

Moment Generating Functions (mgf)

Let x be the random variable then mgf of x

is denoted by $\boxed{M_x(t)}$ and is defined as

$$M_x(t) = \sum_{i=1}^{\infty} e^{tx_i} P(x=x_i) \rightarrow \text{if } x \text{ is discrete R.V}$$

$$= \int_{-\infty}^{\infty} e^{tx} f(x) dx \rightarrow \text{if } x \text{ is continuous R.V.}$$

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

Here ' t ' is a parameter.

Theorem:

The n^{th} derivative of $M_x(t)$ at $t=0$ gives $E(x^n)$.

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

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

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

$$\downarrow \\ M_x(t) = E(1) + t E(x) + \frac{t^2 E(x^2)}{2!} + \dots + \frac{t^n E(x^n)}{n!} + \dots$$

diff w.r.t. to ' t '

+ ...

$$\frac{d}{dt} M_x(t) = 0 + E(x) + \frac{2t}{2!} E(x^2) + \dots + \frac{n t^{n-1}}{n!} E(x^n)$$

Put $t = 0$

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

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

Similarly

$$M''_x(t) = 0 + E(x^2) + \dots +$$

$$M''_x(t) \Big|_{t=0} = E(x^2)$$

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

$$\Rightarrow \text{Similarly, } M^n_x(t) \Big|_{t=0} = E(x^n)$$

$$\Rightarrow V(x) = E(x^2) - (E(x))^2$$

$$V(x) = M''_x(0) - [M'_x(0)]^2$$

Properties of mgf:

1) $M_{ax}(t) = M_x(at)$ ✓

Proof: $M_{ax}(t) = E(e^{t(ax)})$ ∵ Defⁿ of mgf.
 $= E(e^{(at)x})$
 $= M_x(at)$

2) $M_{ax+b}(t) = e^{bt} M_x(at)$ ✓ ∵ a & b are constants

Proof: $M_{ax+b}(t) = E(e^{t(ax+b)})$
 $= E(e^{axt} \cdot e^{bt})$
 $= e^{bt} E(e^{axt})$
 $= e^{bt} M_x(at)$ ✓

3) If x & y are independent random variables

then $M_{x+y}(t) = M_x(t) \cdot M_y(t)$,

Proof: $M_{x+y}(t) = E(e^{(x+y)t})$
 $= E(e^{xt} \cdot e^{yt})$
 $= E(e^{xt}) \cdot E(e^{yt})$ ∵ $E(xy) = E(x) \cdot E(y)$
when x & y are independent

$M_{x+y}(t) = M_x(t) \cdot M_y(t).$

Note: $E(x^n)$ is the coefficient of $\frac{t^n}{n!}$ in the expansion
of $E(e^{tx})$ /mgf

Proof: $E(e^{tx}) = E \left\{ 1 + tx + \frac{t^2 x^2}{2!} + \dots + \frac{t^n x^n}{n!} + \dots \right\}$

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

$+ \dots - \cancel{\dots}$

\Rightarrow $E(x^n)$ = coefficient of $\frac{t^n}{n!}$ in the above eqⁿ.

Note 2: whenever a random variable is given to find $E(x)$, using mgf we can directly find, instead of finding pdf & proceeding.

\Rightarrow This is an alternative way to find mean & variance.

Examples:

1) Let x be the outcome when a die is rolled.
Find mgf of x , also find mean & variance
of x .

Solⁿ: $X: 1, 2, 3, 4, 5, 6 \quad \checkmark$
 $P(x): \frac{1}{6} \frac{1}{6} \frac{1}{6} \frac{1}{6} \frac{1}{6} \frac{1}{6} \quad \checkmark$

$$\begin{aligned} M_X(t) &= E(e^{tx}) \\ &= \sum_{x=1}^6 e^{tx} P(x) \quad \therefore \text{Def } E(x) \\ &= \frac{1}{6} e^t + \frac{1}{6} e^{2t} + \frac{1}{6} e^{3t} + \frac{1}{6} e^{4t} + \frac{1}{6} e^{5t} + \frac{1}{6} e^{6t} \\ M_X(t) &= \frac{1}{6} \left\{ e^t + e^{2t} + e^{3t} + e^{4t} + e^{5t} + e^{6t} \right\} - \text{⊗} \end{aligned}$$

