

Probability & Statistics

~~Assignment : 3~~

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Q.1 = For RV, 'x' : pdf = $f_x(x)$.
& $Y = aX + b$.

(1) finding $f_Y(y)$?

⇒ First let us find out $F_Y(y)$ i.e.

$$P(Y \leq y) = P(aX + b \leq y)$$

$$F_Y(y) = P(X \leq \left(\frac{y-b}{a}\right))$$

Since x is a continuous RV

$$\Rightarrow f_x(x) = \frac{1}{a} e^{-\frac{x-b}{a}}$$

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

$$F_Y(y) = F_X\left(\frac{y-b}{a}\right)$$

Differentiating both sides w.r.t y ;

$$f_Y(y) = f_X\left(\frac{y-b}{a}\right) \left|\frac{1}{a}\right|.$$

$$\therefore Y = ax + b \quad \& \quad a \neq 0, b \neq 0.$$

(2) If X is exponential (λ)

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x} & ; x \geq 0 \\ 0 & ; \text{otherwise} \end{cases}$$

$$\therefore f_Y(y) = f_X\left(\frac{y-b}{a}\right) = \begin{cases} \lambda e^{-\lambda\left(\frac{y-b}{a}\right)} & ; y \geq b \\ 0 & ; \text{otherwise} \end{cases}$$

Thus,

RV ' Y ' will be exponential

only when $b=0$ & $a>0$

but $a < b$ are non-zero.

So, Y can never be exponential RV.

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

that is

$$f_X(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x-\mu}{\sigma} \right)^2} \quad \forall x \in \mathbb{R}$$

and from first part

$$f_Y(y) = \frac{1}{a} f_X\left(\frac{y-b}{a}\right)$$

$$f_Y(y) = \frac{1}{(a\sigma)\sqrt{2\pi}} \cdot e^{-\frac{1}{2} \left(\frac{y-(a\mu+b)}{a\sigma} \right)^2}$$

$\forall y \in \mathbb{R}$

$$f_Y(y) \sim N(a\mu+b, (a\sigma)^2)$$

(!! always !!)

$$Q.2 = X \sim N(\mu, \sigma^2) \rightarrow$$

that means if $Z = N(0, 1)$

then, $X = \sigma Z + \mu$.

And already $\Phi(-z) = 1 - \Phi(z)$; $\Phi(z) = CDF$
And $Y = e^X$. of Z

$$(I) F_Y(y) = P(Y \leq y) = P(e^X \leq y) = P(X \leq \ln y)$$

$$\text{So, } F_Y(y) = F_X(\ln y) \quad \text{--- (1)}$$

&

$$F_X(x) = P(X \leq x) = P(\sigma Z + \mu \leq x)$$

$$\text{So, } F_X(x) = F_Z\left(\frac{x-\mu}{\sigma}\right) = \Phi\left(\frac{x-\mu}{\sigma}\right)$$

--- (II)

from (1) & (II)

$$F_Y(y) = F_X(\ln y) = \Phi\left(\frac{\ln y - \mu}{\sigma}\right)$$

So;

$$F_Y(y) = \begin{cases} \Phi\left(\frac{\ln y - \mu}{\sigma}\right) & ; y > 0 \\ 0 & ; \text{otherwise} \end{cases}$$

$$\text{where } \Phi(m_1) = P(Z \leq m_1) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{m_1} e^{-\frac{m^2}{2}} dm.$$

(2)

Without method of transformation :-

As we know, $f_Y(y) = F'_Y(y)$

From part 1 of this question:

$$F_Y(y) = \phi \left[\frac{(\ln y - u)}{\sigma} \right] = \phi[ky]$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^K e^{-\frac{m^2}{2}} dm$$

where $k = (\ln y - u)/\sigma$; $y > 0$

Otherwise 0.

$$f_Y(y) = F'_Y(y)$$

$$= \phi'[k] \cdot \frac{dk}{dy}$$

$$= \frac{1}{\sqrt{2\pi}} e^{-k^2/2} \cdot \frac{dk}{dy}$$

And ; $dk = \frac{1}{\sigma} \frac{1}{y} dy$

$$-\frac{1}{2} (\ln y - u)^2$$

$$f_Y(y) = \begin{cases} \frac{1}{(\sqrt{2\pi}) (\sigma y)} e^{-\frac{1}{2} (\ln y - u)^2} & ; y > 0 \\ 0 & ; \text{otherwise} \end{cases}$$

$$\text{or } y = e^x$$

with method of transformation:-

$Y = e^X = g(x)$ is continuous & differentiable function hence we can use method of transformation.

[Also, it is strictly increasing].

$$f_Y(y) = \begin{cases} \frac{f_X(x_1)}{g'(x_1)} = f_X(x_1) \cdot \frac{dx_1}{dy}; y = g(x_1) \\ 0 \end{cases}$$

0 ; if $y = g(x)$ does not have solution.

And since $X \sim N(\mu, \sigma^2)$;
hence,

$$f_X(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\left(\frac{x-\mu}{\sigma}\right)^2/2}; \forall x$$

And $g'(x) = dy/dx = e^x$

$$\text{So, } f_Y(y) = \left(\frac{1}{\sigma\sqrt{2\pi}} e^{-\left(\frac{x-\mu}{\sigma}\right)^2/2} \right) / e^x$$

Putting $x = \ln y$

$$f_Y(y) = \begin{cases} \frac{1}{\sigma\sqrt{2\pi}} \frac{e^{-(\ln y - \mu)^2/2}}{(y)}, y = g(x) \\ 0, \text{ otherwise} \end{cases}$$

$$Q. 3 = S O I \Rightarrow$$

We can use Normal approximation of binomial distribution because $n = 12,000$ & $p = \frac{1}{6}$

$$\text{So, } np = 2000 > 5$$

$$\text{& } n(1-p) = 10,000 > 5.$$

Therefore, also according to central limit theorem

$$Z = \frac{Y - np}{\sqrt{np(1-p)}} \xrightarrow{\text{mean}} N(0,1)$$

\downarrow

$\sqrt{np(1-p)} \xrightarrow{\text{Variance}}$

where Y is RV for no. of 2 in given experiment i.e. of rolling a die 12,000 times.

