

Random Variable :-

A rule that assigns a real number to each outcome is called random variable. The rule is nothing but a function of the variable, say, X that assigns a unique value to each outcome of the random experiment.

Discrete Random Variable :-

A random variable X , which can take only a finite number of values in an interval of the domain is called discrete random variable.

- Ex. ① Number appearing on top of a dice when it is thrown.
- ② The number of telephone calls received per day.
- ③ Number of mistakes in a page.
- ④ Number of defective items in a lot.

Discrete Probability Distribution :-

If a random variable x can assume a discrete set of values say x_1, x_2, \dots, x_n with respect to probabilities p_1, p_2, \dots, p_n such that

$$p_1 + p_2 + \dots + p_n = 1 \quad \text{i.e. } \sum_{i=1}^n p_i = 1$$

then occurrence of values x_i with respective probability p_i is called the discrete probability distribution of x .

For example, In a throw of a pair of dice the sum X is discrete random variable which is an integer between 2 and 12 with probabilities $P(X)$,

sample space are

~~(1,1), (1,2), (1,3), (1,4), (1,5), (1,6)~~

~~(2,1), (2,2), (2,3), (2,4), (2,5), (2,6)~~

~~(3,1), (3,2), (3,3), (3,4), (3,5), (3,6)~~

~~(4,1), (4,2), (4,3), (4,4), (4,5), (4,6)~~

~~(5,1), (5,2), (5,3), (5,4), (5,5), (5,6)~~

~~(6,1), (6,2), (6,3), (6,4), (6,5), (6,6)~~

$P(X)$ given as

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 constitute a discrete probability distribution.

Probability Mass Function (Pmf) :-

Probability function of probability mass function of a random variable X which is an integer mathematical function $p(x)$ which gives the probabilities corresponding to different possible discrete set of values say x_1, x_2, \dots, x_n of variable x . i.e. $p(x_i) = p(x=x_i)$

= probability that random variable x assumes values x_i

The function $p(x)$ satisfies the condition :-

$$(i) \quad p(x_i) \geq 0 \quad (ii) \quad \sum p(x_i) = 1$$

Cumulative Distribution function :-

If X is a random variable then $P(X \leq x)$ is called the cumulative distribution (cdf) denoted by $F(x) = P(X \leq x)$.

Expectation of a discrete Random Variable :-

If x is discrete random variable which assumes the discrete set of values x_1, x_2, \dots, x_n with respective probabilities p_1, p_2, \dots, p_n then the expectation or expected value of x is denoted by $E(x)$ and defined as

$$E(x) = x_1 p_1 + x_2 p_2 + x_3 p_3 + \dots + x_n p_n = \sum_{i=1}^n x_i p_i$$

similarly the expected value of x^2 is

$$\text{defined as } E(x^2) = \sum_{i=1}^n x_i^2 p_i$$

Properties :-

- | | |
|----------------------|---|
| (i) $E(a) = a$ | (2) $E(X \pm Y) = E(X) \pm E(Y)$ |
| (ii) $E(ax) = aE(X)$ | (3) $E(XY) = E(X)E(Y)$ if X and Y are independent |
| (iii) $E(X-a) = 0$ | (4) $E(Y) = aE(X) + b$ [if $y = ax + b$] |

Variance and standard Deviation of discrete R.V.

The variable of discrete random variable X is expected value of $(X - \mu)^2$ where μ is mean of variable X

$$\begin{aligned}
 \text{Var}(X) &= V(x) = E(X - \mu)^2 = \sum_{i=1}^n p_i (x_i - \mu)^2 \\
 &= \sum p (x - \mu)^2 = \sum p (x^2 + \mu^2 - 2\mu x) \\
 &= \sum p x^2 + \mu^2 \sum p - 2\mu \sum p x \\
 &= \sum p x^2 + \mu^2 \cdot 1 - 2\mu \cdot \mu \\
 &= \sum p x^2 + \mu^2 - 2\mu^2 \\
 &= \sum p x^2 - \mu^2
 \end{aligned}$$

$\left[\begin{array}{l} \sum p = 1 \\ \mu = E(x) \\ = \sum p x \end{array} \right]$

$$\boxed{\text{Var}(X) = E(X^2) - \{E(X)\}^2}.$$

The standard deviation (SD) of a random variable x is denoted by s , then

$$\boxed{SD(X) = \sqrt{\text{Var}(X)}}.$$

Ex. ① A pair of two coins is tossed, what is the expected value?