We know that $E(x) = M_X'(0)$

$$\begin{aligned} &= \frac{1}{6} \left\{ e^{(0)} + 2e^{2(0)} + 3e^{3(0)} + 4e^{4(0)} + 5e^{5(0)} + 6e^{6(0)} \right\} \\ \Rightarrow E(x) &= \frac{1+2+3+4+5+6}{6} = \boxed{\frac{21}{6}} \quad \checkmark \end{aligned}$$

Next $V(x) = E(x^2) - (E(x))^2$

$$\begin{aligned} \Rightarrow E(x^2) &= M_X''(0) \\ &= \frac{1}{6} \left\{ e^{(0)} + 4e^{2(0)} + 9e^{3(0)} + 16e^{4(0)} + 25e^{5(0)} + 36e^{6(0)} \right\} \\ &= \frac{91}{6} \end{aligned}$$

$$\Rightarrow V(x) = \frac{91}{6} - \left(\frac{21}{6}\right)^2 = \boxed{\frac{35}{12}} = 2.9166 \quad \checkmark$$

2) If x is a random variable taking the values $0, 1, 2, \dots \infty$ and $P(x) = ab^x$ where a, b are +ve numbers such that $a+b=1$

i) find the mgf of x

ii) if $E(x) = m_1$, $E(x^2) = m_2$ Show that

$$m_2 = m_1(2m_1 + 1)$$

→ we need to prove

Soln: We know that $M_x(t) = E(e^{tx})$

Given : $P(x) = ab^x$ $\forall x: 0, 1, 2, \dots \infty$

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

$$= \sum_{x=0}^{\infty} e^{tx} P(x)$$

$$= \sum_{x=0}^{\infty} e^{tx} ab^x$$

$$= a \sum_{x=0}^{\infty} (be^t)^x$$

$$= a \left\{ 1 + be^t + b^2 e^{2t} + b^3 e^{3t} + \dots \right\}$$

$$M_x(t) = a \left(\frac{1}{1-be^t} \right)$$

Geometric progression
where $|be^t| < 1$

Now

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

$$= a \left\{ \frac{(1-be^t)(0) - (-be^t)}{(1-be^t)^2} \right\}_{t=0} \quad (*)$$

$$\frac{u}{\sqrt{v}} = \frac{u'v - v'u}{\sqrt{v^2}}$$

$$= \left. \frac{abe^t}{(1-be^t)^2} \right|_{t=0} = \frac{ab}{(1-b)^2} = \boxed{\frac{b}{a}} \quad ; \quad a+b=1$$

$$M'_X(t) = \frac{ab e^t}{(1-b e^t)^2} \quad \text{---} \quad \text{**}$$

$$E(x) = M''_X(t) \Big|_{t=0} = a \left\{ \frac{(1-b e^t)^2 (b e^t) - (b e^t) 2(1-b e^t)(-b e^t)}{(1-b e^t)^4} \right\}_{t=0}$$

$$= a \left\{ \frac{(1-b)^2 b + b^2 2(1-b)}{(1-b)^4} \right\}$$

$$= a \left\{ \cancel{(1-b)} \left[(1-b)b + 2b^2 \right] \right\}_{(1-b)^4}$$

$$= a \left\{ \frac{b - b^2 + 2b^2}{(1-b)^3} \right\}$$

$$= \frac{a(1+b)b}{(1-b)^3}$$

$$E(x^2) = \frac{(1-b)(1+b)b}{(1-b)^3} = \frac{b(1+b)}{(1-b)^2} = \frac{b(1+b)}{a^2}$$

ii) Given $E(x) = m_1$, $E(x^2) = m_2$

$$m_2 = \frac{b(1+b)}{a^2} = \frac{b}{a^2} (1+b)$$

$$= \frac{b}{a} \left(\frac{1+b}{a} \right)$$

$$= m_1 \left(\frac{1}{a} + \frac{b}{a} \right) \checkmark$$

$$m_2 = m_1 \left(\frac{a+b}{a} + \frac{b}{a} \right) = m_1 \left(1 + \frac{2b}{a} \right)$$

$$\boxed{\frac{b}{a} = m_1}$$

$$\boxed{a+b = 1}$$



$$\boxed{m_2 = m_1 (1 + 2m_1)}$$

3) Find the mgf of a random variable which is uniformly distributed from $-a < x < a$. Hence evaluate $E(x^{2n})$

