

$$\therefore \lim_{n \rightarrow \infty} p(x) = \frac{m(m-0)(m-0) \dots (m-0)}{x!} \cdot \lim_{n \rightarrow \infty} \frac{(1 - \frac{m}{n})^n}{(1 - \frac{m}{n})^n} = \frac{m^x}{x!} \cdot e^{-m} = e^{-m} \cdot \frac{m^x}{x!}, \text{ which is poisson distribution.}$$

Thus Poisson distribution can be treated as binomial distribution with large  $n$  and small  $p$ .

#### Observations :

Poisson distribution is used in all cases where Bernoulli trials are involved i.e., where binomial distribution type of situation arises but with the difference that

- there is no fixedness for the number of trials or the number of trials is large.
- number of failures is of no concern.

A great variety of natural and man-made phenomena are found to follow Poisson's distribution. Some common features of such phenomena are :

- the probability of success during a small interval (of time or space) is proportional to the interval.
- the probability of more than one success in the interval is negligible.
- the probability of success in a given interval does not depend on what happened prior to that interval.

If we take into consideration a time interval  $T$  and divide it into  $n$  sub-intervals of duration  $\Delta t$ , then  $T = n \cdot \Delta t$ .

Suppose that the probability of a success in the small interval  $\Delta t$  is  $\alpha \cdot \Delta t$  where  $\alpha$  is constant. The interval  $T$  has  $n$  sub-intervals of duration  $\Delta t$  and the probability of success in each sub-interval is  $p = \alpha \Delta t$ . We have now a Bernoulli trial in each sub-interval and there are  $n$  such trials over the period  $T$ . Thus we have a binomial distribution with  $n = \frac{T}{\Delta t}$  and  $p = \alpha \Delta t$  since  $n$  is large and  $\Delta t$  small, we can approximate by Poisson distribution with  $m = \text{mean} = np$  i.e.,  $= \frac{T}{\Delta t} \cdot \alpha \Delta t$  i.e., with  $m = \alpha T$ . Then  $p(x) = \frac{e^{-m} m^x}{x!}$  gives the probability for  $x$  successes in the interval  $T$ .

Some phenomena which fit to this model are :

- arrival of telephone calls at a switchboard,
- passing by of trains at an unmanned gate,
- number of deaths due to accidents in a month on a National Highway.

#### PROBABILITY

- number of typographical mistakes in a page of this book.
- number of  $\alpha$ -particles emitted by a radioactive source in a unit of time.
- number of electrons released from the cathode of a vacuum tube in a unit of time.

These are some examples in respect of time intervals. We may also have space intervals. Some examples are

- number of blood cells visible under the microscope. Here the surface area is  $T$ .
- number of stars appearing a portion of the Milky way.
- number of particles of impurities found in a quantity of a molten substance. Here the volume is  $T$ .
- number of germinating seeds on a plot of land. Here the area of the plot is  $T$ .
- number of deaths due to an epidemic in a given locality.
- number of germs in a bacteria culture experiment.

(e) Recurrence relation  $p(x+1) = \frac{m}{x+1} \cdot p(x)$ .

Proof : By definition of Poisson distribution,

$$p(x) = \frac{e^{-m} \cdot m^x}{x!}.$$

Changing  $x$  to  $x+1$ , we have

$$p(x+1) = \frac{e^{-m} \cdot m^{x+1}}{(x+1)!}$$

We can write  $p(x+1) = \frac{e^{-m} \cdot m \cdot m^x}{(x+1) \cdot x!}$

$$\text{i.e., } = \frac{m}{x+1} \cdot \frac{e^{-m} m^x}{x!}$$

$$= \frac{m}{x+1} \cdot p(x) \text{ by above.}$$

$$\text{Thus } p(x+1) = \frac{m}{x+1} \cdot p(x).$$

#### WORKED EXAMPLES

**Example 5.67** With the usual notation prove that for a Poisson distribution,

$$\mu_{r+1} = m \left( r \cdot \mu_{r-1} + \frac{d}{dm} \mu_r \right)$$