

5. Uniform distribution

A continuous random variable X is said to follow a continuous uniform distribution over an interval $[a, b]$ if its probability density function of is given by

$$f(x) = \frac{1}{b-a}, \quad a \leq x \leq b$$

The cumulative distribution function is given by

$$F(x) = \frac{x-a}{b-a}, \quad a \leq x \leq b$$

The uniform distribution with support $[a, b]$ is denoted by $U(a, b)$. This distribution is also called the *rectangular* distribution because of the rectangular shape of its pdf.

1. Moments.

$$\mu'_r = \int_a^b \frac{x^r dx}{b-a} = \frac{1}{b-a} \left[\frac{x^{r+1}}{r+1} \right]_a^b = \frac{1}{b-a} \left[\frac{b^{r+1} - a^{r+1}}{r+1} \right]$$

$$\text{Mean} = \mu'_1 = \frac{1}{b-a} \left[\frac{b^2 - a^2}{2} \right] = \frac{b+a}{2}$$

$$\mu'_2 = \frac{1}{b-a} \left[\frac{b^3 - a^3}{3} \right] = \frac{b^2 + ab + a^2}{3}$$

$$\text{Variance} = \mu_2 = \mu'_2 - (\mu'_1)^2 = \frac{b^2 + ab + a^2}{3} - \left(\frac{b+a}{2} \right)^2 = \frac{(b-a)^2}{12}.$$

2. Moment generating function

$$M_x(t) = \int_a^b \frac{e^{tx} dx}{b-a} = \frac{1}{b-a} \left[\frac{e^{tx}}{t} \right]_a^b = \frac{e^{bt} - e^{at}}{t(b-a)}.$$

6. Exponential distribution

A continuous random variable X is said to follow a exponential distribution with parameter λ if its probability density function of a is given by

$$f(x) = \lambda e^{-\lambda x}, \quad x \geq 0.$$

The cumulative distribution function is given by

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

$$f(x) = \lambda e^{-\lambda(x-\mu)}, x \geq \mu.$$

If we take $\lambda = \frac{1}{b}$ the density become

$$f(x) = \frac{1}{b} e^{-\frac{x}{b}}, x \geq 0, b > 0.$$

1. Moments.

$$\mu'_r = \int_0^\infty x^r \lambda e^{-\lambda x} dx = \left[\frac{\lambda \Gamma(r+1)}{-\lambda^{r+1}} \right]_0^\infty = \frac{\Gamma(r+1)}{\lambda^r}$$

$$\text{Mean} = \mu'_1 = \frac{\Gamma(1+1)}{\lambda} = \frac{1}{\lambda}$$

$$\mu'_2 = \frac{\Gamma(2+1)}{\lambda^2} = \frac{2}{\lambda^2}$$

$$\text{Variance} = \mu_2 = \mu'_2 - (\mu'_1)^2 = \frac{2}{\lambda^2} - \left(\frac{1}{\lambda}\right)^2 = \frac{1}{\lambda^2}.$$

2. Moment generating function

$$M_x(t) = \int_0^\infty e^{tx} \lambda e^{-\lambda x} dx = \int_0^\infty \lambda e^{-x(\lambda-t)} dx = \frac{\lambda}{\lambda-t} = \left(1 - \frac{t}{\lambda}\right)^{-1}.$$

3. Reproductive property

If X_i ($i=1, 2, \dots, k$) are identically independently distributed exponential random variate with parameters λ . Then $S_k = \sum_{i=1}^k X_i$ has a gamma distribution with parameters ($k, 1/\lambda$). (This property is called additive or reproductive property).

$$\text{Given } X_i \sim E(\lambda) \text{ So } M_{x_i}(t) = \left(1 - \frac{t}{\lambda}\right)^{-1}.$$

$$M_{S_k}(t) = \prod_{i=1}^k M_{x_i}(t)$$

$$= \prod_{i=1}^k \left(1 - \frac{t}{\lambda}\right)^{-1} = \left(1 - \frac{t}{\lambda}\right)^{-k}$$

Which is the mgf of gamma distribution (k, λ) . So $S_k = \sum_{i=1}^k X_i$ has a gamma distribution with parameters $(k, 1/\lambda)$.

