

3/7/19

## Random Theory of Variables :-

- \* A function from the sample space  $S$  to the real no. system  $R$  is called a Random Variables.

i.e., a mapping  $X: S \rightarrow R$  where  $S$  is the sample space and  $R$  is the Real no. system is called as RV.

- \* eg: If two coins are tossed

$$S = \{ \overset{S_1}{HH}, \overset{S_2}{HT}, \overset{S_3}{TH}, \overset{S_4}{TT} \}$$

define  $X: S \rightarrow R$  as  $X(s) = \text{"no. of heads"}$ .

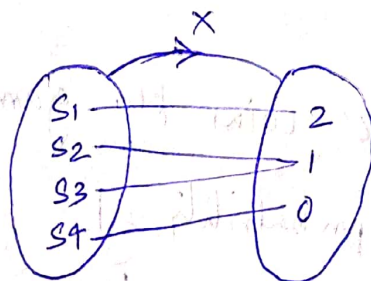
$$\text{then, } X(S_1) = 2$$

$$X(S_2) = 1$$

$$X(S_3) = 1$$

$$X(S_4) = 0$$

Therefore.



Here,  $X$  assumes the values  $\{0, 2, 1\}$  which are real no.'s. Therefore  $X$  is a Random Variable.

### \* Types of Random Variables :-

There are two types of Random Variables :-

- ① Discrete
- ② Continuous

### Discrete RV :-

- The random variable 'X' which assumes only finite values in the given interval i.e., it assumes only the set  $0, 1, 2, \dots, n$  is called a Discrete RV.
- In other words, a discrete RV is a variable which can only take a countable no of values.

eg: Rolling a die, tossing a coin

### Continuous RV :-

- The random variable 'X' which assumes the infinite value in the given interval is called a continuous RV.
- In other words, a continuous RV is a variable which the data can take infinitely many values.

eg:- Time, Temperature

### \* Discrete Probability Distribution

- Let, 'X' be a discrete RV for the sample space of tossing two coins. Then X assumes the values  $\{0, 1, 2\}$

◦ Now,

$$P(X=0) = \frac{1}{4}$$

$$P(X=1) = \frac{2}{4} = \frac{1}{2}$$

$$P(X=2) = \frac{1}{4}$$



X	0	1	2
P(X)	$\frac{1}{4}$	$\frac{1}{2}$	$\frac{1}{4}$

Now, Sum of probability = 1, Hence,  $P(X)$  is called PMF

This is called Discrete probability distribution

(PMF: probability Mass function)

① Find the DPD for the sum of the dice if two are rolled

Sol Let 'S' be the sample space of rolling 2 dice

$$n(S) = 36$$

define  $X: S \rightarrow R$  as  $X(S) = \text{"sum of 2 dice"}$

Hence 'X' assumes the values  $\{2, 3, 4, 5, 6, \dots, 12\}$

$$P(X=2) = \frac{1}{36} \quad P(X=7) = \frac{6}{36} \quad P(X=12) = \frac{1}{36}$$

$$P(X=3) = \frac{2}{36} \quad P(X=8) = \frac{5}{36}$$

$$P(X=4) = \frac{3}{36} \quad P(X=9) = \frac{4}{36}$$

$$P(X=5) = \frac{4}{36} \quad P(X=10) = \frac{3}{36}$$

$$P(X=6) = \frac{5}{36} \quad P(X=11) = \frac{2}{36}$$

X	2	3	4	5	6	7	8	9	10	11	12
P(X)	$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$	$\frac{4}{36}$	$\frac{5}{36}$	$\frac{6}{36}$	$\frac{5}{36}$	$\frac{4}{36}$	$\frac{3}{36}$	$\frac{2}{36}$	$\frac{1}{36}$

This is the req. probability distribution.

Q2

Find the DPD for the min. no of the 2 dice i.e.,  
 $X: S \rightarrow R$  as  $X(s) = \min\{a, b\}$  if 2 dice are rolled.

Let 'S' be the sample space of rolling 2 dice.

$$n(S) = 36$$

Define  $X: S \rightarrow R$  as  $X(s) = \min\{a, b\}$ .

Hence  $X$  assumes values.

$$\{1, 2, 3, 4, 5, 6\}$$

$$P(X=1) = \frac{1}{36}$$

$$P(X=2) = \frac{2}{36}$$

$$P(X=3) = \frac{3}{36}$$

$$P(X=4) = \frac{4}{36}$$

$$P(X=5) = \frac{2}{36}$$

$$P(X=6) = \frac{1}{36}$$

X	1	2	3	4	5	6
P(X)	$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$	$\frac{4}{36}$	$\frac{2}{36}$	$\frac{1}{36}$



Note! —

④

Consider the discrete probability distribution in the following table :

$x$	$x_1$	$x_2$	$\dots$	$x_n$
$P(x)$	$p_1$	$p_2$	$\dots$	$p_n$

①  $\sum p_i = 1$  [ $P(x)$  is called p.m.f]

② Mean  $\mu = \sum x_i P(X=x_i) = E(x) = \text{Expectation of } x$

③ Variance,  $\sigma^2 = \sum (x_i - \mu)^2 p_i$

$$\sigma^2 = \sum x_i^2 p_i - \mu^2$$

The positive sq. root of variance is called 'Standard Deviation'.

~~MD~~ Mean deviation,  $MD = \sum |x_i - \mu| P(X=x_i)$

① A random variable  $x$  has the following probability distribution

$x$	0	1	2	3	4	5	6	7
$P(x)$	0	$k$	$2k$	$2k$	$3k$	$k^2$	$2k^2$	$7k^2 + k$

(i) Find the value of  $k$

(ii)  $P(X < 6)$

(iii)  $P(X \geq 6)$

(iv)  $P(0 < X < 5)$

## \* Continuous Probability Distribution

Let  $X$  be a continuous random variable, since the continuous random variable  $X$  assumes the infinite values in the given interval. i.e., it assumes the values  $(-\infty \text{ to } \infty)$

Then, the probability function  $f(x)$  is defined on  $f: (-\infty, \infty)$  such that,

$$\int_{-\infty}^{\infty} f(x) dx = 1 \quad (\text{p.d.f})$$

If the above satisfies, then the fn  $f(x)$  is called probability density function.

$$\text{Mean, } \mu = \int_{-\infty}^{\infty} xf(x) dx$$

$$\begin{aligned} \text{Variance, } \sigma^2 &= \int_{-\infty}^{\infty} (x-\mu)^2 f(x) dx \\ &= \int_{-\infty}^{\infty} x^2 f(x) dx - \mu^2 \end{aligned}$$

$$\text{Mean deviation, M.D} = \int_{-\infty}^{\infty} |x-\mu| f(x) dx.$$

Ex 19

$$\textcircled{*} \text{ if } P(a < x < b) = \int_a^b f(x) dx.$$

Distribution fn:  $(x)$  Cumulative distribution fn.

$$F(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt.$$

$$\boxed{f(x) = \frac{d}{dx} F(x)}$$

$x = \text{const}$   
 $-\infty$  is the lower limit of fn.

(Ans in terms of  $x$ )