Soln. Expected value or mean value = $E(X) = \bar{x} = \sum_{i=1}^n p_i x_i$

In tossing of two coins, probability distribution is represented in tabular form as follows

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

HH	HT	TH	TT
X=2	X=1	X=1	X=0

$$\therefore E(X) = \frac{1}{4} \times 0 + \frac{1}{2} \times 1 + \frac{1}{4} \times 2 = 1$$

$$\boxed{E(X) = 1}$$

As the probability of getting no head, one head and two heads is respectively $\frac{1}{4}, \frac{1}{2}, \frac{1}{4}$.

In the tossing of two coins, we have

$$\sum p x^2 = \frac{1}{4} \cdot (0)^2 + \frac{1}{2} \cdot (1)^2 + \frac{1}{4} \cdot 2^2 = \frac{1}{2} + 1 = \frac{3}{2}$$

$$\therefore \text{Variance} = \sigma^2 = \sum p x^2 - \bar{x}^2 = \frac{3}{2} - 1^2 \\ = \frac{1}{2}.$$

(7)

Ex. ② A pair of dice is thrown together, find the expected value.

Soln. In a throw of pair of dice the sum (X) is a discrete random variable which is an integer between 2 and 12 with the probability as given below:

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

$$\begin{aligned} \therefore \text{Expected Value} &= E(X) = \sum x P(x) \\ &= \frac{2}{36} + \frac{3}{36} + \frac{4}{36} + \frac{5}{36} + \frac{6}{36} + \frac{5}{36} + \frac{4}{36} + \frac{3}{36} + \frac{2}{36} + \frac{1}{36} \\ &\quad + \frac{3}{36} + \frac{2}{36} + \frac{1}{36} \\ &= \frac{252}{36} = 7. \end{aligned}$$

In case of a pair of dice we have

$$\begin{aligned} \sum p x^2 &= \frac{1}{36} \cdot 4 + \frac{2}{36} \cdot 9 + \frac{3}{36} \cdot 16 + \frac{4}{36} \cdot 25 + \frac{5}{36} \cdot 36 + \frac{6}{36} \cdot 49 \\ &\quad + \frac{5}{36} \cdot 64 + \frac{4}{36} \cdot 81 + \frac{3}{36} \cdot 100 + \frac{2}{36} \cdot 121 + \frac{1}{36} \cdot 144 \\ &= \frac{1}{36} [4 + 18 + 48 + 100 + 180 + 294 + 320 + 324 + 300 \\ &\quad + 242 + 144] \\ &= \frac{1}{36} [1974] = \frac{329}{6} \end{aligned}$$

$$\text{Variance} = \sigma^2 = \frac{329}{6} - 7^2$$

$$= \frac{329 - 294}{6}$$

$$= \frac{35}{6}$$

$$\text{Standard deviation} = \sigma = \sqrt{\frac{35}{6}}$$

Ex. ③ A random variable X has the following distribution

X	-2	-1	0	1	2	3
$P(X)$	0.1	K	0.2	$2K$	0.3	K

Determine : (i) k , (ii) Mean, (iii) Variance.

Solⁿ :- (i) since $\sum p(x) = 1$

$$\Rightarrow 0.1 + k + 0.2 + 2k + 0.3 + k = 1$$

$$\Rightarrow 0.6 + 4k = 1$$

$$\Rightarrow 4k = 0.4$$

$$\Rightarrow \boxed{k = 0.1}$$

Now the probability distribution of the random variable X is

X	-2	-1	0	1	2	3
$P(X)$	0.1	0.1	0.2	0.2	0.3	0.1

④ find the mean and variance of uniform probability distribution $f(x) = \frac{1}{n}$ for $x = 1, 2, \dots, n$. ⑤

Soln.

