

# Continuous-time Markov Chain: Poisson Process 2



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The state probabilities are given by

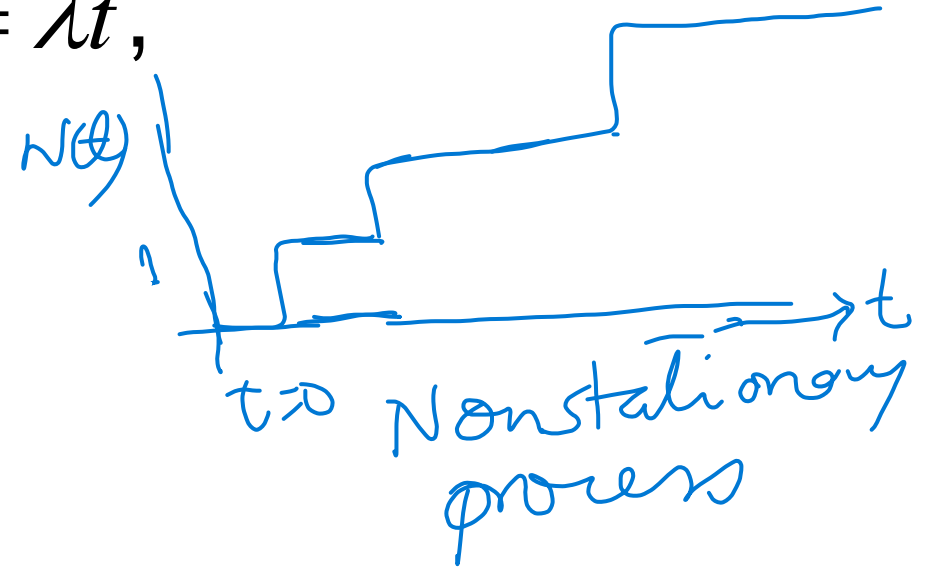
$$p_j(t) = P(N(t) = j) = e^{-\lambda t} \frac{(\lambda t)^j}{j!}, j = 0, 1, \dots$$

The statistics of  $\{N(t)\}$  is given by

$$EN(t) = \lambda t, \quad \text{var}(N(t)) = \lambda t,$$

$$C_N(t_1, t_2) = \lambda \min(t_1, t_2)$$

$$R_N(t_1, t_2) = \lambda \min(t_1, t_2) + \lambda^2 t_1 t_2$$



# MS Continuity and Differentiability of a Poisson Process

For a Poisson process  $\{N(t)\}$ ,

$$R_N(t_1, t_2) = \lambda \min(t_1, t_2) + \lambda^2 t_1 t_2$$

$$\therefore R_N(t, t) = \lambda t + \lambda^2 t^2$$

Now

$$\begin{aligned} \lim_{t_1 \rightarrow t, t_2 \rightarrow t} R_N(t_1, t_2) &= \lambda \lim_{t_1 \rightarrow t, t_2 \rightarrow t} \min(t_1, t_2) + \lambda^2 \lim_{t_1 \rightarrow t, t_2 \rightarrow t} t_1 t_2 \\ &= \lambda t + \lambda^2 t^2 \end{aligned}$$

Thus the autocorrelation function of a Poisson process is continuous at each  $(t, t)$  implying that a Poisson process is m.s. continuous everywhere.

## Differentiability

For a Poisson process  $\{N(t)\}$ ,

$$R_N(t_1, t_2) = \lambda \min(t_1, t_2) + \lambda^2 t_1 t_2$$

$$\therefore R_N(0, t_2) = \lambda \min(0, t_2)$$

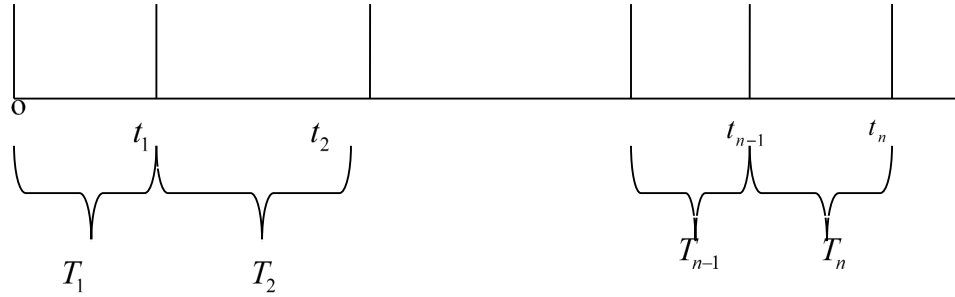
$$= \begin{cases} \lambda t_2 & \text{if } t_2 < 0 \\ 0 & \text{otherwise} \end{cases}$$

$$\therefore \frac{\partial R_N(0, t_2)}{\partial t_2} = \begin{cases} \lambda & \text{if } t_2 < 0 \\ 0 & \text{if } t_2 > 0 \\ \text{does not exist} & \text{if } t_2 = 0 \end{cases}$$

$$\therefore \frac{\partial^2 R_N(t_1, t_2)}{\partial t_1 \partial t_2} \text{ does not exist at } (t_1 = 0, t_2 = 0)$$

Thus,  $\{N(t)\}$ , is not m.s. differentiable.

