

①

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

It is a
Prob. Dist.

$$1. 0 \leq P(E) \leq 1$$

$$2. \sum_i P(E_i) = 1$$

②

X	1	2	3
P(X)	$-\frac{1}{4}$	$\frac{3}{4}$	$\frac{2}{4}$

Not a Prob. Dist.

③

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

Not a prob Dist.

Probability Function or Probability Mass Function (pmf)

Probability function or pmf of a random variable (R.V.)

X is a mathematical function $p(x)$ which gives the probability corresponding to different possible discrete set of values say x_1, x_2, \dots, x_n of variable x .

$$\text{i.e. } p(x_i) = p(x = x_i)$$

The fn $p(x)$ must satisfy the conditions

$$\textcircled{1} \quad p(x_i) \geq 0$$

$$\textcircled{2} \quad \sum p(x_i) = 1$$

Cumulative Distribution Function (Distribution Function)

If X is a R.V. the $P(X \leq x)$ is called the cumulative distribution function (cdf) and is denoted $F(x)$.

$$F(x) = P(X \leq x)$$

Expectation of a Discrete R.V.

If X is a R.V. 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 is defined as

$$E[X] = \sum_{i=1}^n x_i p_i = x_1 p_1 + x_2 p_2 + \dots + x_n p_n$$

$$E[X^2] = \sum_{i=1}^n x_i^2 p_i$$

$$E[X^m] = \sum_i x_i^m p_i$$

Properties

$X, Y \leftarrow \text{R.V.}$

$a, b \leftarrow \text{constants.}$

$\mu \leftarrow \text{mean}$

① $E[a] = a$

② $E[ax] = a E[x]$

③ $E[X - \mu] = 0$

} Try to prove it.

$$\textcircled{4} \quad E[X \pm Y] = E[X] \pm E[Y]$$

$$\textcircled{5} \quad E[XY] = E[X]E[Y] \quad \text{if } X \text{ \& } Y \text{ are independent R.V.}$$

$$\textcircled{6} \quad \begin{aligned} \text{If } Z &= aX + b \quad \text{then} & E[Z] &= E[aX + b] \\ & \text{or} & & \\ Z &= aX + b & & = aE[X] + b \end{aligned}$$

Variance and Standard Deviation

The variable of discrete R.V. X is expected value of

$(X - \mu)^2$ where μ is mean of the variable X .

$$\overset{\sigma^2}{\curvearrowright} \text{Var}(X) = E[(X - \mu)^2] \quad \left| \text{Also denoted by } \sigma^2 \right.$$

$$\text{SD}(X) = \sqrt{\text{Var}(X)} = \sqrt{E[X^2] - (E[X])^2}$$

$$\curvearrowright \sigma$$

* Q. Prove that $\text{Var}(X) = E[X^2] - (E[X])^2$

Q. Find the expected value of getting head when a pair of coins is tossed.

Solⁿ: $S = \{HH, HT, TH, TT\}$

Let, $X = \text{no. of heads}$

Possible values of X are — $X = 0, 1, 2$

Probability distribution table will be given by

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

Expected value of getting head, $E[X] = (0 \times \frac{1}{4}) + (1 \times \frac{2}{4}) + (2 \times \frac{1}{4})$
 $= 1$

Alternate

$$E[X] = \sum_i x_i p_i = x_1 p_1 + x_2 p_2 + x_3 p_3$$

$$= (0 \cdot \frac{1}{4}) + (1 \cdot \frac{2}{4}) + (2 \cdot \frac{1}{4}) = 1.$$

way ahead →

$$\text{var}(X) = E[X^2] - (E[X])^2$$

$$= \left(\sum_i x_i^2 p_i \right) - (1)^2$$

$$= (x_1^2 p_1 + x_2^2 p_2 + x_3^2 p_3) - 1$$

$$= \left[(0^2 \cdot \frac{1}{4}) + (1^2 \cdot \frac{2}{4}) + (2^2 \cdot \frac{1}{4}) \right] - 1$$

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

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

(positive)

$$SD(X) = \sqrt{\text{Var}(X)} = \sqrt{Y_2}$$

$$= 0.7071$$