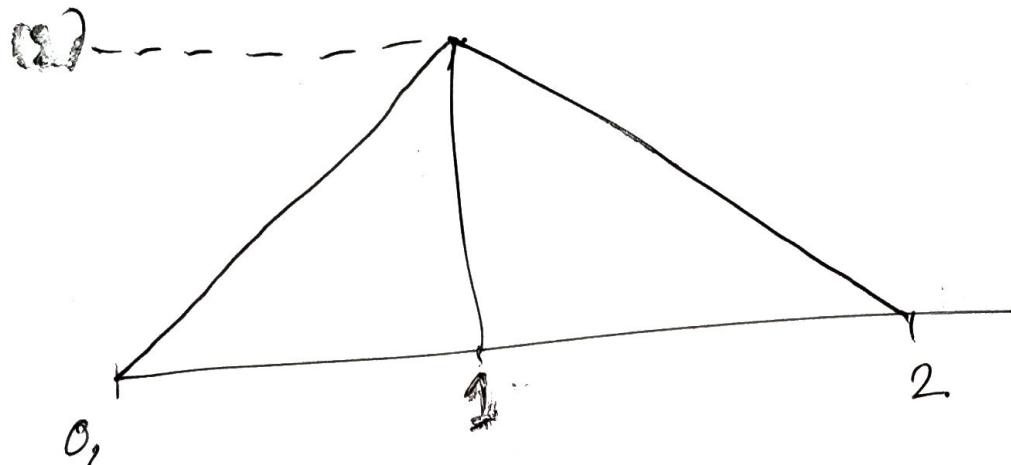


$$\text{for } s \in [1,2]$$

Hence, Density of  $S$  is .

$$f_S(s) = \frac{d}{ds} F(s) = \begin{cases} s & \text{if } s \in [0,1] \\ 2-s & \text{if } s \in [1,2] \end{cases}$$

Hence, the density is



②  $X$  has density

$$f(x) = \frac{1+\alpha x}{2} \quad -1 \leq x \leq 1 \\ -1 \leq \alpha \leq 1$$

$$E[X] = \int_{-1}^1 x \cdot \frac{1+\alpha x}{2} \cdot dx$$

$$= \int_{-1}^1 \frac{x}{2} + \alpha \cdot \int_{-1}^1 \frac{x^2}{2} \cdot dx$$

$$= 0 + \frac{\alpha}{2} \left[ \frac{x^3}{3} \right]_{-1}^1 = \frac{\alpha}{2} \times \frac{2}{3} \\ = \frac{\alpha}{3}$$

$$E[X^2] = \int_{-1}^1 x^2 \cdot \frac{1+\alpha x}{2} \cdot dx$$

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

$$= \frac{1}{2} \cdot \left[ \frac{x^3}{3} \right]_{-1}^1 + 0$$

$$= \frac{1}{2} \times \frac{2}{3} = \frac{1}{3}$$

$$\therefore \text{Var}(Y) = E[X^2] - (E[X])^2$$

$$= \frac{1}{3} - \frac{\alpha^2}{9}$$

③  $U_1, U_2, \dots, U_n \sim \text{Unif}[0, 1]$

$$U_{(n)} = \max(U_1, U_2, \dots, U_n)$$

Fix  $s \in [0, 1]$

$$P(U_{(n)} \leq s) = P(U_1 \leq s, U_2 \leq s, \dots, U_n \leq s)$$

$$= P(U_1 \leq s) P(U_2 \leq s) \dots P(U_n \leq s)$$

$$= s^n$$

∴ density of  $U_{(n)}$  is

$$f_{(n)}(s) = \frac{\partial}{\partial s} s^n = n \cdot s^{n-1} \quad s \in [0, 1]$$

