

وزارة التعليم العالي والبحث العلمي

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المرحلة الثالثة

Chapter Four

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CHAPTER FOUR

Expectation and Variance:

Def: Let x be a. r. v either d.r.v. or c.r.v. " $E(x)$ " is called the "Expectation of x " or "expected Value of x " or "mean of x " And denoted by μ .

Defined as follows:

1- $E(x) = \sum_{\forall x} x f(x)$, when x is a d.r.v.

2- $E(x) = \int_{-\infty}^{\infty} xf(x) . dx$ when x is ac.r.v.

Note: 1- the value of $E(x)$ is constant.

2- $E(x)$ exists if $\sum_{\forall x} |x| f(x) < \infty$, when x is d.r.v.

3- $E(x)$ exists if $\int_{-\infty}^{\infty} |x| f(x) dx < \infty$, when x is c.r.v.

Ex "1": Given p.m.f $f(x) = \begin{cases} \frac{\binom{3}{x}}{8} & \text{for } x = 0, 2, 3, 4 \\ 0 & \text{o.w} \end{cases}$

Find $E(X)$?

Sol: $E(x) = \sum_{x=0}^3 x f(x)$

X	$f(x) = \frac{\binom{3}{x}}{8}$	$x.f(x)$
0	$\frac{\binom{3}{0}}{8} = \frac{1}{8}$	0
1	$\frac{\binom{3}{1}}{8} = \frac{3}{8}$	$\frac{3}{8}$

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2	$\binom{3}{2} / 8 = \frac{3}{8}$	$\frac{6}{8}$
3	$\binom{3}{3} / 8 = \frac{1}{8}$	$\frac{3}{8}$
	$\sum_{x=0}^3 f(x) = 1$	$\sum_{x=0}^3 x f(x) = \frac{12}{8}$

$$\therefore E(x) = \sum_{x=0}^3 x f(x) = \frac{3}{2}$$

H.W: given a.p. M.f $f(x) = \begin{cases} \frac{x}{15} & \text{for } x = 1, 2, 3, 4, 5 \\ 0 & \text{o.w.} \end{cases}$

Find the expected Value of x.

Ex 2: Given a p.d.f $f(x) = \begin{cases} \frac{x}{8} & \text{for } 0 < x < 4 \\ 0 & \text{o.w.} \end{cases}$ Find $E(x)$?

Sol: $E(x) = \int_{-\infty}^{\infty} x f(x) dx = \int_0^4 x \frac{x}{8} dx = \int_0^4 \frac{x^2}{8} dx$

$$= \frac{1}{8} \frac{x^3}{3} \Big|_0^4 = \frac{1}{24} [64 - 0] = \frac{64}{24} = \frac{8}{3}$$

H.W: Given a.p. d.f $f(x) = \begin{cases} 3x^2 & \text{for } 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$

Find $E(x)$?

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← عظمى

✓ Ex: Given ap. d.f $f(x) = \begin{cases} \frac{1}{x} & \text{for } 0 < x < e \\ 0 & \text{o.w.} \end{cases}$ $+1 < x < e$ $-1 < x < 2.72 \sim e^1$

Dose $E(x)$ exist?

Sol: $E(x) = \int_1^e x \cdot \frac{1}{x} dx = \int_1^e dx = 1 - e$

$\therefore E(x)$ is exist.

✓ Ex: Given ap. d.f $f(x) = \begin{cases} \frac{1}{x^2} & x > 1 \\ 0 & \text{o.w.} \end{cases}$ $x > 1$

Dose $E(x)$ exist?

Sol: $E(x) = \int_1^{\infty} x \cdot \frac{1}{x^2} dx = \int_1^{\infty} \frac{1}{x} dx = \ln x \Big|_1^{\infty} = [\infty - 0] = \infty$

$\therefore E(x)$ is not exist.

Expectation of a function of x:

Def: Let x be ar.v. and let $g(x)$ be a function of x then

$E[g(x)] = \sum_{-\infty}^{\infty} g(x) f(x)$, if x is d.r.v.

$= \int_{-\infty}^{\infty} g(x) f(x) dx$, if x is c.r.v.

Ex "1": Given ap.m.f $f(x) = \begin{cases} \frac{x}{10} & \text{for } x = 1, 2, 3, 4 \\ 0 & \text{o.w.} \end{cases}$

find $E(x^2)$?

sol: $g(x) = x^2$

$E[g(x)] = \sum_{-\infty}^{\infty} g(x) f(x)$

$E[x^2] = \sum_{x=1}^4 (x)^2 \cdot \frac{x}{10} = \sum_{x=1}^4 \frac{x^3}{10} = \left[\frac{1}{10} + \frac{8}{10} + \frac{27}{10} + \frac{64}{10} \right]$

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$$= \frac{100}{10} = 10$$

✓ Ex"2": Given ap.d.f $f(x) = \begin{cases} \frac{x}{8} & \text{for } 0 < x < 4 \\ 0 & \text{o.w.} \end{cases}$

find $E(\sqrt{x})$?

sol: $g(x) = \sqrt{x} = x^{\frac{1}{2}}$

$$E[g(x)] = E(x^{\frac{1}{2}}) = \int_0^4 x^{\frac{1}{2}} \cdot \frac{x}{8} dx = \frac{1}{8} \int_0^4 x^{\frac{3}{2}} dx = \frac{1}{8} \cdot \frac{2}{5} x^{\frac{5}{2}} \Big|_0^4$$

$$= \frac{1}{20} \left[(\sqrt{4})^5 - 0 \right] = \frac{1}{20} [32] = \frac{32}{20}$$

Note: if $f(x)$ be a p. d.f of a r.v. x , then $E(b)=b$, where b is constant.

Proof: case "1": If x is a d.r.v. with p.m.f. $f(x)$

$$E(b) = \sum_{x} b f(x) = b \sum_{x} f(x) = b$$

Case "2": If x is a c.r.v. with p.d.f. $f(x)$.

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

Properties of Expectation :

Theorem "1" If x is ar.v. have ap.f. $f(x)$, and $E(x)$ exists.

Let $y=ax+b$, $a, b \in R$, then $E(y)=aE(x)+b$.

Sol: case : "1" If x is a c. r.v. With P.d.f $f(x)$

$$Y=g(x)=ax+b$$

$$E(y)=E[g(x)] = \int_{-\infty}^{\infty} g(x) f(x) dx$$

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$$E[ax + b] = \int_{-\infty}^{\infty} (ax + b) f(x) dx = \int_{-\infty}^{\infty} ax f(x) dx + \int_{-\infty}^{\infty} b f(x) dx$$

$$= a \int_{-\infty}^{\infty} x f(x) dx + b \int_{-\infty}^{\infty} f(x) dx = a E(x) + b$$

case "2": If x is ad.r.v. with p.m.f $f(x)$.

$$y = g(x) = ax + b$$

$$E(y) = E[g(x)] = \sum_{\tau} g(x) f(x)$$

$$E[ax + b] = \sum_{\tau} (ax + b) f(x) = \sum_{\tau} ax f(x) + \sum_{\tau} b f(x)$$

$$= a \sum_{\tau} x f(x) + b \sum_{\tau} f(x) = a E(x) + b$$

theorem "2": let x be a r.v. if $u(x)$ and $v(x)$ are two functions of x . then:

$$E[u(x) \mp v(x)] = E[u(x)] \mp E[v(x)]$$

proof: " case "1": If x is ad.r.v. with p.m.f $f(x)$ let $g(x) = u(x) \mp v(x)$

$$E[u(x) \mp v(x)] = E[g(x)] = \sum_{\tau} g(x) f(x)$$

$$= \sum_{\tau} [u(x) \mp v(x)] f(x) = \sum_{\tau} u(x) f(x) \mp \sum_{\tau} v(x) f(x)$$

$$= E[u(x)] \mp E[v(x)]$$

case "2": If x is ac.r.v. with p.d.f $f(x)$.

$$\text{let } g(x) = u(x) \mp v(x)$$

$$E[u(x) \mp v(x)] = E[g(x)] = \int_{-\infty}^{\infty} g(x) f(x) dx$$

$$= \int_{-\infty}^{\infty} [u(x) \mp v(x)] f(x) dx = \int_{-\infty}^{\infty} u(x) f(x) dx \mp \int_{-\infty}^{\infty} v(x) f(x) dx$$

$$= E[u(x)] \mp E[v(x)]$$

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Ex "1" Given ap.d.f $f(x) = \begin{cases} \frac{x+2}{18} & \text{for } -2 < x < 4 \\ 0 & \text{o.w.} \end{cases}$

find $E[2x^3 - 1]$ & $E[(x+2)^2]$

Sol: $E[2x^3 - 1] \Rightarrow 1. g(x) = 2x^3 - 1 \Rightarrow E[g(x)] = \int_{-2}^4 g(x) f(x) dx$

Or 2. $E[2x^3 - 1] = 2E(x^3) - 1$

$E(x^3) = \int_{-2}^4 x^3 \left(\frac{x+2}{18} \right) dx = \frac{1}{18} \int_{-2}^4 (x^4 + 2x^3) dx$

$= \frac{1}{18} \left[\frac{x^5}{5} + \frac{x^4}{2} \right]_{-2}^4 = \frac{1}{18} \left[\left(\frac{1024}{5} + \frac{256}{2} \right) - \left(\frac{-32}{5} + \frac{16}{2} \right) \right]$

$= \frac{1}{18} \left(\frac{1056}{5} + \frac{240}{2} \right) = \frac{1}{18} \left(\frac{1056}{5} + 120 \right)$

$E[(x+2)^2] \Rightarrow 1. g(x) = (x+2)^2 \Rightarrow E[g(x)] = \int_{-2}^4 g(x) f(x) dx$

Or 2. $E[x^2 + 4x + 4] = E(x^2) + 4E(x) + 4$

$= \int_{-2}^4 x^2 f(x) dx + 4 \int_{-2}^4 x f(x) dx + 4$

$= \dots$

H.W: 1 Given ap.d.f., $f(x) = \begin{cases} \frac{x}{2} & 0 < x < 4 \\ 0 & \text{o.w.} \end{cases}$

Find $E[(x+2)^2]$?

2. Given ap. m.f, $f(x) = \begin{cases} \frac{x}{10} & \text{for } x = 1, 2, 3, 4 \\ 0 & \text{o.w.} \end{cases}$ find $E[2x^3 - 1] E[(x-1)^3]$?

$(x-1)^3 = (x^3 - 3x^2 + 3x - 1)$

theorem "3": let x be ar.v.

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a. If $\exists(a)$ such that $p(x \geq a) = 1$, then $E(x) \geq a$.

b. If $\exists(b)$ such that $p(x \leq b) = 1$, then $E(x) \leq b$.

c. If $p(a \leq x \leq b) = 1$, then $a \leq E(x) \leq b$.

Proof: a. case "1": If x is ac.r.v. with p.d.f $f(x)$

$$P(x \geq a) = 1 \Rightarrow \int_a^{\infty} f(x) dx = 1$$

$$\therefore f(x) > 0 \quad \text{for } a \leq x < \infty$$

$$= 0 \quad \text{o.w.}$$

$$R_x = \{x: a \leq x < \infty\}$$

$$E(x) = \int_a^{\infty} x f(x) dx \geq \int_a^{\infty} a f(x) dx = a \int_a^{\infty} f(x) dx = a$$

$$\therefore E(x) \geq a$$

Case (2) if x is d.r.v. with p.m.f $f(x)$

$$E(x) = \sum_{x=a}^{\infty} x f(x) \geq \sum_{x=a}^{\infty} a f(x) = a \sum_{x=a}^{\infty} f(x) = a$$

$$\therefore E(x) \geq a.$$

b. Similary $\bullet f(a)$.

c. Case "1": If x is ad.r.v. have ap.m.f $f(x)$

$$P(a \leq x \leq b) = 1 \Rightarrow \sum_{x=a}^b f(x) = 1$$

$$\therefore f(x) > 0 \quad \text{for } a \leq x \leq b$$

$$= 0 \quad \text{o.w.}$$

$$E(x) = \sum_{x=a}^b x f(x) \geq \sum_{x=a}^b a f(x) = a \sum_{x=a}^b f(x) = a$$

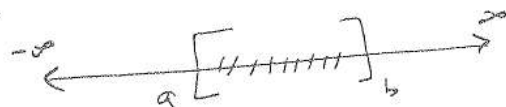
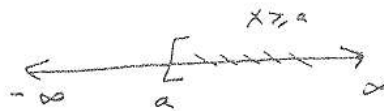
$$\therefore E(x) \geq a \dots \dots \dots (1)$$

$$E(x) = \sum_{x=a}^b x f(x) \leq \sum_{x=a}^b b f(x) = b \sum_{x=a}^b f(x) = b$$

$$\therefore E(x) \leq b \dots \dots \dots (2)$$

$\therefore$ by "1" & "2" we get $a \leq E(x) \leq b$

theorem 4": If $p(x \geq a) = 1$ and $E(x) = a$ then $p(x = a) = 1$ and $p(x > a) = 0$.



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proof: case "1": If x is ac.r.v. from ap.d.f $f(x)$

$$\therefore p(x \geq a) = 1 \Rightarrow \int_a^{\infty} f(x) dx = 1$$

$$\therefore f(x) > 0 \text{ for } a \leq x < \infty$$

$$= 0 \text{ ow}$$

$$E(x) = \int_a^{\infty} x f(x) dx = a = a.1 = a \int_a^{\infty} f(x) dx = \int_a^{\infty} a f(x) dx$$

$$\int_a^{\infty} x f(x) dx = a \int_a^{\infty} f(x) dx \Rightarrow \int_a^{\infty} x f(x) dx = \int_a^{\infty} a f(x) dx$$

..this inequality hold only when $x=a$

$$\text{i.e. } \{x > a\} = \phi \Rightarrow P(X > a) = 0$$

$$p(x \geq a) = p[(x = a) \cup (x > a)] = p(x = a) + p(x > a) \Rightarrow 1 = p(x = a) + 0 \Rightarrow p(x = a) = 1$$

case "2": If x is ad.r.v from ap.m.f $f(x)$

$$\therefore p(x \geq a) = 1 \Rightarrow \sum_{x=a}^{\infty} f(x) = 1$$

$$\therefore f(x) > 0 \text{ for } a \leq x < \infty$$

$$= 0 \text{ ow}$$

$$E(x) = \sum_{x=a}^{\infty} x f(x) = a = a.1 = a \sum_{x=a}^{\infty} f(x) = \sum_{x=a}^{\infty} a f(x)$$

$$\sum_{x=a}^{\infty} x f(x) = \sum_{x=a}^{\infty} a f(x)$$

This inequality hold only when $x=a$

$$\text{i.e. } \{x > a\} = \phi \Rightarrow p(x > a) = 0$$

$$p(x \geq a) = [(x = a) \cup (x > a)] = p(x = a) + p(x > a)$$

$$\Rightarrow 1 = p(x = a) + 0 \Rightarrow p(x = a) = 1$$

variance of random variable:

Def: let x be ar.v. the variance of x denoted by $v(x)$ or δ^2 is defined as

$$\therefore v(x) = E \{ [x - E(x)]^2 \}$$

$$\therefore E(x) = \mu, \text{ then } v(x) = E[(x - \mu)^2]$$

$$\text{note: since } (x - \mu)^2 \geq 0 \text{ then } E[(x - \mu)^2] \geq 0$$

$$\therefore v(x) \geq 0 \text{ always.}$$

properties of variance:

theorem "5" let x be ar.v., then $v(x) = E(x^2) - [E(x)]^2$

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Proof: $v(x) = E\{[x - E(X)]^2\}$

$$\therefore V(x) = E\{x^2 - 2E(x)x + [E(x)]^2\} \quad \{E(x) \text{ constant}\}$$

$$v(x) = E(x^2) - 2E(x)E(x) + [E(x)]^2$$

$$v(x) = E(x^2) - 2[E(x)]^2 + [E(x)]^2 \Rightarrow v(x) = E(x^2) - [E(x)]^2$$

$$\text{note: } 1. v(x) = 0 \Leftrightarrow E(x^2) = E[x]^2$$

$$2. \therefore v(x) \geq 0 \Rightarrow \therefore E(x^2) \geq [E(x)]^2$$

$$3. v(b) = 0, b \text{ is constant.}$$

Theorem "6" let x be ar.v. and $v(x)$ exist, If $y = ax + b$;

$$a, b \in R \text{ then } v(y) = a^2 v(x).$$

$$\text{Proof: } y = ax + b \Rightarrow E(y) = aE(x) + b$$

$$V(y) = E\{[y - E(y)]^2\} = E\{[(ax + b) - (aE(x) + b)]^2\}$$

$$= E\{[ax - aE(x)]^2\} = E\{a^2[x - E(x)]^2\}$$

$$= a^2 E\{[x - E(x)]^2\} = a^2 V(x)$$

Ex: Given ap.d.f $f(x) = \begin{cases} 3x^2 & \text{for } 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$

a. find $E(x)$ & $V(x)$. b. If $y = 1 - 2x$, then find $E(y)$, $V(y)$.

SOL.

$$a. E(x) = \int_0^1 x(3x^2) dx = 3 \frac{x^4}{4} \Big|_0^1 = \frac{3}{4}$$

$$v(x) = E(x^2) - [E(x)]^2$$

$$E(x^2) = \int_0^1 x^2(3x^2) dx = \frac{3}{5} x^5 \Big|_0^1 = \frac{3}{5}$$

$$\therefore v(x) = \frac{3}{5} - \left(\frac{3}{4}\right)^2 = \frac{48 - 45}{80} = \frac{3}{80}$$

$$b. \therefore y = (-2)x + 1 \Rightarrow \therefore E(y) = (-2)E(x) + 1$$

$$\therefore E(y) = (-2)\left(\frac{3}{4}\right) + 1 = \frac{-3}{2} + 1 = -\frac{1}{2}$$

$$v(y) = (-2)^2 v(x) = 4 \cdot \left(\frac{3}{80}\right) = \frac{12}{80}$$

H.w.: Given ap.d.f $f(x) = \begin{cases} 6x(1-x) & \text{for } 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$

If $y = 2 - 3x$, then find $E(y)$ & $v(y)$

Theorem "7" $v(x) = 0$ iff $\exists K$ where k is constant such that $p(x = k) = 1$

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$\Leftarrow$ sup pose that $p(x=k)=1, \text{ t.p } V(X)=0$

Poof: $f(x) > 0$ for $x=k$
 $=0$ o.w.

$\Rightarrow$ sup pose $V(X)=0$ by note.
 t.p $p(x=k)=1$

$$\therefore V(X)=0 \Rightarrow E\{X-E(X)\}^2 = 0$$

$$Y = [X-E(X)]^2 \Rightarrow E(Y)=0$$

$$\text{by th. "4"} \Rightarrow [E(X)=a \rightarrow p(X=a)=1]$$

$$\therefore p(Y=0)=1 \Rightarrow p\{[X-E(X)]^2=0\}=1$$

$$\therefore p\{X-E(X)=0\}=1 \Rightarrow \therefore p[X=E(X)]=1$$

$$\therefore X=k \Rightarrow E(X)=E(k)=k \Rightarrow \therefore p[X=k]=1$$

* Existence of Mean and Variance:

$E(X)$ exists iff $E(|X|) < \infty$

$\therefore V(X) = E(X^2) - [E(X)]^2 \Rightarrow \therefore V(X)$ exists iff $E(X)$ & $E(X^2)$ exist

Ex. "1": Given acauchy p.d.f $f(x) = \frac{1}{\pi(1+x^2)}$ for $-\infty < x < \infty$

Show that $E(X)$ dose not exist?

$$\begin{aligned} \text{Sol.: } E(|X|) &= \int_{-\infty}^{\infty} |x| \frac{1}{\pi(1+x^2)} dx = 2 \int_0^{\infty} x \frac{1}{\pi(1+x^2)} dx \\ &= \frac{1}{\pi} \int_0^{\infty} \frac{2x}{1+x^2} dx = \frac{1}{\pi} \ln(1+x^2) \Big|_0^{\infty} = \frac{1}{\pi} [\ln \infty - \ln 1] \neq \infty \end{aligned}$$

$E(|X|) \neq \infty \Rightarrow E(X)$ dose not exist

Exercises: 1. Let x be a c.r.v. have a p.d.f $f(x)$ where

$f(x) > 0$ for $0 < x < b < \infty$, Show that $E(X) = \int_0^b [1-F(x)]dx$
 $= 0$ o.w.

Hint.: $f(x) = \frac{dF(x)}{dx} \Leftrightarrow f(x)dx = dF(x)$

2. If x is d.r.v. have p.M.f $f(x) > 0$ for $x = -1, 0, 1$
 $= 0$ o.w.

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a. If $f(0) = \frac{1}{2}$, Find $E(x^2)$. $\sum f(x) = 1 \Rightarrow f(-1) + f(0) + f(1) = 1$

b. If $f(0) = \frac{1}{2}$, and $E(x) = \frac{1}{6}$, Find $f(-1)$, $f(1)$

3. Given a p.d.f $f(x) = \begin{cases} 1 - |x| & \text{for } |x| < 1 \\ 0 & \text{o.w.} \end{cases}$

a. Find $E(x)$ & $V(x)$, b. If $y = 2 - 3x$, find $E(y)$ & $V(y)$.

Hint $f(x) = \begin{cases} 1 - |x| & -1 < x < 1 \\ 0 & \text{o.w.} \end{cases} = \begin{cases} 1 + x & -1 < x < 0 \\ 1 - x & 0 \leq x < 1 \\ 0 & \text{o.w.} \end{cases}$

Moments of Random Variables:

Def.: Let x be a r.v. either d.r.v. or c.r.v., let $k \in \mathbb{I}^+$ then $E(x^k)$ is called "the k^{th} moment of x " or "the moment of order k of x ".

when $k = 1 \Rightarrow E(x^1) = 1^{\text{st}}$ moment of $x = \mu$

$E(x^2) = 2^{\text{nd}}$ moment of x

Note: $E(x^k)$ exists iff $E(|x|^k) < \infty$

Theorem "8": If $E(x^k)$ exists then $E(x^j)$ exists, $j < k$ and $j, k \in \mathbb{I}^+$

Proof: case "1" If x is c.r.v. with p.d.f $f(x)$

$\therefore E(x^k)$ exists $\Rightarrow \therefore E(|x|^k) < \infty$

T.p $E(x^1)$ exists, T.p $E(|x|^1) < \infty$

$E(|x|^j) = \int_{-\infty}^{\infty} |x|^j f(x) dx$

$E(|x|^j) = \int_{|x| \leq 1} |x|^j f(x) dx + \int_{|x| > 1} |x|^j f(x) dx$

$\int_{-1}^1 |x|^j dx$

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2nd central moment of R.V.X is equal to $V(x)$.

✓ Ex.: Let x be a r.v.s.t $E(x) = 1$, $E(x^2) = 2$ and $E(x^3) = 5$. Find the 3rd central moment of x .

Sol.: $E[(x-\mu)^3] = E\{x^3 - 3\mu x^2 + 3\mu^2 x - \mu^3\}$
 $= E(x^3) - 3\mu E(x^2) + 3\mu^2 E(x) - \mu^3$
 $= 5 - 3 \cdot 1 \cdot 2 + 3 \cdot 1 \cdot 1 - 1 = 1$

Exercises: 1. If $x \sim$ uniform (a, b) ; $a, b \in \mathbb{R}$. Find the value of 1st central moment of x and also find the 2nd central moment of x .

2. Let $\mu = E(x)$ and $\delta^2 = V(x)$ show that $E[(x-\mu)^4] \geq \delta^4$
i.e. 4th central moment of x is greater than or equal the square of variance of x .

Moment Generating Function (M.g.f).

Def.: A moment generating function (M.g.f) of a r.v. x is a function that determines all moments of x , denoted by $M_x(t)$. suppose that $t \in (-h, h)$, $h > 0$.

If $E[e^{tx}]$ exists $\forall t \in (-h, h)$ then $M_x(t) = E[e^{tx}]$, $-h < t < h$

There are two cases of $M_x(t)$.

Case "1": If x is a d.r.v. from a p.m.f $f(x)$

$$M_x(t) = E[e^{tx}] = \sum_{x \in \mathbb{N}} e^{tx} f(x)$$

Case "2": If x is a c.r.v. have a p.d.f $f(x)$

$$M_x(t) = E[e^{tx}] = \int_{-\infty}^{\infty} e^{tx} f(x) dx$$

✓ Ex.: Given a p.d.f $f(x) = \begin{cases} e^{-x} & \text{for } x > 0 \\ 0 & \text{o.w} \end{cases}$

Find $M_x(t)$ and sketch its graph.

Sol.: $M_x(t) = E(e^{tx}) = \int_0^{\infty} e^{tx} \cdot e^{-x} dx = \int_0^{\infty} e^{-(1-t)x} dx$

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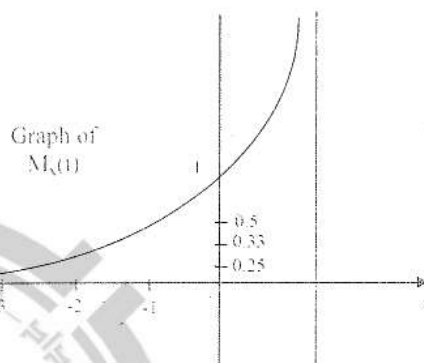
This integration exist only when $(1-t) > 0$, i.e. $t < 1$

$$M_x(t) = \frac{-1}{1-t} e^{-(1-t)x} \Big|_0^\infty = \frac{-1}{1-t} [e^{-\infty} - e^0] = \frac{1}{1-t}$$

$$M_x(t) = \frac{1}{1-t} \quad \text{for } t < 1$$

t	$M_x(t)$
1	∞
0	1
-1	0.5
-2	0.33
-3	0.25
$-\infty$	0

Graph of $M_x(t)$



Ex.: Given a p.m.f.

$$f(x) = \begin{cases} \frac{x}{15} & \text{for } x = 1, 2, 3, 4, 5 \\ 0 & \text{o.w} \end{cases}$$

Find $M_x(t)$.

$$\text{Sol.: } M_x(t) = E(e^{tx}) = \sum_{x=1}^5 e^{tx} f(x) = \sum_{x=1}^5 \frac{x}{15} e^{tx}$$

$$= \frac{1}{15} \sum_{x=1}^5 x e^{tx} = \frac{1}{15} [1 \cdot e^t + 2 \cdot e^{2t} + 3 \cdot e^{3t} + 4 \cdot e^{4t} + 5 \cdot e^{5t}] \quad \text{for } -\infty < t < \infty$$

Theorem "9": Let x be a r.v. have a M.g.f. $M_x(t)$, then

$$M_x(0) = 1, M'_x(0) = E(x), M''_x(0) = E(x^2), M'''_x(0) = E(x^3), \dots,$$

$$M^{(k)}_x(0) = E(x^k)$$

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

by Macclaurin series

$$e^x = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \dots + \frac{x^k}{k!} + \dots$$

$$\therefore e^{tx} = 1 + tx + \frac{t^2 x^2}{2!} + \frac{t^3 x^3}{3!} + \dots + \frac{t^k x^k}{k!} + \dots$$

$$M_x(t) = E[1 + tx + \frac{t^2 x^2}{2!} + \frac{t^3 x^3}{3!} + \dots + \frac{t^k x^k}{k!} + \dots]$$

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

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$$M_x(t=0) = M_x(0) = 1$$

$$* M'_x(t) = \frac{dM_x(t)}{dt} = E(x) + \frac{2t}{2!} E(x^2) + \frac{3t^2}{3!} E(x^3) + \dots + \frac{kt^{k-1}}{k!} E(x^k) + \dots$$

$$M'_x(0) = E(x)$$

$$* M''_x(t) = E(x^2) + \frac{6t}{3!} E(x^3) + \dots + \frac{k(k-1)t^{k-2}}{k!} E(x^k) + \dots$$

$$M''_x(0) = E(x^2)$$

Simillary we can find $E(x^3), E(x^4), \dots, E(x^k), \dots$

Note: $M_x(t) = M_x(0) + tM'_x(0) + \frac{t^2}{2!} M''_x(0) + \dots + \frac{t^k}{k!} M^{(k)}_x(0)$

This series is called the M. g.f. by Macclaurin series.

Ex.: Given $M_x(t) = \frac{1}{1-2t}, t < \frac{1}{2}$ Find $E(x)$ and $V(x)$

Sol.: $M_x(t) = (1-2t)^{-1}, t < \frac{1}{2}$

$$M'_x(t) = -(1-2t)^{-2} \cdot (-2) = 2(1-2t)^{-2}$$

$$\therefore E(x) = M'_x(0) = 2(1-0)^{-2} = 2$$

$$V(x) = E(x^2) - [E(x)]^2$$

$$M''_x(t) = 8(1-2t)^{-3}$$

$$\therefore E(x^2) = M''_x(0) = 8(1-0)^{-3} = 8$$

$$\therefore V(x) = 8 - 2^2 = 4 \geq 0$$

Theorem 10: Let x be a r.v. have a M.g.f $M_x(t)$. If $Y = ax + b, a, b, \in R$ then $M_y(t) = e^{bt} \cdot M_x(at)$

Proof: $Y = ax + b$

$$M_y(t) = E(e^{tx}) \text{ (by def.)} = E[e^{t(ax+b)}] = E[e^{(at)x + bt}]$$

$$M_y(t) = [e^{(at)x} \cdot e^{bt}] = e^{bt} E[e^{(at)x}] = e^{bt} \cdot M_x(at)$$

$$\text{Since } M_x(t) = E(e^{tx}) \Rightarrow M_x(at) = E[e^{(at)x}]$$

Ex: Given $M_x(t) = \frac{1}{1-3t}$ for $t < \frac{1}{3}$, If $y = 1-2x$, find $M_y(t)$.

