

Moment Generating Function (m.g.f.)

$$E(e^{tx})$$

$$M_X(t) = \int e^{tx} f(x) dx$$

Q. Find the m.g.f. of the random variable  $X$  having the probability density function

$$f(x) = \begin{cases} x & , 0 \leq x < 1 \\ 2-x & , 1 \leq x < 2 \\ 0 & , \text{otherwise} \end{cases}$$

Find mean and variance of  $X$  using m.g.f.

Soln:

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

$$= \int_0^1 e^{tx} x dx + \int_1^2 e^{tx} (2-x) dx$$

$$= \left[ x \int e^{tx} dx \right]_0^1 - \int_0^1 \left( \frac{d}{dx} x \int e^{tx} dx \right) dx +$$

$$\left[ (2-x) \int e^{tx} dx \right]_1^2 - \int_1^2 \left( \frac{d}{dx} (2-x) \int e^{tx} dx \right) dx$$

$$= \left[ x \left( \frac{e^{tx}}{t} \right) \right]_0^1 - \int_0^1 \frac{e^{tx}}{t} dx + \left[ (2-x) \cdot \frac{e^{tx}}{t} \right]_1^2 + \int_1^2 \frac{e^{tx}}{t} dx$$

$$M_x(t) = \left( \frac{1 \cdot e^t}{t} - 0 \right) - \left[ \frac{e^{tn}}{t^2} \right]_0^1 + \left[ 0 - \frac{e^t}{t} \right] + \left[ \frac{e^{tn}}{t^2} \right]_1^2$$

$$= \frac{e^t}{t} - \left( \frac{e^t}{t^2} - \frac{1}{t^2} \right) - \frac{e^t}{t} + \left( \frac{e^{2t}}{t^2} - \frac{e^t}{t^2} \right)$$

$$= -\frac{e^t}{t^2} + \frac{1}{t^2} + \frac{e^{2t}}{t^2} - \frac{e^t}{t^2}$$

$$= \frac{e^{2t}}{t^2} - \frac{2e^t}{t^2} + \frac{1}{t^2} = \frac{1}{t^2} [e^{2t} - 2e^t + 1] \rightarrow \textcircled{1}$$

$$= \frac{(e^t)^2 - 2e^t + 1}{t^2}$$

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



Expanding  $M_X(t)$  using ①

$$M_X(t) = \frac{1}{t^2} \left[ \left( 1 + 2t + \frac{(2t)^2}{2!} + \frac{(2t)^3}{3!} + \frac{(2t)^4}{4!} + \dots \right) - \right. \\ \left. 2 \left( 1 + t + \frac{t^2}{2!} + \frac{t^3}{3!} + \frac{t^4}{4!} + \dots \right) + 1 \right]$$

$$= \frac{1}{t^2} \left[ t^2 + t^3 + \frac{7}{12} t^4 + \dots \right]$$

$$M_X(t) = 1 + t + \frac{7}{12} t^2 + \dots$$

Mean  $= \mu'_1 =$  coefficient of  $t$  in  $M_X(t) = 1$

$$\mu'_2 = \text{coefficient of } \frac{t^2}{2!} \text{ in } M_X(t) = 2! \frac{7}{12} = \frac{7}{6}$$

