

## Memoryless property of Exp dist

$$P[X > t + s \mid X > s] = P[X > t]$$

If  $x$  life time of an instrument, the probability that the instrument lives for more than  $t+s$  years given it has already lasted more than  $s$  years is the same as initial prob that it lives for more  $t$  yrs.

Proof

$$\begin{aligned} \text{RHS} = P[X > t] &= \int_t^{\infty} f(x) dx \\ &= \int_t^{\infty} \lambda e^{-\lambda x} dx \\ &= \lambda \left[ \frac{e^{-\lambda x}}{-\lambda} \right]_t^{\infty} \\ &= -[0 - e^{-\lambda t}] \end{aligned}$$

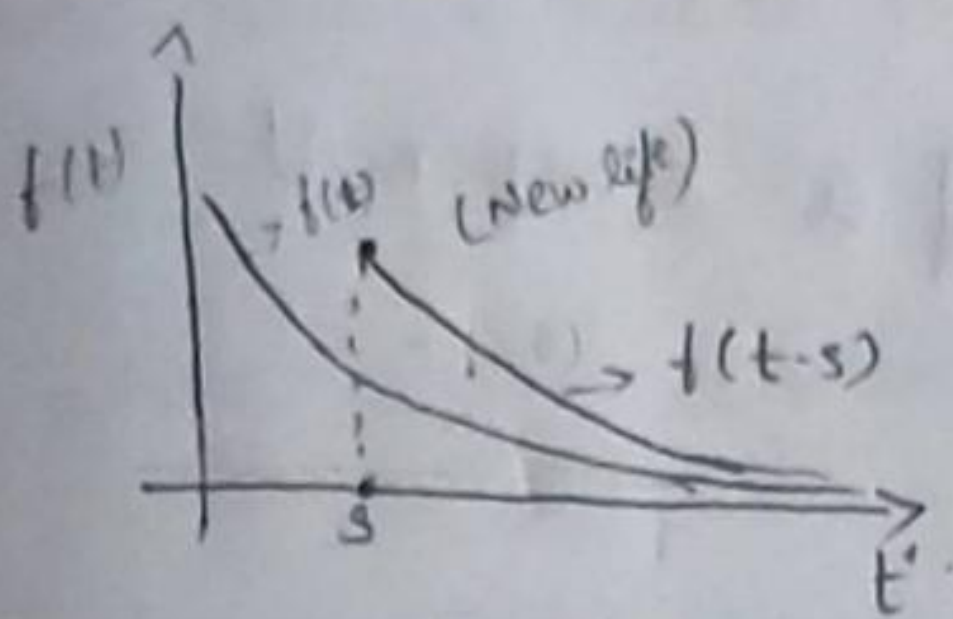
$$P[X > t] = e^{-\lambda t}$$

LHS

$$\begin{aligned} P[X > t+s \mid X > s] &= \frac{P[X > t+s, X > s]}{P[X > s]} \\ &= \frac{P[X > t+s]}{P[X > s]} \\ &= \frac{e^{-(t+s)\lambda}}{e^{-\lambda s}} \end{aligned}$$

$$P[X > t+s \mid X > s] = e^{-\lambda t}$$





$$f(x) = \lambda e^{-\lambda x}, \quad x \geq 0$$

Suppose the ~~amt~~ time one spends in a bank follows ED with a means of 10 mins. What is the probability that a customer will spend more than 15 mins in the bank?

What is the prob that she will spend more than 15 mins given she is still in the bank after 10 mins.

$x \sim$  time spent in the bank (in mins)      mean  $\frac{1}{\lambda} = 10$

$$x \sim \text{Expo} \left( \lambda = \frac{1}{10} \right)$$

WKT

$$P[x > t] = e^{-\lambda t}$$

$$P[x > 15] = e^{-\frac{1}{10}(15)} = e^{-1.5}$$

$$P[x > 10+5 \mid x > 10] = P[x > 5] \quad (\text{Memoryless property})$$

$$= e^{-\frac{1}{10}(5)} = e^{-0.5}$$

\* the question before special continuous dists: guest arriving

$\lambda = 5$  per hour

$\lambda = \frac{5}{60}$  per minutes

$x$  - time for the arrival of next guest

$$x \sim \text{Expo} \left( \lambda = \frac{1}{12} \right)$$

→ guest not arriving till 10 mins  
→ guest not arriving till 10 mins and arriving after 2 mins

$$1 - P[x > 10+2 \mid x > 10] = 1 - P[x > 2] = 1 - e^{-\frac{1}{12}(2)} = 1 - e^{-\frac{1}{6}}$$

$\downarrow$   
 $P[\text{Next guest arriving after 2 mins}]$



Weibull distribution:

$x$ : time to failure of a component

If pdf of  $x$  is

$$f(x) = \alpha \beta x^{\beta-1} e^{-\alpha x^\beta} \quad ; x \geq 0$$

then  $x$  follows weibull distribution

parameters  $\alpha$  and  $\beta$

scale parameter

shape parameter

Mean & variance:

$$\begin{aligned} E[x^r] &= \int_0^\infty x^r f(x) dx \\ &= \int_0^\infty x^r \alpha \beta x^{\beta-1} e^{-\alpha x^\beta} dx \end{aligned}$$

$$\begin{aligned} \text{Take } u = \alpha x^\beta &\rightarrow x^\beta = \frac{u}{\alpha} \quad \Rightarrow \quad x = \left(\frac{u}{\alpha}\right)^{1/\beta} \\ du &= \alpha \beta x^{\beta-1} dx \end{aligned}$$

$$= \int_0^\infty e^{-u} \left(\frac{u}{\alpha}\right)^{r/\beta} du$$

$$= \frac{1}{\alpha^{r/\beta}} \int_0^\infty e^{-u} u^{r/\beta} du$$

WKT

$$n-1 = \frac{r}{\beta}$$

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

$$= \frac{1}{\alpha^{r/\beta}} \Gamma\left(\frac{r}{\beta} + 1\right)$$

$$\boxed{\gamma=1} \Rightarrow E(x) = \frac{1}{\alpha^{1/\beta}} \Gamma\left(\frac{1}{\beta} + 1\right)$$

$$\boxed{\gamma=2} \Rightarrow E(x^2) = \frac{1}{\alpha^{2/\beta}} \Gamma\left(\frac{2}{\beta} + 1\right)$$



Failure rate | Hazard rate  $\lambda(t)$

$$\lambda(t) = \alpha t^{\beta-1}$$

where  $\alpha$  is constant

$\beta < 1 \rightarrow \lambda(t)$  is decreasing  
the equipment strengthens  
over time.

$\beta = 1 \rightarrow \lambda(t) = \alpha$ , constant  
Becomes exponential  
distribution in which  
the memoryless property  
prevails

$\beta > 1 \rightarrow \lambda(t)$  is increasing,  
components wear out  
over time.

Weibull

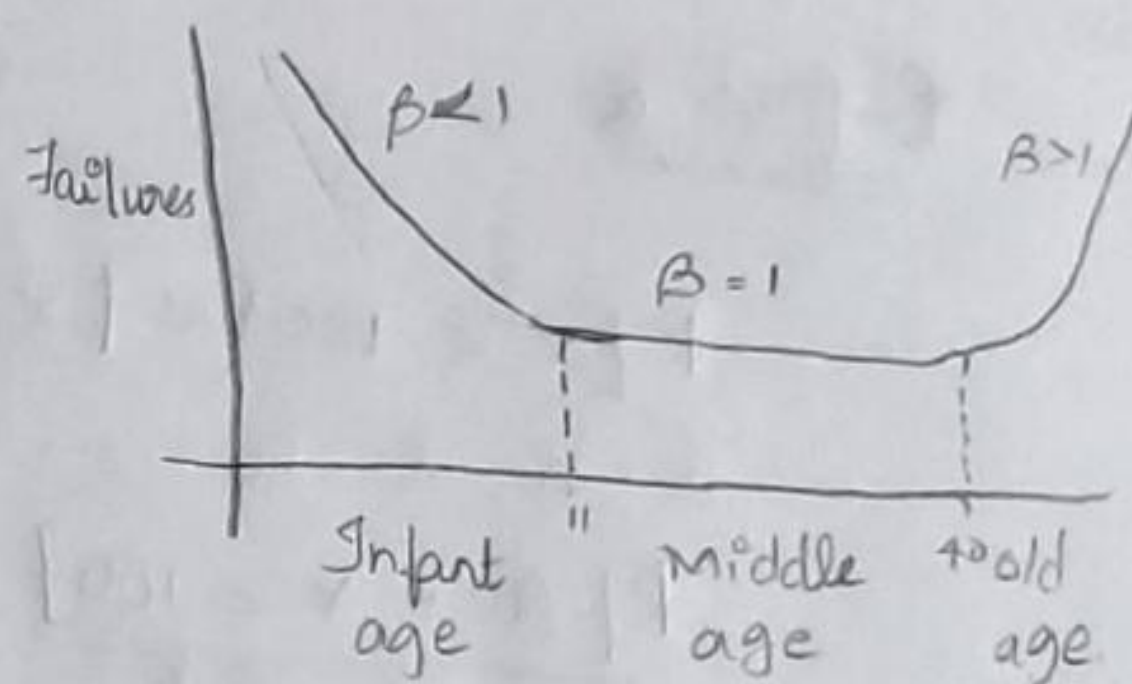
$x \Rightarrow$  time to  
failure

$$f(x) = \alpha \beta x^{\beta-1} e^{-\alpha x^\beta}$$

$\alpha \Rightarrow$  scale

$\beta \Rightarrow$  shape

$$E(x^r) = \frac{1}{\alpha^{r/\beta}} \Gamma\left(\frac{r}{\beta} + 1\right)$$



Suppose the life time (in hrs) of a certain  
Weibull with  $\alpha = 0.1$  &  $\beta = 0.5$

$X$  : life time (in hrs)

$X \sim \text{Weibull}(\alpha = 0.1, \beta = 0.5)$

$$(i) P[X > 300] = e^{-(0.1)(300)^{0.5}} \Rightarrow P[X > t] = e^{-\alpha t^\beta}$$

$$= e^{-1.732} = 0.176$$

$$(ii) \text{Mean life time } E(X) = \frac{1}{(0.1)^{1/0.5}} \Gamma\left(\frac{1}{0.5} + 1\right)$$

$$E(X^r) = \frac{1}{\alpha^{r/\beta}} \Gamma\left(\frac{r}{\beta} + 1\right)$$

$$= 100 \times \Gamma(2+1)$$

$$= 100 \times \Gamma(2) = 100 \times 2!$$

$$= 200!$$



Suppose the life time (in hrs) of a component weibull with  $\beta = 2$  from past experience it is known that 15% of the components that have lasted 90 hrs. Find before 100 hrs. Determine the value of  $\alpha$ .