Sol: $y = (-2)x + 1, a = -2, b = 1$

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By th. (10) $\Rightarrow M_y(t) = e^{bt} M_x(t) = e^t M_x(-2t)$

$$\therefore M_x(t) = \frac{1}{1-3t} \text{ for } t < \frac{1}{3}$$

$$M_x(-2t) = \frac{1}{1-3(-2t)} = \frac{1}{1+6t} \text{ for } t > \frac{-1}{6}$$

$$M_y(t) = e^t \cdot \frac{1}{1+6t} \text{ for } t > \frac{-1}{6}$$

The bound of probability:

Theorem "11": ((Markov inequality))

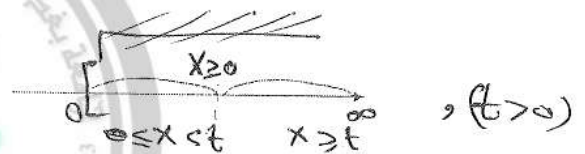
If x is a r.v. and if $P(x \geq 0) = 1$ then $P(x \geq t) \leq \frac{E(x)}{t}$ for $t > 0$

Proof: case "1": if x is a c.r.v with a p.d.f $f(x)$

$$\because p(x \geq 0) = 1 \Rightarrow \int_{-\infty}^{\infty} f(x) dx = 1$$

$$f(x) > 0 \text{ for } x \geq 0$$

O.W



$$E(x) = \int_{-\infty}^{\infty} x f(x) dx = \int_{-\infty}^t x f(x) dx + \int_t^{\infty} x f(x) dx \geq \int_t^{\infty} x f(x) dx \geq \int_t^{\infty} t f(x) dx = t \int_t^{\infty} f(x) dx = t P(x \geq t)$$

$$\therefore E(x) \geq t \int_t^{\infty} f(x) dx \Rightarrow E(x) \geq t P(x \geq t)$$

$$E(x) \geq t P(x \geq t)$$

$$\therefore P(x \geq t) \leq \frac{E(x)}{t}$$

Case "2": If x is a d.r.v. with a p.m.f. $f(x)$ by the same method.

Theorem "12": "Chebyshev inequalities"

Let x be a r.v. where $V(x)$ exists, then:

$$1. P(|x - \mu| \geq t) \leq \frac{V(x)}{t^2}, \quad t > 0, \quad 2. P(|x - \mu| < t) \geq 1 - \frac{V(x)}{t^2}, \quad t > 0,$$

Note: 1. $\frac{V(x)}{t^2}$ is called the upper bound of $P(|x - \mu| \geq t)$

2. $1 - \frac{V(x)}{t^2}$ is called the lower bound of $P(|x - \mu| < t)$

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Proof: 1. $V(x) = E[(x - \mu)^2] \geq 0$

Let $y = (x - \mu)^2 \geq 0 \Rightarrow E(y) = E[(x - \mu)^2] = V(x) \geq 0$

$\therefore$ by Markov inequality, we get

$$P(y \geq t^2) \leq \frac{E(y)}{t^2} \Rightarrow P[(x - \mu)^2 \geq t^2] \leq \frac{V(x)}{t^2}$$

$$P(|x - \mu| \geq t) \leq \frac{V(x)}{t^2}$$



2. $p(A^c) = 1 - p(A)$

$$P(|x - \mu| \geq t) = 1 - P(|x - \mu| < t) \leq \frac{V(x)}{t^2}$$

$$P(|x - \mu| < t) \geq 1 - \frac{V(x)}{t^2}$$

Ex. 1: If $x \sim \text{unif.}(-\sqrt{3}, \sqrt{3})$ then:

a. Find the upper bound of $P(|x - \mu| \geq \frac{3}{2})$

b. Find the value of $P(|x - \mu| \geq \frac{3}{2})$

Sol.:

$$\text{a. } f(x) = \begin{cases} \frac{1}{\sqrt{3} + (\sqrt{3})} = \frac{1}{2\sqrt{3}} & \text{for } -\sqrt{3} < x < \sqrt{3} \\ 0 & \text{o.w.} \end{cases}$$

$$u.b = \frac{V(x)}{t^2}, \quad t = \frac{3}{2}$$

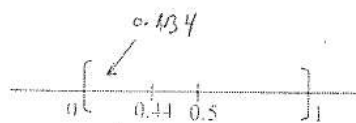
$$E(x) = \int_{-\sqrt{3}}^{\sqrt{3}} x \cdot \frac{1}{2\sqrt{3}} dx = \frac{1}{4\sqrt{3}} x^2 \Big|_{-\sqrt{3}}^{\sqrt{3}} = 0$$

$$E(x^2) = \int_{-\sqrt{3}}^{\sqrt{3}} x^2 \cdot \frac{1}{2\sqrt{3}} dx = \frac{1}{6\sqrt{3}} x^3 \Big|_{-\sqrt{3}}^{\sqrt{3}} = \frac{1}{6\sqrt{3}} [(\sqrt{3})^3 - (-\sqrt{3})^3] = \frac{1}{6\sqrt{3}} [3\sqrt{3} + 3\sqrt{3}] = 1$$

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$$\therefore V(x) = E(x^2) - [E(x)]^2 = 1 - 0 = 1$$

$$\therefore \text{u.b} = \frac{V(x)}{t^2} = \frac{1}{\left(\frac{3}{2}\right)^2} = \frac{4}{9} = 0.44$$



$$\text{b. } p(|x - \mu| \geq \frac{3}{2}) = p(|x - 0| \geq \frac{3}{2}) = p(|x| \geq \frac{3}{2})$$

$$= 1 - p(|x| < \frac{3}{2}) = 1 - p(-\frac{3}{2} < x < \frac{3}{2}) = 1 - \int_{-\frac{3}{2}}^{\frac{3}{2}} \frac{1}{2\sqrt{3}} dx$$

$$= 1 - \frac{1}{2\sqrt{3}} x \Big|_{-\frac{3}{2}}^{\frac{3}{2}} = 1 - \frac{1}{2\sqrt{3}} \left(\frac{3}{2} + \frac{3}{2} \right) = 1 - \frac{3}{2\sqrt{3}} = 1 - \frac{\sqrt{3}}{2}$$

$$= 1 - 0.866 = 0.134$$

Ex. "2": Given a p.d.f $f(x) = \begin{cases} \frac{2x}{9} & \text{for } 0 < x < 3 \\ 0 & \text{o.w.} \end{cases}$

a. Find the lower bound of $p(\frac{5}{4} < x < \frac{11}{4})$

b. Find the value of $p(\frac{5}{4} < x < \frac{11}{4})$

Sol. a. L.b = $1 - \frac{V(x)}{t^2}$

$$E(x) = \int x \cdot \frac{2x}{9} dx = \frac{2}{9} \int x^2 dx = \frac{2}{9} \left(\frac{x^3}{3} \right) \Big|_0^3 = \frac{2}{27} (27 - 0) = 2$$

$$E(x^2) = \int x^2 \cdot \left(\frac{2x}{9} \right) dx = \frac{2}{9} \int x^3 dx = \frac{2}{9} \left(\frac{x^4}{4} \right) \Big|_0^3 = \frac{1}{18} (81) = \frac{9}{2} = 4.5$$

$$\therefore V(x) = E(x^2) - [E(x)]^2$$

$$= 4.5 - 4 = 0.5 = \frac{1}{2}$$

$$p\left(\frac{5}{4} < x < \frac{11}{4}\right) = p\left(\frac{5}{4} - 2 < x - 2 < \frac{11}{4} - 2\right) = p\left(-\frac{3}{4} < x - 2 < \frac{3}{4}\right)$$

$$= p(|x - 2| < \frac{3}{4})$$

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

$$\therefore \text{L.b} = 1 - \frac{V(x)}{t^2} = 1 - \frac{\frac{1}{2}}{\frac{9}{16}} = 1 - \frac{1}{2} \times \frac{16}{9} = 1 - \frac{8}{9} = \frac{1}{9}$$

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$$\begin{aligned} \text{b. } p\left(\frac{5}{4} < x < \frac{11}{4}\right) &= \int_{\frac{5}{4}}^{\frac{11}{4}} \frac{2x}{9} dx \\ &= \frac{1}{9} x^2 \Big|_{\frac{5}{4}}^{\frac{11}{4}} = \frac{1}{9} \left[\frac{121}{16} - \frac{25}{16} \right] = \frac{1}{9} \left[\frac{96}{16} \right] = \frac{6}{9} \end{aligned}$$



Median of Distribution of r.v.

Def.: The median (m) is a value of x such that satisfying the two following inequalities:

$$p(x < m) < \frac{1}{2} \quad \& \quad p(x \leq m) \geq \frac{1}{2}$$

By properties of c.d.f $F(x)$

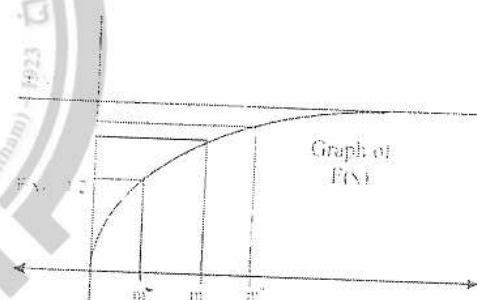
$$p(x < m) = F(\bar{m}), \quad p(x \leq m) = F(m) = F(\bar{m}^+)$$

$$\text{i.e.: } p(x < m) = F(\bar{m}) < \frac{1}{2}, \quad p(x \leq m) = F(\bar{m}^+) \geq \frac{1}{2}$$

Note ① If $\bar{m} = m = \bar{m}^+$ then

$$p(x \leq m) = F(m) = \frac{1}{2}$$

② The value of median is unique



Ex.: Given a p.m.f $f(x) = \begin{cases} \frac{x}{15} & \text{for } x = 1, 2, 3, 4, 5 \\ 0 & \text{o.w.} \end{cases}$
Find the median of x .

$$\text{Sol.: } p(x < m) < \frac{1}{2} \quad \& \quad p(x \leq m) \geq \frac{1}{2}$$

Suppose that $m = 1$

$$p(x < 1) = 0 < \frac{1}{2}, \quad p(x \leq 1) = f(1) = \frac{1}{15} < \frac{1}{2}$$

$\therefore m \neq 1$

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Suppose that $m = 2$

$$p(x < 2) = f(1) = \frac{1}{15} < \frac{1}{2}$$

$$p(x \leq 2) = f(1) + f(2) = \frac{1}{15} + \frac{2}{15} = \frac{3}{15} \neq \frac{1}{2}$$

$\therefore m \neq 2$

Suppose that $m = 3$

$$p(x < 3) = f(1) + f(2) = \frac{1}{15} + \frac{2}{15} = \frac{3}{15} < \frac{1}{2}$$

$$p(x \leq 3) = f(1) + f(2) + f(3) = \frac{1}{15} + \frac{2}{15} + \frac{3}{15} = \frac{6}{15} \neq \frac{1}{2}$$

$\therefore m \neq 3$

Suppose that $m = 4$

$$p(x < 4) = f(1) + f(2) + f(3) = \frac{6}{15} < \frac{1}{2}$$

$$p(x \leq 4) = f(1) + f(2) + f(3) + f(4) = \frac{10}{15} \geq \frac{1}{2}$$

$\therefore m = 4$ is median

$$\frac{12}{30} < \frac{15}{30}$$

Ex.: Given a p.d.f

$$f(x) = \begin{cases} \frac{1}{x^2} & \text{for } x > 1 \\ 0 & \text{o.w.} \end{cases}$$

Find the median of x ?

Sol.: $p(x < m) < \frac{1}{2}$ & $p(x \leq m) \geq \frac{1}{2}$

$$p(x < m) = \int_1^m \frac{1}{x^2} dx = -\frac{1}{x} \Big|_1^m = -\left[\frac{1}{m} - 1\right] < \frac{1}{2}$$

$$-\frac{1}{m} + 1 < \frac{1}{2} \Rightarrow -\frac{1}{m} < -\frac{1}{2} \Rightarrow m < 2 \quad \dots(1)$$

$$p(x \leq m) = \int_1^m \frac{1}{x^2} dx \geq \frac{1}{2}$$

$$-\left[\frac{1}{m} - 1\right] \geq \frac{1}{2} \Rightarrow m \geq 2 \quad \dots(2)$$

From (1) & (2) $\Rightarrow$ median = 2

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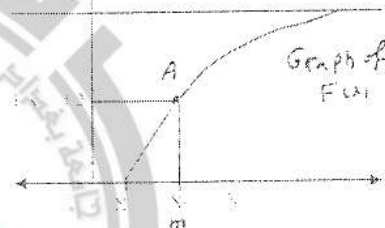
Note: We can also find the value of (Median) from the graph of $F(x)$ such that. At the point $F(x) = \frac{1}{2}$, draw a line to cut the curve of $F(x)$ at A, draw a line to cut the X-axis at m (median) from ex. "2" above

$$f(x) = \begin{cases} \frac{1}{x^2} & \text{for } x > 1 \\ 0 & \text{o.w.} \end{cases}$$

$$F(x) = p(X \leq x) = \int_1^x \frac{1}{t^2} dt = 1 - \frac{1}{x} \Rightarrow F(x) = \begin{cases} 1 - \frac{1}{x} & \text{for } x \geq 1 \\ 0 & \text{o.w.} \end{cases}$$

$$\text{or : } F(x) = \frac{1}{2}$$

$$1 - \frac{1}{x} = \frac{1}{2} \Rightarrow -\frac{1}{x} = -\frac{1}{2} \Rightarrow x = 2$$



To Find the Mode of Dist. Of R.V.X

Def.: Mode is a value of a r.v.x that maximize $f(x)$

i.e. If $x_1 = \text{mode}$, then $f(x_1)$ is a Max.

Note: $F(x_1)$ is Max. $\Leftrightarrow f'(x_1) < 0$

Ex.: Given a p.d.f. $f(x) = \begin{cases} 12x^2(1-x) & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$

Find the mode of x?

Sol.: $0 < \text{mode} < 1$

$$f(x) = 12x^2 - 12x^3 \Rightarrow f'(x) = 24x - 36x^2$$

$$\Rightarrow 12x(2-3x) = 0 \quad \text{either } x = 0 \quad \text{or } x = \frac{2}{3}$$

$$f''(x) = 24 - 72x \Rightarrow f''(0) = 24 > 0 \Rightarrow f(0) \text{ is min.}$$

$$f''\left(\frac{2}{3}\right) = 24 - 72\left(\frac{2}{3}\right) = -24 < 0 \Rightarrow f\left(\frac{2}{3}\right) \text{ is max}$$

$$x_1 = \frac{2}{3} = \text{mode}$$

THE FUTURE

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Def.: Percentile

It is a value of x say (x_0) such that $p(x \leq x_0) = \frac{t}{100}$ $0 < t < \infty$

denoted by P_t , $0 < t < 100$

i.e.: $P_t = x_0$

Ex.: Given a p.d.f $f(x) = \begin{cases} \frac{x}{2} & 0 < x < 2 \\ 0 & \text{o.w.} \end{cases}$

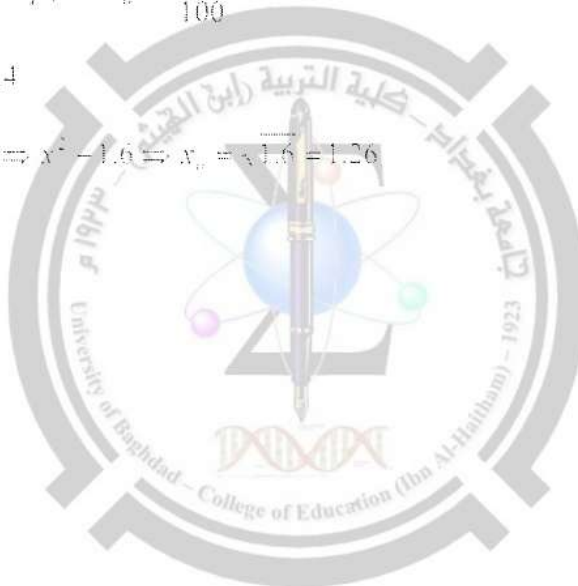
Find P_{40} , P_{65}

Sol.: let $p_{40} = x_0$, $t = 40$

$$p(x \leq x_0) = \frac{t}{100}, \quad p(x \leq x_0) = \frac{40}{100} = 0.4$$

$$\int_0^{x_0} \frac{x}{2} dx = \frac{x^2}{4} = 0.4$$

$$[x^2 - 0^2] = 4(0.4) \Rightarrow x^2 = 1.6 \Rightarrow x_0 = \sqrt{1.6} = 1.26$$



Exercises

① Given a p.f.

$$f(-1) = \frac{1}{8}, f(0) = \frac{6}{8}, f(1) = \frac{1}{8}$$

Find the u.b. of $P(|X| \geq 2\sigma)$.

sol.

$$E(X) = \sum_{x=-1}^1 x f(x) = (-1)f(-1) + (0)f(0) + (1)f(1) \\ = -\frac{1}{8} + 0 + \frac{1}{8} = 0$$

$$\boxed{E(X) = \mu = 0}$$

$$E(X^2) = \sum_{x=-1}^1 x^2 f(x) = (-1)^2 f(-1) + (0)^2 f(0) + (1)^2 f(1) \\ = \frac{1}{8} + 0 + \frac{1}{8} = \frac{1}{4}$$

$$\sigma^2 = V(X) = E(X^2) - [E(X)]^2 = \frac{1}{4} - 0 = \frac{1}{4}$$

$$\boxed{\sigma = \frac{1}{2}}$$

$$P(|X| \geq 2\sigma) = P(|X - \mu| \geq 2 \cdot \frac{1}{2}) = P(|X - \mu| \geq 1) \\ = P(|X - \mu| \geq t) \leq \frac{V(X)}{t^2} \Rightarrow t = 1$$

$$u.b = \frac{V(X)}{t^2} = \frac{(\frac{1}{4})}{1} = \frac{1}{4}$$

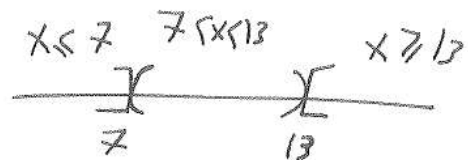
Note If we have to find the exact value of above pr. :

$$P(|X| \geq 2 \cdot \frac{1}{2}) = P(|X| \geq 1) = 1 - P(|X| < 1) \\ = 1 - P(-1 < X < 1) = 1 - P(X=0) = 1 - f(0) \\ = 1 - \frac{6}{8} = \frac{2}{8}$$

② If X is ar.v. with $E(X) = 10$, $P(X \geq 7) = 0.1$, $P(X \geq 13) = 0.3$, then show that $V(X) \geq \frac{9}{2}$.

sol.

$$P(|X - \mu| < t) \geq 1 - \frac{V(X)}{t^2} \\ P(|X - \mu| \geq t) \leq \frac{V(X)}{t^2}$$



$$P[(X \leq 7) \cup (7 \leq X \leq 13) \cup (X \geq 13)] = P(S)$$

$$P(X \leq 7) + P(7 \leq X \leq 13) + P(X \geq 13) = 1$$

$$0.2 + P(7 \leq X \leq 13) + 0.3 = 1$$

$$P(7 \leq X \leq 13) = 0.5 = \frac{1}{2}$$

$$\therefore E(X) = \mu = 10$$

④ Let $X \sim b(2, p)$ & $Y \sim b(4, p)$. If $P(X \geq 1) = \frac{5}{9}$, Find $P(Y \geq 1)$.

Sol. $\because X \sim b(2, p)$

$$\because f(x, 2, p) = \begin{cases} \binom{2}{x} p^x (1-p)^{2-x} & \text{for } x=0, 1, 2 \\ 0 & \text{o.w.} \end{cases}$$

$$P(X \geq 1) = \frac{5}{9} \Rightarrow 1 - P(X < 1) = \frac{5}{9} \Rightarrow 1 - P(X=0) = \frac{5}{9}$$

$$\because P(X=0) = \frac{4}{9} = f(0)$$

$$\binom{2}{0} p^0 (1-p)^2 = \frac{4}{9}$$

$$1 \cdot 1 \cdot (1-p)^2 = \frac{4}{9}$$

$$(1-p)^2 = \frac{4}{9}$$

$$(1-p) = \pm \frac{2}{3}$$

$$\text{If } 1-p = \frac{2}{3}$$

$$\text{then } \boxed{p = \frac{1}{3}}$$

$$\text{If } 1-p = -\frac{2}{3}$$

$$\text{Since } p = 1 + \frac{2}{3} = \frac{5}{3} > 1$$

$$0 < p < 1$$

$$\therefore \boxed{p = \frac{1}{3}}$$

$$\therefore Y \sim b(4, p)$$

$$\therefore f(y, 4, p) = \begin{cases} \binom{4}{y} p^y (1-p)^{4-y} & ; y=0, \dots, 4 \\ 0 & \text{o.w.} \end{cases}$$

$$\because p = \frac{1}{3} \Rightarrow 1-p = \frac{2}{3}$$

$$P(Y \geq 1) = 1 - P(Y < 1) = 1 - P(Y=0)$$

$$= 1 - f(0)$$

$$= 1 - \binom{4}{0} \left(\frac{1}{3}\right)^0 \left(\frac{2}{3}\right)^4$$

$$\therefore P(Y \geq 1) = 1 - \left(\frac{2}{3}\right)^4 = 1 - \frac{16}{81} = \frac{65}{81} < 1$$

⑤ If $X \sim G(\frac{1}{2})$, find Median of X

Sol. $X \sim G(p)$; $p = \frac{1}{2}$, $q = 1 - \frac{1}{2} = \frac{1}{2}$

$$f(x; p) = \begin{cases} \frac{1}{2} \left(\frac{1}{2}\right)^{x-1} & \text{for } x=1, 2, 3, \dots \\ 0 & \text{o.w.} \end{cases}$$

$$P(X \leq m) \leq \frac{1}{2} \text{ \& } P(X \leq m) \geq \frac{1}{2} , m \equiv \text{median}$$

$$\text{If } m=1, P(X \leq 1) = P(X=0) = 0 < \frac{1}{2}$$

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

$$\text{then } \boxed{m=1}$$

Can $m=2$?

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

$$P(X \leq 2) = P(X = 1, 2) = P(X = 1) + P(X = 2) = \frac{1}{2} + \frac{1}{2} \left(\frac{1}{2}\right) = \frac{3}{4} > \frac{1}{2}$$

$$\therefore \boxed{m = 2}$$

There are two values of Median ≤ 2 .

⑥ Given $M_x(t) = \left(\frac{2}{3} + \frac{1}{3}e^t\right)^9$, Show that:

$$P(\mu - 2\sigma < X < \mu + 2\sigma) = \sum_{x=1}^5 \binom{9}{x} \left(\frac{1}{3}\right)^x \left(\frac{2}{3}\right)^{9-x} = P(0 < X < 6)$$

Sol. $M_x(t) = \left(\frac{2}{3} + \frac{1}{3}e^t\right)^9 \Rightarrow p = \frac{1}{3}, n = 9, 1-p = \frac{2}{3}$

$$\therefore X \sim b\left(9, \frac{1}{3}\right)$$

$$M_x(t) = (1-p + pe^t)^n$$

$$f(x, 9, \frac{1}{3}) = \begin{cases} \binom{9}{x} \left(\frac{1}{3}\right)^x \left(\frac{2}{3}\right)^{9-x} & \text{for } x = 0, 1, 2, \dots, 9 \\ 0 & \text{o.w.} \end{cases}$$

$$E(X) = \mu = np, \sigma^2 = V(X) = np(1-p)$$

$$E(X) = 9\left(\frac{1}{3}\right) = 3 \quad \text{and} \quad V(X) = 3\left(\frac{2}{3}\right) = 2$$

$$\mu = 3$$

$$\sigma^2 = 2 \\ \sigma = \sqrt{2}$$

$$\mu - 2\sigma = 3 - 2\sqrt{2} = 3 - 2(1.4) = 3 - 2.8 = 0.2$$

$$\boxed{\sqrt{2} = 1.4}$$

$$\mu + 2\sigma = 3 + 2.8 = 5.8$$

$$P(\mu - 2\sigma < X < \mu + 2\sigma) = P(0.2 < X < 5.8) = P(1 \leq X \leq 5)$$

$$= \sum_{x=1}^5 f(x)$$

$$= \sum_{x=1}^5 \binom{9}{x} \left(\frac{1}{3}\right)^x \left(\frac{2}{3}\right)^{9-x}$$

$$= P(0 < X < 6)$$

⑦ If $X \sim b(n, p)$, and $Y = \frac{X}{n}$, Prove that :

$$\forall \varepsilon > 0, \lim_{n \rightarrow \infty} P(|Y - p| \geq \varepsilon) = 0$$

Sol. $P(|X - \mu| \geq t) \leq \frac{V(X)}{t^2} = \text{u.b.}$

$$P(|X - \mu| < t) \geq 1 - \frac{V(X)}{t^2} = \text{l.b.}$$

$$P(|\mu - p| \geq \varepsilon) = P\left(\left|\frac{X}{n} - p\right| \geq \varepsilon\right) = P\left(\left|\frac{X - np}{n}\right| \geq \varepsilon\right) = P(|X - np| \geq n \cdot \varepsilon)$$

$$\boxed{\mu = np}$$

$$P(|X - np| \geq n \cdot \varepsilon) \geq \frac{V(X)}{t^2}, \quad \boxed{V(X) = np(1-p)}$$

$$= \frac{np(1-p)}{n^2 \varepsilon^2}$$

$$= \frac{p(1-p)}{n \varepsilon^2}$$

$$\therefore n \rightarrow \infty; P(|X - np| \geq n \cdot \varepsilon) \rightarrow 0$$

$$\therefore \lim_{n \rightarrow \infty} P(|X - \mu| \geq t) = \lim_{n \rightarrow \infty} P(|X - np| \geq n \cdot \varepsilon) = 0 \quad \text{By } (0 \leq P(A) \leq 1)$$

$$\begin{aligned} P(|Y - p| < \varepsilon) &= P\left(\left|\frac{X}{n} - p\right| < \varepsilon\right) = P(|X - np| < n \cdot \varepsilon) \\ &= P(|X - np| < n \cdot \varepsilon); \mu = np, t = n \cdot \varepsilon \\ &< 1 - \frac{V(X)}{t^2}; V(X) = np(1-p) \\ &= 1 - \frac{np(1-p)}{n^2 \varepsilon^2} = 1 - \frac{p(1-p)}{n \varepsilon^2} \end{aligned}$$

$$\therefore \lim_{n \rightarrow \infty} P(|Y - p| < \varepsilon) = \lim_{n \rightarrow \infty} \left(1 - \frac{p(1-p)}{n \varepsilon^2}\right) = 1 - 0 = 1$$

⑧ Suppose that $X_1, X_2, \dots, X_n$ are n -independent random variables, and for $i = 1, 2, \dots, n$ let ψ_i be the M.g.f. of X_i . Let $Y = \sum_{i=1}^n X_i$ and let ψ be the M.g.f. of Y . Then for any t s.t. $\psi_i(t)$ exist,

$$\psi(t) = \prod_{i=1}^n \psi_i(t)$$

Sol. X_i ($i = 1, 2, \dots, n$) is indep. r.v.s

ψ_i ($i = 1, 2, \dots, n$) be M.g.f. of X_i

6 ✓

If $Y = \sum_{i=1}^n X_i$, then if ψ be the M.g.f. of ψ , then
 for every t s.t. $(\psi_i(t))$ exists, then

$$\begin{aligned}\psi(t) &= \prod_{i=1}^n \psi_i(t) \quad \text{s.t. } Y = \sum_{i=1}^n X_i \\ \psi_Y(t) &= E(e^{tY}) = E[e^{(\sum X_i)t}] = E[e^{tX_1 + tX_2 + \dots + tX_n}] \\ &= E(e^{tX_1}) \cdot E(e^{tX_2}) \dots E(e^{tX_n}) \\ &= (\psi_1(t)) (\psi_2(t)) \dots (\psi_n(t)) \\ &= \prod_{i=1}^n \psi_i(t)\end{aligned}$$

⑨ Let X be a r.v. with M.g.f. ψ_1 , Let $Y = aX + b$ where $a, b \in \mathbb{R}$ and let ψ_2 denote the M.g.f. of Y , then for any value of t such that $\psi_1(at)$ exists, $\psi_2(t) = \exp(bt) \cdot \psi_1(at)$.

Sol.

$$\begin{aligned}\psi_2(t) &= E(e^{tY}) = E[e^{(ax+b)t}] = E[e^{axt} \cdot e^{bt}] = e^{bt} E(e^{axt}) \\ &= e^{bt} \cdot \psi_1(at) = \exp(bt) \psi_1(at)\end{aligned}$$

Since $\psi_1(t) = E(e^{tx})$
 $\therefore \psi_1(at) = E(e^{atx})$

⑩ If X is a r.v. with $M_X(t)$ exists for $t \in (-h, 0)$, then
 $P(X \leq a) \leq \exp(-at) \cdot M_X(t)$

Sol. $M_X(t) = E(e^{tx})$

$$Y = e^{tx} > 0$$

$$\therefore P(X \geq t) \leq \frac{E(Y)}{t} \quad \text{by Markov}$$

$$P(Y \geq e^{at}) \leq \frac{E(Y)}{e^{at}}$$

$$P(e^{tx} \geq e^{at}) \leq \exp(at) \cdot M_X(t)$$

$$\therefore P(X \leq a) \leq \exp(-at) \cdot M_X(t)$$

45

X
 jesus

- 7
- ⑪ Given a m.g.f. $M_X(t) = e^{3t^2+2t}$ for $-\infty < t < \infty$
Find the L.b. of $P(-\frac{1}{2} < X < \frac{9}{2})$.

Sol.

$$M_X(t) = (e^{3t^2+2t})(6t+2)$$

$$E(X) = M'_X(0) = e^0(0+2)$$

$$M''_X(t) = (e^{3t^2+2t})(6) + (6t+2)^2 e^{3t^2+2t}$$

$$E(X^2) = (e^0)(6) + (2)^2(e^0) = 6+4=10$$

$$V(X) = E(X^2) - [E(X)]^2 = 10 - (2)^2 = 6$$

$$\therefore P(-\frac{1}{2} < X < \frac{9}{2}) = P(-\frac{1}{2} - 2 < X - 2 < \frac{9}{2} - 2)$$

$$= P(-2.5 < X - 2 < 2.5)$$

$$= P(|X - 2| < 2.5) \Rightarrow t = \frac{5}{2} = 2.5$$

$$\text{L.b.} = 1 - \frac{V(X)}{t^2} = 1 - \frac{6}{(\frac{5}{2})^2} = 1 - 6 \cdot \frac{4}{25} = 1 - \frac{24}{25} = \frac{1}{25} \approx 0.04$$

- ⑫ Given a p.m.f. $f(x) = \begin{cases} \frac{x}{10} & \text{for } x=0,1,2,3,4 \\ 0 & \text{o.w.} \end{cases}$
Find the pr. Percentile at the point $x=3.5$.

