

Chapter-5

5.2: Exponential Distribution

$X \rightarrow$ A continuous Random Variable

$\lambda \rightarrow$ parameter

(Exponential distribution)

PDF (Probability density function),

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & ; x \geq 0 \\ 0 & ; x < 0 \end{cases}$$

CDF (Cumulative Distribution Function),

$$F(x) = \int_{-\infty}^x f(y) dy = \begin{cases} 1 - e^{-\lambda x} & ; x \geq 0 \\ 0 & ; x < 0 \end{cases}$$

Proof:

$$F(x) = \int_{-\infty}^x f(y) dy$$

$$= \int_{-\infty}^x \lambda e^{-\lambda y} dy$$

$$= \lambda \left[\frac{e^{-\lambda x}}{-\lambda} \right]_{\infty}^x$$

$$= 1 - e^{-\lambda x}$$

$$P\{X \leq x\} = 1 - e^{-\lambda x} = F(x)$$

$$P\{X > x\} = 1 - F(x) = e^{-\lambda x}$$

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

$$= \int_0^{\infty} \lambda x e^{-\lambda x} dx$$

$$= \left[\lambda x \frac{e^{-\lambda x}}{-\lambda} - \frac{e^{-\lambda x}}{(-\lambda)^2} \lambda \right]_0^{\infty} = \frac{1}{\lambda}$$

$$\therefore E[X] = \frac{1}{\lambda}$$

// mean of Exponential distbⁿ.

$$\rightarrow \varphi(t) = E[e^{tx}] = \int_0^{\infty} e^{tx} \lambda e^{-\lambda x} dx$$

$$= \lambda \left[\frac{e^{(t-\lambda)x}}{t-\lambda} \right]_0^{\infty} = \frac{\lambda}{\lambda - t}$$

$$\boxed{\varphi(t) = \frac{\lambda}{\lambda - t} \quad t < \lambda}$$

$\varphi(t) \rightarrow$ Moment Generating Function

$$E[X^2] = \frac{d^2}{dt^2} [\varphi(t)] \Big|_{t=0}$$

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

$$= \frac{2\lambda}{(\lambda - t)^3} \Big|_{t=0}$$

$$= \frac{2}{\lambda^2}$$

$$\text{Var}(X) = E[X^2] - (E[X])^2$$

$$= \frac{2}{\lambda^2} - \left(\frac{1}{\lambda} \right)^2$$

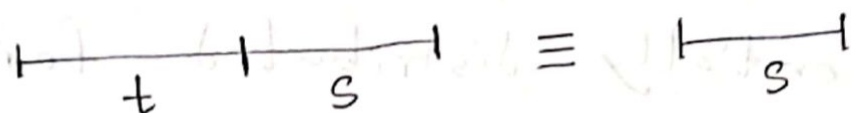
$$= \frac{1}{\lambda^2}$$

$E[X] = \frac{1}{\lambda}$	$\varphi(t) = \frac{\lambda}{\lambda - t}$
$E[X^2] = \frac{2}{\lambda^2}$	$P\{X \leq x\} = 1 - e^{-\lambda x}$
$\text{Var}(X) = \frac{1}{\lambda^2}$	$P\{X > x\} = e^{-\lambda x}$

Properties of Exponential Distribution

A random variable X is said to be without memory or memoryless if

$$P\{X > s+t \mid X > t\} = P\{X > s\}, \quad s, t \geq 0$$



→ Doesn't deteriorate with time.

→ An item that has been used for 't' hours is as good as a new item in regards to the amount of time remaining until the item fails.

→ The item doesn't remember that it has already been in use for time t.

$$\text{So, } P\{X > s+t \mid X > t\} = P\{X > s\}$$

$$\Rightarrow \frac{P\{X > s+t, X > t\}}{P\{X > t\}} = P\{X > s\}$$

$$\therefore P\{X > s+t\} = P\{X > s\} P\{X > t\}$$

Since, $P\{X > s+t\}$ and $P\{X > t\}$
 // means $P\{X > s+t\}$
 This eqⁿ is satisfied if X

is exponentially distributed for

$$e^{-\lambda(s+t)} = e^{-\lambda s} e^{-\lambda t}, \text{ it follows}$$

that exponentially distributed RVs
 are memoryless.

$$P\{X > s+t\} = P\{X > s\} P\{X > t\}$$

$$e^{-\lambda(s+t)} = e^{-\lambda s} \cdot e^{-\lambda t}$$

Example: 5.2

X : Amount of time a customer spends in a bank.

