

Gamma Distribution:

The random variable 'X' has a gamma dist if its p.d.f is

$$f(x; \beta, \alpha) = \frac{1}{\Gamma(\alpha) \beta^\alpha} x^{\alpha-1} e^{-\frac{x}{\beta}}, \quad 0 \leq x < \infty, \quad \alpha, \beta > 0.$$

1. Pdf

$$\int_{-\infty}^{\infty} f(x) dx = 1 = \frac{1}{\Gamma(\alpha) \beta^\alpha} \int_0^{\infty} x^{\alpha-1} e^{-\frac{x}{\beta}} dx$$

$$\text{Let } y = \frac{x}{\beta} \Rightarrow x = \beta y$$

$$dy = \frac{dx}{\beta} \Rightarrow dx = \beta dy$$

$$\text{when } x=0 \Rightarrow y=0$$

$$x=\infty \Rightarrow y=\infty$$

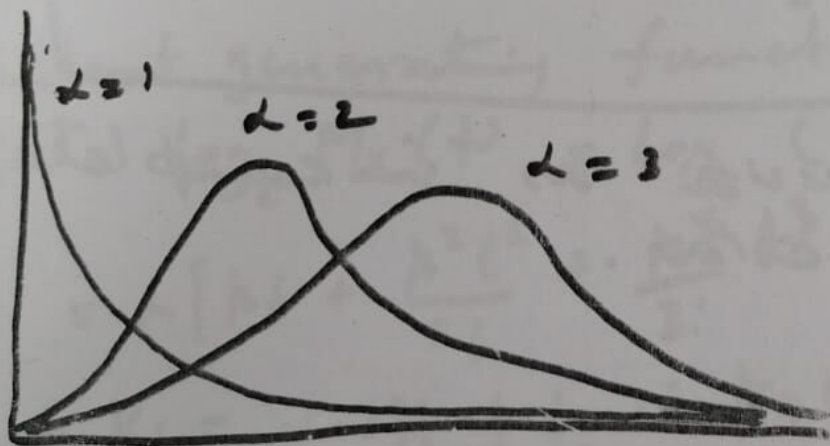
$$= \frac{1}{\Gamma(\alpha) \beta^\alpha} \int_0^{\infty} (\beta y)^{\alpha-1} e^{-y} (\beta dy)$$

$$= \frac{1}{\Gamma(\alpha) \beta^\alpha} \int_0^{\infty} \beta^\alpha y^{\alpha-1} e^{-y} dy$$

$$= \frac{1}{\Gamma(\alpha)} \Gamma(\alpha) = 1$$

$$\therefore \int_0^{\infty} x^{\alpha-1} e^{-\frac{x}{\beta}} dx = \Gamma(\alpha) \beta^\alpha; \quad \alpha > 0$$

2. Shape



Incomplete Gamma Function:

We know that

$$F(x) = \frac{1}{\Gamma(\alpha)} \int_0^x x^{\alpha-1} e^{-x} dx$$

For different values of x and α where α is a +ve integer.

$$\therefore e^{-x} = 1 - x + \frac{x^2}{2!} - \frac{x^3}{3!} + \frac{x^4}{4!} - \dots$$

$$\begin{aligned} F(x) &= \frac{1}{\Gamma(\alpha)} \int_0^x x^{\alpha-1} \left[1 - x + \frac{x^2}{2!} - \frac{x^3}{3!} + \frac{x^4}{4!} - \dots \right] dx \\ &= \frac{1}{\Gamma(\alpha)} \left[\frac{x^\alpha}{\alpha} - \frac{x^{\alpha+1}}{\alpha+1} + \frac{1}{2} \frac{x^{\alpha+2}}{\alpha+2} - \frac{1}{3!} \frac{x^{\alpha+3}}{\alpha+3} + \frac{1}{4!} \frac{x^{\alpha+4}}{\alpha+4} - \dots \right] \end{aligned}$$

$$F(x) = \frac{1}{\Gamma(\alpha)} \sum_{i=0}^{\infty} (-1)^i \frac{x^{\alpha+i}}{\alpha+i} \cdot \frac{1}{i!}$$

which is known as incomplete gamma function.

1. C d f

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It is called incomplete gamma function and is only obtained by tables, Mathematically:

$$F(x) = \begin{cases} \frac{1}{\Gamma(\alpha) \beta^\alpha} \int_0^x x^{\alpha-1} e^{-\frac{x}{\beta}} dx & ; x > 0 \\ 0 & , \text{e.w.} \end{cases}$$