Approximation \Rightarrow

$$P(1900 < Y < 2150) = ? \quad (\text{by continuity correction})$$

$$= P(1900.5 \leq Y \leq 2149.5)$$

$$= P\left(\frac{1900.5 - 2000}{\sqrt{2000 \times \frac{5}{6}}} \leq Z \leq \frac{2149.5 - 2000}{\sqrt{2000 \times \frac{5}{6}}}\right)$$

$$\Rightarrow P\left(\frac{-99.5}{40.8248} < Z < \frac{149.5}{40.8248}\right)$$

$$\Rightarrow P(-2.43724 < Z < 3.66199)$$

$$\Rightarrow \phi(3.66199) - \phi(-2.43724)$$

$$\Rightarrow \phi(3.66199) + \phi(2.43724) - 1$$

(using Normal distribution
Property)

$$\therefore \phi(3.66) + \phi(2.44) - 1$$

$$\therefore 0.99987 + 0.99266 - 1$$

$$\therefore 0.99253 \approx 0.992 \quad \text{Ans}$$

$$8 \cdot 4 = 80^n$$

- (1) Let B represents boy & G represents girl. Then the possible set for two children is
- $$U = \{ (B, B), (B, G), (G, B), (G, G) \}$$

Given one of them is boy, now sample space reduces to

$$S = \{ (B, G), (B, B), (G, B) \}$$

$P(\text{other is a girl})$ in the above sample space S is

$$P(B, G) + P(G, B)$$

$$= \frac{1}{3} + \frac{1}{3} = \frac{2}{3} \text{ Ans}$$

✓

- (2) Given that the child, Atsi is boy, now S' will either be $\{ (B, G), (B, B) \}$ so

$$P(\text{other child is girl in } S) = P(B, G)$$
$$= \frac{1}{2} \text{ Ans}$$

✓

(3) Yes, the answer for part a
to part b is different.
The reason is:-
When we named the boy
i.e. Atsi is a boy, sample
space got reduced from
 $S \rightarrow S'$ by deleting the
option of having
due to unique identification of
Atsi being a boy.

Hence the probability decreased
from $\frac{2}{3}$ to $\frac{1}{2}$.

(4)

Since, we named the child
let Atsi be 1st,

Without loss of generality

$$S = \{ \text{GB}, \text{GB}, \text{BG}, \text{BB} \}$$

Let A be an event Atsi is a boy.
E be other s.t. other child
in a girl given one is boy

$$P(A) = \frac{P(\{\text{BB}, \text{BG}\})}{P(S)} = \frac{3}{4}$$

$$P(E) = \frac{2}{3} \text{ from 1st part}$$

Now if other event B is defined
as Atsi's sibling is a girl
given Atsi is a boy.
Then

$$P(B) = P(A \cap E) = P(A) P(E)$$

as A & E are independent

$$P(B) = \frac{3}{4} \times \frac{2}{3} = \frac{1}{2}$$

An

Q.5 = (1)

$$Z = X + Y = G(X, Y)$$

$X = \text{Exponential}(\lambda)$, $Y = \text{Exponential}(\lambda)$

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x}; & x > 0 \\ 0; & x \leq 0 \end{cases}$$

$$f_Y(y) = \begin{cases} \lambda e^{-\lambda y}; & y > 0 \\ 0; & y \leq 0 \end{cases}$$

We can observe, $0 < Z$. *Always*.

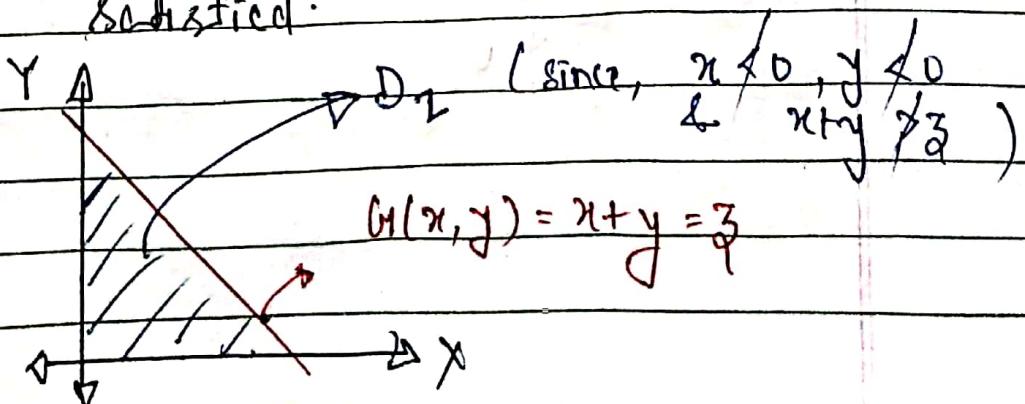
In general,

$$F_Z(z) = P(Z \leq z) = P(G(X, Y) \leq z)$$

$$= P((X, Y) \in D_z)$$

$$= \iint_{x, y \in D_z} f_{XY}(x, y) dx dy$$

where D_z is the region in the $X-Y$ plane where $G(X, Y) \leq z$ is satisfied.



Therefore,

$$F_Z(z) = \int_0^z \int_{y=0}^x f_{XY}(x,y) dx dy$$

Now, finding $f_{XY}(x,y)$

since, x & y are independent

$$f_{XY}(x,y) = f_X(x) \cdot f_Y(y)$$

$$= \lambda e^{-\lambda x} \cdot \lambda e^{-\lambda y}$$

$$= \lambda^2 e^{-\lambda x} e^{-\lambda y}$$

$$\text{So, } F_Z(z) = \int_{y=0}^z \int_{x=0}^{\infty} \lambda^2 e^{-\lambda y} \lambda e^{-\lambda x} dx dy$$

$$= \int_{y=0}^z \lambda^2 e^{-\lambda y} \left[\frac{1}{\lambda} - \frac{e^{-\lambda(z-y)}}{\lambda} \right] dy$$

$$= \lambda \int_{y=0}^z (e^{-\lambda y} - e^{-\lambda z}) dy$$

$$= \lambda \left[\frac{e^{-\lambda z}}{-\lambda} - (e^{-\lambda z})y \right]^3$$

$$= \lambda \left[\left(\frac{e^{-\lambda z}}{-\lambda} \right) - 3(e^{-\lambda z}) + \frac{1}{\lambda} \right]$$

$$= 1 - e^{-\lambda z} (1 + 3\lambda)$$

$$; \quad z > 0.$$

So, $f_z(z) = \frac{d}{dz} F_z(z)$

$$= \frac{d}{dz} (1 - e^{-\lambda z} (1 + 3\lambda))$$

$$= \lambda e^{-\lambda z} (1 + 3\lambda) - \lambda e^{-\lambda z}$$

$$= \lambda e^{-\lambda z} (1 + 3\lambda - 1)$$

$$f_z(z) = \begin{cases} \lambda^2 z e^{-\lambda z} & ; \quad z > 0 \\ 0 & ; \quad \text{otherwise} \end{cases}$$