X	1	2	3	\dots	n
$P(X)$	$\frac{1}{n}$	$\frac{1}{n}$	$\frac{1}{n}$	\dots	$\frac{1}{n}$

$$\text{Mean} = E(X) = \frac{1}{n} + \frac{2}{n} + \frac{3}{n} + \dots + \frac{n}{n}$$

$$= \frac{1+2+3+\dots+n}{n}$$

$$= \frac{1}{n} \left[\sum n \right]$$

$$= \frac{1}{n} \left[\frac{n(n+1)}{2} \right]$$

$$= \frac{n+1}{2}$$

$$\text{Variance} = E(X^2) - \{E(X)\}^2 = \frac{1^2 + 2^2 + \dots + n^2}{n} - \left[\frac{n+1}{2} \right]^2$$

$$= \frac{1}{n} \cdot \frac{n(n+1)(2n+1)}{6} - \frac{(n+1)^2}{4}$$

$$= \frac{n+1}{2} \left[\frac{2n+1}{3} - \frac{(n+1)}{2} \right]$$

$$= \left(\frac{n+1}{2} \right) \left[\frac{4n+2-3n-3}{6} \right]$$

$$= \frac{(n+1)}{2} \left[\frac{(n-1)}{6} \right] = \frac{n^2-1}{12}$$

(5) For discrete probability distribution

x	0	1	2	3	4	5	6	7
f	0	k	$2k$	$2k$	$3k$	k^2	$2k^2$	$7k^2 + k$

Determine (i) k (ii) mean (iii) variance (iv) smallest value of x such that $P(X \leq x) > \frac{1}{2}$.

Soln. (i) $\sum f(x) = 1$

$$\Rightarrow 0 + k + 2k + 2k + 3k + k^2 + 2k^2 + 7k^2 + k = 1$$

$$\Rightarrow 10k^2 + 9k - 1 = 0$$

$$\Rightarrow 10k^2 + 10k - k - 1 = 0$$

$$\Rightarrow 10k(k+1) - 1(k+1) < 0$$

$$\Rightarrow (k+1)(10k-1) = 0$$

$$\Rightarrow k = -1, k = \frac{1}{10}$$

$\therefore k = \frac{1}{10}$ since probability can never be negative.

x	0	1	2	3	4	5	6	7
f	0	$\frac{1}{10}$	$\frac{2}{10}$	$\frac{2}{10}$	$\frac{3}{10}$	$\frac{1}{100}$	$\frac{2}{100}$	$\frac{7}{100} + \frac{1}{10} = \frac{17}{100}$

$$\begin{aligned}
 \text{(ii) Mean} &= \sum xf(x) = 0 \times 0 + 1 \times \frac{1}{10} + 2 \times \frac{2}{10} + 3 \times \frac{2}{10} + 4 \times \frac{3}{10} + 5 \times \frac{1}{100} \\
 &\quad + 6 \times \frac{2}{100} + 7 \times \frac{17}{100} \\
 &= \frac{10 + 40 + 60 + 120 + 5 + 12 + 119}{100} = \frac{366}{100} = 3.66
 \end{aligned}$$

(11)

$$\text{(iii) Variance} = E(x^2) - \{E(x)\}^2$$

$$\sigma^2 = \left[0 + 1^2 \times \frac{1}{10} + 2^2 \times \frac{2}{10} + 3^2 \times \frac{2}{10} + 4^2 \times \frac{3}{10} + 5^2 \times \frac{1}{10} + 6^2 \times \frac{2}{100} + 7^2 \times \frac{17}{100} \right] - \{3.66\}^2$$

$$= \frac{10 + 80 + 180 + 480 + 25 + 72 + 833}{100} - 13.3956$$

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$$= \frac{1680}{100} - 13.3956$$

$$= 16.8 - 13.3956 = 3.40$$

(iv)

$$P(x \leq 0) = f(0) = 0$$

$$P(x \leq 1) = f(0) + f(1) = 0 + 0.1 = 0.1$$

$$P(x \leq 2) = 0 + 0.1 + 0.2 = 0.3$$

$$P(x \leq 3) = 0.3 + 0.2 = 0.5 = \frac{1}{2}$$

$$P(x \leq 4) = 0.5 + 0.3 = 0.8$$

\therefore Smallest value of x such that $P(x \leq x) > \frac{1}{2}$

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Moment of Random Variable :-

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Expectation of random variable

Let X be a random variable defined over the sample space Ω .

