

①

## Probability Theory

There are generally two types of experiment in our common life

- (i) Deterministic experiment
- (ii) Random experiment

(i) Deterministic experiment → The outcomes of such experiments are always certain, if the experiment is repeated again and again, the result also remains approximately the same

For example → If we repeat the experiment of gas law  $PV=nRT$  with different value of P, V, & T, we will always obtain the same value of R.

(ii) Random experiment → If such type of experiment is repeated again & again, the result always may not be the same that is there is no certainty about their result.

For example → If a coin is tossed again & again, it is not certain, when the head or tail will come.

Concept of probability → When the chance of happening an event is expressed numerically, it is called probability.

Def → According to Laplace "Probability is the ratio of favourable cases to the total number of likely cases."

① Trial → The performance of a random experiment for one event. time is called a trial, the outcome of trial is called

For example → Tossing of a coin is a trial. Head (H) or tail (T) falls by tossing a coin. H & T are two events.

② Sample points → The outcomes of a random experiment are known as sample points.

③ Sample space → The set of all possible outcomes or sample points of random experiment is known as sample space.

We denote a sample space by S. Each element of the sample space is a sample point, if H & T come in tossing of a coin.

Then  $S = \{H, T\}$ , where H & T are its two sample point.

(2)

④ Discrete sample space → A sample space, whose sample points (whose number is finite or can be arranged into a sample sequence according to positive integers) is known as discrete sample space.

⑤ Continuous sample space → A sample space whose sample points are uncountably infinite is known as continuous sample space.

⑥ Event → The subset of a sample space is known as event. The subset in which sample points are simple is known as simple or elementary event.

⑦ Equally likely events → Two or more events are said to be equally likely, if the chance of their happening are same.

For example, → 1, 2, 3, 4, 5 & 6 are equally likely events in throwing of a dice.

⑧ Independent events → Two or more events are said to be independent events if the happening of one does not affect the happening of others.

For example, → If two dice are thrown simultaneously, the number coming on first will not affect the number coming on second.

⑨ Dependent events → The events, which is not independent is called dependent events.

⑩ Mutually exclusive events → Two or more events are said to be mutually exclusive events if the happening of one prevents the happening of others.

In other words, mutually exclusive events can not happen simultaneously.

For example

(a) Tossing of a coin H & T are two mutually exclusive events.

(b) Throwing a dice 1, 2, 3, 4, 5 & 6 are mutually exclusive events.

(c) If one card is drawn from a deck of cards then there will be total 52 mutually exclusive events.

③

⑪ Exhaustive events: → All possible outcomes (events) of a random experiment are known as exhaustive events.

For example: → If a bag contains 10 balls & one ball is drawn at random, then there will be 10 exhaustive events.

"In other words two or more events are said to be exhaustive events if their union is equal to sample space."

⑫ Favourable cases: → The outcomes from all equally likely outcomes of a random experiment which are favourable for the happening of a particular event are called favourable cases.

For example: → Throwing a dice, there will be three favourable cases (2, 4, 6) for coming an even number.

Probability of the happening of the event

$$p = \frac{\text{Number of favourable cases}}{\text{Total number of equally likely cases}}$$

Similarly

Probability of not happening of the event

$$q = \frac{\text{No. of cases not favourable}}{\text{Total no. of equally likely cases}}$$

$$p = \frac{m}{m+n} \quad \& \quad q = \frac{n}{m+n}$$

$$p+q = \frac{m+n}{m+n} \Rightarrow \boxed{p+q=1}$$

Example: → A single letter is selected at random from the word "PROBABILITY" what is the probability that it is vowel.

Sol<sup>n</sup>: → Total letters  $n=11$

No. of vowels  $m=4$

$$\text{Required probability} = \frac{m}{n} = \frac{4}{11}$$

④ Example → what is the chance that a non leap year should have 53 Mondays.

Sol<sup>n</sup> → There are 365 days in a non-leap year that is 52 complete weeks & one day. Any day from 8 even days of week (Monday, Tuesday, Wednesday, Thursday, Friday, Saturday) may come 53<sup>rd</sup> times.

So, equally likely cases = 7

Favourable ways for 53 Mondays = 1

Required probability =  $\frac{\text{Favourable cases}}{\text{Total cases}} = \frac{1}{7}$

Example → Find the probability of 53 Saturdays in a leap year.

Sol<sup>n</sup> → There are 366 days in a leap year

No. of days in 52 weeks =  $52 \times 7 = 364$  days

There may be following pairs for remaining two days

- 1) Sunday & Monday
- 2) Monday & Tuesday
- 3) Tuesday & Wednesday
- 4) Wednesday & Thursday
- 5) Thursday & Friday
- 6) Friday & Saturday
- 7) Saturday & Sunday

Total permutations of two days = 7

Permutation of Saturday = 2

∴ Probability of 53 Saturdays =  $\frac{2}{7}$  Ans.

Example → Two dice are drawn

- (i) What is the prob. of 5 on both dice  
(ii) What is prob. that the sum of numbers on the two dice is 9.

Sol<sup>n</sup> → Total no. of cases =  $6 \times 6 = 36$

(i) Favourable no. of cases for 5 on both dice = 1

$$P(5) \text{ on both dice} = \frac{1}{36}$$

- (ii) Total no. on both will be 9 if the points are (6, 3), (5, 4), (4, 5) & (3, 6).

So, favourable cases = 4.

$$\text{Prob. of getting 9} = \frac{4}{36} = \frac{1}{9}$$

⑤

Law of probability → There are two main laws of probability

(i) Addition law of probability

(ii) Multiplication law of probability

(i) Addition law of probability: → If events are not mutually exclusive that is there are some common elements.

Then the formula for addition law will be

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

(ii) Multiplication law of probability: →

If A, B & C are three independent events, then

$$P(A \cap B \cap C) = P(A) \cdot P(B) \cdot P(C)$$

Example: → Find the probability that a leap year selected at random will contain 53 Monday or 53 Tuesday?

Sol: → There are 366 days in a leap year, that is 364 days of 52 weeks & 2 days.

Hence all days occurs 52 times, Any one day may occur in remaining two days.

(i) Monday & Tuesday (ii) Tuesday & Wednesday

(iii) Wednesday & Thursday (iv) Thursday & Friday

(v) Friday & Saturday (vi) Saturday & Sunday

(vii) Sunday & Monday

Prob. of occurring 53 Mondays  $P(A) = \frac{2}{7}$

Prob. of " " Tuesdays  $P(B) = \frac{2}{7}$

Prob. of occurring 53 Mondays & 53 Tuesdays

$$\text{i.e } P(A \cap B) = \frac{1}{7}$$

Req. Prob. i.e  $P(A \cup B) = P(A) + P(B) - P(A \cap B)$

$$\Rightarrow P(A \cup B) = \frac{2}{7} + \frac{2}{7} - \frac{1}{7} = \frac{4-1}{7} = \frac{3}{7} \text{ qns.}$$

⑥ Example → The probability that a question can be solved by A is  $\frac{1}{3}$  & B is  $\frac{1}{4}$ . What is the probability that the question will be solved by any one of them.

Sol →  $P(A) = \frac{1}{3}, P(B) = \frac{1}{4}$

$$P(A \cap B) = P(A) \cdot P(B) \quad \left\{ \begin{array}{l} \text{Both events are} \\ \text{independent} \end{array} \right\}$$

$$= \frac{1}{3} \cdot \frac{1}{4} = \frac{1}{12}$$

Therefore, the prob. that the question will be solved by any one of them.

$$P(A \cup B) = P(A) + P(B) - P(A \cap B) = \frac{1}{3} + \frac{1}{4} - \frac{1}{12}$$

$$= \frac{4+3-1}{12} = \frac{6}{12} = \frac{1}{2} \text{ ans.}$$

Conditional probability → when the happening of an event A depends on the happening of event B. The prob. of event A is called conditional prob. & it is denoted by  $P(A|B)$ .

thus 
$$\boxed{P(A|B) = \frac{P(A \cap B)}{P(B)}} \Rightarrow \boxed{P(A \cap B) = P(A|B) \cdot P(B)}$$

Note → (i) If  $P(A|B) = P(A)$ , then A & B will be independent

(ii) If  $P(B|A) = P(B)$ , then B & A will be independent.

Example → A dice is thrown once, what is the conditional prob. that multiple of three is obtained when it is known that it is an even number.

Sol → let A be the event that a multiple of 3 is obtained.  $A = \{3, 6\} \therefore n(A) = 2$

& let B be the event that an even number is obtained

$$B = \{2, 4, 6\} \therefore n(B) = 3, A \cap B = \{6\} \therefore n(A \cap B) = 1$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{n(A \cap B)/n(S)}{n(B)/n(S)} = \frac{\frac{1}{6}}{\frac{3}{6}} = \frac{1}{3}$$

$$= \frac{1}{6} \times \frac{1}{3} = \frac{1}{18} \text{ ans.}$$

(7)

Example → Find the prob. of drawing of two aces one after other without replacement from a well shuffled 52 cards.

Sol → Let A be an event that an ace will be drawn for first time & B be the event that an ace will be drawn for second time

$$P(A) = \frac{4}{52} = \frac{1}{13}$$

Now, there are 3 aces in 51 cards in the deck

$$P(B|A) = P(B) = \frac{3}{51} \quad \left\{ \because A \text{ & } B \text{ are independent events} \right\}$$

$P(\text{An ace will be drawn for first & second time})$   
i.e  $P(A \cap B) = P(A) \cdot P(B|A) = \frac{1}{13} \cdot \frac{1}{17} = \frac{1}{221}$ .

Baye's theorem → Let  $B_1, B_2, \dots, B_n$  be  $n$  mutually exclusive & exhaustive events, whose union be a sample space & assume A be an arbitrary event in space &  $P(A) \neq 0$  &  $P(B_1), P(B_2), \dots, P(B_n)$  &  $P(A|B_1), P(A|B_2), \dots, P(A|B_n)$  are known.

Then

$$P(B_i|A) = \frac{P(B_i) \cdot P(A|B_i)}{\sum_{i=1}^n P(B_i) \cdot P(A|B_i)}$$

Example → There are two urns. In the first urn there are 3 white, & 2 black balls & in the second 1 white & 4 black balls. One of the urn is selected & one ball is drawn. It turns out to be white, what is the probability that it came from the second urn.

Sol → Let A = Event that a white ball is drawn

$B_1$  = Event that first urn is selected

$B_2$  = Event that second urn is selected

$$P(B_1) = P(B_2) = \frac{1}{2}, P(A|B_1) = \frac{3}{5}, P(A|B_2) = \frac{1}{5}$$

⑧ According to Baye's theorem

$$P(B_i^o/A) = \frac{P(B_i^o) \cdot P(A/B_i^o)}{\sum_{i=1}^n P(B_i^o) \cdot P(A/B_i^o)}$$

$$\Rightarrow P(B_2/A) = \frac{P(B_2) \cdot P(A/B_2)}{P(B_1) \cdot P(A/B_1) + P(B_2) \cdot P(A/B_2)}$$

$$= \frac{\frac{1}{2} \cdot \frac{1}{5}}{\frac{1}{2} \cdot \frac{3}{5} + \frac{1}{2} \cdot \frac{1}{5}} = \frac{\frac{1}{10}}{\frac{3}{10} + \frac{1}{10}} = \frac{\frac{1}{10}}{\frac{4}{10}} = \frac{1}{4} \text{ Ans.}$$

Example → A box contains 3 blue & 2 red balls while another box contains 2 blue & 5 red balls. A ball is drawn at random from one of the boxes turns to be blue. What is the prob. that it came from the first box.

Sol → Let  $A$  be the event that a blue ball is drawn & let us assume  $B_1$  &  $B_2$  be the events that first & second box will be selected respectively.

$$P(B_1) = P(B_2) = \frac{1}{2}, P(A/B_1) = \frac{3}{5}, P(A/B_2) = \frac{2}{7}$$

By Baye's theorem, we get

$$P(B_1^o/A) = \frac{P(B_1^o) P(A/B_1^o)}{\sum_{i=1}^n P(B_i^o) \cdot P(A/B_i^o)}$$

$$\Rightarrow P(B_1/A) = \frac{P(B_1) \cdot P(A/B_1)}{P(B_1) \cdot P(A/B_1) + P(B_2) \cdot P(A/B_2)}$$

$$= \frac{\frac{1}{2} \cdot \frac{3}{5}}{\frac{1}{2} \cdot \frac{3}{5} + \frac{1}{2} \cdot \frac{2}{7}} = \frac{\frac{3}{10}}{\frac{3}{10} + \frac{1}{7}} = \frac{\frac{3}{10}}{\frac{21+10}{70}} = \frac{3/10}{31}$$

$$= \frac{3}{14} \times \frac{70}{31} = \frac{21}{31} \text{ Ans.}$$

9

Example,  $\rightarrow$  Four cards are drawn from a pack of cards.  
 Find the probability that (i) all are diamonds,  
 (ii) there is one card of each suit, and (iii) there are two spades  
Sol,  $\rightarrow$  4 cards can be drawn from a pack of 52 cards  
 in  ${}^{52}C_4$  ways.

$$\therefore \text{Exhaustive number of cases} = {}^{52}C_4 = \frac{13}{4} \times \frac{25}{3} \times \frac{21}{2} \times \frac{19}{1} \times \frac{18}{17} \times \frac{17}{16} \times \frac{15}{15}$$

$$= 270725.$$

(i) There are 13 diamonds in the pack and 4 can be drawn out of them in  ${}^{13}C_4$  ways.

$$\therefore \text{Favourable number of cases} = {}^{13}C_4 = \frac{13 \times 12 \times 11 \times 10 \times 9!}{4 \times 3 \times 2 \times 1 \times 9!} = 715.$$

$$\text{Required Probability} = \frac{715}{270725} = \frac{11}{495} = \frac{1}{45} \quad \begin{matrix} \text{(iii) 2 Spades out of 13} \\ \text{can be drawn in } 13C_2 \text{ ways.} \end{matrix}$$

(ii) There are 4 suits, each containing 13 cards.

$$\therefore \text{Favourable number of cases} = {}^{13}C_1 \times {}^{13}C_1 \times {}^{13}C_1 \times {}^{13}C_1 = 13 \times 13 \times 13 \times 13 \quad \begin{matrix} \text{2 Hearts out of 13} \\ \text{can be drawn in } 13C_2 \text{ ways.} \end{matrix}$$

$$\text{Required probability} = \frac{13 \times 13 \times 13 \times 13}{270725} = \frac{2197}{20825} \quad \begin{matrix} \text{Favourable no. of} \\ \text{cases} = {}^{13}C_2 \times {}^{13}C_2 \\ = 70 \times 70 \\ = \frac{70 \times 70}{270725} = \frac{4900}{20825} \end{matrix}$$

Example,  $\rightarrow$  A and B take turns in throwing dice, the first to be awarded the prize. Show that if A has the first throw, their chances of winning are in the ratio 12:11.

Sol,  $\rightarrow$  The combinations of throwing 10 from two dice can be (6+4), (4+6), (5+5)

The number of combinations is 3.

Total combinations from two dice =  $6 \times 6 = 36$

$$\Rightarrow \text{The probability of not getting 10} = 1 - \frac{1}{12} = \frac{11}{12} = \varphi$$

If A is to win, he should throw 10 in either the first, the third, the fifth --- throws.

Their respective probabilities are  $\beta$ ,  $\varphi^2\beta$ ,  $\varphi^4\beta$  ---

$$\left(\frac{1}{12}\right), \left(\frac{1}{12}\right)^2 \cdot \frac{1}{12}, \left(\frac{1}{12}\right)^4 \cdot \frac{1}{12} \quad \dots$$

A's total probability of winning

$$= \left(\frac{1}{12}\right) + \left(\frac{1}{12}\right)^2 \cdot \frac{1}{12} + \left(\frac{1}{12}\right)^4 \cdot \frac{1}{12} \quad \dots$$

$$= \frac{\frac{1}{12}}{1 - \left(\frac{1}{12}\right)^2} = \frac{12}{23} \quad \left\{ \begin{matrix} \text{This is infinite G.P series} \\ \text{Its sum} = \frac{a}{1-\varphi} \end{matrix} \right\}$$

B can win in either 2<sup>nd</sup>, 4<sup>th</sup>, 6<sup>th</sup> --- throws

So B's total chance of winning =  $\varphi\beta + \varphi^3\beta + \varphi^5\beta \dots$

$$= \left(\frac{1}{12}\right)\left(\frac{1}{12}\right) + \left(\frac{1}{12}\right)^3\left(\frac{1}{12}\right) + \left(\frac{1}{12}\right)^5\left(\frac{1}{12}\right) \quad \dots$$

$$= \frac{\left(\frac{1}{12}\right) \cdot \left(\frac{1}{12}\right)}{1 - \left(\frac{1}{12}\right)^2} = \frac{11}{23}, \quad \text{Hence A's chance to B's chance} \\ = \frac{12}{23} : \frac{11}{23} = 12:11 \quad \underline{\text{Proved}}$$

(10)

Example → If the probability that the man aged 60 will live till 70 is 0.6, what is the probability that out of 10 men aged 60, 9 men will live upto 70.