Q. 5 \Rightarrow

(ii) $Z = X + Y = G(X, Y)$

$X = \text{Uniform}(0, 1), Y = \text{Uniform}(0, 1)$

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

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

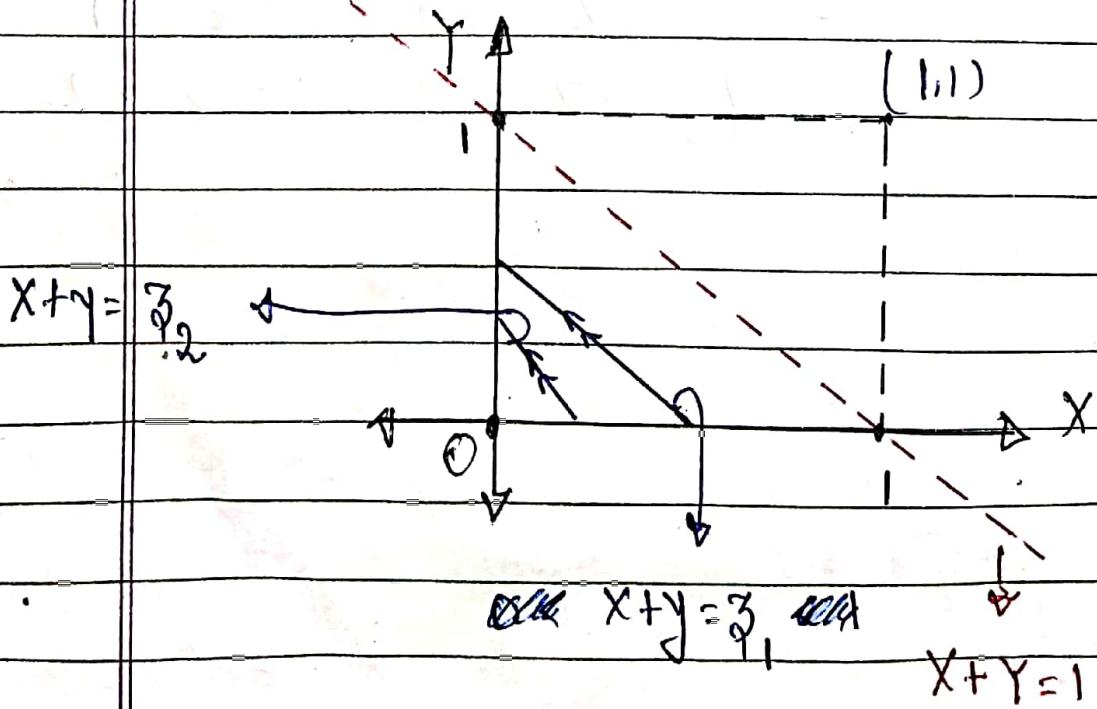
We can directly say $0 \leq Z \leq 2$.

Let's break Z in two

range $① 0 \leq Z \leq 1$

$② 1 \leq Z \leq 2$

Case: 1 :- $0 \leq Z \leq 1$



In general,

$$F_Z(z) = P(Z \leq z) = P(G(X,Y) \leq z)$$

$$= P((X,Y) \in D_z)$$

$$= \iint_{x,y \in D_z} f_{XY}(x,y) dx dy$$

where D_z in the X,Y plane represents

region such that $G_{xy}(x,y) \leq z$ is satisfied.

Now observing the graph:

We can directly say $1 > z_1 > z_2$.

Therefore one D_z will be area formed by following lines:-

$$(X=0), (Y=0) \text{ & } (X+Y=z ; 0 < z < 1.)$$

$$\text{So, } F_Z(z) = \iint_{\substack{x \\ y \\ x+y=z \\ y=0 \quad x=0}} f_{XY}(x,y) dx dy \quad \text{--- (1)}$$

As X, Y are independent;

$$f_{X,Y}(x,y) = f_X(x) \cdot f_Y(y)$$

$$= 1 \cdot 1 = 1.$$

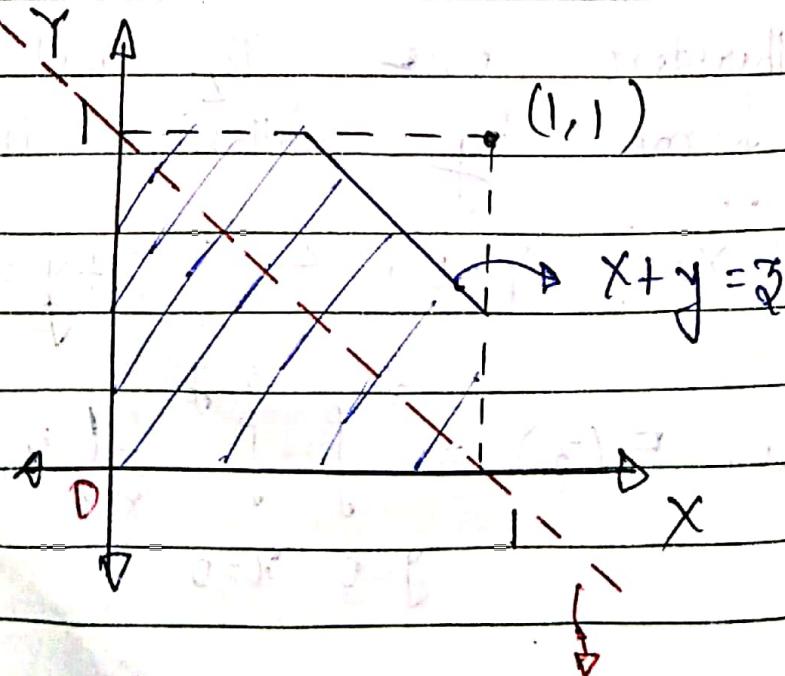
Therefore,

$$F_Z(z) = \int_{y=0}^z \int_{x=0}^{z-y} 1 \cdot dx dy$$

$$= \int_{y=0}^z (z-y) dy$$

$$= z^2/2.$$

Case 2 :- $1 < z < 2$.



