

Univariate distributions :-

Sums :-

1. A random variable X can assume the values $-1, 0, 1$ with probabilities $\frac{1}{3}, \frac{1}{2}, \frac{1}{6}$ respectively. Find the distribution function.
2. Evaluate the distⁿ function of the following distⁿ :-
Spectrum of the r.v. X is $\{-1, 0, 2, 3\}$ with probabilities
 $P(X=-1) = \frac{1}{7}, P(X=0) = \frac{2}{7}, P(X=2) = \frac{3}{7}, P(X=3) = \frac{1}{7}$
3. Find the constant K such that the funcⁿ $f(x)$ given by
$$f_x(x) = K|x|, \quad -2 < x < 2$$
$$= 0, \quad \text{elsewhere} \quad K = \frac{1}{4}$$
is a possible pdf and find its distⁿ function.
4. Find the value of K so that the function $f(x)$ given by
$$f(x) = Kx(2-x), \quad 0 < x < 2$$
$$= 0, \quad \text{elsewhere}$$
is a pdf. Construct the dF and compute $P(X > 1)$.
$$K = \frac{3}{4}$$
$$P(X > 1) = \frac{1}{2}$$
5. A radioactive source emits on the average 2.5 particles per second. Calculate the probability that 2 or more particles will be emitted in an interval of 4 seconds.
6. Find K so that the function $f(x)$ defined by
$$f(x) = K(x - [x]), \quad 1 < x < 2$$
$$= 0, \quad \text{elsewhere}$$
is a pdf. Also find the distⁿ function. [ANS: $1 - 11e^{-10}$]
7. If X follows poisson(μ) distribution, find the distribution of $cx+d$ (c, d are positive constants)
8. X is uniformly distributed over the interval $(-2, 2)$. Find the distribution of the r.v. $\min\{X, 1\}$.

9. If X is uniformly distributed in the interval $(-1, 1)$, find the distribution of $|X|$.
10. The r.v. X is distributed uniformly over the interval $(0, 2)$. Find the d.f. of the larger root of the quadratic equation $t^2 + 2t - X = 0$.
11. The pdf of X is $2xe^{-x^2}$, $x > 0$ and zero otherwise. Find the pdf of $Y = X^2$.
12. If X is Normal $(0, 1)$, find the distribution of e^X .

Transformation of random variable [continuous case]

Let X be a r.v. having an absolutely continuous probability distⁿ given by its p.d.f $f(x)$, $-\infty < x < \infty$.

Let $Y = g(X)$ be another r.v. which is strictly monotonic such that $X = g^{-1}(Y)$ exists. Then the p.d.f. of Y will be

$$f_Y(y) = f_X(x) \left| \frac{dx}{dy} \right| = f\{g^{-1}(y)\} \left| \frac{dx}{dy} \right|$$

Proof: case-I

Let $y = g(x)$ be monotonic increasing function of x , having one-to-one correspondence. Here $\frac{dy}{dx} > 0$

$$\text{Now } X \leq x \Leftrightarrow g(X) \leq g(x)$$

\therefore the events $X \leq x$ and $Y \leq y$ are identical

$$\therefore P(X \leq x) = P(Y \leq y)$$

$$\therefore F(x) = F(y)$$

Taking differentials, we have

$$dF(x) = dF(y) = dF(g(x))$$

If p.d.f of Y be $f(y)$, then we have

$$f(y) dy = f(x) dx = f\{g^{-1}(y)\} dx$$

$$\therefore f(y) = f(x) \frac{dx}{dy} = f\{g^{-1}(y)\} \frac{dx}{dy} \quad \text{--- (1)}$$

case-II Let $y = g(x)$ be monotonic decreasing function i.e. $\frac{dy}{dx} < 0$.

$$\text{Now } X \leq x \Leftrightarrow g(X) \geq g(x)$$

$$\therefore Y \geq y$$

$$\therefore P(Y \geq y) = P(X \leq x)$$

$$1 - P(Y < y) = P(X \leq x)$$

$$\text{or, } 1 - F(y) = F(x)$$

Taking differential, we have

$$-dF(y) = dF(x)$$

$$-f(y)dy = f(x)dx = f\{g^{-1}(y)\}dx$$

$$\therefore, f(y) = -f(x) \frac{dx}{dy} = -f\{g^{-1}(y)\} \frac{dx}{dy} \quad \text{--- (2)}$$

combining (1) and (2), we have

$$f(y) = f(x) \left| \frac{dx}{dy} \right| = f\{g^{-1}(y)\} \left| \frac{dx}{dy} \right|$$

Discrete case

Here the spectrum only changes, the corresponding probabilities masses remaining the same.

