

# Probability Distributions: (Standard) (Parametric)

## ∴ Discrete Distribution:

### 1 Bernoulli Distribution : [Bernoulli Trials]

The Bernoulli distribution is named after swiss mathematicians James Bernoulli (1654-1705). This distribution is associated with the Bernoulli trial

- Bernoulli trial is an experiment with two outcome which are random, named as success and failure
- Prob. success is denoted by  $P$  and prob. of failure is denoted by  $q$ .
- $P+q=1$

Example: Many real life instances, which can be viewed as Bernoulli trial

1. A coin is tossed. It has two outcomes, H & T.

$$P = \frac{1}{2} \neq q = \frac{1}{2}$$

2. A fair die is tossed. consider getting <sup>no</sup> 6 as a success and any other no. as failure  $P = \frac{1}{6}$  &  $q = \frac{5}{6}$ .

3. A card is drawn from a pack of cards. consider getting a card of spade as a success and a card of any other suit as a failure, then  $P = \frac{1}{4}$  &  $q = \frac{3}{4}$ .

4. Two person A and B, are contesting in an election. A public opinion survey is conducted. A voter is selected at random, and he (or she) is asked whether he (or she) would vote for the candidate A.

Note: Bernoulli trials feature questions with "yes" or "no" type of answers. "yes" corresponds to a success, "no" corresponds to a failure.

Def<sup>n</sup>: Bernoulli Distribution:

A r.v  $x$  is said to have a Bernoulli distribution if it can take only two values 0 and 1, with 1 being considered as a success and 0 as a failure, with Prob.  $P$  &  $q$  resp.

$$\text{ie } f(x; p) = P(x=x) = \begin{cases} p & \text{if } x=1 \\ q=1-p & \text{if } x=0 \end{cases}$$

$$\text{Mean : } E(x) = p$$

$$\text{Var}(x) : \text{Var}(x) = p q$$

$$\begin{aligned} \text{Mean} = M = E(x) &= \sum x \cdot P(x=x) \\ &= 0 \times f(0) + 1 \times f(1) = 0 + p = p \end{aligned}$$

$$\begin{aligned} E(x^2) &= \sum x^2 p(x=x) \\ &= 0^2 \times f(0) + 1^2 \times f(1) = p \end{aligned}$$

$$\begin{aligned} \therefore \text{Var}(x) &= E(x^2) - (E(x))^2 \\ &= p^2 - p^2 = 0 \cdot p(1-p) = pq \end{aligned}$$

## Binomial Distribution:

It was discovered by James Bernoulli in 1700.  
 It is a discrete probability distribution of the no. of successes in a sequence of n independent Bernoulli trials with a constant prob. of success P in all the trials.

For eg: Consider a sequence of 3 Bernoulli trials with const. prob. of success P in each trial

Then

$$P(X=0) = P(\text{All failure}) \\ = q \times q \times q = q^3$$

$$P(X=1) = P(2 \text{ failure and 1 success}) \\ = P(3FF) + P(FSF) + P(FFS) \\ = Pq^2 + Pq^2 + Pq^2 \\ = 3Pq^2$$

$$P(X=2) = P(1 \text{ failure and 2 success}) \\ = P^2q + P^2q + P^2q \\ = 3P^2q$$

$$P(X=3) = P(3 \text{ success}) = P(SSS) \\ = P \times P \times P = P^3$$

∴ For the sequence of n Indept Bernoulli trials we have

$$P(X=x) = P(x \text{ success and } (n-x) \text{ failure})$$

The x successes in n trials can occur in  $nC_x$  ways  
 The prob. for each of these ways is  $P^x q^{n-x}$

$$\therefore P(X=x) = nC_x P^x q^{n-x}, x=0,1,2,\dots,n$$

Def<sup>n</sup>: A r.v. X is said to follow binomial distri. denoted by  $B(n,p)$  if it assumes only non-negative values and its PMF is given by

