

# Assignment 4

## Probability and Statistics

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Q.1

$$f(x) = 4x e^{-2x} = \frac{(2)^2 x^{(2-1)} e^{-2x}}{\Gamma(2)} ; x > 0$$

Comparing with  $x \sim \text{Gamma}(\alpha, \lambda)$ ;

i.e.  $f(x) = \begin{cases} \frac{\lambda^\alpha x^{\alpha-1} e^{-\lambda x}}{\Gamma(\alpha)} & ; x > 0 \\ 0 & ; \text{otherwise} \end{cases} \quad \alpha, \lambda > 0$

W.C. g.t.,  $\alpha = \lambda = 2$  for s.t.  $f(x) = f_{\text{gt.}}(x)$

(a) Finding  $E(X)$ ;  $* \Gamma(\alpha+1) = \alpha \Gamma(\alpha)$

$$E(X) = \int_{-\infty}^{\infty} x f(x) dx = \int_0^{\infty} x f(x) dx$$

$$E(X) = \int_0^{\infty} x \cdot 4x^2 e^{-2x} dx = 4 \cdot \frac{\Gamma(3)}{(2)^3} = 1.$$

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

&  $\Gamma(3) = 2 \Gamma(2) = 2 \times (1!) = 2.$

Similarly;

Finding  $E(X^2)$ ;

$$E[X^2] = \int_0^\infty x^2 f(x) dx = \int_0^\infty 4x^3 e^{-2x} dx$$

$$E[X^2] = \frac{4 \Gamma(4)}{(2)^4} = \frac{4}{16} \times (3!)$$

$$E[X^2] = 3/2 \quad \text{Ans}$$



Similarly finding  $E[X^3]$ ;

$$E[X^3] = \int_0^\infty x^3 f(x) dx = \int_0^\infty 4x^4 e^{-2x} dx$$

$$E[X^3] = \frac{4 \Gamma(5)}{(2)^5} = \frac{4}{32} \times (4!)$$

$$\ast \Gamma(n) = (n-1)!$$

$$E[X^3] = 3 \quad \text{Ans}$$

~~\*REMEMBER~~



(b) Driving  $E(X^n)$

$$E[X^n] = \int_0^\infty x^n f(x) dx = \int_0^\infty 4x^{n+1} e^{-2x} dx$$

Using Gamma function proportion as in  
part a;  $E[X^n] = \frac{4 \Gamma(n+2)}{(2)^{n+1}}$

Thus,

$$E[X^n] = (2)^{-n} (n+1)!$$

Q. 2

$$(a) \Gamma(\frac{7}{2}) = (\frac{5}{2}) \Gamma(\frac{5}{2})$$

\* by using  $\Gamma(k) = (k-1) \Gamma(k-1)$

$$\begin{aligned} \Gamma(\frac{7}{2}) &= (\frac{5}{2})(\frac{3}{2}) \Gamma(\frac{3}{2}) \\ &= (\frac{5}{2})(\frac{3}{2})(Y_2) \Gamma(Y_2) \end{aligned}$$

$$\text{As } \Gamma(Y_2) = \sqrt{\pi}$$

$$\begin{aligned} \text{So, } \Gamma(\frac{7}{2}) &= (\frac{5}{2})(\frac{3}{2})(Y_1) \sqrt{\pi} \\ &= \frac{15\sqrt{\pi}}{8} \text{ Ans} \end{aligned}$$

$$(b) I = \int_0^\infty x^6 e^{-5x} dx$$

As we know for  $X$  ~ Gamma( $\alpha, \lambda$ ),  $x > 0$

$$f_X(x) = \frac{1}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x}$$

Definition of Gamma function:

$$\Gamma(\alpha) = \int_0^\infty x^{\alpha-1} e^{-x} dx$$

$$+ \frac{\Gamma(\alpha)}{\lambda^\alpha} = \int_0^\infty x^{\alpha-1} e^{-\lambda x} dx ; \lambda > 0$$

comparing we get  $\alpha = 7$  &  $\lambda = 5$

$$\text{So, } I = \frac{\Gamma(\alpha)}{\Gamma(\alpha - 1)} ; \alpha = 7, \lambda = 5$$

$$I = \frac{\Gamma(7)}{5^7}$$

$$I = \frac{(6)!}{(5)^7} \quad (\text{using } \Gamma(\alpha) = (\alpha-1)!)$$

$$I = \frac{144}{15,625} \approx 0.009216 \quad \text{Ans}$$

$$(8 \cdot 3) \\ \underline{\underline{=}}$$

$$f_{XY}(x,y) = \begin{cases} C(x^2 + y^2) & \text{if } x \in \{1, 2, 4\} \\ & \text{& } y \in \{1, 3\} \\ 0 & \text{otherwise} \end{cases}$$

(a) For RV  $X, Y$  sample space  $S$  is  $\{(1,1), (1,3), (2,1), (2,3), (4,1), (4,3)\}$

$$\text{So, } \sum_{(x,y) \in S} f_{XY}(x,y) = 1$$

$$\Rightarrow C [ C(1^2 + 1^2) + (1^2 + 3^2) + (2^2 + 1^2) + (2^2 + 3^2) + (4^2 + 1^2) + (4^2 + 3^2) ] = 1$$

$$\Rightarrow \frac{1}{c} = 2 + 10 + 5 + 13 + 17 + 25$$

$$\Rightarrow \frac{1}{c} = 72$$

$$\Rightarrow c = 1/72 \text{ Ans}$$

$\Leftarrow$

(b) For sample space S

$P(Y < X)$  means equivalent to  
finding  $\sum P(x,y)$  for 3 ordered  
pairs  $(x,y)$  i.e.  $(2,1), (4,1), (4,3)$

$$\text{So, } P(Y < X) = P(x=2, y=1) + \\ P(x=4, y=1) + \\ P(x=4, y=3)$$

$$= f_{xy}(2,1) + f_{xy}(4,1) + f_{xy}(4,3)$$

$$= c [ (4+1) + (16+1) + (16+9) ]$$

$$= \frac{1}{72} [ 5 + 17 + 25 ]$$

$$P(Y < X) = \frac{47}{72} \text{ Ans}$$

$\Leftarrow$

(c)  $Y > X$  is satisfied by only 2  
( $x,y$ ) ordered pairs in S &  
they are  $(1,3), (2,3)$

$$\begin{aligned}
 P(Y > X) &= f_{XY}(1,3) + f_{XY}(2,3) \\
 &= C [(1+9) + (4+9)] \\
 &= \frac{1}{72} (23)
 \end{aligned}$$

$$P(Y > X) = \frac{23}{72} \quad \text{Ans}$$

(d) Only 1  $(x,y)$  ordered pair  
in  $S$  satisfy  $x=y$  i.e.  $(1,1)$

$$\therefore P(X=Y) = f_{XY}(1,1) = C(1+1)$$

$$P(X=Y) = 2C = \frac{2}{36} \quad \text{Ans}$$

$$(e) P(Y=3) = f_{XY}(1,3) + f_{XY}(2,3) + f_{XY}(4,3)$$

$$\begin{aligned}
 P(Y=3) &= C [(1+9) + (4+9) + (16+9)] \\
 &\div \frac{1}{72} \times 48
 \end{aligned}$$

$$P(Y=3) = \frac{48}{72} = \frac{4}{6} = \frac{2}{3} \quad \text{Ans}$$

(f) Finding marginal PMF  $P_X(x)$ .

