

28/09/23

(unit-4)

Special Continuous distribution

Topic

- *) Normal distribution and its MGF.
-) Gamma distribution and its MGF.
-) Exponential distribution and its MGF.

Normal distⁿ

*) $n = \text{large}$, $n \rightarrow \infty$

for \hookrightarrow continuous data.

$n \rightarrow \text{small}$

\hookrightarrow B.D

$n \rightarrow \text{large}$

\hookrightarrow P.D

\downarrow

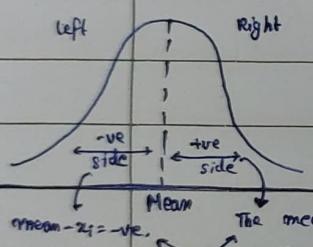
Discrete

mean \rightarrow avg value or mid value.

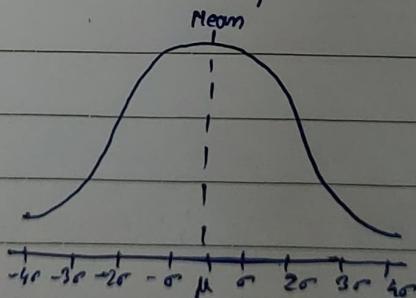
variance \rightarrow difference / deviation b/w

original data and mean data.

* if will tell you above the curve



*) Normal distⁿ is also known as "Gaussian distⁿ" or "Bell shape curve distⁿ"



*) A normal Random Vari^{bbe} is said to have a normal distⁿ with parameters μ (mean) and σ^2 (variance); if its PDF is given by:

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

we denote the value of the RV 'x' as:-

$$N(\mu, \sigma^2)$$

or

$$n(x; \mu, \sigma)$$

$$\text{i.e., } x \sim N(\mu, \sigma^2) \text{ or } X \sim n(x; \mu, \sigma)$$

* Properties of Normal distribution.

• The mean, median and mode are all equal or coinciding.

$$\boxed{\text{mean} = \text{median} = \text{mode}} \rightarrow \text{equal for normal distri.}$$

↳ arrange in ascending

L, then the middle term.

$$\text{Even} = \frac{\text{mth} + (\text{m+1})\text{th}}{2}$$

avg of
two
middle
value

$$\text{odd} = \frac{\text{mth}}{2}$$

some
sort of
thing

• The curve is symmetric about the origin ($\mu = \text{mean}$)

• The total area under the curve and above the

Horizontal axis is equal = 1.

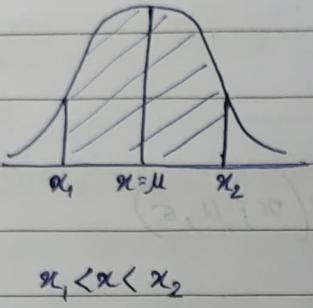
$$\text{i.e. } \int_{-\infty}^{\infty} f(x) dx = 1$$

• Area to the left and to the right of the mean (μ) is 0.5*

• The mode, which is the point of the horizontal axis, where the curve is maximum occurs at $[x = \mu (\text{mean})]$

↳

* Area under the curve :-



$$P(x_1 < x < x_2) = \int_{x_1}^{x_2} f(x) dx$$

$$= \int_{x_1}^{x_2} \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

$$= \frac{1}{\sigma \sqrt{2\pi}} \int_{x_1}^{x_2} \exp \left\{ \frac{-(x-\mu)^2}{2\sigma^2} \right\} dx. \quad \text{--- (1)}$$

It is
a very lengthy

approach so, we do have another method.

* If $x \sim N(\mu, \sigma^2)$ or $x \sim n(x; \mu, \sigma)$ for normal distn variable X with mean = 0 and var = 1; then the new variable Z is given by.

$\boxed{\begin{aligned} Z &= \frac{x-\mu}{\sigma} \\ \mu=0, \sigma^2=1 \end{aligned}}$, where Z is called standard normal variate

whenever, X assumes a value x , the corresponding value of Z is given by $\boxed{Z = \frac{x-\mu}{\sigma}}$