$$x+y=1.$$

In this case ;
 for finding $F_z(z)$, only D_z will be the
 area in the above $x-y$ plane
 where (x,y) lie inside square
 formed by points $(0,0), (0,1), (1,0), (1,1)$
 And $x+y \leq z$.
 That is shaded region.

Calculating that way is very complex.
 Otherwise we can say,

$$F_z(z) = P(Z < z) = 1 - P(Z > z)$$

$$\therefore P(Z > z) = P((x,y) \in D_z^I)$$

$$= \iint_{x,y \in D_z^I} f_{xy}(x,y) dx dy$$

Where D_z^I is the $x-y$ plane represents
 region s.t $g(x,y) > z$ & $0 < x < 1$ &
 $0 < y < 1$.

i.e. unshaded part in the formed square.

$$\begin{aligned} \text{So, } P(Z > z) &= \int_0^1 \int_0^1 f_{xy}(x,y) dx dy \\ &\quad \text{from CND(I)} \\ &= \int_{y=z-1}^1 \int_{x=z-y}^1 1 \cdot dx \cdot dy \quad \left\{ \begin{array}{l} f_{xy}(x,y) \\ = 1 \end{array} \right. \\ &\quad \text{from CND(I)} \end{aligned}$$

$$\text{So, } P(Z > \bar{z}) = \int_{\bar{z}}^1 (1 - \bar{z} + y) dy$$

$$y = \bar{z} +$$

$$= 1 - \frac{(2-\bar{z})^2}{2}; \quad$$

$$1 \leq \bar{z} < 2.$$

$$\text{So, } F_Z(\bar{z}) = 1 - [P(Z > \bar{z})]$$

$$= 1 - (2-\bar{z})^2/2; \quad 1 \leq \bar{z} < 2.$$

Since,

$$f_Z(\bar{z}) = \frac{d}{d\bar{z}} [F_Z(\bar{z})]$$

$$= \begin{cases} \bar{z} & ; 0 \leq \bar{z} < 1 \\ 2-\bar{z} & ; 1 \leq \bar{z} \leq 2 \\ 0 & ; \text{otherwise} \end{cases}$$

An

$\text{Q. 6} =$

$$(1) \quad T = T_1 - T_2 = G(T_1, T_2)$$

$T_1 = \text{Exponential}(\lambda), T_2 = \text{Exponential}(\lambda)$

$$T_1 = \begin{cases} \lambda e^{-\lambda t_1}; & t_1 \geq 0 \\ 0; & \text{otherwise} \end{cases}$$

$$T_2 = \begin{cases} \lambda e^{-\lambda t_2}; & t_2 \geq 0 \\ 0; & \text{otherwise} \end{cases}$$

We can observe $\forall T \in \mathbb{R}$

Finding $F_T(t) \quad \forall t \geq 0$?

$$F_T(t) = P(T \leq t) = P(G(T_1, T_2) \leq t)$$

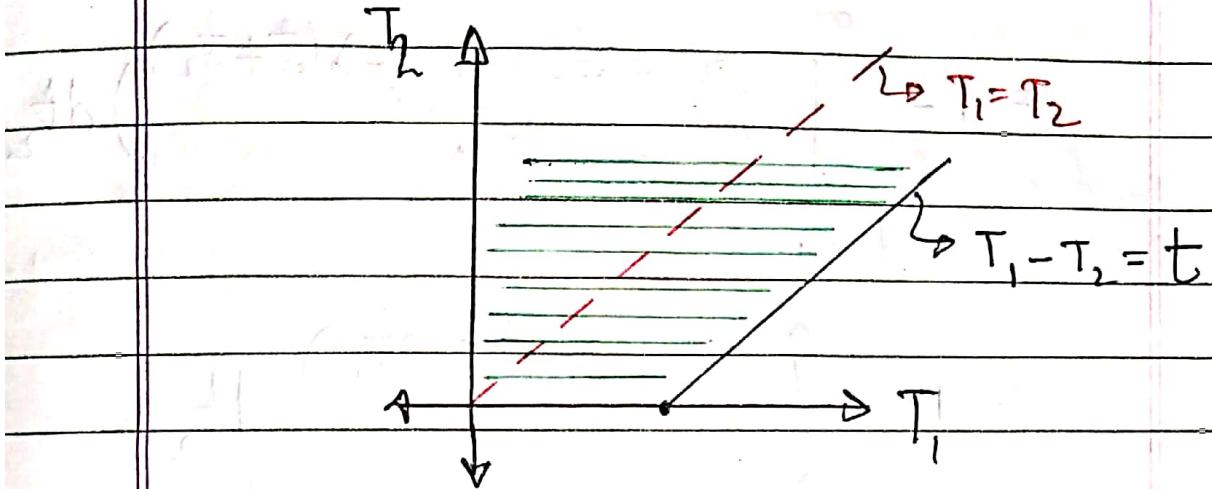
$$= P((t_1, t_2) \in D_2)$$

$$= \iint_{\substack{t_1, t_2 \\ (t_1, t_2) \in D_2}} f_{T_1, T_2}(t_1, t_2) dT_1 dT_2$$

where D_2 is the regions in
the $T_1 - T_2$ plane where $G(t_1, t_2)$

$\leq t$ is satisfied.

D_2 = ~~the~~ shaded portion of graph below.