$$P(x) = P(X=x) = nC_x P^x q^{n-x}, x=0,1,2,\dots,n \\ = 0, \text{ otherwise}$$

$$q = 1-p$$

### Binomial Frequency Distribution:

Suppose that  $n$  trials constitute an experiment and if this experiment is repeated  $N$  times, the frequency fun of the binomial distribution is given by

$$N.P(x) = N \times nC_x P^x q^{n-x}, \quad x = 0, 1, 2, \dots, n$$

#### Properties:

A binomial experiment must possess the following

1. Each trial results in two mutually disjoint outcomes, termed as success and failure
2. The trials must be independent of each other
3. All trials have same const. prob. of success
4. The no. of trials  $n$  is finite

Note: Let  $p$  = prob. of success

$$P(\text{all success}) = p^n$$

$$P(\text{all failure}) = q^n$$

$$P(\text{at least one success}) = 1 - q^n$$

$$P(\text{at least one failure}) = 1 - p^n$$

### Mean of Binomial Distribution:

$$\begin{aligned} M_1 &= \text{Mean} = E(X) = \sum_{x=0}^n x \cdot P(x) \\ &= \sum_{x=0}^n x \cdot nC_x P^x q^{n-x} \\ &= \sum_{x=0}^n x \cdot \frac{n!}{(n-x)! x!} P^x q^{n-x} \\ &= \sum_{x=0}^n x \cdot \frac{n(n-1)!}{(n-x)!(x+1)!} P^x q^{n-x} \end{aligned}$$

$$\begin{aligned}
 \therefore E(x) &= \sum_{x=1}^n \frac{n(n-1)!}{[n-(x-1)]!(x-1)!} p^{x-1} \cdot P \cdot q^{[n-(x-1)]} \\
 &= np \sum_{x=1}^{\infty} {}^{n-1} C_{x-1} p^{x-1} q^{[n-(x-1)]} \\
 &= np (q+p)^{n-1} \\
 &= np \quad (\because q+p=1)
 \end{aligned}$$

### Variance of Binomial Distribution:

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

where

$$\begin{aligned}
 E(x^2) &= E[x(x-1)+x] \\
 &= E \sum_{x=0}^n [x(x-1)+x] P(x) \\
 &= \sum_{x=0}^n x(x-1) P(x) + \sum_{x=0}^n x P(x) \\
 &= \sum_{x=0}^n x(x-1) {}^n C_x p^x q^{n-x} + E(x) \\
 &= \sum_{x=0}^n x(x-1) \frac{n!}{(n-x)! x(x-1)(x-2)!} p^x q^{n-x} + E(x) \\
 &= \sum_{x=2}^n \frac{n(n-1)(n-2)!}{[(n-2)-(x-2)]!(x-2)!} p^{x-2} q^{[n-(x-2)]} + E(x) \\
 &= n(n-1)p^2 \sum_{x=2}^{\infty} {}^{n-2} C_{x-2} p^{x-2} q^{[n-(x-2)]} + E(x) \\
 &= n(n-1)p^2 (q+p)^{n-2} + E(x) \\
 &= n(n-1)p^2 + np \\
 &= np^2 + np
 \end{aligned}$$

$$\begin{aligned}\mu_2 &= \text{Var}(X) = n^2 p^2 - np^2 + np - (np)^2 \\ &= np - np^2 \\ &= np(1-p) = npq\end{aligned}$$

Note: -  $\mu_3 = npq(q-p)$

-  $\mu_4 = npq(1-6pq+3npq)$

- Mean of the binomial distribution is always greater than its variance

Ex:1 The mean and variance of a binomial distribution are 4 and 3 resp. Find  $P(X=0)$ ,  $P(X=1)$  and  $P(X \geq 2)$