Then the expectation of X is defined as the weighted average of all possible values of X .

where the weights are the probabilities of the corresponding outcomes.

Mathematically, if X is a discrete random variable taking values x_1, x_2, \dots, x_n with probabilities p_1, p_2, \dots, p_n respectively, then the expectation of X is given by:

$$E(X) = x_1 p_1 + x_2 p_2 + \dots + x_n p_n$$

If X is a continuous random variable with probability density function $f(x)$, then the expectation of X is given by:

$$E(X) = \int x f(x) dx$$

where the integral is taken over the entire range of X .

The expectation of a random variable is also known as the first moment of the distribution.

Properties of Expectation:

1. Linearity: If a and b are constants, then $E(aX + b) = aE(X) + b$.

2. Additivity: If X_1, X_2, \dots, X_n are independent random variables, then $E(X_1 + X_2 + \dots + X_n) = E(X_1) + E(X_2) + \dots + E(X_n)$.

3. Multiplicativity: If a is a constant, then $E(aX) = aE(X)$.

4. Non-negativity: The expectation of a non-negative random variable is always non-negative.

5. Uniqueness: The expectation of a random variable is unique and is determined by its probability distribution.

6. Invariance under shift: If X is a random variable and a is a constant, then $E(X + a) = E(X) + a$.

7. Invariance under scaling: If X is a random variable and a and b are constants, then $E(aX + b) = aE(X) + b$.

8. Monotonicity: If $X_1 \leq X_2$ for all i , then $E(X_1) \leq E(X_2)$.

9. Jensen's inequality: If X is a random variable and g is a convex function, then $E(g(X)) \geq g(E(X))$.

10. Fubini's theorem: If X is a random variable and Y is a non-negative random variable, then $E(XY) = E(X)E(Y)$.

11. Covariance: If X and Y are random variables, then $Cov(X, Y) = E(XY) - E(X)E(Y)$.

12. Variance: If X is a random variable, then $V(X) = E(X^2) - E(X)^2$.

Moment of Random Variable

12
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The "moments" of a random variable (or its distribution) are expected values of powers or related functions of the random variable.

The r th moment of x is $\mu_r^1 = E(x^r) = \sum x^r p(x=x)$

In particular, the first moment is mean

$$\mu_x = E(x) = \mu_1^1$$

The r th central moment of x is $\mu_r = E(x - \mu_x)^r$

In particular, the second central moment is Variance $\sigma_x^2 = \text{Var } x = E(x - \mu_x)^2$.

Relation Between r th moment of random variable and r th central moment of random variable

$$\mu_1 = E(x - \mu_x) = E(x) - E(\mu_x) = E(x) - E(x) = 0$$

$$\begin{aligned}\mu_2 &= E(x - \mu_x)^2 = E(x^2 + \mu_x^2 - 2x\mu_x) \\ &= E(x^2) + \mu_x^2 - 2\mu_x E(x) \\ &= E(x^2) + \mu_x^2 - 2\mu_x^2 \\ &= E(x^2) - \mu_x^2\end{aligned}$$

$$\mu_2 = \mu_2^1 - \mu_1^1 \cdot \mu_1^1 \quad [\text{first central moment is mean}]$$

$$\begin{aligned}\mu_3 &= E(x - \mu_x)^3 = E(x^3 + \mu_x^3 - 3x^2\mu_x + 3\mu_x^2 x) \\ &= E(x^3) + E(\mu_x^3) - 3E(x^2)\mu_x + 3\mu_x^2 E(x)\end{aligned}$$