Solⁿ: Let $x \sim U(-a, a)$ \because given

$$\Rightarrow f(x) = \frac{1}{2a} \quad \checkmark$$

$$f(x) = \frac{1}{b-a}$$

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

$$= \int_{-a}^a e^{tx} \cdot \frac{1}{2a} dx = \frac{1}{2a} \cdot \left[\frac{e^{tx}}{t} \right]_{-a}^a$$

$$M_x(t) = \frac{1}{at} \left(\frac{e^{at} - e^{-at}}{2} \right) \quad \star$$

$$M_x(t) = \frac{1}{at} \sinhat$$

$$\therefore \sinhat = \frac{e^x - e^{-x}}{2}$$

To find $E(x^{2n})$ we know that

$$E(x^n) = \text{Coefficient of } \frac{t^n}{n!} \text{ of } M_x(t)$$

Similarly, $E(x^{2n}) = \text{Coefficient of } \frac{t^{2n}}{2n!} \text{ of } M_x(t)$

$$\Rightarrow M_x(t) = \frac{\sinhat}{at} \quad \checkmark$$

$$M_x(t) = \frac{\sin \hat{a}t}{at} \quad \checkmark$$

check
power series
for $\sinh x$

$$= \frac{1}{at} \left\{ at + \frac{(at)^3}{3!} + \frac{(at)^5}{5!} + \dots + \frac{(at)^{2n+1}}{(2n+1)!} + \dots \right\}$$

$$= \frac{at}{at} \left\{ 1 + \frac{(at)^2}{3!} + \frac{(at)^4}{5!} + \dots + \frac{(at)^{2n}}{(2n+1)!} + \dots \right\}$$

$$M_x(t) = \left\{ 1 + \frac{(at)^2}{3!} + \frac{(at)^4}{5!} + \dots + \frac{a^{2n}}{(2n+1)} \cdot \frac{t^{2n}}{2n!} + \dots \right\}$$

$$\frac{3!}{3^2 \cdot 2!}$$

$$\Rightarrow E(x^{2n}) = \text{co-eff}^{ht} \text{ of } \frac{t^{2n}}{2n!} \text{ in } M_x(t)$$

From \star

$$E(x^{2n}) = \frac{a^{2n}}{2n+1}$$

$$E(x^2) = \frac{a^2}{3}$$

$$E(x) = \frac{a}{2}$$

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

$$= \frac{a}{2}$$

4) Find the mgf of Binomial distribution.
Hence find the mean & variance of Binomial distribution.

Solⁿ: If $X \sim B(n, P)$

$$P(x) = {}^n C_x P^x q^{n-x}, \quad x=0, 1, 2, \dots, n$$

$$p+q=1$$

$$M_X(t) = E(e^{tx}) - \textcircled{A}$$

$$= \sum_{x=0}^n e^{tx} \cdot {}^n C_x P^x q^{n-x} \rightarrow P(t)$$

$$= \sum_{x=0}^n (e^t p)^x \cdot {}^n C_x q^{n-x} - \textcircled{*}$$

From the Binomial expansion we know that

$$\sum_{x=0}^{\infty} {}^n C_x P^x q^{n-x} = (p+q)^n - \textcircled{**}$$

\Rightarrow From ear $\textcircled{**}$

$M_X(t) = (pe^t + q)^n - \textcircled{1}$

$$\Rightarrow E(X) = M'_X(0)$$

$$M'_X(t) = n(pe^t + q)^{n-1} \cdot pe^t$$

from ① $M'_x(t) = n(p e^t + q)^{n-1} \cdot p e^t$ — ②

when $t=0$

$$M'_x(0) = n(p+q)^{n-1} \cdot p \quad \checkmark$$

$$= n(1)p$$

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

3

$\Rightarrow V(x) = E(x^2) - (E(x))^2$

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

From ② $M''_x(t) = np \left[e^{2t} (n-1) (p e^t + q)^{n-2} \cdot p e^t + (p e^t + q)^{n-1} \cdot e^t \right]$