$$\text{Since, } P_X(x) = \sum_{y=-\infty}^{\infty} P_{XY}(x, y)$$

$$\text{So, } P_X(1) = f_{XY}(1,1) + f_{XY}(1,3) = \frac{12}{72}$$

$$P_X(2) = f_{XY}(2,1) + f_{XY}(2,3) = \frac{18}{72}$$

$$P_X(4) = f_{XY}(4,1) + f_{XY}(4,3) = \frac{42}{72}$$

$$\text{So, } P_X(x) = \begin{cases} 12/72 ; & x=1 \\ 18/72 ; & x=2 \\ 42/72 ; & x=4 \\ 0 & \text{otherwise} \end{cases}$$

Similarly,

$$P_Y(1) = f_{XY}(1,1) + f_{XY}(2,1) + f_{XY}(4,1) = 24/72$$

$$P_Y(3) = f_{XY}(1,3) + f_{XY}(2,3) + f_{XY}(4,3) = 48/72$$

$$\text{So, } P_Y(y) = \begin{cases} 24/72 ; & y=1 \\ 48/72 ; & y=3 \\ 0 & \text{otherwise} \end{cases}$$

$$(g) E[X] = \sum_{x=-\infty}^{\infty} x \cdot P_X(x) ; \text{ for any R/T X.}$$

Using Marginal PMF for  $X \neq Y$  as found in part (f) we can write

$$E[X] = 1 \cdot \frac{12}{72} + 2 \cdot \frac{18}{72} + 4 \cdot \frac{42}{72} = 3 \text{ Ans}$$

$$\leftarrow E[Y] = 1 \cdot \frac{24}{72} + 3 \cdot \frac{48}{72} = \frac{7}{3} \text{ Ans}$$

Let  $Z = XY$  then range of  $Z$   
will be  $R_Z = \{1, 2, 3, 4, 6, 12\}$

$$P_Z(1) = P_{XY}(1,1) = \frac{2}{72}$$

$$P_Z(2) = P_{XY}(2,1) = \frac{5}{72}$$

$$P_Z(3) = P_{XY}(1,3) = \frac{10}{72}$$

$$P_Z(4) = P_{XY}(4,1) = \frac{17}{72}$$

$$P_Z(6) = \cancel{P_{XY}(2,2)} P_{XY}(2,3) = \frac{13}{72}$$

$$P_Z(12) = P_{XY}(4,3) = \frac{25}{72}$$

$$\therefore E[XY] = E[Z] = \sum_{z \in R_Z} z P_Z(z)$$

$$E[XY] = 1 \cdot \frac{2}{72} + 2 \cdot \frac{5}{72} + 3 \cdot \frac{10}{72} +$$

$$4 \cdot \frac{17}{72} + 6 \cdot \frac{13}{72} + 12 \cdot \frac{25}{72}$$

$$= 61/9 \text{ Ans}$$

$$(f) E[X^2] = \sum_{x=-\infty}^{\infty} (x)^2 \cdot p(x); \text{ for any RV } X.$$

Using Marginal PMF for  $X$  &  $Y$  as found  
in point (f) we can write

$$E[X^2] = (1)^2 \cdot \frac{12}{72} + (2)^2 \cdot \frac{18}{72} + (4)^2 \cdot \frac{42}{72} = \frac{21}{2}$$

—  $\star$

$$E[Y^2] = (1)^2 \cdot \frac{24}{72} + (3)^2 \cdot \frac{48}{72} = \frac{57}{9}$$

$$\text{So, } \text{Var}(X) = E[X^2] - (E[X])^2 = \frac{21}{2} - (3)^2$$

$$\text{Var}(X) = \frac{3}{2} \text{ Ans} \quad (\text{found in (g)})$$

$$\& \text{Var}(Y) = E[Y^2] - (E[Y])^2 = \frac{57}{9} - \left(\frac{7}{3}\right)^2$$

$$\text{Var}(Y) = \frac{8}{9} \text{ Ans} \quad (E[Y] \text{ from (g)})$$

$$\text{Now, } \text{Var}(X+Y) = E[(X+Y)^2] - (E[X+Y])^2 \quad (1)$$

Finding  $(E[X+Y])^2$   
Using Linearity,  $E[X+Y] = E[X] + E[Y]$

$$\text{So, } (E[X+Y])^2 = \left(3 + \frac{7}{3}\right)^2 \quad (\text{from part (g)}) \\ = 256/9$$

Now,  $E[(X+Y)^2] = E[X^2 + Y^2 + 2XY]$

Using Linearity again,

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

Put  $E[X^2]$  &  $E[Y^2]$  from above eq  $\star$   
&  $E[XY]$  from part (g)

$$\text{So, } E[(X+Y)^2] = \frac{21}{2} + \frac{57}{9} + 2 \cdot \frac{61}{9} \\ = \frac{547}{18}$$

$$\text{Therefore, } VAM(X+Y) = \frac{547}{18} - \frac{256}{9}$$

$$VAM(X+Y) = \frac{35}{18}$$

(i) Event 'A' is  $Y \leq X$ .

Sample space for A is  $S_A$  then

$$S_A = \{(2,1), (1,1), (4,1), (4,3)\}$$

~~$P(X=2|A)$~~  finding  $P(A)$ ;

$$P(A) = f_{XY}(2,1) + f_{XY}(1,1) + f_{XY}(4,1) + f_{XY}(4,3)$$

$$P(A) = \frac{5}{72} + \frac{2}{72} + \frac{17}{72} + \frac{25}{72} = \frac{49}{72}.$$

$$P(X=2|A) = \frac{P((X=2) \cap A)}{P(A)} = \frac{5/72}{49/72} = 5/49$$

Similarly,

$$P(X=1|A) = \frac{2/72}{49/72} = 2/49$$

$$P(X=4|A) = \frac{17/72 + 25/72}{49/72} = 42/49$$

So, we can use ~~law of~~ ~~of~~ conditional expectation that is

$$E[X|A] = \sum x_i P(x_i|A)$$

$$x_i = R_X$$

$$\text{Hence, } R_X = \{2, 1, 4\}$$

$$\text{So, } E[X|A] = 2 \cdot \frac{5}{49} + 1 \cdot \frac{2}{49} + 4 \cdot \frac{42}{49}$$

$$E[X|A] = 180/49 \quad \underline{\text{Ans}}$$

Now,

$$V_{AM}(X|A) = E[X^2|A] - (E[X|A])^2$$

$$\text{since, } E[X^2|A] = (2)^2 \cdot \frac{5}{49} + (1)^2 \cdot \frac{2}{49} +$$

$$(4)^2 \cdot \frac{42}{49}$$

$$= 694/49$$

$$\text{So, } V_{AM}(X|A) = \frac{694}{49} - \left(\frac{180}{49}\right)^2$$

$$= \frac{1606}{2401}$$

Ans

Q.4

Given:  $f(q) = \begin{cases} 6q(1-q); & 0 \leq q \leq 1 \\ 0; & \text{otherwise} \end{cases}$

&  $P(X=1 | \Omega=q) = q$ .  
 $\Rightarrow P(X=0 | \Omega=q) = 1-q$

Let's first find  $P(X=1)$ .