\therefore If X takes b/w the values, $x_1 = x_1$ and $x_2 = x_2$, then the random variable Z , will fall b/w the corresponding values,

$$Z_1 = \frac{x_1 - \mu}{\sigma} \quad \& \quad Z_2 = \frac{x_2 - \mu}{\sigma}$$

from eq ①

$$-\frac{(x-\mu)^2}{2\sigma^2} \rightarrow -\frac{z^2}{2} \rightarrow ②$$

$$\hookrightarrow \text{so from ①} \rightarrow P(x_1 < x < x_2) = \int_{z_1}^{z_2} \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{z^2}{2}} dz$$

for $z \rightarrow$
mean = 0
variance = 1

$$= \frac{1}{\sqrt{2\pi}} \int_{z_1}^{z_2} e^{-\frac{z^2}{2}} dz$$

$$\boxed{P(x_1 < x < x_2) = \frac{1}{\sqrt{2\pi}} \int_{z_1}^{z_2} e^{-\frac{z^2}{2}} dz}$$

$$P(x_1 < x < x_2) = P(z_1 < z < z_2)$$

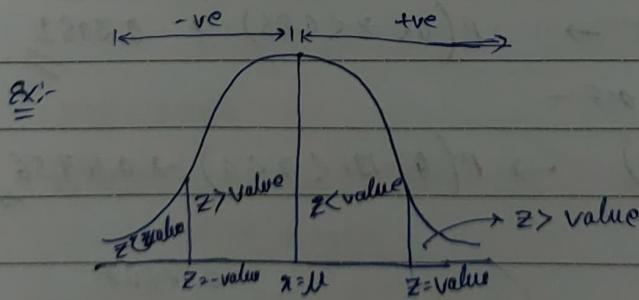
Q. Given a standard normal distribution, find the area given under the curve, which lies to the right of z ,

a) to the right of $z = 1.84$ and

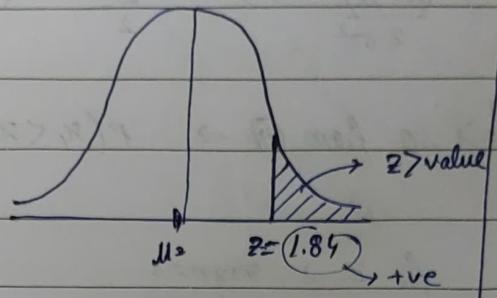
$< >$ → 1st Table

b/w $z = -1.97$ and $z = 0.86$

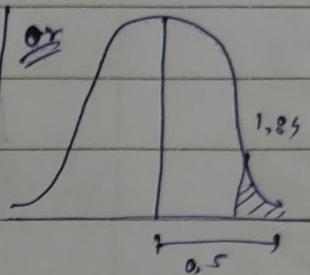
$0 \rightarrow$ value → third table.



a) to the right of $Z = 1.84$

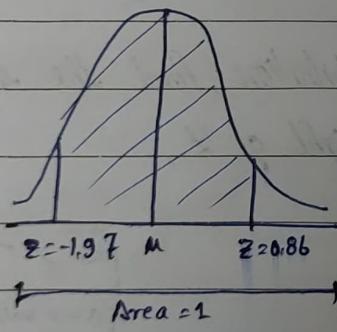


$$\begin{aligned} &= 1 - P(Z < 1.84) \\ &= 1 - \underbrace{0.9671}_{\text{+ve}} \\ &= 0.0329 \end{aligned}$$



$$\begin{aligned} P(Z > 1.84) &= 0.5 - P(0 < z < 1.84) \\ &= 0.5 - 0.4671 \\ &= 0.0329 \end{aligned}$$

b) b/w $Z = -1.97$ and $Z = 0.86$.



~~2 table~~ $P(Z < 0.86) \rightarrow P(0 < z < 0.86) \rightarrow 0.3051$

~~= 0.51~~

$$P(Z > -1.97) \rightarrow P(-1.97 < z < 0) \rightarrow 0.4756$$

~~= 0.7807~~

or

$$P(Z < 0.86) - P(Z < -1.97)$$

$$= 0.8051 - 0.0244$$

~~20.7807~~

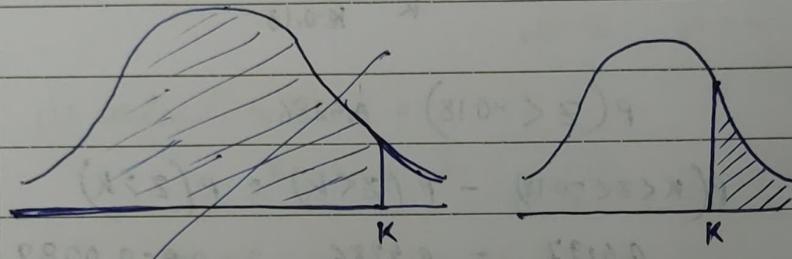
05/10/23

Q. Given a standard normal distribution, find the value of K such that

a) $P(Z > K) = 0.3015$

b) $P(K < Z < -0.18) = 0.4197$

a)