We can write :-

$$F_T(t) = 1 - P((T_1, T_2) \in D_2)$$

where D_2 is region where $G(T_1, T_2) > t$.

$$\text{So, } F_T(t) = \int_{t_2=0}^{\infty} \int_{t_1=0}^{t+t_2} f_{T_1 T_2}(t_1, t_2) dt_1 dt_2$$

Since, T_1 & T_2 are independent hence

$$\begin{aligned} f_{T_1 T_2}(t_1, t_2) &= f_{T_1}(t_1) \cdot f_{T_2}(t_2) \\ &= \lambda^2 e^{-\lambda t_1} e^{-\lambda t_2} \end{aligned}$$

$$So, F_T(t) = \int_0^\infty \lambda^2 e^{-\lambda t_2} \left[\int_{t_1=0}^{t+t_2} e^{-\lambda t_1} dt_1 \right] dt_2$$

$$F_T(t) = \int_0^\infty \lambda^2 e^{-\lambda t_2} \left[1 - e^{-\lambda(t+t_2)} \right] dt_2$$

$$F_T(t) = \lambda \int_0^\infty [e^{-\lambda t_2} - e^{-\lambda(t+2t_2)}] dt_2$$

$$F_T(t) = \lambda \left[\frac{e^{-\lambda t_2}}{(-\lambda)} - \frac{e^{-\lambda(t+2t_2)}}{(-2\lambda)} \right]_0^\infty$$

$$F_T(t) = \lambda \left[(0 - 0) - \left(\frac{1}{(-\lambda)} - \frac{e^{-\lambda t}}{(-2\lambda)} \right) \right]$$

$$F_T(t) = 1 - \frac{e^{-\lambda t}}{2} ; t \geq 0.$$

$$(2) \quad T < 0$$

$$T = T_1 - T_2$$

$$\text{let assume } T' = T_2 - T_1$$

then

$$P(T > t) = P(T' > t)$$

due to independence of T_1, T_2

and also due to symmetry.

$$\begin{aligned} \text{Now, } P(T < t) &= P(T' > -t) \\ &= P(T > -t) \\ &= 1 - P(T \leq (-t)) \end{aligned}$$

$$P(T < t) + P(T \leq (-t)) = 1.$$

So,

$$F_T(t) + F_T(-t) = 1$$

Thus,

$$F_T(-t) = 1 - F_T(t)$$

$$\text{as for } t \geq 0, F_T(t) = 1 - \frac{e^{-\lambda t}}{2}.$$

$$\text{So, } F_T(-t) = \frac{e^{-\lambda t}}{2}; \quad t \geq 0$$

or

$$F_T(q) = \frac{e^{-\lambda q}}{2}; \quad q \leq 0.$$

An

$$(3) \quad F_T(t) = \begin{cases} 1 - \frac{e^{\lambda t}}{2}; & t \geq 0 \\ \frac{e^{\lambda t}}{2}; & t < 0 \end{cases}$$

$$f_T(t) = \frac{d}{dt}(F_T(t))$$

$$= \begin{cases} \frac{\lambda e^{-\lambda t}}{2}; & t \geq 0 \\ \frac{\lambda e^{\lambda t}}{2}; & t < 0 \end{cases}$$

so,

$$f_T(t) = \left(\frac{\lambda}{2}\right) e^{-|\lambda|t}$$

\neq

\checkmark

$\cancel{\checkmark}$

(4) It is Laplace distribution
pdf where $\mu = 0$ & $b = 1$.

Laplace Pdf :

$$f(x | \mu, b) = \frac{1}{2b} e^{-\frac{|x-\mu|}{b}}$$

$$Q \cdot T = 8 \cdot 1^7 \Rightarrow$$

(1) Let us consider the waiting time of Atsi for the next train to be T minutes where $T \in [0, 10]$)

Now as the distribution of arrival of Atsi is uniform, values taken by T are also uniformly distributed between 0 to 10.

So, T is a RV in range $[0, 10]$.

Then,

$$\text{mean waiting time} = E[T]$$

$$\text{so, } E[T] = \int_0^{10} t \cdot \left(\frac{1}{10}\right) dt$$

$$= \frac{1}{20} \left[t^2 \right]_0^{10}$$

$$= \cancel{\frac{100}{20}}$$

$$= 5 \text{ min} \quad \checkmark$$

$$(2) \text{ Answer} = P(T \geq 6+3 \mid T \geq 6)$$

$$= \frac{P((T \geq 9) \cap (T \geq 6))}{P(T \geq 6)}$$

$$= \frac{P(T \geq 9)}{P(T \geq 6)}$$

$$\frac{10-9}{10-6}$$

$$\frac{10-9}{10-6}$$

$$\frac{1}{10}$$

$$\frac{4}{10}$$

$$= Y_4 \text{ Ans}$$

(3) Let the time after which the train arrives after Atsi reach the station be T .

Now it's given that T' is exponential & $E[T'] = 10 \text{ min}$ and where T' is time of arrival of second train after 1st train left.

Since, exponential distribution is memoryless. Atsi's waiting time i.e. T will also be exponential with same mean $E[T] = 10 \text{ min}$.

Let λ be parameter of distribution.

$$E[T] = \infty \int_0^{\infty} t \cdot \lambda e^{-\lambda t} dt = 10$$

$$\Rightarrow \left[-t e^{-\lambda t} \right]_0^\infty + \int_0^\infty e^{-\lambda t} dt = 10$$

$$\Rightarrow \left[\frac{e^{-\lambda t}}{-\lambda} \right]_0^\infty = 10$$

$$\Rightarrow \frac{1}{\lambda} = 10 \Rightarrow \lambda = \frac{1}{10} = 0.1$$

Hence, Atsi's waiting time \sim Exponential (0.1)

Q.8 = sol^n

(1) Claim : For a poisson process with parameter λ , the time from a new arrival is exponentially distributed with same parameter λ .

Proof :-

Let for a poisson process (λ) an event occur at time t .

Assume exactly after t unit time from now another event occur. i.e. at ' $t + \frac{1}{2}t$ ' time.

Required to show : $f(t) = \lambda e^{-\lambda t}$.

where $\lambda' = \lambda$.

where T is RV for waiting time.