Sol<sup>n</sup>: Given probability of a man aged 60 will live upto 70 is  $p = 0.6$  &  $q = 1-p = 1-0.6 = 0.4$   
 $n=10$ , &  $x=9$

Also, using Binomial distribution

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

$$\begin{aligned} \therefore \text{Required probability} &= P_{\text{Bin}}(9 \text{ men will live upto 70}) \\ &= {}^{10} C_9 \cdot (0.6)^9 \cdot (0.4)^1 \\ &= 10 \times \left(\frac{6}{10}\right)^9 \times \frac{4}{10} = 0.0403 \quad \underline{\text{Ans.}} \end{aligned}$$

Example → An insurance company insured 2000 scooter drivers, 4000 car drivers and 6000 truck drivers.

The probability of an accident involving a scooter driver, a car driver and a truck driver is 0.01, 0.03 & 0.15 respectively. One of the insured persons meets with an accident. What is the prob. that he is a scooter driver.

Sol<sup>n</sup>: Let  $E_1$  = The event that a person chosen is a scooter driver  
 $E_2$  = The event that a person chosen is a car driver

$E_3$  = The event that a person chosen is a truck driver

$A$  = The event that a person meets with an accident

Now, total no. of persons =  $2000 + 4000 + 6000 = 12000$

$$P(E_1) = \frac{2000}{12000} = \frac{1}{6}, \quad P(E_2) = \frac{4000}{12000} = \frac{1}{3}, \quad P(E_3) = \frac{6000}{12000} = \frac{1}{2}$$

Also,  $P(A/E_1)$  = The prob. that a person meets with an accident given that he is a scooter driver

Similarly,  $P(A/E_2) = 0.03$ , &  $P(A/E_3) = 0.15$ ,

We are asked to find  $P(E_1/A)$  given that the person meets with an accident what is the prob. that he was a scooter driver

∴ By Bayes theorem

$$\begin{aligned} P(E_1/A) &= \frac{P(E_1) \cdot P(A/E_1)}{P(E_1) \cdot P(A/E_1) + P(E_2) \cdot P(A/E_2) + P(E_3) \cdot P(A/E_3)} \\ &= \frac{\frac{1}{6} \times 0.01}{\frac{1}{6} \times 0.01 + \frac{1}{3} \times 0.03 + \frac{1}{2} \times 0.15} = \frac{0.01}{\frac{1}{6}} \times \frac{\frac{1}{6}}{0.01 + 2 \times (0.03) + 3 \times (0.15)} \end{aligned}$$

$$\Rightarrow P(E_1/A) = \frac{0.01}{0.01 + 0.06 + 0.45} = \frac{1}{1 + 6 + 45} = \frac{1}{52}$$

$$\Rightarrow \boxed{P(E_1/A) = \frac{1}{52}}$$

(11)

Example: → The odds against a solving a certain problem are 5 to 7 and the odds in favour of solving the same problem are 3 to 4. What is the probability that if both of them try, the problem would be solved.

Sol: → The prob. that A can not solve the problem is  $\frac{5}{5+7} = \frac{5}{12}$

The prob. that B can not solve the problem is  $\frac{5}{3+4} = \frac{5}{7}$

Hence, the probability that the problem will be solved is

$$= 1 - \frac{5}{21} = \frac{16}{21}. \text{ Ans.}$$

Example: → In a toy factory, machines A, B and C manufacture respectively 25%, 35%, and 40% of the total of their output 5, 4, 2 percents are respectively defective. A toy is drawn at random from the total production. What is the probability that the toy drawn is defective? Also find the probability that it was manufactured by machine A.

Sol: → Let us suppose x toy is produced in a factory

$$\text{No. of toys produced by } A = \frac{25x}{100} = \frac{x}{4}$$

$$\text{No. of toys produced by } B = \frac{35x}{100} = \frac{7x}{20}$$

$$\text{No. of toys produced by } C = \frac{40x}{100} = \frac{2x}{5}$$

$$\text{No. of defective toys produced by } A = \frac{5}{100} \times \frac{x}{4} = \frac{x}{80}$$

$$\text{No. of defective toys produced by } B = \frac{7}{100} \times \frac{7x}{20} = \frac{49x}{2000}$$

$$\text{No. of defective toys produced by } C = \frac{2}{100} \times \frac{2x}{5} = \frac{4x}{500}$$

$$\begin{aligned} P(\text{Toy drawn is defective}) &= \frac{\text{Total no. of defective toys}}{\text{Total production}} \\ &= \frac{\frac{x}{80} + \frac{49x}{2000} + \frac{4x}{500}}{x} = \frac{500x + 560x + 320x}{20000x} = \frac{1380x}{20000x} \\ &= \frac{690}{20000} = \frac{345}{10000} = \frac{69}{2000} = 0.0345 \end{aligned}$$

$$\begin{aligned} \text{Prob. (Defective/A)} &= \frac{P(\text{Defective}/A) \times P(\text{Machine A})}{P(\frac{\text{Def.}}{A}) \times P(A) + P(\frac{\text{Def.}}{B}) \times P(B) + P(\frac{\text{Def.}}{C}) \times P(C)} \\ &= \frac{\frac{5}{100} \times \frac{25}{100}}{\frac{125}{10000} \times \frac{10000}{345}} = \frac{125/10000}{69/345} = \frac{25}{69} \text{ Ans.} \end{aligned}$$

(12)

Example → In a bolt factory, there are four machines A, B, C, D manufacturing 20%, 15%, 25%, and 40% of the total output respectively of their outputs 5%, 4%, 3%, and 2% from the factory's production and is found defective. What is the probability that the bolt was manufactured by machine A or machine D?

Sol. → Let  $E_1, E_2, E_3$  and  $E_4$  denote the events that a bolt selected at random is manufactured by the machine A, B, C and D respectively and let A denote the event of its being defective. Then we have  $P(E_1) = 0.20$ ,  $P(E_2) = 0.15$ ,  $P(E_3) = 0.25$  and  $P(E_4) = 0.40$ . The probability of drawing a defective bolt manufactured by machine A is  $P(A/E_1) = 0.05$ , similarly  $P(A/E_2) = 0.04$ ,  $P(A/E_3) = 0.03$ , and  $P(A/E_4) = 0.02$ . Hence the probability that a defective bolt manufactured by machine A is given by

$$P(E_1/A) = \frac{P(E_1) P(A/E_1)}{\sum_{i=1}^n P(E_i) P(A/E_i)} = \frac{0.20 \times 0.05}{0.20 \times 0.05 + 0.15 \times 0.04 + 0.25 \times 0.03 + 0.40 \times 0.02}$$

$$= \frac{100}{100 + 60 + 75 + 80} = \frac{100}{315} = 0.3175.$$

Similarly  $P(E_4/A) = \frac{P(E_4) P(A/E_4)}{\sum_{i=1}^4 P(E_i) P(A/E_i)} = \frac{0.40 \times 0.02}{0.0315} = \frac{80}{315} = 0.254.$

Ans.

Example → There are three bags, first containing 1 white, 2 red, 3 green balls, second, 2 white, 3 red, 1 green balls and third 3 white, 1 red, 2 green balls. Two balls are drawn from a bag chosen at random. These are found to be one white and one red. Find the probability that the balls so drawn come from the second bag.

Sol. → Let  $B_1, B_2, B_3$  pertain to the first, second, third bags chosen and A, the two balls are white and red.

$$\text{Now } P(B_1) = P(B_2) = P(B_3) = \frac{1}{3}$$

$P(A/B_1) = P(\text{a white and a red ball are drawn from the first bag})$

$$= \frac{14 \times 24}{6C_2} = \frac{2}{15}$$

Similarly,  $P(A/B_2) = \frac{24 \times 34}{6C_2} = \frac{2}{5}$

$$P(A/B_3) = \frac{34 \times 14}{6C_2} = \frac{1}{5}$$

By Baye's theorem

$$P(B_2/A) = \frac{P(B_2) \cdot P(A/B_2)}{P(B_1) \cdot P(A/B_1) + P(B_2) \cdot P(A/B_2) + P(B_3) \cdot P(A/B_3)}$$

$$= \frac{\frac{1}{3} \times \frac{2}{5}}{\frac{1}{3} \times \frac{2}{15} + \frac{1}{3} \times \frac{2}{5} + \frac{1}{3} \times \frac{1}{5}} = \frac{6}{11} \quad \underline{\underline{\text{Ans.}}}$$

(13)

Random variable → If such type of variable repeated again & again, but the result always may not be the same, that is there is no certainty about their result.

For example → If a coin is tossed again & again, it is not certain when the head or tail will come.

There are two types of random variable

- 1) Discrete random variable
- 2) Continuous random variable

→ If a random variable takes a finite set of values, then it is called discrete variable

→ On the other hand, if it takes an infinite set of values is called continuous variable.

Discrete probability distribution →

Let a random variable  $X$  assume values  $x_1, x_2, x_3, \dots, x_n$  with probabilities

$p_1, p_2, p_3, \dots, p_n$  respectively, for each  $x_i$

$$p_1 + p_2 + p_3 + \dots + p_n = \sum_{i=1}^n p_i = 1$$

Then  $X : x_1 \quad x_2 \quad x_3 \quad \dots \quad x_n$

$P(X) : p_1 \quad p_2 \quad p_3 \quad \dots \quad p_n$

is called the discrete probability distribution for  $X$ .

# ⑭ The probability density function of a variable  $X$  is

$X :$	0	1	2	3	4	5	6
$P(X) :$	$k$	$3k$	$5k$	$7k$	$9k$	$11k$	$13k$

i) Find  $P(X < 2)$  ii)  $P(X \geq 5)$  iii)  $P(3 < X \leq 6)$

ii) what will be the minimum value of  $k$  so that  $P(X \leq 2) > 0.3$ .

⇒ i) If  $X$  is a random variable, then

$$\sum_{i=0}^6 P(X=i) = 1, \text{ i.e. } k + 3k + 5k + 7k + 9k + 11k + 13k = 1 \\ \Rightarrow 49k = 1 \Rightarrow k = \frac{1}{49}$$

ii)  $P(X < 2) = k + 3k + 5k + 7k = 16k = \frac{16}{49}$

iii)  $P(X \geq 5) = 11k + 13k = 24k = \frac{24}{49}$

iv)  $P(3 < X \leq 6) = 9k + 11k + 13k = 33k = \frac{33}{49}$

v)  $P(X \leq 2) = k + 3k + 5k = 9k > 0.3 \Rightarrow k > \frac{1}{30}$

thus the minimum value of  $k = \frac{1}{30}$

# Example → A random variable  $X$  has the following probability function

$X :$	0	1	2	3	4	5	6	7
$P(X) :$	$k$	$2k$	$2k$	$3k$	$k^2$	$2k^2$	$7k^2$	$7k^2$

i) find the value of  $k$  ii) evaluate  $P(X \leq 6)$ ,  $P(X \geq 6)$

iii)  $P(0 < X \leq 5)$ .

Given  $\sum_{i=0}^7 P(X=i) = 1$ , i.e  
if  $X$  is random variable, then

$$0+k+2k+2k+3k+k^2+2k^2+7k^2+k^2=1 \Rightarrow k=\frac{1}{10} \\ \Rightarrow 7k^2+9k-1=0 \text{ i.e. } (10-k)(k+1)=0, \text{ i.e. } k=\frac{1}{10}$$

ii)  $P(X \leq 6) = 0+k+2k+2k+3k+k^2 = 8k+k^2 = \frac{8}{10} + \frac{1}{100} = \frac{81}{100}$

iii)  $P(X \geq 6) = 2k^2+7k^2+k = \frac{9}{100} + \frac{1}{10} = \frac{19}{100}$

iv)  $P(0 < X \leq 5) = k+2k+2k+3k = 8k = \frac{8}{10} = \frac{4}{5}$

Simp (15)

### Continuous probability distribution

Let  $x$  be a continuous random variable

taking values in the interval  $(-\infty, \infty)$ , let  $f(x)$  be a function satisfying the following properties

i)  $f(x)$  is integrable on  $(-\infty, \infty)$

ii)  $f(x) \geq 0$  for all  $x$  in  $(-\infty, \infty)$  i.e.  $x \in \mathbb{R}$ ,

iii)  $\int_{-\infty}^{\infty} f(x) dx = 1.$

Then  $f(x)$  is called the probability distribution (or density) function (P.D.F) of the continuous random variable  $x$ .

### Mean & variance of a random variable

a) Let  $x: x_1, x_2, x_3, \dots, x_n$

$p(x_i) : p_1, p_2, p_3, \dots, p_n$

be a discrete probability distribution

then we denote the mean by  $\mu$  & define

$$\text{as } \mu = \frac{\sum p_i x_i}{\sum p_i} = \sum p_i x_i, \text{ where } \sum p_i = 1$$

other names of the mean are average or expected value  $E(X)$ .

We denote the variance by  $\sigma^2$  & define

$$\text{as } \sigma^2 = \sum p_i (x_i - \mu)^2$$

standard deviation  $\sigma = \sqrt{\text{variance}}$

b) If  $x$  is a continuous random variable with probability density function  $f(x)$

$$\text{then } \boxed{\text{mean} = E(x) = \int_{-\infty}^{\infty} x f(x) dx}$$

(16)

similarly  $E[x^2] = \int_{-\infty}^{\infty} x^2 \cdot f(x) dx$

in general  $E[x^n] = \int_{-\infty}^{\infty} x^n \cdot f(x) dx$

$$\boxed{\text{variance}(x) = \int_{-\infty}^{\infty} (x - \bar{x})^2 \cdot f(x) dx}$$
$$= E(x^2) - [E(x)]^2$$

mean deviation from the mean

$$= \int_{-\infty}^{\infty} |x - \bar{x}| \cdot f(x) dx$$

⑯ Example →  
 If  $X$  is a continuous random variable  
 with probability density function given by

$$f(x) = \begin{cases} kx & 0 \leq x < 5 \\ k(10-x) & 5 \leq x < 10 \\ 0 & \text{otherwise} \end{cases}$$

i) Find the value of  $k$ , ii) mean of  $X$

iii)  $P(5 < X < 12)$

i) ~~Soln~~  
 In order that  $f(x)$  should be a probability  
 density function  $\int_{-\infty}^{\infty} f(x) dx = 1$ .

$$\Rightarrow \int_0^5 kx dx + \int_5^{10} k(10-x) dx = 1.$$

$$\Rightarrow k \left[ \frac{x^2}{2} \right]_0^5 + 10k \left[ x \right]_5^{10} - k \left[ \frac{x^2}{2} \right]_5^{10} = 1$$

$$\Rightarrow \frac{25k}{2} + 100k - 50k - \frac{100}{2}k + \frac{25}{2}k = 1$$

$$\Rightarrow \left( \frac{25}{2} + 100 - 50 - 50 + \frac{25}{2} \right) k = 1$$

$$\Rightarrow \left( \frac{25 + 200 - 100 - 100 + 25}{2} \right) k = 1$$

$$\Rightarrow \frac{50}{2} k = 1 \Rightarrow \boxed{k = \frac{1}{25}}$$

ii) mean of  $\bar{X} = \int_0^{10} x f(x) dx$

$$= \int_0^5 x \cdot f(x) dx + \int_5^{10} x \cdot f(x) dx$$

$$= \int_0^5 x \cdot kx dx + \int_5^{10} x \cdot k(10-x) dx$$

$$= \frac{1}{25} \left[ \frac{x^3}{3} \right]_0^5 + \frac{1}{25} \left[ 10 \cdot \frac{x^2}{2} - \frac{x^3}{3} \right]_5^{10}$$

(18)

$$\begin{aligned}
 &= \frac{1}{25} \cdot \frac{125}{3} + \frac{1}{25} \left[ 500 - \frac{1000}{3} - 125 + \frac{125}{3} \right] \\
 &= \frac{1}{25} \left[ \frac{125}{3} + 500 - \frac{1000}{3} - 125 + \frac{125}{3} \right] \\
 &= \frac{1}{25} \left[ \frac{125 + 1500 - 1000 - 375 + 125}{3} \right] = \frac{1}{25} \times \frac{5}{3} \\
 &\Rightarrow \boxed{x=5}
 \end{aligned}$$

(ii)  $P(5 < X \leq 12) = \int_5^{12} f(x) dx$

$$\begin{aligned}
 &= \int_5^{10} f(x) dx + \int_{10}^{12} f(x) dx = \int_5^{10} k(10-x) dx \\
 &= \frac{1}{25} \left[ 10x - \frac{x^2}{2} \right]_5^{10} \\
 &= \frac{1}{25} \left[ 100 - 50 - 50 + \frac{25}{2} \right] = \frac{1}{25} \times \frac{25}{2} = \frac{1}{2}
 \end{aligned}$$

$$\boxed{P(5 < X \leq 12) = \frac{1}{2}}$$

# 28 Example → the function defined as follows on density function?  $f(x) = \begin{cases} e^{-x} & x \geq 0 \\ 0 & \text{otherwise} \end{cases}$

Soln.  $\Rightarrow f(x) \geq 0$ , for every  $x$  in  $(-\infty, \infty)$

$$f \int_{-\infty}^{\infty} f(x) dx = \int_{-\infty}^0 f(x) dx + \int_0^{\infty} f(x) dx$$

$$= \int_{-\infty}^0 0 dx + \int_0^{\infty} e^{-x} dx = 0 + \left[ \frac{e^{-x}}{-1} \right]_0^{\infty} = -1(0-1) = 1$$

$\Rightarrow f(x)$  satisfies the requirement of density function

$\Rightarrow f(x)$  is a density function

$$P(1.5 < x \leq 2) = \int_1^2 e^{-x} dx = \left[ \frac{e^{-x}}{-1} \right]_1^2 = e^{-1} - e^{-2}$$

$$= 0.368 - 0.135$$

$$= 0.233$$

(19)

Example  $\rightarrow$   $x$  is a continuous random variable given by with probability density function

$$f(x) = \begin{cases} kx, & \text{if } 0 \leq x < 2 \\ 2k, & \text{if } 2 \leq x < 4 \\ -kx + 6k, & \text{if } 4 \leq x < 6. \end{cases}$$

Find  $k$  and mean value of  $x$ .