$x$  : lifetime (in hrs)

$x \sim \text{weibull} (\alpha = ?, \beta = 2)$

To find  $\alpha$

$$P[x < 100 \text{ hrs} \mid x > 90 \text{ hrs}] = 15\% = 0.15$$

$$\frac{P[90 < x < 100]}{P[x > 90]} = 0.15$$

$$\frac{F(100) - F(90)}{P[x > 90]} = 0.15$$

$$\frac{(1 - e^{-\alpha(100)^2}) - (1 - e^{-\alpha(90)^2})}{e^{-\alpha(90)^2}} = 0.15$$

$$\frac{e^{-8100\alpha} - e^{-10000\alpha}}{e^{-8100\alpha}} = 0.15$$

$$1 - e^{-1900\alpha} = 0.15$$

$$e^{-1900\alpha} = 0.85$$

Taking  $\ln$  on both sides,

$$-1900\alpha = \ln(0.85)$$

$$\alpha = \frac{\ln(0.85)}{-1900}$$

$$= 0.00009$$

$$P(x > t) = e^{-\alpha t^\beta}$$

$$F(t) = 1 - e^{-\alpha t^\beta}$$

$\Downarrow$

$$P[x \leq t]$$



The life time of a certain components from weibull with  $\beta = 2$ . Find the value of  $\alpha$  given that the prob that the component, life exceeds 5 yrs is  $e^{-0.25}$ . Find the mean & variance.

$X$  : life time (in yrs)

