

Random variables

1. Random Variables

A real number associated with outcome of an experiment is known as a random variable.

Example(1)

Suppose two fair coins are tossed. Then $S = \{HH, HT, TH, TT\}$

$$X = \begin{cases} 0, & \text{if at least one head occurs} \\ 1, & \text{otherwise} \end{cases}$$

Outcome	HH	HT	TH	TT
X	0	0	0	1

Let X be the random variable corresponding to number of heads. Then X takes values 0,1,2.

Outcome	HH	HT	TH	TT
X	2	1	1	0

Range of random variable

The set of all real numbers of a random variate X is called range of X .

In **example(1)**, Range of $X=\{0,1,2\}$

Example(3)

Consider the random experiment of throwing a dice twice.

The corresponding sample space is

$$S=\{(a,b) \mid a=1,2,3,4,5,6 \text{ and } b=1,2,3,4,5,6\}$$

Let $X=a+b$ be a random variable, then

$$\text{Range of } X=\{2,3,4,5,6,7,8,9,10,11,12\}$$

Over the same sample space, we can define

$$Y = \begin{cases} 1 & \text{if } a + b \text{ is even} \\ -1 & \text{if } a + b \text{ is odd} \end{cases}$$

Range of $Y = \{-1, 1\}$

Thus infinitely many random variables can be defined on a given sample space.

2. Discrete and Continuous random variable

If a random variable X takes at most a countable number of values $x_1, x_2, x_3, x_4, \dots, x_n$, then it is called a discrete random variable (DRV).

Examples: (1) and (2).

A random variable X is said to be a continuous random variable (CRV) if it can take any value in an interval which may be finite or infinite.

Example: weight of articles

3. Discrete Probability distribution

Let X be a discrete random variable assuming the values $x_1, x_2, x_3, x_4, \dots, x_n$. With each possible outcome x_i we associate a number $p_i = P(X=x_i) = P(x_i)$ called the probability of x_i .

Then $P(x_i)$ is called the **probability mass function** (PMF) of the random variable X if the following conditions are satisfied.

- (i) $P(x_i) \geq 0 \quad \forall i$
- (ii) $\sum P(x_i) = 1$

The set $\{P(x_i)\}$ is called the probability distribution of the random variable.

Example:

Suppose two fair coins are tossed. Let X be the random variable corresponding to number of heads.

Outcome	HH	HT	TH	TT
X	2	1	1	0

The probability distribution for number of heads is given by

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

Clearly $P(x_i) \geq 0 \quad \forall i$ and $\sum P(x_i) = 1$

$$E(X) = 0 + (1/2) + (1/2) = 1 = \text{Mean}????$$

Example:

Suppose **THREE** fair coins are tossed. Let X be the random variable corresponding to number of heads.

Range of $X = \{0, 1, 2, 3\}$

$S = \{HHH, HHT, HTH, THH, TTT, TTH, THT, HTT\}$

The probability distribution for number of heads is given by

X	0 TTT	1 TTH, THT, HTT	2 HHT, HTH, THH	3 HHH
$P(X)$	1/8	3/8	3/8	1/8

$$\begin{aligned} \text{Mean} = E(X) &= (3/8) + (3/4) + (3/8) \\ &= (3+6+3)/8 = 12/8 = 1.5 \end{aligned}$$

4. Continuous Probability distribution

Let X be a continuous random variable assuming values x over an interval.

We assign a real number $f(x)$ satisfying the conditions

- (i) $f(x) \geq 0 \quad \forall x$
- (ii) $\int_{-\infty}^{\infty} f(x) dx = 1$

Then $f(x)$ is called the probability density function (PDF) of the continuous variable X .

If (a,b) is a subinterval of the range space of X then the probability of x which lies in (a,b) is denoted by $P(a \leq X \leq b)$ and is defined as

$$P(a \leq X \leq b) = \int_a^b f(x) dx$$

5. Cumulative distribution function

Let X be a random variable (discrete or continuous). We define $F(x)$ to be the cumulative distribution function (CDF) or simply distribution function if

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

If X is a discrete random variable, then

$$\begin{aligned} F(x_i) &= P(X \leq x_i) \\ &= P(x_1) + P(x_2) + \cdots P(x_i) \end{aligned}$$