According to Law of total probability:

$$P(X=1) = \int_0^1 P(X=1 | \Omega=q) dq$$

$$P(X=1) = \int_0^1 q dq = y_2 \quad * \text{ because } P_\Omega \in [0,1].$$

Similarly,  $P(X=0) = \int_0^1 1-q dq = y_2$ .

Now,

$$\underset{\Omega|X}{P}(q|x) = \frac{P(\Omega=q \text{ And } X=x)}{P(X=x)} - ①$$

&

$$\underset{X|\Omega}{P}(x|q) = \frac{P(X=x \text{ And } \Omega=q)}{P(\Omega=q)} - ②$$

① ÷ ②

We get:

$$\frac{P_{\theta|x}(q|x)}{P_{x|\theta}(x|q)} = \frac{P(\theta=q)}{P(x=x)} \quad \text{Eq. (11)}$$

for  $x=1$ ; in eq. (11)

$$P_{\theta|x}(q|1) = \frac{P(\theta=1)}{P(x=1)} P_{x|\theta}(1|q)$$

$$= \frac{6q(1-q)}{Y_2} \cdot q$$

$$= 12q^2(1-q) ; 0 \leq q \leq 1$$

& for  $x=0$ ; in eq. (11)

$$P_{\theta|x}(q|0) = \frac{P(\theta=0)}{P(x=0)} P_{x|\theta}(0|q)$$

$$= \frac{6q(1-q)}{Y_2} \cdot (1-q)$$

$$= 12q(1-q)^2 ; 0 \leq q \leq 1$$

Therefore,  ~~$\forall q \in [0, 1]$~~

$$P_{\theta|x}(q|x) = \begin{cases} 12q^2(1-q) & ; x=1 \text{ } 0 \leq q \leq 1 \\ 12q(1-q)^2 & ; x=0 \text{ } 0 < q \leq 1 \\ 0 & ; \text{otherwise} \end{cases}$$

Q. 5

Let the dam have its centre at origin.  
 Then, for any point on the dam  
 $(x, y) \rightarrow x^2 + y^2 \leq r^2$  i.e. circle  
 (say C)

Range of  $X \in [-r, r]$  }  $X \neq Y$  and RY  
 " "  $Y \in [-r, r]$  } representing x & y  
 co-ordinates resp.

Since, all the points of impact are  
 equally likely hence,

$$f_{XY}(x, y) = \begin{cases} \frac{1}{\pi r^2}; & x^2 + y^2 \leq r^2 \\ 0 & \text{otherwise} \end{cases}$$

Now finding conditional Pmf  $f_{X|Y}(x|y)$ .

$$f_{X|Y}(x=y) = f_{X,Y}(x=y)$$

$$f_Y(y)$$

$$f_{X|Y}(x|y) = \frac{Y}{\pi r^2}$$

$$f_Y(y)$$

finding  $f_Y(y)$ ; that marginal PDF.

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

$$f_Y(y) = \int_{-t}^t \frac{1}{\pi r^2} dx \quad ; \quad t = \sqrt{r^2 - y^2}$$

$\text{if } |y| < r$

$$f_Y(y) = 2t/\pi r^2$$

Therefore,

$$f_{X|Y}(x|y) = \frac{1}{\pi r^2} f_Y(y)$$

$$= \frac{1}{2\pi/\pi r^2}$$

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

$$f_{X|Y}(x|y) = \frac{1}{2\sqrt{r^2-y^2}}, \text{ if } x^2+y^2 \leq r^2$$

For each value  $y$ ,  $f_{X|Y}(x|y)$  is uniform.

$$\text{So, } f_{X|Y}(x|y) = \begin{cases} \frac{1}{2\sqrt{r^2-y^2}} & ; x^2+y^2 \leq r^2 \\ 0 & ; \text{Otherwise} \end{cases}$$

Q.6

Total ways to distribute 4 balls in 4 bins =  $4^4$   
because each ball have 4 options.

(i) Let  $X$  be number of balls in 3<sup>rd</sup> bin.

Favourable cases =  ${}^4C_x$ .

Choosing 'x' balls for 3<sup>rd</sup> bin  $\rightarrow$  distributing other  $(4-x)$  balls in bin 1, 2, 4.

Thus,  $P_X(x) = \begin{cases} {}^4C_x \cdot 3^{4-x} & ; x \in \{0, 1, 2, 3, 4\} \\ 0 & ; \text{otherwise.} \end{cases}$

Thus,

$$P_X(x) = \begin{cases} 81/256 & ; x=0 \\ 27/256 & ; x=1 \\ 27/128 & ; x=2 \\ 3/64 & ; x=3 \\ 1/256 & ; x=4 \\ 0 & ; \text{otherwise} \end{cases}$$

(ii) Let  $N$  be number of non-empty bins.

$n=0$  is not possible because every ball needs to be in a bin.

Favourable cases for ( $n=1$ ) = Selecting 1 bin =  ${}^4C_1 = 4$   
to store all ball

Favourable cases for ( $n=2$ ) = selecting 2 bin  $\times$   
(ways of assignment)

$$= \frac{4}{C_2} [2^4 - 2] \quad \begin{array}{l} \text{Subtracted because of 2} \\ \text{Cause exist when all} \\ \text{balls goes in 1 bin.} \end{array}$$

Favourable cases for (n=3) = select 3 bin x  
(way of Assignment)

using Inclusion-Exclusion Principle

For this case, Way of assignment =

$$3^4 - \frac{3}{2}c_1^4 + \frac{3}{2}c_2^4 = 36$$

$$\text{So, Favorable case (n=3)} = {}^4C_3 \times 36 = 144$$

Favourable cases for ( $n=4$ ) = distributing balls

s.t. no bin is empty

$$= \frac{4}{1} - \frac{4}{C_1^3} + \frac{4}{C_2^2} - \frac{4}{C_3^1} \quad \left. \begin{array}{l} \text{Inclusion} \\ \text{Exclusion} \end{array} \right\}$$

1 = 24

$$P_A(n) = \begin{cases} 1/64 & ; n=1 \\ 21/64 & ; n=2 \\ 9/16 & ; n=3 \\ 3/32 & ; n=4 \\ 0 & ; \text{otherwise} \end{cases}$$

(iii) A = 1<sup>st</sup> ball sent to same-label bin

$$80, P(A) = \frac{4^3}{4^4} = \frac{1}{4} \rightarrow 1 \text{ bin is favourable.}$$

$\rightarrow 4 \text{ bin available}$

$$P(A) = \frac{1}{4}.$$

\* Now  $x=1 \& n=2$

$\xrightarrow{3^{\text{rd}} \text{ bin}}$   $\xrightarrow{1^{\text{st}} \text{ bin}}$   $\xrightarrow{2^{\text{nd}} \text{ bin}}$  2 empty bins

has 1 ball only.

Due to A, 1<sup>st</sup> bin has at least 1 ball.