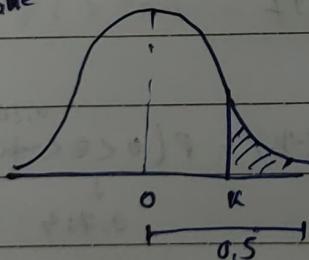
$$Z > K = 0.3015.$$

$$\begin{aligned} Z < K &= 1 - 0.3015 \\ &= 0.6985 \end{aligned}$$

$$20.8985$$

$$K = 0.52$$

2nd Table



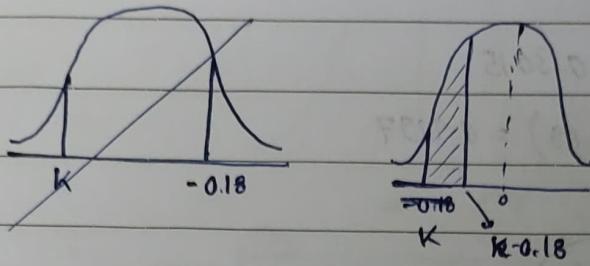
$$Z > K = 0.3015$$

$$P(0 < Z < K) = 0.5 - 0.3015$$

$$= 0.1985$$

$$\Rightarrow K = 0.52$$

$$b) P(K < Z < -0.18) = 0.4197$$

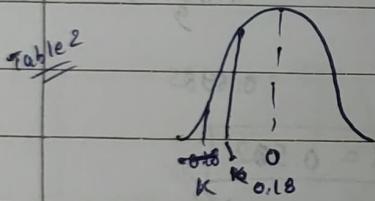


$$P(Z < -0.18) = 0.4286.$$

$$P(K < Z < -0.18) = P(Z < K) = P(Z > K)$$

$$0.4197 - 0.4286 = \cancel{0.42} - 0.0089$$

$$\boxed{K = 0.1867} \quad \boxed{K = 2.37}$$



$$P(K < Z < -0.18) = 0.4197$$

$$P(K < 0)$$

$$P(K < Z < -0.18) = P(0 < Z < -K) = P(0 < Z < \cancel{K}) = 0.4197$$

$\downarrow \qquad \downarrow$
 0.714

$$P(0 < Z < -K) = 0.4197 + 0.714$$

$$= 0.2993 + 0.3488 \quad (491)$$

$$K = \boxed{2.37}$$

Q X is normally distributed, and mean of $X = 12$, $\sigma = 4$.

Find the Probability of the following.

(i) $X \geq 20$

(ii) $X \leq 20$

(iii) $0 \leq X \leq 12$

$$\mu = 12 ; \sigma = 4$$

(i) $P(X \geq 20)$

$$X = 20 \rightarrow Z = \frac{X - \mu}{\sigma} = \frac{20 - 12}{4} = \frac{8}{4} = 2$$

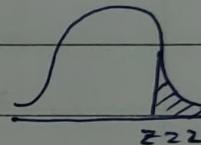
$$P(X \geq 20) \Rightarrow P(Z \geq 2)$$

Table 1

$$P(Z \geq 2) = 1 - P(Z < 2)$$

$$= 1 - 0.9772$$

$$= 0.0228$$



(ii) $P(X \leq 20) = P(Z \leq 2)$

$$= 1 - P(Z > 2)$$

$$= 1 - 0.0228$$

$$= 0.9772$$

Table 2

$$P(Z \geq 2) = 0.5 -$$

$$P(0 < Z < 2)$$

$$0.5 - 0.4772$$

$$0.0228$$

(iii) $X_1 = 0 , X_2 = 12$

$$X_1 = 0 \Rightarrow Z_1 = \frac{X_1 - \mu}{\sigma} = \frac{0 - 12}{4} = -3$$

$$-3 < Z < 0$$

$$X_2 = 12 \Rightarrow Z_2 = \frac{X_2 - \mu}{\sigma} = \frac{12 - 12}{4} = 0$$

Table 2
 $P(-3 < Z < 0) = 0.4987$

MGF of Normal Distribution

$$M_x(t) = \int_{-\infty}^{\infty} e^{tx} f(x) dx$$

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

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

$$= \frac{e^{\mu t}}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{t(x+\mu)} \cdot \exp\left\{-\frac{z^2}{2}\right\} dz$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{t(\mu + \sigma z)} \cdot e^{-\frac{z^2}{2}} dz$$

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

$$= \frac{e^{t\mu}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(z^2 - 2\sigma z t)} dz$$

$$z = \frac{x-\mu}{\sigma}$$

$$\sigma z = x - \mu$$

$$\sigma z + \mu = x$$

differentiate

$$\sigma dz + 0 = dx$$

$$\sigma dz = dx$$

→ performing, Now

$$\Rightarrow \frac{e^{t\mu}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(z^2 + \sigma^2 t^2 - 2\sigma z t - \sigma^2 t^2)} dz$$

(Completing Square.)