$$\therefore E[U_{(n)}] = \int_0^1 s \cdot n \cdot s^{n-1} \, ds$$

$$= n \cdot \int_0^1 s^n \, ds$$

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

$$P(U_{(1)} < s)$$

$$= 1 - P(U_{(1)} \geq s)$$

$$= 1 - P(U_1 \geq s, U_2 \geq s, \dots, U_n \geq s)$$

$$= 1 - (1-s)^n$$

∴ Density of  $U_{(1)}$  is

$$f_{(1)}(s) = \frac{\partial}{\partial s} \cdot P(U_{(1)} < s)$$

$$= \frac{\partial}{\partial s} \cdot [1 - (1-s)^n]$$

$$= n (1-s)^{n-1}$$

$$\therefore E[U_{(1)}] = \int_0^1 n (1-s)^{n-1} \cdot s \cdot ds$$

$$= n \int_0^1 s (1-s)^{n-1} \cdot ds$$

Let

$$x = 1-s$$

$$\Rightarrow dx = -ds$$

$$= n \int_1^0 (1-x) \cdot x^{n-1} \cdot (-dx)$$

$$= n \int_0^1 (1-x) x^{n-1} \cdot dx$$

$$\geq E[U_{(n)} - U_{(1)}]$$

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

④ Let  $X$  be  $\text{Unif}[0,1]$  the side length of the square.

The Area of the Square is  $X^2$   
∴ Expected Area of Square

$$\begin{aligned} \text{is } E[X^2] &= \int_0^1 x^2 \cdot 1 \, dx \\ &= \left[ \frac{x^3}{3} \right]_0^1 \, dx \\ &= \frac{1}{3}. \end{aligned}$$

⑤  $X \sim \text{Poisson}$  distribution with mean  $\lambda$

$$\therefore P(X=x) = e^{-\lambda} \cdot \frac{\lambda^x}{x!} \quad x=0, 1, 2, \dots$$

Hence,

$$E\left[\frac{1}{X+1}\right] = \sum_{k=0}^{\infty} \frac{1}{k+1} \cdot e^{-\lambda} \cdot \frac{\lambda^k}{k!}$$

$$= \sum_{k=0}^{\infty} e^{-\lambda} \cdot \frac{\lambda^k}{(k+1)!}$$

$$= \frac{e^{-\lambda}}{\lambda} \cdot \sum_{k=0}^{\infty} \frac{\lambda^{k+1}}{(k+1)!}$$

$$= \frac{e^{-\lambda}}{\lambda} \left[ \sum_{k=1}^{\infty} \frac{\lambda^{k'}}{k'!} \right]$$

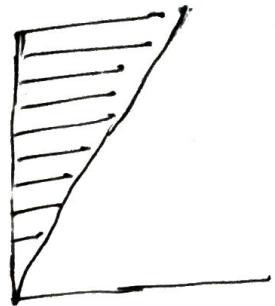
$$= \frac{e^{-\lambda}}{\lambda} \cdot \left[ \underbrace{\sum_{k'=0}^{\infty} \frac{\lambda^{k'}}{k'!}}_{= e^\lambda} - 1 \right]$$

$$= \frac{e^{-\lambda}}{\lambda} [e^\lambda - 1]$$

$$= \boxed{\frac{1 - e^{-\lambda}}{\lambda}}$$

⑥ X, Y have joint p.d.f

$$f(x,y) = e^{-y} \quad 0 < x < y$$



a)  $\text{Cov}(X,Y) = E[XY] - E[X]E[Y]$

$$E[XY] = \int_0^\infty \int_0^y xy e^{-y} dx dy$$

$$= \int_0^\infty y e^{-y} \int_0^y x dx dy$$

$$= \int_0^\infty y e^{-y} \left[ \frac{x^2}{2} \right]_0^y dy$$

$$= \int_0^\infty y e^{-y} \frac{y^2}{2} dy$$

$$= \frac{1}{2} \int_0^\infty y^3 e^{-y} dy$$

$$= \frac{1}{2} \Gamma(4) = \frac{1}{2} \times 3! = 3$$

$$\begin{aligned}
 E[X] &= \int_0^\infty \int_0^y x \cdot e^{-y} \cdot dx dy \\
 &= \int_0^\infty e^{-y} \int_0^y x \cdot dx \cdot dy \\
 &= \int_0^\infty e^{-y} \cdot \frac{y^2}{2} \cdot dy = \frac{1}{2} \Gamma(3) \\
 &= \frac{1}{2} \cdot 2! = 1
 \end{aligned}$$

$$\begin{aligned}
 E[Y] &= \int_0^\infty \int_0^y y \cdot e^{-y} dx dy \\
 &= \int_0^\infty y \cdot e^{-y} \int_0^y 1 \cdot dx \cdot dy \\
 &= \int_0^\infty y \cdot e^{-y} \cdot y \cdot dy \\
 &= \int_0^\infty y^2 \cdot e^{-y} dy = \Gamma(3) = 2! = 2
 \end{aligned}$$