$$\text{variance } (\mu_2) = \mu'_2 - \mu_1'^2$$

$$= \frac{7}{6} - (1)^2$$

$$= \frac{1}{6}$$

# NORMAL DISTRIBUTION

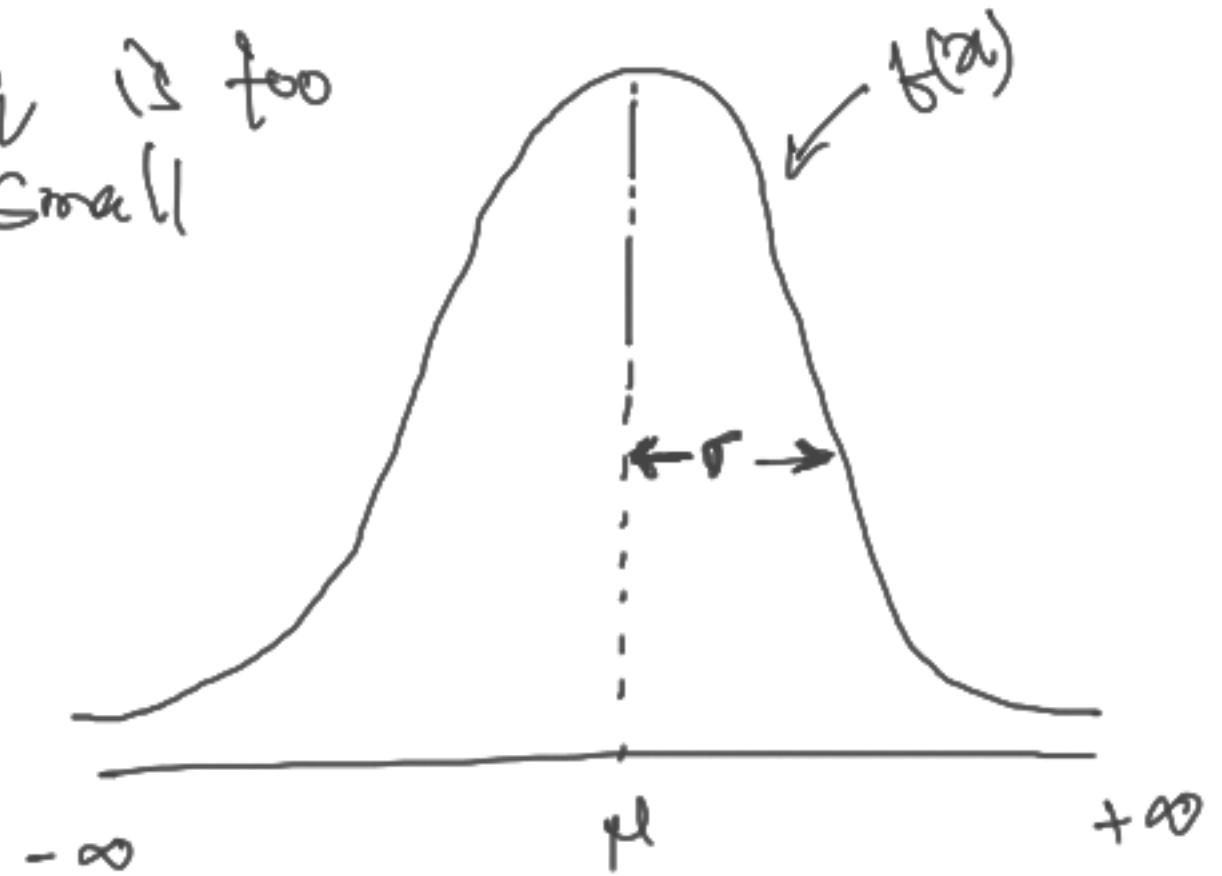
Limiting case of Binomial Distribution

$p \rightarrow q$   $\leftarrow$  neither  $p$  nor  $q$  is too small

$n \rightarrow \infty$   $\leftarrow$  infinitely large

Discovered by De Moivre in 1733

It is continuous R.V. distr.



$X \leftarrow$  continuous R.V.

Then  $X$  is said to have normal distribution if its p.d.f. is defined as

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

$\sigma > 0$

Here  $\mu, \sigma \leftarrow$  parameters of Normal Distribution

## Binomial

1. Discrete R.V.

2. 2 parameters  
( $p$  or  $q$  and  $n$ )

3. —

$$4. P(x) = {}^n C_x p^x q^{n-x}$$

## Poisson

1. Discrete R.V.

One parameter  
( $\lambda$ )

Limiting case of  
Binomial dist.

$$p \rightarrow 0, n \rightarrow \infty$$

$$P(x) = \frac{e^{-\lambda} \lambda^x}{x!}$$

## Normal.

Continuous R.V.

Two parameters ( $\mu, \sigma$ )

Limiting case binomial  
dist.

$$p \rightarrow 0.5, n \rightarrow \infty$$

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

pdf



$$\begin{array}{l} \xleftarrow{B.D} \\ \Sigma \quad \text{mean} = np \\ \quad \text{var} = npq \end{array}$$

$$\begin{array}{l} \xleftarrow{P.D.} \\ \text{mean} = \lambda \\ \text{var} = \lambda \end{array}$$

$$\begin{array}{l} \xleftarrow{N.D} \\ \text{mean} = \mu \\ \text{var} = \sigma^2 \end{array} \quad \left. \begin{array}{l} \nearrow \\ \nearrow \end{array} \right\} \begin{array}{l} \text{values det.} \\ \text{from formula} \\ \text{given at (i)} \end{array}$$

Q. Prove that the mean and variance of the normal distribution

$$\begin{array}{l} \text{Mean} = \mu \\ \text{Variance} = \sigma^2 \end{array}$$

Ex. For normal distribution

$$\text{Mean} = \mu$$

$$\text{Median} = \mu$$

$$\text{Mode} = \mu$$

\* In case of normal distribution,  $\text{mean} = \text{median} = \text{mode}$ .