$$u_3 = E(x^3) - u_1'^3 - 3u_1' u_2' + 3u_1'^2 u_2'$$

$$= u_3' - 3u_1' u_2' + 2u_2'^3$$

$$u_4 = u_4' + 6u_1'^2 u_2' - 4u_1' u_2' - 3u_1'^4$$

Let x be a discrete random variable having probability mass function

$$p_x(x) = \begin{cases} \frac{1}{2} & \text{if } x=1 \\ \frac{1}{3} & \text{if } x=2 \\ \frac{1}{6} & \text{if } x=3 \\ 0 & \text{otherwise} \end{cases}$$

find the third moment of x ,

$$\text{Third moment} = E(x^3)$$

$$= \sum x^3 p_x(x)$$

$$= \frac{1}{2} \cdot 1^3 + \frac{1}{3} \cdot 2^3 + \frac{1}{6} \cdot 3^3 + 0$$

$$= \frac{1}{2} + \frac{8}{3} + \frac{27}{6}$$

$$= \frac{3+16+27}{6}$$

$$= \frac{46}{6}$$

$$= \frac{23}{3}$$

Probability Density Function :-

14

The probability density function (P.d.f.) of a random variable X usually denoted by $f_x(x)$ or simply $f(x)$ has following properties :-

$$f(x) \geq 0, \quad -\infty < x < \infty$$

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

A continuous random variables X has a probability density function defined by

$$f(x) = \begin{cases} \frac{1}{16} (3+x)^2 & \text{if } -3 \leq x \leq -1 \\ \frac{1}{16} (6-2x^2) & \text{if } -1 \leq x \leq 1 \\ \frac{1}{16} (3-x)^2 & \text{if } 1 < x \leq 3 \\ 0 & \text{elsewhere.} \end{cases}$$

Verify $f(x)$ is a density function, and also find the mean of the random variable X .

Since $f(x)$ is density function $\int_{-\infty}^{\infty} f(x) dx = 1$.

$$\begin{aligned} \int_{-\infty}^{\infty} f(x) dx &= \int_{-\infty}^{-3} f(x) dx + \int_{-3}^{-1} f(x) dx + \int_{-1}^{1} f(x) dx + \int_{1}^{3} f(x) dx + \int_{3}^{\infty} f(x) dx \\ &= \int_{-\infty}^{-3} 0 dx + \int_{-3}^{-1} \frac{1}{16} (3+x)^2 dx + \int_{-1}^{1} \frac{1}{16} (6-2x^2) dx \\ &\quad + \int_{1}^{3} \frac{1}{16} (3-x)^2 dx + \int_{3}^{\infty} 0 dx \\ &= \frac{1}{16} \left[\left\{ \frac{(3+x)^3}{3} \right\}_{-3}^{-1} + \left\{ 6x - 2 \cdot \frac{x^3}{3} \right\}_{-1}^1 + \left\{ \frac{(3-x)^3}{3} \right\}_1^3 \right] \end{aligned}$$

$$= \frac{1}{16} \left[\left(\frac{8}{3} - 0 \right) + \left(6 - \frac{2}{3} \right) - \left(-6 + \frac{2}{3} \right) - \left(0 - \frac{8}{3} \right) \right]$$

$$= \frac{1}{16} \left[\frac{8}{3} + 6 - \frac{2}{3} + 6 - \frac{2}{3} + \frac{8}{3} \right]$$

$$= \frac{1}{16} \left[\frac{16 - 4 + 36}{3} \right] = \frac{1}{16} \times \frac{48}{3} = \frac{48}{48} = 1$$

$\therefore \boxed{\int_{-\infty}^{\infty} f(x) dx = 1}$

Hence $f(x)$ is a density function.