Sol  $\rightarrow$  since total prob. = 1

we have  $\int_0^6 f(x) dx = 1$

$$\Rightarrow \int_0^2 f(x) dx + \int_2^4 f(x) dx + \int_4^6 f(x) dx = 1$$

$$\Rightarrow \int_0^2 kx dx + \int_2^4 2k dx + \int_4^6 (-kx + 6k) dx = 1$$

$$\Rightarrow k \left[ \frac{x^2}{2} \right]_0^2 + 2k \left[ x \right]_2^4 + \left[ -\frac{k}{2} x^2 + 6kx \right]_4^6 = 1$$

$$\Rightarrow k(2-0) + 2k(4-2) + (-18k + 36k) - (-8k + 24k) = 1$$

$$\Rightarrow 2k + 4k + 18k - 16k = 1 \Rightarrow 8k = 1 \Rightarrow \boxed{k = \frac{1}{8}}$$

mean value of  $x = E(x) = \int_0^6 x \cdot f(x) dx$

$$= \int_0^2 x \cdot kx dx + \int_2^4 x \cdot 2k dx + \int_4^6 x(-kx + 6k) dx$$

$$= k \left[ \frac{x^3}{3} \right]_0^2 + 2k \left[ \frac{x^2}{2} \right]_2^4 + \left[ -\frac{k}{3} x^3 + 6k \cdot \frac{x^2}{2} \right]_4^6$$

$$= k \left( \frac{8}{3} \right) + k \left( 12 \right) - \frac{k}{3} (152) + 3k (20)$$

$$= 24k = 24 \times \frac{1}{8} = 3$$

$$\Rightarrow \boxed{k = \frac{1}{8}} \quad \boxed{\text{mean} = 3}$$

(20)

Example → Determine the value of  $k$ , if the probability function of a random variable  $x$  is given by

$$P(x) = \begin{cases} \frac{kx}{20}, & x = 1, 2, 3, 4 \\ 0, & \text{for other integers} \end{cases}$$

Sol →  $P(x) = \begin{cases} \frac{kx}{20}, & x = 1, 2, 3, 4 \\ 0, & \text{for other integers} \end{cases}$

By discrete distribution function, we have

$$\begin{aligned} \Rightarrow \sum_{i=0}^{\infty} P(x_i) = 1 &\Rightarrow P(x_0) + P(x_1) + P(x_2) + P(x_3) + P(x_4) = 1 \\ \Rightarrow 0 + \frac{k}{20} + \frac{2k}{20} + \frac{3k}{20} + \frac{4k}{20} &= 1 \\ \Rightarrow \frac{10k}{20} &= 1 \Rightarrow [k=2] \end{aligned}$$

Example → A random variable  $x$  has the following density function

$$\begin{array}{ccccccc} x: & -2 & -1 & 0 & 1 & 2 & 3 \\ f(x): & 0.1 & k & 0.2 & 2k & 0.3 & k \end{array}$$

Find the value  $k$  and calculate mean & variance.

Sol → For the value of  $k$   $\sum P(x) = 1$ .

$$\begin{aligned} \Rightarrow 0.1 + k + 0.2 + 2k + 0.3 + k &= 1 \\ \Rightarrow 4k &= 0.4 \Rightarrow [k=0.1] \end{aligned}$$

$$\text{Mean} = \mu = \sum x_i p(x_i) = -2 \times 0.1 - 1 \times 0.1 + 0 \times 0.2 + 2 \times 0.1 + 1 + 0.3 \times 2 + 3 \times 0.1 = -0.2 - 0.1 + 0.2 + 0.6 + 0.3 = 0.8$$

$$\begin{aligned} \text{Variance} (\sigma^2) &= \sum (x_i - \mu)^2 \cdot p(x_i) \\ &= (-2 - 0.8)^2 \times 0.1 + (-1 - 0.8)^2 \times 0.1 \\ &\quad + (0 - 0.8)^2 \times 0.2 + (1 - 0.8)^2 \times 0.2 + (2 - 0.8)^2 \times 0.3 \\ &\quad + (3 - 0.8)^2 \times 0.1 \\ &= 0.1 [(-2.8)^2 + (-1.8)^2 + (-0.8)^2 \times 2 + (0.2)^2 \times 2 + (1.2)^2 \times 3 \\ &\quad + (2.8)^2] = 0.1 [7.84 + 3.24 + 1.28 + 0.08 + 4.32 + 7.84] \\ &= 2.23 \end{aligned}$$

$$\Rightarrow [k=0.1] \quad [\mu=0.8] \quad [\sigma^2=2.23]$$

(2) Example → The density function of a random variable  $x$  is given by  $f(x) = kx(2-x)$ ,  $0 \leq x \leq 2$ . Find (i)  $k$  (ii) mean (iii) variance (iv) mean deviation about the mean.

Soln: → (i) since total probability = 1, we have  
 $\int_0^2 f(x) dx = 1$ ,

$$\Rightarrow \int_0^2 kx(2-x) dx = 1 \Rightarrow k \left[ x^2 - \frac{x^3}{3} \right]_0^2 = 1$$

$$\Rightarrow k(4 - \frac{8}{3}) = 1 \Rightarrow k \cdot \frac{4}{3} = 1 \Rightarrow k = \frac{3}{4}$$

$$\text{Hence } f(x) = \frac{3}{4}x(2-x), \quad 0 \leq x \leq 2.$$

$$\begin{aligned} \text{(ii) Mean} &= E(x) = \int_0^2 x f(x) dx = \frac{3}{4} \int_0^2 x^2 (2-x) dx \\ &= \frac{3}{4} \left[ \frac{2x^3}{3} - \frac{x^4}{4} \right]_0^2 = \frac{3}{4} \left( \frac{16}{3} - 4 \right) = \frac{3}{4} \cdot \frac{4}{3} = 1 \end{aligned}$$

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

$$= \int_0^2 x^2 f(x) dx - (1)^2 = \frac{3}{4} \int_0^2 x^3 (2-x) dx - 1$$

$$= \frac{3}{4} \left[ 2 \cdot \frac{x^4}{4} - \frac{x^5}{5} \right]_0^2 - 1 = \frac{3}{4} \left( 8 - \frac{32}{5} \right) - 1$$

$$\Rightarrow \frac{3}{4} \left[ 2 \cdot \frac{x^4}{4} - \frac{x^5}{5} \right]_0^2 - 1 = \frac{3}{4} \cdot \frac{28}{5} - 1 = \frac{6}{5} - 1 = \frac{1}{5}$$

$$\Rightarrow \boxed{\text{Var}(x) = \frac{1}{5}}$$

(iv) Mean deviation about the mean

$$= \int_0^2 |x - \bar{x}| f(x) dx = \int_0^2 |x - 1| f(x) dx$$

$$= \int_0^1 |x - 1| f(x) dx + \int_1^2 |x - 1| f(x) dx$$

$$= \int_0^1 (1-x) \cdot \frac{3}{4}x(2-x) dx + \int_1^2 (x-1) \cdot \frac{3}{4}x(2-x) dx$$

$$= \frac{3}{4} \int_0^1 (2x - 3x^2 + x^3) dx + \frac{3}{4} \int_1^2 (-2x + 3x^2 - x^3) dx$$

$$= \frac{3}{4} \left[ x^2 - x^3 + \frac{x^4}{4} \right]_0^1 + \frac{3}{4} \left[ -x^2 + x^3 - \frac{x^4}{4} \right]_1^2,$$

$$= \frac{3}{4} \left( \frac{1}{4} \right) + \frac{3}{4} \left( \frac{1}{4} \right) = \frac{3}{8}, \quad \underline{\text{Ans}}$$

(22)

Moment generating function (m.g.f.):  $\rightarrow$

The moment generating function of a random variable  $x$  about the point 'a' is defined as

$$M_a(t) = E[e^{t(x-a)}] = \int_{-\infty}^{\infty} e^{t(x-a)} \cdot f(x) dx$$

where  $t$  is a real number &  $x$  is continuous

Moment about the origin,  $\rightarrow$  we have

$$M_0(t) = E[e^{tx}] = \int_{-\infty}^{\infty} e^{tx} \cdot f(x) dx \text{ if } x \text{ is continuous}$$

$$\& M_0(t) = E[e^{tx}] = \sum_{x=1}^{\infty} e^{tx} f(x) \text{ if } x \text{ is discrete}$$

Now,

$$\begin{aligned} M_a(t) &= E[e^{t(x-a)}] = E\left[1 + t(x-a) + \frac{t^2}{2!}(x-a)^2 + \frac{t^3}{3!}(x-a)^3 + \dots + \frac{t^8}{8!}(x-a)^8 + \dots\right] \\ &= E[1] + tE(x-a) + \frac{t^2}{2!}E(x-a)^2 + \frac{t^3}{3!}E(x-a)^3 + \dots + \frac{t^8}{8!}E[(x-a)^8] \end{aligned}$$

where  $\mu'_s = E[(x-a)^s]$  is the  $s$ th moment about the point  $a$ .

Further  $\rightarrow \frac{d}{dt}\{M_a(t)\} = \mu'_1 + \mu'_2 t + \mu'_3 \frac{t^2}{2} + \dots$

$$\Rightarrow \left| \frac{d}{dt}\{M_a(t)\} \right|_{at t=0} = \mu'_1$$

Again

$$\frac{d^2}{dt^2}\{M_a(t)\} = \mu'_2 + \mu'_3 t + \dots$$

$$\Rightarrow \left| \frac{d^2}{dt^2} M_a(t) \right|_{at t=0} = \mu'_2$$

Note,  $\rightarrow$  where  $\mu'_1$  is moment about mean and  $\mu'_s$  is moment about any number.

$$\text{In general } \left| \frac{d^s}{dt^s} M_a(t) \right|_{at t=0} = \mu'_s$$

Remark  $\rightarrow$  ① Mean =  $E[x] = \mu'_1$   
if  
 $\text{Var}[x] = \mu'_2 - (\mu'_1)^2$

② Properties,  $\rightarrow$  (i)  $M_{cx}(t) = M_x(ct)$

$$(ii) M_a(t) = e^{at} M_0(t)$$

(iii) If  $x_1, x_2$  are two independent random variable, then  $M_{x_1+x_2} = M_{x_1}(t) \cdot M_{x_2}(t)$

(23)

$$\textcircled{3} \quad \mu_2 = \mu'_2 - (\mu'_1)^2, \quad \mu_3 = \mu'_3 - 3\mu'_1\mu'_2 + 2\mu'^3 \\ \mu_4 = \mu'_4 - 4\mu'_1\mu'_3 + 6\mu'_2\mu'^2 - 3\mu'^4$$

$$\textcircled{4} \quad \beta_1 = \frac{\mu'_3}{\mu'^2_2} \quad \text{if } \beta_1 = 0 \Rightarrow \text{the distribution is symmetrical} \\ \beta_2 = \frac{\mu'_4}{\mu'^2_2}$$

Example → Find the moment generating function of the random variable  $x$  having the density function  
 $f(x) = \begin{cases} x, & 0 < x < 1. \\ 2-x, & 1 \leq x < 2. \\ 0 & \text{elsewhere} \end{cases}$ . Also determine  $\mu'$ ,  $\mu'_1$  &  $\mu_2$ .

Sol → We have  $M_x(t) = \int_0^\infty e^{tx} f(x) dx$

$$M_x(t) = \int_0^1 x \cdot e^{tx} dx + \int_1^2 (2-x) e^{tx} dx + \int_2^\infty 0 \cdot e^{tx} dx$$

$$= \left\{ \frac{x e^{tx}}{t} - \frac{e^{tx}}{t^2} \right\}_0^1 + \left\{ \frac{2 e^{tx}}{t} - \frac{x e^{tx}}{t} + \frac{e^{tx}}{t^2} \right\}_1^\infty$$

$$= \frac{e^t}{t} - \frac{e^t}{t^2} + \frac{1}{t^2} + \left[ \left\{ \frac{2 e^{2t}}{t} - \frac{2 e^{2t}}{t} + \frac{e^{2t}}{t^2} \right\}_1^\infty - \left\{ \frac{2 e^t}{t} - \frac{e^t}{t} + \frac{e^t}{t^2} \right\}_0^1 \right]$$

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

$$= \frac{(e^t - 1)^2}{t^2} = \frac{1 + t + \frac{t^2}{2} + \frac{t^3}{3!} + \dots - 1}{t^2}$$

$$\Rightarrow M_x(t) = 1 + t + t^2 + \dots$$

$$\mu' = \left[ \frac{d}{dt} \{ M_x(t) \} \right]_{at \ t=0} = 1$$

$$\mu'_2 = \left[ \frac{d^2}{dt^2} \{ M_x(t) \} \right]_{at \ t=0} = 2$$

$$\therefore \mu_2 = \mu'_2 - \mu'^2_1 = 2 - 1 = 1 \Rightarrow \boxed{\mu_2 = 1}$$

$$\Rightarrow \boxed{\mu' = 1} \quad \boxed{\mu'_1 = 2} \quad \text{and} \quad \boxed{\mu_2 = 1} \quad \text{Ans.}$$

(24)

Example → A random variable  $x$  has probability function  $p(x) = \frac{1}{2^x}$ ,  $x=1, 2, 3, \dots$ . Find its moment generating function about origin.

Sol<sup>n</sup>,  $\rightarrow p(x) = \frac{1}{2^x}$ ,  $x=1, 2, 3, \dots$

The moment generating function of  $x$  is

$$M_X(t) = E(e^{tx}) = \sum_{x=1}^{\infty} e^{tx} \cdot \frac{1}{2^x} = \sum_{x=1}^{\infty} \left(\frac{e^t}{2}\right)^x$$

$$\begin{aligned} M_X(t) &= \frac{e^t}{2} + \left(\frac{e^t}{2}\right)^2 + \left(\frac{e^t}{2}\right)^3 + \dots \\ &= \frac{e^t}{2} \left[1 + \frac{e^t}{2} + \left(\frac{e^t}{2}\right)^2 + \dots\right] \quad \{G.P. series\} \end{aligned}$$

$$= \frac{e^t}{2} \left[\frac{1}{1 - \frac{e^t}{2}}\right] = \frac{e^t}{2} \cdot \frac{1}{\frac{2 - e^t}{2}} = \frac{e^t}{2 - e^t}$$

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

$$\begin{aligned} \frac{d}{dt} [M_X(t)] &= \frac{d}{dt} \left[ \frac{e^t}{2 - e^t} \right] = \frac{(2 - e^t) \cdot e^t - e^t \cdot (-e^t)}{(2 - e^t)^2} \\ &= \frac{2e^t - e^{2t} + e^{2t}}{(2 - e^t)^2} = \frac{2 \cdot e^t}{(2 - e^t)^2} \end{aligned}$$

$$Y_1 = \left\{ \frac{d}{dt} [M_X(t)] \right\}_{at t=0} = \frac{2}{1^2} = 2 \Rightarrow \boxed{Y_1 = 2}$$

$\Rightarrow \boxed{Y_1 = 2}$ , where  $Y_1$  = first moment about origin  
Similarly  $\frac{d^2}{dt^2} [M_X(t)] = 2 \left[ \frac{e^t \cdot (2 - e^t)^2 + 2 \cdot e^t (2 - e^t) \cdot (-e^t)}{(2 - e^t)^3} \right]$

$$\begin{aligned} \Rightarrow \frac{d^2}{dt^2} [M_X(t)] &= 2 \left[ \frac{e^t (2 - e^t) + 2 \cdot e^{2t}}{(2 - e^t)^3} \right] = \frac{4e^t - 2e^{2t} + 4e^{2t}}{(2 - e^t)^3} \\ &= \frac{4e^t + 2 \cdot e^{2t}}{(2 - e^t)^3} \end{aligned}$$

$$Y_2 = \left[ \frac{d^2}{dt^2} \{M_X(t)\} \right]_{at t=0} = \frac{6}{1} = 6$$

$$\Rightarrow \boxed{Y_2 = 6}$$

(25) Example: Find the moment generating function of the distribution  $f(x) = \frac{1}{c} e^{-x/c}$ ,  $0 \leq x < \infty$ ,  $c > 0$ , about origin. Hence find its mean and standard deviation.