$$\begin{aligned}
 \therefore \text{Cov}(X, Y) &= E[XY] - E[X] \cdot E[Y] \\
 &= 3 - 1 \cdot 2 \\
 &= 1
 \end{aligned}$$

$$\begin{aligned}
 \text{Var}(x) &= E[x^2] - (E[x])^2 \\
 &= E[x^2] - 1 \\
 &= \int_0^\infty \int_0^y x^2 \cdot e^{-y} dx dy - 1 \\
 &= \int_0^\infty e^{-y} \frac{y^3}{3} dy - 1 \\
 &= \frac{1}{3} \Gamma(4) = \frac{3!}{3} - 1 = 1
 \end{aligned}$$

$$\begin{aligned}
 \text{Var}(y) &= E[y^2] - (E[y])^2 \\
 &= \int_0^\infty \int_0^y y^2 \cdot e^{-y} dx dy - 4 \\
 &= \int_0^\infty y^2 \cdot e^{-y} y dy - 4 \\
 &= \int_0^\infty y^3 \cdot e^{-y} dy - 4 \\
 &= \Gamma(4) - 4 \\
 &= 6 - 4 = 2
 \end{aligned}$$

$$\therefore \text{Cor}(x, y) = \frac{\text{Cov}(x, y)}{\sqrt{\text{Var}(x) \cdot \text{Var}(y)}}$$

$$= \frac{1}{\sqrt{2}}$$

(b) The conditional density of  $X$  given  $Y=y$  is  $\frac{f(x, y)}{f_Y(y)}$  where  $f_Y(y)$  is marginal density of  $y$ .

Marginal Density of  $Y$  is given by

$$f_Y(y) = \int_0^y e^{-x} \cdot dx = y \cdot e^{-y} \quad y \in [0, \infty)$$

Hence conditional density of  $X$  given  $Y=y$  is.

$$f_{X|Y}(x) = \frac{e^{-y}}{y \cdot e^{-y}} \quad \text{for } 0 < x < y, \quad y \in \mathbb{R}^+ \text{ fixed}$$

$$= \frac{1}{y} \quad \text{for } 0 < x < y$$

i.e.  $f_{X|Y}(x) \sim \text{Unif}[0, y]$

$$\text{Ex. } E[X|Y=y] = \frac{y}{2}$$

Marginal of  $X$  is given by

$$\frac{f(x,y)}{f_x(x)} \quad \text{where } f_x(x) \text{ is marginal}$$

density of  $X$ .

$$f_x(x) = \int_x^{\infty} e^{-y} \cdot dy$$

$$= [-e^{-y}]_x^{\infty} = e^{-x} \quad x \in [0, \infty)$$

conditional density of

$Y$  given  $X=x$  is given by

$$\frac{f_{(x,y)}}{f_x(x)} = e^{-y+x} \quad \text{for } y > x$$

$x \in \mathbb{R}^+ \text{ fixed}$

$$\therefore E[Y | X=x] = \int_x^{\infty} y \cdot e^{-y+x} \cdot dy$$

$$Z = y - x$$

$$dz = dy$$

$$= \int_0^{\infty} (z+x) e^{-z} \cdot dz$$

$$= \int_0^{\infty} z \cdot e^{-z} dz + x \cdot \int_0^{\infty} e^{-z} dz$$

⑥ note

$$E[X|Y] = \frac{Y}{2}$$

$$E[Y|X] = 1+X$$

Note that

Marginal density of  $Y$  is

$$f_Y(y) = y e^{-y} \cdot \boxed{y > 0}$$

Hence, Marginal density of  $U = \frac{Y}{2}$  is

given by

Due to Jacobian

$$f_U(u) = 2u \cdot e^{-2u} \times 2^u \quad u > 0$$

$$= 4u e^{-2u} \quad u > 0$$

---

Marginal density of  $X$  is  $f_X(x) = e^{-x}, x > 0$

Marginal density of  $V = 1+X$  is.