when $t=0$

$$M''_x(0) = np \left[(n-1)(1)p + (1) \right]$$

$$= np [np - p + 1]$$

$$M''_x(0) = n^2 p^2 - np^2 + np = E(x^2) - ④$$

$\Rightarrow V(x) = n^2 p^2 - np^2 + np - n^2 p^2$

$$= np(1-p)$$

$$\therefore p+q=1$$

$V(x) = npq$

$$1-p=q$$

$$V(x) = npq$$

\Rightarrow mgf of Poisson Distribution

Let $x \sim P(\alpha)$

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

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

Defⁿ.

$$= \sum_{x=0}^{\infty} e^{tx} \cdot \frac{e^{-\alpha} \cdot \alpha^x}{x!}$$

— A

$$\begin{aligned}
 &= \bar{e}^\alpha \left\{ 1 + \frac{e^t \alpha}{1!} + \frac{(e^t \alpha)^2}{2!} + \dots + \frac{(e^t \alpha)^n}{n!} + \dots \right\} \\
 &= \bar{e}^\alpha e^{e^t \alpha} \\
 &= \frac{e^{e^t \alpha - \alpha}}{e^{\alpha(e^t - 1)}} \quad \xrightarrow{(1)} \text{mgf of P. D. . .}
 \end{aligned}$$

To find mean

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

from ① ,

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

when $t = 0$

$$M'_x(0) = \alpha = E(x) \quad \xrightarrow{(3)}$$

$$\Rightarrow V(x) = E(x^2) - (E(x))^2$$

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

from ② ,

$$M''_x(t) = \alpha \left[e^t \cdot e^{\alpha(e^t - 1)} + e^t e^{\alpha(e^t - 1)} \cdot \alpha e^t \right]$$

when $t = 0$

$$M''_x(0) = \alpha(1 + \alpha)$$

$$\Rightarrow V(x) = \alpha + \alpha^2 - \alpha^2$$

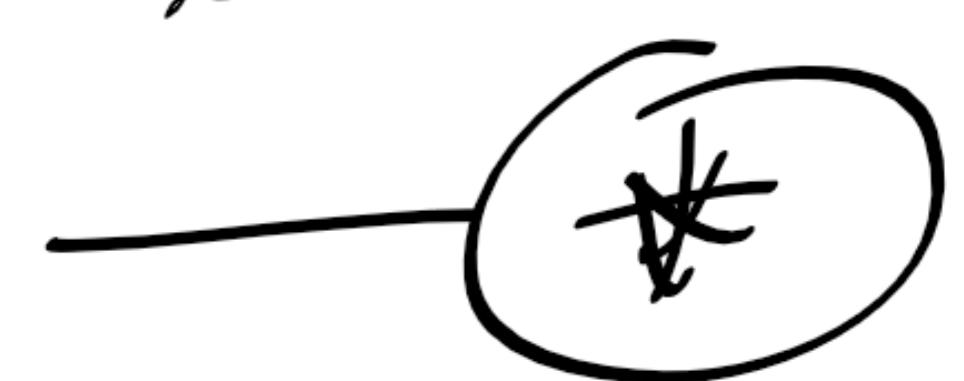
$$V(x) = \alpha \quad \xrightarrow{(4)}$$

\Rightarrow Moment generating function of Gamma Distribution

Let $x \sim G(r, \alpha)$

pdf of x is given by

$$f(x) = \frac{\alpha^{-\alpha x} (x^\alpha)^{r-1}}{\Gamma(r)}, \quad x > 0$$



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

$$= \int_{-\infty}^{\infty} e^{tx} f(x) dx \quad \rightarrow \quad \because \text{by Defn of } E(x)$$