Let $y = g(x)$ be a continuous and strictly monotonic function so that a unique inverse function $x = g^{-1}(y)$ exists.

$$\text{set } y_i = g(x_i) \quad \text{--- (1)}$$

Since the transformation has a unique inverse

$$(x = x_i) \Leftrightarrow g(x) = g(x_i)$$

$$\text{so } y = y_i$$

Hence the spectrum of y consists of the pts y_i given by (1).

$$\text{and } P(x = x_i) = P(y = y_i)$$

$$\text{or, } p_{x_i} = p_{y_i}$$

Important discrete distribution

1. Binomial (n, p) -distⁿ:

A discrete distribution of a random variable X is said to be binomial if the spectrum is the set $\{0, 1, 2, \dots, n\}$ i.e. $X = x_i = i$ $i = 0, 1, 2, \dots, n$ and probability mass function (p.m.f) is given by

$$f_i = P(X = x_i) = \binom{n}{i} p^i (1-p)^{n-i}, \quad i = 0, 1, 2, \dots, n$$

where n ($n \in \mathbb{N}$) and p ($0 < p < 1$) are two parameters of the distribution.

This distribution is known as binomial (n, p) -distribution.

$$\text{Here, } \sum_{i=0}^n f_i = \sum_{i=0}^n \binom{n}{i} p^i (1-p)^{n-i} = \{p + (1-p)\}^n = 1.$$

The number of success X in a Bernoullian sequence of n independent trials follows binomial distribution with parameters n and p . p is the probability of success in each trial.

2. Poisson distribution:

The prob. mass. function (p.m.f) of the distⁿ is given by $P_x = \frac{e^{-\mu} \mu^x}{x!}$, $x = 0, 1, 2, \dots$

where $\mu (> 0)$ is the only parameter of the distⁿ.

$$\text{Note: (i) } P_x \geq 0 \text{ and } \sum_{x=0}^{\infty} \frac{e^{-\mu} \mu^x}{x!} = e^{-\mu} \left(1 + \mu + \frac{\mu^2}{2!} + \frac{\mu^3}{3!} + \dots \right) = e^{-\mu} e^{\mu} = 1$$

(ii) Following are examples of poisson variate.

- 1) Number of errors in type per page in a book.
- 2) Number of defective articles per box of 100 articles.
- 3) Number of telephone calls in a booth in an interval of time.
- 4) Number of car passing by in a crossing of a busy road for a given interval of time.

6. Uniform distribution:-

The probability mass function of discrete uniform distribution is given by

$$f_i = P(X=x_i) = \frac{1}{n}, \quad i=1, 2, \dots, n$$

The spectrum of the random variable X is $X = x_1, x_2, \dots, x_n$. Here, there is n step points x_1, x_2, \dots, x_n with equal probabilities, that is why it is called uniform.

The basic need for pmf $f_i > 0$ and $\sum_{i=1}^n f_i = 1$ satisfied.

If $x_m \leq x \leq x_{m+1}$, $m < n$, Then

$$P(X \leq x) = \frac{m}{n} \quad \text{and} \quad P(X \leq x_n) = 1$$

The distribution function is given by

$$F(x) = m/n, \quad x_m \leq x \leq x_{m+1}, \quad m < n$$

$$= 1, \quad x = x_n.$$

Important Continuous distribution

1. Uniform or Rectangular distribution

A continuous random variable X is said to have uniform distribution in the interval $a < x < b$ if the probability density function is defined by

$$f(x) = \frac{1}{b-a}, \quad a < x < b$$
$$= 0, \quad \text{elsewhere}$$

(i) we have, $f(x) \geq 0$ as $b > a$

$$\text{and } \int_a^b f(x) dx = \int_a^b \frac{1}{b-a} dx = \frac{1}{b-a} [x]_a^b = \frac{b-a}{b-a} = 1.$$