Proof :-

$$P(T > a) = P(\text{no. of arrival at time } t = \text{no. of arrival at time } t + a)$$

$$= P(\text{no. of arrival in time interval } [t, t+a] = 0)$$

$$= [(\lambda a)^0 e^{-\lambda a}] / (0)!$$

$$\text{So, } P(T > a) = e^{-\lambda a}$$

So,

$$P(T \leq t) = 1 - P(T > t)$$

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

Differentiating both side w.r.t t ;

$$f(t) = \frac{d}{dt} \left[1 - e^{-\lambda t} \right] \sim \text{exponential}(\lambda).$$

Hence, proved that waiting time
for a poisson distribution $\text{P}(t)$ is
Exponentially distributed (λ')

$$\lambda' = \lambda$$

(2)

Claim : The waiting time for a Bernoulli process is geometric.

Proof :-

For a bernoulli process; [RV 'X']

$$P(x) = \begin{cases} p &; x=1 \text{ i.e. arrived} \\ 1-p &; x=0 \text{ i.e. not arrived} \\ 0 &; \text{otherwise} \end{cases}$$

Let 'Y' be RV which shows the waiting time.

$Y=y$ implies that no animal occurred till $(y-1)^{\text{th}}$ time unit & arrived at y^{th} unit time.

So,

$$P(Y=y) = \begin{cases} (1-p)^{(y-1)} p &; y=1, 2, 3, \dots \\ 0 &; \text{otherwise} \end{cases}$$

Thus, we can see $f_Y(y) = P(Y=y)$ is a geometric distribution.

Hence, our claim is correct.

(3) Claim :- An exponential distribution
can be obtained as a
limit of a geometric Distribution.

Proof :-

As we know , Geometric
Distribution is

$$f_x(x) = p(1-p)^x ; x \geq 0$$

and can be interpreted as
number of failures in a
sequence of Bernoulli trials until
the 1st success.

Now, say p becomes very small
& let $n \times p = \lambda$; n is
total trials & $n \rightarrow \infty$.

$$\text{So, } P(X=x) = f_X(x) = \left(\frac{\lambda}{n}\right) \left(1 - \frac{\lambda}{n}\right)^{n-x}$$

Now when limit $n \rightarrow \infty$, $\sum_{x=0}^{\infty} f_X(x) = ?$

$$\lim_{n \rightarrow \infty} \sum_{x=0}^{\infty} \left(\frac{\lambda}{n}\right) \left(1 - \frac{\lambda}{n}\right)^{x + \frac{1}{n} \cdot n + \frac{1}{n} \cdot x}$$

~~$$= \lim_{n \rightarrow \infty} \sum_{x=0}^{\infty} (\lambda) \left(1 - \frac{\lambda}{n}\right)^{\frac{(n)(x)}{n}} \left(\frac{1}{n}\right)$$~~

~~$$= \lim_{n \rightarrow \infty} \sum_{y=0}^{\infty} (\lambda) \left(1 - \frac{y}{n}\right)^{\frac{n \cdot y}{n}} \left(\frac{1}{n}\right)$$~~

(Assumed)

$$y = \frac{x}{n}$$

Therefore,

We can't write

$$\lim_{n \rightarrow \infty} \sum_{x=0}^{\lambda} f(x) = \lim_{n \rightarrow \infty} \sum_{y=0}^{\left(\frac{n}{\lambda}\right) \cdot y \cdot \lambda} (\lambda) \left(1 - \frac{\lambda}{n}\right)^{\frac{(n)}{n}} \left(\frac{1}{n}\right)$$

$$\text{where, } y = \frac{x}{\lambda}.$$

\Rightarrow

$$\lim_{n \rightarrow \infty} \sum_{x=0}^{\lambda} f(x) =$$

$$\int_0^{\lambda} \lim_{n \rightarrow \infty} \left(1 - \frac{\lambda}{n}\right)^{\frac{n}{n} \cdot y \cdot \lambda} dy$$

$$= \int_0^{\lambda} \lambda \cdot e^{\lim_{n \rightarrow \infty} \frac{n}{n} y \cdot \lambda \cdot (-\frac{1}{n})} dy$$

$$= \int_0^{\lambda} \lambda \cdot e^{-y\lambda} dy$$

Therefore

for $n \rightarrow \infty$ & $p < 1$,

$$F_X(\lambda) = \int_0^{\lambda} \lambda \cdot e^{-y\lambda} dy = 1 - e^{-\lambda}$$

So, $f_X(\lambda) = \lambda e^{-\lambda}$ i.e. exponential distribution.

$$8 \cdot 9 = 801^n \quad (1)$$

$$(a) R(x; \sigma) = \frac{x}{\sigma^2} e^{-\frac{x^2}{2\sigma^2}}; x > 0$$

$$\text{mean} = \int_0^\infty x \cdot R(x; \sigma) dx = E$$

$$E = \int_{-\infty}^\infty \left(\frac{x^2}{\sigma^2}\right) e^{-\frac{x^2}{2\sigma^2}} dx$$

$$\text{say } x = \sigma y \text{ then } dx = \sigma dy$$

$$so, E = \int_0^\infty y^2 e^{-y^2/2} dy = \sigma \int_0^\infty y^2 e^{-y^2/2} dy$$

$$\text{say, } y^2 = 3 \Rightarrow dy = dz$$

$$\Rightarrow dy = \frac{dz}{2\sqrt{3}}$$

$$so, E = \int_0^\infty 3 e^{-3/2} \frac{dz}{2\sqrt{3}}$$

$$E = \frac{\pi}{2} \int_0^\infty \sqrt{3} e^{-z^2/2} dz$$

$$E = \sigma \sqrt{2} \int_0^{\infty} x \Gamma\left(\frac{3}{2}\right) e^{-\frac{x^2}{2}} dx$$

$$E = \sigma \sqrt{2} \underbrace{\Gamma\left(\frac{3}{2}\right)}_{\text{Gamma function}}$$

$$E = \sigma \sqrt{2} \frac{1}{2} \Gamma\left(\frac{1}{2}\right) \quad \left\{ \begin{array}{l} \Gamma\left(\frac{1}{2}x + 1\right) \\ = d \Gamma\left(x\right). \end{array} \right.$$

$$E[n] = E = \sigma \sqrt{\frac{\pi}{2}} \text{ Am}$$

$$(b) \text{ Variance } = ? = E[x^2] - (E[x])^2$$

Finding $E[x^2]$ $(x \geq 0)$

$$E[x^2] = \int_{-\infty}^{\infty} x^2 \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{x^2}{2\sigma^2}} dx$$

$$E[x^2] = \int_0^{\infty} x^2 \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{x^2}{2\sigma^2}} dx$$

$$\text{Let } \frac{x^2}{2\sigma^2} = y$$

\Rightarrow

$$xdx = \sigma^2 dy$$

$$\text{So, } E[x^2] = \int_0^\infty 2y e^{-\frac{y}{2}} \sigma^2 dy$$

$$E[x^2] = 2\sigma^2 \int_0^\infty y e^{-\frac{y}{2}} dy$$

$$E[x^2] = 2\sigma^2 \Gamma(2) = 2\sigma^2$$

because $\Gamma(2) = 1! = 1$.