↳ coeff of z^2 must be 1

$$\Rightarrow \frac{e^{t\mu}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}((z - \sigma t)^2 - \sigma^2 t^2)} dz$$

↳ coeff of z = $2\sigma t$
↳ divide by $\frac{1}{2}$

$$\Rightarrow \frac{e^{t\mu}}{\sqrt{2\pi}} \cdot e^{\frac{1}{2}(\sigma^2 t^2)} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(z - \sigma t)^2} dz$$

$\Rightarrow \frac{(2\sigma t)^{\frac{1}{2}}}{\sqrt{\pi}}$
 $= \sigma t$

$$\Rightarrow \frac{e^{(tu + \frac{1}{2}\sigma^2 t^2)}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(b)^2} \cdot db$$

$$\begin{cases} z - st = b \\ dz - 0 = db \end{cases}$$

$$\Rightarrow \frac{e^{(tu + \frac{1}{2}\sigma^2 t^2)}}{\sqrt{2\pi}} \cdot 2 \int_0^{\infty} e^{-\frac{b^2}{2}} \cdot db$$

$$\int_a^a f(x) dx = 2 \int_0^a f(x) dx$$

$$\Rightarrow \frac{e^{(tu + \frac{1}{2}\sigma^2 t^2)}}{\sqrt{2\pi}} \cdot 2 \int_0^{\infty} e^{-c} \cdot \frac{1}{\sqrt{2c}} \cdot dc$$

$$\frac{b^2}{2} = c$$

$$\frac{1}{2}b \cdot b = dc \quad \frac{1}{2}b \cdot db = dc$$

$$\Rightarrow \frac{e^{(tu + \frac{1}{2}\sigma^2 t^2)}}{\sqrt{2\pi}} \int_0^{\infty} e^{-c} \cdot c^{-\frac{1}{2}} \cdot dc$$

$$db = \frac{dc}{b} \rightarrow \sqrt{2c}$$

$$\Rightarrow \frac{e^{(tu + \frac{1}{2}\sigma^2 t^2)}}{\sqrt{\pi}} \int_0^{\infty} c^{-\frac{1}{2}} \cdot c^{-\frac{1}{2}} \cdot e^{-c} \cdot dc$$

* Gamma function.

$$\Gamma_{\frac{1}{2}} = \sqrt{\pi}$$

value of Gamma $\frac{1}{2} = \sqrt{\pi}$

$$\Gamma_{\frac{1}{2}} \Rightarrow \int_0^{\infty} c^{-\frac{1}{2}} \cdot e^{-c} \cdot dc$$

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

$$\boxed{\Gamma_{\alpha} = \int x^{\alpha-1} \cdot e^{-x} \cdot dx, x > 0}$$

$$\Rightarrow \frac{e^{(tu + \frac{1}{2}\sigma^2 t^2)}}{\sqrt{\pi}} \cdot \Gamma_{\frac{1}{2}} \cdot \sqrt{\pi}$$

$$\text{ex:- } \Gamma_3 = \int x^2 \cdot e^{-x^3} \cdot dx$$

$$\boxed{M_x(t) = e^{(tu + \frac{1}{2}\sigma^2 t^2)}}$$

Normal Distribution as Limiting case of Binomial Distribution.

Normal Distribution is the another limiting case of Binomial Distribution just like poisson's Distribution.

The conditions are :-

- * (i) n , the no. of trials is indefinitely very large, ($i.e. n \rightarrow \infty$)
- (ii) neither p nor q is small. ($p, q \neq 0$)

If capital "X" is a binomial r.v with mean (μ) = np

and variance (σ^2) = npq , then the Limiting form of the

distribution is

$$Z = \frac{X - np}{\sqrt{npq}}$$

as $n \rightarrow \infty$, is the standard Normal Distribution, $N(Z; 0, 1)$

Remark:- The ND with $\mu = np$ and $\text{var} = npq$ not only provides a very accurate approximation to the binomial distribution, when n is Large, and p is not extremely close to 0 or 1, but also provides a fairly good approximation, even when n is small and p is close to $\frac{1}{2}$.