## Inter-arrival time and Waiting time for the Poisson Process



$P(\text{No event occur during } [0, t])$

$$P(T_1 > t) = e^{-\lambda t}$$
$$f_T(t) = \lambda \cdot e^{-\lambda t} \quad t \geq 0$$

Let  $T_n$  = time elapsed between the  $(n-1)$ th event and the  $n$ th event. The random process  $\{T_n, n = 1, 2, \dots\}$  represent the *inter-arrival time* of the From the CTMC theory, it is clear that  $T_n \sim \exp(\lambda), n = 1, 2, \dots$  (identically distributed)

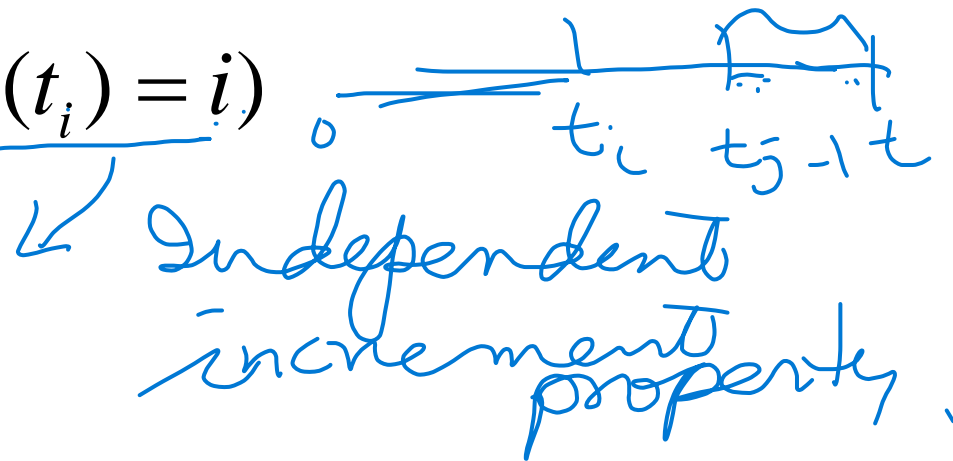
To prove independence of  $T_n$ s, consider the conditional probability  $P(T_j > t \mid T_i = t_i)$

$$P(T_j > t \mid T_i = t_i) = P(T_j > t \mid N(t_i) = i)$$

$$= P(N(t_{j-1}, t] = 0 \mid N(t_i) = i)$$

$$= P(N(t_{j-1}, t] = 0)$$

$$= P(T_j > t)$$



Thus  $T_n$ s are iid RVs with

$$f_{T_n}(t) = \lambda e^{-\lambda t} \quad n > 0$$

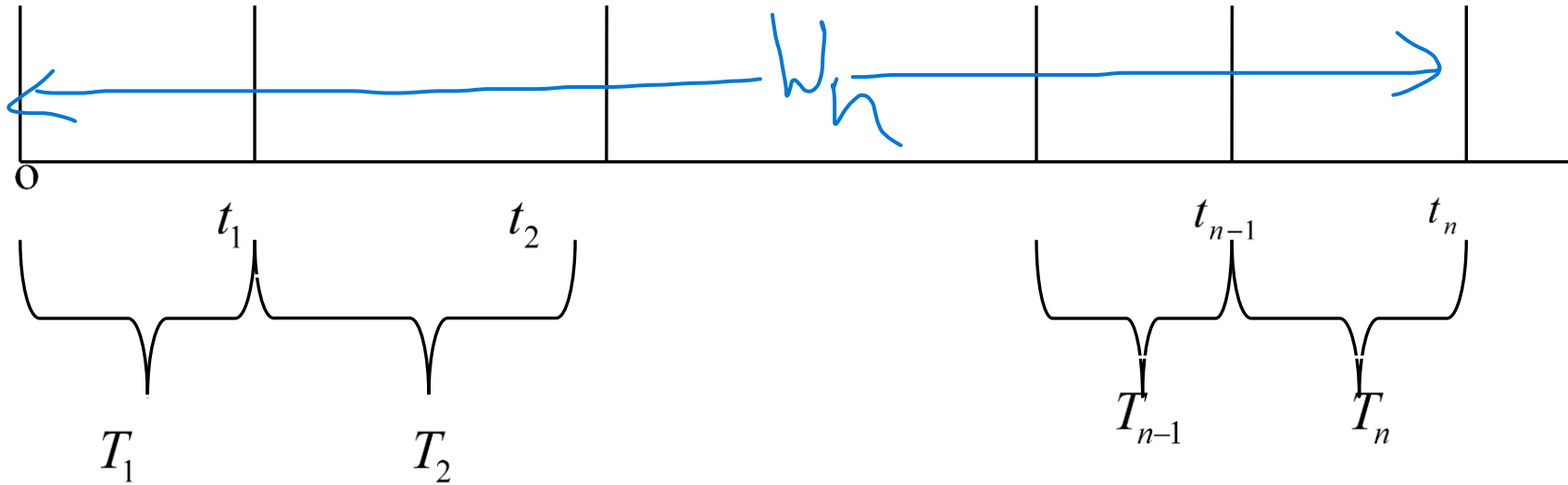
Memoryless property.