$$\text{So, Variance} = 2\sigma^2 - \left(\frac{\sigma\sqrt{\pi}}{2}\right)^2$$

$$= \sigma^2 \left(2 - \frac{\pi}{2}\right) \cdot \text{Ans}$$

\approx

(c) Mode of distribution = ?

Mode is the value of x at which pdf is max.

$$-\frac{x^2}{2\sigma^2}$$

$$R(x; \sigma) = \frac{x}{\sigma^2} c ; (x \geq 0)$$

for max/min: $\frac{d}{dx} R(x; \sigma) = 0$.

$$\Rightarrow \frac{d}{dx} \left(\frac{x}{\sigma^2} e^{-x^2/2\sigma^2} \right) = 0.$$

$$\Rightarrow -\frac{x^2}{\sigma^2} + \frac{x}{\sigma^2} e^{-x^2/2\sigma^2} = 0$$

$$\Rightarrow \frac{e^{-x^2/2\sigma^2}}{\sigma^2} + \frac{x}{\sigma^2} e^{-x^2/2\sigma^2} = 0$$

$$\Rightarrow \frac{e^{-x^2/2\sigma^2}}{\sigma^2} \left[1 - \frac{x^2}{\sigma^2} \right] = 0.$$

$$\Rightarrow \frac{e^{-x^2/2\sigma^2}}{\sigma^2} = 0 \quad \text{or} \quad x^2 = \sigma^2$$

↓ ↓

at $x \rightarrow \infty$

$x = \sigma$ ($x > 0$).
($\sigma > 0$)

at $x \rightarrow \infty$

$$\frac{d^2}{dx^2} R(x; \sigma) \rightarrow 0.$$

at $x = \sigma$

$$\frac{d^2}{dx^2} R(x; \sigma) < 0 \quad \text{i.e. maximum}$$

at $x = \sigma$

Hence, mode $= \sigma$,

Ans

$$0.9 = (2) \quad \text{so} \gamma = ?$$

$$C(x; \gamma) = -\frac{1}{\pi \gamma \left[1 + \left(\frac{x - x_0}{\gamma} \right)^2 \right]}$$

$$(a) \text{ mean} = E[x] = \int_{-\infty}^{+\infty} x d_n$$

$$\frac{1}{\pi \gamma \left[1 + \left(\frac{x - x_0}{\gamma} \right)^2 \right]}$$

$$\text{Let } \frac{x - x_0}{\gamma} = z$$

$$dx = \gamma dz$$

$$\text{So, } E[x] = \int_{-\infty}^{\infty} x \left(\gamma z + x_0 \right) dz$$

$$\frac{1}{\pi \gamma \left[1 + z^2 \right]}$$

$$\pi E[x] = \int_{-\infty}^{\infty} \frac{\gamma z dz}{1+z^2} + \int_{-\infty}^{\infty} \frac{x_0 dz}{1+z^2}$$

$$\pi E[x] = \frac{\gamma}{2} \left[\ln(1+z^2) \right]_{-\infty}^{\infty} + \dots$$

$$x_0 \left[\tan^{-1} z \right]_{-\infty}^{\infty}$$

$$\text{Therefore, } E[X] = \frac{1}{2\pi} \left[\ln(1+z^2) \right]_{-\infty}^{+\infty} + x_0 \pi$$

$$\text{Since, } \left[\ln(1+z^2) \right]_{-\infty}^{+\infty} = \ln(\infty) - \ln(\infty) \\ = \infty - \infty \\ (\text{undefined})$$

Hence, mean for Cauchy's distribution is undefined and does not exist.

$$E((z^2 + 1)^{-1}) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{1}{1+x^2} dx$$

(b) Characteristic function of

$$C(x; \gamma) = \frac{1}{\pi \gamma \left[1 + \left(\frac{x - x_0}{\gamma} \right)^2 \right]}$$

$$\begin{aligned} \text{i} \quad \phi_x(t; x_0, \gamma) &= E[e^{ix_0 t}] \\ &= \int_{-\infty}^{\infty} e^{ix_0 t} C(x; \gamma) dt \end{aligned}$$

$$\phi_x(t; x_0, \gamma) = e^{ix_0 t - \gamma t^2}$$

i) $t \rightarrow 0^+$ ($t = 0 + h$, $h > 0$ & $h \ll 1$)

$$\begin{aligned} \frac{d}{dt} \phi_x(t; x_0, \gamma) &= \frac{d}{dh} \left(e^{ix_0 t - \gamma h^2} \right) \Big|_{h=0} \\ &= (ix_0 - \gamma) e^{(ix_0 - \gamma)h} \Big|_{h=0} \\ &= ix_0 - \gamma \end{aligned}$$

(ii) $t = 0^-$ ($t = 0 - h$, $h > 0$, $h \ll 1$, $|t| = h$)

$$dt = -dh$$

$$\frac{d}{dt}(\phi_x(t; x_0, \gamma)) = \frac{d}{dt} \left(e^{i x_0 (-h)} - \gamma^h \right) \Big|_{h=0}$$

$$= -\frac{d}{dh} \left(e^{i x_0 (-h)} - \gamma^h \right) \Big|_{h=0}$$

$$= -(-ix_0 - \gamma)$$

$$= ix_0 + \gamma$$

Hence, derivative does not exist.