Mean of the random variable X is

$$\begin{aligned} E(x) &= \int_{-\infty}^{\infty} x f(x) dx \\ &= \frac{1}{16} \int_{-3}^{-1} x(3+x)^2 dx + \frac{1}{16} \int_{-1}^1 x(6-2x^2) dx + \frac{1}{16} \int_1^3 x(3-x)^2 dx \\ &= \frac{1}{16} \left[\int_{-3}^{-1} x(9+6x+x^2) dx + 0 + \int_1^3 x(9+x^2-6x) dx \right] \\ &= \frac{1}{16} \left[\int_{-3}^{-1} (9x+6x^2+x^3) dx + \int_1^3 (9x+x^3-6x^2) dx \right] \\ &= \frac{1}{16} \left[\left(\frac{9x^2}{2} + \frac{6x^3}{3} + \frac{x^4}{4} \right) \Big|_{-3}^1 + \left(\frac{9x^2}{2} + \frac{x^4}{4} - \frac{6x^3}{3} \right) \Big|_1^3 \right] \\ &= \frac{1}{16} \left[\left(\frac{9}{2} - 2 + \frac{1}{4} - \frac{81}{2} + 54 - \frac{81}{4} \right) + \left(\frac{81}{2} + \frac{81}{4} - 54 - \frac{9}{2} - \frac{1}{4} + 2 \right) \right] = 0 \end{aligned}$$

Ex. ② A continuous random variable x has

$$f(x) = \begin{cases} \frac{1}{2}(x+1) & \text{for } -1 < x < 1 \\ 0 & \text{elsewhere} \end{cases}$$

represents the density, find the mean and standard deviation of x .

If $f(x)$ is density function, then it satisfies

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

$$\begin{aligned} \int_{-\infty}^{\infty} f(x) dx &= \int_{-\infty}^{-1} f(x) dx + \int_{-1}^1 f(x) dx + \int_1^{\infty} f(x) dx \\ &= \int_{-\infty}^{-1} 0 \cdot dx + \int_{-1}^1 \frac{1}{2}(x+1) dx + \int_1^{\infty} 0 \cdot dx \end{aligned}$$

$$= \frac{1}{2} \left[\frac{x^2}{2} + x \right]_{-1}^1$$

$$= \frac{1}{2} \left[\left(\frac{1}{2} + 1 \right) - \left(\frac{1}{2} - 1 \right) \right]$$

$$= \frac{1}{2} [2]$$

$$= 1.$$

Hence $f(x)$ is a density function.

Mean of the random variable X is

$$\begin{aligned}
 E(X) &= \int_{-\infty}^{\infty} x f(x) dx \\
 &= \int_{-\infty}^{-1} 0 \cdot dx + \int_{-1}^1 x \cdot \frac{1}{2}(x+1) dx + \int_1^{\infty} 0 \cdot dx \\
 &= \frac{1}{2} \int_{-1}^1 (x^2+x) dx \\
 &= \frac{1}{2} \left[\frac{x^3}{3} + \frac{x^2}{2} \right]_{-1}^1 \\
 &= \frac{1}{2} \left[\left(\frac{1}{3} + \frac{1}{2} \right) - \left(-\frac{1}{3} + \frac{1}{2} \right) \right] \\
 &= \frac{1}{2} \left[\frac{1}{3} + \frac{1}{2} + \frac{1}{3} - \frac{1}{2} \right] \\
 &= \frac{1}{2} \times \cancel{\frac{2}{3}} = \frac{1}{3}.
 \end{aligned}$$

Therefore the mean of the random variable X is $\frac{1}{3}$.

\therefore The variance of the random variable X is $V(X) = E(X^2) - [E(X)]^2$

Now $E(X^2) = \int_{-\infty}^{\infty} x^2 f(x) dx$

$$= \int_{-1}^1 x^2 \cdot \frac{1}{2}(x+1) dx$$

$$E(x^2) = \frac{1}{2} \left\{ \int_{-1}^1 (x^3 + x^2) dx \right\}$$

$$= \frac{1}{2} \left[0 + 2 \int_0^1 x^2 dx \right]$$

x^3 is odd function
 x^2 is even function

$$= \left(\frac{x^3}{3} \right)_0^1 = \frac{1}{3}$$

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

$$= \frac{1}{3} - \left(\frac{1}{3}\right)^2 = \frac{1}{3} - \frac{1}{9}$$

$$\text{Var}(x) = \frac{3-1}{9} = \frac{2}{9}$$

$$\therefore \text{standard deviation of } x = \frac{\sqrt{2}}{3}$$

If the probability density function

$$f(x) = \begin{cases} kx^2 & \text{if } 0 \leq x \leq 3 \\ 0 & \text{elsewhere} \end{cases}$$

find the value 'k' and find the probability between $x = \frac{1}{2}$ and $x = \frac{3}{2}$.