Thus, # balls in bin 1 & 3 is 3, 1 resp.

So, Favorable count =  ${}^3C_1 = 3$  = choosing which ball goes to bin 3

out of 2, 3rd & 4<sup>th</sup> ball.

\*  $\xrightarrow{3^{\text{rd}} \text{ bin has 2 balls only}}$

Now,  $x=2 \& n=2 \Rightarrow$  2 empty bins

Due to A, 1<sup>st</sup> bin has ball no. 1.

Favorable count = Arranging 3<sup>rd</sup> bin 2 ball out

of 2<sup>nd</sup>, 3<sup>rd</sup> & 4<sup>th</sup> ball =  ${}^3C_2 = 3$ .

\*  $x=3 \& n=2$  (2 empty bins)

1<sup>st</sup> bin  $\Rightarrow$  1 ball (ball no. 1) due to A

3<sup>rd</sup> bin has exactly 3 ball.

Only 1 favorable case = 1<sup>st</sup> ball in  
bin 1 & all other in 3<sup>rd</sup> bin = 1.

\*  $x=1 \& n=3$  (only 1 empty bin)

1<sup>st</sup> bin  $\Rightarrow$  ball no. 1.

3<sup>rd</sup> bin  $\Rightarrow$  1 ball only; ways =  ${}^3C_1$  out of 2, 3, 4

Selecting other filled bin =  ${}^2C_1$ ,

Favorable count =  ${}^3C_1 \cdot {}^2C_1 \cdot (2-1) = 18$

\*  $x=2 \& n=3$  (only 1 empty bin)

1<sup>st</sup> bin  $\rightarrow$  1<sup>st</sup> ball

3<sup>rd</sup> bin  $\Rightarrow$  2 ball only ways =  ${}^3C_2$ , out of 2  
1<sup>st</sup>, 2<sup>nd</sup> & 3<sup>rd</sup> ball.

Selecting other bin =  ${}^2C_1$

Favorable count =  ${}^3C_2 {}^2C_1 = 6$ .

\*  $x=3 \& n=3$  (only 1 empty bin)

1<sup>st</sup> bin  $\rightarrow$  1<sup>st</sup> ball

3<sup>rd</sup>  $\Rightarrow$  3 balls only ways =  ${}^3C_3$ , 3 out of 2<sup>nd</sup>, 3<sup>rd</sup> & 4<sup>th</sup> ball.

Selecting other bin is not possible because no such case can exist because all the ball gets exhausted.

$$\text{Now, As } P_{X,N|A}(x,n) = \frac{P((x,n) \cap A)}{P(A)} = \frac{c(x,n)}{4^4} = \frac{1}{4}$$

;  $c(x,n)$  is favorable count.

We have following table giving

$P_{X,N|A}(x,n)$  for corr. x & n values:-

<del>x^n</del>	1. 2	2. 3	3. 4
1	$\frac{3}{4}^3 = \frac{3}{64}$	<del><math>\frac{10}{4}^3 = \frac{9}{32}</math></del>	
2	$\frac{3}{4}^3 = \frac{3}{64}$	$\frac{6}{4}^3 = \frac{3}{32}$	
3	$\frac{1}{4}^3 = \frac{1}{64}$	0.	

Q. 7

(a)

Let 'S' be RV denoting total no. of steps. And 'T' is time when it be passed out.

Acc. to Law of total expectation:

$$E[S] = \sum_{t=1} E[S|T=t] P_T(t)$$

Finding  $E[S|T=1]$ ? At  $T=1$  :-

All possible steps = {0, 1, 2}

So,

$$E[S|T=1] = 0 \cdot \left(1 - \frac{1}{4} - \frac{1}{2}\right) + 1 \cdot \frac{1}{2} + 2 \cdot \frac{1}{4}$$

$$= 1$$

$$\text{&} P(1) = \gamma_3.$$

Finding  $E[S|T=2]$ ? at  $T=2$  :-

All possible steps = {0, 1, 2, 3, 4}

So,

$$E[S|T=2] = 0 \cdot \frac{1}{4} \cdot \frac{1}{4} + 1 \cdot \left(\frac{1}{4} \cdot \frac{1}{2} + \frac{1}{2} \cdot \frac{1}{4}\right) +$$

$$2 \cdot \left(\frac{1}{4} \cdot \frac{1}{4} + \frac{1}{4} \cdot \frac{1}{4} + \frac{1}{2} \cdot \frac{1}{2}\right) +$$

$$3 \cdot \left(\frac{1}{2} \cdot \frac{1}{4} + \frac{1}{4} \cdot \frac{1}{2}\right) + 4 \cdot \left(\frac{1}{4} \cdot \frac{1}{4}\right)$$

$$= \left(\frac{1}{8} + \frac{1}{8}\right) + \left(\frac{1}{8} + \frac{1}{8} + \frac{1}{2}\right) + \left(\frac{3}{4}\right) + \left(\frac{1}{4}\right)$$

$$E[S|T=2] = \frac{1}{4} + \frac{1}{4} + \frac{1}{2} + 1 = 2$$

$$\& P_p(2) = Y_3.$$

Finding  $E[S|T=3]$  ?

At  $T=3$  :-

All possible steps =  $\{0, 1, 2, 3, 4, 5, 6\}$

$$P_{S_i}(0) = \cancel{\frac{1}{4} \cdot \frac{1}{4} \cdot \frac{1}{4}} = \frac{1}{64} \quad (0, 0, 0)$$

$$P_{S_i}(1) = 3 \times \cancel{\frac{1}{2} \cdot \frac{1}{4} \cdot \frac{1}{4}} = \frac{3}{32} \quad (0, 1, 0)$$

$$P_{S_i}(2) = 3 \times \left( \cancel{\frac{1}{4} \cdot \frac{1}{4} \cdot \frac{1}{4}} \right) + \quad (0, 0, 2), \\ (0, 1, 1)$$

$$3 \times \left( \cancel{\frac{1}{4} \cdot \frac{1}{2} \cdot \frac{1}{2}} \right) = \frac{15}{64} \quad \cancel{3x}$$

$$P_{S_i}(3) = 6 \times \left( \cancel{\frac{1}{4} \cdot \frac{1}{2} \cdot \frac{1}{4}} \right) + \quad (0, 1, 0) \} 6x$$

$$\left( \cancel{\frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2}} \right) = \frac{10}{32} \quad (1, 1, 1)$$

$$P_{S_i}(4) = 3 \times \left( \cancel{\frac{1}{4} \cdot \frac{1}{4} \cdot \frac{1}{4}} \right) + \quad (0, 2, 0) \} 3x$$

$$3 \times \left( \cancel{\frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{4}} \right) = \frac{15}{64} \quad (1, 1, 2) \} 3x$$

$$P_{S_i}(5) = 3 \times \cancel{\frac{1}{2} \cdot \frac{1}{4} \cdot \frac{1}{4}} = \frac{3}{32} \quad (1, 2, 1) \} 3x$$

$$P_{S_i}(6) = 1 \cdot \left( \cancel{\frac{1}{4} \cdot \frac{1}{4} \cdot \frac{1}{4}} \right) = \frac{1}{64} \quad (2, 2, 1)$$

$$\text{So, } E[S|T=3] = 0 \cdot P_{S_i^0}(0) + 1 \cdot P_{S_i^0}(1) + \\ 2 \cdot P_{S_i^1}(2) + 3 \cdot P_{S_i^1}(3) + 4 \cdot P_{S_i^1}(4) + \\ 5 \cdot P_{S_i^2}(5) + 6 \cdot P_{S_i^2}(6)$$

