

## 1.2.3 Expectation of Bernoulli Distribution

⇒ What is expectation?

→ Roughly, the expectation is the average value of the random variable where each value is weighted according to its probability.

⇒ Expectation of Discrete random variables;

$$E(X) = \sum_i x_i P(X_i)$$

- $E(X)$  is the expectation value of the discrete random variable  $x$
- $x$  is the value of the discrete random variable  $x$
- $P(x)$  is the probability mass function of  $x$ .

Example: Let  $R$  be the value that comes up with you roll a fair 6-sided die.

⇒ Then the expected value of  $R$  is;

$x$	1	2	3	4	5	6
$P(x)$	$1/6$	$1/6$	$1/6$	$1/6$	$1/6$	$1/6$

$$\begin{aligned}
 \Rightarrow E(R) &= 1 \times \frac{1}{6} + 2 \times \frac{1}{6} + 3 \times \frac{1}{6} + 4 \times \frac{1}{6} \\
 &\quad + 5 \times \frac{1}{6} + 6 \times \frac{1}{6} \\
 &= \frac{1}{6} (42) = \underline{\underline{\frac{7}{2}}}
 \end{aligned}$$

e.g. for tossing a coin  $S = \{H, T\}$ .  
 $x$ : getting head  
 $x = \{0, 1\}$

$$\therefore E[X] = 1 \times 0 \times P(T) + 1 \times P(H) \\ = P(H) = \frac{1}{2}$$

⇒ Expectation of Bernoulli Distributions

$$\rightarrow P(X=1) = p \quad P(X) = p^x (1-p)^{1-x} \\ P(X=0) = 1-p$$

$$\therefore E[X] = P(X=1) \cdot 1 + P(X=0) \cdot 0 \\ = p \cdot 1 + (1-p) \cdot 0 \\ = p$$

$$\therefore \underline{E[X] = p}$$

→ Thus, the mean or expected value of a Bernoulli distribution is given by  $E[X] = p$ .

⇒ Variance of Discrete Random variables:

→ The variance can be defined as the difference of the mean of  $X^2$  and the square of the mean of  $X$ .

→ Mathematically, this statement can be written as follows:

$$\underline{\underline{Var[X] = E[X^2] - (E[X])^2}}$$



$$\therefore E[X] = \sum x_i \cdot P(x_i)$$

$$\therefore E[X^2] = \sum x_i^2 \cdot P(x_i)$$

$$E[\phi(X)] = \sum \phi(x_i) \cdot P(x_i)$$

e.g.: For variance of rolling die & getting 6.  
 $\Rightarrow$  Find the variance & standard deviation when a fair die is rolled.

$X$	1	2	3	4	5	6
$P(X)$	$1/6$	$1/6$	$1/6$	$1/6$	$1/6$	$1/6$
$X^2$	1	4	9	16	25	36

$$E[X] = \frac{7}{2} \text{ [from previous question]}$$

$$E[X^2] = \frac{1}{6} (1 + 4 + 9 + 16 + 25 + 36)$$

$$= \frac{1}{6} (n(n+1)(2n+1))$$

$$= \frac{1}{36} (6)(7)(13)$$

$$E[X^2] = \frac{91}{6}$$

$\therefore$  Variance;

$$V[X] = \frac{91}{6} - \left(\frac{7}{2}\right)^2$$

$$V[X] = \frac{91 - 49}{6} = \frac{42}{6}$$

$$= \frac{35}{2}$$

→ Using the properties of  $E[x^2]$ , we get,

$$E[x^2] = \sum x^2 \cdot P(x=x)$$

$$\therefore E[x^2] = 1^2 \cdot p + 0^2 \cdot (1-p) = \underline{\underline{p}}$$

→ Substituting this value in  $\text{Var}[x]$   

$$= E[x^2] - (E[x])^2$$

$$\therefore \text{var}[x] = p - p^2$$

$$\therefore \underline{\underline{\text{Var}[x] = p(1-p)}}$$

→ Hence, the variance of a Bernoulli distribution is  $\text{Var}[x] = p(1-p)$

GATE-2011: If the difference between expectation of the square of a random variable ( $E[x^2]$ ) and the square of the expectation of the random variable ( $(E[x])^2$ ) is denoted by  $R$ , then?

(A.)  $R=0$  (B.)  $R<0$  (C.)  $R \geq 0$  (D.)  $R>0$

$$\rightarrow R = E[x^2] - (E[x])^2 = \underline{\underline{p(1-p)}}$$

→ The difference between  $E[x^2]$  and  $(E[x])^2$  is called variance of a random variable.

→ Variance measures how far a set of numbers is spread out. (A variance of zero indicates that all the values are identical).

→ A non-zero Variance is always positive.

→ Properties of variance:

→  $\text{Var}(K) = 0$ ; when  $K$  is a constant.

→  $\text{Var}(aX + b) = \underline{\underline{a^2 \text{Var}(X)}}$ .