:-

$$\begin{aligned}\text{Mean} &= np & \text{Var}(X) &= npq \\ \text{i.e } 4 &= np & \text{i.e } 3 &= npq \\ && \therefore 3 &= (4)q \quad \therefore q = \frac{3}{4} \\ && & \therefore P = 1 - q \\ && & = \frac{1}{4}\end{aligned}$$

$$\text{Mean} = np$$

$$\therefore 4 = n\left(\frac{1}{4}\right) \quad \therefore n = 16$$

$$P(X=x) \leq nC_x p^x q^{n-x}$$

$$\therefore P(X=0) = 16C_0 \left(\frac{1}{4}\right)^0 \left(\frac{3}{4}\right)^{16-0} = \left(\frac{3}{4}\right)^{16} = 0.01$$

$$P(X=1) = 16C_1 \left(\frac{1}{4}\right)^1 \left(\frac{3}{4}\right)^{15} = 16 \times \frac{1}{4} \times \left(\frac{3}{4}\right)^{15} = 0.053$$

$$\begin{aligned}P(X \geq 2) &= 1 - P(X < 2) = 1 - [P(X=0) + P(X=1)] \\ &= 1 - [0.01 + 0.053] \\ &= 0.93\end{aligned}$$

Ex: 2 The mean and std. deviation of a binomial distribution are 6 and 2 resp. Determine the distribution.

$$\text{mean} = np = 6$$

$$\text{s.d.} = 2 \Rightarrow \text{var}(x) = (\text{s.d.})^2 = 4 = npq$$

$$\therefore p = \frac{1}{3}, q = \frac{2}{3}, n = 18$$

The binomial distribution is

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

$$= 18C_x \left(\frac{1}{3}\right)^x \left(\frac{2}{3}\right)^{n-x}, x=0,1,\dots,18$$

Q

Moment Generating function of Binomial Distribution:

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

$$= \sum_{x=0}^n e^{tx} nC_x p^x q^{n-x}$$

$$= \sum_{x=0}^n nC_x (pe^t)^x q^{n-x}$$

$$= q^n + nC_1 (pe^t)^1 q^{n-1} + nC_2 (pe^t)^2 q^{n-2} + \dots \\ \dots + (pe^t)^n$$

$$= (q + pe^t)^n$$

Mean of M.Gf fun

$$E(x) = M'_1 = \left[ \frac{d}{dt} M_X(t) \right]_{t=0}$$

$$= \left[ \frac{d}{dt} (q + pe^t)^n \right]_{t=0}$$

$$= [n(q + pe^t)^{n-1} \cdot pe^t]_{t=0}$$

$$= np(p+q)^{n-1}$$

$$= np$$

$$\begin{aligned}
 E(X^2) &= M'_2 = \left[ \frac{d^2}{dt^2} M_X(t) \right]_{t=0} \\
 &= \left[ \frac{d}{dt} (n(q+pe^t)^{n-1} pe^t) \right]_{t=0} \\
 &= npe^t(q+pe^t)^{n-1} + n(n-1)(q+pe^t)^{n-2}(pe^t)^2 \\
 &= np(q+p)^{n-1} + n(n-1)p^2(q+p)^{n-2} \\
 &= np + n(n-1)p^2 \\
 &= np + (n^2-n)p^2
 \end{aligned}$$

$$\begin{aligned}
 \text{Var}(X) &= E(X^2) - (E(X))^2 \\
 &= np + n^2 p^2 - np^2 - (np)^2 \\
 &= np - np^2 \\
 &= np(1-p) = npq
 \end{aligned}$$

Ex For a r.v  $X$ ,  $M_X(t) = \frac{1}{81}(e^t + 2)^4$ . Find  $P(X \leq 2)$

$\Rightarrow$  For a binomial r.v  $Y$  with parameters  $n$  and  $p$  the moment generating function of  $X$  is given by

$$M_Y(t) = (q + pe^t)^n$$

Given that

$$\begin{aligned}
 M_X(t) &= \frac{1}{81}(e^t + 2)^4 \\
 &= \left(\frac{2}{3} + \frac{1}{3}e^t\right)^4
 \end{aligned}$$