Sol.

$$P(X \leq x_0) = \frac{t}{100}$$

$$P(X \leq 3.5) = \frac{6}{100} \Rightarrow \sum_{x=0}^3 f(x) = \frac{t}{100}$$

$$f(0) + f(1) + f(2) + f(3) = \frac{t}{100}$$

$$\frac{0}{10} + \frac{1}{10} + \frac{2}{10} + \frac{3}{10} = \frac{t}{100} \Rightarrow \frac{6}{10} = \frac{t}{100} \Rightarrow 10t = 600$$

$$\boxed{t = 60\%}$$

- ⑬ Find the mean, median and mode of Cauchy dist. if exists.

Sol. $f(x) = \frac{1}{\pi(1+x^2)} \quad -\infty < x < \infty \quad (\text{Cauchy dist.})$

Mean: $E(X)$ exist only when $E(|X|) < \infty$

$$E(|X|) = \int_{-\infty}^{\infty} |x| \frac{1}{\pi(1+x^2)} dx = \int_0^{\infty} \frac{2x}{\pi(1+x^2)} dx = \frac{1}{\pi} \ln(1+x^2) \Big|_0^{\infty}$$

$$= \frac{1}{\pi} [\ln(\infty) - \ln(0)] = \infty$$

$E(X)$ not exist.

Median $P(X \leq m) \geq \frac{1}{2}$ & $P(X \leq m) \leq \frac{1}{2}$

$$P(X \leq m) \leq \frac{1}{2}$$

$$P(X \leq m) \geq \frac{1}{2}$$

$$\int_{-\infty}^m \frac{1}{\pi(1+x^2)} dx \leq \frac{1}{2}$$

$$\frac{1}{\pi} \int_{-\infty}^m \frac{1}{1+x^2} dx \leq \frac{1}{2}$$

$$\frac{1}{\pi} \tan^{-1} x \Big|_{-\infty}^m \leq \frac{1}{2}$$

$$\frac{1}{\pi} \left[\tan^{-1}(m) - \frac{\tan^{-1}(\infty)}{-\pi/2} \right] \leq \frac{1}{2}$$

$$\tan^{-1}(m) + \frac{\pi}{2} \leq \frac{\pi}{2}$$

$$\tan^{-1}(m) \leq 0$$

$$m \leq \tan(0)$$

$$\therefore m \leq 0$$

$$\int_{-\infty}^m \frac{1}{\pi(1+x^2)} dx \geq \frac{1}{2}$$

$$\frac{1}{\pi} \int_{-\infty}^m \frac{1}{(1+x^2)} dx \geq \frac{1}{2}$$

$$\frac{1}{\pi} \tan^{-1} x \Big|_{-\infty}^m \geq \frac{1}{2}$$

$$\frac{1}{\pi} [\tan^{-1}(m) - \tan^{-1}(\infty)] \geq \frac{1}{2}$$

$$\tan^{-1}(m) \geq 0$$

$$m \geq \tan(0)$$

$$\therefore m \geq 0$$

$$\therefore \boxed{m_e = 0} \text{ median.}$$

Mode $f(x) = \frac{1}{\pi} (1+x^2)^{-1} \Rightarrow f'(x) = -\frac{1}{\pi} (1+x^2)^{-2} (2x) = 0$

$$\frac{-2x}{\pi(1+x^2)^2} = 0 \Rightarrow x=0$$

$$f''(x) = \frac{2}{\pi} (1+x^2)^{-3} (2x)(2x) - \frac{1}{\pi} (1+x^2)^{-2} (2)$$

$$f''(0) = 0 - \frac{2}{\pi} < 0 \Rightarrow \text{mode} = 0$$

$$\therefore \boxed{m_o = 0} \text{ mode.}$$

(14) Given a p.d.f. $f(x) = \begin{cases} e^x & \text{for } x \leq 0 \\ 0 & \text{o.w.} \end{cases}$

Find $M_x(t)$ and sketch its graph. Also, find $E(x)$ by two methods.

Sol.

$$M_x(t) = E(e^{tx}) = \int_{-\infty}^0 e^{tx} e^x dx = \int_{-\infty}^0 e^{x(1+t)} dx$$

This integration exists only when $(1+t) > 0 \Rightarrow \boxed{t > -1}$

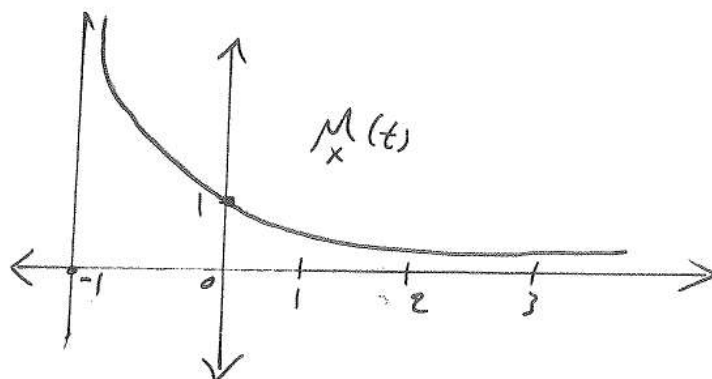
$$M_X(t) = \frac{1}{1+t} e^{x(1+t)} \Big|_{-\infty}^0 = \frac{1}{1+t} [e^0 - e^{-\infty}] = \frac{1}{1+t} [1 - 0] = \frac{1}{1+t}$$

$$M_X(t) = \begin{cases} \frac{1}{1+t} & \text{for } t > -1 \\ 0 & \text{o.w.} \end{cases}$$

must $1+t > 0$

$t > -1$

t	$M_X(t)$
-1	∞
0	1
1	$\frac{1}{2}$
2	$\frac{1}{3}$
3	$\frac{1}{4}$
$\vdots$	
∞	0



$$\begin{aligned} \textcircled{1} E(X) &= \int_{-\infty}^{\infty} x f(x) dx = \int_{-\infty}^0 x \cdot e^x dx \\ &= uv \Big|_{-\infty}^0 - \int_{-\infty}^0 v du \\ &= x e^x \Big|_{-\infty}^0 - \int_{-\infty}^0 e^x dx \\ &= [0 \cdot e^0 - (-\infty) e^{-\infty}] - [e^x]_{-\infty}^0 = [0 - 0] - [e^0 - e^{-\infty}] \\ &= -1 \end{aligned}$$

$u = x, dv = e^x$
 $du = 1, v = e^x$

$$M_X(t) = \frac{1}{1+t} = (1+t)^{-1} ; t > -1$$

$$M_X'(t) = -1(1+t)^{-2} (1)$$

$$E(X) = M_X'(0) = (-1)[1+0]^{-1} = -1$$

(15) If $X \sim \text{Unif}(0, 2)$, find the u.b. of $P(|X - \mu| \geq 2)$.

Sol: u.b. = $\frac{V(X)}{t^2}$, $t = 2$

$$f(x) = \begin{cases} \frac{1}{2-0} = \frac{1}{2} & 0 < x < 2 \\ 0 & \text{o.w.} \end{cases}$$

$$\mu = E(X) = \int_0^2 x f(x) dx = \int_0^2 x \cdot \frac{1}{2} dx = \frac{1}{2} \cdot \frac{x^2}{2} \Big|_0^2 = \frac{1}{4} [4 - 0] = \frac{4}{4} = 1$$

$$\boxed{\mu = E(X) = 1}$$

$$E(X^2) = \int_0^2 x^2 \cdot \frac{1}{2} dx = \frac{1}{2} \frac{x^3}{3} \Big|_0^2 = \frac{1}{6} [8 - 0] = \frac{8}{6} = \frac{4}{3}$$

$$\begin{aligned}
 V(X) &= E(X^2) - [E(X)]^2 \\
 &= \frac{4}{3} - (1)^2 \\
 &= \frac{4}{3} - \frac{3}{3} = \frac{1}{3}
 \end{aligned}$$

$$U.b. = \frac{V(X)}{t^2} = \frac{(\frac{1}{3})}{(2)^2} = \frac{1/3}{4} = \frac{1}{12}$$

①6 Find the pr. percentile at the point $x=1$, if ~~given~~ Given a p.d.f.

$$f(x) = \begin{cases} \frac{3}{8}x^2 & \text{for } 0 \leq x \leq 2 \\ 0 & \text{o.w.} \end{cases}$$

Sol. $P(X \leq 1) = \frac{t}{100} \Rightarrow \int_0^1 f(x) dx = \frac{t}{100} \Rightarrow \frac{3}{8} \int_0^1 x^2 dx = \frac{t}{100}$

$$\frac{3}{8} \cdot \frac{x^3}{3} \Big|_0^1 = \frac{1}{8} = \frac{t}{100} \Rightarrow t = \frac{100}{8} = 12.5\% = 0.125$$

①7 A coin is tossed 4-times, $X \equiv$ number of heads. Find the mean, median and mode of X (if exists).

Sol. $f(x) = \begin{cases} \frac{\binom{4}{x}}{16} & x=0,1,2,3,4 \\ 0 & \text{o.w.} \end{cases}$

x	$f(x) = P(X=x)$
0	$f(0) = \frac{1}{16}$
1	$f(1) = \frac{4}{16}$
2	$f(2) = \frac{6}{16}$
3	$f(3) = \frac{4}{16}$
4	$f(4) = \frac{1}{16}$

$$\sum_{x=0}^4 f(x) = \frac{16}{16} = 1$$

$f(2)$ is a Maximum
∴ mode = 2

Median

$$\neq m = 0$$

$$P(X < 0) \leq \frac{1}{2} ; P(X \leq 0) \geq \frac{1}{2}$$

$$0 \leq \frac{1}{2}$$

$$\therefore m \neq 0$$

$$f(0) = \frac{1}{16} \neq \frac{1}{2}$$

If $m=1$

$$P(X < 1) \leq \frac{1}{2} ; P(X \leq 1) \geq \frac{1}{2}$$

$$f(0) = \frac{1}{16} < \frac{1}{2} ; f(0) + f(1) \geq \frac{1}{2}$$

$$\therefore m \neq 1$$

$$\frac{1}{16} + \frac{4}{16} \geq \frac{1}{2}$$

$$\frac{5}{16} \neq \frac{1}{2}$$

if $m=2$

$$P(X \leq 2) \leq \frac{1}{2}$$

$$f(0) + f(1) \leq \frac{1}{2}$$

$$\frac{1}{16} + \frac{4}{16} \leq \frac{1}{2}$$

$$\frac{5}{16} \leq \frac{1}{2}$$

$$P(X \geq 2) \geq \frac{1}{2}$$

$$f(0) + f(1) + f(2) \geq \frac{1}{2}$$

$$\frac{1}{16} + \frac{4}{16} + \frac{6}{16} \geq \frac{1}{2}$$

$$\frac{11}{16} \geq \frac{1}{2}$$

$$\therefore \text{Median} = \boxed{m=2}$$

Mean

$$E(X) = \sum_{x=0}^4 x f(x) = 0 \cdot f(0) + 1 \cdot f(1) + 2 \cdot f(2) + 3 \cdot f(3) + 4 \cdot f(4)$$

$$= 1 \cdot \frac{4}{16} + 2 \cdot \frac{6}{16} + 3 \cdot \frac{4}{16} + 4 \cdot \frac{1}{16}$$

$$= \frac{4}{16} + \frac{12}{16} + \frac{12}{16} + \frac{4}{16} = \frac{32}{16} = 2 = \mu$$

$$\therefore \boxed{\mu=2}$$

(18) If $X \sim b(n, p)$, then find the value of (n) s.t.

$$P\left(\left|\frac{X}{n} - p\right| < 0.1\right) \geq 0.95$$

Sol. By theorem (Chebyshev's theorem)

$$P\left(\left|\frac{X}{n} - p\right| < 0.1\right) \geq 0.95 \iff P\left(\left|\frac{X}{n} - \mu\right| < t\right) \geq 1 - \frac{V(X)}{t^2}$$

$$= P(|X - np| < (0.1)n) = P(|X - \mu| < \frac{n}{10}) \Rightarrow \boxed{t = \frac{n}{10}}, \mu = np$$

$$L.b. = 1 - \frac{V(X)}{t^2}$$

$$V(X) = np(1-p)$$

$$L.b. = 1 - \frac{np(1-p)}{\left(\frac{n}{10}\right)^2} = 1 - \left[np(1-p) \frac{100}{n^2} \right]$$

$$\geq 1 - \frac{100 p(1-p)}{n} \geq 0.95$$

$$\frac{100 p(1-p)}{n} \leq 0.05 = \frac{5}{100} = \frac{1}{20}$$

$$2000 p(1-p) \leq n$$

$$\boxed{n \geq 2000 p(1-p)}$$

$$\forall p \text{ } 0 < p < 1$$

①9 If X has a M.g.f. as follows:

$$M_X(t) = \frac{2e^t}{5(1 - \frac{2}{5}e^t)} \text{ for } t < \ln(\frac{5}{2}), \text{ then find } P(X > 7 | X > 3).$$

Sol.

$$M_X(t) = \frac{(\frac{2}{5})e^t}{(1 - \frac{2}{5}e^t)} = \frac{pe^t}{1 - qe^t}$$

memoryless

$$\therefore X \sim G(p = \frac{2}{5})$$

$$\therefore P(X > 7 | X > 3) = P(X > 4) = (q)^4 = (\frac{3}{5})^4.$$

②0 If X has a p.m.f.

$$f(x) = \begin{cases} (\frac{2}{3}) (\frac{1}{3})^{x-1} & \text{for } x = 1, 2, 3, \dots \\ 0 & \text{otherwise} \end{cases}$$

Find $P(X > 5 | X > 2)$

Sol. $X \sim G(\frac{2}{3})$, $p = \frac{2}{3}$, $q = \frac{1}{3}$

$$\therefore P(X > 5 | X > 2) = P(X > 3) = (q)^3 = (\frac{1}{3})^3 = \frac{1}{27}$$

②1 If $X \sim P(\lambda)$, then show that:

$$E(X) = V(X) = \lambda$$

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

$$M'_X(t) = e^{\lambda(e^t - 1)} (\lambda e^t)$$

$$E(X) = M'_X(0) = e^0 (\lambda) e^0 = \lambda$$

$$M''_X(t) = e^{\lambda(e^t - 1)} (\lambda e^t) + (\lambda e^t) e^{\lambda(e^t - 1)} (\lambda e^t)$$

$$E(X^2) = M''_X(t=0) = 1 \cdot (\lambda) + (\lambda) \cdot (1) \cdot (\lambda)$$

$$E(X^2) = \lambda + \lambda^2$$

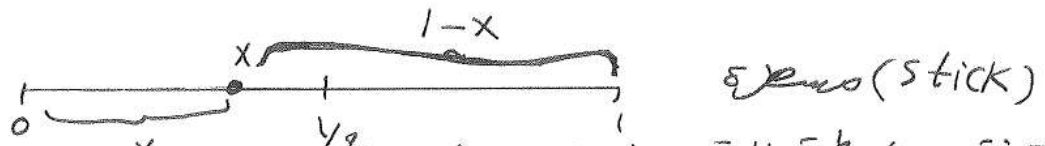
$$V(X) = E(X^2) - [E(X)]^2 = \lambda + \lambda^2 - \lambda^2 = \lambda$$

$$\boxed{E(X) = V(X) = \lambda}$$

derive the Expected value of X and the variance of X

(22) A point X is chosen at random from a stick of length one unite, and then the stick is broken at the chosen point into two unequal parts, find the expected value of longer part.

Sol.



تقسم المسطرة الى قسمين غير متساويين، فاحد الاقسام صغير X ، والاخر كبير $(1-X)$

∴ X is shorter part, $(1-X)$ is longer part.

∴ $X \in (0, \frac{1}{2})$

∴ $X \sim \text{unif.}(0, \frac{1}{2})$

$$f(x) = \begin{cases} \frac{1}{\frac{1}{2}-0} = 2 & \text{for } 0 < x < \frac{1}{2} \\ 0 & \text{o.w.} \end{cases}$$

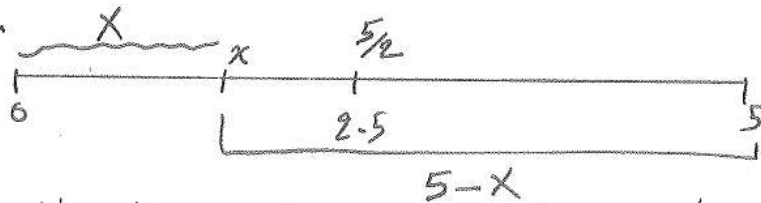
Find $E[\text{longer part}]$

i.e. Find $E(1-X)$?

$$\begin{aligned} E(1-X) &= 1 - E(X) \\ &= 1 - \int_0^{\frac{1}{2}} x \cdot (2) dx = 1 - \frac{2}{2} x^2 \Big|_0^{\frac{1}{2}} = 1 - \left[\left(\frac{1}{2} \right)^2 - 0 \right] \\ &= 1 - \frac{1}{4} = \frac{3}{4} \end{aligned}$$

$$\therefore E(1-X) = \frac{3}{4} \quad (\text{القيمة المتوقعة للجزء الأكبر})$$

(23) A point X is chosen on a line of long 5 cm. This chosen point divides the line into two unequal parts. Find the expectation of shorter part.



Sol.

Let X divide the line into unequal two parts.

∴ X is the shorter part, $(5-X)$ is the longer part.

∴ $X \in (0, \frac{5}{2})$

∴ $X \sim \text{unif.}(0, \frac{5}{2})$

$$f(x) = \begin{cases} \frac{1}{\frac{5}{2}-0} = \frac{2}{5} & \text{for } 0 < x < \frac{5}{2} \\ 0 & \text{o.w.} \end{cases}$$

Find $E(\text{shorter part}) = E(X)$?

$$E(X) = \int x f(x) dx = \int_0^{5/2} x \left(\frac{2}{5}\right) dx = \frac{2}{5} \cdot \frac{x^2}{2} \Big|_0^{5/2} \\ = \frac{1}{5} \left[\left(\frac{5}{2}\right)^2 - 0 \right] = \frac{1}{5} \cdot \frac{25}{4} = \frac{5}{4} = 1.25$$

(24) Given a p.d.f.
 $f(x) = \begin{cases} x^2 & \text{for } 0 \leq x \leq 1 \\ 0 & \text{o.w.} \end{cases}$
 Find mode of X .

Sol. $f(x) = 12x^2 - 2x^3$
 $f'(x) = 24x - 36x^2 = 0$
 $12x(2 - 3x) = 0$
 $(12x = 0) \text{ or } (2 - 3x = 0)$
 $x_1 = 0 \text{ or } x_2 = \frac{2}{3}$

$f''(x) = 24 - 72x$
 g.f. $(x=0) \Rightarrow f''(0) = 24 \Rightarrow f(0)$ is min. then mode $\neq 0$
 g.f. $(x=\frac{2}{3}) \Rightarrow f''(\frac{2}{3}) = 24 - 72(\frac{2}{3}) = -24 < 0$ then $f(\frac{2}{3})$ is Max.
 $\therefore \text{mode} = \frac{2}{3}$

ليست
 البتة

$$\int 12x^2 = \frac{12}{3} x^3 = 4x^3$$

$$\int 2x^3 = \frac{2}{4} x^4 = \frac{1}{2} x^4$$

(25) If X has a poisson dist. with parameter m , then
 $E(X) = V(X) = m$.

Sol. $M_X(t) = e^{m(e^t - 1)}$ (by theorem)

$$M'_X(t) = e^m (e^t - 1) \cdot m e^t$$

$$\therefore E(X) = M'_X(0) = m e^m (e^0 - 1) = m$$

$$M''_X(t) = e^m (e^t - 1) \cdot m e^t + m e^t \cdot e^m (e^t - 1)$$

$$E(X^2) = M''_X(0) = m e^0 + m e^0 \cdot m e^0 = m + m^2$$

$$V(X) = E(X^2) - [E(X)]^2$$

$$V(X) = (m + m^2) - m^2$$

$$V(X) = m = E(X)$$

Then
 derive

- (26) Given a m.g.f. of X , $M_X(t) = \frac{1}{1-2t}$ for $t < \frac{1}{2}$. Find the mean and variance of X .

Sol. $M_X(t) = \frac{1}{1-2t} = (1-2t)^{-1}$

$$M'_X(t) = -(1-2t)^{-2}(-2) = 2(1-2t)^{-2}$$

$$M''_X(t) = -4(1-2t)^{-3}(-2) = 8(1-2t)^{-3}$$

$$M'_X(0) = 2, \quad M''_X(0) = 8$$

$$\therefore E(X) = 2, \quad E(X^2) = 8 \quad (\text{by theorem})$$

$$V(X) = E(X^2) - [E(X)]^2 = 8 - (2)^2 = 4$$

- (27) If X has a uniform dist. on $(-\sqrt{3}, \sqrt{3})$, then find the upper bound of $P(|X - \mu| > \frac{3}{2})$.

Sol. U.b. = $\frac{V(X)}{t^2} = \frac{1}{\frac{9}{4}} = \frac{4}{9}$ for $-\sqrt{3} < X < \sqrt{3}$
 $f(x) = \frac{1}{\sqrt{3} - (-\sqrt{3})} = \begin{cases} \frac{1}{2\sqrt{3}} & \text{for } -\sqrt{3} < x < \sqrt{3} \\ 0 & \text{o.w.} \end{cases}$

$$E(X) = \int_{-\sqrt{3}}^{\sqrt{3}} x \left(\frac{1}{2\sqrt{3}} \right) dx = 0$$

$$E(X^2) = \int_{-\sqrt{3}}^{\sqrt{3}} x^2 \left(\frac{1}{2\sqrt{3}} \right) dx = 1$$

$$\therefore V(X) = E(X^2) - (E(X))^2 = 1 - 0^2 = 1 \Rightarrow V(X) = 1$$

$$U.b. = \frac{V(X)}{t^2} = \frac{1}{\left(\frac{3}{2}\right)^2} = \frac{4}{9} = 0.44$$

- (28) If X has a geometric dist. with $p = \frac{1}{4}$, then find $P(X > 8 | X > 3)$.

Sol. $p = \frac{1}{4} \Rightarrow q = \frac{3}{4}$

$$P(X > 8 | X > 3) = P(X > 5) = [P(A^c)]^5 = q^5 = \left(\frac{3}{4}\right)^5$$

(29) If $X \sim b(n, p)$, then the M.g.f. of X is
 $M_X(t) = (1-p+pe^t)^n$

Sol.

$\because X \sim b(n, p)$

$\therefore f(x, n, p) = \begin{cases} \binom{n}{x} p^x (1-p)^{n-x} & \text{for } x=0, 1, 2, \dots, n \\ 0 & \text{o.w.} \end{cases}$

$M_X(t) = E(e^{tx}) = \sum_{x=0}^n e^{tx} f(x)$

$= \sum_{x=0}^n e^{tx} \binom{n}{x} p^x (1-p)^{n-x}$
 $= \sum_{x=0}^n \binom{n}{x} (pe^t)^x (1-p)^{n-x}$

Since $(a+b)^n = \sum_{x=0}^n \binom{n}{x} b^x a^{n-x}$

Let $b = pe^t$ and $a = 1-p$

then $(1-p+pe^t)^n = \sum_{x=0}^n \binom{n}{x} (pe^t)^x (1-p)^{n-x} = M_X(t)$

$\therefore \boxed{M_X(t) = (1-p+pe^t)^n}$

(30) Given a p.d.f. $f(x) = \begin{cases} e^{-x} & \text{for } x > 0 \\ 0 & \text{o.w.} \end{cases}$
 Find and sketch graph of $M_X(t)$.

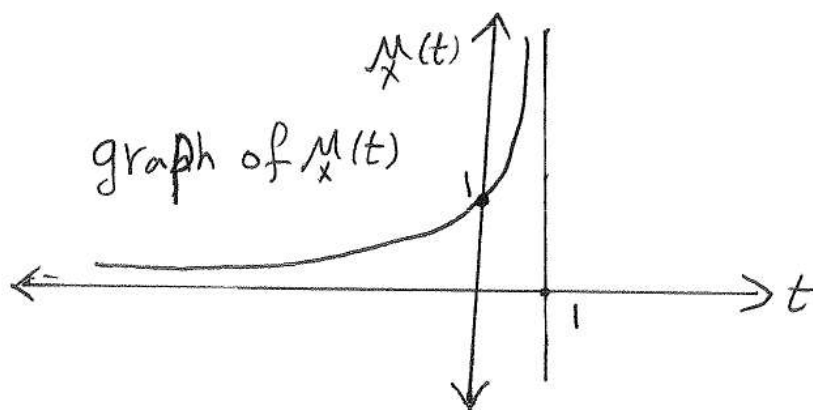
Sol.

$M_X(t) = E(e^{tx}) = \int_0^{\infty} e^{tx} (e^{-x}) dx = \int_0^{\infty} e^{-(1-t)x} dx$

$e^{-(1-t)x}$ exists when $t < 1$

$M_X(t) = \frac{-1}{1-t} \int_0^{\infty} e^{-(1-t)x} (-1) dx = \frac{1}{1-t}$ for $t < 1$

t	$M_X(t) = \frac{1}{1-t}$
1	∞
0	1
-1	.5
$\vdots$	
$-\infty$	0



✓ which distributes that x have
 @ derive M.H.
 (3) E.H. ✓

(31) Given $M_x(t) = \frac{1}{1-3t}$ for $t < \frac{1}{3}$, if $y = 1-2x$, then find $M_y(t)$.

Sol. $M(t) = e^{bt} M_x(at)$, $y = ax+b$
 $M_x(t) = \frac{1}{1-3t}$ for $t < \frac{1}{3}$
 $M_y(t) = e^t M_x(-2t)$, $y = 1-2x$
 where $M_x(t) = \frac{1}{1+6t}$ for $t > -\frac{1}{6}$

(32) If a r.v. X with $M_x(t)$ for $-h < t < h$
 show that $P(X \geq a) \leq e^{-at} M_x(t)$ for $0 < t < h$
 and $P(X \leq a) \leq e^{-at} M_x(t)$ for $-h < t < 0$

Proof $M_x(t) = E(e^{tx})$; $y = e^{tx}$
 $e^{at} > 0$; $a > 0$ & $t > 0$

$$P(X \geq t) \leq \frac{E(X)}{t} \text{ (by theorem)}$$

$$P(Y \geq e^{at}) \leq \frac{E(Y)}{e^{at}}$$

$$P(e^{tx} \geq e^{at}) \leq \frac{E(e^{tx})}{e^{at}}$$

$$P(\ln e^{tx} \geq \ln e^{at}) \leq \frac{M_x(t)}{e^{at}} = e^{-at} M_x(t)$$

$$P(tx \geq at) \leq e^{-at} M_x(t) \text{ for } 0 < t < h$$

$$P(X \geq a) \leq e^{-at} M_x(t) \text{ — ①}$$

$$P(tx \geq at) \leq e^{-at} M_x(t)$$

$$P(X \leq a) \leq e^{-at} M_x(t) \text{ for } -h < t < 0 \text{ — ②}$$

(33) If X has a geometric dist. with parameter p , then the M.g.f. of X is $M_x(t) = \frac{pe^t}{1-qe^t}$ for $t < \ln(\frac{1}{q})$.

Sol. $X \sim G(p, q) \Rightarrow f_X(x) = \begin{cases} p \cdot q^{x-1} & , x = 1, 2, 3, \dots \\ 0 & \text{o.w.} \end{cases}$

derive

{6}

$$\begin{aligned}
 M_X(t) &= E(e^{tx}) \\
 &= \sum_{x=1}^{\infty} e^{tx} f(x) \\
 &= \sum_{x=1}^{\infty} e^{tx} (pq^{x-1}) = \frac{p}{q} \sum_{x=1}^{\infty} (qe^t)^x \\
 &= \frac{p}{q} [qe^t + (qe^t)^2 + \dots + (qe^t)^n + \dots] \\
 &= \frac{p}{q} \cdot qe^t [1 + qe^t + q^2e^{2t} + \dots + q^{n-1}e^{(n-1)t} + \dots]
 \end{aligned}$$

$$\therefore r = \frac{qe^t}{1} = qe^t$$

$$|r| < 1 \Rightarrow qe^t < 1 \Rightarrow e^t < \frac{1}{q} \Rightarrow \ln e^t < \ln\left(\frac{1}{q}\right)$$

$$\Rightarrow t < \ln\left(\frac{1}{q}\right)$$

$$S = \underbrace{1 - qe^t + (qe^t)^2 + \dots}_{\text{Geometric series}}, \quad S = \frac{1}{1-r} = \frac{1}{1-qe^t}$$

$$M_X(t) = \frac{pe^t}{1-qe^t} \quad \text{for } t < \ln\left(\frac{1}{q}\right)$$

(34) $X \sim N(\mu, \sigma^2)$, $Z = \frac{X-\mu}{\sigma} \sim N(0,1)$ find $E(Z)$, $V(Z)$.

Sol. $E(Z) = E\left(\frac{X-\mu}{\sigma}\right)$

$$\begin{aligned}
 &= E\left(\frac{X}{\sigma} - \frac{\mu}{\sigma}\right) \\
 &= E\left(\frac{X}{\sigma}\right) - \frac{\mu}{\sigma} \\
 &= \frac{1}{\sigma} E(X) - \frac{\mu}{\sigma} \\
 &= \frac{\mu}{\sigma} - \frac{\mu}{\sigma} = 0
 \end{aligned}$$

$$\begin{aligned}
 \text{Var}(Z) &= V\left(\frac{X}{\sigma} - \frac{\mu}{\sigma}\right) \\
 &= \frac{V(X)}{\sigma^2} - V\left(\frac{\mu}{\sigma}\right) \\
 &= \frac{\sigma^2}{\sigma^2} - 0 \\
 &= 1
 \end{aligned}$$

$$\begin{aligned}
 M_Z(t) &= e^{\frac{tZ}{\sigma}} = E\left[e^{\left(\frac{X}{\sigma} - \frac{\mu}{\sigma}\right)t}\right] \\
 &= e^{-\frac{\mu}{\sigma}t} E\left[e^{\frac{X}{\sigma}t}\right] \\
 &= e^{-\frac{\mu}{\sigma}t} M_X\left(\frac{t}{\sigma}\right) \\
 &= e^{-\frac{\mu}{\sigma}t} e^{\frac{\mu}{\sigma}t + \frac{\sigma^2 t^2}{2\sigma}} \\
 &= e^{\frac{\sigma^2 t^2}{2\sigma^2}} = e^{t^2/2} = e^{0 + \frac{t^2}{2}}
 \end{aligned}$$

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

✓ [6]

(35) If a r.v. X has Gamma dist. with parameters (a) & (b) , then show that $V(X) = b \cdot E(X)$

Sol. $\because X \sim G(a, b)$

$$\because E(X) = a \cdot b \text{ \& } V(X) = a \cdot b^2$$

$$\begin{aligned} \because V(X) &= (a \cdot b) \cdot b \\ &= E(X) \cdot b \\ &= b E(X) \end{aligned}$$

OR $\because X \sim G(a, b)$

$$\because M_X(t) = (1 - bt)^{-a}$$

$$M_X'(t) = (-a)(1 - bt)^{-a-1} \cdot (-b) = (ab)(1 - bt)^{-(a+1)}$$

$$M_X'(0) = (a \cdot b)(1 - 0)^{-(a+1)} = a \cdot b = E(X); \text{ (Since } E(X) = M_X'(0) \text{)}$$

$$E(X^2) = ?$$

$$E(X^2) = M_X''(0)$$

$$M_X''(t) = (a \cdot b)(a+1)(1 - bt)^{-(a+2)} \cdot (-b) = (ab^2)(a+1)(1 - bt)^{-(a+2)}$$

$$E(X^2) = M_X''(0) = ab^2(a+1)(1 - 0) = a^2b^2 + ab^2$$

$$\begin{aligned} \because V(X) &= E(X^2) - [E(X)]^2 \\ &= (a^2b^2) + (ab^2) - (ab)^2 = ab^2 \end{aligned}$$

$$\because V(X) = (a \cdot b) \cdot b = E(X) \cdot b \Rightarrow V(X) = b E(X)$$

(36) If a m.g.f. of X is as follows:

$M_X(t) = (1 - 2t)^{-7}$, then find the p.d.f. of X , $E(X)$ & $V(X)$.