λ : Arrival rate, $\lambda = \frac{1}{10}$ customer/sec

$$\begin{aligned}\textcircled{i} \quad P\{X > 15\} &= e^{-\lambda x} \\ &= e^{-\lambda \times 15} = e^{-\frac{1}{10} \times 15} \\ &= e^{-3/2}\end{aligned}$$

$$\begin{aligned}\textcircled{ii} \quad P\{X > 15 \mid X > 10\} \\ &= P\{X > \underbrace{10}_{\textcircled{5} + \textcircled{10}} + 5 \mid X > 10\} \quad \oplus\end{aligned}$$

$$\begin{aligned}&= P\{X > 5\} \\ &= e^{-\frac{1}{10} \times 5} \\ &= e^{-1/2}\end{aligned}$$

15 second এর জন্য
সম্ভাব্য probability given
10 second already passed
is equal to
5 second এর জন্য
সম্ভাব্য probability
for a new customer
(Memory less)

Exponential Distribution is the only memoryless distribution

Proof:

Let X is memoryless and $\bar{F}(x) = P\{X > x\}$

$$\bar{F}(s+t) = P\{X > s+t\} = P\{X > s\} P\{X > t\}$$

$$\bar{F}(s+t) = \bar{F}(s) \bar{F}(t)$$

$\bar{F}(x)$ satisfies the functional eqⁿ,

$$g(s+t) = g(s) g(t)$$

The only solution of this equation is

$$g(x) = e^{-\lambda x}$$

$$\therefore g(s+t) = e^{-\lambda(s+t)} = e^{-\lambda s} \cdot e^{-\lambda t} = g(s) g(t)$$

Since, a distribution function is always right continuous, we must have,

$$\bar{F}(x) = e^{-\lambda x}$$

$$\therefore F(x) = P\{X \leq x\} = 1 - e^{-\lambda x}$$

So, X is exponentially distributed
[Proved]

Example:

Lifetime of a bulb is exponentially distributed with mean 10 hours.

$$P\{\text{lifetime} > t+s \mid \text{lifetime} > t\}$$

$$= \frac{1 - F(t+s)}{1 - F(t)} \quad // \text{ if lifetime was not exponential.}$$

But now,

$$P\{\text{lifetime} > t+s \mid \text{lifetime} > t\}$$

$$= P\{\text{lifetime} > s\}$$

$$= 1 - F(s)$$

$$= 1 - (1 - e^{-\lambda \times s})$$

$$= 1 - (1 - e^{-\left(\frac{1}{10} \times s\right)})$$

$$\lambda = \frac{1}{10}$$

$$\lambda = \frac{1}{\text{mean}}$$

$$= e^{-\frac{1}{2}}$$

Note:

$$P\{X \leq s\} = F(s) \quad \text{and} \quad P\{X > s\} = \bar{F}(s) = 1 - F(s)$$

Further Properties of Exp. Distr.

1 ଟି Bulb average 5 hours ଭୁଲ,
ତାହାଙ୍କର " " " 10 " " .

ପ୍ରଥମ Bulb ଟି / ଦ୍ୱିତୀୟ Bulb ଯଦି ଗୋଟିଏ
ସମୟ ଭୁଲ, ସମ୍ଭାବନା କେତେ?