By the uniqueness theorem of M.G.F  $M_X(t) = M_Y(t)$   
 $\Leftrightarrow F_X = F_Y$

$$\begin{aligned}
 \therefore M_X(t) &= M_Y(t) \\
 &= (q + pe^t)^n \\
 &= \left(\frac{1}{3}e^t + \frac{2}{3}\right)^4
 \end{aligned}$$

$$\therefore P = \frac{1}{3}, q = \frac{2}{3}, n = 4$$

$$P(X=x) = {}^4C_x P^x q^{4-x}$$

$$= {}^4C_x \left(\frac{1}{3}\right)^x \left(\frac{2}{3}\right)^{4-x}, \quad x=0,1,2,3,4$$

$$P(X \leq 2) = P(X=0) + P(X=1) + P(X=2)$$

$$= q^4 + 4pq^3 + 6p^2q^2$$

$$= \frac{16}{81} + \frac{32}{81} + \frac{8}{27} = \frac{8}{9}$$

- Ex:** A box contains 100 cellphones, 20 of which are defective. 10 cellphones are selected for inspection. Find the probability that (i) at least one is defective  
(ii) at the most three are defective  
(iii) all the ten are defective  
(iv) none of the ten is defective

$\therefore x$  — no. of defective cellphone

$$n = 10$$

$$p = \frac{20}{100} = \frac{1}{5} \quad \therefore q = 1 - \frac{1}{5} = \frac{4}{5}$$

$$P(X=x) = {}^{10}C_x \left(\frac{1}{5}\right)^x \left(\frac{4}{5}\right)^{10-x}, \quad x=0,1,2,\dots,10$$

$$= {}^{10}C_x \frac{4^{10-x}}{5^{10}}$$

(i) at least one cellphone defective

$$P(X \geq 1) = 1 - P(X=0)$$

$$= 1 - \left(\frac{4}{5}\right)^{10} = 0.8926$$

(ii) at the most three cellphones are defective

$$P(X \leq 3) = P(X=0) + P(X=1) + P(X=2) + P(X=3)$$

$$= \frac{1}{5^{10}} [4^{10} + (10 \times 4^9) + (45 \times 4^8) + (120 \times 4^7)]$$

$$= 0.8791$$

(iii) All ten cellphones are defective

$$P(X=10) = 1 \times \left(\frac{1}{5}\right)^{10} = 1.0240 \times 10^{-7}$$

(iv) None of the ten is defective is

$$P(X=0) = 1 \times \left(\frac{4}{5}\right)^{10} = 0.1074$$

**Ex:** It is known that 5% of the books bound at a certain bindery have defective bindings. Find the prob. that 2 of 100 books bound by this bindery will have defective bindings.

$\therefore$   $X$  - no. of books with defective bindings.

$$n = 100$$

$$p = \frac{5}{100} = \frac{1}{20}$$

$$P(X=x) = {}^{100}C_x \left(\frac{1}{20}\right)^x \left(\frac{19}{20}\right)^{100-x}, \quad x = 0, 1, 2, \dots, 100$$

$$\therefore P(X=2) = {}^{100}C_2 \left(\frac{1}{20}\right)^2 \left(\frac{19}{20}\right)^{98} = 0.0812$$

**Ex:** Six dice are thrown 729 times. How many times do you expect at least 3 dice to show 5 or 6?

$\therefore$  Success - Event getting a 5 or 6

$$\therefore p = \frac{2}{6} = \frac{1}{3} \quad \therefore q = \frac{2}{3}$$

When 6 dice are thrown, the no. of successes  $X$ , follows a binomial distribution with

$$n = 6$$

$$p = \frac{1}{3}$$

$$P(X=x) = {}^6C_x \left(\frac{1}{3}\right)^x \left(\frac{2}{3}\right)^{6-x}, \quad x = 0, 1, \dots, 6$$

$$\begin{aligned} P(X \geq 3) &= 1 - P(X < 3) = 1 - [P(X=0) + P(X=1) + P(X=2)] \\ &= 1 - \frac{1}{3^6} [2^6 + (6 \times 2^5) + (10 \times 2^4)] = \frac{233}{729} \end{aligned}$$

Thus, when 6 dice are thrown 729 times, the no. of times we expect at least 3 dice to show 5 or 6 is

$$N \times P(X \geq 3) = 729 \times \frac{233}{729} = 233$$

Ex: In a certain town 20% samples of the population are literate. Assume that 200 investigators each take sample of 10 individuals to see whether they are literate. How many investigators would you expect to report that 3 people or less are literate in the samples?