Q The prob that a patient recovers from a rare blood disease. (0.4)

If 100 people are known to have contacted this disease, what is the prob that fewer than 30 patient survive.

$$\rightarrow n = 100$$

$$p = 0.4$$

$$q = 0.6$$

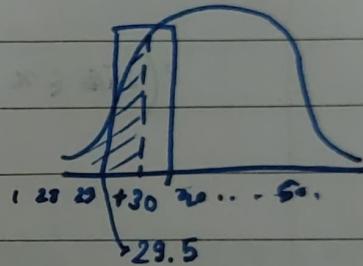
$$P(X < 30)$$

$$np = 100 \times 0.4 = 40$$

$$\sqrt{npq} = \sqrt{100 \times 0.4 \times 0.6} = 2\sqrt{6}$$

$$z = \frac{x - np}{\sqrt{npq}} \rightarrow x = ? = 29.5$$

$$z = \frac{29.5 - 40}{2\sqrt{6}} = -2.14$$



$$P(X \leq 30) = P(z \leq -2.14)$$

$$= 0.0162$$

Q A MCQ Quiz has 200 Question

each with 4 possible answers of which only one is correct.

What is the prob. of sheer (guesswork) yields from 25 to 30 correct answers for the 80 of the 200 Q problems about which the player has no knowledge.

$$\rightarrow n = 80$$

X : correct Q/A.

$$p = \frac{1}{4}, q = \frac{3}{4}$$

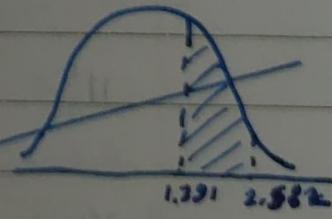
$$np = 80 \times \frac{1}{4} = 20$$

$$\sqrt{npq} = \sqrt{80 \times \frac{1}{4} \times \frac{3}{4}} = \sqrt{15} = 3.872$$

$$P(25 \leq X \leq 30) \\ = P(1.291 \leq Z \leq 2.582)$$

$$Z_1 = \frac{25 - 20}{\sqrt{15}} = \frac{5}{\sqrt{15}} \approx 1.291$$

$$Z_2 = \frac{30 - 20}{\sqrt{15}} = \frac{10}{\sqrt{15}} \approx 2.582$$

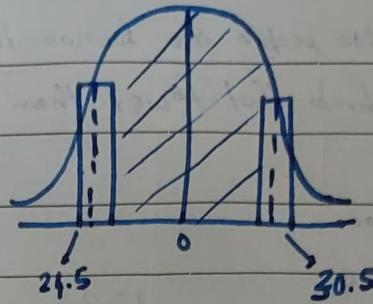


$$\Rightarrow n = 80$$

$$p = \frac{1}{4}, q = \frac{3}{4}$$

$$np = 20$$

$$\sqrt{npq} = \sqrt{15}$$



$$x_1 = 24.5$$

$$z_1 = \frac{24.5 - 20}{\sqrt{15}} = \frac{4.5}{\sqrt{15}} = \frac{4.5}{3.872} \approx 1.162$$

$$x_2 = 30.5$$

$$z_2 = \frac{30.5 - 20}{\sqrt{15}} = \frac{10.5}{\sqrt{15}} = \frac{10.5}{3.872} \approx 2.71$$

$$P(25 \leq x \leq 30) = P(1.162 \leq z \leq 2.71)$$

$$= P(-\infty < z \leq 2.71) - P(-\infty < z \leq 1.162)$$

$$= 0.4966 - 0.3770$$

$$\approx 0.1196$$

* Mean and Variance of Normal Distribution using MGF.

↳ Derivation of Mean and Variance.

$$\mu'_2 = E[x^2] = \left. \frac{d^2}{dx^2} M_x(t) \right|_{t=0}$$

(i) Mean

$$\mu = E(x) = \left. \frac{d}{dt} M_x(t) \right|_{t=0}$$

$$\text{Var}, \sigma^2 = E[X^2] = \frac{d^2}{dx^2} M_X t \Big|_{t=0}$$

(i) Mean $\Rightarrow E[X] = \frac{d}{dt} M_X t \Big|_{t=0}$

$$= \frac{d}{dt} \left[e^{\mu t + \frac{1}{2} \sigma^2 t^2} \right]_{t=0}$$

$$= e^{(\mu t + \frac{1}{2} \sigma^2 t^2)} \cdot (\mu + \frac{1}{2} \sigma^2 t) \Big|_{t=0}$$