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

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

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

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

$$\text{Put } (\alpha-t)x = u \quad \Rightarrow x = \frac{u}{\alpha-t}$$

$$dx = \frac{du}{\alpha-t}$$

$$\Rightarrow M_x(t) = \frac{\alpha^r}{\Gamma(r)} \int_0^{\infty} \bar{e}^u \left(\frac{u}{\alpha-t}\right)^{r-1} \frac{du}{\alpha-t}$$

$$= \frac{\alpha^r}{\Gamma(r) (\alpha-t)^r} \int_0^{\infty} \bar{e}^u u^{r-1} du \rightarrow \Gamma(r)$$

$$= \frac{\alpha^r}{\cancel{\alpha} (\alpha-t)^r} \quad \cancel{\text{if } t < \alpha}$$

mgf of
G. D.

$M_x(t) = \frac{\alpha^r}{(\alpha-t)^r}$

— ①

To find $E(x)$

From ①, $M'_x(t) = \alpha^r \left\{ -\frac{(\alpha-t)^{r-1} \cdot r(-1)}{(\alpha-t)^{2r}} \right\}$

$$= \frac{r\alpha^r (\alpha-t)^{r-1}}{(\alpha-t)^{2r}} = \frac{r\alpha^r}{(\alpha-t)^{r+1}} \quad — ②$$

$$M'_x(0) = \frac{r\alpha^r}{\alpha^{r+1}} = \frac{r}{\alpha}.$$

$M'_x(0) = \frac{r}{\alpha} = E(x)$

— ③

From ②

$$M''_x(t) = \frac{r\alpha^r (r+1) (\alpha-t)^{r-2}}{(\alpha-t)^{2r+2}}$$

when $t=0$

$$M''_x(0) = \frac{r(r+1)}{\alpha^2} = E(x^2).$$

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

$$V(x) = \frac{r^2 + r}{\alpha^2} - \frac{r^2}{\alpha^2} = \boxed{\frac{r}{\alpha^2}} \quad — ④$$

\Rightarrow mgf of Exponential Distribution

Let $X \sim \text{Exp}(\alpha)$ then,

pdf of X is given by

$$f(x) = \alpha e^{-\alpha x} \quad \forall x > 0$$

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

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

$$= \alpha \left[\frac{e^{-(\alpha-t)x}}{-(\alpha-t)} \right]_0^\infty$$

$$= \frac{\alpha}{-(\alpha-t)} \left[-e^\infty - e^0 \right]$$

$$M_X(t) = \frac{\alpha}{\alpha-t}$$

— (1)

$$E(X) = M'_X(0)$$

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

$$= \frac{\alpha}{t-\alpha} (\bar{e}^\alpha - e^\alpha)$$

$$M_x(t) = \frac{\alpha}{\alpha-t}$$

1

H. 11) find $E(x)$ & $V(x)$ when $x \sim \exp(\alpha)$

\Rightarrow mgf of χ^2 -distribution

Let $x \sim \chi^2(n)$

\Rightarrow pdf of x is given by

$$f(x) = \frac{e^{-x/2}}{2^{n/2} \Gamma(n/2)} x^{n/2-1} \quad x \geq 0$$

A

$$M_x(t) = \int_0^\infty e^{tx} \frac{e^{-x/2}}{2^{n/2} \Gamma(n/2)} x^{n/2-1} dx$$

$$= \frac{1}{2^{n/2} \Gamma(n/2)} \int_0^\infty e^{-(y_2-t)x} x^{n/2-1} dx$$

$$= \frac{1}{2^{n/2} \Gamma(n/2)} \int_0^\infty e^{(t-y_2)x} x^{n/2-1} dx$$

$$= \frac{1}{2^{n/2} \Gamma(n/2)} \cdot \frac{\Gamma(n/2)}{-(t-y_2)^{n/2}}$$

using
property

$$M_x(t) = (1-2t)^{-n/2}$$

1

of gamma
function

H. 12)
find $E(X)$ & $V(X)$

\Rightarrow mgf of Normal Distribution

Let $x \sim N(\mu, \sigma^2)$

Pdf of x is given by

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx$$

$-\infty < x < \infty$

$$M_x(t) = E(e^{tx}) \quad \text{— By defn.}$$

$$= \int_{-\infty}^{\infty} e^{tx} f(x) dx$$

$$= \int_{-\infty}^{\infty} e^{tx} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx$$

$$= \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{tx} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx$$