$X \sim \text{weibull}(\alpha = ?, \beta = 2)$

To find  $\alpha$

given  $P[X > 5] = e^{-0.25}$

$$P[X > t] = e^{-\alpha t^\beta}$$

$$e^{-\alpha(5)^2} = e^{-0.25}$$

$$25\alpha = 0.25$$

$$\boxed{\alpha = 0.01}$$

$$E(X) = \frac{1}{\alpha^{1/\beta}} \Gamma\left(\frac{\beta}{\beta} + 1\right)$$

$$= \frac{1}{(0.01)^{1/2}} \Gamma\left(\frac{1}{2} + 1\right)$$

$$= \frac{1}{0.1} \Gamma\left(\frac{1}{2} + 1\right)$$

$$\Gamma(n+1) = n \Gamma(n)$$

$$= 10 \cdot \frac{1}{2} \Gamma\left(\frac{1}{2}\right)$$

$$\Gamma\left(\frac{1}{2}\right) = \sqrt{\pi}$$

$$= 5\sqrt{\pi}$$

$$E(X^2) = \frac{1}{(0.01)^{2/2}} \Gamma\left[\frac{2}{2} + 1\right]$$

$$= 100 \Gamma(2)$$

$$= 100 \times 1! = 100$$

$$\text{Variance}(X) = E(X^2) - E(X)^2$$

$$= 100 - 25\pi$$



## Normal / Gaussian Distribution. (Z)

pdf

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

parameters

$\mu$

&

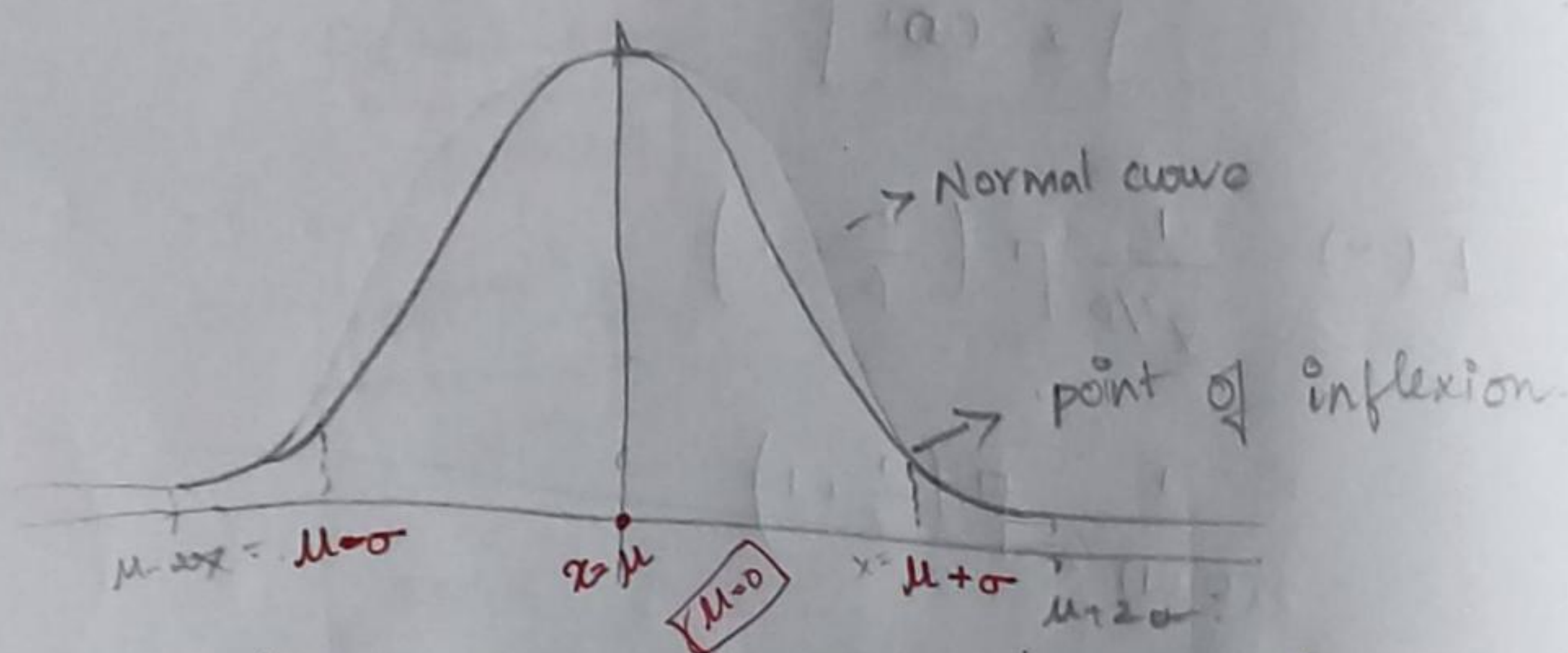
$\sigma$

mean

std deviation

Special case : Standard normal distribution

It is a normal distribution with  $\mu=0$  &  $\sigma=1$ .



Normal curve : → graph of the pdf of normal distribution.