If X is a continuous variable with PDF $f(x)$ then

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

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

6. Mathematical Expectation $E(X)$:

For a DRV X having the possible values $x_1, x_2, x_3, x_4, \dots, x_n$, the expectation of X is defined as

$$E(X) = \sum_{i=1}^n x_i P(x_i)$$

Similarly, expectation of X^2 is defined as

$$E(X^2) = \sum_{i=1}^n x_i^2 P(x_i)$$

In general,

$$E(\phi(X)) = \sum_{i=1}^n \phi(x_i) P(x_i)$$

For a CRV X , having density function $f(x)$ the expectation of X is defined as

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

In general,

$$E(\phi(x)) = \int_{-\infty}^{\infty} \phi(x) f(x) dx$$

7. Mean & Variance:

The expectation of X is known as mean value of the probability distribution and it is denoted as μ .

$$\therefore \mu = E(X)$$

Variance of a distribution is defined as

$$Var(X) = \sigma^2 = E[(X - \mu)^2]$$

For a DRV

$$\sigma^2 = \sum_{i=1}^n (x_i - \mu)^2 P(x_i)$$

For a CRV

$$\sigma^2 = \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx$$

NOTE:

The positive square root of the variance is called the **standard deviation** and is given by $\sigma = \sqrt{Var(X)} = \sqrt{E[(X - \mu)^2]}$

8. Laws of Expectation:

a) $E(cX) = cE(X)$ where c is constant

b) If X and Y are random variables, then

$$E(X + Y) = E(X) + E(Y)$$

c) If X and Y are independent random variables, then $E(XY) = E(X)E(Y)$

Problems

a. Prove that $\sigma^2 = E(X^2) - [E(X)]^2$

$$\sigma^2 = E[(X - \mu)^2]$$

$$\Rightarrow \sigma^2 = \sum_{i=1}^n (x_i - \mu)^2 P(x_i)$$

$$\Rightarrow \sigma^2 = \sum_{i=1}^n (x_i^2 + \mu^2 - 2x_i\mu) P(x_i)$$

$$\sigma^2 = \sum_{i=1}^n (x_i^2 P(x_i) + \mu^2 P(x_i) - 2x_i\mu P(x_i))$$

$$\sigma^2 = \sum_{i=1}^n x_i^2 P(x_i) + \mu^2 \sum_{i=1}^n P(x_i) - 2\mu \sum_{i=1}^n x_i P(x_i)$$

$$\sigma^2 = E(X^2) + \mu^2 - 2\mu^2$$

$$\sigma^2 = E(X^2) - [E(X)]^2$$

- b. The probability distribution of a random variable X is given by the following table. Find k and evaluate the mean.

X	0	1	2	3	4	5
$P(X=x)$	k	$5k$	$10k$	$10k$	$5k$	k

Soln: The set $\{P(x_i)\}$ is called the probability distribution of the random variable if it satisfies the following conditions.

$$(i) \quad P(x_i) \geq 0 \quad \forall i$$

$$(ii) \quad \sum P(x_i) = 1$$

$$\therefore k + 5k + 10k + 10k + 5k + k = 1$$

$$\therefore 32k = 1 \quad \Rightarrow \quad k = \frac{1}{32}$$

$$\text{Mean, } \mu = E(X) = \sum_{i=1}^n x_i P(x_i)$$

$$\Rightarrow \mu = 0(k) + 1(5k) + 2(10k) + 3(10k) + 4(5k) + 5(k)$$

$$\Rightarrow \mu = 80k = \frac{80}{32} = \frac{5}{2}$$

c. The probability density function of a variate X is

X	0	1	2	3	4	5	6
$P(X)$	k	$3k$	$5k$	$7k$	$9k$	$11k$	$13k$

Find k

(i) $P(X < 4)$, $P(X \geq 5)$, $P(3 < X < 6)$

(ii) Mean and variance

d. A random variable X has the following probability function

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

Find (i) k (ii) $P(X \geq 6)$ (iii) $P(X < 6)$ (iv) $P(1 \leq X < 5)$ (v) $E(X)$

Soln:

$$k + 2k + 2k + 3k + k^2 + 2k^2 + 7k^2 + k = 1$$

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

$$k = \frac{-9 \pm \sqrt{81 + 40}}{20} = \frac{-9 \pm 11}{20}$$

$$k = -1 \text{ or } 0.1$$

But, $P(x_i) \geq 0$ and hence $k = 0.1$

X	1	2	3	4	5	6	7
$P(X)$	0.1	0.2	0.2	0.3	0.01	0.02	0.17

$$ii) P(X \geq 6) = P(6) + P(7) = 0.19$$

$$iii) P(X < 6) = 1 - P(X \geq 6) = 0.81$$

$$\begin{aligned} iv) P(1 \leq X < 5) \\ &= P(1) + P(2) + P(3) + P(4) \\ &= 0.8 \end{aligned}$$

$$iv) E(X) = \sum_{i=1}^n x_i P(x_i) = 3.66$$

- e. A company has five applicants for two positions. Two women and three men have applied. Suppose that no preference is given for choosing either sex. Let the random variable X be the number of women chosen to fill the two positions. Write the probability mass function of X .

Soln:

$$S = \{M_1M_2, M_1M_3, M_2M_3, W_1W_2, M_1W_1, M_1W_2, M_2W_1, M_2W_2, M_3W_1, M_3W_2\}$$

X	0		1		2	
	0W	2M	1W	1M	2W	0M
P(X)	3/10		6/10		1/10	

Method-02 {without writing S}

X	0		1		2	
	0W	2M	1W	1M	2W	0M
P(X)	$\frac{{}^2C_0 \cdot {}^3C_2}{{}^5C_2}$		$\frac{{}^2C_1 \cdot {}^3C_1}{{}^5C_2}$		$\frac{{}^2C_2 \cdot {}^3C_0}{{}^5C_2}$	
	3/10		6/10		1/10	

- f. Find $E[X]$, $E[X^2]$ and σ^2 for the probability function $P(X)$ defined by the following table

x_i	1	2	3	n
$P(x_i)$	k	2k	3k	nk

Where k is an appropriate constant.

- g. Find which of the following is a probability density function.

$$a) f_1(x) = \begin{cases} 2x, & 0 < x < 1 \\ 0, & \text{otherwise} \end{cases}$$

$$b) f_2(x) = \begin{cases} 2x, & -1 < x < 1 \\ 0, & \text{otherwise} \end{cases}$$

$$c) f_3(x) = \begin{cases} |x|, & |x| \leq 1 \\ 0, & \text{otherwise} \end{cases}$$

$$d) f_4(x) = \begin{cases} 2x, & 0 < x \leq 1 \\ 4 - 4x, & 1 < x < 2 \\ 0, & \text{otherwise} \end{cases}$$

- j. A random variable X has the density function $f(x) = \frac{k}{1+x^2}$, $-\infty < x < \infty$ determine k and evaluate $P(X \geq 0)$.

- k. A continuous random variable X has pdf given by $f(x) = \begin{cases} 2e^{-2x}, & x > 0 \\ 0, & x \leq 0 \end{cases}$
Evaluate (i) $E(x)$ and (ii) $E(x^2)$. Hence find the standard deviation.

- l. If a continuous random variable X has pdf

$$f(x) = \begin{cases} \frac{1}{4}, & -2 < x < 2 \\ 0, & \text{elsewhere} \end{cases}$$

- Obtain (i) $P[X < 1]$ (ii) $P[|X| > 1]$
(iii) $P[2X + 3 > 5]$

m. Let the continuous random variable X have the pdf $f(x) = \begin{cases} \frac{k}{1+x^2}, & 1 < x < \infty \\ 0, & \text{elsewhere} \end{cases}$. Find the distribution function $F(x)$.

n. The pdf of a random variable X is given by

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

Find (i) Probability distribution function $F(X)$ and (ii) $P(X \geq 1.5)$.

O. Is the function $f(x) = \begin{cases} e^{-x}, & 0 \leq x < \infty \\ 0, & \text{elsewhere} \end{cases}$ a density function of the continuous random variable X ?

i). If so, determine $P(1 \leq X \leq 2)$.

ii). Also, find the probability distribution function $f(x)$ at $x = 2$.

- p. The diameter of an electric cable is assumed to be a continuous random variable with pdf $f(x) = 6x(1-x)$, $0 \leq x \leq 1$, 0 elsewhere. Verify that the above is a pdf. Also find its mean and variance.
- q. A random variable X has the probability function $p(x) = \frac{1}{2^x}$; $x = 1, 2, 3, 4 \dots$
Find (i) Mean & (ii) variance.