Sol. $\beta = 2, \alpha = 7 \rightarrow r = 14$

$\because X \sim \chi^2(14)$ chi-squar with (14) d.f.

$$\because f(x) = \begin{cases} \frac{x^6 e^{-\frac{x}{2}}}{\Gamma(7) 2^7} & \text{for } 0 < x < \infty \\ 0 & \text{o.w.} \end{cases}$$

$$E(X) = r = 14, \sigma^2 = V(X) = 2r = 2(14) = 28.$$

Find $P(X \leq a) = 0.95$
the value of a

(37) Given a p.d.f.

$$f(x) = \begin{cases} \frac{2x}{9} & \text{for } 0 < x < 3 \\ 0 & \text{o.w.} \end{cases}$$

Find the lower bounded of $P(\frac{5}{4} < X < \frac{11}{4})$.

Sol. l.b = $1 - \frac{V(X)}{t^2}$

We must find $E(X)$, $E(X^2)$:

$$\begin{aligned} E(X) &= \int_0^3 x f(x) dx = \int_0^3 x \left(\frac{2}{9}x\right) dx = \frac{2}{9} \left[\int_0^3 x^2 dx \right] \\ &= \frac{2}{9} \left[\frac{x^3}{3} \right]_0^3 = \frac{2}{27} [27 - 0] = 2 \end{aligned}$$

$$\begin{aligned} E(X^2) &= \int_0^3 x^2 f(x) dx = \int_0^3 x^2 \left(\frac{2}{9}x\right) dx = \frac{2}{9} \left[\int_0^3 x^3 dx \right] \\ &= \frac{2}{36} \left[\frac{x^4}{4} \right]_0^3 = \frac{1}{18} [81] = \frac{9}{2} = 4.5 \end{aligned}$$

$$\begin{aligned} V(X) &= E(X^2) - [E(X)]^2 \\ &= 4.5 - (2)^2 \\ &= 4.5 - 4 = 0.5 = \frac{1}{2} \end{aligned}$$

$$\begin{aligned} P\left(\frac{5}{4} < X < \frac{11}{4}\right) &= P\left(\frac{5}{4} - 2 < X - 2 < \frac{11}{4} - 2\right) \\ &= P\left(-\frac{3}{4} < X - 2 < \frac{3}{4}\right) \\ &= P\left(|X - 2| < \frac{3}{4}\right) \end{aligned}$$

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

$$\begin{aligned} \text{l.b} &= 1 - \frac{V(X)}{t^2} = 1 - \frac{\frac{1}{2}}{\left(\frac{3}{4}\right)^2} = 1 - \frac{\frac{1}{2}}{\frac{9}{16}} = 1 - \frac{8}{9} \\ &= \frac{1}{9} \end{aligned}$$

$$\therefore \boxed{\text{l.b} = \frac{1}{9}}$$

(38) Given a p.d.f. of X

$$f(x) = \begin{cases} \frac{1}{4} e^{-\frac{x}{4}} & \text{for } x > 0 \\ 0 & \text{o.w.} \end{cases}$$

① Find $M_X(t)$ ② If $Y = 2 - X$, then find $M_Y(t)$

Sol.

Sol.

$$\begin{aligned} \textcircled{1} M_X(t) &= E(e^{tx}) = \frac{1}{4} \int_0^{\infty} e^{tx} \cdot e^{-\frac{1}{4}x} dx \\ &= \frac{1}{4} \int_0^{\infty} e^{-(\frac{1}{4}-t)x} dx \end{aligned}$$

This integration exist only when $(\frac{1}{4}-t) > 0$

i.e. $\frac{1}{4} > t \rightarrow t < \frac{1}{4}$

$$\begin{aligned} \therefore M_X(t) &= \frac{1}{4} \cdot \frac{-1}{(\frac{1}{4}-t)} e^{-(\frac{1}{4}-t)x} \Big|_0^{\infty} = \frac{-1}{4(\frac{1}{4}-t)} [e^{-\infty} - e^0] \\ &= \frac{-1}{4(\frac{1}{4}-t)} [0-1] \end{aligned}$$

$$= \frac{1}{4(\frac{1}{4}-t)} \text{ for } t < \frac{1}{4}$$

$\textcircled{2} y = 2-x, b=2, a=-1$

$$\begin{aligned} M_Y(t) &= e^{bt} \cdot M_X(at) \\ &= e^{2t} \cdot M_X(t) \end{aligned}$$

$$M_X(-t) = \frac{1}{4(\frac{1}{4}+t)} \text{ for } -t < \frac{1}{4}$$

$$M_X(-t) = \frac{1}{4(\frac{1}{4}+t)} \text{ for } t > -\frac{1}{4}$$

$$\therefore M_Y(t) = e^{2t} \cdot \frac{1}{4(\frac{1}{4}+t)} \text{ for } t > -\frac{1}{4}$$

وزارة التعليم العالي والبحث العلمي

جامعة ديالى

كلية تربية المقداد / قسم الرياضيات

محاضرات مادة

الإحصاء والاحتمالية

للعام الدراسي (2023-2024)

المرحلة الثالثة

Chapter Five

مدرس المادة: م.م هند إبراهيم محمد

Notes (Pr. dist. = Pr. fun. = f) Probability Distribution of Two Random Variables

(I) Joint Probability Density Function (J.P.d.f.)

Def. Let X and Y be two c.r.v.'s. A function f defined over XY -plane is a joint p.d.f. of X and Y if for a subset $R \in XY$ -plane, then:

$$P[(x,y) \in R] = \iint_R f(x,y) dx dy$$

if $R = \{ (x,y) : a \leq x \leq b \text{ and } c \leq y \leq d \}$

Then

$$\begin{aligned} P[(x,y) \in R] &= \int_c^d \int_a^b f(x,y) dx dy \\ &= \int_a^b \int_c^d f(x,y) dy dx \end{aligned}$$

Note:

- ① $f(x,y)$ is a solid over XY -plane. $\longrightarrow f(x)$ is a curve
- ② R can be $\square, \square, \Delta, O, \bigcirc, \dots \longrightarrow (a,b) \xrightarrow{a \quad b}$
- ③ $P[(x,y) \in R]$ is the volume inside solid of $f(x,y)$. $\longrightarrow$ area under the curve

Properties of Joint p.d.f.

① A joint p.d.f. $f(x,y)$ satisfies two conditions:

- a) $f(x,y) \geq 0 \quad \forall (x,y) \in R$
- b) $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y) dx dy = 1$

② Also :

①

REVIEW

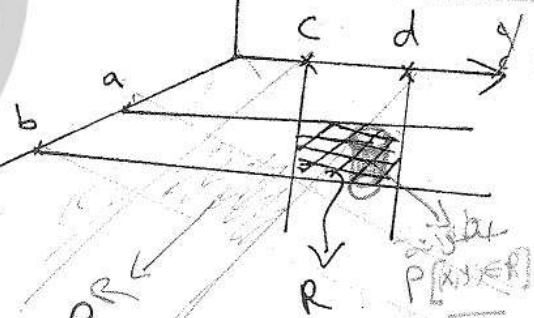
ch.5 (Pr. dist. of X,Y)

$f(x,y)$ is J.P.d.f.
① $f(x,y) \geq 0 \quad \forall (x,y) \in R$

$Z = f(x,y)$
② $\iint_R f(x,y) dx dy = 1$

$f_1(x) = \int_c^d f(x,y) dy$

$f_2(y) = \int_a^b f(x,y) dx$



- Regions R
- ① $R = \{ (x,y) : 0 < x < 1, 0 < y < 1 \}$
 - ② $R = \{ (x,y) : 0 < x < y < 1 \}$
 - ③ $R = \{ (x,y) : x^2 + y^2 \leq 1 \}$

∴ $f(x, y)$ is a function of x and y , from $f(x, y)$, we can find a function of x alone $f_1(x)$ and a function of y alone $f_2(y)$. s.t.

$f_1(x)$ is called Marginal p.d.f. of x

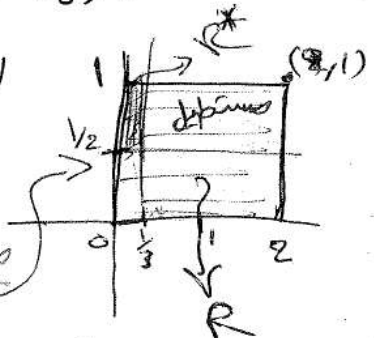
where $f_1(x) = \int_{\square} f(x, y) dy$; limits of integral follows limits of y

$f_2(y)$ is called Marginal p.d.f. of y

where $f_2(y) = \int_{\square} f(x, y) dx$; limits of integral follows limits of x

Example 8 Given a joint p.d.f. (J.P.d.f.) $f(x, y)$:

$$f(x, y) = \begin{cases} Kx^2y & \text{for } 0 < x < 2, 0 < y < 1 \\ 0 & \text{o.w.} \end{cases}$$



- (a) Find the value of K , (b) Find $P(0 < x < \frac{1}{2}, \frac{1}{2} < y < 1)$, (c) Find $P(x+y \geq 1)$, (d) Find $f_1(x)$ and $f_2(y)$

Sol. By Cond. (2)

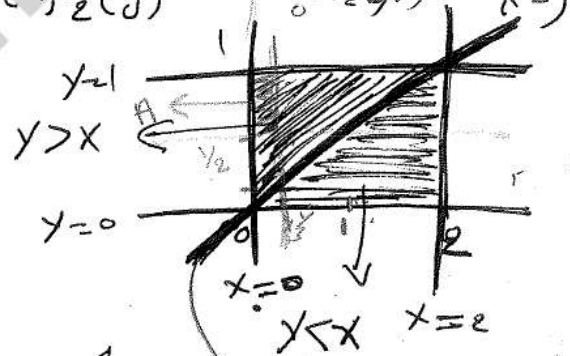
$$\textcircled{a} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = 1$$

$$R = \{(x, y) : 0 < x < 2, 0 < y < 1\}$$

$$\int_0^1 \int_0^2 Kx^2y dx dy = 1 \Rightarrow \int_0^1 Ky \left(\frac{x^3}{3} \Big|_0^2 \right) dy = 1$$

$$\frac{1}{3} \int_0^1 Ky (8-0) dy = 1 \Rightarrow \frac{8K}{3} \int_0^1 y dy = 1$$

$$\frac{8K}{3} \left[\frac{y^2}{2} \Big|_0^1 \right] = 1 \Rightarrow \frac{8K}{6} [1-0] = 1 \Rightarrow \frac{4}{3} K = 1 \Rightarrow \boxed{K = \frac{3}{4}}$$



الخط (المحور) لا يملك كثافة مشتركة (Joint density function) ولا يملك كثافة حدية (Marginal density function)

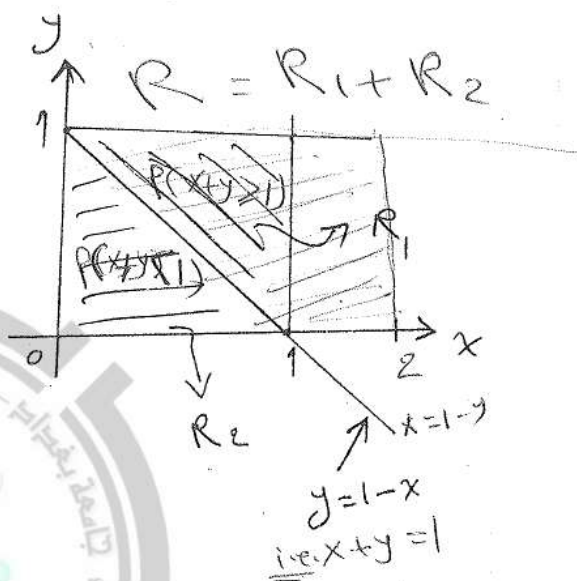
$$(b) f(x,y) = \begin{cases} \frac{3}{4}x^2y & 0 < x < 2, 0 < y < 1 \\ 0 & \text{o.w} \end{cases}$$

$$P(0 < x < \frac{1}{2}, \frac{1}{2} < y < 2) = \int_{\frac{1}{2}}^1 \int_0^{\frac{1}{2}} \frac{3}{4}x^2y \, dx \, dy = \dots = \frac{1}{32(9)} = \frac{1}{288}$$

$$(c) P(x+y \geq 1) = ? \quad (d) P(x+y < 1) = ?$$

$$\text{consider } (x+y)=1 \Rightarrow y=1-x$$

x	y=1-x
0	1
1	0



$P(x+y \geq 1)$ = Volum of $f(x,y)$ over R_1

$P(x+y < 1)$ = Volum of $f(x,y)$ over R_2

$$P(x+y \geq 1) = 1 - P(x+y < 1)$$

$$P(x+y < 1) = \int_0^1 \int_0^{1-x} f(x,y) \, dy \, dx$$

$$= \int_0^1 \left[\int_0^{1-x} \frac{3}{4}x^2y \, dy \right] dx = \int_0^1 \frac{3}{4}x^2 \frac{y^2}{2} \Big|_0^{1-x} dx$$

$$= \int_0^1 \frac{3}{8}x^2[(1-x)^2 - 0] dx$$

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

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

$$= \frac{3}{8} \left[\left(\frac{1}{3} - \frac{1}{2} + \frac{1}{5} \right) - (0-0+0) \right] = \frac{3}{8} \left[\frac{10-15+6}{30} \right]$$

$$= \frac{3}{8} \left(\frac{1}{30} \right) = \frac{1}{80}$$

$$P(x+y < 1) = \frac{1}{80} \Rightarrow P(x+y \geq 1) = 1 - \frac{1}{80} = \frac{79}{80} \in [0,1]$$

(d) To find $f_1(x)$ & $f_2(y)$

$$f_1(x) = \int_0^1 f(x,y) \, dy$$

$$f_1(x) = \int_0^1 \frac{3}{4} x^2 y dy = \frac{3}{4} x^2 \frac{y^2}{2} \Big|_0^1 = \frac{3}{8} x^2 [1-0] = \frac{3}{8} x^2$$

$$f_1(x) = \begin{cases} \frac{3}{8} x^2 & \text{for } 0 \leq x \leq 2 \\ 0 & \text{o.w.} \end{cases}$$

$$f_2(y) = \int_0^2 f(x,y) dx$$

$$= \int_0^2 \frac{3}{4} x^2 y dx = \frac{3}{4} y \frac{x^3}{3} \Big|_0^2 = \frac{1}{4} y [8-0] = 2y$$

$$= 2y$$

$$f_2(y) = \begin{cases} 2y & \text{for } 0 \leq y \leq 1 \\ 0 & \text{o.w.} \end{cases}, P(Y < \frac{1}{3}) = \int_0^{\frac{1}{3}} 2y dy = \frac{2}{2} y^2 \Big|_0^{\frac{1}{3}} = \frac{1}{9}$$

Example 4) Given a J.P.d.f.

$$f(x,y) = \begin{cases} 2 & \text{for } 0 \leq x \leq y \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

Find $f_1(x)$, $f_2(y)$, $P(X > \frac{1}{2})$, $P(Y < \frac{1}{2})$, $P(X < \frac{1}{2})$, Find $P(X > \frac{1}{2})$

Sol. $P(X > \frac{1}{2}, Y < \frac{1}{2}) = \iint_{R} f(x,y) dy dx = \int_{\frac{1}{2}}^1 \int_{\frac{1}{2}}^y 2 dy dx = \int_{\frac{1}{2}}^1 (2y - 1) dx = \frac{1}{4}$

$$R = \{ (x,y) : 0 \leq x \leq y \leq 1 \}$$

$$f_1(x) = \int_x^1 2 dy \text{ for } 0 \leq x \leq 1$$

$$= 2y \Big|_x^1 = 2[1-x]$$

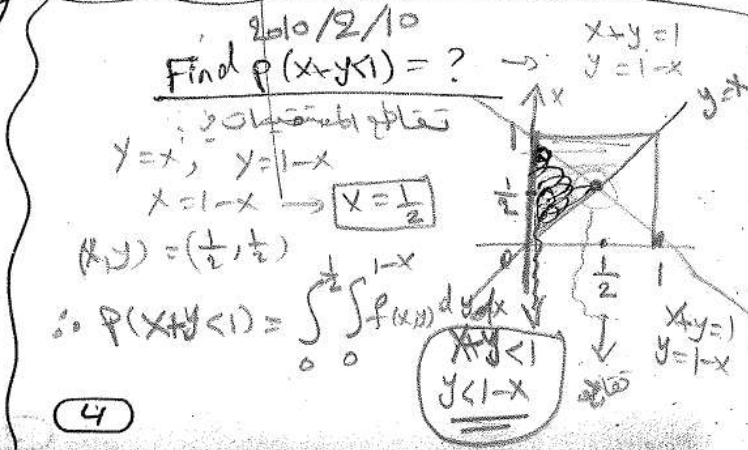
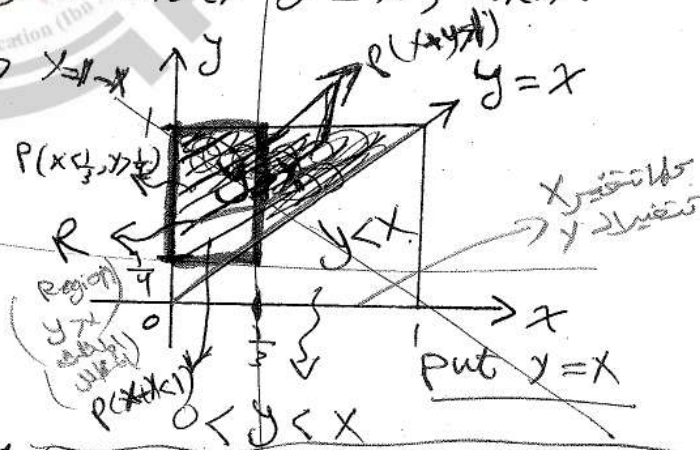
$$f_1(x) = \begin{cases} 2(1-x) & \text{for } 0 \leq x \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

Hint: Prove that f_1 & f_2 are pr. func. o.w.

$$f_2(y) = \int_0^y 2 dx \text{ for } 0 \leq y \leq 1$$

$$= 2x \Big|_0^y = 2y$$

$$f_2(y) = \begin{cases} 2y & \text{for } 0 \leq y \leq 1 \\ 0 & \text{o.w.} \end{cases}$$



$$P(X > \frac{1}{3}) = \int_{\frac{1}{3}}^1 f_1(x) dx$$

$$= \int_{\frac{1}{3}}^1 2(1-x) dx = \frac{8}{18}$$

$P(X < \frac{1}{3})$ calc = $\int_0^{\frac{1}{3}} f_1(x) dx$ calc = $\int_0^{\frac{1}{3}} 2(1-x) dx$
 $P(Y > \frac{1}{4})$ implies $\int_{\frac{1}{4}}^1 f_2(y) dy$

Find $P(X+Y \geq 1) = ?$

$$P(Y < \frac{1}{2}) = \int_0^{\frac{1}{2}} 2y dy = \dots = \frac{1}{4}$$

$$P(X < \frac{1}{3}, Y > \frac{1}{4}) = \int_{\frac{1}{4}}^{\frac{1}{2}} \int_{\frac{1}{3}}^1 2 dx dy = \dots = \frac{1}{2}$$

II Joint Probability Mass Function (J.P.M.F.)

Def. Let X and Y be two d.r.v. A function $f(x,y)$ is a joint p.m.f. of X and Y iff

$$f(x,y) = P(X=x, Y=y) \quad \forall (x,y) \in \mathbb{R}$$

Consider f.p.
 $f(x,y)$ is p.m.f.
 $f(x) = P(X=x)$
 $f(2) = P(X=2)$

and satisfies two conditions:-

(a) $f(x,y) \geq 0 \quad \forall (x,y) \in \mathbb{R}$

(b) $\sum_{y \in \mathbb{R}} \sum_{x \in \mathbb{R}} f(x,y) = 1$

Also,

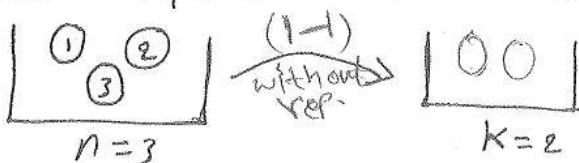
$f_1(x) = \sum_{y \in \mathbb{R}} f(x,y)$, limit of summation follows limit of y .

$f_2(y) = \sum_{x \in \mathbb{R}} f(x,y)$, limit of summation follows limit of x .

where $f_1(x)$ and $f_2(y)$ are called Marginal p.m.f. of X and Y respectively.

Example 3 A box has (3) balls (1), (2), (3). Choose two balls one by one without replac. & find $P(x,y)$.

Sol.



$$S = \left\{ \begin{matrix} (1,2) & (2,1) & (3,1) \\ (1,3) & (2,3) & (3,2) \\ (3,3) & (2,2) & (1,1) \end{matrix} \right\}; S \text{ has "6" elts.}$$

OR A box has (3) balls which are numbered 1, 2, 3.

$P_2^3 = \frac{3!}{1!} = 6 = n(s)$
 $X = 1^{st} \text{ chosen ball ; } x = 1, 2, 3$
 $Y = 2^{nd} \text{ chosen ball ; } y = 1, 2, 3 \Rightarrow X \neq Y$

$P(X=2, Y=1) = \frac{1}{6} = f(2,1)$

$P(X=3, Y=2) = \frac{1}{6} = f(3,2)$

$f(x,y) = P(X=x; Y=y) = \frac{1}{6}$

$f(x,y) = \begin{cases} \frac{1}{6} & \text{for } x=1,2,3; y=1,2,3; x \neq y \\ 0 & \text{o.w.} \end{cases}$

$\text{Find: } f_1(x) = \sum_y f(x,y) = \frac{2}{6}, x=1,2,3 \quad f_2(y) = \sum_x f(x,y) = \frac{2}{6}, y=1,2,3$

Example 4 Given a J.P.M.F. $f(x,y)$

$f(x,y) = \begin{cases} \frac{1}{21}(x+y) & \text{for } x=1,2,3 \\ & y=1,2 \\ 0 & \text{o.w.} \end{cases}$

Find $P(X \geq 2, Y \leq 2), f_1(x), f_2(y)$

$\text{Sol. } P(X \geq 2, Y \leq 2) = \sum_{y=1}^2 \sum_{x=2}^3 \frac{1}{21}(x+y)$
 $= \frac{1}{21} \sum_{y=1}^2 [(2+y) + (3+y)]$
 $= \frac{1}{21} \sum_{y=1}^2 [2y+5] = \frac{1}{21} [(2(1)+5) + 2(2)+5]$
 $= \frac{1}{21} [7+9] = \frac{16}{21}$

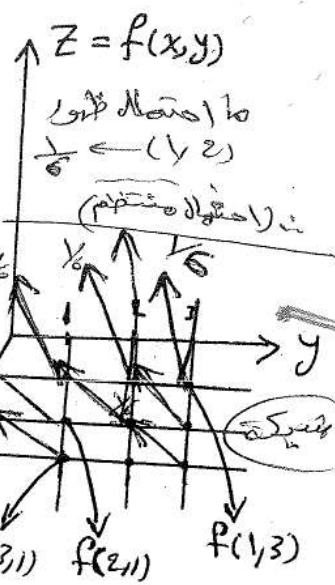
$P(X=3, Y=1) = f(3,1) = \frac{1}{21}(3+1) = \frac{4}{21}$

$f_1(x) = \sum_y f(x,y)$

$f_1(x) = \sum_{y=1}^2 \frac{1}{21}(x+y) = \frac{1}{21} [(x+1) + (x+2)]$
 $= \frac{1}{21} [3+2x]$

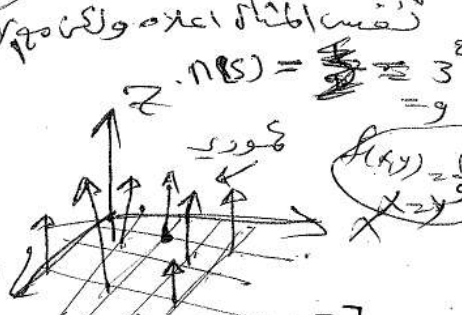
$f_1(x) = \begin{cases} \frac{1}{21} [3+2x] & \text{for } x=1,2,3 \\ 0 & \text{o.w.} \end{cases}$

Graph of $f(x,y)$



$P(X=1) = f_1(1) = \frac{2}{6}$ (3-dim.)
 $\sum_x f(x,y) = \sum_{x=1}^3 \frac{1}{6} = \frac{3}{6} = \frac{1}{2}$

$P(X \geq 2) = \sum_{x=2}^3 f_1(x) = f_1(2) + f_1(3) = \frac{2}{6} + \frac{2}{6} = \frac{4}{6} = \frac{2}{3}$
 $P(Y \leq 2) = ? = \frac{2}{3} + \frac{2}{3} = 1$

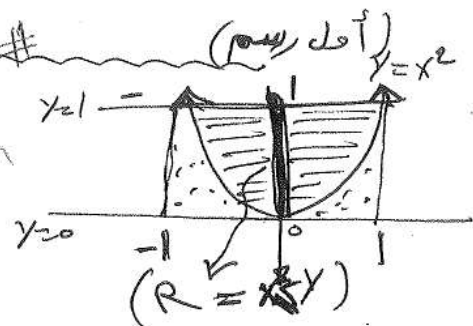


$P(X \geq 2, Y \leq 2) = \sum_{x=2}^3 \sum_{y=1}^2 \frac{1}{6}$
 $= \sum_{x=2}^3 (\frac{1}{6} + \frac{1}{6})$
 $= (\frac{2}{6}) [\frac{1}{6} + \frac{1}{6}]$

$$\begin{aligned}
 f_2(y) &= \sum_x f(x,y) \\
 &= \sum_{x=1}^3 \frac{1}{2!} (x+y) \\
 &= \frac{1}{2!} [(1+y) + (2+y) + (3+y)] \\
 &= \frac{1}{2!} (6+3y) = \frac{3}{2!} (y+2) = \frac{1}{x} (y+2)
 \end{aligned}$$

$$f_2(y) = \begin{cases} \frac{1}{x} (y+2) & \text{for } y=1, 2 \\ 0 & \text{o.w.} \end{cases}$$

Example 8 Given a J.P.d.f.
 $f(x,y) = \begin{cases} cx^2y & \text{for } (x^2 \leq y \leq 1) \\ 0 & \text{o.w.} \end{cases}$ Region



(a) Find the value of c (b) Find $P(X \leq Y)$

Sol.

$$R = \{(x,y) : 0 \leq x^2 \leq y \leq 1\} = \begin{cases} 0 \leq x^2 \leq y, & x^2 \leq y \leq 1 \\ 0 \leq x^2 \leq 1, & 0 \leq y \leq 1 \end{cases}$$

suppose $y = x^2 \Rightarrow x = \pm \sqrt{y}$

$$0 \leq x^2 \leq y$$

$$|x| \leq \sqrt{y}$$

$$-\sqrt{y} \leq x \leq \sqrt{y}$$

$$x^2 \leq 1$$

$$|x| \leq 1$$

$$-1 \leq x \leq 1$$

$$0 \leq x^2 \leq y < 1$$

$$0 \leq y < 1$$

$$y = x^2$$

$$x = \pm \sqrt{y}$$

$$by \sqrt{x^2} = |x|$$

$$R = \begin{cases} -\sqrt{y} \leq x \leq \sqrt{y} & ; x^2 \leq y \leq 1 \\ -1 \leq x \leq 1 & ; 0 \leq y \leq 1 \end{cases}$$

x	y = x^2
0	0
$\pm \frac{1}{2}$	$\frac{1}{4}$
± 1	1

Consider $y = x^2$

$y = x^2, y = x$: لإيجاد نقاط التقاطع

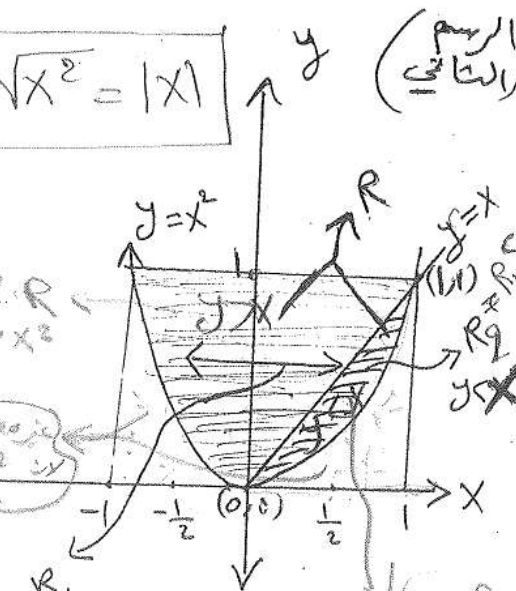
$$x^2 = x$$

$$x^2 - x = 0$$

$$x(x-1) = 0$$

$$x = 0, x = 1$$

$$\Rightarrow (0,0), (1,1)$$



R_1 ($x \leq y$)

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

R

R_2 ($x > y$)

$P(X \leq Y) = 1 - P(X > Y)$

لذلك

الحل

$P(X \leq Y) = 1 - P(X > Y)$

(a) By Conda. (2)