Result

The exponential distribution has the memoryless (forgetfulness) property.

A exponential random variable X has the memoryless property if for all nonnegative integers s and t ,

$$P(X \geq s+t | X \geq t) = P(X \geq s)$$

Proof:

The probability mass function for a exponential random variable X is

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

Then

$$P(X \geq x) = e^{-\lambda x}$$

$$\begin{aligned} P(X \geq s+t | X \geq t) &= \frac{P(X \geq s+t \cap X \geq t)}{P(X \geq t)} = \frac{P(X \geq s+t)}{P(X \geq t)} \\ &= \frac{e^{-\lambda(s+t)}}{e^{-\lambda t}} = e^{-\lambda s} = P(X \geq s) \end{aligned}$$

which proves the memoryless property.

7. Gamma distribution

A continuous random variable X is said to follow a gamma distribution with parameter a and b if its probability density function is given by

$$f(x) = \frac{1}{b^a \Gamma(a)} x^{a-1} e^{-x/b}, x \geq 0.$$

If we take $b=1$, then we get one parameter gamma distribution or standard gamma distribution. That is in this case the pdf is

$$f(x) = \frac{1}{\Gamma(a)} x^{a-1} e^{-x}, x \geq 0.$$

Mean	ab
Variance	ab^2
Mode	$b(a-1), a > 1$
Coefficient of variation	$\frac{1}{\sqrt{a}}$
Coefficient of Skewness	$\frac{2}{\sqrt{a}}$
Coefficient of Kurtosis	$3 + \frac{6}{a}$
Moment generating function	$(1 - bt)^{-a}, t < \frac{1}{b}$.

$$f(x) = \frac{1}{b^a \Gamma(a)} x^{a-1} e^{-x/b}, x \geq 0.$$

1. Moments.

$$\begin{aligned} \mu'_r &= \int_0^\infty x^r \frac{1}{b^a \Gamma(a)} x^{a-1} e^{-x/b} dx = \int_0^\infty \frac{1}{b^a \Gamma(a)} x^{a+r-1} e^{-x/b} dx \\ &= \frac{1}{b^a \Gamma(a)} \frac{\Gamma(a+r)}{\left(\frac{1}{b}\right)^{a+r}} = \frac{b^r \Gamma(a+r)}{\Gamma(a)} \end{aligned}$$

$$\text{Mean} = \mu'_1 = \frac{b^1 \Gamma(a+1)}{\Gamma(a)} = ab.$$

$$\mu'_2 = \frac{b^2 \Gamma(a+2)}{\Gamma(a)} = a(a+1)b^2.$$

$$\text{Variance} = \mu_2 - (\mu_1')^2 = a(a+1)b^2 - (ab)^2 = ab^2.$$

2 Moment generating function

$$\begin{aligned} M_x(t) &= \int_0^{\infty} e^{tx} \frac{1}{b^a \Gamma(a)} x^{a-1} e^{-x/b} dx \\ &= \int_0^{\infty} \frac{1}{b^a \Gamma(a)} x^{a-1} e^{-x(\frac{1}{b}-t)} dx = \frac{1}{b^a \Gamma(a)} \frac{\Gamma(a)}{(\frac{1}{b}-t)^a} = (1-bt)^{-a}. \end{aligned}$$

3 Reproductive property

If X_i ($i=1, 2, \dots, k$) are identically independently distributed gamma random variate with parameters (a_i, b) . Then $S_k = \sum_{i=1}^k X_i$ has a gamma distribution with parameters $(\sum_{i=1}^k a_i, b)$.

Given $X_i \sim G(a_i, b)$ So $M_{x_i}(t) = (1-bt)^{-a_i}$.

$$\begin{aligned} M_{S_k}(t) &= \prod_{i=1}^k M_{x_i}(t) \\ &= \prod_{i=1}^k (1-bt)^{-a_i} = (1-bt)^{-\sum a_i} \end{aligned}$$

Hence $S_k = \sum_{i=1}^k X_i$ has a gamma distribution with parameters $(\sum_{i=1}^k a_i, b)$.