$\Rightarrow E[S|T=3] = 0 + \frac{3}{32} + \frac{15}{32} + \frac{30}{32} +$

$$\frac{30}{32} + \frac{15}{32} + \frac{3}{32}$$

$$= \frac{96}{32} = 3$$

$$\& P(3) = y_3$$

thus,

$$E[S] = 1 \cdot \frac{1}{3} + 2 \cdot \frac{1}{3} + 3 \cdot \frac{1}{3}$$

$$E[S] = 6/3 = 2 \quad \text{Ans}$$

✓

Q.7

(b) Let 's' be RV denoting his position when he passes out & T be that time.

$$E[S^5] = \sum_{s_i \in R_s} (s_i)^5 P(S=s_i)$$

from Law of Total Probability;  $\rightarrow Y_3$

$$P(S=s_i) = \sum_{t=1}^3 P(S=s_i | T=t) P(T=t),$$

— (1)

Let  $A_1, A_2, A_3$  represents set of position of drunker when passes at  $T=1, 2, 3$  respectively.  $R_s$  will be Union of  $A_1, A_2, A_3$   
i.e.

$$R_s = A_1 \cup A_2 \cup A_3$$

We can observe,  $A_1 = \{0, 1, -2\}$

$$\& A_2 = \{-4, -2, -1, 0, 1, 2\}$$

$$\& A_3 = \{-6, -4, -3, -2, -1, 0, 1, 2, 3\}$$

$$\text{Thus, } R_s = \{-6, -4, -3, -2, -1, 0, 1, 2, 3\}$$

Now using eq. (1);

\*  $\sum_{i=1}^3 P(S=-6 | T=i) \frac{1}{3}$

$$P(S=-6) = \sum_{i=1}^3 P(S=-6 | T=i) \frac{1}{3} = P(S=-6 | T=3) \frac{1}{3}$$

$$\Rightarrow P(S=-6) = \frac{1}{3} \cdot \frac{1}{4} \cdot \frac{1}{4} \cdot \frac{1}{4} = \frac{1}{3 \cdot 64}$$

Similarly;

\*  $\sum_{i=1}^3 P(S=-4 | T=i) \frac{1}{3}$

$$P(S=-4) = \frac{1}{3} P(S=-4 | T=2) + \frac{1}{3} P(S=-4 | T=3)$$

$$= \frac{1}{3} \cdot \frac{1}{16} + \frac{1}{3} \cdot \frac{3}{64} = \frac{1}{3} \cdot \frac{7}{64}$$

\*  $\sum_{i=1}^3 P(S=-3 | T=i) \frac{1}{3}$

$$P(S=-3) = \frac{1}{3} P(S=-3 | T=3) = \frac{1}{3} \cdot \frac{3}{32}$$

\*  $\sum_{i=1}^3 P(S=-2 | T=i) \frac{1}{3}$

$$P(S=-2) = \frac{1}{3} P(S=-2 | T=1) + \frac{1}{3} P(S=-2 | T=2)$$

$$+ \frac{1}{3} P(S=-2 | T=3)$$

$$= \frac{1}{3} \cdot \frac{1}{4} + \frac{1}{3} \cdot \frac{1}{8} + \frac{1}{3} \cdot \frac{3}{64}$$

$$= \frac{1}{3} \cdot \frac{16+8+3}{64} = \frac{1}{3} \cdot \frac{27}{64}$$

\*  $\sum_{i=1}^3 P(S=-1 | T=i) \frac{1}{3}$

$$P(S=-1) = \frac{1}{3} P(S=-1 | T=2) + \frac{1}{3} P(S=-1 | T=3)$$

$$= \frac{1}{3} \cdot \frac{1}{4} + \frac{1}{3} \cdot \frac{6}{32} = \frac{1}{3} \cdot \frac{14}{32}$$

\*

$$P(S=0) = \frac{1}{3} P(S=0 | T=1) + \frac{1}{3} P(S=0 | T=2) + \frac{1}{3} P(S=0 | T=3)$$

$T =$  Let say  $k$ .

\*

$$P(S=1) = \frac{1}{3} \sum_{i=1}^3 P(S=1 | T=i)$$

$$= \frac{1}{3} \left[ \frac{1}{2} + \frac{1}{4} + \frac{3}{32} \right] = \frac{1}{3} \cdot \frac{16+8+3}{32}$$

$$= \frac{1}{3} \cdot \frac{27}{32}$$

\*

$$P(S=2) = \frac{1}{3} \sum_{i=2}^3 P(S=2 | T=i)$$

$$= \frac{1}{3} \left[ \frac{1}{4} + \frac{3}{16} \right] = \frac{1}{3} \cdot \frac{7}{16}$$

\*

$$P(S=3) = \frac{1}{3} P(S=3 | T=3) = \frac{1}{3} \cdot \frac{1}{8}$$

Now,

$$\begin{aligned} E[S] &= (-6)^5 P(S=-6) + (-4)^5 P(S=-4) + \\ &\quad (-3)^5 P(S=-3) + (-2)^5 P(S=-2) + \\ &\quad (-1)^5 P(S=-1) + (0)^5 P(S=0) + \\ &\quad (1)^5 P(S=1) + (2)^5 P(S=2) + \\ &\quad (3)^5 P(S=3) \end{aligned}$$

$$= \left( \frac{-7776}{3.64} \right) + \left( \frac{-1024.7}{3.64} \right) +$$

$$\left( \frac{-243.3}{3.32} \right) + \left( \frac{-32.27}{3.64} \right)$$

$$+ \left( \frac{-14}{3.32} \right) + 0 + \left( \frac{27}{3.32} \right) +$$

$$\left( \frac{32.7}{3.16} \right) + \left( \frac{243}{3.8} \right)$$

$$= \frac{1}{3.64} [ -7776 - 7168 + 1458 \\ - 864 - 28 + 0 + ] \\ 54 + 896 + 1944 ]$$

$$= \frac{1}{3.64} (-14400)$$

$$= -75$$

Ans

Q.9

Total throws = 4

Range of RV 'X' =  $R_X = \{0, 1, 2, 3, 4\}$

Range of RV 'Y' =  $R_Y = \{0, 1, 2, 3, 4\}$

and  $X+Y \leq 4$ .

Let joint PMF of  $X, Y = f_{XY}(x, y)$

;  $x \in R_X, y \in R_Y, x+y \leq 4$ .