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

$$\int_{-1}^1 \int_{x^2}^1 c x^2 y dy dx = 1$$

$$1 = \frac{c}{2} \int_{-1}^1 x^2 y^2 \Big|_{x^2}^1 dx = \frac{c}{2} \int_{-1}^1 x^2 (1 - x^4) dx$$

$$1 = \frac{c}{2} \int_{-1}^1 [x^2 - x^6] dx = \frac{c}{2} \left[\frac{x^3}{3} - \frac{x^7}{7} \right]_{-1}^1$$

$$1 = \frac{c}{2} \left[\left(\frac{1}{3} - \frac{1}{7} \right) - \left(-\frac{1}{3} - \frac{-1}{7} \right) \right]$$

$$1 = \frac{c}{2} \left[\left(\frac{1}{3} - \frac{1}{7} \right) + \left(\frac{1}{3} - \frac{1}{7} \right) \right]$$

$$1 = \frac{c}{2} \left(\frac{2}{3} - \frac{2}{7} \right) = c \left(\frac{1}{3} - \frac{1}{7} \right) = c \left(\frac{7-3}{21} \right) = \frac{4}{21} c$$

$$\Rightarrow \boxed{c = \frac{21}{4}}$$

$$\therefore f(x,y) = \begin{cases} \frac{21}{4} x^2 y & \text{for } x^2 \leq y \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

(الرسم الثاني)

(b) To find $P(X \leq Y)$:

$$R = \left\{ \begin{array}{l} -\sqrt{y} \leq x \leq \sqrt{y}, \quad x^2 \leq y \leq 1 \\ -1 \leq x \leq 1, \quad 0 \leq y \leq 1 \end{array} \right\}$$

$P(X \leq Y)$ = volum of $f(x,y)$ over R ,

but $P(Y \geq X) = 1 - P(Y < X)$

$P(Y < X)$ = volum of $f(x,y)$ over R^c

$$P(Y < X) = \int_0^1 \int_{x^2}^x \frac{21}{4} x^2 y dy dx$$

$$\left(\begin{array}{l} x=0 \\ x=1 \end{array} \right) \leftarrow \begin{array}{l} y \\ x \end{array} = \frac{21}{8} \int_0^1 x^2 y^2 \Big|_{x^2}^x dx$$

$$= \frac{21}{8} \int_0^1 x^2 [x^2 - x^4] dx = \frac{21}{8} \int_0^1 [x^4 - x^6] dx$$

$$\begin{aligned}
 &= \frac{21}{8} \left[\frac{x^5}{5} - \frac{x^7}{7} \right]_0^1 \\
 &= \frac{21}{8} \left[\left(\frac{1}{5} - \frac{1}{7} \right) - (0-0) \right] \\
 &= \frac{21}{8} \left[\frac{7-5}{35} \right] = \frac{21}{8} \cdot \frac{2}{35} = \left(\frac{3}{20} \right)
 \end{aligned}$$

$$\begin{aligned}
 \therefore P(X \leq Y) &= 1 - P(X > Y) \approx \int_0^1 (x^2 - x) dx \quad (\text{احتمال مضاعف}) \\
 &= 1 - P(Y < X) \\
 &= 1 - \frac{3}{20} = \left(\frac{17}{20} \right)
 \end{aligned}$$

© To find $f_1(x)$ and $f_2(y)$

$$R = \left\{ \begin{array}{l} -\sqrt{y} \leq x \leq \sqrt{y}, \quad x^2 \leq y \leq 1 \\ -1 \leq x \leq 1, \quad 0 \leq y \leq 1 \end{array} \right\}$$

$$\begin{aligned}
 f_1(x) &= \int_{x^2}^1 \frac{21}{4} x^2 y dy \quad \text{for } -1 \leq x \leq 1 \\
 &= \frac{21}{4} x^2 \frac{y^2}{2} \Big|_{x^2}^1 = \frac{21}{8} x^2 (1 - x^4)
 \end{aligned}$$

$$f_1(x) = \begin{cases} \frac{21}{8} x^2 (1 - x^4) & \text{for } -1 \leq x \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

$$f_2(y) = \int_{-\sqrt{y}}^{\sqrt{y}} \frac{21}{4} x^2 y dx \quad \text{for } 0 \leq y \leq 1$$

$$= \frac{21}{4} y \frac{x^3}{3} \Big|_{-\sqrt{y}}^{\sqrt{y}} = \frac{7}{4} y \left[y^{\frac{3}{2}} + y^{\frac{3}{2}} \right] = \frac{7}{2} y^{\frac{5}{2}}$$

$$f_2(y) = \begin{cases} \frac{7}{2} y^{\frac{5}{2}} & 0 \leq y \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

$$\boxed{P(X+Y < 1) = ?} \quad (\text{H.w.})$$

Stochastic Independence

الاستقلال الاحصائي

Def: X and Y are stochastically independence iff:

$$f_1(x) f_2(y) = f(x, y) \quad \forall (x, y)$$

مقارنة
A & B are indep. events:
 $P(A)P(B) = P(AB)$

Denoted by (s. indep.)

شرح بـ (1/2 ساعة)
الوقت
للدراسة

Example: Given a J.P.M.F

$x \backslash y$	1	2	3	4	$f_1(x)$ الوان
1	$f(1,1) = .1$	$f(1,2) = 0$	$f(1,3) = .1$	$f(1,4) = 0$	$f_1(1) = .1 + .1 = .2$
2	$f(2,1) = .3$	$f(2,2) = 0$	$f(2,3) = .1$	$f(2,4) = .2$	$f_1(2) = .3 + .1 + .2 = .6$
3	$f(3,1) = 0$	$f(3,2) = .2$	$f(3,3) = 0$	$f(3,4) = 0$	$f_1(3) = .2$
$f_2(y)$ الوان	$f_2(1) = .4$	$f_2(2) = .2$	$f_2(3) = .2$	$f_2(4) = .2$	$\sum_{x=1}^3 f_1(x) = \sum_{y=1}^4 f_2(y) = 1$

$f(x, y) = P(X=x, Y=y)$

a. T.P. $f_1(x) \cdot f_2(y) = f(x, y) \quad \forall (x, y)$

Note $\sum_x f_1(x) = 1 \rightarrow f_1(x)$ is p.m.f. of x
 $\sum_y f_2(y) = 1 \rightarrow f_2(y)$ is p.m.f. of y (Prove that)

$$f_1(1) \cdot f_2(1) = (.2)(.4) = .08$$

$$f(1,1) = .1$$

$$\therefore f_1(1) f_2(1) \neq f(1,1) \text{ \& } f_1(2) f_2(2) \neq f(2,2)$$

$\therefore X$ and Y are not s. indep.

b. Find $P(X \geq 2, Y \leq 3)$

$$P(X \geq 2, Y \leq 3) = \sum_{y=1}^3 \sum_{x=2}^3 f(x, y) = \sum_{y=1}^3 [f(2, y) + f(3, y)]$$

$$= [f(2, 1) + f(2, 2) + f(2, 3)] + [f(3, 1) + f(3, 2) + f(3, 3)]$$

$$= .03 + 0 + .1 + 0 + .2 + 0$$

$$= .6$$

c. Find $P(X \geq 2) = \sum_{x=2}^3 f_1(x)$

$$= f_1(2) + f_1(3)$$

$$= 0.6 + 0.2 = 0.8$$

d. $P(Y \leq 3) = \sum_{y=1}^3 f_2(y)$

$$= f_2(1) + f_2(2) + f_2(3)$$

$$= .4 + .2 + .2 = .8$$

Uniform Distribution of two R.V.'s

تعريف
إذا كان X و Y متغيرين عشوائيين
مستقلين ومتوزعين توزيعاً
متساوياً على المنطقة R فإن
المتكافئة لـ $f(x, y)$ هي

Def. If a c.r.v. X and Y have a uniform distribution on a region R , then the joint p.d.f. of X and Y is:

$$f(x, y) = \frac{1}{\text{Area}(R)} \rightarrow \text{(constant)}$$

Example 6: X and Y have a uniform dist. on a triangle Δ with vertices $(0, 0)$, $(0, 1)$ and $(1, 1)$. Find $f(x, y)$.

Sol.

R is a triangle Δ

$$y \geq x \quad x \geq 0$$

$$y \leq 1$$

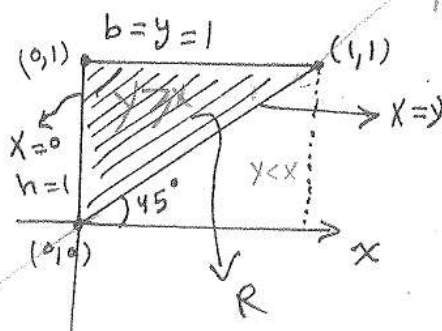
$$\Rightarrow 0 \leq x \leq y \leq 1$$

$$R = \{(x, y); 0 \leq x \leq y \leq 1\}$$

$$\text{Area}(\Delta) = \frac{1}{2} \cdot h \cdot b$$

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

II



X و Y متغيرين عشوائيين
مستقلين ومتوزعين توزيعاً
متساوياً على المنطقة R فإن
المتكافئة لـ $f(x, y)$ هي

$$f(x,y) = \frac{1}{\text{Area}(R)} = \frac{1}{\frac{1}{2}} = \begin{cases} 2 & \text{for } 0 \leq x \leq y \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

Example 7 A point (x,y) is chosen from a region R which is bounded by a curve $y=x^2$ and a line $y=x$, find $f(x,y)$.

Sol. Sketch $y=x$ and $y=x^2$

A point $(x,y) \in R$ must be above the curve $y=x^2$ and under the line $y=x$, then $x^2 \leq y \leq x$.

Also that point must be between $0 \leq x \leq 1, 0 \leq y \leq 1$
 $(x^2=x \Rightarrow x=0, x=1 \Rightarrow y=0, y=1)$

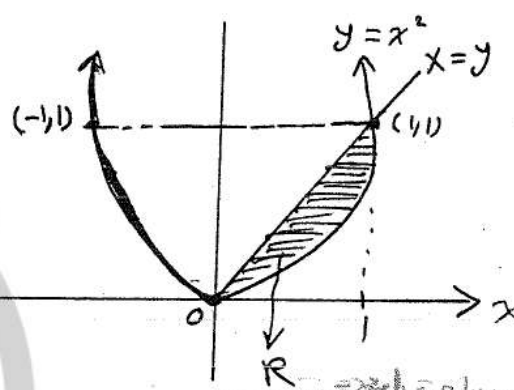
$$R = \{(x,y); 0 \leq x^2 \leq y \leq x \leq 1\}$$

$$\text{Area}(R) = \int_0^1 \int_{x^2}^x dy dx = \int_0^1 (x - x^2) dx$$

$$= \int_0^1 (x - x^2) dx$$

$$= \left(\frac{x^2}{2} - \frac{x^3}{3} \right) \Big|_0^1 = \frac{1}{2} - \frac{1}{3} = \frac{1}{6}$$

$$f(x,y) = \frac{1}{1/6} = \begin{cases} 6 & \text{for } 0 \leq x^2 \leq y \leq x \leq 1 \\ 0 & \text{o.w.} \end{cases}$$



$$\int_0^1 (x^2 - x) dx = \int_0^1 (x - x^2) dx$$

Example 8 A point (x,y) is chosen from a region R where

$$R = \{(x,y); x^2 + y^2 \leq 1\}$$

a. find $f(x,y)$

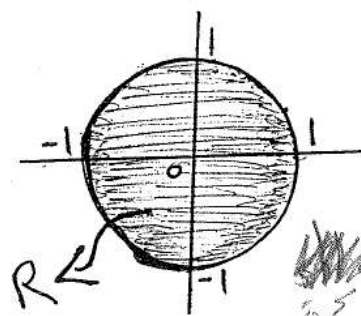
b. find $f_1(x)$ & $f_2(y)$

Sol.

a. Let $x^2 + y^2 = 1 \Rightarrow r=1$ نصف دائرة

$$\text{Area}(R) = \pi r^2 = \pi$$

$$\text{then } f(x,y) = \begin{cases} \frac{1}{\pi} & \text{for } x^2 + y^2 \leq 1 \\ 0 & \text{o.w.} \end{cases}$$



Sphere
كرة
Kura

b. We must find $f_1(x)$ & $f_2(y)$:

$$f_1(x) = \int_{-\sqrt{1-x^2}}^{\sqrt{1-x^2}} \frac{1}{\pi} dy \quad \text{for } -1 \leq x \leq 1$$

$$= \frac{1}{\pi} y \Big|_{-\sqrt{1-x^2}}^{\sqrt{1-x^2}} = \frac{1}{\pi} [\sqrt{1-x^2} + \sqrt{1-x^2}] = \frac{2}{\pi} \sqrt{1-x^2} \quad \text{for } -\sqrt{1-x^2} \leq y \leq \sqrt{1-x^2}$$

$$f_2(y) = \int_{-\sqrt{1-y^2}}^{\sqrt{1-y^2}} \frac{1}{\pi} dx \quad \text{for } -1 \leq y \leq 1$$

$$= \frac{1}{\pi} x \Big|_{-\sqrt{1-y^2}}^{\sqrt{1-y^2}} = \frac{2}{\pi} \sqrt{1-y^2} \quad \text{for } -\sqrt{1-y^2} \leq x \leq \sqrt{1-y^2}$$

$$f_1(x) \cdot f_2(y) = \frac{2}{\pi} \cdot \frac{2}{\pi} \sqrt{(1-y^2)(1-x^2)} = \frac{4}{\pi^2} \sqrt{(1-y^2)(1-x^2)}$$

$$f(x,y) \neq f_1(x) f_2(y) \quad (\text{since } f(x,y) = \frac{1}{\pi})$$

* Conditional Function and Conditional probability

Def. Let X and Y be two r.v.'s $f(y|x)$ denoted the conditional p.d.f or p.m.f. of y given $X=x$.

$f(x|y)$ denoted the conditional p.d.f. or p.m.f. of x given $y=y$.

where

$f(y x) = \frac{f(x,y)}{f_1(x)} ; f_1(x) \neq 0$
$f(x y) = \frac{f(x,y)}{f_2(y)} ; f_2(y) \neq 0$

(Joint Probability)

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

* Note:

$$P(a < x \leq b | y=c) = \int_a^b f(x|y=c) dx$$

$$= \sum_{x=a}^b f(x|y=c)$$

$$P(c < y < d | x=a) = \int_c^d f(y|x=a) dy$$

$$= \sum_{y=c}^d f(y|x=a)$$

✓ Example 8 Given a J.P.d.f.

$$f(x,y) = \begin{cases} 8xy & \text{for } 0 < x < y < 1 \\ 0 & \text{o.w.} \end{cases}$$

(1) Find $f(x|y)$ & $f(y|x)$

(2) Find $P(x > \frac{1}{4} | y = \frac{1}{2})$, $P(y < \frac{1}{4} | x = \frac{1}{2})$.

Sol. $R = \{(x,y); 0 < x < y < 1\}$

$$R = \left\{ \begin{array}{ll} 0 < x < y & x < y < 1 \\ 0 < x < 1 & 0 < y < 1 \end{array} \right\}$$

(1) we must find $f_1(x)$ and $f_2(y)$

$$f_1(x) = \int_x^1 8xy dy = 8x \left[\frac{y^2}{2} \right]_x^1 = 4x(1-x^2)$$

$$= 4x(1-x^2) \text{ for } 0 < x < 1$$

$$f_2(y) = \int_0^y 8xy dx = \frac{8}{2} y x^2 = 4y^3 \text{ for } 0 < y < 1$$

$$f(x|y) = \frac{f(x,y)}{f_2(y)} = \frac{8xy}{4y^3} = \begin{cases} \frac{2x}{y^2} & \text{for } \begin{cases} 0 < x < y \\ 0 < y < 1 \end{cases} \\ 0 & \text{o.w.} \end{cases}$$

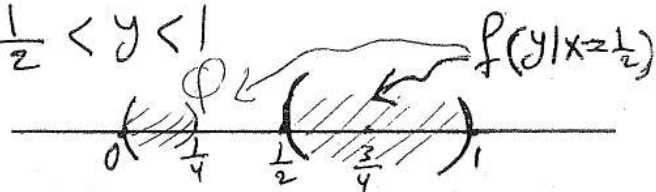
$$f(y|x) = \frac{f(x,y)}{f_1(x)} = \frac{8xy}{4x(1-x^2)} = \frac{2y}{1-x^2}$$

$$\therefore f(y|x) = \begin{cases} \frac{2y}{1-x^2} & \text{for } x < y < 1 \text{ \& } 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

وہاں $\leftarrow$
 y (یا x) کے لیے
 x (یا y) کے لیے
 x (یا y) کے لیے
 x (یا y) کے لیے

$$(2) P(y < \frac{1}{4} | x = \frac{1}{2}) = \int_{-\infty}^{\frac{1}{4}} f(y|x = \frac{1}{2}) dy = \int_0^{\frac{1}{4}} 0 dy = 0$$

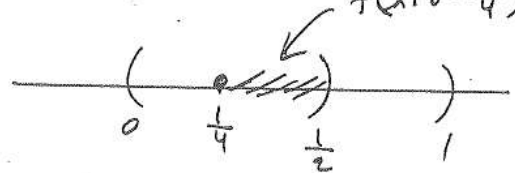
$$x < y < 1, x = \frac{1}{2} \Rightarrow \frac{1}{2} < y < 1$$



$$P(X > \frac{1}{4} | Y = \frac{1}{2}) = \int_0^y f(x|y=\frac{1}{2}) dx$$

← (نعوض $y = \frac{1}{2}$ في $f(x|y)$ ثم نأخذ التكامل)
 $f(x|y=\frac{1}{2})$

$$P(X > \frac{1}{4} | Y = \frac{3}{4}) = \int_{\frac{1}{4}}^y \frac{2x}{y} dx$$



$$0 < x < y, y = \frac{3}{4} \Rightarrow \frac{1}{4} < x < \frac{3}{4}$$

$$0 < x < y, y = \frac{1}{2} \Rightarrow 0 < x < \frac{1}{2}$$

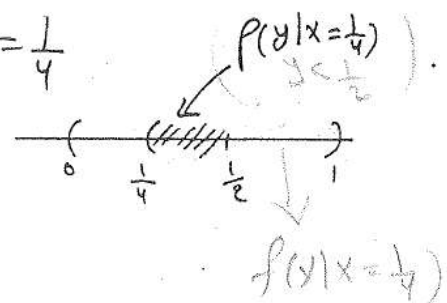
$$= \int_{\frac{1}{4}}^{\frac{1}{2}} \frac{2x}{(\frac{1}{2})^2} dx = 4 \cdot 2 \cdot \frac{x^2}{2} \Big|_{\frac{1}{4}}^{\frac{1}{2}}$$

$$= 4 \left[\frac{1}{4} - \frac{1}{16} \right] = 4 \cdot \left(\frac{3}{16} \right) = \frac{3}{4}$$

$$P(Y < \frac{1}{2} | X = \frac{1}{4}) = ?$$

$$x < y < 1, x = \frac{1}{4}$$

$$\frac{1}{4} < y < 1$$



$$P(Y < \frac{1}{2} | X = \frac{1}{4}) = \int_x^{\frac{1}{2}} f(y|x=\frac{1}{4}) dy$$

$$= \int_x^{\frac{1}{2}} \frac{2y}{1-x^2} dy = \int_{\frac{1}{4}}^{\frac{1}{2}} \frac{2y}{1-(\frac{1}{4})^2} dy$$

$$= \int_{\frac{1}{4}}^{\frac{1}{2}} \frac{2y}{15/16} dy = \frac{16}{15} \cdot 2 \cdot \frac{y^2}{2} \Big|_{\frac{1}{4}}^{\frac{1}{2}}$$

$$= \frac{16}{15} \left(\frac{1}{4} - \frac{1}{16} \right) = \frac{16}{15} \cdot \frac{3}{16} = \frac{1}{5}$$

Example 50 If $X \sim \text{unif}(0,1)$ and
 $f(y|x) = \begin{cases} \frac{1}{1-x} & \text{for } x < y < 1, 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$

(a) find $f(x,y)$?

$$\because X \sim \text{unif}(0,1)$$

$$f_1(x) = \frac{1}{1-0} = 1 \Rightarrow f_1(x) = \begin{cases} 1 & \text{for } 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$f(y|x) = \frac{f(x,y)}{f_1(x)}$$

$$f(x,y) = f_1(x) \cdot f(y|x)$$

$$= (1) \left(\frac{1}{1-x} \right) = \begin{cases} \frac{1}{1-x} & \text{for } x < y < 1, 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$\therefore f(y|x) = f(x,y)$$

$$b. \text{ Find } P\left(X > \frac{1}{2} \mid Y = \frac{3}{4}\right) = \int_0^y f(x|y = \frac{3}{4}) dx$$

$$f(x|y) = \frac{f(x,y)}{f_2(y)}$$

$$f_2(y) = \int_0^y f(x,y) dx = \int_0^y \frac{1}{1-x} dx = -\ln(1-x) \Big|_0^y$$

$$= -[\ln(1-y) - \ln(1-0)]$$

$$= -\ln(1-y) \text{ for } 0 < y < 1$$

$$\therefore f_2(y) = -\ln(1-y) \text{ for } 0 < y < 1$$

$$\therefore f(x|y) = \frac{\frac{1}{1-x}}{-\ln(1-y)}$$

$$f(x|y) = \begin{cases} \frac{1}{-\ln(1-y)} \cdot \frac{1}{1-x} & \text{for } 0 < x < y \text{ \& } 0 < y < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$P\left(X > \frac{1}{2} \mid Y = \frac{3}{4}\right) = \int_{\frac{1}{2}}^y f(x|y = \frac{3}{4}) dx$$

$$\left(0 < X < Y, Y = \frac{3}{4}\right) = \int_{\frac{1}{2}}^{\frac{3}{4}} \frac{-1}{\ln(1-y)} \cdot \frac{1}{1-x} dx = \int_{\frac{1}{2}}^{\frac{3}{4}} \frac{-1}{\ln(1-\frac{3}{4})} \cdot \left(\frac{1}{1-x}\right) dx$$

$$\Rightarrow 0 < X < \frac{3}{4}$$

$$X > \frac{1}{2}$$

$$\Rightarrow \frac{1}{2} < X < \frac{3}{4}$$

$$= \frac{-1}{\ln 1 - \ln 4} \cdot [-\ln(1-x)]_{\frac{1}{2}}^{\frac{3}{4}}$$

$$= \frac{-1}{-\ln 4} [\ln(1-\frac{3}{4}) - \ln(1-\frac{1}{2})]$$

$$= \frac{-1}{\ln 4} [\ln(\frac{1}{4}) - \ln(\frac{1}{2})]$$

$$= \frac{-1}{\ln 2^2} [(\ln 1 - \ln 4) - (\ln 1 - \ln 2)]$$

$$= \frac{-1}{2 \ln 2} (-2 \ln 2 + \ln 2) = \frac{-1}{2 \ln 2} (-\ln 2) = \frac{1}{2}$$

H.w Find $P(Y < \frac{3}{4} | X = \frac{1}{2})$

Expectation of two Random Variables Review

① Def. $E(xy)$ is the expectation or the mean of both x and y such that:

$$E(xy) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f(x, y) dx dy \quad \text{if } x \text{ and } y \text{ are C.R.V.s}$$

$$= \sum_x \sum_y xy f(x, y) \quad \text{if } x \text{ and } y \text{ are d.r.v.s}$$

Note: ① $E(x) \equiv \text{mean of } x = \mu_x$

s.t. $E(x) = \int_{-\infty}^{\infty} x f_1(x) dx \quad \text{if } x \text{ is C.R.V.}$

$$= \sum_x x f_1(x) \quad \text{if } x \text{ is d.r.v.}$$

② $E(y) \equiv \text{mean of } y = \mu_y$

s.t. $E(y) = \int_{-\infty}^{\infty} y f_2(y) dy \quad \text{if } y \text{ is C.R.V.}$

$$= \sum_y y f_2(y) \quad \text{if } y \text{ is d.r.v.}$$

③ (a) $V(x) = E(x^2) - [E(x)]^2$

(b) $V(y) = E(y^2) - [E(y)]^2$

④ Covariance of both x & y

We use the symbol $\text{CoV}(x, y)$ (it's mean covariance), s.t.

$$\text{CoV}(x, y) = E\{[x - E(x)][y - E(y)]\}.$$

Theorem (1) If X and Y are Random Variables, then:

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

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

$$= E\{XY - YE(X) - XE(Y) + E(X) \cdot E(Y)\}$$
$$= E(XY) - E(X)E(Y) - E(X)E(Y) + E(X)E(Y)$$
$$= E(XY) - E(X)E(Y)$$

Theorem (2) If X and Y are independent, then

$$E(XY) = E(X)E(Y)$$

Proof ^{or} Case ① $\text{Cov}(X, Y) = 0$ If X and Y are c.r.v.s

∵ X & Y are indep. $\Rightarrow f_1(x)f_2(y) = f(x, y)$

$$E(XY) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (xy) f(x, y) dx dy$$
$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f_1(x) f_2(y) dx dy$$
$$= \int_{-\infty}^{\infty} x f_1(x) dx \cdot \int_{-\infty}^{\infty} y f_2(y) dy \quad \text{by Comm. Property}$$
$$= E(X) \cdot E(Y)$$

Case ② By the same method.

Note By Th. 2 above, if X & Y are indep., then

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

Example: Given a J.P.d.f.

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

Are X & Y indep.? (If $\text{Cov}(X, Y) = 0$ Then

① By using $\text{Cov}(X, Y)$ ② By using $E(XY)$ ③ $f_1(x)f_2(y) = f(x, y)$ $\Rightarrow X$ & Y are indep.)

$$\begin{aligned}
 E(xy) &= \int_0^1 \int_0^x xy (2) dy dx \\
 &= \int_0^1 \left[\frac{2xy^2}{2} \right]_0^x dx \\
 &= \int_0^1 x \cdot x^2 dx \\
 &= \left[\frac{x^4}{4} \right]_0^1 \\
 &= \frac{1}{4}
 \end{aligned}$$

$$\begin{aligned}
 \rho_{xy} &= \frac{\text{Cov}(x,y)}{\sqrt{V(x) \cdot V(y)}} \\
 &= \frac{\sigma_{x,y}}{\sigma_x \sigma_y} \\
 &= \rho_{xy} \text{ Correlation coefficient}
 \end{aligned}$$

$$E(x^2) = \int_0^1 x^2 (2x) dx = \left[\frac{2}{4} x^4 \right]_0^1 = \frac{2}{4} = \frac{1}{2}$$

$$V(x) = \frac{1}{2} - \left(\frac{2}{3}\right)^2 = \frac{1}{2} - \frac{4}{9} = \frac{9-8}{18} = \frac{1}{18}$$

$$\begin{aligned}
 E(y^3) &= \int_0^1 y^3 [2(1-y)] dy = 2 \left[\frac{y^4}{4} - \frac{y^5}{5} \right]_0^1 \\
 &= 2 \left[\frac{1}{4} - \frac{1}{5} \right] \\
 &= 2 \left[\frac{5-4}{20} \right] \\
 &= \frac{2}{10} = \frac{1}{5}
 \end{aligned}$$

$$\begin{aligned}
 V(y) &= \frac{1}{6} - \left[\frac{2}{6} \right]^2 \\
 &= \frac{1}{6} - \frac{4}{36} \\
 &= \frac{6}{36} - \frac{4}{36} \\
 &= \frac{2}{36} = \frac{1}{18}
 \end{aligned}$$

$$R = \left\{ \begin{array}{ll} 0 < y < x & , \quad y < x < 1 \\ 0 < y < 1 & , \quad 0 < x < 1 \end{array} \right\}$$

$$E(XY) = \int_0^1 \int_y^1 xy \cdot f(x,y) dx dy$$

$$= \int_0^1 \int_y^1 xy(2) dx dy = \int_0^1 \int_0^x (xy) f(x,y) dy dx$$

$$= \int_0^1 2y \left. \frac{x^2}{2} \right|_y^1 dy = \int_0^1 y(1-y^2) dy$$

$$= \int_0^1 [y - y^3] dy = \left. \frac{y^2}{2} - \frac{y^4}{4} \right|_0^1$$

$$= \frac{1}{2} - \frac{1}{4} = \frac{1}{4}$$

$$E(X) = \int_0^1 x f_1(x) dx, \quad E(Y) = \int_0^1 y f_2(y) dy$$

$$f_1(x) = \int_0^x 2 dy = \int_0^x 2 dy = 2y \Big|_0^x = 2x \Rightarrow f_1(x) = \begin{cases} 2x & \text{for } 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$f_2(y) = \int_y^1 f(x,y) dx = \int_y^1 2 dx = 2x \Big|_y^1 = 2[1-y] \Rightarrow f_2(y) = \begin{cases} 2(1-y) & \text{for } 0 < y < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$E(X) = \int_0^1 x f_1(x) dx = \int_0^1 x(2x) dx = \int_0^1 2x^2 dx = \left. \frac{2}{3} x^3 \right|_0^1 = \frac{2}{3}$$

$$E(Y) = \int_0^1 y f_2(y) dy = \int_0^1 2y(1-y) dy = \int_0^1 (2y - 2y^2) dy = \left. y^2 - \frac{2}{3} y^3 \right|_0^1 = 1 - \frac{2}{3} = \frac{1}{3}$$

$$\therefore \text{Cov}(X,Y) = E(XY) - E(X)E(Y)$$

$$= \frac{1}{4} - \left(\frac{2}{3}\right) \cdot \left(\frac{1}{3}\right) = \frac{9-8}{36} = \frac{1}{36}$$

$\text{Cov}(X,Y) \neq 0 \Rightarrow X \text{ and } Y \text{ are dependent.}$

~~$\text{Cov}(X,Y) > 0$ all ways~~

(Error)

$$\left. \begin{array}{l} V(X) = 1/18 \\ V(Y) = 1/18 \end{array} \right\} \rho_{X,Y} = \frac{\text{Cov}(X,Y)}{\sqrt{V(X) \cdot V(Y)}} = \frac{\frac{1}{36}}{\sqrt{\frac{1}{18} \cdot \frac{1}{18}}} = \frac{\frac{1}{36}}{\frac{1}{18}} = \frac{1}{2}$$

Correlation Coefficient

Def. when X and Y are dependent and we want to know how much X depends on Y we use the correlation to measure it, denoted by " $\rho_{x,y}$ " read (Rho) where:

$$\rho_{x,y} = \frac{\text{Cov}(X,Y)}{\sqrt{V(X)V(Y)}} = \frac{E(XY) - E(X)E(Y)}{\sqrt{V(X)V(Y)}} = \frac{\text{Cov}(X,Y)}{\sigma_x \sigma_y}$$

$\text{Cov}(X,Y) \neq 0$ if X & Y are dependent
where $-1 \leq \rho \leq 1$

H.w. Find $\rho_{x,y}$ from above example. $\rightarrow \text{Cov}(X,Y) = \frac{1}{2} \neq 0$

② Conditional Mean and Conditional Variance

$E(Y|X)$ denote to the conditional mean of Y given $X=x$.