$$= e^0 \cdot \mu$$

$$= \boxed{\mu}$$

(ii) Variance

$$\Rightarrow \sigma^2 = E[X^2] = \frac{d^2}{dx^2} M_X t \Big|_{t=0}$$

$$\Rightarrow \frac{d^2}{dt^2} \left[e^{\mu t + \frac{1}{2} \sigma^2 t^2} \right]_{t=0} = \frac{d}{dt} \left(e^{(\mu t + \frac{1}{2} \sigma^2 t^2)} \cdot (\mu + \frac{1}{2} \sigma^2 t) \right)$$

$$\Rightarrow (\mu + \frac{1}{2} \sigma^2 t) \left\{ \frac{d}{dt} \left(e^{\mu t + \frac{1}{2} \sigma^2 t^2} \right) \right\}$$

$$\Rightarrow \left(e^{\mu t + \frac{1}{2} \sigma^2 t^2} \right) \cdot \left(\frac{1}{2} \sigma^2 \right) + \left(\mu + \frac{1}{2} \sigma^2 t^2 \right) \cdot \left\{ e^{\mu t + \frac{1}{2} \sigma^2 t^2} \cdot \left(\mu + \frac{1}{2} \sigma^2 t^2 \right) \right\}_{t=0}$$

$$= e^0 \cdot \sigma^2 + e^0 \cdot \mu^2 = \boxed{\sigma^2 + \mu^2}$$

12/10/23

Gamma and Exponential Distribution.

Discrete Data				Continuous Data		
B.D	NGD	Geo	Posi	N.D	Gamma	Exp.
$n = \text{trial}$ $\rightarrow \text{small}$ $p \rightarrow \text{given}$ $\hookrightarrow \text{small}$	check for k^{th} trial success in n^{th} trial	check for k^{th} trial success	"on an avg of 1st success given question"	$n \rightarrow \text{large}$ $b \rightarrow \text{large}$ "Normally distributed form"	K^{th} event occurs in what time = ?	after K^{th} event the time period for the <u>$K+1^{\text{th}}$</u> event

K^{th} event occurs

in what particular time = ??

* Gamma Distribution is used to find time until
 K events occur.

Gamma function is defined.

Ex:- The time until K customers arrived in a
particular restaurant.

$$\star \quad \Gamma(x) = \int_0^{\infty} t^{x-1} \cdot e^{-t} dt, \quad x > 0$$

Ex:- The time you have been invited to
cake party.

properties.

$$i) F_n = (n-1)(n-2) \dots 1$$

$$ii) F_n = (n-1)!$$

$$iii) F_n = \Gamma_1 = 1 \quad 0!$$

$$iv) \Gamma_{1/2} = \sqrt{\pi}$$

Definition :- The continuous RV ' X ' has gamma distribution with parameters α and β ,

if its PDF is given by :-

$$\text{where } x > 0 \quad \left. \begin{array}{l} \\ f(x; \alpha, \beta) = \begin{cases} \frac{1}{\beta^\alpha \Gamma(\alpha)} x^{\alpha-1} e^{-x/\beta}, & x > 0 \\ 0, & \text{elsewhere} \end{cases} \end{array} \right\} \quad \beta > 0 \quad \star$$

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Remark :- The relation b/w λ and β poisson distribution.

$$\Rightarrow \lambda = \frac{1}{\beta}$$

where λ = Average no. of events per unit time.

λ = No. of Events.

β = Avg time b/w no. of events.

Exponential Distribution. :- The continuous RV ' X ' has an exp. distribution with parameter " β ", if its pdf is given by.

It occurs when $\alpha = 1$

$$f(x; \beta) = \begin{cases} \frac{1}{\beta} e^{-x/\beta}, & x > 0 \\ 0, & \text{elsewhere} \end{cases} \quad **$$

	Mean	Variance.
Gamma distribution	$\alpha\beta$	$\alpha\beta^2$
Exponential distribution	β	β^2

MGF of Gamma distribution.

$$M_{X,t} = E[e^{tx}] = \int_0^\infty e^{tx} f(x) dx$$

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

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

$$M_{X,t} = E[e^{tx}] = \int_0^\infty e^{tx} f(x) dx$$

$$\text{Put } \left(\frac{1}{\beta}-t\right)x = u$$

$$x = \frac{u}{\frac{1}{\beta}-t}$$

$$dx = \frac{du}{\frac{1}{\beta}-t}$$

$$\begin{aligned}
 &= \frac{1}{\beta^\alpha \Gamma(\alpha)} \int_0^\infty e^{-u} \left(\frac{u}{\beta} - t \right)^{\alpha-1} \cdot \frac{du}{\frac{u}{\beta} - t} \\
 &= \frac{1}{\beta^\alpha \Gamma(\alpha)} \cdot \frac{1}{\left(\frac{u}{\beta} - t \right)^{\alpha-1}} \cdot \frac{1}{\left(\frac{u}{\beta} - t \right)} \cdot \int_0^\infty e^{-u} \cdot u^{\alpha-1} \cdot du \\
 &= \frac{1}{\beta^\alpha \Gamma(\alpha)} \cdot \frac{\beta^{\alpha-1}}{1-t}
 \end{aligned}$$

$$M_{X(t)} = \frac{1}{(1-\beta t)^\alpha}$$

Derivation of Mean and Variance of Gamma distribution by using MGF.