Put $\frac{x-\mu}{\sigma} = u \Rightarrow$

$$\boxed{x = \sigma u + \mu}$$
$$\boxed{dx = \sigma du}$$

$$= \frac{\sigma}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{t(u+\sigma u)} e^{-\frac{u^2}{2}} du$$

$$= \frac{e^{ut}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{tu} e^{-\frac{u^2}{2}} du$$

$$= \frac{e^{ut}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(u^2 - 2t u \sigma)} du$$

Completing the square by
adding & subtracting the
term $(t\sigma)^2$

i.e.

$$\begin{aligned} u^2 + 2t u \sigma + (t\sigma)^2 - (t\sigma)^2 \\ \Rightarrow (u + t\sigma)^2 - (t\sigma)^2 \end{aligned}$$

$$\Rightarrow \frac{e^{ut}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}[(u+t\sigma)^2 - (t\sigma)^2]} du$$

$$\Rightarrow \frac{e^{ut + \frac{\sigma^2 t^2}{2}}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{(u+t\sigma)^2}{2}} du$$

put $u + t\sigma = s$

$$du = ds$$

$$\Rightarrow \frac{e^{ut + \frac{\sigma^2 t^2}{2}}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{s^2}{2}} ds$$

even funⁿ.

$$\Rightarrow \frac{e^{ut + \frac{\sigma^2 t^2}{2}}}{\sqrt{2\pi}} \underbrace{(2)}_{0}^{\delta} e^{-\frac{s^2}{2}} ds$$

Put $\frac{s^2}{2} = y \Rightarrow s = \sqrt{2y}$

$$\frac{2s ds}{2} = dy$$

$$s ds = dy$$

$$\Rightarrow \frac{\sqrt{2}}{\sqrt{\pi}} \cdot e^{ut + \frac{\sigma^2 t^2}{2}} \int_0^{\infty} e^{-y} \frac{dy}{\sqrt{2y}}$$

$$\Rightarrow \sqrt{\frac{2}{\pi}} e^{\mu t + \frac{\sigma^2 t^2}{2}} \cdot \frac{1}{\sqrt{2}} \int_0^\infty e^{-y} y^{\frac{1}{2}} dy$$

$$= \frac{1}{\sqrt{\pi}} e^{\mu t + \frac{\sigma^2 t^2}{2}} \cdot r(y_2)$$

$$= \frac{1}{\sqrt{\pi}} e^{\mu t + \frac{\sigma^2 t^2}{2}} \cdot \sqrt{\pi}$$

$M_x(t) =$

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

$$M'_x(t) = e^{\mu t + \frac{\sigma^2 t^2}{2}} \left(\mu + \frac{1}{2} \sigma^2 t \right)$$

when $t=0 \Rightarrow M'_x(0) = \mu = E(x)$

$$E(x^2) = \bar{x}^2 + \sigma^2 \quad (\text{Verify})$$

$$V(x) = \bar{x}^2 + \sigma^2 - \bar{x}^2$$

$$\boxed{V(x) = \sigma^2}$$

- check later.

H.W.

\Rightarrow Let x be the random variable & if its mgf is given by $M_x(t) = \left(\frac{1}{3} + \frac{2}{3}e^t\right)^5$ then

find probability at $x=2$,

$$\& x=3$$

$$\text{Soln: } M_x(t) = \left(\frac{1}{3} + \frac{2}{3}e^t\right)^5$$

it is similar to $(pe^t + q)^n$

where $n=5$, $p=\frac{2}{3}$, $q=\frac{1}{3}$

$$\Rightarrow X \sim B(n, p)$$

$$P(X=x) = {}^n C_x p^x q^{n-x}$$

$$P(X=2) = {}^5 C_2 \left(\frac{2}{3}\right)^2 \left(\frac{1}{3}\right)^3$$

$$= 0.1646$$

$$P(X=3) = 0.3292$$