$E(X|Y)$ denote to the conditional mean of X given $Y=y$.

where $E(Y|X) = \int y f(y|X) dy$, limit of integral follows, limit of y .

$g(x) = \text{func. of } X$

$E(X|Y) = \int x f(x|Y) dx$, limit of integral follows, limit of x .

Note Since $f(y|X)$ defined for interval of y interm of X , then $E(Y|X)$ is a function of X . Also $E(X|Y)$ is a function of Y .

$V(Y|X)$ denote to the conditional variance of Y given $X=x$.

$V(X|Y)$ denote to the conditional variance of X given $Y=y$.

where

$$V(y|x) = E\{[y - E(y|x)]^2 | x\}$$

$$= \int [y - E(y|x)]^2 f(y|x) dy$$

$$V(x|y) = E\{[x - E(x|y)]^2 | y\}$$

$$= \int [x - E(x|y)]^2 f(x|y) dx \quad \text{by def. of } E(x|y)$$

Example: Given a J.P.d.f $f(x, y)$

$$f(x, y) = \begin{cases} 2 & \text{for } 0 \leq x+y \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

Find $E(x|y), E(y|x), V(x|y), V(y|x)$

Sol.

$$R = \{(x, y); 0 \leq x+y \leq 1\} = \left\{ \begin{array}{l} 0 \leq x \leq 1-y, 0 \leq y \leq 1-x \\ 0 \leq x \leq 1, 0 \leq y \leq 1 \end{array} \right\}$$

$$E(x|y) = \int x f(x|y) dx$$

$$f(x|y) = \frac{f(x, y)}{f_2(y)}$$

$$f_2(y) = \int_0^{1-y} f(x, y) dx = \int_0^{1-y} 2 dx = 2x \Big|_0^{1-y} = 2(1-y)$$

$$f_2(y) = \begin{cases} 2(1-y) & \text{for } 0 \leq y \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

$$f(x|y) = \frac{2}{2(1-y)} = \frac{1}{1-y}$$

$$f(x|y) = \begin{cases} \frac{1}{1-y} & \text{for } 0 \leq x \leq 1-y \\ 0 & \text{o.w.} \end{cases} \quad \left(\begin{array}{l} \text{من } 0 \text{ إلى } 1-y \\ \text{و } 0 \leq y \leq 1 \end{array} \right)$$

$$E(x|y) = \int_0^{1-y} x \cdot \frac{1}{(1-y)} dx = \frac{1}{(1-y)} \cdot \frac{x^2}{2} \Big|_0^{1-y} = \frac{(1-y)^2}{2(1-y)} = \frac{(1-y)}{2} \quad 0 \leq y \leq 1$$

$$V(x|y) = \int_0^{1-y} [x - E(x|y)]^2 f(x|y) dx = \int_0^{1-y} \left[x - \frac{(1-y)}{2} \right]^2 dx$$

$$= \int_0^{1-y} \left[x^2 - (1-y)x + \frac{(1-y)^2}{4} \right] dx = \frac{1}{12} (1-y)^2 \quad \text{for } 0 \leq y \leq 1$$

Find $E(y|x)$ & $V(y|x)$

H.w

or

$$E(x^2|y) = \int x^2 f(x|y) dx$$

$$= \frac{x^3}{3(1-y)} \Big|_0^{1-y} = \frac{(1-y)^3}{3(1-y)} = \frac{(1-y)^2}{3}$$

$$V(x|y) = E(x^2|y) - [E(x|y)]^2$$

$$= \frac{(1-y)^2}{3} - \left(\frac{(1-y)}{2} \right)^2$$

Theorem (3) If X and Y are two Random Variables, then:

a. $E[E(Y|X)] = E(Y)$, b. $E[E(X|Y)] = E(X)$

Proof Case (1): If X and Y are c.r.v. with p.d.f $f(x)$.

$E(Y|X)$ is a function of X .

Suppose that $E(Y|X) = g(x)$ as a function of x

$$\begin{aligned} E[E(Y|X)] &= E[g(x)] = \int_{-\infty}^{\infty} g(x) \cdot f_1(x) dx = \int_{-\infty}^{\infty} [E(Y|X)] f_1(x) dx \\ &= \int_{-\infty}^{\infty} \left[\int_{-\infty}^{\infty} y f(y|x) dy \right] f_1(x) dx = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y f(y|x) f_1(x) dy dx \end{aligned}$$

$$\begin{aligned} f(y|x) &= \frac{f(x,y)}{f_1(x)} \Rightarrow f(y|x) \cdot f_1(x) = f(x,y) \\ \int_{-\infty}^{\infty} y \int_{-\infty}^{\infty} f(x,y) dx dy &= \int_{-\infty}^{\infty} y f_2(y) dy = E(Y) \end{aligned}$$

Case (2): If X and Y are d.r.v. with p.m.f. $f(x)$.

$E(Y|X)$ is a function of X .

Suppose that $E(Y|X) = g(x)$

$$\begin{aligned} E[E(Y|X)] &= E(g(x)) = \sum_x g(x) \cdot f_1(x) = \sum_x [E(Y|X)] f_1(x) \\ &= \sum_x \left[\sum_y y f(y|x) \right] f_1(x) = \sum_x \sum_y y f(y|x) f_1(x) \end{aligned}$$

$$f(y|x) = \frac{f(x,y)}{f_1(x)} \Rightarrow f(y|x) \cdot f_1(x) = f(x,y)$$

$$\sum_y y \sum_x f(x,y) = \sum_y y f_2(y) = E(Y)$$

b. By the same method.

Example: Given a J. p.d.f. $f(x,y)$

$$f(x,y) = \begin{cases} 2 & \text{for } 0 \leq x \leq y \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

Find $E[E(Y|X)]$.

Sol.

$$E[E(Y|X)] = E(Y) \text{ (by Theorem 3)}$$

$$E(Y) = \int_{-\infty}^{\infty} y f_2(y) dy = \int_0^1 \int_0^y f(x,y) dx dy = \int_0^1 2 dy = 2y \Big|_0^1 = 2$$

$$f_2(y) = \begin{cases} 2y & \text{for } 0 \leq y \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

$$\therefore E(Y) = \int_0^1 y (2y) dy = 2 \int_0^1 y^2 dy = \frac{2}{3} y^3 \Big|_0^1 = \left(\frac{2}{3} \right)$$

Joint Distribution function (J.d.f) of two a.r.v.s X and Y

Def. The J.d.f of X and Y is defined to a function such that for all values of X and Y ($-\infty < x < \infty, -\infty < y < \infty$) then:

$$F(x, y) = P(X \leq x, Y \leq y)$$

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

① $\int_{-\infty}^x f(t) dt$ if X is c.r.v.

② $\sum_{x_j \leq x} f(x_j)$ if X is d.r.v.

Remarks

$$1. P(a < X \leq b, c < Y \leq d) = [F(b, d) - F(a, d)] - [F(b, c) - F(a, c)]$$

نقارن اعلاه للمتغير العشوائى الواحد (X)

$$\begin{cases} 1. F(x) = P(X \leq x) \\ 2. P(a < X \leq b) = F(b) - F(a) \end{cases}$$

$$2. (i) F_1(x) = P(X \leq x) = \lim_{y \rightarrow \infty} P(X \leq x \text{ and } Y \leq y) = \lim_{y \rightarrow \infty} F(x, y)$$

$$(ii) F_2(y) = \lim_{x \rightarrow \infty} F(x, y)$$

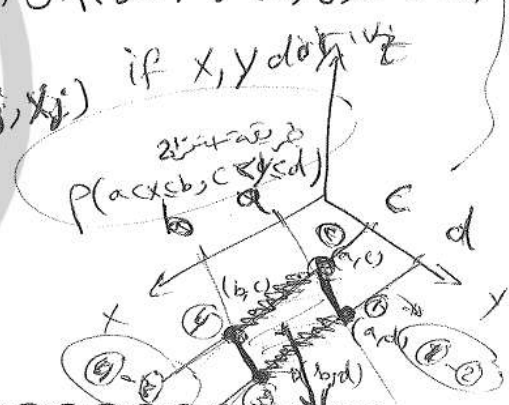
3. If X and Y have cont. J.d.f $F(x, y)$ with J.P.d.f $f(x, y)$, then for any values of X and Y the J.d.f. is

$$① F(x, y) = \sum_{x_j \leq x} \sum_{y_j \leq y} f(x_j, y_j) \text{ if } x, y \text{ d.r.v.}$$

$$② F(x, y) = \int_{-\infty}^y \int_{-\infty}^x f(v, s) dv ds \text{ therefore}$$

Note

$$* f(x, y) = \frac{\partial^2 F(x, y)}{\partial x \partial y}, \quad f(x, y) \text{ diff.}$$



Example: Given a J.d.f $F(x, y)$

$$F(x, y) = \begin{cases} 0 & \text{for } x < 0, y < 0 \\ \frac{1}{16} xy(x+y) & \text{for } 0 \leq x \leq 2, 0 \leq y \leq 2 \\ 1 & \text{for } x > 2, y > 2 \end{cases}$$

$$F(x) = P(X \leq x, Y \leq y)$$

Find $F_1(x), F_2(y), f(x, y)$.

$$\begin{aligned} \text{sol. } F_1(x) &= \lim_{y \rightarrow \infty} F(x, y) = \lim_{y > 2} F(x, y) = F(x, 2) \\ &= \frac{1}{16} x(2)(x+2) = \frac{x(x+2)}{8} \end{aligned}$$

$$F_1(x) = \begin{cases} 0 & \text{for } x < 0 \\ \frac{x(x+2)}{8} & \text{for } 0 \leq x \leq 2 \\ 1 & \text{for } x > 2 \end{cases}$$

$$F_2(y) = \lim_{x \rightarrow \infty} F(x, y) = \lim_{x > 2} F(x, y) = F(2, y) = \frac{1}{16} (2) y (2+y) = \frac{y(2+y)}{8}$$

$$F_2(y) = \begin{cases} 0 & \text{for } y < 0 \\ \frac{y(2+y)}{8} & \text{for } 0 \leq y \leq 2 \\ 1 & \text{for } y > 2 \end{cases}$$

$$f(x, y) = \frac{\partial^2 F(x, y)}{\partial x \partial y} = \frac{\partial}{\partial x} \left[\frac{\partial F(x, y)}{\partial y} \right]$$

$$F(x, y) = \frac{1}{16} (x^2 y + x y^2)$$

$$f(x, y) = \frac{\partial}{\partial x} \left[\frac{1}{16} (x^2 + 2xy) \right] = \left[\frac{1}{16} (2x + 2y) \right] = \frac{1}{8} (x+y)$$

$$f(x, y) = \begin{cases} \frac{x+y}{8} & \text{for } 0 \leq y \leq 2, 0 \leq x \leq 2 \\ 0 & \text{o.w.} \end{cases}$$

Ans. (1) Prove that $E(X+Y) = E(X) + E(Y)$

(2) $E(XY) = E(X)E(Y)$

Proof (1) If X & Y are independent d.r.v.s,

$$\begin{aligned} E(X_1 + X_2) &= \sum_{x_1} \sum_{x_2} (x_1 + x_2) f(x_1, x_2) \\ &= \sum_{x_1} \sum_{x_2} (x_1 f(x_1, x_2) + x_2 f(x_1, x_2)) \\ &= \sum_{x_1} \sum_{x_2} x_1 f(x_1, x_2) + \sum_{x_1} \sum_{x_2} x_2 f(x_1, x_2) \\ &= \left(\sum_{x_1} x_1 f_1(x_1) \right) \left(\sum_{x_2} f_2(x_2) \right) + \left(\sum_{x_2} x_2 f_2(x_2) \right) \left(\sum_{x_1} f_1(x_1) \right) \\ &= E(X_1) \cdot (1) + (28) E(X_2) \cdot (1) \end{aligned}$$

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$$\rho_{x,y} = \frac{\text{Cov}(x,y)}{\sigma_x \sigma_y} = \frac{E(xy) - E(x)E(y)}{\sqrt{\sigma_x^2} \sqrt{\sigma_y^2}}$$

$$\sigma_x^2 = E(x^2) - [E(x)]^2$$

$$\sigma_y^2 = E(y^2) - [E(y)]^2$$

$$E(x^2) = \left. \frac{\partial^2 M(t_1, 0)}{\partial t_1^2} \right|_{t_1=0} = -2(1-t_1)^{-3}(-1) \Big|_{t_1=0} = \boxed{2}$$

$$E(y^2) = \left. \frac{\partial^2 M(0, t_2)}{\partial t_2^2} \right|_{t_2=0} = -2(1-t_2)^{-3}(-1) \Big|_{t_2=0} = \boxed{2}$$

$$\sigma_x^2 = 2 - (1)^2 = \boxed{1} \Rightarrow \boxed{\sigma_x = 1}$$

$$\sigma_y^2 = 2 - (1)^2 = \boxed{1} \Rightarrow \boxed{\sigma_y = 1}$$

$$\rho_{x,y} = \frac{1 - (1)(1)}{(1)(1)} = 0$$

$\therefore X$ & Y are independent (since $\text{Cov}(x,y) = 0$).

H.w. Given a J.P.F of X & Y :

$$f(x,y) = \begin{cases} e^{-y} & \text{for } 0 \leq x < y < \infty \\ 0 & \text{o.w.} \end{cases}$$

Find: ① the M.G.F. of X, Y $[M_{x,y}(t_1, t_2)]$.

② $E(xy)$

③ $f_1(x)$ & $f_2(y)$

④ $M_x = E(x)$ & $M_y = E(y)$ (by two methods).

⑤ $\text{Cov}(X, Y)$.

Note: $R = \left\{ \begin{array}{ll} 0 \leq x < y & , x < y < \infty \\ 0 \leq x < \infty & , 0 \leq y < \infty \end{array} \right\}$

$$M.g.f \text{ of } X = M_{X,Y}(t_1, 0) = M_X(t_1) = \frac{1}{1-t_1}$$

$$\mu_X = E(X) = \left. \frac{\partial M(t_1, 0)}{\partial t_1} \right|_{t_1=0} = \left. \frac{\partial M_X(t_1)}{\partial t_1} \right|_{t_1=0}$$

$$= \left. \frac{1}{(1-t_1)^2} \right|_{t_1=0} = \boxed{1}$$

$$M.g.f \text{ of } Y = M_{X,Y}(0, t_2) = M_Y(t_2) = \frac{1}{1-t_2}$$

$$\mu_Y = E(Y) = \left. \frac{\partial M(0, t_2)}{\partial t_2} \right|_{t_2=0} = \left. \frac{\partial M_Y(t_2)}{\partial t_2} \right|_{t_2=0}$$

$$= \left. \frac{1}{(1-t_2)^2} \right|_{t_2=0} = \boxed{1}$$

$$\therefore M.g.f \text{ of } X, Y = M_{X,Y}(t_1, t_2) = \frac{1}{(1-t_1)(1-t_2)}$$

$$E(XY) = \left. \frac{\partial^2 M(t_1, t_2)}{\partial t_1 \partial t_2} \right|_{t_1=t_2=0}$$

$$= \left. \frac{\partial}{\partial t_1} \left(\frac{\partial M(t_1, t_2)}{\partial t_2} \right) \right|_{t_1=t_2=0}$$

$$= \left. \frac{\partial}{\partial t_1} \left(\frac{1}{(1-t_1)(1-t_2)^2} \right) \right|_{t_1=t_2=0}$$

$$= \left. \frac{1}{(1-t_1)^2(1-t_2)^2} \right|_{t_1=t_2=0}$$

$$= \boxed{1}$$

The variance of X or Y by using $M_{X,Y}(t_1, t_2)$:

$$\begin{aligned}\text{Var}(X) = \sigma_x^2 &= \frac{\partial^2 M(0,0)}{\partial t_1^2} - \left(\frac{\partial M(0,0)}{\partial t_1} \right)^2 \\ &= E(X^2) - [E(X)]^2\end{aligned}$$

$$\begin{aligned}\text{Var}(Y) = \sigma_y^2 &= \frac{\partial^2 M(0,0)}{\partial t_2^2} - \left(\frac{\partial M(0,0)}{\partial t_2} \right)^2 \\ &= E(Y^2) - [E(Y)]^2\end{aligned}$$

Example: Let $f(x,y)$ be the J.P.d.f of X, Y :

$$f(x,y) = \begin{cases} e^{-(x+y)} & \text{for } 0 < x < \infty, 0 < y < \infty \\ 0 & \text{o.w.} \end{cases}$$

Find the M.g.f of X, Y and also, find $M_x, M_y, E(XY)$, and $\rho_{X,Y}$.

Sol.

$$\begin{aligned}M_{X,Y}(t_1, t_2) &= E[e^{t_1 X + t_2 Y}] \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{t_1 x + t_2 y} f(x,y) dy dx \\ &= \int_0^{\infty} \int_0^{\infty} e^{t_1 x + t_2 y} e^{-(x+y)} dy dx \\ &= \left(\int_0^{\infty} e^{-x(1-t_1)} dx \right) \left(\int_0^{\infty} e^{-y(1-t_2)} dy \right) \\ &= \frac{1}{(1-t_1)} \cdot \frac{1}{(1-t_2)} = \frac{1}{(1-t_1)(1-t_2)}\end{aligned}$$

(Pg. 2)

Moment Generating Function of two R.V.'s X & Y

② Let $f(x,y)$ be the J.P.f of X, Y ; and consider that:

① $E[e^{t_1x+t_2y}]$ exists for $|t_1| < h_1, |t_2| < h_2$; h_1, h_2 are positive constants, then:

$E[e^{t_1x+t_2y}]$ is the m.g.f of $X & Y$ and is denoted

by $M_{x,y}(t_1, t_2)$ and is defined as follows:

$$M_{x,y}(t_1, t_2) = E[e^{t_1x+t_2y}] = \begin{cases} \sum_{v_x} \sum_{v_y} e^{t_1x+t_2y} f(x,y) & \text{if } x,y \text{ are d.r.v.s} \\ \iint_{x,y} e^{t_1x+t_2y} f(x,y) dy dx & \text{if } x,y \text{ are c.r.v.s} \end{cases}$$

and

$$M_x(t_1) = M_x(t_1, 0) = E[e^{t_1x}] = \begin{cases} \sum_{v_x} e^{t_1x} f_1(x) & \text{if } x \text{ is d.r.v.} \\ \int_{-\infty}^{\infty} e^{t_1x} f_1(x) dx & \text{if } x \text{ is c.r.v.} \end{cases}$$

$$M_y(t_2) = M_y(0, t_2) = E[e^{t_2y}] = \begin{cases} \sum_{v_y} e^{t_2y} f_2(y) & \text{if } y \text{ is d.r.v.} \\ \int_{-\infty}^{\infty} e^{t_2y} f_2(y) dy & \text{if } y \text{ is c.r.v.} \end{cases}$$

and

③ $\frac{\partial M(0,0)}{\partial t_1} = M_x = E(X)$; $\frac{\partial M(0,0)}{\partial t_2} = M_y = E(Y)$

$$\frac{\partial^2 M(0,0)}{\partial t_1^2} = E(X^2) ; \frac{\partial^2 M(0,0)}{\partial t_2^2} = E(Y^2)$$

$$\frac{\partial^2 M(0,0)}{\partial t_1 \partial t_2} = E(XY)$$

Exercises About Ch.5

Q₁: Given a J.P. f. :

$$f(x,y) = \begin{cases} cy^2 & \text{for } 0 \leq x \leq 2 \text{ \& } 0 \leq y \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

Find :

① The value of (c)

② $P(X+Y > 2)$

③ $P(Y < \frac{1}{2})$, ④ $P(X \leq 1)$

$$\begin{aligned} \int_0^2 \int_0^1 y^2 dy dx &= c \int_0^2 \left[\frac{y^3}{3} \right]_0^1 dx \\ &= \frac{c}{3} \times 1^3 \times 2 \\ &= \frac{2}{3}c \end{aligned}$$

Q₂: Let the J.P.m.f of X & Y be:

$$f(x,y) = \begin{cases} \frac{xy^2}{3} & \text{for } x=1,2,3 \text{ \& } y=1,2 \\ 0 & \text{o.w.} \end{cases}$$

(i) Find the marginal p.f. of X?

(ii) Find the marginal p.f. of Y?

(iii) Are X & Y independent?

Q₃: Let X & Y have the J.P.m.f. be:

$$f(x,y) = \begin{cases} \frac{x+2y}{18} & , \quad x=1,2, y=1,2 \\ 0 & \text{o.w.} \end{cases}$$

Find: $\mu_x = E(X)$, $\mu_y = E(Y)$, & $\text{Cov}(X,Y)$.

Q₄: Suppose that X, Y are C.R.V.s with J.P.d.f. :

$$f(x,y) = \begin{cases} e^{-(x+y)} & \text{for } x > 0, y > 0 \\ 0 & \text{o.w.} \end{cases}$$

Find: $M_{X,Y}(t_1, t_2)$, $M_X(t_1)$, $M_Y(t_2)$.

* $f(x,y)$ for x,y , $f_X(x)$, $f_Y(y)$
 * X and Y indep. pg. 1 check with by Serma's method

Q5: Suppose X and Y are two r.v.s having the following J.p.d.f.:

$$f(x,y) = \begin{cases} x+y & \text{for } 0 < x < 1, 0 < y < 1 \\ 0 & \text{o.w.} \end{cases}$$

- (i) Find the conditional p.d.f $f(x|y)$ and $f(y|x)$.
 (ii) Are these Conditional function $f(x|y)$ and $f(y|x)$ Pr. d.f.? (prove that).

Q6: Consider the following J.P.M.F. of X and Y whose values are given by:

$y \backslash x$	2	3	4	5	6	$f_2(y)$
0	$1/9$	0	$4/27$	0	$2/27$	$1/3$
1	$2/27$	0	$8/81$	0	$4/81$	$2/9$
2	$4/27$	0	$16/81$	0	$8/81$	$4/9$
$f_1(x)$	$1/3$	0	$4/9$	0	$2/9$	1

- Find: ① $f(x|y)$ & $f(y|x)$
 ② $P(X|Y=1)$ & $P(Y|X=2)$

Q7: If J.P.M.F.:

$$P(1,1) = \frac{1}{9}, \quad P(2,1) = \frac{1}{9}, \quad P(3,1) = \frac{1}{9}$$

$$P(1,2) = \frac{1}{9}, \quad P(2,2) = 0, \quad P(3,2) = \frac{1}{18}$$

$$P(1,3) = 0, \quad P(2,3) = \frac{1}{6}, \quad P(3,3) = \frac{1}{9}$$

Are X & Y independent?

Q8: If X and Y are independent with m.p.d.f.:

$$f_1(x) = \begin{cases} 3x^2 & \text{for } 0 \leq x \leq 1 \\ 0 & \text{o.w.} \end{cases}, \quad f_2(y) = \begin{cases} 2y & \text{for } 0 \leq y \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

Find $E(XY)$.

Q₉: Suppose that J.P.d.f of X & Y is:

$$f(x,y) = \begin{cases} C \sin x & \text{for } 0 \leq x \leq \pi, 0 \leq y \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

Find the value of C . ($C = \frac{1}{2\pi}$)

Q₁₀: Let X and Y have the joint p.m.f. be:

(x,y)	(1,1)	(1,2)	(1,3)	(2,1)	(2,2)	(2,3)
$P(x,y)$	$\frac{2}{15}$	$\frac{4}{15}$	$\frac{3}{15}$	$\frac{1}{15}$	$\frac{1}{15}$	$\frac{4}{15}$

Find the Correlation Coefficient $\rho_{x,y}$.

Q₁₁: If $f(x,y) = \begin{cases} e^{-x-y} & \text{for } 0 < x < \infty \text{ \& } 0 < y < \infty \\ 0 & \text{o.w.} \end{cases}$

is the J.P.d.f of two r.v.s X & Y .

Show that X & Y are stochastically independent.

Q₁₂: Let X & Y be independent Y.V.'s with the following distribution:

X	1	2
$f_1(x)$	.6	.4

Y	10	15	5
$f_2(y)$	.5	.3	.2

① Find the Joint pr. f of X & Y

② also, find $E(XY)$.

Q₁₃: Let $f(x|y) = \begin{cases} \frac{C_1 x}{y^2} & \text{for } 0 < x < y, 0 < y < 1 \\ 0 & \text{o.w.} \end{cases}$

& $f_2(y) = \begin{cases} C_2 y^4 & \text{for } 0 < y < 1 \\ 0 & \text{o.w.} \end{cases}$; then:

$$f(x,y) = \dots \quad (0 < x < y < 1)$$

Find:

(1) Determine the constants C_1 and C_2 .

(2) the $f_1(x)$ p.d.f. of X .

(3) the J.P.d.f. of X & Y ; $f(x,y)$

(4) $P(-\frac{1}{2} < X < \frac{1}{2} \mid Y = \frac{5}{8})$

(5) $P(\frac{1}{4} < X < \frac{1}{2})$.

Q14: Let $f(x,y) = \begin{cases} \frac{1}{16} & \text{for } x=1,2,3,4 \\ & y=1,2,3,4 \\ 0 & \text{o.w.} \end{cases}$

be the J.P.f. of X & Y .

Show that X & Y are stochastically independent.

Q15: Given a J.P.f. of X & Y :

$$f(x,y) = \begin{cases} x+y & \text{for } 0 < x < 1 \text{ & } 0 < y < 1 \\ 0 & \text{o.w.} \end{cases}$$

Find the Correlation Coefficient of X & Y ($\rho_{x,y}$).

Solution of Q₁:

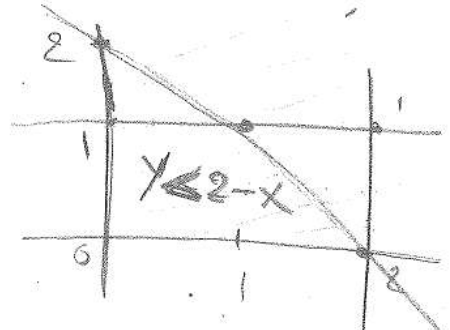
① by cond. (2) $\Rightarrow \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y) dx dy = 1$

$C \int_0^2 \int_0^1 y^2 dy dx = 1 \Rightarrow \boxed{C = \frac{3}{2}}$

② $P(X+Y > 2) = 1 - P(X+Y \leq 2)$

Let $x+y = 2 \Rightarrow y = 2-x$

$x+y \leq 2 \Rightarrow y \leq 2-x$



$R = \left\{ \begin{array}{l} 0 \leq y \leq 2-x \\ 0 \leq x \leq 2 \end{array} \right\}$

$P(X+Y \leq 2) = \int_0^2 \int_0^{2-x} \frac{3}{2} y^2 dy dx = \dots = \frac{1}{2}$

③ $P(Y < \frac{1}{2}) = ?$

To find $f_2(y)$ in the first.

$f_2(y) = \int_0^2 f(x,y) dx = \frac{3}{2} \int_0^2 y^2 dx = \dots = \int_0^1 3y^2 dx \text{ for } 0 \leq y \leq 1$

$P(Y < \frac{1}{2}) = 3 \int_0^{\frac{1}{2}} y^2 dy = \dots = \frac{1}{8}$

④ $P(X \leq 1) = ?$

To find $f_1(x)$ in the beginning.