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

$$\text{Var}(T_n) = \frac{1}{\lambda^2}$$

$$P(T > s+t \mid T > s) = P(T > t)$$

# Waiting time



Now let us analyse the *waiting time*  $W_n$ . This is the time that elapses before the  *$n$ th* event occurs.

$\therefore W_n = \sum_{i=1}^n T_i$  ( sum of  $n$  independent exponential random variables)

*Gamma RV.*

## Proof

We note that

$$F_{W_n}(t) = P(\{W_n \leq t\})$$

$$= P(\{N(t) \geq n\})$$

$$= \sum_{k=n}^{\infty} \frac{(\lambda t)^k e^{-\lambda t}}{k!}$$

$$\therefore f_{W_n}(t) = \frac{d}{dt}(F_{W_n}(t))$$

$$= \sum_{k=n}^{\infty} \left( \frac{\lambda(\lambda t)^{k-1} e^{-\lambda t}}{k-1!} - \frac{\lambda(\lambda t)^k e^{-\lambda t}}{k!} \right)$$

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

Thus  $W_n$  is a gamma random variable.

$$W_n = \sum_{i=1}^n T_i$$

$$E W_n = \sum_{i=1}^n E T_i$$
$$Var(W_n) = \sum_{i=1}^n Var(T_i)$$

$$= \sum_{i=1}^n \frac{1}{\lambda} = \frac{n}{\lambda}$$
$$= \sum_{i=1}^n Var(T_i) = \frac{n}{\lambda^2}$$

$$\begin{aligned} & \lambda e^{-\lambda t} \frac{(\lambda t)^{n-1}}{(n-1)!} - \lambda e^{-\lambda t} \frac{(\lambda t)^n}{n!} \\ & + \lambda e^{-\lambda t} \frac{(\lambda t)^{n-1}}{(n-1)!} - \dots \end{aligned}$$



$$\int_0^{\infty} f_{w_n}(t) dt = \frac{\int_0^{\infty} \lambda e^{-\lambda t} (\lambda t)^{n-1} dt}{\Gamma(n)}$$

Gamma with integer shape parameter is Erlang distribution

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

$$\lambda t = u \quad \Gamma(n) = \int_0^{\infty} e^{-x} x^{n-1} dx$$



## Example 2

The number of customers arriving at a service station is a Poisson process with a rate of 2 customers per minute.

- (a) What is the mean arrival time of the customers?
- (b) What is the probability that the first customer arrives after 30 second.
- (c) Given that there is no arrival before 1 min, what is the probability that first arrival will be after 3 min.
- (d) Given that the third customer has arrived at  $t=2$  min, what is the probability that fourth customer will arrive after  $t=4$  min?
- (e) What is the average waiting time before the 10<sup>th</sup> customer arrives?

$$\begin{aligned} \lambda &= 2/\text{min} \\ E T_i &= \frac{1}{\lambda} = \frac{1}{2} \text{ min} = 30 \text{ sec} \\ P(T_1 > 30) &= e^{-\lambda t} = e^{-2 \times \frac{1}{2}} = e^{-1} \end{aligned}$$

$$P(T_1 > 3 | T_1 > 1) \stackrel{[Memoryless \text{ property}]}{=} P(T_1 > 2)$$

$$P(T_4 > 4 | T_1 = 2) \stackrel{[Independence \text{ properties of } T_i]}{=} P(T_4 > 4)$$

$$EW_{10} = 10 \times \frac{1}{2} = 5 \text{ min}$$