$$M'_x = E[X^t] = \frac{d}{dt} M_x(t) \Big|_{t=0}$$

$$\mu = E[X] = \frac{d}{dt} M_x(t) \Big|_{t=0}$$

$$\Rightarrow \frac{d}{dt} \left[\frac{1}{(1-\beta t)^\alpha} \right] \Bigg|_{t=0}$$

$x' = 1$
 $\frac{1}{x} = x^{-1}$ or $=^{-1}x^{-2}$

$$\cancel{\frac{1}{(1-\beta t)^{\alpha-1}}} \cancel{\frac{1}{\alpha-1}}$$

$$\Rightarrow \frac{d}{dt} \left[(1-\beta t)^{-\alpha} \right] \Bigg|_{t=0}$$

$$\Rightarrow (-\alpha) \cdot (1-\beta t)^{\alpha-1} \cdot (0-\beta) \Bigg|_{t=0}$$

$$\Rightarrow (-\alpha), (1-\alpha), (-\beta)$$

$$\boxed{\mu = \alpha\beta} \quad \star\star$$

Variance by MGF

$$\begin{aligned}\sigma^2 &= E[X^2] - (E[X])^2 \\ &= \frac{d^2}{dt^2} \left[\frac{1}{(1-\beta t)^\alpha} \right] \Big|_{t=0} \Rightarrow \frac{d^2}{dt^2} \left[(1-\beta t)^{-\alpha} \right] \Big|_{t=0}\end{aligned}$$

$$\Rightarrow \left(\frac{d}{dt} \left[\alpha \beta (1-\beta t)^{-\alpha-1} \right] \right) \Big|_{t=0}$$

$$\Rightarrow \alpha \beta \cdot ((-\alpha-1)(1-\beta t)^{-\alpha-2} \cdot (0-\beta))$$

$$\Rightarrow \alpha \beta \cdot ((-\alpha-1)(1-\alpha) \cdot (-\beta))$$

$$\Rightarrow \alpha \beta [\alpha \beta + \beta]$$

$$\Rightarrow \alpha^2 \beta^2 + \alpha \beta^2$$

$$\text{Now, } \sigma^2 = \alpha^2 \beta^2 + \alpha \beta^2 - \alpha^2 \beta^2$$

$$\boxed{\sigma^2 = \alpha \beta^2} \quad \star\star$$

MGF of Exponential Distribution.

$$\Rightarrow M_X(t) = E[e^{tx}] = \int_0^\infty e^{tx} f(x) dx$$

$$\Rightarrow \int_0^\infty e^{tx} \left(\frac{1}{\beta} e^{-\frac{x}{\beta}} \right) dx \Rightarrow \frac{1}{\beta} \int_0^\infty e^{tx} e^{-\frac{x}{\beta}} dx \Rightarrow \frac{1}{\beta} \int_0^\infty e^{-(\frac{1}{\beta}-t)x} dx$$

$$\Rightarrow \frac{1}{\beta} \int_0^\infty e^{-u} \cdot \frac{du}{\frac{1}{\beta}-t} \quad \text{but } \Rightarrow \left(\frac{1}{\beta}-t \right) x = u$$

$$\Rightarrow \frac{1}{\beta} \cdot \frac{1}{\frac{1}{\beta}-t} \int_0^\infty (\mu^{\alpha-1}) e^{-u} \cdot du \quad \left(\text{Since in Exp } (\alpha=1) \right) \quad x = \frac{u}{\frac{1}{\beta}-t}$$

$$\text{we insert } \mu^{\alpha-1} \Rightarrow \mu^{\alpha-1} = \mu^0$$

$$dx = \frac{du}{\frac{1}{\beta}-t}$$

$$\Rightarrow \frac{1}{P} \cdot \frac{\beta}{\beta t + \beta t} \cdot \int_0^{\infty} e^{-\beta t} \cdot \mu^{t-1} dt$$