→ Given:

$$P(\text{An individual is literate}) = \frac{20}{100} = 0.2$$

$$\text{i.e., } P = 0.2 \quad \text{and } q = 1 - P = 0.8$$

$$n = 9 \times 10 = 90$$

$X$  - no. of literate

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

$$= 10C_x (0.2)^x (0.8)^{10-x}$$

P(Investigator reporting 3 or less as literate)

$$P(X \leq 3) = P(X=0) + P(X=1) + P(X=2) + P(X=3)$$

$$= 10C_0 (0.2)^0 (0.8)^{10} + 10C_1 (0.2)^1 (0.8)^9$$

$$+ 10C_2 (0.2)^2 (0.8)^8 + 10C_3 (0.2)^3 (0.8)^7$$

$$= (0.8)^7 [ (0.2)^3 + 10 (0.2)(0.8)^2 + 45 (0.2)^2 (0.8)^2 + 120 (0.2)^3 ]$$

$$= 0.2097 (0.512 + 1.28 + 1.44 + 0.96)$$

$$= 0.8790$$

The no. of investigators reporting 3 or less as literate out of 200 =  $200 \times 0.8790 = 175.82 = 176$ .

Ex: A variate takes values  $0, 1, 2, \dots, n$  with frequencies proportional to the binomial coefficients  $1, nC_1, nC_2, \dots, nC_n$ . Find the mean and the variance of the distribution and show that the variance is half of the mean.

$$\therefore \text{The total frequency} = 1 + nC_1 + nC_2 + \dots + nC_n \\ = (1+1)^n = 2^n$$

Given: The variate takes values  $0, 1, 2, \dots, n$  with frequencies proportional to the binomial coeffi.  $1, nC_1, nC_2, \dots, nC_n$

$\therefore$  The probabilities for  $0, 1, 2, \dots, n$  values of the variates are  $\frac{1}{2^n}, \frac{nC_1}{2^n}, \frac{nC_2}{2^n}, \dots, \frac{nC_n}{2^n}$  resp.

The above are the terms of a binomial distribution  $(\frac{1}{2} + \frac{1}{2})^n$  where  $P = \frac{1}{2}$  and  $q = \frac{1}{2}$

$$\text{Mean} = np = \frac{n}{2}$$

$$\text{Vari} = npq = n \times \frac{1}{2} \times \frac{1}{2} = \frac{n}{4} = \frac{\text{mean}}{2}$$

Ex: An irregular six-faced die is thrown and the expectation that in 10 throws it will give 5 even nos. is twice the expectation that it will give 4 even nos. How many times in 10000 sets of 10 throws (each) would you expect it to give no even nos?

$\therefore$  Let the r.v.  $X$  denote the no. of even nos

$$P(\text{getting } x \text{ even nos}) = P(X=x)$$

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

Given:  $n=10$

$$\therefore P(X=x) = 10C_x P^x q^{10-x}, \quad x=0, 1, 2, \dots, 10$$

$P(\text{getting 5 even no.}) = 2 P(\text{getting 4 even no.})$

$$P(X=5) = 2 P(X=4)$$

$$\therefore {}^{10}C_5 P^5 q^5 = 2 ({}^{10}C_4 P^4 q^6)$$

$$\text{i.e. } \frac{P}{5} = \frac{q}{3} \Rightarrow 3P = 5q$$

$$\Rightarrow 3P = 5(1-P)$$

$$\Rightarrow 8P = 5$$

$$\therefore P = \frac{5}{8}$$

$$\therefore q = \frac{3}{8}$$

$$\therefore P(\text{getting } x \text{ even no.}) = {}^{10}C_x \left(\frac{5}{8}\right)^x \left(\frac{3}{8}\right)^{10-x} \quad x=0,1,\dots,10$$