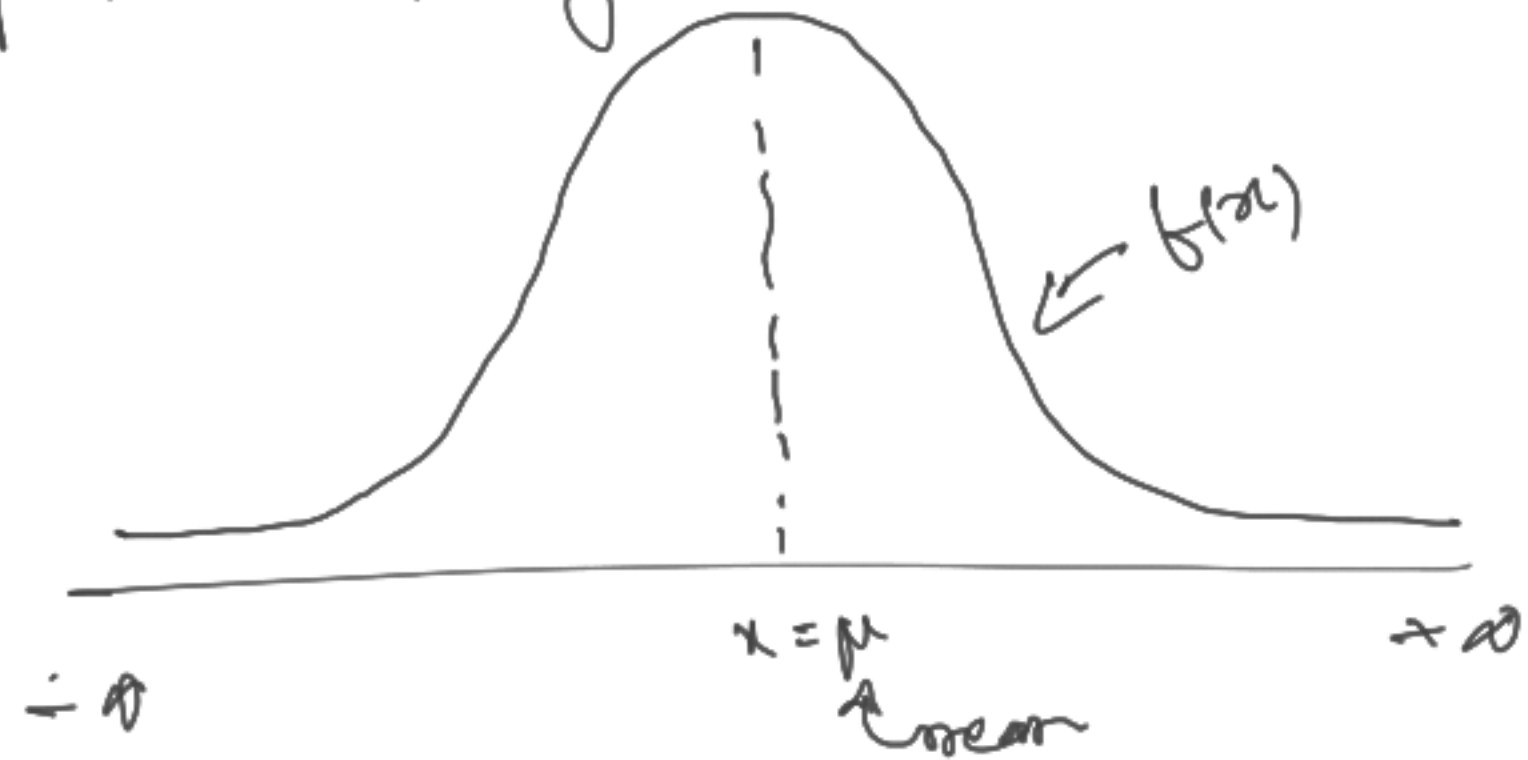
### Properties of Normal Distribution

① The normal probability curve with mean  $\mu$  and standard deviation  $\sigma$  is given by

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

2. The curve is bell-shaped and symmetrical about the line  $x = \mu$

3. Mean, median and mode of the normal distribution coincides i.e. unimodal.



4.  $f(x)$  decreases rapidly as  $x$  increases.

5.  $x$ -axis is an asymptote to the curve.

6. Maximum probability occurs at the point  $x = \mu$

and  $\text{max}^m \text{ prob} = \frac{1}{\sigma \sqrt{2\pi}}$

7.  $M.D. (\bar{x}) = \frac{4}{5} \sigma$  or  $M.D. (\mu) = \frac{4}{5} \sigma$

8. The point of inflexion of curve are at  $x = \mu \pm \sigma$

(i) Area of the normal curve between  $(\mu - \sigma)$  and  $(\mu + \sigma)$  is

0.6826 i.e.  $P(\mu - \sigma < X < \mu + \sigma) = 0.6826$

(i)  $P(\mu - 2\sigma < X < \mu + 2\sigma) = 0.9544$

(ii)  $P(\mu - 3\sigma < X < \mu + 3\sigma) = 0.9973$

Area betn  $\mu - 2\sigma$  and  $\mu + 2\sigma$