$f_1(x) = \int_0^1 f(x,y) dy = \frac{3}{2} \int_0^1 y^2 dy = \dots = \frac{1}{2} \text{ for } 0 \leq x \leq 2$

$P(X \leq 1) = \int_0^1 f_1(x) dx$

$= \int_0^1 \frac{1}{2} dx = \dots = \frac{1}{2}$

⑤ $P(X=3Y) = \int_0^2 \int_{\frac{1}{3}x}^{\frac{2}{3}x} f(x,y) dy dx = \int_0^2 \int_{\frac{1}{3}x}^{\frac{2}{3}x} \frac{2}{3} y^2 dy dx$
 $\Rightarrow \boxed{Y = \frac{1}{3}X}$
 $= \dots = \frac{2}{27}$

Q2: If $f_1(x) \cdot f_2(y) = f(x,y)$, then X & Y are indep.

$$f_1(x) = \sum_{y=0}^3 f(x,y) = \sum_{y=0}^3 \frac{1}{30} (x+y)$$

$$= \frac{1}{30} [x + (x+1) + (x+2) + (x+3)]$$

$$= \frac{1}{30} (4x+6) = \begin{cases} \frac{1}{15} (2x+3) & \text{for } x=0,1,2 \\ 0 & \text{o.w.} \end{cases}$$

$$f_2(y) = \sum_{x=0}^2 f(x,y) = \sum_{x=0}^2 \frac{1}{30} (x+y)$$

$$= \frac{1}{30} [y + (1+y) + (2+y)]$$

$$= \frac{1}{30} (3+3y) = \frac{1}{10} (y+1)$$

$$f_2(y) = \begin{cases} \frac{1}{10} (y+1) & \text{for } y=0,1,2,3 \\ 0 & \text{o.w.} \end{cases}$$

$$f_1(x) f_2(y) = \frac{1}{15} (2x+3) \cdot \frac{1}{10} (y+1)$$

$$\neq \frac{1}{30} (x+y) = f(x,y)$$

$$\therefore f_1(x) \cdot f_2(y) \neq f(x,y)$$

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$$\frac{x+y}{30} \quad \begin{matrix} y=0,1,2,3 \\ x=0,1,2 \end{matrix}$$

Q3: To find μ_1 & μ_2 in the first time.

$$\mu_1(x) = \sum_{x_2=1}^2 \frac{1}{18} (x_1 + 2x_2)$$

$$= \frac{1}{18} \sum_{x_2=1}^2 x_1 + \frac{1}{18} \sum_{x_2=1}^2 2x_2$$

$$= \dots = \begin{cases} \frac{x_1+3}{9} & \text{for } x_1=1,2 \\ 0 & \text{o.w.} \end{cases}$$

$$\mu_{x_1} = E(x_1) = \sum_{x_1=1}^2 x_1 f_1(x_1) = \sum_{x_1=1}^2 x_1 \left(\frac{x_1+3}{9} \right) = \dots = \frac{14}{9}$$

$$f_2(x_2) = \sum_{x_1=1}^2 \frac{1}{18} (x_1 + 2x_2) = \begin{cases} \frac{1}{18} (3+4x_2) & \text{for } x_2=1,2 \\ 0 & \text{o.w.} \end{cases}$$

$$\mu_{x_2} = E(x_2) = \sum_{x_2=1}^2 x_2 f_2(x_2) = \sum_{x_2=1}^2 x_2 \left(\frac{3+4x_2}{18} \right) = \dots = \frac{29}{18}$$

$$\text{Cov}(X_1, Y) = E(X_1 X_2) - E(X_1)E(X_2)$$

$$= \mu_{X_1 X_2} - \mu_{X_1} \mu_{X_2}$$

$$\mu_{X_1 X_2} = \sum_{X_1=1}^2 \sum_{X_2=1}^2 f(X_1, X_2) \cdot X_1 X_2$$

$$= \frac{1}{18} \sum_{X_1=1}^2 \sum_{X_2=1}^2 X_1 X_2 (X_1 + 2X_2)$$

$$= \frac{1}{18} [(1)(1)(1+2(1)) + (1)(2)(1+2(2)) + (2)(1)(2+2(1)) + (2)(2)(2+2(2))] = \frac{9}{2}$$

$$\rho_{xy} = \frac{\text{Cov}(X, Y)}{\sigma_x \sigma_y} = \frac{E(XY) - E(X)E(Y)}{\sqrt{V(X)} \sqrt{V(Y)}}$$

$$\sigma_{X_1} = \sqrt{\sigma_{X_1}^2} ; \sigma_{X_1}^2 = E(X_1^2) - [E(X_1)]^2$$

$$\sigma_{X_2} = \sqrt{\sigma_{X_2}^2} ; \sigma_{X_2}^2 = E(X_2^2) - [E(X_2)]^2$$

$$E(X_1^2) = \sum_{X_1=1}^2 X_1^2 f_1(X_1) = \sum_{X_1=1}^2 X_1^2 \left(\frac{1+3}{9}\right) = \frac{24}{9}$$

$$E(X_2^2) = \sum_{X_2=1}^2 X_2^2 f_2(X_2) = \sum_{X_2=1}^2 X_2^2 \left(\frac{1}{18} (3+4X_2)\right) = \frac{51}{18}$$

$$\mu_{X_1}^2 = [E(X_1)]^2 \text{ و } \mu_{X_2}^2 = [E(X_2)]^2$$

$$\sigma_{X_1}^2 = E(X_1^2) - [E(X_1)]^2$$

$$= \frac{24}{9} - \left(\frac{14}{9}\right)^2 = \dots$$

$$\sigma_{X_2}^2 = \frac{51}{18} - \left(\frac{29}{18}\right)^2 = \dots$$

(الكمالات)

Q6: ① $f(X|Y) = \frac{f(X,Y)}{f_2(Y)}$

, $f(Y|X) = \frac{f(X,Y)}{f(X)}$

② $f(X|Y=1) = \frac{f(X,Y=1)}{f_2(Y=1)} = \frac{f(X_i|Y=1)}{2/9}$, $X_i = 2, 3, \dots, 6$.

$$= \begin{cases} \frac{2/27}{2/9} = \frac{1}{3} & \text{for } X=2 \\ \frac{0}{2/9} = 0 & \text{for } X=3 \\ \frac{8/81}{2/9} = \frac{4}{9} & \text{for } X=4 \\ \frac{0}{2/9} = 0 & \text{for } X=5 \\ \frac{4/81}{2/9} = \frac{2}{9} & \text{for } X=6 \end{cases}$$

$$f(y|x=2) = \frac{f(y|x=2)}{f(x=2)} = \frac{f(y|x=2)}{\frac{1}{3}}, y=0,1,2$$

$$= \begin{cases} \frac{1/9}{1/3} = \frac{1}{3} & \text{for } y=0 \\ \frac{2/27}{1/3} = \frac{2}{9} & \text{for } y=1 \\ \frac{4/27}{1/3} = \frac{4}{9} & \text{for } y=2 \end{cases}$$

Hence $f(y) = f(y|x=2)$; therefore X & Y are independent.

Q7:

$y \backslash x$	1	2	3	$f_2(y)$
1	$1/9$	$1/3$	$1/9$	$5/9 = f_2(1)$
2	$1/9$	0	$1/18$	$3/18 = f_2(2)$
3	0	$1/6$	$1/9$	$15/54 = f_2(3)$
$f_1(x)$	$2/9$ $f_1(1)$	$3/6$ $f_1(2)$	$3/18$ $f_1(3)$	1

$$f(1,1) = \frac{1}{9} \neq f_1(1)f_2(1) = \left(\frac{2}{9}\right)\left(\frac{5}{9}\right)$$

$$f(2,1) = \frac{1}{3} \neq f_1(2)f_2(1) = \left(\frac{3}{6}\right)\left(\frac{5}{9}\right)$$

$$f(3,1) = \frac{1}{9} \neq f_1(3)f_2(1) = \left(\frac{3}{18}\right)\left(\frac{5}{9}\right)$$

$$f(1,2) = \frac{1}{9} \neq f_1(1)f_2(2) = \left(\frac{2}{9}\right)\left(\frac{3}{18}\right)$$

$\therefore X$ & Y are dependent.

Q8: Since X & Y are independent; then:

$$f_1(x)f_2(y) = f(x,y)$$

$$f_1(x) \cdot f_2(y) = \begin{cases} 6yx^2 & \text{for } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

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

$$= \int_0^1 \int_0^1 (6yx^2) xy dy dx = \dots = \frac{1}{2}$$

Sol.
Q10

$$f_1(x) = \sum_{y=1}^3 f(x,y)$$

$$= \begin{cases} 9/15 & \text{for } x=1 \\ 6/15 & \text{for } x=2 \\ 0 & \text{o.w.} \end{cases}$$

$$f_2(y) = \sum_{x=1}^2 f(x,y)$$

$$= \begin{cases} 3/15 & \text{for } y=1 \\ 5/15 & \text{for } y=2 \\ 7/15 & \text{for } y=3 \\ 0 & \text{o.w.} \end{cases}$$

$$E(X) = \sum_{x=1}^2 x f_1(x)$$

$$= (1)(9/15) + (2)(6/15)$$

$$= \frac{21}{15}$$

$$E(Y) = \sum_{y=1}^3 y f_2(y)$$

$$= (1)(3/15) + (2)(5/15) + (3)(7/15)$$

$$= \frac{34}{15}$$

$$\rho_{XY} = \text{Cov}(X,Y) / \sigma_X \sigma_Y = \dots$$

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J.P.F.
 $f(x,y)$

X \ y	1	2	$f_2(y)$	
1	2/15	1/15	3/15	$f_2(1)$
2	4/15	1/15	5/15	$f_2(2)$
3	3/15	4/15	7/15	$f_2(3)$
$f_1(x)$	9/15	6/15	1	
	$f_1(1)$	$f_1(2)$		

Q12: ① $f(1,5) = f_1(1) f_2(5) = (.6)(.2) = .12$

$$f(1,10) = f_1(1) f_2(10) = (.6)(.5) = .30$$

$$f(1,15) = f_1(1) f_2(15) = (.6)(.3) = .18$$

$$f(2,10) = f_1(2) f_2(10) = (.4)(.5) = .20$$

$$f(2,5) = f_1(2) f_2(5) = (.4)(.2) = .08$$

$$f(2,15) = f_1(2) f_2(15) = (.4)(.3) = .12$$

Since $(X \text{ \& } Y \text{ are independent r.v.s})$

$$\Rightarrow f_1(x) f_2(y) = f(x,y)$$

→

$$\therefore f(x,y) = \begin{cases} .12 & \text{for } x=1, y=5 \\ .08 & \text{for } x=2, y=5 \\ .30 & \text{for } x=1, y=10 \\ .20 & \text{for } x=2, y=10 \\ .18 & \text{for } x=1, y=15 \\ .12 & \text{for } x=2, y=15 \\ 0 & \text{o.w.} \end{cases}$$

$$\textcircled{2} E(XY) = \sum_x \sum_y XY f(x,y)$$

دلیل

Q13: Sol. $\textcircled{*}$ $\textcircled{a}$ $\int_0^y f(x|y) dx = 1$ (By cond. ②) of pr. f. $\frac{C_1}{2y^2} (y^2) = 1$

$$C_1 \int_0^y \frac{x}{y^2} dx = \frac{C_1}{2y^2} x^2 \Big|_0^y = 1 \Rightarrow C_1 = 2y^2 \Rightarrow C_1 = 2$$

$$\therefore f(x|y) = \begin{cases} \frac{2x}{y^2} & \text{for } 0 < x < y & 0 < y < 1 \\ 0 & \text{o.w.} \end{cases}$$

$\textcircled{*}$ $\int_0^1 f_2(y) dy = 1$ (by Cond. ②) of pr. f.)

$$C_2 \int_0^1 y^4 dy = 1 \Rightarrow C_2 = 5$$

$$\therefore f_2(y) = \begin{cases} 5y^4 & \text{for } 0 < y < 1 \\ 0 & \text{o.w.} \end{cases}$$

c) $f(x,y) = \frac{f(x,y)}{f_2(y)} \Rightarrow f(x,y) = f(x|y) \cdot f_2(y) = (2x)(5y^4) = 10xy^4$ for $0 < x < y < 1$ o.w.

d) $f_1(x) = \int f(x,y) dy = 10 \int_0^1 xy^4 dy = \int 2x$ for $0 < x < 1$ o.w.

e) $P(-\frac{1}{2} < x < \frac{1}{2} | y = \frac{5}{8}) = \int_{-1/2}^{1/2} f(x|y = \frac{5}{8}) dx$
 $= \int_{-1/2}^{1/2} 10x (\frac{5}{8})^4 dx = \dots$

e) $P(\frac{1}{4} < x < \frac{1}{2}) = \int_{1/4}^{1/2} f_1(x) dx = \int_{1/4}^{1/2} 2x dx = \dots = \frac{3}{16}$

وزارة التعليم العالي والبحث العلمي

جامعة ديالى

كلية تربية المقداد / قسم الرياضيات

محاضرات مادة

الإحصاء والاحتمالية

للعام الدراسي (2023-2024)

المرحلة الثالثة

Chapter Six

مدرس المادة: م.م هند إبراهيم محمد

Some Special Distribution:

Discrete Distribution:

1) Discrete uniform distribution:

The r.v. X is said to have an uniform distribution if P.m.f. is give by

$$f(X, N) = \begin{cases} \frac{1}{N} & ; x = 1, 2, 3, \dots, N \\ 0 & ; o.w. \end{cases} \quad ; \text{Where the parameter } N \geq 1 \text{ Natural numbere}$$

And denoted by $X \sim D_u(N)$

Clear that

$$f(x) \geq 0 \quad \text{and} \quad \sum_{x=1}^N f(x) = 1$$

Then $f(x)$ satisfies the condition of begin a P.m.f. of discrete type of random variable x .

Note : the mean $\mu = \frac{N+1}{2}$ and the variance $\sigma^2 = \frac{N^2-1}{12}$ (proof)

The moment generating function of uniform distribution is given by

$$\begin{aligned} M_x(t) &= E(e^{tx}) = \sum_{\forall x} e^{tx} f(x) = \sum_{\forall x} e^{tx} \frac{1}{N} = \frac{1}{N} \sum_{\forall x} e^{tx} \\ &= \frac{1}{N} [e^t + e^{2t} + e^{3t} + \dots + e^{Nt}] \\ &= \frac{e^t}{N} [1 + e^t + e^{2t} + \dots + e^{(N-1)t}] \\ &= \frac{e^t}{N} \left[\frac{1 - e^{Nt}}{1 - e^t} \right] \end{aligned}$$

Example : If $X \sim D_u(6)$; find ❶ $P(x > 1)$ ❷ $P(x \geq 3)$ ❸ mean and variance

Solution :

$$X \sim D_u(6) \Rightarrow f(x) = \begin{cases} \frac{1}{6} & ; x = 1, 2, 3, \dots, 6 \\ 0 & ; o.w. \end{cases}$$

❶ $P(x > 1) = 1 - P(x \leq 1) = 1 - P(x=1) = 1 - 1/6 = 5/6$

❷ $P(x \geq 3) = P(x=3) + P(x=4) + P(x=5) + P(x=6) = 1/6 + 1/6 + 1/6 + 1/6 = 4/6$

Or $= 1 - P(x < 3) = 1 - [P(x=1) + P(x=2)] = 1 - [1/6 + 1/6] = 1 - 2/6 = 4/6$

❸ Mean $\mu = \frac{N+1}{2} = \frac{6+1}{2} = \frac{7}{2}$

Variance $\sigma^2 = \frac{N^2-1}{12} = \frac{36-1}{12} = \frac{35}{12}$

2) Bernoulli distribution:

The r.v. X is said to have a Bernoulli distribution with parameter p if P.m.f.

is give by

$$f(X, p) = \begin{cases} p^x q^{1-x} & ; x = 0, 1 \\ 0 & ; o.w. \end{cases} ; \text{Where the parameter } 0 \leq p \leq 1, q = 1 - p$$

And denoted by $X \sim Ber(p)$

Clear that

$$f(x) \geq 0 \quad \text{and} \quad \sum_{x=0}^1 f(x) = p + q = 1$$

Then $f(x)$ satisfies the condition of being a P.m.f. of discrete type of random variable x

Note : the mean $\mu = p$ and the variance $\sigma^2 = p q$ (proof)

The moment generating function of uniform distribution is given by

$$\begin{aligned} M_x(t) &= E(e^{tx}) = \sum_{\forall x} e^{tx} f(x) \\ &= \sum_{\forall x} e^{tx} p^x q^{1-x} = \sum_{x=0}^1 e^{tx} p^x q^{1-x} \\ &= e^{0t} p^0 q^{1-0} + e^{1t} p^1 q^{1-1} = q + p e^t \end{aligned}$$

3) Binomial distribution:

The r.v. X is said to have a Binomial distribution with two parameters n and p if

P.m.f. is given by

$$f(X, n, p) = \begin{cases} C_x^n p^x q^{n-x} & ; x = 0, 1, 2, \dots, n \\ 0 & ; o.w. \end{cases}$$

And denoted by $X \sim b(n, p)$

Where the two parameters n and p satisfy the following conditions:

- 1- n is a positive integer
- 2- $0 \leq p \leq 1$
- 3- $q = 1 - p$ or $p + q = 1$

Under the conditions, it is clear that

$$f(x) \geq 0 \quad \forall x \quad \text{and} \quad \sum_{x=0}^n f(x) = \sum_{x=0}^n C_x^n p^x q^{n-x} = (p + q)^n = 1$$

Then $f(x)$ satisfies the condition of being a P.m.f. of discrete type of random variable x

Useful relationship:

$$\text{Binomial Formula is } \sum_{x=0}^n C_x^n a^x b^{n-x} = (a+b)^n.$$

Note : the mean $\mu = np$ and the variance $\sigma^2 = npq$

Proof :

$$\begin{aligned} \mu &= E(x) = \sum_{\forall x} x f(x) = \sum_{x=0}^n x \cdot C_x^n p^x q^{n-x} \\ \therefore C_x^n &= \frac{n!}{x! (n-x)!} \\ \therefore \mu &= \sum_{x=0}^n x \cdot C_x^n p^x q^{n-x} = \sum_{x=0}^n x \frac{n!}{x! (n-x)!} p^x q^{n-x} \\ &= \sum_{x=1}^n x \frac{n!}{x! (n-x)!} p^x q^{n-x} \\ &= \sum_{x=1}^n \frac{n!}{(x-1)! (n-x)!} p^x q^{n-x} \\ &= np \sum_{x=1}^n \frac{(n-1)!}{(x-1)! (n-x)!} p^{x-1} q^{n-x} \end{aligned}$$

let $y = x - 1$ and $m = n - 1$

$$\begin{aligned} \Rightarrow \mu &= np \sum_{y=0}^m \frac{m!}{y! (m-y)!} p^y q^{m-y} \\ \therefore \mu &= np \sum_{y=0}^m C_y^m p^y q^{m-y} = np(p+q)^m = np 1^m = np \end{aligned}$$

$$\therefore \mu = np$$

$$\therefore \sigma^2 = \text{var}(x) = E(x^2) - E^2(x)$$

We must compute $E(x^2)$ to that we compute $E[x(x-1)] = E(x^2) - E(x)$

$$\begin{aligned} E[x(x-1)] &= \sum_{\forall x} x(x-1) f(x) = \sum_{x=0}^n x(x-1) C_x^n p^x q^{n-x} \\ \therefore E[x(x-1)] &= \sum_{x=2}^n x(x-1) \frac{n!}{x!(n-x)!} p^x q^{n-x} \\ &= \sum_{x=2}^n \frac{n!}{(x-2)!(n-x)!} p^x q^{n-x} \\ &= n(n-1)p^2 \sum_{x=2}^n \frac{(n-2)!}{(x-2)!(n-x)!} p^{x-2} q^{n-x} \\ &= n(n-1)p^2 \sum_{x=2}^n C_{x-2}^{n-2} p^{x-2} q^{n-x} \end{aligned}$$

let $y = x - 2$ and $m = n - 2$

$$= n(n-1)p^2 \sum_{y=0}^m C_y^m p^y q^{m-y} = 1$$

$$E[x(x-1)] = E(x^2) - E(x) = n(n-1)p^2$$

$$\Rightarrow E(x^2) = np + n^2p^2 - np^2$$

$$\begin{aligned} \therefore \sigma^2 &= E(x^2) - E^2(x) = np + \cancel{n^2p^2} - np^2 - \cancel{n^2p^2} \\ &= np - np^2 = np(1-p) = npq \end{aligned}$$

$$\therefore \sigma^2 = npq$$

The moment generating function of Binomial distribution is given by

$$\begin{aligned} M_x(t) &= E(e^{tx}) = \sum_{\forall x} e^{tx} f(x) = \sum_{\forall x} e^{tx} C_x^n p^x q^{n-x} = \sum_{x=0}^n C_x^n e^{tx} p^x q^{n-x} \\ &= \sum_{x=0}^n C_x^n (e^t p)^x q^{n-x} = (q + p e^t)^n \end{aligned}$$

The properties of binomial distribution is:

- 1- two possible outcome
- 2- n trials
- 3- independent trials

Example: Suppose X is binomially distributed with parameter n and p, further suppose

$$E(X) = 5 \text{ and } \text{var}(X) = 4, \text{ find } n \text{ and } p.$$

Solution: $X \sim b(n, p) \Rightarrow E(x) = np = 5 \text{ and } \text{var}(x) = npq = 4$

$$\Rightarrow q = \frac{4}{5} \Rightarrow p = 1 - \frac{4}{5} = \frac{1}{5}$$

$$\therefore np = 5 \Rightarrow n = \frac{5}{\frac{1}{5}} = 25$$

$$\therefore X \sim b\left(25, \frac{1}{5}\right)$$

4) Poisson distribution:

The r.v. X is said to have a Poisson distribution with parameter λ if

P.m.f. is give by

$$f(X, \lambda) = \begin{cases} \frac{e^{-\lambda} \lambda^x}{x!} & ; x = 0, 1, 2, \dots, \infty \\ 0 & ; o.w. \end{cases}, \text{ where } \lambda > 0$$

And denoted by $X \sim Po(\lambda)$

$$\text{Since } \lambda > 0 \Rightarrow f(x) \geq 0 ; \forall x$$

$$\text{Since } \sum_{x=0}^{\infty} \frac{\lambda^x}{x!} = e^{\lambda} \Rightarrow \sum_{\forall x} f(x) = 1$$

Then f(x) satisfies the condition of begin a P.m.f. of discrete type of random variable x

Note : the Mean $\mu = \lambda$ and the Variance $\sigma^2 = \lambda$

Proof :

$$\begin{aligned}
 \text{Mean} = E(X) &= \sum_{x=0}^{\infty} x f(x; \lambda) = \sum_{x=0}^{\infty} x \frac{e^{-\lambda} \lambda^x}{x!} \\
 &= e^{-\lambda} \sum_{x=0}^{\infty} x \frac{\lambda \lambda^{x-1}}{x(x-1)!} \\
 &= e^{-\lambda} \lambda \sum_{x=1}^{\infty} \frac{\lambda^{x-1}}{(x-1)!} \\
 &= e^{-\lambda} \lambda \left[\frac{\lambda^0}{0!} + \frac{\lambda^1}{1!} + \frac{\lambda^2}{2!} + \frac{\lambda^3}{3!} + \dots + \infty \right] \\
 &= e^{-\lambda} \lambda e^{\lambda} \\
 &= \lambda
 \end{aligned}$$

$\therefore$ Mean of the Poisson distribution is λ

And

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

$\therefore$ Mean of the Poisson distribution is λ From (1)

$$\text{Var}(X) = E(X^2) - [\lambda]^2 \dots \dots \dots (1)$$

$$\begin{aligned}
 E(X^2) &= \sum_{x=0}^{\infty} x^2 f(x; \lambda) = \sum_{x=0}^{\infty} x^2 \frac{e^{-\lambda} \lambda^x}{x!} \\
 &= \sum_{x=0}^{\infty} [x(x-1) + x] \frac{e^{-\lambda} \lambda^x}{x!} \\
 &= \sum_{x=0}^{\infty} x(x-1) \frac{e^{-\lambda} \lambda^x}{x!} + \sum_{x=0}^{\infty} x \frac{e^{-\lambda} \lambda^x}{x!} = E(X) = \lambda \\
 &= e^{-\lambda} \sum_{x=2}^{\infty} x(x-1) \frac{\lambda^2 \lambda^{x-2}}{x(x-1)(x-2)!} + \lambda
 \end{aligned}$$

$$\begin{aligned}
&= e^{-\lambda} \lambda^2 \sum_{x=2}^{\infty} \frac{\lambda^{x-2}}{(x-2)!} + \lambda \\
&= e^{-\lambda} \lambda^2 \left[\frac{\lambda^0}{0!} + \frac{\lambda^1}{1!} + \frac{\lambda^2}{2!} + \frac{\lambda^3}{3!} + \dots + \infty \right] + \lambda \\
&= e^{-\lambda} \lambda^2 e^{\lambda} + \lambda
\end{aligned}$$

$$\Rightarrow E(X^2) = \lambda^2 + \lambda \dots \dots \dots (2)$$

Putting (2) in (1) we get

$$\begin{aligned}
\text{Var}(X) &= E(X^2) - [\lambda]^2 \\
&= \lambda^2 + \lambda - \lambda^2 \\
&= \lambda
\end{aligned}$$

$\therefore$ Variance of the Poisson distribution is λ

Note: In Poisson distribution the mean and variance are equal i.e. λ

The properties of Poisson distribution is:

- 1- The event occurring randomly in time.
- 2- The number of event in non-overlapping time period are independent.
- 3- $\lambda > 0$, rate of Poisson process is average number of events per unite time.

Example : Suppose the number of flaws in a 100-foot roll of paper is a Poisson random variable with $\lambda = 10$. Then the probability that there are eight flaws in a 100-foot roll is:

Solution:

$$P(X=8 | \lambda=10) = \frac{e^{-\lambda} \lambda^x}{x!} = \frac{e^{-10}(10)^8}{8!} = \frac{(2.71828)^{-10}(100,000,000)}{40,320} = .1126$$

The probability of seven flaws in a 50-foot roll is:

$$P(X=7 | \lambda=5) = \frac{e^{-\lambda} \lambda^x}{x!} = \frac{e^{-5}(5)^7}{7!} = \frac{(2.71828)^{-5}(78,125)}{5,040} = .1044$$

Note:

The Poisson distribution can be limiting case of a binomial distribution under certain conditions.

1. Number of trials i.e. n is indefinitely large i.e. $n \rightarrow \infty$
2. p , the probability of success in each trial is indefinitely small i.e. $p \rightarrow 0$
3. $np = \lambda$ is finite.

Proof: (للاطلاع)

If X is a binomial distribution then the probability mass function is given by

$$P(X=x) = C_x^n p^x q^{n-x}, \quad x = 0, 1, 2, \dots, n$$

Under the above conditions

$$\begin{aligned}
 \lim_{n \rightarrow \infty} b(x; n, p) &= \lim_{n \rightarrow \infty} C_x^n p^x q^{n-x} \\
 &= \lim_{n \rightarrow \infty} \frac{n!}{x!(n-x)!} \left(\frac{\lambda}{n}\right)^x \left(1 - \frac{\lambda}{n}\right)^{n-x} \quad [\because np = \lambda] \\
 &= \lim_{n \rightarrow \infty} \frac{n(n-1)(n-2)\dots(n-x+1)(n-x)!}{x!(n-x)!} \left(\frac{\lambda}{n}\right)^x \left(1 - \frac{\lambda}{n}\right)^{n-x} \\
 &= \frac{\lambda^x}{x!} \lim_{n \rightarrow \infty} \frac{n^x \left[1 \left(1 - \frac{1}{n}\right) \left(1 - \frac{2}{n}\right) \dots \left(1 - \frac{x-1}{n}\right)\right] \left(1 - \frac{\lambda}{n}\right)^n}{n^x \left(1 - \frac{\lambda}{n}\right)^x} \\
 &= \frac{\lambda^x}{x!} \lim_{n \rightarrow \infty} 1 \cdot \frac{\left(1 - \frac{\lambda}{n}\right)^n}{\left(1 - \frac{\lambda}{n}\right)^x} \\
 &= \frac{\lambda^x}{x!} \frac{\lim_{n \rightarrow \infty} \left(1 - \frac{\lambda}{n}\right)^n}{\lim_{n \rightarrow \infty} \left(1 - \frac{\lambda}{n}\right)^x} = \frac{\lambda^x}{x!} \frac{e^{-\lambda}}{1^x} = \frac{e^{-\lambda} \lambda^x}{x!}
 \end{aligned}$$

Example : If $n=5000$, $p=0.001$ and $X \sim b(n, p)$ find $P(x=7)=f(7)$

Solution : $X \sim b(n, p) = X \sim Po(\lambda)$ $\lambda = np = 5000 * 0.001 = 5$

$$P(X=7 | \lambda=5) = \frac{e^{-\lambda} \lambda^x}{x!} = \frac{e^{-5} (5)^7}{7!} = \frac{(2.71828)^{-5} (78,125)}{5,040} = .1044$$

$$\therefore f(7) = 0.1044$$

The moment generating function of Binomial distribution is given by

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

Proof :

$$\begin{aligned} M_x(t) &= E(e^{tx}) = \sum_{x=0}^{\infty} e^{tx} f(x) = \sum_{x=0}^{\infty} e^{tx} \frac{e^{-\lambda} \lambda^x}{x!} \\ &= e^{-\lambda} \sum_{x=0}^{\infty} \frac{(\lambda e^t)^x}{x!} = e^{-\lambda} e^{\lambda e^t} = e^{\lambda(e^t - 1)} \end{aligned}$$

Example : Let X be r.v. whose M.g.f is give by $e^{2e^t - 2}$ find $P(X \geq 1)$

Solution : $X \sim Po(2)$

$$\therefore f(X; \lambda) = \begin{cases} \frac{e^{-2} 2^x}{x!} & ; x = 0, 1, 2, \dots, \infty \\ 0 & ; o.w. \end{cases}$$

$$\begin{aligned} \therefore P(X \geq 1) &= 1 - P(X < 1) \\ &= 1 - P(X = 0) \\ &= 1 - \frac{e^{-2} 2^0}{0!} = 1 - e^{-2} = 0.864 \end{aligned}$$

Question: If x has Poisson distribution and $P(X=0)=1/2$ find $E(x)$?