Let  $y=0$  then  $f_{XY}(x, 0) = 4c_x \left(\frac{1}{6}\right)^x \left(\frac{4}{6}\right)^{4-x}$   
 $; x \in R_X$

Similarly for  $y=1$ ,  $f_{XY}(x, 1) = 4c_x \left(\frac{1}{6}\right)^x \left(\frac{1}{6}\right)^{3-x} c_1 \left(\frac{1}{6}\right)^1 \left(\frac{4}{6}\right)^3$   
 $; x \in R_X = \{1, 2, 3\}$

$y=2$ ,  $f_{XY}(x, 2) = 4c_x \left(\frac{1}{6}\right)^x \left(\frac{1}{6}\right)^2 \left(\frac{4}{6}\right)^{2-x}$   
 $; x \in R_X = \{0, 1, 2, 3\}$

$y=3$ ,  $f_{XY}(x, 3) =$

$4c_x \left(\frac{1}{6}\right)^x \left(\frac{1}{6}\right)^3 \left(\frac{4}{6}\right)^{1-x}$   
 $; x \in R_X = \{0, 1, 2, 3\}$

for  $y=4$ ;  $f_{xy}(0,4) = \left(\frac{1}{6}\right)^4$   
 (or will be 0)  
 Only)

Therefore, we can write

$$f_{xy}(x,y) = 4C_x\left(\frac{1}{6}\right)\left(\frac{4}{6}\right)^{4-x}; x \in R_x, y=0$$

$$4C_1 \cdot 3C_x\left(\frac{1}{6}\right)\left(\frac{1}{6}\right)\left(\frac{4}{6}\right)^{3-x}; x \in R_x - \{4\}, y=1$$

$$4C_2 \cdot 2C_x\left(\frac{1}{6}\right)\left(\frac{1}{6}\right)^2\left(\frac{4}{6}\right)^{2-x}; x \in R_x - \{3,4\}, y=2$$

$$4C_3 \cdot 1C_x\left(\frac{1}{6}\right)\left(\frac{1}{6}\right)^3\left(\frac{4}{6}\right)^{1-x}; x \in \{0,1\}, y=3$$

$$\left(\frac{1}{6}\right)^4; y=4$$

0 ; otherwise.

0.9 remaining

Hence, Joint PMF is

<del>X\Y</del>	0	1	2	3	4
0	$\frac{16}{81}$	$\frac{16}{81}$	$\frac{2}{27}$	$\frac{1}{81}$	$\frac{1}{1296}$
1	$\frac{16}{81}$	$\frac{4}{27}$	$\frac{1}{27}$	$\frac{1}{324}$	0
2	$\frac{2}{27}$	$\frac{1}{27}$	$\frac{1}{216}$	0	0
3	$\frac{1}{81}$	$\frac{1}{324}$	0	0	0
4	$\frac{1}{1296}$	0	0	0	0

Q. 10

$$E[N] = c \quad \& \quad V_{AM}(N) = v$$

$$E[X_i] = k \quad \& \quad V_{AM}(X_i) = m$$

Since,  $Y = \sum_i X_i$

We can write, (if  $N$  is stochastic)  
using law of total probability for expectation;

$$E[Y] = \sum_{n_i \in R_N} E[Y|N=n_i] P_N(n_i)$$

$$E[Y] = E[E[Y|N]]$$

because  $E[Y|N]$  will be function  
of  $N$  i.e.  $g(N)$ .

$$So, E[Y] = E\left[E\left[\sum_i X_i | N\right]\right]$$

$$= E\left[\sum_{i=1}^N E[X_i | N]\right]$$

; linearity of expectation

$$= E\left[\sum_{i=1}^N E[X_i]\right]$$

;  $X_i$ 's &  $N$  are independent

$$\Rightarrow E[Y] = E[N] E[X_i]$$

$E[X_i]$  are equal

$$\Rightarrow E[Y] = E[X_i] E[N] \quad \therefore \text{since } E[X_i]$$

does not vary

$$\Rightarrow E[Y] = k c. \quad \text{for given } N \text{ & } c$$

= Ans is not random

Now, using law of total variance;

$$\text{Var}(Y) = E[\text{Var}(Y|N)] + \text{Var}(E[Y|N])$$

$$= E[\text{Var}(Y|N)] + \text{Var}(N E[X_i])$$

$$= E[\text{Var}(Y|N)] + k^2 \text{Var}(N) \quad (\text{as above})$$

$$= E[\text{Var}(Y|N)] + k^2 \nu$$

$$\text{As, } \text{Var}(Y|N) = \sum_{i=1}^N \text{Var}(X_i|N) = \sum_{i=1}^N \text{Var}(X_i)$$

$$(\because X_i's \text{ are independent of } N) = N \text{Var}(X_i) = Nm$$

$$\text{So, } \text{Var}(Y) = E(Nm) + k^2 \nu = m E(N) + k^2 \nu$$

$$\text{Var}(Y) = mc + k^2 \nu \quad \text{Ans}$$

Q.11

Let  $T$  be the random temperature (in  $^{\circ}\text{C}$ ) of the city, s.t.  $T \sim N(10, \frac{1}{10})$   
where  $\mu = 10 = 10^{\circ}\text{C}$ .

So,

$$T \sim N(10, 100)$$

$$P(T \leq 59^{\circ}\text{F}) = P(T \leq 15^{\circ}\text{C}) = ? = Q.$$

We can write,

$$f_T(t) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{t-\mu}{\sigma}\right)^2}$$

$$f_T(t) = \frac{1}{10 \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{t-10}{10}\right)^2} \quad t \in \mathbb{R}$$

$$\text{So, } Q = \int_{-\infty}^{15} f_T(t) dt \quad \text{(Shaded)}$$

$$Q = \int_{-\infty}^{15} \frac{1}{10 \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{t-10}{10}\right)^2} dt$$

$$\text{put } \frac{t-10}{10} = z \Rightarrow dt = 10dz$$

$$\text{So, } Q = \int_{-\infty}^{0.5} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} z^2} dz = \Phi(0.5) \quad \text{thus}$$

$$Q = 0.6915 \quad \text{Ans} \quad \text{CDF of standard normal distribution.}$$

Q.12

Let 'R' be RV denoting amount of rice.

Let 'C' be RV denoting quantity served.

Let 'W' be RV denoting weight (total).

ATQ:  $R \sim N(370 \text{ g}, (24)^2 \text{ g}^2)$

$$C \sim N(170 \text{ g}, (7)^2 \text{ g}^2)$$

&  $W = R + C$ .