Sol:  $f(x) = \frac{1}{c} e^{-x/c}$ ,  $0 \leq x < \infty$ ,  $c > 0$

The moment generating function about the origin is

$$M_0(t) = \int_0^\infty e^{tx} \cdot \frac{1}{c} \cdot e^{-x/c} dx = \frac{1}{c} \int_0^\infty e^{(t - \frac{1}{c})x} dx$$

$$= \frac{1}{c} \frac{1}{(t - \frac{1}{c})} \Big|_0^\infty = (1 - ct)^{-1}$$

$$\Rightarrow M_0(t) = 1 + ct + c^2 t^2 + c^3 t^3 + \dots$$

$$\mu' = \left[ \frac{d}{dt} \{M_0(t)\} \right]_{at \ t=0} = (c + 2c^2 t + 3c^3 t^2 + \dots)_{at \ t=0} = c$$

$$\Rightarrow \boxed{\mu' = c} \quad \mu'_2 = \left[ \frac{d^2}{dt^2} \{M_0(t)\} \right]_{at \ t=0} = 2c^2$$

$$\therefore \mu_2 = \mu'_2 - (\mu')^2 = 2c^2 - c^2 = c^2$$

$$\text{Hence } \boxed{\text{mean} = c} \quad \& \quad \text{s.d} = \sqrt{\mu_2} \Rightarrow \boxed{\text{s.d} = c}$$

Example: For the continuous distribution  $df = y_0 x(2-x) dx$ ,  $0 \leq x \leq 2$ .

- (i) Find mean, variance,  $\beta_1$ ,  $\beta_2$  and hence show that the distribution is symmetrical  
(ii) Also find mean deviation about mean.

Sol: since  $\frac{df}{dx} = f(x) = y_0 x(2-x)$ .

$$\text{we know that } \int_{-\infty}^{\infty} f(x) dx = 1 \Rightarrow \int_0^2 y_0 x(2-x) dx = 1$$

$$\Rightarrow y_0 = \frac{3}{4}, \text{ Hence the p.d.f is } f(x) = \frac{3}{4} x(2-x), 0 \leq x \leq 2.$$

Now, we know  $\sigma$ th moment  $\mu'_\sigma = E[x^\sigma] = \int_0^2 x^\sigma \cdot f(x) dx$

$$\Rightarrow \mu'_\sigma = \frac{3}{4} \int_0^2 x^\sigma \cdot x(2-x) dx \Rightarrow \mu'_\sigma = \frac{3}{4} \int_0^2 x^{\sigma+1} (2-x) dx$$

$$= \frac{3 \cdot 2^{\sigma+1}}{(\sigma+2)(\sigma+3)}$$

Put  $\sigma = 1, 2, 3, 4, \dots$  we obtain

$$\mu' = \frac{3 \cdot 2^2}{3 \cdot 4} = 1, \quad \mu'_2 = \frac{3 \cdot 2^3}{4 \cdot 5} = \frac{6}{5}, \quad \mu'_3 = \frac{3 \cdot 2^4}{5 \cdot 6} = \frac{8}{5}$$

$$\mu'_4 = \frac{3 \cdot 2^5}{6 \cdot 7} = \frac{16}{7},$$

(26)

$$\text{variance} = \mu_2 = \mu'_2 - (\mu'_1)^2 = \frac{6}{5} - (1)^2 = \frac{1}{5}$$

$$\mu_3 = \mu'_3 - 3\mu'_2\mu'_1 + 2\mu'^3_1 = \frac{8}{5} - 3 \cdot \frac{6}{5} \cdot 1 + 2 \cdot 1^3 = 0$$

$$\mu_4 = \mu'_4 - 4\mu'_3\mu'_1 + 6\mu'_2\mu'^2_1 - 3\mu'^4_1$$

$$= \frac{16}{17} - 4 \cdot \frac{8}{5} \cdot 1 + 6 \cdot \frac{6}{5} \cdot 1 - 3 \cdot 1^4 = \frac{3}{35}$$

Now  $\beta_1 = \frac{\mu'_3}{\mu'^2_2} = 0 \Rightarrow$  the distribution is symmetrical

$$\beta_2 = \frac{\mu_4}{\mu'^2_2} = \frac{3/35}{(1/5)^2} = \frac{15}{7}$$

(ii) Mean deviation about mean  $= E[|x - \mu'|]$

$$\Rightarrow E[|x - \mu'|] = \int_0^2 |x - 1| f(x) dx$$

$$= \int_0^1 |x - 1| f(x) dx + \int_1^2 |x - 1| f(x) dx$$

$$= \frac{3}{4} \left[ \int_0^1 (1-x) \cdot x \cdot (2-x) dx + \int_1^2 (1-x) \cdot x \cdot (2-x) dx \right]$$

$$= \frac{3}{4} \left[ \left[ \left| x^2 - x^3 + \frac{x^4}{4} \right| \right]_0^1 + \left[ \left| x^3 - \frac{x^4}{4} - x^2 \right| \right]_1^2 \right] = \frac{3}{8}.$$

Example,  $\rightarrow$  A variate  $x$  has the probability distribution

$$x : -3 \quad 6 \quad 9$$

$$P[x=x] : \quad y_1 \quad y_2 \quad y_3$$

Find  $E[x]$  &  $E[x^2]$  and using the law of expectation, evaluate  $E[2x+1]^2$ .

$$\text{Soln}, \rightarrow E[x] = \sum x P(x) = -3 \cdot \frac{1}{6} + 6 \cdot \frac{1}{2} + 9 \cdot \frac{1}{3} = \frac{11}{2}$$

$$E[x^2] = \sum x^2 P(x) = 9 \cdot \frac{1}{6} + 36 \cdot \frac{1}{2} + 81 \cdot \frac{1}{3} = \frac{93}{2}$$

$$\therefore E[2x+1]^2 = E[4x^2 + 4x + 1]$$

$$= 4(E[x]^2) + 4E[x] + 1 = 4 \times \frac{93}{2} + 4 \times \frac{11}{2} + 1 \\ = 209.$$

(27)

## Skewness and Kurtosis

- ① Symmetry → A distribution is said to be symmetrical when its mean, median and mode are identical i.e.  $\boxed{\text{Mean} = \text{Median} = \text{Mode}}$



- ② Skewness → Skewness denotes the opposite of symmetry. i.e. Skewness means lack of symmetry.

Skew symmetrical distribution → A distribution which is not symmetrical is said to be skew symmetrical distribution.

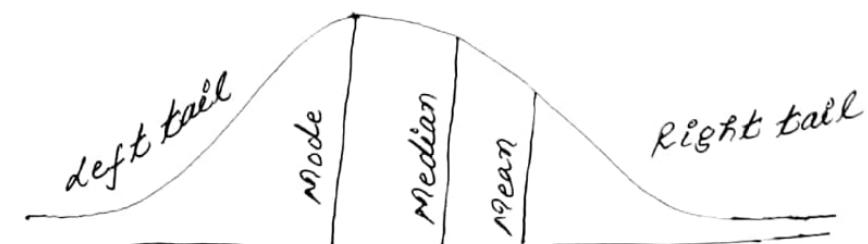
In skew symmetrical distribution, the left and the right tail are not of equal length.

One tail will be larger than the other.

- (a) Negative skew distribution → The left tail is larger than the right tail.



- (b) Positive skew distribution → The right tail of the curve will be longer than the left tail.



- Uses of skewness → ① It gives the nature of the curve ② It gives nature and concentration of observation about the mean.

(28)

Measure of Skewness  $\xrightarrow{x}$  Measure of skewness is known as the measure of symmetry.

There are two types of measure of skewness.

1. Absolute measure,  $\rightarrow$  Absolute measure = |Mean - Mode|
2. Relative measure,  $\rightarrow$  There are four type of relative measure of skewness.

(i) Karl Pearson's coefficient of skewness

(ii) Bowley's coefficient of skewness

(iii) Kelly's coefficient of skewness

(iv) measure of skewness based on the moments

$$\boxed{\text{Mode} = 3\text{Median} - 2\text{Mean}}$$

Moments,  $\rightarrow$  The  $\gamma$ th moment of a variable  $x$  about the mean  $\mu$  is usually denoted by  $\mu_\gamma$  is given by

$$\mu_\gamma = \frac{1}{N} \sum f_i (x_i - \mu)^\gamma \quad \sum f_i = N$$

The  $\gamma$ th moment of a variable  $x$  about any point 'a' is defined by

$$\mu'_\gamma = \frac{1}{N} \sum f_i (x_i - a)^\gamma$$

Moment about mean,  $\rightarrow$  Let  $\mu$  be the mean

$$\text{Then } \mu_\gamma = \frac{1}{N} \sum_{i=1}^n f_i (x_i - \mu)^\gamma \quad \gamma = 1, 2, 3, \dots$$

$$\text{If } \gamma = 0, \quad \mu_0 = \frac{1}{N} \sum_{i=1}^n f_i (x_i - \mu)^0 = 1$$

$$\text{If } \gamma = 1, \quad \mu_1 = \frac{1}{N} \sum_{i=1}^n f_i (x_i - \mu) = 0$$

$$\text{If } \gamma = 2, \quad \mu_2 = \frac{1}{N} \sum_{i=1}^n f_i (x_i - \mu)^2$$

$$\text{If } \gamma = 3, \quad \mu_3 = \frac{1}{N} \sum_{i=1}^n f_i (x_i - \mu)^3$$

$$\text{If } \gamma = 4, \quad \mu_4 = \frac{1}{N} \sum_{i=1}^n f_i (x_i - \mu)^4$$

Moment about point,  $\rightarrow$  Let 'a' be any arbitrary point.

$$\mu'_\gamma = \frac{1}{N} \sum_{i=1}^n f_i (x_i - a)^\gamma \quad \gamma = 0, 1, 2, \dots$$

$$\text{If } \gamma = 0, \quad \mu'_0 = \frac{1}{N} \sum_{i=1}^n f_i (x_i - a)^0 = 1$$

$$\text{If } \gamma = 1, \quad \mu'_1 = \frac{1}{N} \sum_{i=1}^n f_i (x_i - a) = \frac{1}{N} \sum f_i x_i - \frac{a}{N} \sum f_i$$

$$\mu'_1 = \mu - a$$

$$(29) \text{ If } \alpha = 2, \mu'_2 = \frac{1}{N} \sum_{i=1}^n f_i (x_i - \alpha)^2$$

$$\text{If } \alpha = 3, \mu'_3 = \frac{1}{N} \sum_{i=1}^n f_i (x_i - \alpha)^3$$

$$\text{If } \alpha = 4, \mu'_4 = \frac{1}{N} \sum_{i=1}^n f_i (x_i - \alpha)^4$$

Moment about the origin  $\rightarrow \nu_\alpha = \frac{1}{N} \sum_{i=1}^n f_i x_i^\alpha$

$$\text{If } \alpha = 0, \nu_0 = \frac{1}{N} \sum_{i=1}^n f_i x_i^0 = 1, \text{ if } \alpha = 1, \nu_1 = \frac{1}{N} \sum_{i=1}^n f_i x_i^1 = \mu$$

$$\text{If } \alpha = 2, \nu_2 = \frac{1}{N} \sum_{i=1}^n f_i x_i^2, \text{ if } \alpha = 3, \nu_3 = \frac{1}{N} \sum_{i=1}^n f_i x_i^3$$

$$\text{If } \alpha = 4, \nu_4 = \frac{1}{N} \sum_{i=1}^n f_i x_i^4$$

Relation between  $\mu_r$  and  $\mu'_r$   $\rightarrow$  we have  $r^{th}$  moment about the mean  $\mu$

$$\mu_r = \frac{1}{N} \sum_{i=1}^{\infty} f_i (x_i - \mu)^r \text{ where } N = \sum f_i$$

$$= \frac{1}{N} \sum_{i=1}^n f_i [(x_i - \mu) + (\mu - \alpha)]^r = \frac{1}{N} \sum_{i=1}^n f_i [(x_i - \alpha) + \mu' - \mu]^r$$

$$\therefore \mu' = \mu - \alpha$$

On expanding by Binomial theorem

$$\mu_r = \frac{1}{N} \sum_{i=1}^n f_i \left[ (x_i - \alpha)^r - r_1 (x_i - \alpha)^{r-1} \mu' + r_2 (x_i - \alpha)^{r-2} \mu'^2 - \dots + (-1)^r \mu'^r \right]$$

$$\mu_r = \mu'_r - r_1 \mu' \mu'_1 + r_2 \mu' \mu'^2 - \dots + (-1)^r \mu'^r$$

put  $r = 2, 3, 4, \dots$  we get

$$\mu_2 = \mu'_2 - 2\mu' \mu'_1 + 2\mu' \mu'^2 \quad (\because \mu'_0 = 1)$$

$$= \mu'_2 - 2\mu'^2 + \mu'^2 = \mu'_2 - \mu'^2$$

$$\Rightarrow \boxed{\mu_2 = \mu'_2 - \mu'^2}$$

$$\mu_3 = \mu'_3 - 3\mu'_2 \mu' \mu'_1 + 3\mu'_1 \mu' \mu'^2 - 3\mu'_0 \mu' \mu'^3$$

$$= \mu'_3 - 3\mu'_2 \mu' \mu'_1 + 3\mu'^3 - \mu'^3 = \mu'_3 - 3\mu'_2 \mu' \mu'_1 + 2\mu'^3$$

$$\mu_4 = \mu'_4 - 4\mu'_3 \mu' \mu'_1 + 6\mu'_2 \mu' \mu'^2 - 4\mu'_1 \mu' \mu'^3 + 4\mu'_0 \mu' \mu'^4$$

$$= \mu'_4 - 4\mu'_3 \mu' \mu'_1 + 6\mu'_2 \mu' \mu'^2 - 4\mu'^4 + \mu'^4$$

$$= \mu'_4 - 4\mu'_3 \mu' \mu'_1 + 6\mu'_2 \mu' \mu'^2 - 3\mu'^4$$

(30)

Thus

$$\mu_4 = 0$$

$$\mu_2 = \mu'_2 - \mu'^2$$

$$\mu_3 = \mu'_3 - 3\mu'_2\mu'_1 + 2\mu'^3$$

$$\mu_4 = \mu'_4 - 4\mu'_3\mu'_1 + 6\mu'_2\mu'^2 - 3\mu'^4$$

Are first four moments

Measure of Skewness based on moment →

Measure of skewness is given by  $\beta_1$

where 
$$\beta_1 = \frac{\mu'_3}{\mu'^2}$$

Relation between  $\mu_r$  and  $\mu'_r$  →

$$\mu'_1 = \mu_1 - a \quad \mu'_3 = \mu_3 + 3\mu_2\mu'_1 + \mu'^3$$

$$\mu'_2 = \mu_2 + \mu'^2 \quad \mu'_4 = \mu_4 + 4\mu_3\mu'_1 + 6\mu_2\mu'^2 + \mu'^4$$

Relation between  $\sigma$  →  $\sigma_1 = \bar{x}$ ,  $\sigma_2 = \mu_2 + \bar{x}^2$   
 $\sigma_3 = \mu_3 + 3\mu_2\bar{x} + \bar{x}^3$

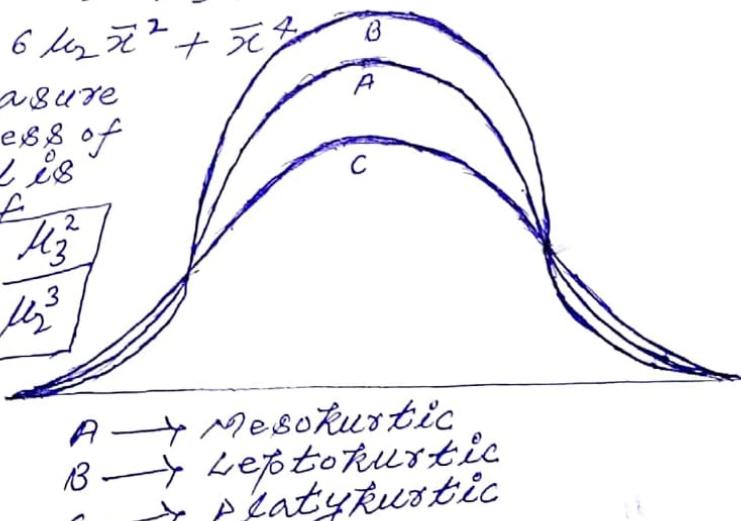
$$\sigma_4 = \mu_4 + 4\mu_3\bar{x} + 6\mu_2\bar{x}^2 + \bar{x}^4$$

Kurtosis → It measure the degree of peakness of a distribution and is given by measure of kurtosis

Kurtosis

$$\beta_2 = \frac{\mu_4}{\mu'^2}$$

$$\beta_1 = \frac{\mu'_3}{\mu'^2}$$



- 1) If  $\beta_2 = 3$ , the curve is normal or mesokurtic
- 2) If  $\beta_2 > 3$ , the curve is peaked or leptokurtic
- 3) If  $\beta_2 < 3$ , the curve is flat topped or platykurtic

Gamma coefficients → 
$$\gamma_1 = \pm \sqrt{\beta_1} \quad \gamma_2 = \beta_2 - 3$$

(31) Example → The first four moments of a distribution about the value 4 of the variate are  $-1.5, 17, -30$ , and 108, calculate the first four moments about the mean and find  $\beta_1$  and  $\beta_2$ .