Q. The required no. of times that in 10000 sets of 10 throws each we get no even nos.

$$= 10000 \times P(X=0)$$

$$= 10000 \times {}^{10}C_0 \left(\frac{5}{8}\right)^0 \left(\frac{3}{8}\right)^{10} = 1 \text{ (approx.)}$$

Ex: If at least 1 child in a family with 2 children is a boy, what is the prob. that both children are boys?

$$\therefore P(\text{child is a boy}) = \frac{1}{2} = p \quad \therefore q = \frac{1}{2}$$

$$n = 2$$

$x$  - no. of boys

$$P(X=x) = {}^2C_x \left(\frac{1}{2}\right)^x, \quad x=0,1,2.$$

$$P(\text{at least 1 boy}) = P(X=1 \text{ or } 2) = {}^2C_1 \left(\frac{1}{2}\right)^2 + {}^2C_2 \left(\frac{1}{2}\right)^2 = \frac{3}{4}$$

$P(\text{both boys} | \text{at least 1 boy})$

$$P(X=2 | X=1 \text{ or } 2) = \frac{P(X=2 \cap X=1 \text{ or } 2)}{P(X=1 \text{ or } 2)}$$

$$= P(X=2) = \frac{2C_2\left(\frac{1}{2}\right)^2}{3^4} = \frac{1}{13}$$

Ex: If the probability of success is  $\frac{1}{100}$ , how many trials are necessary in order that probability of at least one success is greater than  $\frac{1}{2}$ ?

$\therefore X$  - no. of successes

$$\text{Given: } P = \frac{1}{100} \therefore q = \frac{99}{100}$$

$$\therefore P(X=x) = nC_x \left(\frac{1}{100}\right)^x \left(\frac{99}{100}\right)^{n-x}, \quad x=0,1,2,\dots,n$$

$$P(X \geq 0) = (0.99)^n = q^n$$

$P(\text{hitting at least once})$  is largest if

$$P(X \geq 1) = 1 - P(X < 1)$$

$$= 1 - q^n$$

$$P(X \geq 1) > \frac{1}{2} \quad \therefore 1 - q^n > \frac{1}{2}$$

$$\text{i.e., } 1 - (0.99)^n > \frac{1}{2}$$

$$\text{i.e., } 1 - \frac{1}{2} > (0.99)^n$$

$$\text{i.e., } \frac{1}{2} > (0.99)^n$$

$$\Rightarrow \log(0.5) > n \log(0.99)$$

$$\Rightarrow \frac{0.3010}{0.0044} < n$$

$$\Rightarrow 68.4 < n$$

$$\Rightarrow n = 69$$

Ex: Ten coins are tossed 1024 times and the following frequencies are observed. compare these frequencies with the expected frequencies:

No. of heads : 0 1 2 3 4 5 6 7 8 9 10

Given Frequencies : 2 10 38 106 188 257 226 128 59 7 3

$\therefore$  Given  $n = 10$

$$N = \sum f = 1024$$

$P$  = Prob. of getting a head in one toss

$$\therefore P = \frac{1}{2}$$

$$\therefore q = \frac{1}{2}$$

The expected frequency of  $x$  heads

$$= 1024 \times {}^{10}C_x \left(\frac{1}{2}\right)^{10-x} \left(\frac{1}{2}\right)^x, \quad x=0,1,2,\dots,10$$

Hence

Observed fre. : 2 10 38 106 188 257 226 128 59 7 3

Expected fre. : 1 10 45 120 210 252 210 120 45 10 1

Ex: A biased coin is tossed 8 times and the no. of heads are noted. The experiment is repeated 100 times and the following frequency distribution is obtained.