We have to find  $P(W \leq 575 \text{ g}) = ?$

As proved in Q.13;

$$W = R + C \sim N(\mu_R + \mu_C, \sigma_R^2 + \sigma_C^2)$$

where  $\mu_R = 370 \text{ g}$ ,  $\mu_C = 170 \text{ g}$

&  $\sigma_R = 24 \text{ g}$  &  $\sigma_C = 7 \text{ g}$

So, if  $Z \sim N(0,1)$  i.e. standard normal RV

then,  $W = (\sigma_R^2 + \sigma_C^2)Z + (\mu_R + \mu_C)$

Now;

$$P(W \leq 575) = P(Z \leq \frac{575 - (\mu_R + \mu_C)}{\sqrt{\sigma_R^2 + \sigma_C^2}})$$

$$= \Phi\left(\frac{575 - (370 + 170)}{\sqrt{(24)^2 + (7)^2}}\right) = \Phi\left(\frac{35}{\sqrt{625}}\right) = \Phi(1.4)$$

$$= 0.919 \quad \text{Ans}$$

Q.13

$$X \sim N(\mu_x, \sigma_x^2) \equiv E[X] = \mu_x \\ \text{& } \text{Var}(X) = \sigma_x^2$$

$$Y \sim N(\mu_y, \sigma_y^2) \equiv E[Y] = \mu_y \\ \text{& } \text{Var}(Y) = \sigma_y^2$$

Let  $Z \sim N(0,1)$ 

RTP:

$$X+Y \sim N(\mu_x + \mu_y, \sigma_x^2 + \sigma_y^2)$$

Proof:

We have to prove that for

$$\text{RV } \Omega = X+Y, \text{ mean} = \mu_x + \mu_y$$

$$\text{and } \text{Var}(\Omega) = \sigma_x^2 + \sigma_y^2 \text{ i.e.}$$

$$\text{Var}(\Omega) = \text{Var}(X) + \text{Var}(Y).$$

Finding m.g.f for RV 'x' i.e.  $N_x(t)$ 

$$M_x(t) = E[e^{xt}] = \int_{-\infty}^{\infty} e^{xt} \frac{1}{\sigma_x \sqrt{2\pi}} e^{-\frac{1}{2} \left( \frac{x-\mu_x}{\sigma_x} \right)^2} dx$$

$$\therefore X = \sigma_x Z + \mu_x$$

$$M_x(t) = \int_{-\infty}^{\infty} e^{(\sigma_x Z + \mu_x)t} \frac{1}{\sqrt{2\pi}} e^{-\frac{Z^2}{2}} dz$$

$$M_x(t) = \frac{e^{\mu_x t}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{\sigma_x Z t} \cdot e^{-\frac{Z^2}{2}} dz$$

$$\text{Let } I = \int_{-\infty}^{\infty} e^{(\sigma_x t)z} \cdot e^{-\frac{z^2}{2}} dz$$

$$\text{So, } I = \int_{-\infty}^{\infty} e^{[(\sigma_x t)z - \frac{z^2}{2}]} dz$$

$$\text{Let } I = \int_{-\infty}^{\infty} e^{[\frac{(\sigma_x t)^2}{2} - (\frac{z}{\sqrt{2}} - \frac{\sigma_x t}{\sqrt{2}})^2]} dz$$

$$\text{Let } \lambda = (z - \sigma_x t)/\sqrt{2}$$

$$\Rightarrow d\lambda = dz/\sqrt{2}$$

$$\text{So, } I = \int_{-\infty}^{\infty} e^{[\frac{(\sigma_x t)^2}{2} - \lambda^2]} d\lambda$$

$$I = \exp\left[\frac{-(\sigma_x t)^2}{2}\right] \int_{-\infty}^{\infty} e^{-\lambda^2} d\lambda$$

$$I = \sqrt{2} e^{\frac{-(\sigma_x t)^2}{2}} \int_{-\infty}^{\infty} e^{-\lambda^2} d\lambda \rightarrow I_1$$

$$(I_1)^2 = \left[ \int_{-\infty}^{\infty} e^{-\lambda^2} d\lambda \right]^2 \left[ \int_{-\infty}^{\infty} e^{-k^2} dk \right]$$

Let  $\gamma = r \sin \theta$ ,  $k = r \cos \theta$

$$(I_1)^2 = \iint_{\gamma=0, \theta=0}^{\infty, 2\pi} e^{-r^2} r d\theta dr$$

$$= 2\pi \int_0^{\infty} r e^{-r^2} dr = -\pi \int_{r=0}^{r=\infty} e^{-r^2} d(-r^2)$$
$$= -\pi \left[ e^{-r^2} \right]_{r=0}^{r=\infty}$$

$$I_1^2 = \pi \Rightarrow I_1 = \sqrt{\pi}$$

$$\text{So, } I = \sqrt{2\pi} e^{(\bar{x} + \bar{y})^2/2}$$

$$\text{So, } M_x(t) = e^{\mu_x t} \cdot e^{(\bar{x} + \bar{y})^2/2}$$

$$\text{Similarly, } M_y(t) = e^{\mu_y t} \cdot e^{(\bar{x} + \bar{y})^2/2}$$

$$\text{Now, } M_B(t) = E(e^{tB}) = E\left(e^{t(X+Y)}\right)$$

$$\Rightarrow M_B(t) = E\left[e^{\bar{X}t} \cdot e^{\bar{Y}t}\right] \quad \left\{ \text{by linearity} \right.$$
$$= E[e^{\bar{X}t}] \cdot E[e^{\bar{Y}t}] \quad \left. \text{--- (1)} \right.$$

$$\Rightarrow M_B(t) = e^{\mu_B t} \cdot e^{(\bar{X} + \bar{Y})^2/2} \quad \text{--- (1)}$$

where  $\mu_B$  is mean &  $\sigma_B^2$  is

Variance of  $B$ .

Solving , eq. ①

$$M_Q(t) = e^{\mu_X t} \cdot e^{(\sigma_X)^2/2} \cdot e^{\mu_Y t} \cdot e^{(\sigma_Y)^2/2}$$

$$\Rightarrow M_Q(t) = \exp [(\mu_X + \mu_Y)t].$$

$$\exp \left[ (\sigma_X^2 + \sigma_Y^2) \frac{t^2}{2} \right]$$

→ ⑪

On comparing above eq. ⑪ with eq ⑪  
we find ;

$$\mu_Q = (\mu_X + \mu_Y)$$

$$\text{and } \sigma_Q^2 = (\sigma_X)^2 + (\sigma_Y)^2.$$

Also, from eq. ⑪ & due to  
uniqueness theorem, we can  
say Q will be normal distribution

Random variable with mean  
 $\mu_Q$  & varying  $\sigma_Q^2$ .

So,

$$Q = X + Y \sim N(\mu_Q, \sigma_Q^2).$$

$$\equiv N(\mu_X + \mu_Y, (\sigma_X^2 + \sigma_Y^2)).$$

Q. 14.

$$-(x+y)$$

$$f_{XY}(x,y) = \begin{cases} 4y(x-y)e^{-(x+y)} & ; 0 < x < \infty \\ 0 & ; 0 \leq y \leq x \\ 0 & ; \text{otherwise} \end{cases}$$

For,

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

$$(\text{due to constraint}) = \int_0^y x f_{X|Y}(x|y) dx$$

i.e. Domain consideration

Finding  $f_{X|Y}(x|y)$ :

$$f_{X|Y}(x|y) = f_{XY}(x,y) / f_Y(y).$$

$$\text{We can say, } f_Y(y) = \int_{-\infty}^{\infty} f_{XY}(x,y) dx$$

$$(\text{due to constraint}) = \int_0^y f_{XY}(x,y) dx$$

$$\text{so, } f_Y(y) = \int_y^{\infty} 4y(x-y)e^{-(x+y)} dx$$

$$f_Y(y) = 4ye^{-y} \int_0^{\infty} (x-y)e^{-\frac{(x+y)^2}{2}} dx$$

$$I = \int_{-\infty}^{\infty} y e^{-x} dx - y \int_{-\infty}^{\infty} e^{-x} dx$$

$$I = [-ye^{-x}]_{-\infty}^{\infty} + \int_{-\infty}^{\infty} e^{-x} dx - y [-e^{-x}]_{-\infty}^{\infty}$$

$$I = ye^{-y} + (1-y)(e^{-y}) = e^{-y}$$

$$\text{So, } f_y(y) = 4ye^{-2y} \quad ; \quad y > 0$$

Thus,

$$f_{x|y}(x,y) = \frac{4y(x-y)e^{-(x+y)}}{4ye^{-2y}}$$

$$f_{x|y}(x,y) = \begin{cases} (x-y)e^{(y-x)} & ; \\ 0 & ; \text{ otherwise} \end{cases}$$