Solution :

$$\therefore X \sim Po(\lambda)$$

$$\therefore f(X, \lambda) = \begin{cases} \frac{e^{-\lambda} \lambda^x}{x!} & ; x = 0, 1, 2, \dots, \infty \\ 0 & ; o.w. \end{cases} \quad [H. W.]$$

5) Negative Binomial distribution

Perform independent Bernoulli trials repeatedly until a given number of success are observed i.e. let x the total number of failure before the r^{th} success, that means the r.v. X represent the number of failure Prior to the r^{th} success the P.m.f of X is:

$$f(X; r, p) = \begin{cases} C_x^{x+r-1} p^r q^x & ; x = 0, 1, 2, \dots \dots \dots \\ 0 & ; o.w. \end{cases}$$

And denoted by $X \sim Nb(r, p)$

Where the parameter r and p satisfy $r=1, 2, 3, \dots \dots \dots$ and $0 \leq p \leq 1$ and $q = 1 - p$

Since $0 \leq p \leq 1$

$$\therefore f(x) \geq 0 \quad \forall x \quad \text{and} \quad \sum_{x=0}^{\infty} f(x) = \sum_{x=0}^{\infty} C_x^{x+r-1} p^r q^x = p^r (1 - q)^{-r} = 1$$

Then $f(x)$ satisfies the condition of begin a P.m.f. of discrete type of random variable x

Useful relationship:
$$\sum_{x=0}^{\infty} C_x^{n+x-1} a^x = (1 - a)^{-n}.$$

Note : the mean $\mu = \frac{rq}{p}$ and the variance $\sigma^2 = \frac{rq}{p^2}$

Proof :

$$\mu = E(x) = \sum_{\forall x} x f(x) = \sum_{x=0}^{\infty} x \cdot C_x^{x+r-1} p^r q^x$$

$$\therefore C_x^n = \frac{n!}{x! (n - x)!}$$

$$\therefore \mu = \sum_{x=1}^{\infty} x \cdot C_x^{x+r-1} p^r q^x = \sum_{x=1}^{\infty} x \frac{(x + r - 1)!}{x! (r - 1)!} p^r q^x$$

$$\begin{aligned}
&= p^r \sum_{x=1}^{\infty} \frac{(x+r-1)!}{(x-1)!(r-1)!} q^x \\
&= r q p^r \sum_{x=1}^{\infty} \frac{(x+r-1)!}{(x-1)! r!} q^{x-1}
\end{aligned}$$

let $y = x - 1$ and $s = r + 1$

$$\begin{aligned}
&= r q p^r \sum_{y=0}^{\infty} \frac{(y+s-1)!}{y! (s-1)!} q^y \\
\Rightarrow \mu &= \frac{r q p^r}{(1-q)^s} = \frac{r q p^r}{p^{r+1}} = \frac{r q}{p} \\
\therefore \mu &= E(x) = \frac{r q}{p}
\end{aligned}$$

$$\therefore \sigma^2 = \text{var}(x) = E(x^2) - E^2(x)$$

We must compute $E(x^2)$ to that we compute $E[x(x-1)] = E(x^2) - E(x)$

$$\begin{aligned}
E[x(x-1)] &= \sum_{\forall x} x(x-1) f(x) = \sum_{x=0}^{\infty} x(x-1) C_x^{x+r-1} p^r q^x \\
\therefore E[x(x-1)] &= \sum_{x=2}^{\infty} x(x-1) \frac{(x+r-1)!}{x! (r-1)!} p^r q^x \\
&= q^2 p^r r(r+1) \sum_{x=2}^{\infty} \frac{(x+r-1)!}{(x-2)! (r+1)!} q^{x-2}
\end{aligned}$$

let $y = x - 2$ and $s = r + 2$

$$\begin{aligned}
&= q^2 p^r r(r+1) \sum_{y=0}^{\infty} \frac{(y+s-1)!}{y! (s-1)!} q^y \\
\Rightarrow E(x^2) - E(x) &= \frac{q^2 p^r r(r+1)}{(1-q)^s} = \frac{q^2 p^r r(r+1)}{p^{r+2}} = \frac{r(r+1)q^2}{p^2}
\end{aligned}$$

$$\begin{aligned}
\Rightarrow E(x^2) &= \frac{r^2 q^2}{p^2} + \frac{r q^2}{p^2} + \frac{r q}{p} \\
\therefore \sigma^2 &= E(x^2) - E^2(x) = \frac{r^2 q^2}{p^2} + \frac{r q^2}{p^2} + \frac{r q}{p} - \frac{r^2 q^2}{p^2} \\
&= \frac{r q^2}{p^2} + \frac{r q}{p} = \frac{r q (q + p)}{p^2} = \frac{r q}{p^2} \\
\therefore \sigma^2 &= \frac{r q}{p^2}
\end{aligned}$$

The moment generating function of **Negative Binomial** distribution is given by

$$\begin{aligned}
M_x(t) &= E(e^{tx}) = \sum_{\forall x} e^{tx} f(x) = \sum_{\forall x} e^{tx} C_x^{x+r-1} p^r q^x \\
&= p^r \sum_{x=0}^{\infty} C_x^{x+r-1} (e^t q)^x = \left(\frac{p}{1 - q e^t} \right)^r
\end{aligned}$$

Other formula Of Negative Binomial distribution given by

$$f(X; r, p) = \begin{cases} C_{r-1}^{x+r-1} p^r q^x & ; x = 0, 1, 2, \dots \dots \dots \\ 0 & ; o.w. \end{cases}$$

Where X denote the number of failures before we get the first r successes

Remark: If X denote the number of trials required to get a total of r successes then

$$f(X; r, p) = \begin{cases} C_{r-1}^{x-1} p^r q^{x-r} & ; x = r, r + 1, r + 2, \dots \dots \dots \\ 0 & ; o.w. \end{cases}$$

6) Geometric Distribution

In case of the binomial distribution, the number of trials was predetermined. Sometimes, however, we wish to know the number of trials needed before a certain outcome occurs.

For example, we wish to play until we win; you roll dice until you get an 11; a mechanic waits for the first plane to arrive at the airport that needs repair; a basketball player shoots until he makes it. These situations fall under the

Geometric distribution. (Special case of Negative Binomial at $r=1$)

The r.v. X is said to have a **Geometric distribution** with parameter p if

P.m.f. is given by

$$f(X; p) = \begin{cases} p q^x & ; x = 0, 1, 2, \dots \dots \dots \\ 0 & ; o.w. \end{cases} \quad \text{where } 0 \leq p \leq 1; \quad q = 1 - p$$

And denoted by $X \sim G(p)$

Since $0 \leq p \leq 1$

$$\therefore f(x) \geq 0 \quad \forall x \quad \text{and} \quad \sum_{x=0}^{\infty} f(x) = \sum_{x=0}^{\infty} p q^x = \frac{p}{(1-q)} = 1$$

Then $f(x)$ satisfies the condition of begin a P.m.f. of discrete type of random variable x

The moment generating function of Geometric distribution is given by

$$\begin{aligned} M_x(t) &= E(e^{tx}) = \sum_{\forall x} e^{tx} f(x) = \sum_{\forall x} e^{tx} p q^x \\ &= p \sum_{x=0}^{\infty} (e^t q)^x = \frac{p}{1 - q e^t} \end{aligned}$$

Note : the mean $\mu = \frac{q}{p}$ and the variance $\sigma^2 = \frac{q}{p^2}$

Proof :

$$M_x(t) = \frac{p}{1 - q e^t} \Rightarrow M'_x(t) = \frac{p q e^t}{(1 - q e^t)^2} \Rightarrow E(x) = M'_x(0) = \frac{p q e^0}{(1 - q e^0)^2} = \frac{q}{p}$$

$$\therefore \mu = E(x) = \frac{q}{p}$$

$$M''_x(t) = \frac{(1 - q e^t)^2 (p q e^t) + 2q((1 - q e^t)(p q e^t))}{(1 - q e^t)^4} \Rightarrow E(x^2) = M''_x(0)$$

$$\therefore E(x^2) = \frac{(1 - q e^0)^2(p q e^0) + 2q((1 - q e^0)(p q e^0))}{(1 - q e^0)^4}$$

$$\therefore E(x^2) = \frac{(1 - q)^2(p q) + 2q((1 - q)(p q))}{(1 - q)^4} = \frac{qp^3 + 2q^2p^2}{p^4} = \frac{qp + 2q^2}{p^2}$$

$$\sigma^2 = \text{var}(x) = E(x^2) - E^2(x) \Rightarrow \sigma^2 = \frac{qp + 2q^2}{p^2} - \frac{q^2}{p^2} = \frac{q(p + q)}{p^2} = \frac{q}{p^2}$$

$$\therefore \sigma^2 = \frac{q}{p^2}$$

The properties Geometric distribution

1. Each event falls into just one of two categories, which are generally referred to as a "success" or "failure."
2. The probability of success, call it p , is the same for each observation.
3. The observations are all independent.
4. The variable of interest is the number of trials required to obtain the **first** success.

Remark: If X denote the number of trials required to get a first success then the P.m.f is:

$$f(X; p) = \begin{cases} p q^{x-1} & ; x = 1, 2, 3, \dots \dots \dots \\ 0 & ; o.w. \end{cases}$$

Note : By this case the mean $\mu = \frac{1}{p}$ and the variance $\sigma^2 = \frac{1}{p^2}$

Proof (H.W.)

Example: On island of Oahu in the small village of Nanakuli, about 80% of the residents are of Hawaiian ancestry . Suppose you fly to Hawaii and visit Nanakuli.

What is the probability that the first villager you meet is Hawaiian?

What is the probability that you do not meet a Hawaiian until the third villager?

(على جزيرة أوهايو في القرية الصغيرة في نانا كولي، حوالي 80 % من السكان من أصل هاواي. افترض أنك سافرت إلى هاواي وزرت نانا كولي. ما هو احتمال أن القروي الأول الذي تلتقيه هاواي؟ ما هو احتمال بأنك لا تلتقي بأي هاواي حتى القروي الثالث؟)

Solution :

$$f(X; p) = \begin{cases} p q^{x-1} & ; x = 1, 2, 3, \dots \dots \dots \\ 0 & ; o.w. \end{cases}$$

$$p = 0.80 = .8 \quad ; \quad q = 1 - p = 1 - .8 = .2$$

$$f(1) = P(X = 1) = (1 - .8)^{1-1} (.80) = (.2)^0 (.80) = .80.$$

$$f(3) = P(X = 3) = (.2)^2 (.80) = .032$$

What is the probability that you will meet at most three people from Hawaiian ancestry (اصل)?

$$\begin{aligned} P(X \leq 3) &= P(X = 1) + P(X = 2) + P(X = 3) \\ &= (.2)^0 (.8) + (.2)^1 (.8) + (.2)^2 (.8) \\ &= .8 + .16 + .032 = .992 \end{aligned}$$

How many people should you expect to meet before you meet the first Hawaiian?

$$\mu = \frac{1}{.80} = 1.25 \Rightarrow \text{First integer number} \geq \mu \Rightarrow \text{Answer} = 2$$

Note : If X is a geometric random variable with probability p of success on each trial, the

expected number of trials necessary to reach the first success is $\mu = E(x) = \frac{1}{p}$.

What is the probability that it takes more than three people before you meet a Hawaiian?

$$P(X > 3) = (1 - .8)^4 = .0016$$

Note, The probability that it takes more than n trials before we see the first success is

$$P(X > n) = (1 - p)^{n+1} \text{ and } P(X \geq n) = (1 - p)^n. \text{ (Proof that H.W.)}$$

Example: Suppose we flip a fair coin until we get a head. Let X be the number of trail to get a head. Find the probability mass function of X .

Solution:

$$X \sim G(p) \quad p = 0.5 \quad \text{and} \quad q = 1 - p = 0.5$$

$$P.m.f. \text{ is } f(X) = \begin{cases} (0.5) (0.5)^{x-1} = (0.5)^x & ; x = 1, 2, \dots \\ 0 & ; o.w. \end{cases}$$

Question: A fair die cast on successive independent trail until the second six is observed.

Find the probability of observing exactly ten non-sixes before the second six is appear

Answer:

$$r = 2, \quad p = \frac{1}{6}, \quad q = \frac{5}{6}$$

$$\therefore X \sim Nb(r, p) \Rightarrow f(X; r, p) = \begin{cases} C_{r-1}^{x+r-1} p^r q^x & ; x = 0, 1, 2, \dots \dots \dots \\ 0 & ; o.w. \end{cases}$$

$$\Rightarrow f(x) = f\left(X; 2, \frac{1}{6}\right) = \begin{cases} C_1^{x+1} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^x & ; x = 0, 1, 2, \dots \dots \dots \\ 0 & ; o.w. \end{cases}$$

$$\begin{aligned} \Rightarrow f(10) = P(x = 10) &= C_1^{11} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^{10} \\ &= 10 \times (0.0278) \times (0.1615) = 0.0449 \end{aligned}$$

Or

$$f(X; r, p) = \begin{cases} C_x^{x+r-1} p^r q^x & ; x = 0, 1, 2, \dots \dots \dots \\ 0 & ; o.w. \end{cases}$$

$$\Rightarrow f(x) = f\left(X; 2, \frac{1}{6}\right) = \begin{cases} C_x^{x+1} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^x & ; x = 0, 1, 2, \dots \dots \dots \\ 0 & ; o.w. \end{cases}$$

$$\begin{aligned} \Rightarrow f(10) = P(x = 10) &= C_{10}^{11} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^{10} \\ &= 10 \times (0.0278) \times (0.1615) = 0.0449 \end{aligned}$$

Theorem: (Memory Loss Property)

If X has the geometric distribution with parameter p then

$$P(X \geq i + j \mid X \geq i) = P(X \geq j) \quad ; \quad i, j = 1, 2, 3, \dots \dots \dots$$

Proof:

$$P(X \geq i + j \mid X \geq i) = \frac{P(X \geq i + j)}{P(X \geq i)} = \frac{(1-p)^{i+j}}{(1-p)^i} = (1-p)^j = P(X \geq j)$$

7) The hyper-geometric distribution :

Suppose that we have population of N objects; D of one type and $N-D$ a second type. A random sample of size n is drawn from the population without replacement. Then if we let x denote the number of objects of the first type selected, we get first of all $\binom{D}{x}$ ways of choosing x object of the first type, $\binom{N-D}{n-x}$ ways of choosing $n-x$ object of the second type, $\binom{N}{n}$ ways of choosing a sample of size n from the population of N objects then the r.v. X has P.m.f is given by:

$$f(x) = f(x; N, D, n) = \begin{cases} \frac{\binom{D}{x} \binom{N-D}{n-x}}{\binom{N}{n}} & ; x = a, a+1, a+2, \dots, b \\ 0 & ; \text{otherwise} \end{cases}$$

where ❶ $a = \max\{0, n+D-N\}$ and $b = \min\{n, D\}$

❷ $0 \leq x \leq D$ ❸ $0 \leq n-x \leq N-D$

And denoted by $X \sim \text{hyp}(N, D, n)$

clear that $f(x) \geq 0 \quad \forall x$

$$\text{and} \quad \sum_{x=0}^n f(x) = \sum_{x=0}^n \frac{\binom{D}{x} \binom{N-D}{n-x}}{\binom{N}{n}} = \frac{1}{\binom{N}{n}} \sum_{x=0}^n \binom{D}{x} \binom{N-D}{n-x} = \frac{\binom{N}{n}}{\binom{N}{n}} = 1$$

Useful relationship: $\sum_{k=0}^m \binom{A}{k} \binom{B}{m-k} = \binom{A+B}{m}$

Then $f(x)$ satisfies the condition of begin a P.m.f. of discrete type of random variable x

Theorem: $X \sim \text{hyp}(N, D, n)$ then $\mu = n \frac{D}{N}$ and $\sigma^2 = n \frac{D}{N} \frac{N-n}{N-1} \frac{N-D}{N}$

Proof :

$$\mu = E(x) = \sum_{\forall x} x f(x) = \sum_{x=0}^n x \cdot \frac{\binom{D}{x} \binom{N-D}{n-x}}{\binom{N}{n}}$$

$$\therefore C_x^m = \frac{m!}{x! (m-x)!}$$

$$\begin{aligned} \therefore \mu &= \sum_{x=1}^n x \cdot \frac{\binom{D}{x} \binom{N-D}{n-x}}{\binom{N}{n}} = \frac{1}{\binom{N}{n}} \sum_{x=1}^n x \frac{D!}{x! (D-x)!} \binom{N-D}{n-x} \\ &= \frac{D}{\binom{N}{n}} \sum_{x=1}^n \frac{(D-1)!}{(x-1)! (D-x)!} \binom{N-D}{n-x} \end{aligned}$$

let $y = x - 1$, $m = n - 1$ and $A = D - 1$

$$\begin{aligned} &= \frac{D}{\binom{N}{n}} \sum_{y=0}^m \frac{A!}{y! (A-y)!} \binom{N-A-1}{m-y} \\ &= \frac{D}{\binom{N}{n}} \sum_{y=0}^m \binom{A}{y} \binom{N-A-1}{m-y} \\ &= \frac{D}{\binom{N}{n}} \binom{N-1}{m} = \frac{D}{\binom{N}{n}} \binom{N-1}{n-1} = \frac{D}{\cancel{\binom{N}{n}}} \frac{n}{N} \cancel{\binom{N}{n}} \\ \Rightarrow \mu &= E(x) = n \frac{D}{N} \end{aligned}$$

$$\therefore \sigma^2 = \text{var}(x) = E(x^2) - E^2(x)$$

We must compute $E(x^2)$ to that we compute $E[x(x-1)] = E(x^2) - E(x)$

$$\begin{aligned} E[x(x-1)] &= \sum_{\forall x} x(x-1) f(x) = \sum_{x=0}^n x(x-1) \frac{\binom{D}{x} \binom{N-D}{n-x}}{\binom{N}{n}} \\ \therefore E[x(x-1)] &= \frac{1}{\binom{N}{n}} \sum_{x=2}^n \frac{D!}{(x-2)! (D-x)!} \binom{N-D}{n-x} \\ &= \frac{D(D-1)}{\binom{N}{n}} \sum_{x=2}^n \frac{(D-2)!}{(x-2)! (D-x)!} \binom{N-D}{n-x} \end{aligned}$$

let $y = x - 2$, $m = n - 2$ and $A = D - 2$

$$\begin{aligned}
 &= \frac{D(D-1)}{\binom{N}{n}} \sum_{y=0}^m \frac{A!}{y! (A-y)!} \binom{N-A-2}{m-y} \\
 &= \frac{D(D-1)}{\binom{N}{n}} \sum_{y=0}^m \binom{A}{y} \binom{N-A-2}{m-y} \\
 &= \frac{D(D-1)}{\binom{N}{n}} \binom{N-2}{m} = \frac{D}{\binom{N}{n}} \binom{N-2}{n-2} \\
 &= \frac{D(D-1)}{\cancel{\binom{N}{n}}} \frac{n(n-1)}{N(N-1)} \cancel{\binom{N}{n}} = \frac{D(D-1)}{N} \frac{n(n-1)}{(N-1)} \\
 \Rightarrow E(x^2) &= \frac{D(D-1)}{N} \frac{n(n-1)}{(N-1)} + \frac{nD}{N} \\
 \therefore \sigma^2 = E(x^2) - E^2(x) &= \frac{D(D-1)}{N} \frac{n(n-1)}{(N-1)} + \frac{nD}{N} - \left(\frac{nD}{N}\right)^2 \\
 &= \frac{nD}{N} \left[\frac{(D-1)(n-1)}{(N-1)} + 1 - \frac{nD}{N} \right] \\
 &= \frac{nD}{N} \left[\frac{N(n-1)(D-1) + N(N-1) - nD(N-1)}{N(N-1)} \right] \\
 &= \frac{nD}{N} \left[\frac{\cancel{nND} - nN - ND + \cancel{N} + N^2 - \cancel{N} - \cancel{nND} + nD}{N(N-1)} \right] \\
 &= \frac{nD}{N} \left[\frac{N^2 - nN - ND + nD}{N(N-1)} \right] \\
 &= \frac{nD}{N} \left[\frac{N(N-n) - D(N-n)}{N(N-1)} \right] \\
 &= \frac{nD}{N} \left[\frac{(N-n)(N-D)}{N(N-1)} \right] \\
 \therefore \sigma^2 &= n \frac{D}{N} \frac{N-n}{N-1} \frac{N-D}{N}
 \end{aligned}$$

The moment generating function of **Negative Binomial** distribution is given by

$$\begin{aligned} M_x(t) &= E(e^{tx}) = \sum_{\forall x} e^{tx} f(x) = \sum_{\forall x} e^{tx} \frac{\binom{D}{x} \binom{N-D}{n-x}}{\binom{N}{n}} \\ &= 1 + \sum_{k=1}^{\infty} [F(k+1) - 1] \frac{p}{k!} \end{aligned}$$

Where: $F(k)$ is the sum of the first k -coefficient of the hyper-geometric series.

Note : If we set $p = \frac{D}{N}$ then the mean of the hyper-geometric distribution coincides with the mean of the binomial distribution, and the variance is $\frac{N-n}{N-1}$ times of the variance of binomial distribution.

Some continues Distribution:

1) Continues uniform distribution (Rectangular distribution) :

The r.v. X is said to have an uniform distribution if P.d.f. is give by

$$f(x, a, b) = \begin{cases} \frac{1}{b-a} & ; a \leq x \leq b \\ 0 & ; o.w. \end{cases} ; \text{Where } a \text{ and } b \text{ it satisfy } -\infty < a < b < \infty$$

And denoted by $X \sim C_u(a, b)$

Clear that satisfy

$$f(x) \geq 0 \quad \forall x \quad \text{and} \quad \int_a^b f(x) dx = 1$$

Then $f(x)$ satisfies the condition of begin a P.d.f. of continues type of random variable x .

Note : the mean $\mu = \frac{a+b}{2}$ and the variance $\sigma^2 = \frac{(b-a)^2}{12}$ (proof)

The moment generating function of uniform distribution is given by

$$\begin{aligned} M_x(t) &= E(e^{tx}) = \int_a^b e^{tx} f(x) dx = \int_a^b e^{tx} \frac{1}{b-a} dx \\ &= \frac{1}{b-a} \int_a^b e^{tx} dx = \frac{e^{bt} - e^{at}}{t(b-a)} \end{aligned}$$

Example : If x is uniform distribution over $(0, 10)$, calculate the probability that :

- ❶ $P(X < 3)$ ❷ $P(3 < X < 8)$ ❸ $P(X > 6)$ ❹ The distribution function

Solution : $X \sim C_u(a, b)$ $\therefore X \sim C_u(0, 10)$

$$f(x, a, b) = \begin{cases} \frac{1}{b-a} & ; a \leq x \leq b \\ 0 & ; o.w. \end{cases} \Rightarrow f(x, 0, 10) = \begin{cases} \frac{1}{10} & ; 0 \leq x \leq 10 \\ 0 & ; o.w. \end{cases}$$

$$\text{❶ } P(X < 3) = \int_0^3 \frac{1}{10} dx = \left[\frac{x}{10} \right]_0^3 = \frac{3}{10}$$

$$\text{❷ } P(X > 6) = \int_6^{10} \frac{1}{10} dx = \left[\frac{x}{10} \right]_6^{10} = \frac{10}{10} - \frac{6}{10} = \frac{4}{10}$$

$$\text{❸ } P(3 < X < 8) = \int_3^8 \frac{1}{10} dx = \left[\frac{x}{10} \right]_3^8 = \frac{8}{10} - \frac{3}{10} = \frac{5}{10}$$

$$\text{❹ } F(x) = P(X < x) = \int_{-\infty}^x f(u) du = \int_0^x \frac{1}{10} du = \left[\frac{u}{10} \right]_0^x = \frac{x}{10}$$

$$\Rightarrow F(x) = \begin{cases} 0 & ; x \leq 0 \\ \frac{x}{10} & ; 0 < x < 10 \\ 1 & ; x \geq 10 \end{cases}$$

2) Normal distribution :

The normal distribution is one of the widely used in application of statistical methods the P.d.f. of X is give by:

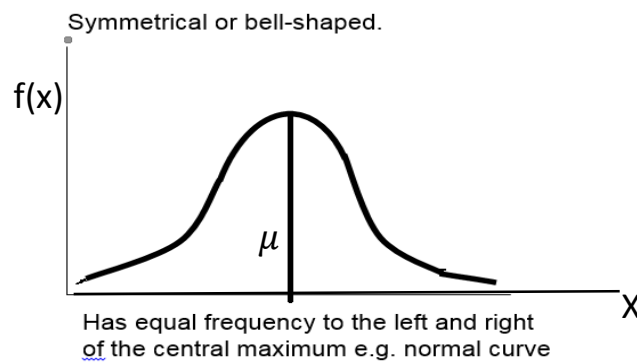
$$f(x; \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad ; -\infty < x < \infty$$

Where the parameters μ and σ^2 satisfy $-\infty < \mu < \infty$ and $\sigma^2 > 0$,

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

The properties of this p.d.f. are

- 1) It is a bell shaped curve is symmetric about μ



- 2) The two parameter that appear in the density μ and σ^2 represent the mean and the variance of the random variable X.

Now

Consider the integral

$$I = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

$$\text{let } y = \frac{x - \mu}{\sigma} \Rightarrow dy = \frac{dx}{\sigma} \Rightarrow dx = \sigma dy$$

$$\therefore -\infty < x < \infty \quad \therefore -\infty < y < \infty \Rightarrow I = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} dy$$

$$\therefore I = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} dx \quad \text{and} \quad I = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} dy$$

$$\therefore I^2 = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} dx \cdot \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} dy$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} dx dy$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-\frac{1}{2}x^2} \cdot e^{-\frac{1}{2}y^2} dx dy$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(x^2 + y^2)} dx dy$$

We change the variables to polar coordination by using

$$x = r \cos \theta \quad \text{and} \quad y = r \sin \theta$$

$$\begin{aligned} \therefore I^2 &= \frac{1}{2\pi} \int_0^{\infty} \int_0^{2\pi} r e^{-\frac{1}{2}r^2} d\theta dr = \frac{1}{2\pi} \int_0^{\infty} r e^{-\frac{1}{2}r^2} [\theta]_0^{2\pi} dr \\ &= \int_0^{\infty} r e^{-\frac{1}{2}r^2} dr = \left[-e^{-\frac{1}{2}r^2} \right]_0^{\infty} = 1 \Rightarrow I = 1 \end{aligned}$$

Then $f(x)$ satisfies the condition of begin a P.d.f. of continues type to random variable x .

The moment generating function of uniform distribution is given by

$$\begin{aligned}
 M_x(t) &= E(e^{tx}) = \int_{-\infty}^{\infty} e^{tx} f(x) dx = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} e^{tx} \cdot e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx \\
 &= \frac{1}{\sqrt{2\pi\sigma^2}} \int_{-\infty}^{\infty} e^{-\frac{1}{2\sigma^2}[2x\sigma^2 t + x^2 - 2\mu x + \mu^2]} dx \\
 &= \frac{1}{\sqrt{2\pi\sigma^2}} \int_{-\infty}^{\infty} e^{-\frac{1}{2\sigma^2}[x^2 - 2x(\mu + \sigma^2 t) + \mu^2]} dx \\
 &= \frac{1}{\sqrt{2\pi\sigma^2}} \int_{-\infty}^{\infty} e^{-\frac{1}{2\sigma^2}[x^2 - 2x(\mu + \sigma^2 t) + \mu^2 + (\mu + \sigma^2 t)^2 - (\mu + \sigma^2 t)^2]} dx \\
 &= \frac{1}{\sqrt{2\pi\sigma^2}} \int_{-\infty}^{\infty} e^{-\frac{1}{2\sigma^2}[x^2 - 2x(\mu + \sigma^2 t) + (\mu + \sigma^2 t)^2]} e^{\frac{-\mu^2 + (\mu + \sigma^2 t)^2}{2\sigma^2}} dx \\
 &= \frac{e^{\frac{-\mu^2 + (\mu + \sigma^2 t)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}} \int_{-\infty}^{\infty} e^{-\frac{1}{2\sigma^2}[x^2 - 2x(\mu + \sigma^2 t) + (\mu + \sigma^2 t)^2]} dx \\
 &= \frac{e^{\frac{-\mu^2 + (\mu + \sigma^2 t)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}} \int_{-\infty}^{\infty} e^{-\frac{1}{2\sigma^2}[x - (\mu + \sigma^2 t)]^2} dx
 \end{aligned}$$

$$\text{If } y = x - (\mu + \sigma^2 t) \Rightarrow \frac{1}{\sqrt{2\pi\sigma^2}} \int_{-\infty}^{\infty} e^{-\frac{1}{2\sigma^2}[x - (\mu + \sigma^2 t)]^2} dx = I = 1$$

$$\therefore M_x(t) = e^{\frac{-\mu^2 + (\mu + \sigma^2 t)^2}{2\sigma^2}} = e^{\frac{-\mu^2 + \mu^2 + 2\mu\sigma^2 t + (\sigma^2 t)^2}{2\sigma^2}}$$

$$\Rightarrow M_x(t) = e^{\mu t + \frac{1}{2}\sigma^2 t^2}$$

Note : The mean μ and the variance σ^2 of normal distribution will be calculated from $M_x(t)$ as following :

Since $M_x(t) = e^{\mu t + \frac{1}{2}\sigma^2 t^2}$, $E(x) = M'_x(0)$ and $E(x^2) = M''_x(0)$

$$M'_x(t) = e^{\mu t + \frac{1}{2}\sigma^2 t^2} (\mu + \sigma^2 t) \Rightarrow M'_x(0) = \mu$$

$$M''_x(t) = e^{\mu t + \frac{1}{2}\sigma^2 t^2} \cdot \sigma^2 + e^{\mu t + \frac{1}{2}\sigma^2 t^2} (\mu + \sigma^2 t)^2 \Rightarrow M''_x(0) = \sigma^2 + \mu^2$$

$$\therefore \text{var}(x) = E(x^2) - E^2(x) \Rightarrow \text{var}(x) = \sigma^2 + \mu^2 - \mu^2 = \sigma^2$$

3) Standard normal distribution

If the normal random variable Z has mean zero and variance one instead of μ and σ^2 it is called standard normal distribution with P.d.f that:

$$f(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} ; -\infty < z < \infty$$

And denoted by $Z \sim N(0,1)$

Theorem : If the r.v $X \sim N(\mu, \sigma^2)$, $\sigma^2 > 0$; then the r.v $Z = \frac{X - \mu}{\sigma} \sim N(0,1)$

Note : ① If the r.v $X \sim N(\mu, \sigma^2)$, $\sigma^2 > 0$; then $P(a < x < b) = Z\left(\frac{b - \mu}{\sigma}\right) - Z\left(\frac{a - \mu}{\sigma}\right)$

② We can write $Z(x)$ as the following forms $N(x)$, $\Phi(x)$ or $F(x)$.

Remark: $N(-x) = 1 - N(x)$ or $Z(-x) = 1 - Z(x)$.

Note: $M_z(t) = e^{\frac{1}{2}t^2}$ and $Z(x) = N(x) = F(x) = Pr(Z \leq x) = Pr\left(\frac{X - \mu}{\sigma} \leq x\right)$

Example: Let $X \sim N(3,4)$. Find $P(X \leq 4)$

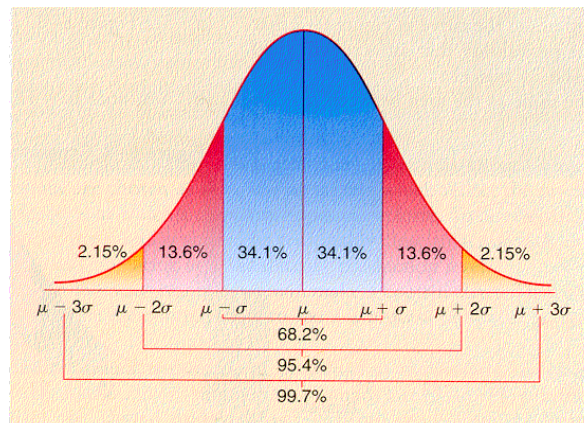
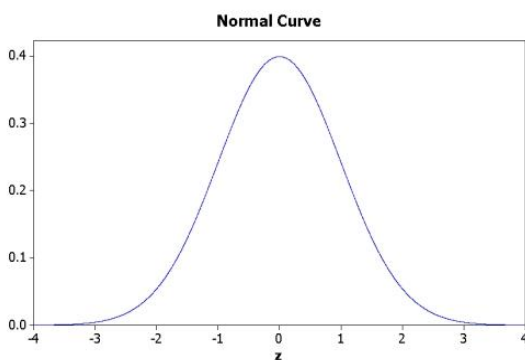
Solution :

$$\text{since } X \sim N(3,4) \Rightarrow \mu = 3 \text{ and } \sigma^2 = 4 \Rightarrow \sigma = 2$$

$$P(X \leq 4) = P\left(\frac{X - \mu}{\sigma} \leq \frac{4 - \mu}{\sigma}\right) = P\left(Z \leq \frac{4 - 3}{2}\right) = Z\left(\frac{1}{2}\right) \text{ or } \Phi\left(\frac{1}{2}\right)$$

From Z table (standard normal table) we get $\Phi\left(\frac{1}{2}\right) = Z\left(\frac{1}{2}\right) = 0.6915$

$$\therefore P(X \leq 4) = 0.6915$$



Areas Under the Normal Curve