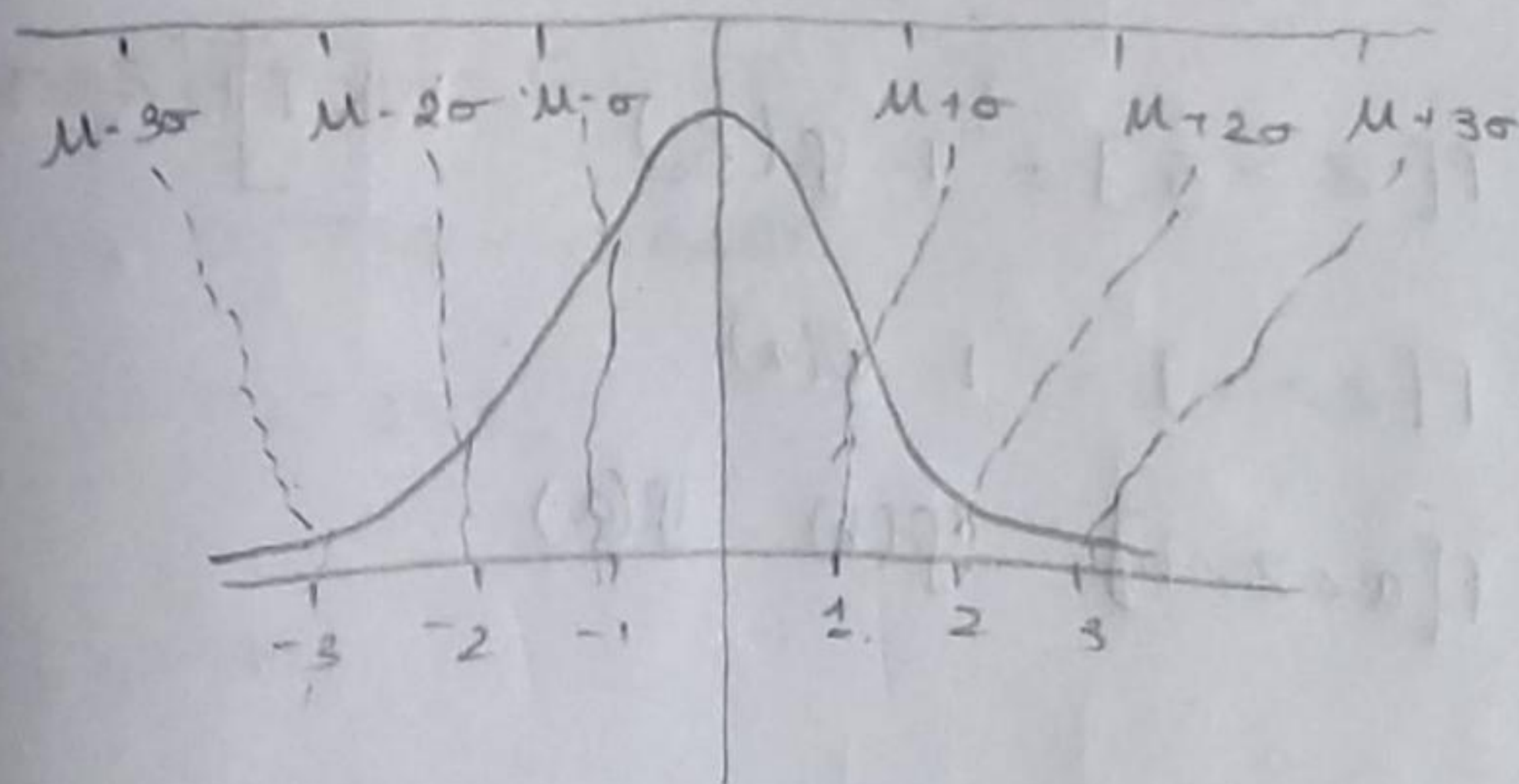
Properties of normal

- Symmetric about the mean  $x = \mu$
- Mean = Median = Mode
- $x = \mu - \sigma$ ,  $x = \mu + \sigma$  are points of inflexion
- x axis is an asymptote
- $\int_a^b f(x) dx = \text{Area under normal curve}$   
 $x=a, x=b$



⇒ Normal table contain probability for standard normal distribution (SND) ·  $\mu=0, \sigma=1$ .

⇒ An arbitrary normal distribution with any mean  $\mu$  and SD  $\sigma$  can be converted into a SND by the transformation  $z = \frac{x - \mu}{\sigma}$  where  $z$  is the SN variable.



$$\mu - 3\sigma = -3$$

$$\mu - 2\sigma = -2$$

$$\mu - \sigma = -1$$

$$\mu = 0$$

$$\mu + 3\sigma = 3$$

$$\mu + 2\sigma = 2$$

$$\mu + \sigma = 1$$

∴ Any ND → SND

Empirical rules:

$$P[\mu - \sigma < x < \mu + \sigma] = 68\%$$

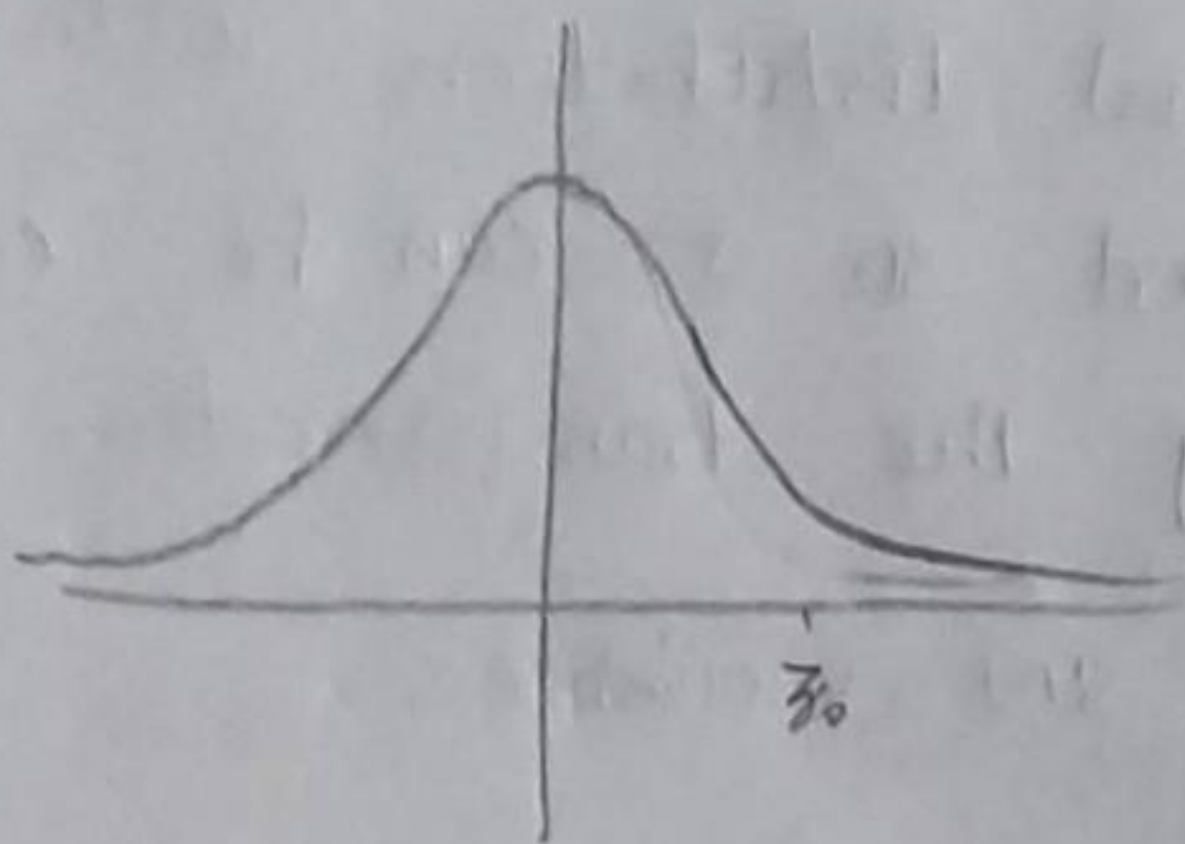
$$P[\mu - 2\sigma < x < \mu + 2\sigma] = 95\%$$