8. Beta distribution of first kind

A continuous random variable X is said to follow a beta distribution with parameter a and b if its probability density function is given by

$$f(x) = \frac{1}{B(a,b)} x^{a-1} (1-x)^{b-1}, \quad 0 \leq x \leq 1.$$

Where $B(a, b) = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)}$.

$$\int_0^1 x^{a-1}(1-x)^{b-1} dx = B(a, b) = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)}$$

Moments

$$\begin{aligned} \mu'_r &= \int_0^1 x^r \frac{1}{B(a,b)} x^{a-1}(1-x)^{b-1} dx \\ &= \int_0^1 \frac{1}{B(a,b)} x^{a+r-1}(1-x)^{b-1} dx = \frac{b(a+r,b)}{B(a,b)} \\ &= \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} \frac{\Gamma(a+r)\Gamma(b)}{\Gamma(a+b+r)} = \frac{\Gamma(a+r)\Gamma(a+b)}{\Gamma(a)\Gamma(a+b+r)}. \end{aligned}$$

Mean	$\frac{a}{a+b}$
Variance	$\frac{ab}{(a+b+1)(a+b)^2}$
Mode	$\frac{a-1}{a+b-2}$
Mean Deviation	$\frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)} \frac{2a^a b^b}{(a+b)^{(a+b+1)}}$
Coefficient of variation	$\frac{\sqrt{b}}{\sqrt{a(a+b+1)}}$
Coefficient of Skewness	$\frac{2(b-a)\sqrt{(a+b+1)}}{(a+b+2)\sqrt{ab}}$
Coefficient of Kurtosis	$\frac{3(a+b+1)[2(a+b)^2+ab(a+b-6)]}{ab(a+b+2)(a+b+3)}$
Characteristic function	$\frac{\Gamma(a+b)}{\Gamma(a)} \sum_{k=0}^{\infty} \frac{\Gamma(a+k)(it)^k}{\Gamma(a+b+k)\Gamma(k+1)}$
Moments about the Origin:	$E(X^k) = \prod_{i=0}^{k-1} \frac{a+i}{a+b+i}, k=1,2,\dots$

Beta distribution of second kind

A continuous random variable X is said to follow a beta distribution of second kind with parameter a and b if its probability density function is given by

$$f(x) = \frac{1}{B(a,b)} \frac{x^{a-1}}{(1+x)^{a+b}}, \quad 0 \leq x < \infty.$$

where

$$B(a, b) = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)}.$$

$$\text{Mean} = \frac{a}{b-1}$$

$$\text{Variance} = \frac{a(a+b-1)}{(b-2)(b-1)^2}.$$

Result

If X follows a beta distribution of first kind with parameter a and b . Then $Y = \frac{X}{1-X}$ follow a beta distribution of second kind with parameter a and b .

Proof

Given X follows a beta distribution of first kind with parameter a and b . So

$$f(x) = \frac{1}{B(a,b)} x^{a-1}(1-x)^{b-1}, \quad 0 \leq x \leq 1.$$

Let $g(y)$ be the pdf of $Y = \frac{X}{1-X}$ So $x = \frac{y}{1+y}$

Now

$$\frac{dx}{dy} = (1-x)^2 = \left(1 - \frac{y}{1+y}\right)^2$$

Hence

$$\begin{aligned}g(y) = f(x) \left| \frac{dx}{dy} \right| &= \frac{1}{B(a,b)} \left(\frac{y}{1+y} \right)^{a-1} \left(1 - \frac{y}{1+y} \right)^{b-1} \left(1 - \frac{y}{1+y} \right)^2 \\ &= \frac{1}{B(a,b)} \frac{y^{a-1}}{(1+y)^{a+b}}, \quad 0 \leq y < \infty.\end{aligned}$$

Which is the pdf of beta distribution of second kind. So Y follow a beta distribution of second kind.