$x$	0	1	2	3	4	5	6	7	8
$f$	2	7	13	15	25	16	11	8	3

Fit a binomial distribution to these data calculate the theoretical frequencies.

$$\therefore \text{Mean} = \mu = \frac{\sum f x}{\sum f} = \frac{404}{100} = 4.04$$

$$\text{Thus } p = \frac{\mu}{n} = \frac{4.04}{8} = 0.505 \quad (\text{Mean} = np)$$

$$\therefore q = 1-p = 0.495$$

Since  $N=100$ , the theoretical frequencies is given by

$$N f(x) = N \cdot {}^n C_x \cdot p^x \cdot q^{n-x} = 100 \cdot {}^8 C_x (0.505)^x (0.495)^{8-x}$$

$x=0,1,2,\dots,8$

$$Nf(0) = 100 \times (0.495)^8 = 0.3604$$

$$Nf(1) = 100 \times 8C_1 (0.505) (0.495)^7 = 2.9414 \text{ and so on } Nf(n)$$

No of heads	observed freq	Theo. freq
0	2	0.3604
1	7	2.9414
2	13	10.5030
3	15	21.4303
4	25	27.3290
5	16	22.3049
6	11	11.3777
7	8	3.3164
8	3	0.4229

## Generalization of Bernoulli Theorem / Multinomial Distribution

If  $A_0, A_1, A_2, \dots, A_n$  are exhaustive and mutually exclusive events associated with a random experiment such that  $P(A_i) = p_i$  where  $p_1 + p_2 + \dots + p_k = 1$  and if the experiment is repeated  $n$  times, then the probability that  $A_1$  occurs  $r_1$  times,  $A_2$  occurs  $r_2$  times,  $\dots, A_k$  occurs  $r_k$  times such that  $r_1 + r_2 + \dots + r_k = n$  is given by

$$P_n(r_1, r_2, \dots, r_k) = \frac{n!}{r_1! r_2! \dots r_k!} p_1^{r_1} p_2^{r_2} \dots p_k^{r_k}$$

e.g. A fair die is rolled 5 times. Find the prob. that 1 shown twice, and 6 shown once.

Given  $r_1 = 5$

$A_1 =$  getting 1

$A_2 =$  getting 6

$P(A_1) = \frac{1}{6} (= p_1)$   $P(A_2) = \frac{1}{6} = p_2$   $P(A_3) = \frac{1}{6} = p_3$

$r_1 = 2, r_2 = 2, r_3 = 1$

$\therefore r_1 + r_2 + r_3 = 5 = n$

Using the multinomial distribution, the required

$$P_n(r_1, r_2, r_3) = \frac{n!}{r_1! r_2! r_3!} p_1^{r_1} p_2^{r_2} p_3^{r_3}$$

$$= \frac{5!}{2! 2! 1!} \left(\frac{1}{6}\right)^2 \left(\frac{1}{6}\right)^2 \left(\frac{1}{6}\right)$$

$$\frac{p_1}{p_2} = \frac{1}{6} = \frac{1}{5} \Rightarrow \frac{1}{5} = 0.0039$$

$$P(A) = P_5(2, 2, 1) = \frac{5!}{2! 2! 1!} \left(\frac{1}{6}\right)^2 \left(\frac{1}{6}\right)^2 \left(\frac{1}{6}\right)$$

Ex: If the moment generating fun<sup>n</sup> of a r.v.  $X$  is  $\left(\frac{2}{3} + \frac{1}{3}e^t\right)^9$

Show that  $P(\mu - 2\sigma < X < \mu + 2\sigma) = \sum_{x=1}^5 q C_x \left(\frac{1}{3}\right)^x \left(\frac{2}{3}\right)^{9-x}$