(ii) The distribution function $F(x)$ is given by

$$F(x) = P(-\infty < X \leq x)$$

Here, $F(x) = 0$ when $x < a$

For $a \leq x \leq b$, $F(x) = P(-\infty < X \leq a) + P(a \leq X \leq x) + P(b < X \leq x)$

$$= 0 + \int_a^x \frac{1}{b-a} dx = \frac{x-a}{b-a}$$

For $x > b$, $F(x) = P(-\infty < X < a) + P(a \leq x \leq b) + P(b < X \leq x)$

$$= 0 + \int_a^b \frac{dx}{b-a} + 0 = 1.$$

$\therefore F(x) = 0$ when $x < a$

$$= \frac{x-a}{b-a} \quad \text{when } a \leq x \leq b$$

$$= 1 \quad \text{when } x > b$$

(iii) A point x is chosen at random in the interval $a \leq x \leq b$ in such a way that the probability that it lies in any sub interval is proportional to the length of the sub interval. Then, we can show that x is uniformly distributed over the interval (a, b) i.e., The random point X has a uniform distribution in the given interval.

2. Normal distribution:

The probability density function (pdf) of this distribution is given by

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2\sigma^2}(x-\mu)^2}$$

where μ and $\sigma (> 0)$ be two parameters of this distⁿ, $-\infty < x < \infty$

In this case, we say that the random variable X follows $N(\mu, \sigma)$ -distribution

(i) From this pdf, we have

$$\int_{-\infty}^{\infty} f(x) dx = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx \quad \text{let } \frac{x-\mu}{\sqrt{2\sigma^2}} = z$$

$$\text{then } \int_{-\infty}^{\infty} f(x) dx = \frac{1}{\sqrt{\pi}} \int_{-\infty}^{\infty} e^{-z^2} dz = \frac{2}{\sqrt{\pi}} \int_0^{\infty} e^{-z^2} dz = \frac{2}{\sqrt{\pi}} \cdot \frac{\sqrt{\pi}}{2} = 1.$$

(ii) The distribution function $F(x)$ is given by

$$F(x) = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

3. Standard normal distribution:

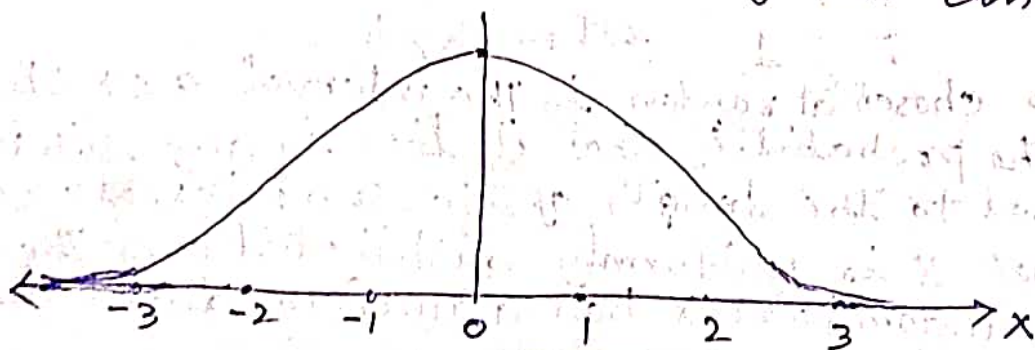
In normal distribution, we take $\mu=0$, $\sigma=1$, then the distribution follows $N(0,1)$.

The pdf of this distribution is given by

$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2}, \quad -\infty < x < \infty$$

The distribution function is $F(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{1}{2}x^2} dx$

The probability density curve of this distribution is



3. Exponential distribution:

The probability density function of this distⁿ is given by

$$f(x) = \frac{1}{\lambda} e^{-\frac{x}{\lambda}}, \quad x > 0, \quad \lambda > 0$$

where $\lambda > 0$ is the only parameter of the distribution. $= 0$, elsewhere

We see that (i) $f(x) \geq 0$ for all x and

$$(ii) \int_{-\infty}^{\infty} f(x) dx = \int_0^{\infty} \frac{1}{\lambda} e^{-\frac{x}{\lambda}} dx = 1$$

Th: Show that poisson distribution as a limiting case of Binomial distribution.