(1)(0-1)(1-0)

$$\Rightarrow M_X(t) = \frac{1}{1-\beta t} \quad \star\star$$

* Mean of Exponential by using MGF

$$\mu' = E[X] = \frac{d}{dt} M_X(t) \Big|_{t=0}$$

$$= \frac{d}{dt} \left(\frac{1}{1-\beta t} \right) \Big|_{t=0} = \frac{d}{dt} \left((1-\beta t)^{-1} \right) \Big|_{t=0}$$

$$= (-1) \cdot (1-\beta t)^{-2} (-\beta) \Big|_{t=0}$$

$$\Rightarrow (-1)(-\beta)$$

$$\boxed{\mu = \beta} \quad \star\star$$

* Variance of Exponential by using MGF

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

$$= \frac{d^2}{dt^2} M_X(t) \Big|_{t=0} = \frac{d^2}{dt^2} \left[\frac{1}{1-\beta t} \right] \Big|_{t=0}$$

$$= \frac{d}{dt} \left((-1)(1-\beta t)^{-2} (-\beta) \right) \Big|_{t=0} = \frac{d}{dt} \left(\beta \cdot (1-\beta t)^{-2} \right) \Big|_{t=0}$$

$$= -2\beta \cdot (1-\beta t)^{-3} (-\beta) \Big|_{t=0}$$

$$\begin{aligned} \sigma^2 &= 2\beta^2 - \beta^2 \\ \sigma^2 &= \beta^2 \end{aligned} \quad \star\star$$

$$E[X^2] = \boxed{2\beta^2}$$

16/10/23

Q Suppose that telephone calls arriving at a call centre with an avg of 5 calls coming per minute. What is the prob that up to a minute, \downarrow collapse by the time two calls have come into the telephone ^{will} calls.

$$\lambda = 5$$

$$\alpha = 2$$

$$\beta = \frac{1}{5}$$

X: Time will elapse before 2 calls.

$$P(X \leq 1) = ?$$

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

$$\Rightarrow \int_0^{\infty} \frac{1}{(1/5)^2} \cdot x^1 \cdot e^{-5x} dx$$

$$\Rightarrow \left(\frac{1}{25} \right) \int_0^{\infty} x^1 e^{-5x} dx$$

$$\Rightarrow \left[\frac{1}{25} \left(x^1 e^{-5x} \Big|_0^\infty - \int_0^\infty 1 e^{-5x} dx \right) \right]$$

function
by part.
ISLATE

$$\Rightarrow 25 \left[x^1 \cdot \frac{e^{-5x}}{-5} \Big|_0^\infty - \int_0^\infty 1 \cdot e^{-5x} dx \right]$$

$$\Rightarrow 25 \left[-5 \left[e^{-5x} \right] \Big|_0^\infty - \frac{1}{25} \cdot e^{-5x} \Big|_0^\infty \right]$$

$$I \int II - \int \left(\frac{d}{dx} I \int II \right) dx$$

$$I \int II - \int \left[\frac{d}{dx} I \int II \right] dx$$

$$e^0 = 1$$

$$Ans = 0.96$$

Q In a bio-medical study with rats are those responds investigation is used to determine the effect of the dose of the toxicants on their survival. The toxicant is one that is frequently discharge into the atmosphere from the jet fuel. for a certain dose of the toxicant the study determines that the survival time in week as a gamma distribution $\alpha = 5$, $\beta = 10$.

What is the prob that a rat survives no longer than 60 weeks.

$$\Rightarrow \alpha = 5, \beta = 10. \quad P(X \leq 60 \text{ week}) = ?$$

$$\int_0^{60} \frac{1}{\beta^{\alpha}} \cdot x^{\alpha-1} \cdot e^{-x/\beta} dx$$

$$P(X \leq 60) = \int_0^{60} \frac{1}{(10^5)(24)} \cdot x^4 \cdot e^{-x/10} dx$$

$$= \frac{1}{(10^5)(24)} \int_0^{60} x^4 \cdot e^{-x/10} dx \quad \begin{matrix} \text{technique when } e^{\text{variable}} \\ \text{diff} \end{matrix}$$

$$= \frac{1}{(10^5)(24)} \left[x^4 \cdot e^{-x/10} - 4x^3 \cdot e^{-x/10} + 12x^2 \cdot e^{-x/10} - 24x \cdot e^{-x/10} + 24 \cdot e^{-x/10} \right]_0^{60}$$

$$\Rightarrow \frac{1}{10^5(24)} \left[-10x^4 \cdot e^{-x/10} - 400x^3 \cdot e^{-x/10} + 12000x^2 \cdot e^{-x/10} - 2400x \cdot e^{-x/10} + 240000 \cdot e^{-x/10} \right]_0^{60}$$

$$\text{Ans} = 0.77$$