$\Rightarrow$  Given: Moment generating fun<sup>n</sup>

$$M_X(t) = (q + pe^t)^n = \left(\frac{2}{3} + \frac{1}{3}e^t\right)^9$$

$$\therefore q = \frac{2}{3} \text{ and } p = \frac{1}{3} \text{ and } n = 9$$

$\therefore X$  follows binomial distribution  $\sim B(9, \frac{1}{3})$

$$E(X) = np = 9 \times \frac{1}{3} = 3 = \mu$$

$$\text{Var}(X) = npq = 9 \left(\frac{1}{3}\right) \left(\frac{2}{3}\right) = 2 = \sigma^2$$

$$\therefore \sigma = \sqrt{\text{Var}(X)} = \sqrt{2}$$

$$\therefore \mu \pm 2\sigma = 3 \pm 2\sqrt{2} = 0.2 \text{ or } 5.8$$

$$\therefore P(\mu - 2\sigma < X < \mu + 2\sigma) = P(0.2 < X < 5.8)$$

$$= P(1 \leq X \leq 5)$$

$$= \sum_{x=1}^5 q C_x p^x q^{n-x}$$

$$= \sum_{x=1}^5 q C_x \left(\frac{1}{3}\right)^x \left(\frac{2}{3}\right)^{9-x}$$

Ex: If the moment generating fun<sup>n</sup> of a r.v.  $X$  is of the form  $(0.4e^t + 0.6)^8$ , find the moment generating fun<sup>n</sup> of  $3X + 2$ .

$\therefore$  Here  $q = 0.6$ ,  $p = 0.4$ ,  $n = 8$

$$\begin{aligned} M_{3X+2}(t) &= E(e^{(3X+2)t}) = E[e^{2t} \cdot e^{3Xt}] \\ &= e^{2t} E(e^{3Xt}) \\ &= e^{2t} E[e^{X(3t)}] \\ &= e^{2t} (0.6 + 0.4e^{3t})^8 \end{aligned}$$

## Recurrence Formula for the Central Moments of the Binomial Distribution.

∴ By def<sup>n</sup>, the  $k^{\text{th}}$  order central moment  $\mu_k$  is given by

$$\begin{aligned}\mu_k &= E\{(x - E(x))^k\} \\ &= \sum_{x=0}^n (x - np)^k n C_x p^x q^{n-x} \quad \text{--- (1)}\end{aligned}$$

Diffling (1) w.r.t P, we get

$$\begin{aligned}\frac{d\mu_k}{dp} &= \sum_{x=0}^n n C_x \left\{ -nk(x-np)^{k-1} p^x q^{n-x} \right. \\ &\quad \left. + (x-np)^k [x p^{x-1} q^{n-x} + p^x (n-x) q^{n-x-1} (-1)] \right\} \\ &= -nk\mu_{k-1} + \sum_{x=0}^n n C_x (x-np)^{k-1} p^x q^{n-x} [xq - (n-x)p] \\ &= -nk\mu_{k-1} + \frac{1}{pq} \sum_{x=0}^n n C_x p^x q^{n-x} (x-np)^{k+1} \\ &= -nk\mu_{k-1} + \frac{1}{pq} \mu_{k+1}\end{aligned}$$

$$\therefore \mu_{k+1} = pq \left( \frac{d\mu_k}{dp} + nk\mu_{k-1} \right)$$

### Moments of higher order

$$\begin{aligned}\text{Put } k=1 \text{ in above, } \mu_2 &= pq \left( \frac{d\mu_1}{dp} + n\mu_0 \right) \\ &= npq \quad (\mu_0=1 \text{ & } \mu_1=0)\end{aligned}$$

$$\begin{aligned}k=2 \quad \mu_3 &= pq \left( \frac{d\mu_2}{dp} + 2n\mu_1 \right) \\ &= pq \frac{d}{dp} (np(1-p)) = npq(1-2p) \\ &= npq(p-1)\end{aligned}$$

$$k=3, \text{ simplifying, } \mu_4 = npq[1+3pq(n-2)]$$

Note: $\mu_2 = \text{Variance}$  $\mu_3 = \text{measure of skewness } (\beta_1)$  $\mu_4 = " " \text{ kurtosis } (\beta_2)$ 

$$\text{where } \mu_1 = \frac{\mu_3^2}{\mu_2^3} \quad \mu_2 = \frac{\mu_4}{\mu_2^2}$$

Note: The mean of binomial distri. is always greater than variance.