Proof: For binomial distribution, we have

$$f_x = P(X=x) = \binom{n}{x} p^x (1-p)^{n-x}, \quad x=0,1,2, \dots, n$$

p is the probability of success in each trial.

Now, we take the number of trials n is large and p be very small. Then, np is moderate in magnitude.

Let $\mu = np$, then $p = \frac{\mu}{n}$

$$\text{and } P(X=x) = \frac{n!}{x! (n-x)!} \left(\frac{\mu}{n}\right)^x \left(1 - \frac{\mu}{n}\right)^{n-x}$$

Taking limit as $n \rightarrow \infty$ and $p \rightarrow 0$, we have

$$\lim_{n \rightarrow \infty} P(X=x) = \lim_{n \rightarrow \infty} \frac{n!}{x! (n-x)!} \left(\frac{\mu}{n}\right)^x \left(1 - \frac{\mu}{n}\right)^{n-x}$$

$$\therefore P_x = \lim_{n \rightarrow \infty} \frac{n(n-1) \dots (n-x+1)}{x!} \frac{\mu^x}{n^x} \left(1 - \frac{\mu}{n}\right)^n \left(1 - \frac{\mu}{n}\right)^{-x}$$

$$= \frac{\mu^x}{x!} \lim_{n \rightarrow \infty} \left(1 - \frac{1}{n}\right) \left(1 - \frac{2}{n}\right) \dots \left(1 - \frac{x-1}{n}\right) \cdot \lim_{n \rightarrow \infty} \left(1 - \frac{\mu}{n}\right)^n \lim_{n \rightarrow \infty} \left(1 - \frac{\mu}{n}\right)^{-x}$$

$$= \frac{\mu^x}{x!} \cdot 1 \cdot \lim_{n \rightarrow \infty} \left[\left\{ 1 + \left(-\frac{\mu}{n}\right) \right\}^{-n} \right] \cdot 1 \quad \text{as } \lim_{n \rightarrow \infty} \left(1 - \frac{\mu}{n}\right)^{-n} = 1$$

$$= \frac{\mu^x}{x!} e^{-\mu}, \quad \text{since } \lim_{x \rightarrow \infty} \left(1 + \frac{1}{x}\right)^x = e$$

$$= \frac{e^{-\mu} \mu^x}{x!}, \quad x=0,1,2, \dots$$

which is the $\frac{1}{x!}$ probability mass function of poisson distribution.

3. Negative Binomial distⁿ:

Let p be the prob. of success at each trial in an infinite sequence of independent Bernoullian trials. The random variable X denotes the number of failures preceeding the r th success ($r > 1$), then the distribution of X is given by the p.d.f

$$f_i = P(X=i) = \binom{i+r-1}{r-1} p^r (1-p)^i, \quad i = 0, 1, 2, \dots$$

where, r and p ($0 < p < 1$) be two parameters of the distⁿ.

4. Geometric distribution:

The r.v. X denote the number of failures preceeding the first success. Then the pdf of geometric distⁿ is given by

$$P_i = (1-p)^i p = q^i p, \quad i = 0, 1, 2, \dots \quad p+q=1.$$

Note: Sometimes it is possible to develop such a formula even for a transformation which is not one-to-one.

Let $y = x^2$ where x is a continuous r.v.

Here the transformation $y = x^2$ is not one-to-one.

We have, $F_Y(y) = P(Y \leq y) = P(x^2 \leq y)$

$$= P(-\sqrt{y} \leq x \leq \sqrt{y})$$

$$= F_X(\sqrt{y}) - F_X(-\sqrt{y})$$

$$\therefore f_Y(y) = F_Y'(y) = \frac{d}{dy} [F_X(\sqrt{y}) - F_X(-\sqrt{y})]$$

$$= \frac{f_X(\sqrt{y}) + f_X(-\sqrt{y})}{2\sqrt{y}}, \quad \text{if } y > 0$$

$$= 0 \quad \text{if } y \leq 0$$