$$f_V(v) = e^{-(v-1)} \quad \boxed{v > 1}$$

$$\textcircled{7} \quad \textcircled{a} \quad \text{Cov}(x_1 + 2x_2, x_2 + x_3 + 1)$$

$$\begin{aligned}
&= \text{Cov}(x_1, x_2) + \text{Cov}(x_1, x_3) + \text{Cov}(x_1, 1) \\
&\quad + 2 \text{Cov}(x_2, x_2) + 2 \text{Cov}(x_2, x_3) \\
&\quad + 2 \text{Cov}(x_2, 1) \\
&= 0 + 0 + 0 + 2 + 0 + 0 = 2
\end{aligned}$$

$$\text{Var}(x_1 + 2x_2)$$

$$\begin{aligned}
&= \text{Var}(x_1) + 4 \cdot \text{Var}(x_2) + 2 \text{Cov}(x_1, 2x_2) \\
&= 1 + 4 + 0 = 5
\end{aligned}$$

$$\text{Var}(x_2 + x_3 + 1)$$

$$\begin{aligned}
&= \text{Var}(x_2 + x_3) = \text{Var}(x_2) + \text{Var}(x_3) \\
&\quad + 2 \text{Cov}(x_2, x_3) \\
&= 1 + 1 + 0 = 2
\end{aligned}$$

$$\therefore \text{Cor}(x_1 + 2x_2, x_2 + x_3 + 1)$$

$$= \frac{2}{\sqrt{2 \times 5}} = \frac{2}{\sqrt{10}}$$

$$\begin{aligned}
 & \text{7) } \textcircled{b} \quad \text{6v}(x_1+x_2+3, 5x_3+x_4) \\
 & = \text{cov}(x_1+x_2, 5x_3+x_4) = \text{cov}(x_1, 5x_3) + \text{cov}(x_1, x_4) \\
 & \quad + \text{cov}(x_2, 5x_3) + \text{cov}(x_2, x_4) \quad = 0 \\
 & \therefore \text{cov}(x_1+x_2+3, 5x_3+x_4) \\
 & = 0
 \end{aligned}$$

(8) The event that we need " $n$ " tries to get 'both' Heads and tails can be broken into two scenario's.

$\underbrace{H H H \dots H}_ {(n-1) \text{ times}} T$  and  $\overbrace{T T T \dots T}^{(n-1) \text{ times}} H$

where  $n \geq 2$ .

$X$  be the "count" that number of tries needed.

$$\begin{aligned}
 \therefore P(X=n) &= P\left(\underbrace{H H \dots H}_{n-1} \cdot T\right) + P\left(\underbrace{T T \dots T}_{n-1} H\right) \\
 &= P^{n-1} \cdot (1-P) + (1-P)^{n-1} \cdot P \\
 &= P(1-P) \left[ P^{n-2} + (1-P)^{n-2} \right] \boxed{n \geq 2}
 \end{aligned}$$

$$E[X] =$$

$$= \sum_{n=2}^{\infty} n P(1-P) \left[ P^{n-2} + (1-P)^{n-2} \right]$$

$$= P(1-P) \cdot \sum_{n=2}^{\infty} \left[ n P^{n-2} + n \cdot (1-P)^{n-2} \right]$$

$$= P(1-P) \left[ \sum_{k=0}^{\infty} (k+2) \cdot P^k + (k+2) \cdot (1-P)^k \right]$$

$$= P(1-P) \left[ 2 \sum_{k=0}^{\infty} P^k + (1-P)^k + \sum_{k=0}^{\infty} k \cdot P^k + \sum_{k=0}^{\infty} k \cdot (1-P)^k \right]$$

$$= P(1-P) \left[ \frac{2}{1-P} + \frac{2}{P} + \frac{1}{(1-P)^2} - \frac{1}{1-P} + \frac{1}{P^2} - \frac{1}{P} \right]$$

$$= P(1-P) \left[ \frac{1}{P^2} + \frac{1}{(1-P)^2} + \frac{1}{1-P} + \frac{1}{P} \right]$$

$$= \frac{1-P}{P} + \frac{P}{1-P} + P + 1-P$$

$$= 1 + \frac{1-P}{P} + \frac{P}{1-P}. \quad (\text{Ans})$$

$$8) \textcircled{b} \quad P(\text{Last flip is heads}) \\ = P(\text{first flip is Tails}) \\ = (1-P)$$

