

3.4 Binomial Probability Distribution

Binomial Experiment:

An experiment for which the following 4 conditions are satisfied is called a binomial experiment.

- 1) The experiment consists of sequence of n smaller experiments called trials, where n is fixed in advance of the experiment.
- 2) Each trial can result in one of the same two possible outcomes, which we generally denote by success (S) and failure (F).
- 3) The trials are independent, so that the outcome on any particular trial does not influence the outcome on any other trial.
- 4) The probability of success $P(S)$ is constant from trial to trial, we denote this probability by p and probability of failure is denoted by $q(=1-p)$.

Binomial random variable:

The binomial random variable X associated with a binomial experiment consisting of n trials is defined as

$X =$ the number of success (S) among the n trials

Example:

For $n=3$, there are eight possible outcomes for the experiment.

$\{SSS, SSF, SFF, FSS, FSF, FFS, SFS, FFF\}$

Then from definition, $X(SSS)=3, X(SSF)=2, X(SFF)=1,$
 $\dots, X(FFF)=0$

\therefore possible values for X in 3-trial experiment
 are $x=0, 1, 2, 3$.

Similarly possible values for X in an n -trial experiment
 are $x=0, 1, 2, \dots, n$.

Some notations:

- 1) $p \rightarrow$ probability of success in a single trial
 $q=1-p \rightarrow$ probability of failure " u " " l " " 1 "
- 2) $X \sim \text{Bin}(n, p) \rightarrow X$ is a binomial r.v. based on n trials
 with probability of success p
- 3) $b(x; n, p) \rightarrow$ pmf of Binomial distribution
 with parameters n and p and r.v. X .
 $B(x; n, p) \rightarrow$ cdf of binomial distribution
- 4) $S \rightarrow$ Success, the occurrence of any event A
 $F \rightarrow$ failure, the non-occurrence of A

The Binomial Distribution:

A discrete random variable X is said to have a binomial distribution with parameters n (+ve integer) and p ($0 < p < 1$) if pmf of X is given by

$$b(x; n, p) = \binom{n}{x} p^x (1-p)^{n-x}, \quad x=0, 1, 2, \dots, n.$$

$$\binom{n}{x} = {}^nC_x = \frac{n!}{x! (n-x)!}$$

The cdf of binomial distribution:

If $X \sim \text{Bin}(n, p)$ then the cdf is given by

$$\begin{aligned} B(x; n, p) &= P(X \leq x) \\ &= \sum_{y=0}^x b(y; n, p), \quad x=0, 1, 2, \dots, n. \end{aligned}$$

Theorem: (Mean and Variance of Binomial distribution) 91

If $X \sim \text{Bin}(n, p)$, then $E(X) = np$, $V(X) = np(1-p) = npq$,
and $\sigma_X = \sqrt{npq}$, where $q = 1-p$.

Proof: If $X \sim \text{Bin}(n, p)$ then the pmf is

$$b(x; n, p) = \binom{n}{x} p^x q^{n-x}, \quad q=1-p, \quad x=0, 1, 2, \dots, n.$$

\therefore Mean, $\mu = E(X) = \sum_x x p(x)$, $p(x)$ = probability function

$$= \sum_{x=0}^n x \binom{n}{x} p^x q^{n-x} = \sum_{x=1}^n x \frac{n!}{x!(n-x)!} p^x q^{n-x}$$

$$\Rightarrow \mu = E(X) = \sum_{x=1}^n \frac{n \cdot (n-1)!}{(x-1)!(n-x)!} p^x q^{n-x}, \quad \text{if } x-1 = r$$

$$= \sum_{r=0}^{n-1} \frac{n(n-1)!}{r!(n-r-1)!} p^{r+1} q^{n-r-1}$$

$$= np \sum_{r=0}^{n-1} \binom{n-1}{r} p^r q^{n-1-r} = np (p+q)^{n-1} = np$$

$$\Rightarrow \boxed{E(X) = np}$$

$$[\because q=1-p \Rightarrow p+q=1, \quad (a+b)^n = \sum_{r=0}^n \binom{n}{r} a^r b^{n-r}]$$

To find $V(X)$, we know that $V(X) = E\{X(X-1)\} + \mu(\mu-1)$, $\mu = E(X)$.

$$\text{Now, } E\{X(X-1)\} = \sum_x x(x-1) p(x) = \sum_{x=0}^n x(x-1) \binom{n}{x} p^x q^{n-x}$$

$$= \sum_{x=2}^n x(x-1) \frac{n!}{x!(n-x)!} p^x q^{n-x}$$

$$= \sum_{x=2}^n \frac{n!}{(x-2)!(n-x)!} p^x q^{n-x}$$

$$= \sum_{r=0}^{n-2} \frac{n!}{r!(n-2-r)!} p^{r+2} q^{n-2-r}, \quad \text{let } x-2 = r$$

$$\Rightarrow E\{x(x-1)\} = n(n-1)p^2 \sum_{r=0}^{n-2} \frac{(n-2)!}{r!(n-2-r)!} p^r q^{n-2-r}$$

$$= n(n-1)p^2 (p+q)^{n-2} = n(n-1)p^2 \quad (\because p+q=1)$$

$$\therefore V(x) = E\{x(x-1)\} + \mu(\mu-1), \quad \mu = E(x)$$

$$= n(n-1)p^2 - np(np-1), \quad [\because \mu = E(x) = np]$$

$$= n^2 p^2 - np^2 - n^2 p^2 + np$$

$$= np(1-p) = npq \quad [\because q=1-p]$$

$$\therefore \sigma_x = \sqrt{V(x)} = \sqrt{npq} \quad (\text{standard deviation})$$