

# Poisson process

## Lemma

*Definition 1  $\implies$  Definition 2*

Proof on board.

## Lemma

*Definition 2  $\implies$  Definition 1*

Self Study: Refer Sheldon Ross, Stochastic processes, Theorem 2.1.1

## Poisson Processes Definition 3

A ctsp  $\{N(t), t \geq 0\}$  is a Poisson process with rate  $\lambda > 0$  if

- ▶  $N(0) = 0$
- ▶  $N(t)$  is a counting process with stationary and independent increments
- ▶  $X_i$ , the time interval between  $i - 1$ th and  $i$ th event is exponentially distributed with parameter  $\lambda$ .

## Lemma

*Definition 1/2  $\implies$  Definition 3*

### Proof:

- ▶ What is  $P(X_1 > t)$  =?

$$P(X_1 > t) = P(N(0, t) = 0) = e^{-\lambda t}$$

- ▶ This implies  $F_{X_1}(t) = P(X_1 \leq t) = 1 - e^{-\lambda t}$  and hence  $X_1$  has exponential distribution.
- ▶ What is  $P(X_2 > t | X_1 = s)$ ?

$$\begin{aligned} P(X_2 > t | X_1 = s) &= P(N(s, t + s] = 0 | X_1 = s) \\ &= P(N(s, t + s] = 0) \text{ (indep. increments)} \\ &= e^{-\lambda t} \text{ (stat. increments)} \end{aligned}$$

- ▶ This implies  $X_2$  is exponential. Repeating the arguments yields the lemma.

## Definition 3 $\implies$ Definition 1

### Lemma

*i.i.d exponential interarrival time implies  $N(0, t)$  has Poisson distribution with rate  $\lambda t$ .*

- ▶ Let  $S_0 = 0$  and  $S_n = \sum_{i=1}^n X_i$
- ▶ If  $S_n = t$ , we say that the  $n$ th renewal happened at time  $t$ .
- ▶  $f_{S_n}(t) = \lambda \left[ \frac{(\lambda t)^{n-1} e^{-\lambda t}}{n-1!} \right]$  and  $F_{S_n}(t) = \int_{x=0}^t \lambda \left[ \frac{(\lambda x)^{n-1} e^{-\lambda x}}{n-1!} \right] dx$

## More on $F_{S_n}(t)$

$$\blacktriangleright F_{S_n}(t) = \int_0^t \lambda \left[ \frac{(\lambda x)^{n-1} e^{-\lambda x}}{n-1!} \right] dx$$

$$\blacktriangleright \text{Integration by parts } (u(x) = e^{-\lambda x}, \quad v'(x) = \lambda \left[ \frac{(\lambda x)^{n-1}}{n-1!} \right])$$

$$\int_a^b u(x) v'(x) dx = [u(x) v(x)]_a^b - \int_a^b u'(x) v(x) dx$$

$$\blacktriangleright F_{S_n}(t) = \left[ \frac{(\lambda x)^n e^{-\lambda x}}{n!} \right]_0^t - \int_0^t \left[ \frac{-\lambda e^{-\lambda x} (\lambda x)^n}{n!} \right] dx$$

$$\blacktriangleright F_{S_n}(t) = \frac{(\lambda t)^n e^{-\lambda t}}{n!} + F_{S_{n+1}}(t)$$

$$F_{S_n}(t) - F_{S_{n+1}}(t) = \frac{(\lambda t)^n e^{-\lambda t}}{n!}$$

## Relation between $S_n$ and $N(t)$

$$N(t) = \sup\{n : S_n \leq t\}$$

$$N(t) \geq n \Leftrightarrow S_n \leq t$$

- ▶  $P\{N(t) \geq n\} = P\{S_n \leq t\}$
- ▶  $P\{N(t) = n\} = P\{N(t) \geq n\} - P\{N(t) \geq n+1\}.$
- ▶  $P\{N(t) = n\} = P\{S_n \leq t\} - P\{S_{n+1} \leq t\}.$
- ▶  $P\{N(t) = n\} = \text{Poisson}(\lambda t).$

### Lemma

*Exponential interarrival times imply  $N(t)$  has Poisson distribution with rate  $\lambda t$*

## Properties of Poisson Process (Self Study)

Merging: Merging two independent Poisson processes with rate  $\lambda_1$  and  $\lambda_2$  leads to a Poisson process with rate  $\lambda_1 + \lambda_2$ .

Splitting: If you label each event point of a  $\text{Poisson}(\lambda)$  process as type A or type B with probability  $p$  or  $1 - p$  respectively, then Events of type A form a  $\text{Poisson}(p\lambda)$  process. Similarly Events of type B form a  $\text{Poisson}((1 - p)\lambda)$  process.

# Conditional distribution of Arrival times

## Lemma

*Given that 1 event of  $P.P.(\lambda)$  has happened by time  $t$ , it is equally likely to have happened anywhere in  $[0, t]$  i.e.,*

$$P\{X_1 < s | N(t) = 1\} = \frac{s}{t}.$$

## Proof.

$$\begin{aligned} P\{X_1 < s | N(t) = 1\} &= \frac{P\{X_1 < s, N(t) = 1\}}{P(N(t) = 1)} \\ &= \frac{P\{N[0, s) = 1, N[s, t] = 0\}}{P(N(t) = 1)} \\ &= \frac{P\{N[0, s) = 1\}P\{N[s, t] = 0\}}{P(N(t) = 1)} \\ &= \frac{\lambda s e^{-\lambda s} e^{-\lambda(t-s)}}{\lambda t e^{-\lambda t}} = \frac{s}{t} \end{aligned}$$





## Conditional distribution of Arrival times

Doubt

Theorem 2.3.1 from Sheldon Ross

### Lemma

*Given that  $N(t) = n$ , the joint distribution of the arrival times of these  $n$  jobs is  $\frac{n!}{t^n}$ .*

Proof: Self Study