$$P[\mu - 3\sigma < x < \mu + 3\sigma] = 99.7\%$$

68 95 99.7 rule



From normal table one can find  $P[Z \leq z_0] = \Phi(z_0)$  for any  $z_0 \in \mathbb{R}$



$$P[Z > z_0] = 1 - \Phi(z_0)$$

$$\text{WKT } P[Z \leq z_0] = \Phi(z_0)$$

$$P[Z > a] = 1 - \Phi(a)$$

$$P[a < Z < b] = \Phi(b) - \Phi(a)$$

A survey indicates that for each trip to the super market, a customer spends an average of 45 mins with a SD of 12 mins. The length of time spent follows ND. What is the prob a randomly selected customer will be in the store.

- (i) b/w 24 and 54 min
- (ii) more than 39 min
- (iii) b/w 33 and 60 min

$X$  = Time spent in the store (in mins)

$$X \sim N(\mu = 45, \sigma = 12)$$

$$Z = \frac{X - \mu}{\sigma}$$



$$(i) P[24 < x < 54]$$

$$P[Z < z_0] = \Phi(z_0)$$

$$= P\left[\left(\frac{24-45}{12}\right) < Z < \left(\frac{54-45}{12}\right)\right]$$

$$= P[-1.75 < Z < 0.75]$$

$$= \Phi(0.75) - \Phi(-1.75)$$

$$= 0.7734 - 0.0401$$

$$= 0.7333$$

$$(ii) P[x > 39]$$

$$= P\left[Z > \frac{39-45}{12}\right]$$

$$= P[Z > -0.5]$$

$$= 1 - \Phi(-0.5) = 1 - 0.3085 = 0.6915$$

$$(iii) P[33 < x < 60]$$

$$= P\left[\left(\frac{33-45}{12}\right) < Z < \left(\frac{60-45}{12}\right)\right]$$

$$= P[-1 < Z < 1.25]$$

$$= \Phi(1.25) - \Phi(-1)$$

$$= 0.8944 - 0.1587$$

$$= 0.7357$$



The scores on a test given to 11k students are normally distributed with mean 500 & SD 100. We should be the score of student be to place them among the top 10%?

$x$  = Scores

$$x \sim ND(\mu = 500, \sigma = 100)$$

To find  $x$  such that

$$P[X \geq x] = 10\%$$

$$P\left[Z \geq \frac{x-500}{100}\right] = 0.1$$

$$P\left[Z < \frac{x-500}{100}\right] = 0.9$$

$$(1 - \alpha)$$

$$P(Z \leq z) = \Phi(z)$$

$$\Phi\left(\frac{x-500}{100}\right) = 0.9$$

(Finding the reverse in the table)

From Normal table for prob = 0.9 (0.887), the  $Z$  score is 1.28

$$\begin{array}{l} 1.28 \Rightarrow 0.8997 \\ \text{Approx to } 0.9 \end{array}$$

$$\frac{x-500}{100} = 1.28$$

$$x = 628$$



In a ND 31% of the items are under 45 & 8% of the items are over 64. Find the mean & variance

$$x \sim N(\mu, \sigma)$$

given  $P[x < 45] = 31\% = 0.31$

$$P\left[Z < \frac{45 - \mu}{\sigma}\right] = 0.31$$

$$\Phi\left(\frac{45 - \mu}{\sigma}\right) = 0.31$$

$$0.3085 = -0.5$$

// from table

$$\frac{45 - \mu}{\sigma} = -0.5$$

$$\mu - 0.5\sigma = 45 \rightarrow \textcircled{1}$$

Also,

$$P[x > 64] = 8\% = 0.08$$

$$P\left[Z > \frac{64 - \mu}{\sigma}\right] = 0.08$$

$$P\left[Z < \frac{64 - \mu}{\sigma}\right] = 0.92$$

$$\Phi\left(\frac{64 - \mu}{\sigma}\right) = 0.92$$

// From table for prob 0.92

Z score = 1.41

$$\frac{64 - \mu}{\sigma} = 1.41$$

$$64 - \mu = 1.41\sigma$$

$$\mu + 1.41\sigma = 64 \rightarrow \textcircled{2}$$

Solving ① & ②

Approx values

$$\sigma = 10$$

$$\mu = 50$$

$$\mu - 0.5\sigma = 45$$

$$- \mu - 1.41\sigma = -64$$

$$+ 1.91\sigma = 19$$

$$\sigma = 9.948$$

$$\mu - 0.5(9.948) = 45$$

$$\mu = 49.974$$



## Moments

Let  $x$  be a r.v with mean  $E(x) = \mu$ .

$E(x^r)$  is called  $r^{\text{th}}$  moment of  $x$  (moment about origin)

$E(x - \mu)^r$  is called  $r^{\text{th}}$  central moment of  $x$  (moment about mean)

Any central moment can be expressed as

$n^{\text{th}}$  moment of origin

$$E(x - \mu)^2 = E(x^2) - E(x)^2$$

$1^{\text{st}}$  moment of origin to the  $n$  power.