Thus,

$$E[X|Y=y] = \int_0^{\infty} x(x-y)e^{-(x+y)} dx$$

$$E[X|Y=y] = e^y \int_0^{\infty} (x^2 - xy)e^{-x} dx \rightarrow I_1$$

$$I_1 = \int_0^{\infty} x^2 e^{-x} dx - y \int_0^{\infty} xe^{-x} dx$$

$$I_1 = \int_{-\infty}^{\infty} -x^2 e^{-y} + (2-y) \int_{-\infty}^{\infty} xe^{-y} dx$$

$$I_1 = y^2 e^{-y} + (2-y) \left[ \left[ -xe^{-y} \right]_{-\infty}^{\infty} + \int_{-\infty}^{\infty} e^{-y} dx \right]$$

$$I_1 = y^2 e^{-y} + (2-y) [ y e^{-y} + e^{-y} ]$$

$$I_1 = y^2 e^{-y} + (2-y)(y+1) e^{-y}$$

$$I_1 = e^{-y} (y^2 + 2y - y^2 - y)$$

$$I_1 = ye^{-y}(ay+2)$$

$$\therefore E[X|Y=y] = e^y I_1 = y+2$$

Ans

For  $\text{Var}(X|Y=y)$  :-Now finding  $E[X^2|Y=y]$  ?

$$E[X^2|Y=y] = \int_{-\infty}^{\infty} (x)^2 f_{X|Y}(x|y) dx$$

Considering domain of  $f_{X|Y}(x|y)$

$$= \int_{-\infty}^{\infty} x^2 (x-y) e^{-y} dx$$

$$\rightarrow I_2$$

$$= e^y \int_y^{\infty} x^2 (x-y) e^{-x} dx$$

$$I_2 = \int_{-\infty}^{\infty} x^3 e^{-x} dx - y \int_{-\infty}^{\infty} x^2 e^{-x} dx$$

$$I_2 = \left[ x^3 e^{-x} \right]_{-\infty}^{\infty} + (3-y) \int_{-\infty}^{\infty} x^2 e^{-x} dx$$

$$I_2 = y^3 e^{-y} + (3-y) \left[ x^2 e^{-x} \right]_{-\infty}^{\infty} +$$

$$\left. \int_{-\infty}^{\infty} x e^{-x} dx \right]$$

$$I_2 = y^3 e^{-y} + (3-y) \left[ y^2 e^{-y} + 2 \left[ -x e^{-x} \right]_{-\infty}^{\infty} + 2 \int_{-\infty}^{\infty} e^{-x} dx \right]$$

$$I_2 = y^3 e^{-y} + (3-y) \left[ y^2 e^{-y} + 2y e^{-y} + 2e^{-y} \right]$$

$$I_2 = e^{-y} \left[ y^3 + (3-y)(y^2 + 2y + 2) \right]$$

$$I_2 = e^{-y} \left[ y^2 + 4y + 6 \right]$$

∴,  $E[x^2 | Y=y] = y^2 + 4y + 6$

$$\text{So, } \text{Var}(X | Y=y) = E[x^2 | Y=y] - E[x | Y=y]^2$$

$$= (y^2 + 4y + 6) - (y^2 + 4y + 4) = 2 \text{ Ans}$$

Q.15

$$X_1 \sim N(2, 3)$$

$$\& X_2 \sim N(1, 4)$$

As found in Q.13,  
mgf of a 'Normal' RV 'X' is

$$M_X(t) = e^{t\mu_X} \cdot e^{\frac{(\sigma_X t)^2}{2}} \quad \text{--- (1)}$$

Therefore,

$$M_{X_1}(t) = \exp[-2t] \cdot \exp[3t^2/2]$$

&

$$M_{X_2}(t) = \exp[t] \cdot \exp[2t^2]$$

Thus,

$$M_Y(t) = e^{[2t + \frac{3t^2}{2}]}$$

$$M_{X_1}(t) = e^{[t + 2t^2]}$$

$$\& M_{X_2}(t) = e^{[t]}$$

(a)  $Y = 2X_1 + 3X_2$

$$\text{Since, } M_Y(t) = E[e^{tY}] = E[e^{t(2X_1 + 3X_2)}]$$

$$M_Y(t) = E[e^{2tX_1 + 3tX_2}]$$

Using Linearity;

$$M_Y(t) = E[e^{2tX_1}] \cdot E[e^{3tX_2}]$$

Thus,

$$M_Y(t) = M_{X_1}(2t) \cdot M_{X_2}(3t)$$

$$M_Y(t) = e^{\left[2(2t) + \frac{3(2t)^2}{2}\right]} \quad \left. \begin{array}{l} \text{Using} \\ \textcircled{A} \end{array} \right\}$$

$$e^{\left[(2t) + 2(3t)^2\right]}$$

$$\Rightarrow M_Y(t) = e^{[7t]} \cdot e^{[24t^2]}$$

On Comparing with eq.  $\textcircled{A}$  we find  
that, (also using Uniqueness theorem)  
 $M_Y(t)$  refers to mgf of a normal  
distribution whose mean =  $7t$   
and variance =  $48$ .

So,  $Y \sim N(7, 48)$  ~~or~~

$$(b) M_Y(t) = E[e^{Yt}] = E[e^{(X_1 + X_2)t}]$$

$$M_Y(t) = E[e^{X_1 t} \cdot e^{X_2 t}]$$

using Linearity;

$$M_Y(t) = E[e^{X_1 t}] \cdot E[e^{-X_2 t}]$$

Thus,

$$M_Y(t) = M_{X_1}(t) \cdot M_{X_2}(-t)$$

Using  $\star$

$$M_Y(t) = e^{[2t + \frac{3t^2}{2}]} \cdot e^{[-t + \frac{3t^2}{2}]} \\ = e^{[t]} \cdot e^{\frac{3t^2}{2}}$$

On comparing with eq. # and also  
using uniqueness theorem,  $M_Y(t)$  refers  
to mgf of a Normal distribution  
whose mean  $t=1$  and variance  $= 3t^2$

So,  $Y \sim N(1, 3t^2)$  Ans

~~11~~

Q.15 remaining

Therefore for part (a):

$$f_Y(y) = \frac{1}{4\sqrt{\pi}} \exp\left(-\frac{1}{2}\left(\frac{y-7}{4\sqrt{3}}\right)^2\right).$$

and for part (b):

$$f_Y(y) = \frac{1}{\sqrt{14\pi}} \exp\left(-\frac{1}{2} \cdot \frac{(y-1)^2}{7}\right).$$

Ans