Soln → We have  $a = 4$ ,  $\mu' = -1.5$ ,  $\mu'_2 = 17$ ,  $\mu'_3 = -30$  &  $\mu'_4 = 108$

Moments about the mean

$$\mu_1 = 0, \mu_2 = \mu'_2 - \mu'^2 = 17 - (-1.5)^2 = 14.75$$

$$\mu_3 = \mu'_3 - 3\mu'_2 \mu' + 2\mu'^3 = -30 - 3(17)(-1.5) + 2(-1.5)^3 = 39.75$$

$$\mu_4 = \mu'_4 - 4\mu'_3 \mu' + 6\mu'_2 \mu'^2 - 3\mu'^4$$

$$= 108 - 4(-30)(-1.5) + 6(17)(-1.5)^2 - 3(-1.5)^4 = 142.31$$

$$\beta_1 = \frac{\mu_3^2}{\mu_2^2} = \frac{(39.75)^2}{(14.75)^2} = 0.4924$$

$$\beta_2 = \frac{\mu_4}{\mu_2^2} = \frac{142.31}{(14.75)^2} = 0.6541$$

Also mean  $\mu = \mu' + a = -1.5 + 4 = 2.5$

Example → The first three moments of a distribution about the value 2 of the variable are 1, 16, and -40. Show that the mean is 3, variance is 15 and  $\mu_3 = -86$ .

Soln → Given  $a = 2$ ,  $\mu' = 1$ ,  $\mu'_2 = 16$ ,  $\mu'_3 = -40$

we know that  $\mu = \mu' + a$ ,  $\Rightarrow \mu = \mu' + a = 1 + 2 = 3$

$\Rightarrow$  Mean = 3, variance =  $\mu'_2 - \mu'^2 = 16 - (1)^2 = 15$ .

and  $\mu_3 = \mu'_3 - 3\mu'_2 \mu' + 2\mu'^3 = -40 - 3(16)(1) + 2(1)^3 = -86$ .

Example → For a distribution, the mean is 10, variance is 16,  $\gamma_1$  is 1 and  $\beta_2$  is 4. Find the first four moments about the origin.

Soln → When  $\mu = 10$ ,  $\mu_2 = 16$ ,  $\gamma_1 = 1$ ,  $\beta_2 = 4$

Now  $\gamma_1 = 1 \Rightarrow \sqrt{\beta_1} = 1 \Rightarrow \beta_1 = 1$

Also  $\beta_1 = 1, \Rightarrow \frac{\mu_3^2}{\mu_2^2} = 1 \Rightarrow \mu_3^2 = \mu_2^3 = (16)^3 = (4 \times 4)^3 = (64)^2$

or  $\mu_3^2 = 64^2 \Rightarrow \mu_3 = 64$  and  $\beta_2 = 4 \Rightarrow \frac{\mu_4}{\mu_2^2} = 4$

$\Rightarrow \mu_4 = 4(\mu_2^2) = 4(16)^2 = 1024$ .

Moments about the origin:  $\mu = \mu = 10$

$$\mu_2 = \mu_2 + \mu^2 = 16 + (10)^2 = 116$$

$$\mu_3 = \mu_3 + 3\mu_2 \mu + \mu^3 = 64 + 3(16)(10) + (10)^3 = 1544$$

$$\mu_4 = \mu_4 + 4\mu_3 \mu + 6\mu_2 \mu^2 + \mu^4$$

$$= 1024 + 4(64)(10) + 6(16)(10)^2 + (10)^4 = 23184.$$

(32)

Example → calculate the variance and third central moment from the following data:

$x_i^o$	0	1	2	3	4	5	6	7	8
$f_i^o$	1	9	26	59	72	52	29	7	1

Sol." → let us take assumed mean  $a = 4$

$x_i^o$	$f_i^o$	$x_i^o - 4$	$f_i^o (x_i^o - 4)$	$f_i^o (x_i^o - 4)^2$	$f_i^o (x_i^o - 4)^3$
0	1	-4	-4	16	-64
1	9	-3	-27	81	-243
2	26	-2	-52	104	-208
3	59	-1	-59	59	-59
4	72	0	0	0	0
5	52	1	52	52	52
6	29	2	58	116	232
7	7	3	21	63	189
8	1	4	4	16	64
	256		-7	507	-37

$$\text{Since } \mu'_1 = \frac{\sum f_i^o (x_i^o - a)^1}{\sum f_i^o}$$

$$\text{Then } \mu'_1 = \frac{\sum f_i^o (x_i^o - a)}{\sum f_i^o} = \frac{-7}{256}$$

$$\mu'_2 = \frac{\sum f_i^o (x_i^o - a)^2}{\sum f_i^o} = \frac{507}{256}$$

$$\mu'_3 = \frac{\sum f_i^o (x_i^o - a)^3}{\sum f_i^o} = \frac{-37}{256}$$

$$\text{variance } \sigma^2 = \mu'_2 - \mu'^2_1 = \frac{507}{256} - \left(\frac{-7}{256}\right)^2 \\ = 1.97972$$

Third central moment

$$\mu'_3 = \mu'_3 - 3\mu'_1 \mu'_2 + 2\mu'^3_1$$

$$= \frac{-37}{256} - 3\left(\frac{507}{256}\right)\left(\frac{-7}{256}\right) + 2\left(\frac{-7}{256}\right)^3 = 0.01789 \quad \underline{\text{Ans.}}$$

(33) Example, → Find out Kurtosis of the data given below

Class interval :	0 - 10	10 - 20	20 - 30	30 - 40
Frequency :	1	3	4	2

Sol.	Class	$f_i^o$	Mid value $x_i^o$	$x_i^o - 25$	$f_i^o(x_i^o - 25)$	$f_i^o(x_i^o - 25)^2$	$f_i^o(x_i^o - 25)^3$	$f_i^o(x_i^o - 25)^4$
	0 - 10	1	5	-20	-20	400	-8000	160000
	10 - 20	3	15	-10	-30	300	-3000	30000
	20 - 30	4	25	0	0	0	0	0
	30 - 40	2	35	10	20	200	2000	20000
		10			-30	900	-9000	210000

Let us assumed mean  $a = 25$

$$\text{since } \mu'_1 = \frac{\sum f_i^o (x_i^o - a)^1}{\sum f_i^o}$$

$$\text{then } \mu'_1 = \frac{\sum f_i^o (x_i^o - 25)}{\sum f_i^o} = \frac{-30}{10} = -3$$

$$\mu'_2 = \frac{\sum f_i^o (x_i^o - 25)^2}{\sum f_i^o} = \frac{900}{10} = 90$$

$$\mu'_3 = \frac{\sum f_i^o (x_i^o - 25)^3}{\sum f_i^o} = \frac{-9000}{10} = -900$$

$$\mu'_4 = \frac{\sum f_i^o (x_i^o - 25)^4}{\sum f_i^o} = \frac{210000}{10} = 21000$$

$$\text{Now } \mu_2 = \mu'_1 - \mu'_1^2 = 90 - (-3)^2 = 81$$

$$\mu_3 = \mu'_3 - 3\mu'_1 \mu'_2 + 2\mu'_1^3 = -900 - 3(90)(-3) + 2(-3)^3 = -144,$$

$$\mu_4 = \mu'_4 - 4\mu'_1 \mu'_3 + 6\mu'_1 \mu'_2 - 3\mu'_1^4$$

$$= 21000 - 4(-900)(-3) + 6(90)(-3)^2 - 3(-3)^4$$

$$= 14817$$

$$\text{Kurtosis } \beta_2 = \frac{\mu_4}{\mu_2^2} = \frac{14817}{(81)^2} = 2.250$$

$$\text{Also } \gamma_2 = \beta_2 - 3 = -0.742.$$

(34)

Binomial distribution:  $\rightarrow$  A random variable  $X$  is said to have a binomial distribution if the discrete density function of  $X$  is given by

$$f_X(x) = \begin{cases} nCx p^x q^{n-x} & \text{for } x=0, 1, 2, \dots, n \\ 0 & \text{otherwise} \end{cases}$$

Here  $n$  is a positive integer  $0 < p < 1$ , and  $q = 1 - p$ . We write  $X \sim B(n, p)$ , where  $n$  and  $p$  are two parameters.

Noted that  $\sum_{x=0}^n f_X(x) = \sum_{x=0}^n nCx p^x q^{n-x} = (p+q)^n = 1$

Theorem:  $\rightarrow$  If a random variable  $X$  has a binomial distribution, then  $E[X] = np$ ,  $\text{Var}[X] = npq$  and  $M_X(t) = (q + p \cdot e^t)^n$

Sol.:  $\rightarrow$  The mgf of  $X$  is given by

$$\begin{aligned} M_X(t) &= E[e^{tx}] = \sum_{x=0}^n e^{tx} \cdot f(x) = \sum_{x=0}^n e^{tx} \cdot nCx p^x q^{n-x} \\ &= \sum_{x=0}^n nCx (pe^t)^x q^{n-x} \\ &= {}^n C_0 (pe^t)^0 \cdot q^n + {}^n C_1 (pe^t)^1 \cdot q^{n-1} + {}^n C_2 (pe^t)^2 \cdot q^{n-2} + \dots + {}^n C_n (pe^t)^n q^0 \end{aligned}$$

$$\Rightarrow M_X(t) = (pe^t + q)^n \quad \{ \text{By using } (a+b)^n = {}^n C_0 a^n + {}^n C_1 a^{n-1} b + \dots + {}^n C_n b^n \}$$

$$\text{Now } \frac{d}{dt} \{M_X(t)\} = n(p e^t + q)^{n-1} \cdot p \cdot e^t$$

$$\frac{d^2}{dt^2} \{M_X(t)\} = n \beta [(n-1)(pe^t + q)^{n-2} \cdot pe^t + (pe^t + q)^{n-1} \cdot e^t]$$

$$\text{At } t=0, \mu'_1 = \left. \frac{d}{dt} \{M_X(t)\} \right|_{at \ t=0} = n \beta (p+q)^{n-1} = np$$

$$\mu''_2 = \left. \frac{d^2}{dt^2} \{M_X(t)\} \right|_{at \ t=0} = n \beta [(n-1) \cdot p + 1] = n(n-1)p^2 + np$$

$$\text{Then } \boxed{E[X] = \mu'_1 = np}$$

$$\text{Var}[X] = \mu''_2 - \mu'^2_1 = n^2 p^2 - np^2 + np - (np)^2 = np(1-p) = npq$$

$$\Rightarrow \boxed{\text{Var}[X] = npq}$$

Remark: ① Binomial distribution is given by  $(p+q)^n$

② If we take  $n=1$  in Binomial distribution, then random variable  $X$  is said to have Bernoulli distribution.

Its discrete density function is given by

$$f_X(x) = \begin{cases} p^x q^{1-x} & \text{for } x=0, 1 \\ 0 & \text{otherwise} \end{cases}$$

(35) Mode of Binomial distribution,  $\rightarrow$  mode is the value of the binomial variate for which the probability is maximum. If  $x$  is the modal value, then  $(n+1)p-1 \leq x \leq (n+1)p$

Case (i)  $\rightarrow$  let  $(n+1)p$  be an integer, then there are two modes  $(n+1)p$  and  $(n+1)p-1$ .

In this case the distribution is said to be bi-modal.

Case (ii)  $\rightarrow$  let  $(n+1)p$  be a fraction

The mode is the integral part of  $(n+1)p$ .

Example  $\rightarrow$  The mean and variance of binomial distribution are 4 and  $\frac{4}{3}$  respectively. Find  $P[X \geq 1]$ .

Sol.  $\rightarrow$  We are given that  $E[X] = np = 4$  and  $\text{var}(X) = npq = \frac{4}{3}$   
 $\Rightarrow npq = \frac{4}{3}$  or  $4q = \frac{4}{3} \Rightarrow q = \frac{1}{3}$  so  $p = 1 - \frac{1}{3} = \frac{2}{3}$   
 and  $n \cdot \frac{2}{3} = 4 \Rightarrow n = 6$ , Hence  $P[X \geq 1] = 1 - P[X=0] = 1 - 6 \cdot \left(\frac{1}{3}\right)^6$   
 $\Rightarrow P[X \geq 1] = 1 - \left(\frac{1}{3}\right)^6 = \frac{728}{729}$

Example  $\rightarrow$  Determine the binomial distribution for which the mean is 4 and variance 3, and find its mode.

Sol.  $\rightarrow$  We are given that  $E[X] = np = 4$  and  $\text{var}(X) = npq = 3$   
 $\Rightarrow npq = 3$  or  $4q = 3 \Rightarrow q = \frac{3}{4}$  so  $p = \frac{1}{4}$  and  $n \cdot \frac{1}{4} = 4 \Rightarrow n = 16$ .

Hence the binomial distribution is  $(p+q)^n = \left(\frac{3}{4} + \frac{1}{4}\right)^n$ .

Now, we have  $(n+1)p = 17 \times \frac{1}{4} = 4.25$

Mode of binomial distribution is integral part of  $(n+1)p = 4$

Example  $\rightarrow$  If the sum and the product of the mean and variance of a binomial distribution are 24 and 128. Find the binomial distribution.

Sol.  $\rightarrow$  We are given  $np + npq = 24 \quad \text{--- (1)}$   $\Rightarrow np = \frac{24}{1+q}$   
 $\text{and } np \cdot (npq) = 128 \quad \text{--- (2)}$   $\Rightarrow n^2 p^2 = \frac{128}{q}$   
 $\Rightarrow \left(\frac{24}{1+q}\right)^2 = \frac{128}{q} \Rightarrow \frac{24^2 \times 24}{(1+q)^2} = \frac{128^2}{q} \Rightarrow 9q = 2(1+q)^2$   
 $\Rightarrow 2q^2 - 5q + 2 \Rightarrow (2q-1)(q-2) = 0 \Rightarrow q = \frac{1}{2} \text{ or } q = 2 \times \text{(Not Possible)}$   
 since  $q = \frac{1}{2}$  so  $p = \frac{1}{2} \Rightarrow n \cdot \frac{1}{2} = \frac{24}{1+\frac{1}{2}} \Rightarrow n = 32$

Hence the binomial distribution is  $(p+q)^n = \left(\frac{1}{2} + \frac{1}{2}\right)^{32}$ .

Example  $\rightarrow$  Six dice are thrown 729 times. How many times do you expect at least three dice to show a 5 or 6.

Sol.  $\rightarrow$  Here  $p = \frac{2}{6} = \frac{1}{3}$   $q = 1 - \frac{1}{3} = \frac{2}{3}$  &  $n = 6$ .

Then  $P[X=x] = {}^6C_x p^x q^{n-x} = {}^6C_x \left(\frac{1}{3}\right)^x \left(\frac{2}{3}\right)^{6-x} = \frac{{}^6C_x \cdot 2^{6-x}}{3^6}$

$$\begin{aligned} \Rightarrow P[\text{At least three success}] &= P[X=3] + P[X=4] + P[X=5] + P[X=6] \\ &= \frac{{}^6C_3 \cdot 2^3}{3^6} + \frac{{}^6C_4 \cdot 2^2}{3^6} + \frac{{}^6C_5 \cdot 2^1}{3^6} + \frac{{}^6C_6 \cdot 2^0}{3^6} = \frac{1}{36} \left[ \frac{6 \cdot 5 \cdot 4 \cdot 3 \cdot 2 \cdot 1}{4 \cdot 3 \cdot 2 \cdot 1} + 6 \cdot 5 \cdot 4 \cdot 3 \cdot 4 + 6 \cdot 2 + 1 \right] \\ &= \frac{233}{729} \end{aligned}$$

Ansl.

(36) Hence the number of times at least 3 success occur  
 $= \frac{729}{729} \times \frac{233}{729} = 233.$

Example,  $\rightarrow$  If on an average 1 vessel in every 10 is wrecked, find the probability that out of 5 vessels expected to arrive at least will arrive safely.

Sol.  $\rightarrow$  Let  $P$  = Probability that vessel to be safe  $= 1 - \frac{1}{10} = \frac{9}{10}$

Also  $n=5$ ,  $P[X=x] = {}^5C_x p^x q^{5-x} = {}^5C_x \left(\frac{9}{10}\right)^x \left(\frac{1}{10}\right)^{5-x} = {}^5C_x q^x$

$$P[\text{At least } x \text{ will arrive safely}] = P[X=4] + P[X=5] \\ = \frac{{}^5C_4 \cdot 9^4}{10^5} + \frac{{}^5C_5 \cdot 9^5}{10^5} = \frac{9^4}{10^5} [{}^5C_4 + {}^5C_5] = 0.91854.$$

Note,  $\rightarrow$  If  $X \sim B(n_1, p)$  and  $Y \sim B(n_2, p)$  are two independent binomial variates, show that  $X+Y \sim B(n_1+n_2, p)$ .