X_1 and X_2 are independent

Random Variables with means $\frac{1}{\lambda_1}, \frac{1}{\lambda_2}$

$$P\{X_1 < X_2\} = \int_0^{\infty} P\{X_1 < X_2, X_1 = x\} dx$$

$$= \int_0^{\infty} P\{X_1 < X_2 | X_1 = x\} \lambda_1 e^{-\lambda_1 x} dx$$

$$= \int_0^{\infty} P\{x < X_2\} \lambda_1 e^{-\lambda_1 x} dx$$

$$= \int_0^{\infty} e^{-\lambda_2 x} \lambda_1 e^{-\lambda_1 x} dx \quad \left| \begin{array}{l} p(X > x) = 1 - F(x) \\ = 1 - (1 - e^{-\lambda_2 x}) \\ = e^{-\lambda_2 x} \end{array} \right.$$

$$= \lambda_1 \int_0^{\infty} e^{-(\lambda_1 + \lambda_2)x} dx$$

$$= \lambda_1 \left[\frac{e^{-(\lambda_1 + \lambda_2)x}}{-(\lambda_1 + \lambda_2)} \right]_0^{\infty}$$

$$= \frac{\lambda_1}{\lambda_1 + \lambda_2}$$

$$\boxed{P\{X_1 < X_2\} = \frac{\lambda_1}{\lambda_1 + \lambda_2}}$$

$$\text{So, } \lambda_1 = \frac{1}{5}, \quad \lambda_2 = \frac{1}{10}$$

$$\therefore P\{X_1 < X_2\} = \frac{\frac{1}{5}}{\frac{1}{5} + \frac{1}{10}} = \frac{2}{3}$$

Example

$$\lambda_1 = \frac{1}{1000}$$

$$\lambda_2 = \frac{1}{500}$$

$$P\{X_1 < X_2\} = \frac{\lambda_1}{\lambda_1 + \lambda_2}$$

$$= \frac{\frac{1}{1000}}{\frac{1}{500} + \frac{1}{1000}}$$

$$= \frac{1}{3}$$

5.3 Counting Process

A stochastic process is said to be a counting process if $N(t)$ represents the total number of events that have occurred up to time t .

Examples:

persons entered store

of births # of goals.

Properties

- (i) $N(t) \geq 0$
- (ii) $N(t)$ is integer
- (iii) if $s < t$, $N(s) \leq N(t)$
- (iv) for $s < t$, $N(t) - N(s) = \#$ of events during interval $(s, t]$



① Independent Increment:

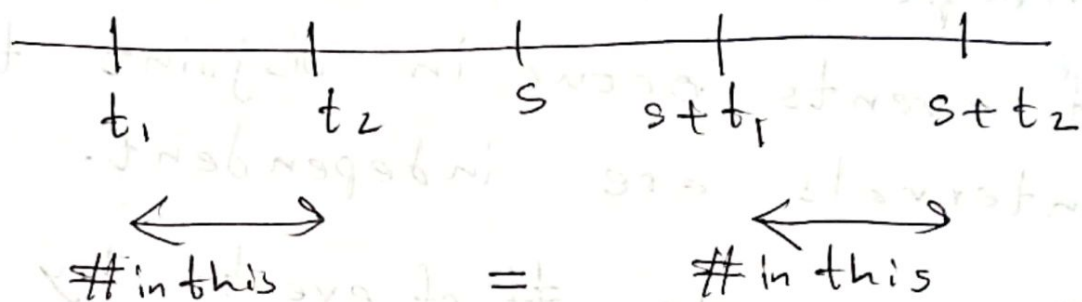
A counting process is said to possess independent increments if the number of events occur in disjoint time intervals are independent.

For example, $\#$ of events by time 10 $\downarrow N(10)$ must be independent of $\#$ of events during $[10, 15]; (N(15) - N(10))$.

② Stationary Increment

କ୍ଷେତ୍ର interval ଥିବା କାଳୀନ କାଳୀନ ଗ
interval ଥିବା length ଥିବା ସମସ୍ତ ସମୟରେ
A counting process is said to possess stationary increments if the distribution of the number of events that occur in any interval of time depends only on the length of the interval.

$$\text{i.e. } N(t_2 + s) - N(t_1 + s) \\ = N(t_2) - N(t_1)$$



5.3

Poisson Process:

A counting process $\{N(t) ; t \geq 0\}$ is said to be a poisson process having rate $\lambda, (\lambda > 0)$ if