⑨ Let  $X_i = 1$  if  $i$ th ball is Red  
 $= 0$  if  $"$  " " not Red

$Y_i = 1$  if  $i$ th ball is Blue  
 $= 0$  " " " not "

$$\therefore X = \sum_{i=1}^{12} X_i \quad Y = \sum_{i=1}^{12} Y_i$$

$$\begin{aligned} \text{Cov}(X, Y) &= \text{Cov}\left(\sum_{i=1}^{12} X_i, \sum_{i=1}^{12} Y_i\right) \\ &= \sum_{i=1}^{12} \text{Cov}(X_i, Y_i) \\ &\quad + 2 \sum_{i < j} \text{Cov}(X_i, X_j) \end{aligned}$$

$$\text{Cov}(X_i, Y_i) = E[X_i Y_i] - E[X_i] E[Y_i]$$

$$= 0 - \frac{10}{30} \times \frac{8}{30}$$

See that

$$E[X_i Y_i] = \frac{1}{30} \times \frac{4}{30} = \frac{4}{900}$$

$$\text{④} \operatorname{Cov}(X_i, Y_j)$$

$$= E[X_i Y_j] - E[X_i]E[Y_j]$$

$$= \frac{10}{30} \times \frac{8}{29} - \frac{10}{30} \times \frac{8}{30}$$

$$= \frac{1}{3} \left[ \frac{8}{29} - \frac{8}{30} \right]$$

$$= \frac{8}{3} \left[ \frac{30 - 1}{29 \times 30} \right] = \underbrace{\frac{8}{3} \times \frac{1}{29 \times 30}}_{\text{Ans}}$$

$$\therefore \operatorname{Cov}(X, Y)$$

$$= -12 \times \frac{4}{15} + 2 \times \binom{12}{2} \times \frac{8}{3} \times \frac{1}{29 \times 30}$$

$$= -12 \times \frac{4}{15} + \frac{12 \times 11 \times 8}{3 \times 29 \times 30}.$$

(Ans).

⑩ Conditional joint pdf of  $Y_1, \dots, Y_n$   
given  $X = x$  is.

$$h(y_1, \dots, y_n | x) = \begin{cases} \frac{1}{x^n} & 0 < y_i < x \\ & \forall i \\ 0 & \text{o.w.} \end{cases}$$

Joint p.d.f of  $X$  and  $Y_1, \dots, Y_n$  is.

$$f(x) \cdot h(y_1, \dots, y_n | x) = \begin{cases} \frac{1}{n!} \exp(-x) & 0 < y_i < x \\ & \forall i \\ 0 & \text{o.w.} \end{cases}$$

ⓐ For  $y_i > 0 \forall i$ , the marginal P.d.f  
of  $Y_1, Y_2, \dots, Y_n$  is.

$$\int_{\max(y_1, \dots, y_n)}^{\infty} \frac{1}{n!} \exp(-x) dx = \frac{1}{n!} \exp\left(-\max_{(y_1, \dots, y_n)}\right)$$

ⓑ Conditional P.d.f of  $X$  given  $Y_i$  is.

$$\frac{\frac{1}{n!} \cdot \exp(-x)}{\frac{1}{n!} \exp(-\max(y_1, \dots, y_n))} = \exp\left(-x - \max_{(y_1, \dots, y_n)}\right) \text{ where } x > \max(y_1, \dots, y_n)$$

(11) We see that  $x \in [0, 2]$   
 This imply that  $y \in [0, 1]$

$$\begin{aligned}
 & P(Y \leq y) \\
 &= P(x^2 - 2x \geq -y) \\
 &= P((x-1)^2 \geq 1-y) \\
 &= P(x-1 \geq \sqrt{1-y}) + P(x-1 \leq -\sqrt{1-y}) \\
 &= P(x \geq 1+\sqrt{1-y}) + P(x \leq 1-\sqrt{1-y}) \\
 &= \int_{1+\sqrt{1-y}}^2 \frac{1}{2} x dx + \int_0^{1-\sqrt{1-y}} \frac{1}{2} x dx \\
 &= 1 - \sqrt{1-y}
 \end{aligned}$$

$\therefore$  Density of  $Y$  is  $\frac{\partial}{\partial y} \cdot P(Y \leq y)$   
 $\therefore = \frac{\partial}{\partial y} \cdot (1 - \sqrt{1-y}) = \frac{1}{2(1-y)^{1/2}}$   
 $y \in (0, 1)$ ,

(12) Let  $Z = X_1$ .