Example,  $\rightarrow$  If  $X$  &  $Y$  are independent binomial variates  $B(5, \frac{1}{2})$  and  $B(7, \frac{1}{2})$ . Find  $P[X+Y=3]$ .

Sol.  $\rightarrow$  Since  $X \sim B(5, \frac{1}{2})$  and  $Y \sim B(7, \frac{1}{2})$  are independent binomial variate then  $X+Y \sim B(5+7, \frac{1}{2})$

Thus  $Z = X+Y$  is a binomial variate with  $n=12$ ,  $p=\frac{1}{2}$ ,  $q=\frac{1}{2}$

$$P[X+Y=3] = {}^{12}C_3 p^3 q^9 = \frac{12 \cdot 11 \cdot 10}{3 \cdot 2 \cdot 1} \left(\frac{1}{2}\right)^3 \left(\frac{1}{2}\right)^9 = 55 \left(\frac{1}{2}\right)^{12}.$$

Example,  $\rightarrow$  Let  $X$  be a  $B(2, p)$  and  $Y$  be a  $B(4, p)$ . If  $P[X \geq 1] = \frac{5}{9}$  find  $P[Y \geq 1]$ .

Sol.  $\rightarrow$  We have  $P[X \geq 1] = 1 - P[X=0] = 1 - 2p^2 q^2 = 1 - q^2$  ( $\because n=2$ )  
 $\therefore P[X \geq 1] = \frac{5}{9} \Rightarrow 1 - q^2 = \frac{5}{9} \Rightarrow q^2 = \frac{4}{9} \Rightarrow q = \frac{2}{3} \text{ & } p = 1 - \frac{2}{3} = \frac{1}{3}$

$$\text{Hence } P[Y \geq 1] = 1 - P[Y=0] = 1 - 4p^4 q^4 = 1 - q^4 (\because n=4) \\ = 1 - \left(\frac{2}{3}\right)^4 = \frac{65}{81}.$$

Example,  $\rightarrow$  If  $X$  and  $Y$  are two independent binomial variates with parameters  $n_1=6$ ,  $p=\frac{1}{2}$  and  $n_2=4$ ,  $p=\frac{1}{2}$  respectively.

Evaluate (i)  $P[X+Y=8]$  (ii)  $P[X+Y \geq 3]$ .

Example,  $\rightarrow$  For a binomial distribution, show that

$$\mu_{r+1} = pq \left[ nr \mu_r + \frac{d \mu_r}{dp} \right]$$

and deduce the values of  $\mu_2$ ,  $\mu_3$  and  $\mu_4$ . Hence find  $\beta$  and  $\gamma$  coefficients for binomial distribution.

Example,  $\rightarrow$  A perfect cubic die is thrown a large number of times in sets of 8, the occurrence of 5 or 6 is called a success. In what proportion of the sets would you expect 3 success?

Ans: Required proportion  $= 0.2731 = 27.31\%$ .

(37) Poisson distribution  $\rightarrow$  A random variable  $X$  is said to have a Poisson distribution if the discrete density function of  $X$  is given by

$$f_X(x) = \begin{cases} \frac{e^{-d} d^x}{x!} & \text{for } x=0, 1, 2, \dots \text{ where } d>0 \\ 0 & \text{otherwise} \end{cases}$$

The variable  $X$  is called Poisson variate, denoted as  $X \sim P(d)$ .

Remark  $\rightarrow$  ① If  $X \sim P(d)$ , then the probability of  $x$  success is given by

$$P[X=x] = \frac{e^{-d} d^x}{x!}$$

In particular,  $P[X=0] = e^{-d}$  and  $P[X=1] = d e^{-d}$

$$\textcircled{2} \quad \sum_{x=0}^{\infty} \frac{e^{-d} d^x}{x!} = e^{-d} \sum_{x=0}^{\infty} \frac{d^x}{x!} = e^{-d} \cdot e^d = 1.$$

Theorem  $\rightarrow$  If  $X$  is a Poisson distributed random variable, then  $E[X]=d$ ,  $\text{Var}[X]=d$  and  $\text{M}_X(t)=e^{d(e^t-1)}$ .

Sol.  $\rightarrow$  Moment generating function of  $X$  is given by

$$\begin{aligned} M_X(t) &= E[e^{tx}] = \sum_{x=0}^{\infty} e^{tx} f_X(x) = \sum_{x=0}^{\infty} e^{tx} \frac{e^{-d} d^x}{x!} \\ &= e^{-d} \sum_{x=0}^{\infty} \frac{(det)^x}{x!} = e^{-d} [1 + (det) + \frac{(det)^2}{2!} + \frac{(det)^3}{3!} + \dots] \\ &= e^{-d} e^{det} = e^{d(e^t-1)} \quad \Rightarrow \boxed{M_X(t) = e^{d(e^t-1)}} \end{aligned}$$

$$\text{Now } \frac{d}{dt} M_X(t) = e^{d(e^t-1)} \cdot det$$

$$\frac{d^2}{dt^2} M_X(t) = d [e^{d(e^t-1)} \cdot det \cdot et + e^{d(e^t-1)} \cdot et]$$

$$\text{At } t=0 \quad \mu'_1 = \frac{d}{dt} M_X(t) \Big|_{at t=0} = d$$

$$\mu''_1 = \frac{d^2}{dt^2} M_X(t) \Big|_{at t=0} = d[d+1] = d^2+d$$

$$\text{Thus } E[X] = \mu'_1 = d \text{ and } \text{Var}(X) = \mu''_1 - \mu'^2_1 = d^2+d-d^2 = d$$

$$\Rightarrow \boxed{E(X) = \text{Var}(X) = d}$$

(38)

Theorem → Poisson distribution as a limiting case of Binomial distribution →

Derive Poisson distribution as the limiting case of Binomial distribution, where  $n \rightarrow \infty$ ,  $p \rightarrow 0$  and  $np = \lambda$  (a finite constant).

Proof → For a Binomial distribution, we have

$$P[X=x] = {}^n C_x p^x q^{n-x} \text{ and } P[0] = q^n$$

$$\text{Also, } \lim_{n \rightarrow \infty} P[0] = \lim_{n \rightarrow \infty} (1-p)^n = \lim_{n \rightarrow \infty} \left(1 - \frac{d}{n}\right)^n = e^{-\lambda} = 0$$

$$\begin{aligned} \text{Now } \frac{P[x+1]}{P[x]} &= \frac{{}^n C_{x+1} p^{x+1} q^{n-x-1}}{{}^n C_x p^x q^{n-x}} && \left\{ \text{using } \lim_{n \rightarrow \infty} \left(1 + \frac{d}{n}\right)^n = e^d \right\} \\ &= \frac{n-x}{x+1} \cdot \frac{p}{q} = \frac{n-x}{x+1} \cdot \frac{\frac{\lambda}{n}}{1 - \frac{\lambda}{n}} = \frac{1 - \frac{x}{n}}{1 - \frac{\lambda}{n}} \cdot \frac{d}{x+1} \end{aligned}$$

$$\lim_{n \rightarrow \infty} \frac{P[x+1]}{P[x]} = \lim_{n \rightarrow \infty} \frac{1 - \frac{x}{n}}{1 - \frac{\lambda}{n}} \cdot \frac{\lambda}{x+1} = \frac{\lambda}{x+1}$$

$$\Rightarrow \frac{P[x]}{P[x-1]} = \frac{\lambda}{x} \quad (2)$$

$$\text{Now } P[x] = \frac{P[x]}{P[x-1]} \cdot \frac{P[x-1]}{P[x-2]} \cdot \frac{P[x-2]}{P[x-3]} \cdots \frac{P[1]}{P[0]} \cdot P[0]$$

Taking  $\lim n \rightarrow \infty$  and using (1) & (2)

$$\lim_{n \rightarrow \infty} P[x] = \frac{\lambda}{x} \cdot \frac{\lambda}{x-1} \cdot \frac{\lambda}{x-2} \cdots \frac{\lambda}{1} \cdot e^{-\lambda} = \frac{e^{-\lambda} \cdot \lambda^x}{x!}$$

Hence the probability of  $x$  success is given by

$$P[X=x] = \frac{e^{-\lambda} \cdot \lambda^x}{x!} \quad x=0, 1, 2, \dots$$

Theorem → Show that for the Poisson distribution with  $\lambda$   $\mu_{r+1} = r\lambda \mu_{r-1} + \lambda \frac{d\mu_r}{d\lambda}$ . Hence deduce the value of  $\mu_2$ ,  $\mu_3$  and  $\mu_4$ . Also find  $\beta$  and  $\gamma$ -coefficient of the Poisson distribution.

Proof → We have  $\mu_r = E[(X-\lambda)^r] = \sum_{x=0}^{\infty} (x-\lambda)^r \frac{e^{-\lambda} \lambda^x}{x!}$

$$\Rightarrow \mu_r = \sum_{x=0}^{\infty} \frac{1}{x!} [e^{-\lambda} \lambda^x (x-\lambda)^r]$$

$$\begin{aligned} \Rightarrow \frac{d\mu_r}{d\lambda} &= \sum_{x=0}^{\infty} \frac{1}{x!} [e^{-\lambda} (\lambda-\lambda)^r \lambda^x + x \lambda^{x-1} e^{-\lambda} (\lambda-\lambda)^{r-1} - r(\lambda-\lambda)^{r-1} e^{-\lambda} \lambda^x] \\ &= \sum_{x=0}^{\infty} \frac{1}{x!} [e^{-\lambda} \lambda^{x-1} (\lambda-\lambda)^r - r(\lambda-\lambda)^{r-1} e^{-\lambda} \lambda^x] \end{aligned}$$

$$= \frac{1}{\lambda} \sum_{x=0}^{\infty} (\lambda-\lambda)^{r+1} e^{-\lambda} \frac{\lambda^x}{x!} - r \sum_{x=0}^{\infty} (\lambda-\lambda)^{r-1} e^{-\lambda} \frac{\lambda^x}{x!}$$

$$= \frac{1}{\lambda} \mu_{r+1} - r \mu_{r-1} \Rightarrow \boxed{\mu_{r+1} = r\lambda \mu_{r-1} + \lambda \frac{d\mu_r}{d\lambda}} \quad (1)$$

(39) Put  $\lambda = 1, 2$  & (3) in (1) we get

$$\mu_2 = \lambda \mu_0 + \lambda \frac{d \mu_1}{d \lambda} = \lambda \quad (\because \mu_0 = 1, \mu_1 = 0)$$

$$\mu_3 = 2\lambda \mu_1 + \lambda \frac{d \mu_2}{d \lambda} = 0 + \lambda \cdot 1 = \lambda$$

$$\mu_4 = 3\lambda \mu_2 + \lambda \frac{d \mu_3}{d \lambda} = 3\lambda^2 + \lambda$$

$$\beta_1 = \frac{\mu_3^2}{\mu_2^2} = \frac{\lambda^2}{\lambda^3} = \frac{1}{\lambda} \quad \text{and } \beta_2 = \frac{\mu_4}{\mu_2^2} = \frac{3\lambda^2 + \lambda}{\lambda^2} = 3 + \frac{1}{\lambda}$$

$$\gamma_1 = \sqrt{\beta_1} = \frac{1}{\sqrt{\lambda}} \quad \gamma_2 = \beta_2 - 3 = \frac{1}{\lambda}$$

Example → If  $X$  is a Poisson variate such that  $P[X=1] = 2P[X=2]$ . Find mean and variance.

Also find  $P[X=0]$ .

Sol. → It is given that  $P[X=1] = 2P[X=2]$

$$\frac{\lambda e^{-\lambda}}{1!} = 2 \frac{\lambda^2}{2!} e^{-\lambda} \Rightarrow \boxed{\lambda = 1}$$

Hence mean = variance = 1 &  $P[X=0] = \frac{1 \cdot e^{-1}}{0!} = e^{-1}$

Example → If  $X$  is a Poisson variate such that

$$P[X=2] = 9P[X=4] + 90P[X=6] \quad (i) \text{ Find } E[X].$$

(ii)  $\beta_1$ , the coefficient of skewness.

Sol. → (i) It is given that  $P[X=2] = 9P[X=4] + 90P[X=6]$

$$\Rightarrow \frac{\bar{e}^\lambda \cdot \lambda^2}{2!} = 9 \cdot \frac{\bar{e}^\lambda \lambda^4}{4!} + 90 \cdot \frac{\bar{e}^\lambda \lambda^6}{6!}$$

$$\frac{1}{2} = \frac{9 \cdot \lambda^2}{4 \cdot 3 \cdot 2} + \frac{90 \cdot \lambda^4}{6 \cdot 5 \cdot 4 \cdot 3 \cdot 2} \Rightarrow \lambda^4 + 3\lambda^2 - 4 = 0$$

$$\Rightarrow (\lambda^2 + 4)(\lambda^2 - 1) = 0 \Rightarrow \lambda^2 - 1 = 0 \Rightarrow \boxed{\lambda = 1}$$

$$\text{Hence } \boxed{E[X] = \lambda = 1}$$

(ii)  $\beta_1 = \text{coefficient of skewness} = \frac{1}{\lambda} = 1$

Ex → Show that in a Poisson distribution with unit mean, mean deviation about mean is  $(\frac{2}{e})$  times the standard deviation.

Sol. → Here  $\lambda = 1$  so that  $\sigma_x = \sqrt{\lambda} = 1$   
Poisson distribution  $P[X=x] = \frac{\lambda^x e^{-\lambda}}{x!} = \frac{e^{-1}}{x!}, x=0, 1, 2, \dots$

Mean deviation about mean is

$$M.D = E[|X-1|] = \sum_{x=0}^{\infty} |x-1| \cdot \frac{e^{-1}}{x!} = |1| \frac{e^{-1}}{0!} + |1-1| \frac{e^{-1}}{1!} + \sum_{x=2}^{\infty} |x-1| \cdot \frac{e^{-1}}{x!}$$

$$= e^{-1} + \sum_{x=2}^{\infty} (x-1) \frac{e^{-1}}{x!} = e^{-1} + e^{-1} \sum_{x=2}^{\infty} \left\{ \frac{1}{(x-1)!} - \frac{1}{x!} \right\} = e^{-1} + e^{-1} = \frac{2}{e}$$

$$M.D = \frac{2}{e} \times 1 = \frac{2}{e} \sigma_x = \frac{2}{e} \times S.D$$

Ex → An insurance company insures 4000 people against loss of both eyes in a car accident based on previous data, the rates were computed on the assumption that on the average 10 persons in 100000 will have car accident each year that result in this type of injury. What is the probability that more than 3 of the injured will collect on their policy in a given year? [Given  $e^{-0.4} = 0.6703$ ].

Sol. → Given  $n = 4000, p = \text{prob. of loss of both eyes in a car accident} = \frac{10}{100000} = 0.0001$ ,

$$\lambda = np = 4000 \times 0.0001 = 0.4$$

Then by Poisson distribution  $P[X=x] = \frac{e^{-\lambda} \lambda^x}{x!} = \frac{e^{-0.4} \cdot (0.4)^x}{x!}$

Prob. [More than 3 of the injured will collect on their policy in a given year]  $x=0, 1, 2, 3, \dots$

$$= P[X > 3] = 1 - \{ P(X=0) + P(X=1) + P(X=2) + P(X=3) \}$$

$$= 1 - e^{-0.4} \left\{ \frac{(0.4)^0}{0!} + \frac{(0.4)^1}{1!} + \frac{(0.4)^2}{2!} + \frac{(0.4)^3}{3!} \right\}$$

$$= 1 - e^{-0.4} (1 + 0.4 + 0.08 + 0.0107) = 1 - 0.6703 \times 1.4907 \\ = 0.0008,$$

Ex → In a book of 520 pages, 390 typographical error occurs. Assuming Poisson law for the number of errors per page, find the prob. that a random sample of 5 pages will contain no error.

Sol. → The average no. of typographical error per page in book  $= \lambda = \frac{390}{520} = 0.75$ .

Then by Poisson distribution  $P[X=x] = \frac{e^{-\lambda} \lambda^x}{x!} = \frac{e^{-0.75} \cdot (0.75)^x}{x!}$

$$P[5 \text{ pages contain no error}] = \{P(X=0)\}^5 = (e^{-0.75})^5 = e^{-3.75}$$

Ans.

(41)

Example → A manufacturer, who produces machine bottles, find that 0.1% of the bottles are defective. The bottles are packed in boxes containing 500 bottles. If the manufacturer buys 100 boxes from the producer of bottle. Using Poisson distribution. Find how many boxes will contain

(i) No defective (ii) At least two defectives {Given  $e^{-0.5} = 0.6065$ }

Sol. → Here  $n = 100$ ,  $n = 500$ ,  $p = 0.1\% = 0.001$  so  $\lambda = np$   
 $\lambda = 500 \times 0.001 = 0.5$

By Poisson distribution. Prob. [x defective bottles in box]

$$\text{i.e. } P[X=x] = \frac{e^{-0.5} \cdot (0.5)^x}{x!}$$

Hence in a consignment of 100 boxes, the number of boxes containing x defective bottles =  $f(x) = n \cdot P[X=x]$

$$= 100 \times \frac{e^{-0.5} \times (0.5)^x}{x!}$$

(i) No. of boxes containing no defective bottles

$$= 100 \times P[X=0] = 100 \times \frac{e^{-0.5} \times (0.5)^0}{0!} = 100 \times 0.6065 = 60.65 \approx 61.$$

(ii) No. of boxes containing at least two defective bottles

$$= 100 \times P[X \geq 2] = 100 \times \{1 - P(X=0) - P(X=1)\} = 100 \times \left\{1 - \frac{e^{-0.5} (0.5)^0}{0!} - \frac{e^{-0.5} (0.5)^1}{1!}\right\}$$

$$= 100 \left[1 - 0.6065 - 0.6065 \times 0.5\right] = 100 \times 0.0925 = 9.25 \approx 9. \text{ Ans.}$$

Example → Six coins are tossed 6,400 times. Using the Poisson distribution. Find the approximate probability of getting six heads 8 times.