- (i) $N(0) = 0$
- (ii) The process has independent increments
- (iii) The process has stationary increments.

i.e. The # of events in an interval of length t is poisson distributed with mean (λt) , $\lambda > 0$

$$P\{N(t+s) - N(s) = n\} = e^{-\lambda t} \frac{(\lambda t)^n}{n!}$$

$$\therefore E[N(t)] = \lambda t$$

$$\forall s, t \geq 0; n = 0, 1, 2, \dots$$

(5) $N(t+s)$ ਤਰ੍ਹਾਂ n ਦੀ count

ਦੀ probability λ ਦੇ t ਦੇ

n ਦੀ count ਦੀ probability same

ਹੈ interval ਤਰ੍ਹਾਂ $(t+s-s=t)$

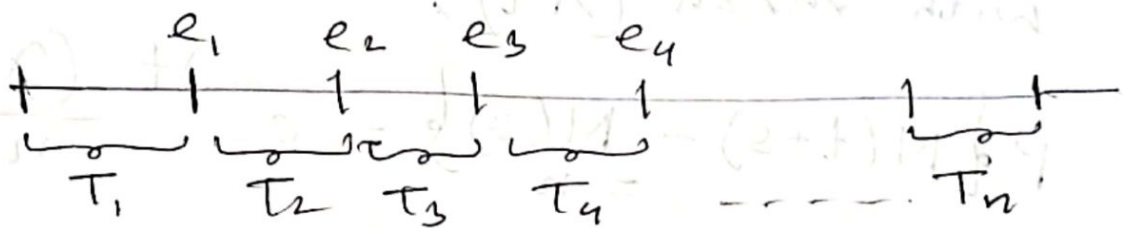
Note:

of events in length t is poisson distributed but the time is exponentially distributed.

λ is rate of the process (Memoryless)

Distribution of Inter-Arrival time

Inter-arrival time : 2nd event or
ଅନ୍ୟତମ ଘଟଣା,



$\{T_n, n = 1, 2, 3, \dots\}$ interarrival time.

$$\begin{aligned} P\{T_1 > t\} &= P\{\text{No events in } (0, t]\} \\ &= P\{N(t) = 0\} = \frac{e^{-\lambda t} (\lambda t)^0}{0!} \\ &= e^{-\lambda t} \quad [n=0] \end{aligned}$$

Hence, T_1 has exponential distribution
with parameter λ , mean $\frac{1}{\lambda}$.

Note

$P\{T_1 > t\}$ means first event
occurs after time
t, so no event in $(0, t]$.

$$\begin{aligned}
 & P \{ T_2 > t \mid T_1 = s \} \\
 &= P \{ \text{No events in } (s, s+t] \mid T_1 = s \} \\
 &= P \{ \text{No events in } (s, s+t] \} \\
 &= P \{ 0 \text{ events in } (0, t] \} \quad \begin{array}{l} \text{[independent increment,} \\ T_1 = s \text{]} \end{array} \\
 &= e^{-\lambda t} \quad \text{[stationary increment]}
 \end{aligned}$$

$\therefore T_2$ is exponentially distributed with mean $\frac{1}{\lambda}$, and T_2 is independent of T_1 .

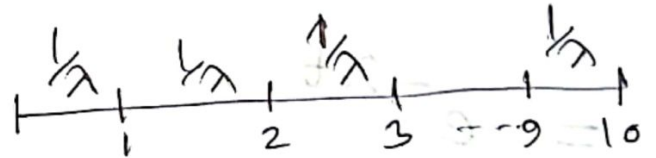
So, T_n is also exp. distr. with parameter λ , mean $\frac{1}{\lambda}$.

$S_n \rightarrow$ the time of n th event.

$$S_n = \sum_{i=1}^n T_i, \quad n \geq 1$$

Example

People migrating into a territory at a poisson rate $\lambda = 1$ per day.



(i) $E[S_{10}]$

$$= \frac{10}{\lambda} = 10 \text{ days}$$

(ii) $P\{T_{11} > 2\} = e^{-2\lambda}$

$$= e^{-2}$$

$$\approx 0.133.$$