Hence  $(X_1, X_2) \longleftrightarrow (Y, Z)$  is one to one transform.

$$X_1 = Z \\ X_2 = Y/Z$$

Therefore

$$J = \det \begin{bmatrix} \frac{\partial X_1}{\partial Y} & \frac{\partial X_1}{\partial Z} \\ \frac{\partial X_2}{\partial Y} & \frac{\partial X_2}{\partial Z} \end{bmatrix}$$

$$= \det \begin{bmatrix} 0 & 1 \\ 1/2 & -\frac{y}{z^2} \end{bmatrix} = -\frac{1}{2}$$

For  $0 < y < z < 1$ , the joint pdf of  $Y, Z$  is.

$$g(y, z) = f(z, \frac{y}{z}) \cdot |J| = \left(z + \frac{y}{z}\right) \cdot \frac{1}{2}.$$

$$= 1 + \frac{y}{z^2} \quad \boxed{0 < y < z < 1}$$

Hence, marginal of  $Y$  is.

$$g_y(y) = \int_y^1 1 + \frac{y}{z^2} \cdot dz$$

$$= 2(1-y) \quad y \in [0, 1].$$

(13) mgf of  $Z$  is.

$$\begin{aligned}\Psi_1(t) &= E[\exp(tz)] \\ &= E[\exp(t(2x - 3y + 4))] \\ &= e^{4t} \cdot E[e^{2tx - 3ty}] \\ &= e^{4t} \cdot E[e^{2tx}] \cdot E[e^{-3ty}] \\ &\quad (\text{As. } X, Y \text{ indep}) \\ &= e^{4t} \Psi(2t) \Psi(-3t) \\ &= e^{4t} \cdot e^{4t^2 + 6t} \cdot e^{9t^2 - 9t} \\ &= e^{13t^2 + t} \quad (\text{Ans})\end{aligned}$$

Ans

(14)  $X$  can take finite no of values  $x_1, x_2, \dots, x_k$  with prob  $P_1, \dots, P_k$   
 then mgf is

$$\psi(t) = P_1 \exp(tx_1) + \dots + P_k \exp(tx_k)$$

Hence by the expression of  $\psi(t)$  we get

$X$  take values 1, 4, 8 with probabilities  
 $\frac{1}{5}, \frac{2}{5}, \frac{2}{5}$  respectively.

(15) P-d.f of  $X$  is  $h(x) = \frac{1}{2} f(x) + \frac{1}{2} g(x)$

$$E[X] = \int_{-\infty}^{\infty} x \left[ \frac{1}{2} f(x) + \frac{1}{2} g(x) \right] dx$$

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

$$= \frac{1}{2} \mu_f + \frac{1}{2} \mu_g \quad \left[ \text{where } \mu_f \text{ & } \mu_g \text{ are expectations of } f \text{ & } g \text{ respectively} \right]$$

Now,  $\int_{-\infty}^1 h(x) dx = \frac{1}{2} \int_{-\infty}^1 f(x) dx = \frac{1}{2}$  &

$$\int_2^\infty h(x) dx = \frac{1}{2} \int_2^\infty g(x) dx = \frac{1}{2}$$

$\therefore$  Any Value between  $[1, 2]$  is a median

⑯ Simple Calculation gives

$$E[X] = \int_0^1 \int_0^2 x - \frac{1}{3}(x+y) dy dx = \frac{5}{9}$$

$$E[Y] = \int_0^1 \int_0^2 y - \frac{1}{3}(x+y) dy dx = \frac{11}{9}$$

$$E[X^2] = \int_0^1 \int_0^2 x^2 - \frac{1}{3}(x+y) dy dx = \frac{7}{18}$$

$$E[Y^2] = \int_0^1 \int_0^2 y^2 - \frac{1}{3}(x+y) dy dx = \frac{16}{9}$$

$$E[XY] = \int_0^1 \int_0^2 xy - \frac{1}{3}(x+y) dy dx = \frac{2}{3}.$$

$$\therefore \text{Var}(X) = E[X^2] - (E[X])^2 = \frac{13}{162}$$

$$\text{Var}(Y) = E[Y^2] - (E[Y])^2 = \frac{23}{81}$$

$$\text{Cov}(X, Y) = E[XY] - E(X)E(Y) = -\frac{1}{81}$$

$$\text{Var}(2X - 3Y + 8)$$

$$= 4\text{Var}(X) + 3^2\text{Var}(Y) - 2 \cdot 2 \cdot 3 \text{Cov}(X, Y)$$

$$= 245/81$$