Sol. → Here  $n = 6400$ ,  $P$  = Prob. of getting six heads in one throw of six coins =  $\left(\frac{1}{2}\right)^6$ .

$$\lambda = np = 6400 \times \left(\frac{1}{2}\right)^6 = 100.$$

By Poisson distribution,

$$P[\text{getting 6 heads 8 times}] = P[X=8] = \frac{e^{-\lambda} \lambda^8}{8!} = \frac{e^{-100} \cdot (100)^8}{8!}$$

Example → In a Poisson frequency distribution, frequency corresponding to 3 success is  $\frac{1}{2}$  times frequency corresponding to 2 success. Find the mean and standard deviation of the distribution.

Sol. → The frequency function is given by  $f(x) = N \cdot P(x) = Nx \cdot P(X=x)$

$$= Nx \frac{e^{-\lambda} \lambda^x}{x!}, \quad x=0, 1, 2 \text{ putting } x=3 \text{ and } x=4$$

$$f(3) = Nx \frac{e^{-\lambda} \lambda^3}{3!} \quad \& \quad f(4) = Nx \frac{e^{-\lambda} \lambda^4}{4!}$$

$$f(3) = \frac{2}{3} f(4) \Rightarrow Nx \frac{e^{-\lambda} \lambda^3}{3!} = \frac{2}{3} Nx \frac{e^{-\lambda} \lambda^4}{4!} \Rightarrow \frac{1}{6} = \frac{2}{3} \times \frac{\lambda}{24} \Rightarrow \lambda = 4$$

$$\Rightarrow \boxed{\lambda = 6} \quad \boxed{\text{mean} = 6} \quad \& \quad \boxed{\text{variance} = 6}$$

$$\boxed{S.D = \sqrt{6}}$$

(42)

Note → ① Mode of the Poisson distribution with parameter  $\lambda$  is given by  $\lambda - 1, \lfloor \lambda \rfloor$ .

Case (i) → If  $\lambda$  is a positive integer then  $\lambda - 1$  and  $\lambda$  are two modes of Poisson distribution. In this case the Poisson distribution is said to be bimodal.

Case (ii) → If  $\lambda$  is a fraction, then mode of P.D is the integral part of  $\lambda$ .

Example → A Poisson distribution has a double mode at  $x=1$ , and  $x=2$ . What is the probability that  $x$  will have one or the other of these two values?

Sol. → If the Poisson distribution is bimodal, then the two modes are at the points  $x=\lambda - 1$  and  $x=\lambda$ , where  $\lambda$  is the parameter of the Poisson distribution. Therefore, since we are given that the two modes are at the points  $x=1$  &  $x=2$ , we find that  $\lambda=2$ .

$$P[x=x] = \frac{e^{-\lambda} \lambda^x}{x!} = \frac{e^{-2} 2^x}{x!} \quad x=0, 1, 2, \dots$$

$$P[x=1] = e^{-2} \cdot 2 \quad \text{and} \quad P[x=2] = \frac{e^{-2} \cdot 2^2}{2!} = 2e^{-2}$$

$$\text{Required probability} = P[x=1] + P[x=2] = 2e^{-2} + 2e^{-2} = 4e^{-2}$$

Example → If  $x$  and  $y$  are independent Poisson variate such that  $P[x=1] = P[x=2]$  and  $P[y=2] = P[y=3]$ . Find the variance of  $x - 2y$ .

Sol. → Let  $x \sim P(\lambda)$  and  $y \sim P(\mu)$

$$\text{then } P(x=x) = \frac{e^{-\lambda} \lambda^x}{x!} \quad x=0, 1, 2, \dots \quad \lambda > 0$$

$$\text{and } P(y=y) = \frac{e^{-\mu} \mu^y}{y!} \quad y=0, 1, 2, \dots \quad \mu > 0$$

Using given condition,

$$P[x=1] = P[x=2] \quad \text{and} \quad P[y=2] = P[y=3]$$

$$\frac{\lambda e^{-\lambda}}{1!} = \frac{\lambda^2 e^{-\lambda}}{2!} \quad \text{and} \quad \frac{\mu^2 e^{-\mu}}{2!} = \frac{\mu^3 e^{-\mu}}{3!}$$

$$\Rightarrow \boxed{\lambda=2}$$

$$\Rightarrow \boxed{\mu=3}$$

$$\text{Now } \text{Var}(x) = 2,$$

$$\text{Var}(y) = 3.$$

$$\therefore \text{Var}(x - 2y) = 1^2 \text{Var}(x) + (-2)^2 \text{Var}(y)$$

$$= 1 \cdot 2 + 4 \cdot 3 = 14 \quad \underline{\underline{\text{Ans.}}}$$

(43) Example, fit a binomial distribution to the following frequency data:

$x$ :	0	1	2	3	4
$f$ :	30	62	46	10	2

Sol. The table is as follows:

$x$	$f$	$f \cdot x$	Mean of observations = $\frac{\sum fx}{\sum f}$
0	30	0	$= \frac{192}{150} = 1.28$
1	62	62	$\Rightarrow np = 1.28$
2	46	92	$\Rightarrow 4f_p = 1.28 \Rightarrow p = 0.32$
3	10	30	$\Rightarrow q = 1 - p = 1 - 0.32 = 0.68$
4	2	8	Also $N = 150 \Rightarrow \sum f = N$
	$\sum f = 150$	$\sum fx = 192$	

Hence, the binomial distribution is  $= N(8 + p)^n$   
 $= 150(0.68 + 0.32)^4$ . Ans.

Example, fit a Binomial distribution of the following data and compare the theoretical frequencies with the actual ones.

$x$ :	0	1	2	3	4	5
$f$ :	2	14	20	34	22	8

Sol. To fit a Binomial distribution means we have to find the Binomial distribution given by  $= N(8 + p)^n$ .

$\Rightarrow$  Mean of the Binomial distribution  $= np = \frac{\sum f_i x_i}{\sum f_i}$

$$\Rightarrow np = \frac{0 + 14 + 40 + 132 + 88 + 40}{2 + 14 + 20 + 34 + 22 + 8} = \frac{284}{100} = 2.84$$

$$\Rightarrow np = 2.84, \Rightarrow p = \frac{2.84}{n} = \frac{2.84}{5} = 0.568$$

$$\text{Also, } q = 1 - p = 1 - 0.568 = 0.432.$$

Hence the Binomial distribution  $= N(8 + p)^n = 100(0.432 + 0.568)^5$ . Further, the Binomial distribution is given by

$$P(x=r) = N \cdot nCr \cdot q^{n-r} \cdot p^r, r=0, 1, 2, 3, 4, 5.$$

The theoretical frequencies are given by putting  $r=0, 1, 2, 3, 4, 5$ .

$$\text{when } r=0, P(x=0) = 100 \times 5C_0 (0.432)^0 (0.568)^0 = 100 \times 1 \times (0.0150) = 1.5045.$$

$$\text{when } r=1, P(x=1) = 100 \times 5C_1 (0.432)^1 (0.568)^1 = 100 \times 5 \times (0.0348) \times (0.568)$$

$$= 9.891.$$

$$\text{when } r=2, P(x=2) = 100 \times 5C_2 (0.432)^2 (0.568)^2 = 100 \times 10 \times (0.0806) \times (0.3226)$$

$$\text{when } r=3, P(x=3) = 100 \times 5C_3 (0.432)^3 (0.568)^3 = 100 \times 10 \times (0.1866) \times (0.1832) = 34.185$$

$$\text{when } r=4, P(x=4) = 100 \times 5C_4 (0.432)^4 (0.568)^4 = 100 \times 5 \times (0.432) \times (0.1040) = 22.484,$$

$$\text{when } r=5, P(x=5) = 100 \times 5C_5 (0.432)^5 (0.568)^5 = 100 \times 1 \times (0.5912) = 5.9120.$$

Tabulating all the above, we have the following Binomial distribution of best fit.

(44)

$x$	Actual frequencies	Theoretical frequencies
0	2	$1.5045 = 2$
1	14	$9.891 = 10$
2	20	$26 = 26$
3	34	$34.185 = 34$
4	22	$22.464 = 22$
5	8	$5.9120 = 6$
	Total = 100	Total = 100

Example → Fit a Poisson distribution to the following data and calculate theoretical frequencies.

(i) $x:$	0	1	2	3	4	f (°) $x:$	0	1	2	3	4
f:	120	60	15	2	1	f:	109	65	22	3	1

Sol" → (i) Mean of given distribution =  $\frac{\sum fx}{\sum f}$   
 $\Rightarrow \lambda = \frac{60 + 30 + 6 + 4}{200} = 0.5$

Required Poisson distribution =  $N \cdot \frac{e^{-\lambda} \cdot \lambda^x}{x!} = 200 \cdot \frac{e^{-0.5} \cdot (0.5)^x}{x!} = (121.306) \cdot \frac{(0.5)^x}{x!}$

$x$	$N \cdot P(x)$	Theoretical frequency
0	$121.306 \cdot \frac{(0.5)^0}{0!} = 121.306$	121
1	$121.306 \cdot \frac{(0.5)^1}{1!} = 60.653$	61
2	$121.306 \cdot \frac{(0.5)^2}{2!} = 15.163$	15
3	$121.306 \cdot \frac{(0.5)^3}{3!} = 2.527$	3
4	$121.306 \cdot \frac{(0.5)^4}{4!} = 0.3159$	0

(ii) Mean of given distribution =  $\frac{\sum fx}{\sum f} = \frac{65 + 44 + 9 + 4}{200} = 0.61$

This is the parameter ( $\lambda$ ) of the Poisson distribution

∴ Required Poisson distribution =  $N \cdot \frac{e^{-\lambda} \cdot \lambda^x}{x!} = 200 \cdot \frac{e^{-0.61} \cdot (0.61)^x}{x!}$   
 $= 108.67 \times \frac{(0.61)^x}{x!}$

$x$	$N \cdot P(x)$	Theoretical frequency
0	$108.67 \times \frac{(0.61)^0}{0!} = 108.67$	109
1	$108.67 \times \frac{(0.61)^1}{1!} = 62.289$	66
2	$108.67 \times \frac{(0.61)^2}{2!} = 20.218$	20
3	$108.67 \times \frac{(0.61)^3}{3!} = 4.0111$	4
4	$108.67 \times \frac{(0.61)^4}{4!} = 0.6269$	1

Total = 200.

(45)

Example → If the probability of a bad reaction from a certain injection is 0.001, determine the chance that out of 1000 individuals more than 2 will be defective.

Sol. → It follows a Poisson distribution as the probability of occurrence is very small.

$$\text{mean}(m) = np = 1000 \times (0.001) = 1.$$

Probability that more than 2 will be defective

$$= 1 - [\text{prob. that no one gets a bad reaction} + \text{prob. that one gets a bad reaction} + \text{prob. that two get a bad reaction}]$$

$$= 1 - \left[ e^{-m} + \frac{m e^{-m}}{1!} + \frac{m^2 e^{-m}}{2!} \right] = 1 - \left[ \frac{1}{e} + \frac{1}{e} + \frac{1}{2e} \right]$$

$$= 1 - \left[ \frac{2+2+1}{2e} \right] = 1 - \frac{5}{2e} = 1 - \frac{5}{2 \times 2.718} = 1 - \frac{5}{5.436}$$

$$= 1 - 0.91979 = 0.080206. \quad \underline{\underline{\text{Ans.}}}$$

### Normal Distribution OR Gaussian Distribution

Defn. → A random variable  $X$ , whose probability density function is given by

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

is called a normal variate with parameters  $\mu$  (called mean) and  $\sigma^2$  (called variance). It is denoted by  $X \sim N(\mu, \sigma^2)$ .

Remark → ① The cumulative distribution function  $X \sim N(\mu, \sigma^2)$  is denoted as  $\Phi(x)$ . Thus

$$\Phi(x) = P[X \leq x] = \int_{-\infty}^x \phi(x) dx$$

② It can be verified that  $\int_{-\infty}^{\infty} \phi(x) dx = 1$ .

③ The curve  $y = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left( \frac{x-\mu}{\sigma} \right)^2}$  is called the normal curve

which passes the following properties:

1. The total area under the standard normal curve is 1.

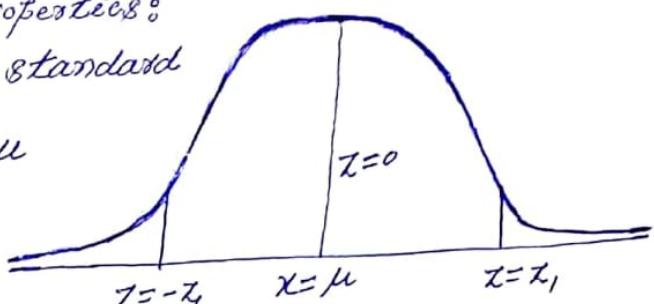
2. The ordinate at mean i.e.  $x = \mu$  divides the area under the standard normal curve into two equal parts.

Symbolically,

$$P[-\infty < Z < 0] = P[0 < Z < \infty] = 0.5.$$

3. Since, the curve is symmetrical

$$\text{thus, } P[0 \leq Z \leq z_1] = P[-z_1 \leq Z \leq 0].$$



(46)

standard normal variate  $\rightarrow$  if  $X \sim N(\mu, \sigma^2)$

then  $Z = \frac{X-\mu}{\sigma}$  is called standard normal variate with  
 $E[Z]=0$  and  $\text{Var}[Z]=1$  and is written as  $Z \sim N(0,1)$

The probability density function (pdf) of standard normal variate  $Z$  is given by  $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2}$ ,  $-\infty < z < \infty$ .

Remark  $\rightarrow$  ① The cumulative distribution function of  $N(0,1)$  is given by  $\Phi(z) = P[Z \leq z] = \int_{-\infty}^z \phi(z) dz$

② Since  $\int_{-\infty}^{\infty} \phi(z) dz = 1$  i.e.  $\int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} dz = 1$ .

$$\Rightarrow \left[ \int_{-\infty}^{\infty} e^{-\frac{1}{2}z^2} dz = \sqrt{2\pi} \right] \quad \left[ \text{Also } \int_{-\infty}^{\infty} e^{-z^2} dz = \sqrt{\pi} \right].$$

Theorem  $\rightarrow$  If  $X \sim N(\mu, \sigma^2)$  then  $E[X] = \mu$ ,  $\text{Var}[X] = \sigma^2$   
&  $M_X(t) = e^{\mu t + \frac{1}{2}\sigma^2 t^2}$ ,

$$\begin{aligned}
 \text{Proof } \rightarrow \text{The m.g.f of } X \text{ is given by } M_X(t) &= E[e^{tx}] = \int_{-\infty}^{\infty} e^{tx} \phi(x) dx \\
 M_X(t) &= \int_{-\infty}^{\infty} e^{tx} \cdot \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{t(\mu+\sigma x)} e^{-\frac{1}{2}\frac{(x-\mu)^2}{\sigma^2}} dx \\
 &= \frac{1}{\sqrt{2\pi}} e^{\mu t} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(x^2 - 2tx + t^2\sigma^2)}. dx \quad \left. \begin{array}{l} \frac{x-\mu}{\sigma} = z \\ \Rightarrow \sigma x + \mu = x \\ \Rightarrow dx = \sigma dz \end{array} \right. \\
 &= \frac{1}{\sqrt{2\pi}} e^{\mu t} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(x^2 - 2\sigma xt + (\sigma t)^2 - (\sigma t)^2)}. dx \\
 &= \frac{1}{\sqrt{2\pi}} e^{\mu t} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(x - \sigma t)^2}. dx = \frac{1}{\sqrt{2\pi}} e^{\mu t + \frac{1}{2}\sigma^2 t^2} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(x - \sigma t)^2} dx. \\
 &= \frac{1}{\sqrt{2\pi}} e^{\mu t + \frac{1}{2}\sigma^2 t^2} \int_{-\infty}^{\infty} e^{-\frac{1}{2}y^2} dy \quad \left. \begin{array}{l} \text{let } x - \sigma t = y \Rightarrow dx = dy \\ \Rightarrow dx = dy \end{array} \right. \\
 &= \frac{1}{\sqrt{2\pi}} e^{\mu t + \frac{1}{2}\sigma^2 t^2} \cdot \sqrt{\pi} = e^{\mu t + \frac{1}{2}\sigma^2 t^2}
 \end{aligned}$$

$$\text{Now } M'_X(t) = e^{\mu t + \frac{1}{2}\sigma^2 t^2} (\mu + \sigma^2 t)$$

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

$$\text{At } t=0, M'_X(t)_{at t=0} = \mu' = \mu \text{ & } M''_X(t)_{at t=0} = \mu_2' = \mu^2 + \sigma^2$$

$$\Rightarrow \text{mean} = E[X] = \mu' = \mu \text{ & } \text{Var}(X) = \mu_2' - \mu'^2$$

$$\Rightarrow \text{Var}(X) = \mu^2 + \sigma^2 - \mu^2 = \sigma^2$$

$$\boxed{E[X] = \mu} \text{ & } \boxed{\text{Var}(X) = \sigma^2} \Rightarrow \boxed{S.D = \sigma}$$

47

Example → Show that the mean deviation from the mean of the general normal distribution is  $\sigma\sqrt{\frac{2}{\pi}}$  or  $\frac{4}{5}\sigma$  (approx.)