$$\text{from the given data, } f(x) = \begin{cases} kx^2, & \text{for } x \leq 3 \\ 0, & \text{elsewhere} \end{cases}$$

If $f(x)$ is a density function, then it satisfies $\int_{-\infty}^{\infty} f(x) dx = 1$.

$$\Rightarrow \int_{-\infty}^0 f(x)dx + \int_0^3 f(x)dx + \int_3^{\infty} f(x)dx = 1.$$

$$\Rightarrow k \cdot \int_0^3 x^3 dx = 1$$

$$\Rightarrow k \left[\frac{x^4}{4} \right]_0^3 = 1$$

$$\Rightarrow k \left[\frac{81}{4} \right] = 1$$

$$\Rightarrow \boxed{k = \frac{4}{81}}$$

$$\text{Now, } f(x) = \begin{cases} \frac{4}{81}x^3 & \text{if } 0 \leq x \leq 3 \\ 0 & \text{elsewhere} \end{cases}$$

$$\begin{aligned}
 (P) \quad P\left(\frac{1}{2} \leq x \leq \frac{3}{2}\right) &= \int_{1/2}^{3/2} \frac{4}{81} x^3 dx \\
 &= \frac{4}{81} \left[\frac{x^4}{4} \right]_{1/2}^{3/2} = \frac{4}{81} \left[x^4 \right]_{1/2}^{3/2} \\
 &= \frac{1}{81} \left[\frac{81}{16} - \frac{1}{16} \right] \\
 &= \frac{1}{81} \times \frac{80}{16} = \frac{5}{81} = 0.0617,
 \end{aligned}$$

Let X be a discrete random variable having probability mass function.

$$P_x(x) = \begin{cases} \frac{1}{2} & \text{if } x=1 \\ \frac{1}{3} & \text{if } x=2 \\ \frac{1}{6} & \text{if } x=3 \\ 0 & \text{otherwise} \end{cases}$$

Find the third moment of X .

$$\text{Third moment} = E(X^3) = \sum x^3 P_x(x)$$

$$= \frac{1}{2} \cdot 1^3 + \frac{1}{3} \cdot 2^3 + \frac{1}{6} \cdot 3^3$$

$$= \frac{1}{2} + \frac{8}{3} + \frac{27}{6}$$

$$= \frac{3+16+27}{6} = \frac{46}{6} = \frac{23}{3}$$

Let X be a discrete random variable with probability mass function

$$P_x(x) = \begin{cases} \frac{3}{4} & \text{if } x=1 \\ \frac{1}{4} & \text{if } x=2 \\ 0 & \text{otherwise} \end{cases}$$

Find the third central moment of X .

$$E(x) = \mu_x = \sum x P_x(x) = 1 \times \frac{3}{4} + 2 \times \frac{1}{4} = \frac{5}{4}$$

The third central moment of X can be

Computed as follows :-

$$\begin{aligned} E(x - \mu_x)^3 &= \sum \left(x - \frac{5}{4} \right)^3 P_x(x) \\ &= \left(1 - \frac{5}{4} \right)^3 \times \frac{3}{4} + \left(2 - \frac{5}{4} \right)^3 \times \frac{1}{4} \\ &= \left(-\frac{1}{4} \right)^3 \times \frac{3}{4} + \left(\frac{3}{4} \right)^3 \times \frac{1}{4} \end{aligned}$$

$$= \left[-\frac{3}{64} + \frac{27}{64} \right]^{\frac{1}{4}} = \frac{\frac{3}{6}}{\frac{24}{64}} \times \frac{1}{4} = \frac{3}{32}$$

22