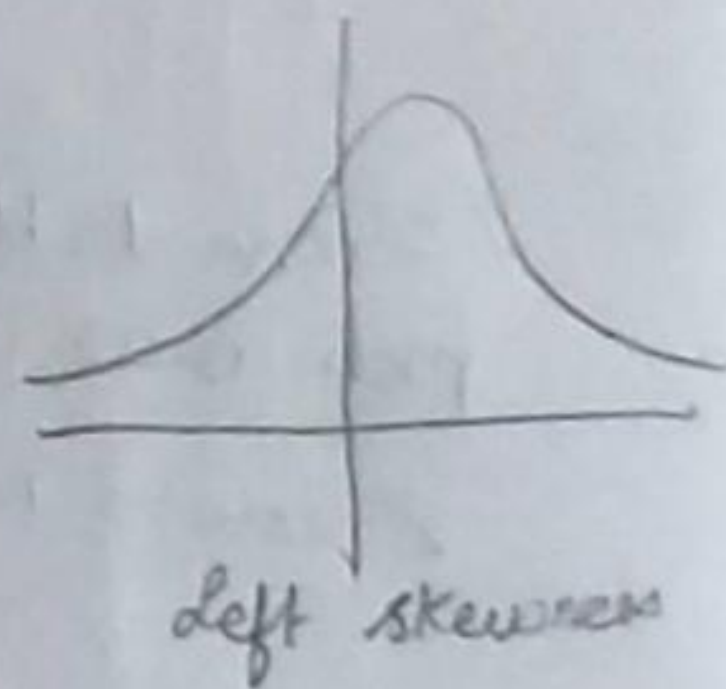
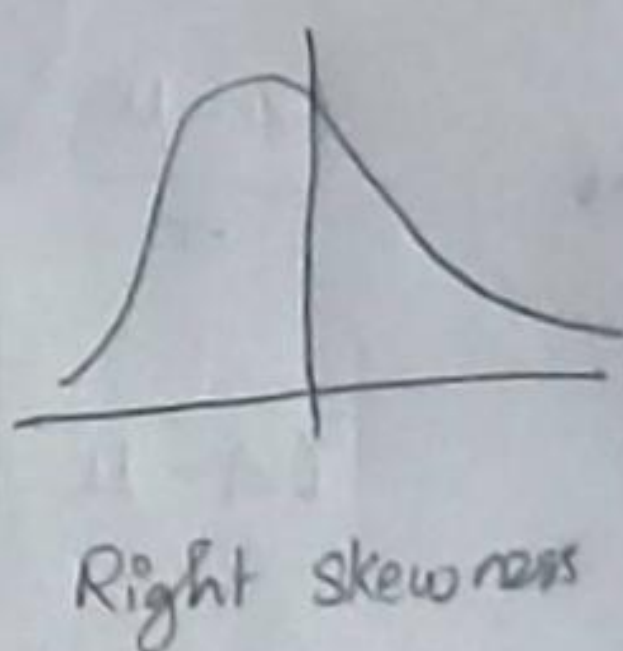
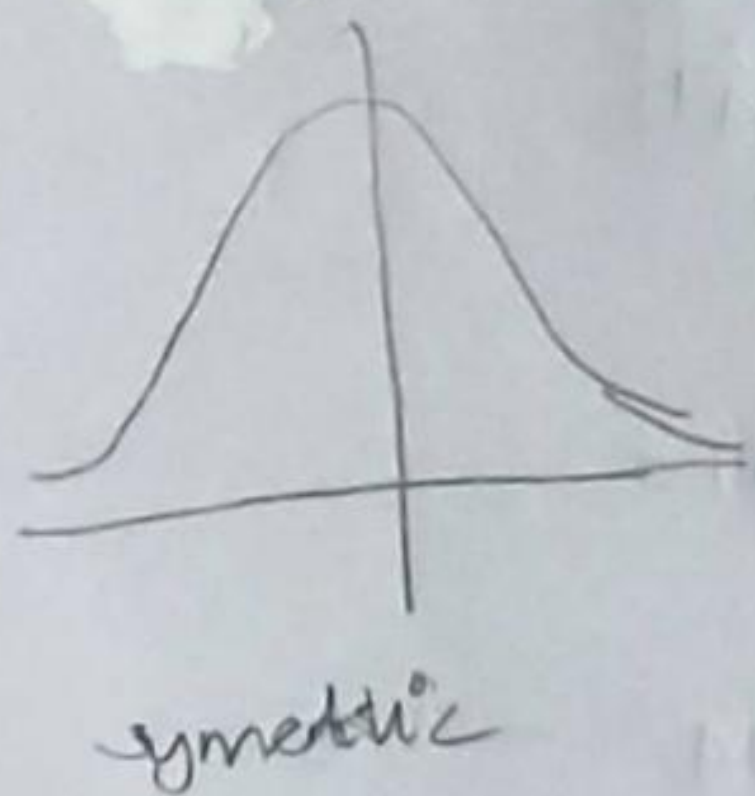
$r=1$   $E(x)$  - the  $1^{\text{st}}$  moment is the mean.