4. Moments:

r^{th} moment about origin:

$$\mu'_r = E(x^r) = \int_{-\infty}^{\infty} x^r f(x) dx \quad (\text{By definition})$$

$$= \frac{1}{\Gamma(\alpha) \beta^\alpha} \int_0^{\infty} x^r \left[x^{\alpha-1} e^{-\frac{x}{\beta}} \right] dx$$

$$\text{Let } y = \frac{x}{\beta} \Rightarrow x = \beta y \\ dx = \beta dy$$

$$= \frac{1}{\Gamma(\alpha) \beta^\alpha} \int_0^{\infty} (\beta y)^{\alpha+r-1} e^{-y} (\beta dy) \quad \text{when } \begin{matrix} x=0 \Rightarrow y=0 \\ x=\infty \Rightarrow y=\infty \end{matrix}$$

$$= \frac{\beta^{\alpha+r}}{\Gamma(\alpha) \beta^\alpha} \int_0^{\infty} y^{\alpha+r-1} e^{-y} dy, \quad \because \int_0^{\infty} x^{\alpha-1} e^{-x} dx = \Gamma(\alpha)$$

$$\mu'_r = \frac{\beta^r}{\Gamma(\alpha)} \Gamma(\alpha+r) ; r = 1, 2, 3, 4, \dots$$

(A)

Note: when $\beta = 1 \Rightarrow \text{Mean} = \text{Variance}$.

$\bar{x} = r^2$ like Poisson dist.

m.g.f

$$M_x(t) = E[e^{tx}] = \int_{-\infty}^{\infty} e^{tx} f(x) dx \quad (\text{By definition})$$

$$= \int_0^{\infty} e^{tx} \left[\frac{1}{\Gamma(\alpha)} \lambda^{\alpha} x^{\alpha-1} e^{-\frac{x}{\lambda}} \right] dx$$

$$= \frac{1}{\Gamma(\alpha)} \lambda^{\alpha} \int_0^{\infty} x^{\alpha-1} e^{-(\frac{1}{\lambda} - t)x} dx$$

(as $\int_0^{\infty} x^{\alpha-1} e^{-\frac{x}{\lambda}} dx = \frac{\Gamma(\alpha)}{\lambda^{\alpha}}$)

$$= \frac{1}{\Gamma(\alpha)} \lambda^{\alpha} \left[\frac{\Gamma(\alpha)}{(\frac{1}{\lambda} - t)^{\alpha}} \right]$$

$$= \frac{1}{\lambda^{\alpha} \cdot \frac{1}{\lambda^{\alpha}} (1 - \lambda t)^{\alpha}} = (1 - \lambda t)^{-\alpha}$$

$$\boxed{M_x(t) = (1 - \lambda t)^{-\alpha}}$$

Characteristic Function:

$$\phi_x(t) = (1 - i\lambda t)^{-\alpha}$$

Cumulant generating function:

$$K_x(t) = \log_e M_x(t) = \log_e (1 - \lambda t)^{-\alpha} = -\alpha \log_e (1 - \lambda t)$$

$$= \alpha \left[\lambda t + \frac{\lambda^2 t^2}{2!} + \frac{\lambda^3 t^3}{3!} + \frac{\lambda^4 t^4}{4!} + \dots \right]$$

Mean = K_1 = coeff of t in $K_x(t) = \alpha \lambda$

$M_2 = K_2 =$ " " $\frac{t^2}{2!}$ " " $= \alpha \lambda^2$

$M_3 = K_3 =$ " " $\frac{t^3}{3!}$ " " $= 2\alpha \lambda^3$

$M_4 = K_4 =$ " " $\frac{t^4}{4!}$ " " $= 6\alpha \lambda^4$

$$M_x(t) = \left(1 - \frac{t}{\alpha}\right)^{-1}$$

Expanding we have

$$M_x(t) = \sum_{r=0}^{\infty} \left(\frac{t}{\alpha}\right)^r = \sum_{r=0}^{\infty} \left(\frac{r!}{\alpha^r}\right) \cdot \frac{t^r}{r!}$$

Comparing the coefficients

$$\mu_r' = \frac{r!}{\alpha^r}$$

Show that

$$\beta_1 = 4 ; r_1 = 2 = \sqrt{\beta_1}$$

$$\beta_2 = 9 ; r_2 = 6 = \beta_2 - 3$$

Cumulant Generating Function

$$K_x(t) = \log_e M_x(t) = \log_e \left(1 - \frac{t}{\alpha}\right)^{-1}$$

$$= -\log_e \left(1 - \frac{t}{\alpha}\right)$$

$$\because \ln(1-x) = -x - \frac{x^2}{2} - \frac{x^3}{3} - \dots$$

$$= \sum_{r=1}^{\infty} \frac{(t/\alpha)^r}{r}$$

$$K_x(t) = \sum_{r=1}^{\infty} \frac{(r-1)!}{\alpha^r} \cdot \frac{t^r}{r!}$$

Comparing the coefficients of $\frac{t^r}{r!}$ we get

$$\mu_1' = k_1 = \frac{1}{\alpha}, \mu_2' = k_2 = \frac{1}{\alpha^2}$$

$$\mu_3' = k_3 = \frac{1}{\alpha^3}, \mu_4' = k_4 + 3k_2^2 = \frac{1}{\alpha^4}$$

$$\beta_1 = \frac{\mu_3'^2}{\mu_2'^2} = 4$$

$$r_1 = \sqrt{\beta_1} = 2$$

$$\beta_2 = \frac{\mu_4'}{\mu_2'^2} = 9$$

$$r_2 = \beta_2 - 3 = 9 - 3 = 6$$

Mean:

Put $r=1$ in Eq (A)

$$\mu_1' = E(X) = \frac{\beta}{\Gamma \alpha} \Gamma(\alpha+1) = \frac{\alpha! \beta}{(\alpha-1)!} = \alpha \beta.$$

Variance:

Put $r=2$ in Eq (A)

$$\begin{aligned} \mu_2' = E(X^2) &= \frac{\beta^2}{\Gamma \alpha} \Gamma(\alpha+2) = \frac{(\alpha+1)\alpha(\alpha-1)! \beta^2}{(\alpha-1)!} \\ &= \alpha(\alpha+1) \beta^2 \end{aligned}$$

$$\sigma^2 = \mu_2 = \mu_2' - \mu_1'^2 = \cancel{\alpha^2 \beta^2} + \alpha \beta^2 - (\cancel{\alpha \beta})^2$$

$$\boxed{\sigma^2 = \mu_2 = \alpha \beta^2}$$

Skewness

$$= \sqrt{\beta_1} = \frac{2}{\sqrt{\alpha}}$$

Kurtosis

$$\beta_2 = 3 \left(1 + \frac{2}{\alpha} \right)$$

$$\alpha_2 = \beta_2 - 3 =$$

$$\mu_3' =$$

$$\mu_4' =$$

$$\mu_3 =$$

$$\mu_4 =$$

$$\beta_1 =$$

$$\beta_2 =$$

Median:

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

$$= \frac{1}{\sigma\sqrt{2\pi}} \left[\int_{-\infty}^{\mu} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx + \int_{\mu}^{\infty} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx \right] = \frac{1}{2}$$

But

$$\frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx = \frac{1}{\sqrt{2\pi}} \int_0^{\infty} e^{-\frac{1}{2}z^2} dz$$

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

$$= \frac{1}{2} + 0 \quad \therefore \frac{1}{\sigma\sqrt{2\pi}} \int_{\mu}^{\mu} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx = 0$$

$$\Rightarrow \boxed{\mu = M} \text{ or } \boxed{M = \mu}$$

Hence $\boxed{\text{Mean} = \text{Median} = \text{Mode}}$

Mean Deviation about mean:

$$M.D._{\bar{x}} = \int_{-\infty}^{\infty} |x - \mu| f(x) dx = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} |x - \mu| e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx$$

$$= \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} |\sigma z| e^{-\frac{1}{2}z^2} \sigma dz$$

$$= \frac{\sigma}{\sqrt{2\pi}} \int_{-\infty}^{\infty} |z| e^{-\frac{1}{2}z^2} dz$$

$$= \frac{\sigma}{\sqrt{2\pi}} \cdot 2 \int_0^{\infty} z e^{-\frac{1}{2}z^2} dz$$

$$= \sqrt{\frac{2}{\pi}} \sigma \left(-e^{-y} \Big|_0^{\infty} \right)$$

$\therefore |z| e^{-\frac{1}{2}z^2}$ is an even function
 $|z| = z$
 $0 < z < \infty$

$$= \sqrt{\frac{2}{\pi}} \sigma$$

$$\boxed{M.D._{\bar{x}} = \frac{4}{5} \sigma}$$

Problem :

Suppose that on average 30 customer per hour arrive at a shop in accordance with a Poisson process. That is, if minutes is our unit, then $\lambda = \frac{1}{2}$.

What is the probability that the shopkeeper will wait more than 5 minutes before both of the first two customers arrive?

Sol.:

Let 'X' denotes the waiting time in minutes until the second customer arrives. Then

$$X \sim \text{gamma} (d=2, \beta = \frac{1}{\lambda} = 2)$$

Hence

$$P(X > 5) = \int_5^{\infty} \frac{x^{d-1} e^{-x/\beta}}{\Gamma(d) \beta^d} dx$$

$$= \frac{1}{\Gamma(2) 2^2} \int_5^{\infty} x^{2-1} e^{-x/2} dx$$

$$= \frac{1}{4} \int_5^{\infty} x e^{-x/2} dx$$

$$= \frac{1}{4} \left[\frac{x e^{-x/2}}{-\frac{1}{2}} - \frac{e^{-x/2}}{(-\frac{1}{2})(-\frac{1}{2})} \right]_5^{\infty}$$

$$= \frac{1}{4} \left[-2 x e^{-x/2} - 4 e^{-x/2} \right]_5^{\infty}$$

$$P(X > 5) = \frac{7}{2} e^{-5/2} = 0.287$$