Proof → M.D about mean  $\mu = \int_{-\infty}^{\infty} |x - \mu| f(x) dx$

$$\begin{aligned} \Rightarrow \mu &= \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} |x - \mu| e^{-\frac{1}{2}(\frac{x-\mu}{\sigma})^2} dx \\ &= \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} |xz| e^{-\frac{1}{2}z^2} \cdot \sigma dz \quad \left\{ \text{Let } \frac{x-\mu}{\sigma} = z \Rightarrow x = \mu + \sigma z \right. \\ &= \frac{\sigma}{\sqrt{2\pi}} \int_{-\infty}^{\infty} |z| e^{-\frac{1}{2}z^2} dz = \frac{2\sigma}{\sqrt{2\pi}} \int_0^{\infty} z e^{-\frac{1}{2}z^2} dz \quad \left\{ \text{Even function} \right\} \\ &= \frac{2\sigma}{\sqrt{2\pi}} \int_0^{\infty} e^{-t} dt \quad \left\{ \text{Take } \frac{z^2}{2} = t \Rightarrow dz = dt \right\} \\ &= \sqrt{\frac{2}{\pi}} \sigma \left[ \frac{-e^{-t}}{-1} \right]_0^{\infty} = \sqrt{\frac{2}{\pi}} \sigma = \frac{4}{5}\sigma \quad (\text{Approx.}) \end{aligned}$$

Example → If  $x$  is a normal variate with mean 30 and S.D 5. Find the probabilities that

$$(i) 26 \leq x \leq 40 \quad (ii) x \geq 45 \quad (iii) |x - 30| > 5.$$

Sol. → Given  $\mu = 30$ ,  $\sigma = 5$  Let  $z = \frac{x-\mu}{\sigma} = \frac{x-30}{5}$

$$\begin{aligned} (i) P[26 \leq x \leq 40] &= P\left[\frac{26-30}{5} \leq \frac{x-30}{5} \leq \frac{40-30}{5}\right] = P[-0.8 \leq z \leq 2] \\ &= P[-0.8 \leq z \leq 0] + P[0 \leq z \leq 2] = P[-0.8 \leq z \leq 0] + 0.4772 \\ &= P[0 \leq z \leq 0.8] + 0.4772 \quad \left\{ \text{Due to symmetry} \right\} \\ &= 0.2881 + 0.4772 = 0.7653. \end{aligned}$$

$$(ii) P[x \geq 45] = P\left[\frac{x-30}{5} \geq \frac{45-30}{5}\right] = P[z \geq 3] = 0.5 - P[z \leq 3] \\ = 0.5 - 0.49865 = 0.00135$$

$$(iii) \text{ since } P[|x-30| \leq 5] = P[25 \leq x \leq 35] = P\left[\frac{25-30}{5} \leq \frac{x-30}{5} \leq \frac{35-30}{5}\right] \\ = P[-1 \leq z \leq 1] = 2P[0 \leq z \leq 1] = 2 \times 0.3413 = 0.6826$$

$$\Rightarrow P[|x-30| > 5] = 1 - P[|x-30| \leq 5] = 1 - 0.6826 = 0.3174.$$

Example → If  $x$  is normally distributed with mean 11 & S.D 1.5. Find the number  $x_0$  such that  $P[x > x_0] = 0.3$

Sol. → Given  $0.3 = P[x > x_0] = 0.5 - P[11 < x < x_0]$

$$\begin{aligned} &= 0.5 - P\left[\frac{11-x_0}{1.5} < \frac{x_0-11}{1.5} < \frac{x_0-11}{1.5}\right] \\ &= 0.5 - P\left[0 < z < \frac{x_0-11}{1.5}\right] \end{aligned}$$

$$\textcircled{48} \Rightarrow P\left[0 < z < \frac{x_0 - 11}{1.5}\right] = 0.2$$

$$\Rightarrow \frac{x_0 - 11}{1.5} = 0.52 \quad (\text{from table})$$

$$\Rightarrow x_0 = 11 + 1.5 \times 0.52 = 11.78.$$

Example → The life of electric tubes of a certain type may be normally distributed with mean 155 hours & standard deviation 19 hours. What is the prob.

(i) That the life of randomly chosen tube is b/w 136 hours and 174 hours.

(ii) That the life of a randomly chosen tube is less than 117 hours.

(iii) that the total life of two randomly chosen tubes will be more than 395 hours.

Sol. Given  $\mu = 155$ ,  $\sigma = 19$ , let  $z = \frac{x - \mu}{\sigma} = \frac{x - 155}{19}$   
where  $x$  denotes the life of electric.

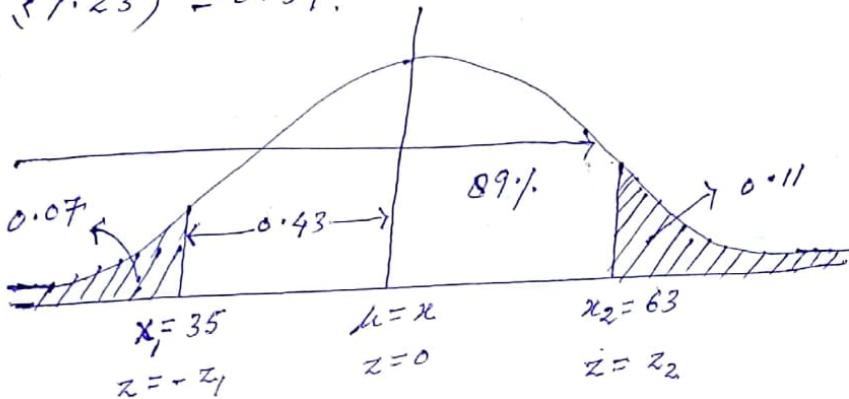
$$\begin{aligned} \text{(i)} \quad P[136 < x < 174] &= P\left[\frac{136 - 155}{19} < \frac{x - 155}{19} < \frac{174 - 155}{19}\right] \\ &= P[-1 < z < 1] \\ &= 2P[0 < z < 1] = 2 \times 0.3413 = 0.6826. \end{aligned}$$

$$\begin{aligned} \text{(ii)} \quad P[x < 117] &= P\left[\frac{x - 155}{19} < \frac{117 - 155}{19}\right] = P[z < -2] \\ &= P[z > 2] \quad \{ \text{By symmetry} \} \\ &= 0.5 - P[0 < z < 2] = 0.5 - 0.4772 \\ &= 0.0228. \end{aligned}$$

$$\begin{aligned} \text{(iii)} \quad P[x > 395] &= P\left[\frac{x - 155}{19} > \frac{395 - 155}{19}\right] \\ &= P[z > 12.63] = 0. \end{aligned}$$

- (47) ~~Ques.~~ Example, ~~in~~
1. In a normal distribution 7% of the items are under 35 and 89% are under 63. Find the mean and s.d of the distribution? Given that  
~~Sol.~~  $P(0 \leq Z \leq 0.18) = 0.07$ ,  $P(0 \leq Z \leq 1.48) = 0.43$ ,

$$P(0 \leq Z \leq 1.23) = 0.39.$$



Let  $\mu$  &  $\sigma$  be the mean & standard deviation of the normal distribution

$$\text{Here } P(X < 35) = 0.07$$

$$\text{when } x = 35, z = \frac{35 - \mu}{\sigma} = z_1 \text{ (say)} \quad \text{--- (1)}$$

$$P(X < 35) = 0.07$$

$$\Rightarrow P(Z < z_1) = 0.07$$

$$\Rightarrow 0.5 - P(z_1 < Z < 0) = 0.07$$

$$\Rightarrow P(0 < Z < z_1) = 0.50 - 0.07 = 0.43$$

$$\Rightarrow \text{or } P(0 < Z < z_1) = 0.43 \text{ corresponds to the value } 1.48 \Rightarrow z_1 = -1.48$$

$$\text{from (1)} \quad 35 - \mu = -1.48\sigma \quad \text{--- (2)}$$

$$\text{again } P(X < 63) = 0.89, \text{ when } x = 63, z = \frac{63 - \mu}{\sigma} = z_2 \text{ (say)} \quad \text{--- (3)}$$

$$\Rightarrow P(Z < z_2) = 0.89$$

$$\Rightarrow 0.5 + P(0 < Z < z_2) = 0.89$$

$$\Rightarrow P(0 < Z < z_2) = 0.89 - 0.5 = 0.39$$

$$\Rightarrow z_2 \text{ corresponds to the value } 1.23, \text{ i.e. } z_2 = 1.23$$

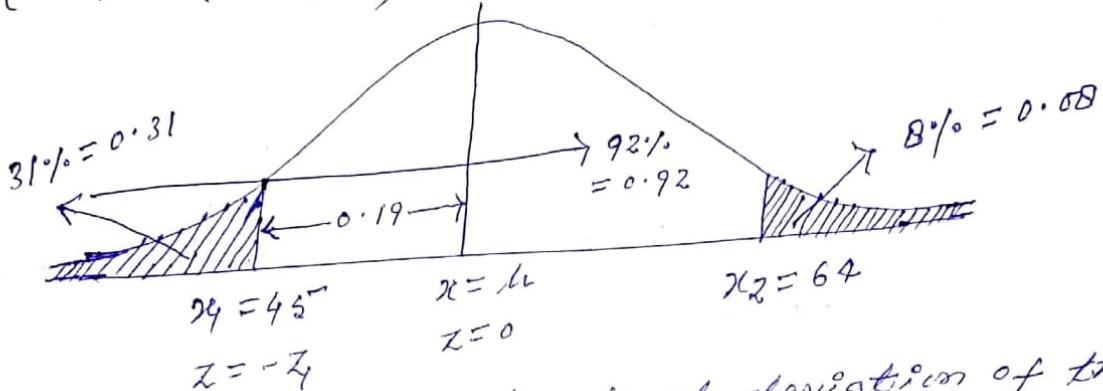
$$\text{from (3)} \quad 63 - \mu = 1.23\sigma \quad \text{--- (4)}$$

$$\text{on solving (2) + (4), we get } \boxed{\sigma = 10.38}$$

$$\therefore \boxed{\mu = 50.3}$$

- ④ Example  
 2. In a normal distribution, 31% of the items are under 45 and 8% are over 64. Find the mean & standard deviation of the distribution.

Given that  $P(0.5z \leq 0.496) = 0.19$   
~~Soln.~~ &  $P(0.5z \leq 1.405) = 0.42$ .



Let  $\mu$  &  $\sigma$  be the mean & standard deviation of the normal distribution when  $x = 45$ ,  $\Rightarrow z = \frac{45 - \mu}{\sigma} = z_1$  (say) — (1)

$$\text{Here } P(x < 45) = 31\% = 0.31, \Rightarrow 0.5 - P(z_1 < z < 0) = 0.31$$

$$P(z < z_1) = 0.31 \Rightarrow 0.5 - P(z_1 < z < 0) = 0.19$$

$$\Rightarrow P(z_1 < z < 0) = 0.5 - 0.31 = 0.19$$

$\Rightarrow P(0 < z < z_1) = 0.19 \Rightarrow z_1$  corresponds to the value 0.496

$$\text{i.e. } z_1 = -0.496, \text{ from } 45 - \mu = -0.496\sigma \quad \text{--- (2)}$$

$$\text{Again } P(x > 64) = 8\% = 0.08 \text{ OR } P(x < 64) = 92\% = 0.92$$

$$\text{when } x = 64 \Rightarrow z = \frac{64 - \mu}{\sigma} = z_2 \text{ (say)} \quad \text{--- (3)}$$

$$\Rightarrow P(x > 64) = 0.92 \Rightarrow 0.5 + P(0 < z < z_2) = 0.92$$

$$\Rightarrow P(0 < z < z_2) = 0.92 - 0.5 = 0.42$$

$\Rightarrow z_2$  corresponds to the value 1.405

$$\text{from (3)} \quad 64 - \mu = 1.405\sigma \quad \text{--- (4)}$$

On solving (2) & (4) we get

$$\boxed{\mu = 50} \quad \& \quad \boxed{\sigma = 10}$$

(51)

Example: → Assuming the mean height of soldiers to be 68.22 inches with a variance of 10.8 inches<sup>2</sup>, find how many soldiers in a regiment of 1000 would you expect to be over 6 feet tall. (Given: Area under the standard normal curve between  $z=0$  &  $z=0.35$  is 0.1368) and between  $z=0$  and  $z=1.15$  is 0.3746)

Sol.: → Given  $\mu = 68.22$ ,  $\sigma^2 = 10.8 \Rightarrow \sigma = 3.206$ , let  $z = \frac{x-\mu}{\sigma}$

$$\Rightarrow z = \frac{x - 68.22}{3.206}, \text{ when } x = 6 \text{ feet} = 72 \text{ inches}$$

$$z = \frac{72 - 68.22}{3.206} = 1.15$$

$$P[X > 72] = P[z > 1.15] = 0.5 - P[0 < z < 1.15] = 0.5 - 0.3746 = 0.1254$$

Hence, the number of soldiers out of 1000 who are over 6 feet tall

$$= 1000 \times 0.1254 = 125.4 \approx 125 \text{ nearly.}$$

Example: → The marks obtained by candidates in statistics in a certain examination are found to be normally distributed. If 12.5% of the candidates obtain 60 or more marks and 39% obtain less than 30 marks. Find the mean number of marks obtained by the candidates.

Sol.: → Let  $X \sim N(\mu, \sigma^2)$ , where  $\mu$  = mean,  $\sigma$  = s.d

It is given that  $P[X > 60] = 0.125$ , and  $P[X < 30] = 0.39$

$$\Rightarrow P[\mu < X < 60] = 0.5 - 0.125 = 0.375 \text{ and } P[30 < X < \mu] = 0.5 - 0.39 = 0.11$$

$$\Rightarrow P\left[\frac{\mu-\mu}{\sigma} < \frac{x-\mu}{\sigma} < \frac{60-\mu}{\sigma}\right] = 0.375 \text{ & } P\left[\frac{30-\mu}{\sigma} < \frac{x-\mu}{\sigma} < \frac{\mu-\mu}{\sigma}\right] = 0.11$$

$$\Rightarrow P\left[0 < z < \frac{60-\mu}{\sigma}\right] = 0.375 \text{ & } P\left[\frac{30-\mu}{\sigma} < z < 0\right] = 0.11$$

$$\Rightarrow \frac{60-\mu}{\sigma} = 1.15 - ① \quad \text{&} \quad P\left[0 < z < \frac{\mu-30}{\sigma}\right] = 0.11 \quad (\text{By symmetry})$$

$$\therefore \frac{\mu-30}{\sigma} = 0.28 - ②$$

on solving both, we get  $\boxed{\mu = 35.87}$  &  $\boxed{\sigma = 20.97}$

Example: → In a distribution exactly 7% of the items are under 35 and 89% are under 63. what are the mean and standard deviation of the distribution.

Sol.: → If  $X \sim N(\mu, \sigma^2)$ , where  $\mu$  = mean &  $\sigma$  = s.d we are given  $P[X < 63] = 0.89$  and  $P[X < 35] = 0.07$

$$P[\mu < X < 63] = 0.89 - 0.5 = 0.39 \text{ and } P[35 < X < \mu] = 0.5 - 0.07 = 0.43$$

$$P\left[\frac{\mu-\mu}{\sigma} < \frac{x-\mu}{\sigma} < \frac{63-\mu}{\sigma}\right] = 0.39 \text{ and } P\left[\frac{35-\mu}{\sigma} < \frac{x-\mu}{\sigma} < \frac{\mu-\mu}{\sigma}\right] = 0.43$$

$$P\left[0 < z < \frac{63-\mu}{\sigma}\right] = 0.39 \text{ and } P\left[\frac{35-\mu}{\sigma} < z < 0\right] = 0.43$$

$$\frac{63-\mu}{\sigma} = 1.23 - ① \quad \text{and} \quad P\left[0 < z < \frac{\mu-35}{\sigma}\right] = 0.43$$

$$\Rightarrow \frac{\mu-35}{\sigma} = 1.48 - ②$$

On solving ① & ②, we get

$$\boxed{\mu = 50.3} \quad \text{and} \quad \boxed{\sigma = 10.33}$$