$r=2$   $E(x - \mu)^2$  - the  $2^{\text{nd}}$  central moment is the variance.

For other ~~variance~~ values of  $r$  also,  $E(x^r)$  is a valuable measure

$E(x) = \mu$	$E(x^r) = \mu_r'$
$E(x - \mu)^r = \mu_r$	

The quantity  $\frac{\mu_3}{\sigma^3}$  is a measure of symmetry (skew)





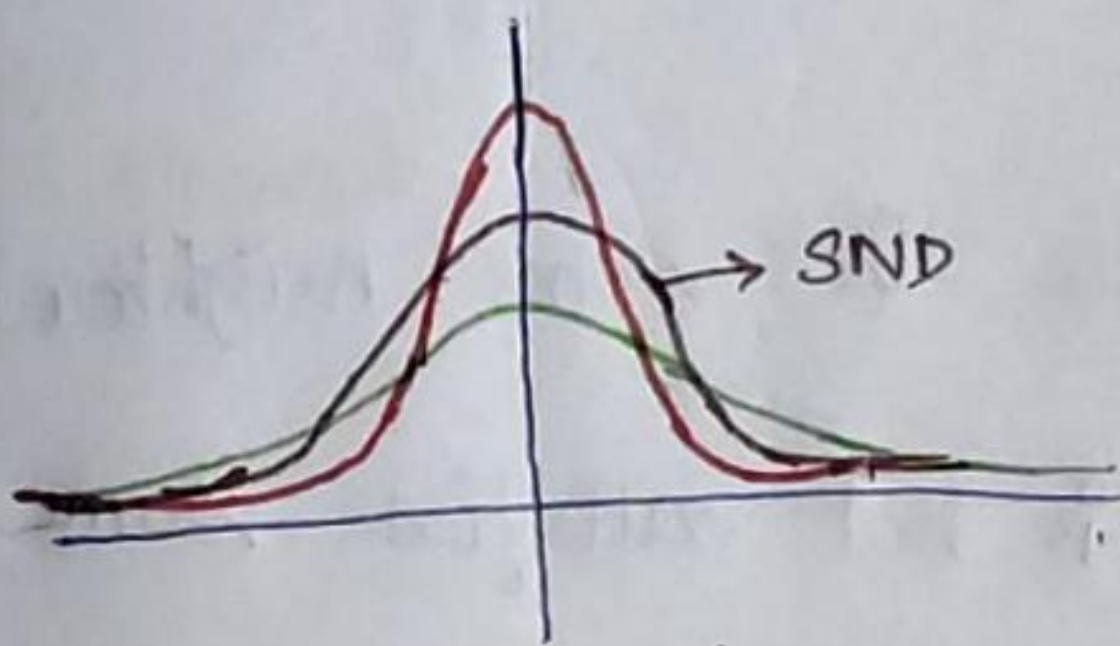
It is called skewness

$= 0$  , then the distribution is symmetric

$> 0$  , then the distribution is skewed on the right

$< 0$  , then the distribution is skewed on the left

the quantity  $\frac{\mu_4}{\sigma^4}$  is a measure of **Kurtosis**  
(Peakness)



$= 3$  for SND

$> 3$  for a distribution which is more peaked than SND

$< 3$  for a distribution which is less peaked than SND

Moment means importance  
With the knowledge of moments we know



## Moment generating function (MGF)

For a r.v.  $X$ ,

$$M_X(t) = E(e^{tx}) \quad e^{tx} \text{ is another r.v.}$$

If  $M_X(t)$  is defined for all values of  $t$  in an interval  $(-\delta, \delta)$  for  $\delta > 0$ , then  $M_X(t)$  is called the **moment generating function** of  $X$ .

Note:

$M_X(t)$  is finite in some neighbourhood of 0  $(-\delta, \delta)$

If this condition is not satisfied some moments may not exist.

Generating moments from mgf

method 1

$$\frac{d^r}{dt^r} [M_X(t)]_{t=0} = E(X^r)$$

the  $r^{\text{th}}$  derivative of mgf about  $t=0$  is the  **$r^{\text{th}}$  moment**

Proof:

$$\begin{aligned} M_X'(t) &= \frac{d}{dt} \{E(e^{tx})\} \\ &= E \left\{ \frac{d}{dt} (e^{tx}) \right\} \\ &= E [x e^{tx}] \end{aligned}$$

$$M_X(t) \Big|_{t=0} = E(X)$$

1<sup>st</sup> derivative of mgf @  $t=0$  is  $E(X)$



Method 2.

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

$$= E\left(1 + \frac{tx}{1!} + \frac{t^2 x^2}{2!} + \frac{t^3 x^3}{3!} + \dots + \frac{t^r x^r}{r!} + \dots\right)$$

$$= 1 + \frac{t}{1!} E(x) + \frac{t^2}{2!} E(x^2) + \frac{t^3}{3!} E(x^3) + \dots + \frac{t^r}{r!} (Ex^r)$$

the coefficient of  $\frac{t^r}{r!}$  in the expansion of mgf is  
the  $r^{\text{th}}$  moment.