* As we know mean is the 1st derivative of characteristic function at $t=0$, this shows that mean doesn't exist.

(C) The Cauchy distribution doesn't have finite moments of any order.

Higher even powered raw moments converge to ∞ while odd powered raw moments, are undefined because their value are essentially equivalent to $\infty - \infty$, since the two half of the integral diverge & have opposite sign, we can also say that central moments & standardized moments are undefined, since they are all based on mean & mean is undefined.

The variance which is the second central moment, is likewise non-existent.

As variance is the expected square of the deviation from mean, it is not defined, mean is not defined.

And .

(c) remaining:-

The consequence undefined variance
is that central limit theorem
doesn't apply & certain parameter
estimation techniques will therefore
breakdown

$$(Q, 10 = 50^\circ)$$

Say the required rectangle have points $P_1(x_0, y_0)$, $P_2(x_1, y_1)$, $A(x_0, y_1)$ & $B(x_1, y_0)$.

Let F be an event in which rectangle fits completely in the circle. Due to symmetry, the probability of F taking first point from 1st quadrant will be equal to probability of F taking 1st point (P_1) from any quadrant.

Let us suppose, P_1 falls on the quarter circle in 1st quadrant. We know,

$$y_0 = \sqrt{1-x_0^2} \quad \text{--- (1)}$$

For rectangle to be fully inside the circle :-

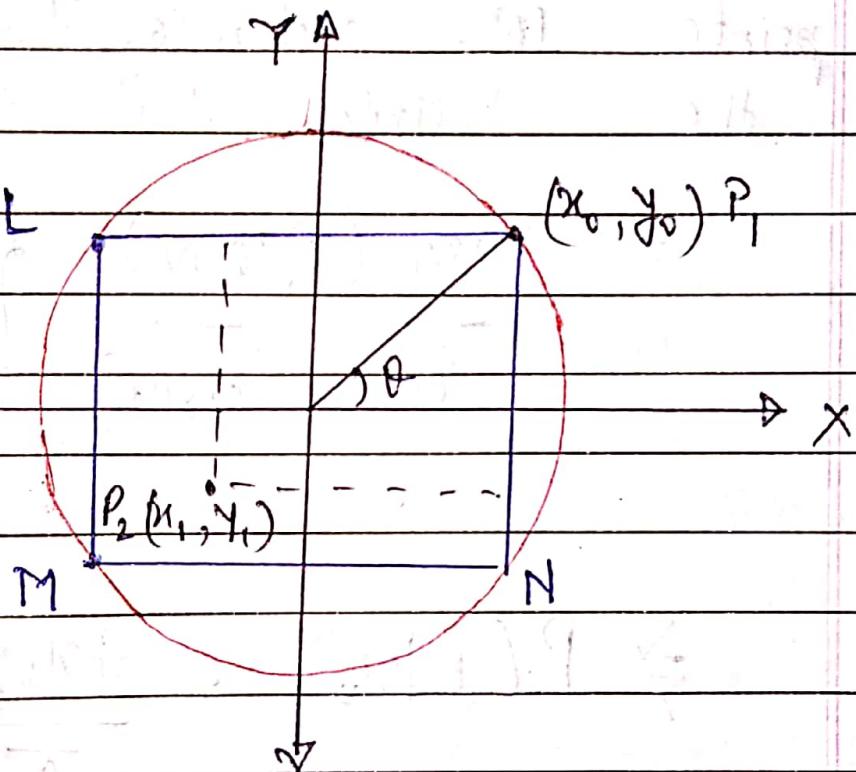
$$x_0^2 + y_1^2 \leq 1 \quad \text{and} \quad x_1^2 + y_0^2 \leq 1$$

because pt. A & B should be inside/on the circle

$$(1-y_0^2) + y_1^2 \leq 1 \quad \& \quad x_1^2 + (1-y_0^2) \leq 1$$

$$\Rightarrow y_1^2 \leq y_0^2 \quad \& \quad x_1^2 \leq x_0^2$$

$$\Rightarrow |y_1| \leq |y_0| \quad \& \quad |x_1| \leq |x_0|$$



$\text{Area } (LP_1NN) = \text{Area formed by}$

$$|y_1| \leq |y_0| \quad \& \quad |x_1| \leq |x_0|$$

Hence, for the required rectangle ~~P₁AB₂B~~,
point P_2 should lie inside/on
rectangle ~~P₁AB₂B~~ P_1LMN

$P(F|x_0)$ is the probability that rectangle P_1LMN (which is asked in Question) falls inside the circle 'C' given the value of x_0 .

Therefore, the probability that no points of rectangle lies outside the circle is

$$= \frac{\text{Area } (P_1LMN)}{\text{Area } (\text{circle})} \quad \left\{ \begin{array}{l} \text{As } P_2(x_1, y_1) \\ \text{can be chosen} \\ \text{anywhere} \\ \text{inside } P_1LMN. \end{array} \right.$$

$$\Rightarrow P(F|x_0) = \frac{4x_0 y_0}{\pi}$$

taking $r = 1$, $x_0 = r \cos \theta$, $y_0 = r \sin \theta$

$$P(F|\theta) = \frac{4 \cos \theta \sin \theta}{\pi}$$

We also know that,

$$P[F/x_0 \mid 1^{\text{st}} \text{ quadrant}] = P[F/x_0 \mid \text{whole circle}]$$

$$\Rightarrow P(F) = \frac{\int_0^{\pi/2} P(F/\theta) d\theta}{\pi/2}$$

$$\Rightarrow P(F) = \frac{2}{\pi} \int_0^{\pi/2} \frac{4 \cos \theta \sin \theta}{\pi} d\theta$$

$$\Rightarrow P(F) = \frac{4}{\pi^2} \int_0^{\pi/2} \sin 2\theta d\theta$$

$$\begin{aligned} \Rightarrow P(F) &= \frac{4}{\pi^2} \left[-\frac{\cos 2\theta}{2} \right]_0^{\pi/2} \\ &= \frac{4}{\pi^2} \left[\left(-\frac{(-1)}{2} \right) - \left(-\frac{1}{2} \right) \right] \end{aligned}$$

$$\Rightarrow P(F) = \frac{4}{\pi^2} \quad \text{Ans}$$