
Poisson Process

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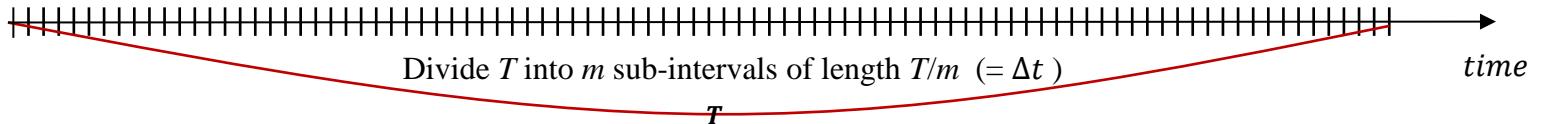
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Definition

- (Homogeneous) Poisson Process
 - The counting process $\{N(t), t \geq 0\}$ is Poisson process having rate λ ($\lambda > 0$) if
 - 1) $N(0) = 0$
 - 2) The events in $(t, t + \Delta t]$ is independent of t
 - 3) The probability of an event in $(t, t + \Delta t]$ is $\lambda \Delta t + o(\Delta t)$
 - 4) The probability of no event in $(t, t + \Delta t]$ is $1 - \lambda \Delta t + o(\Delta t)$
where $o(\Delta t)$ is any function such that $\lim_{\Delta t \rightarrow 0} \frac{o(\Delta t)}{\Delta t} = 0$
 - When $\Delta t \rightarrow 0$, it implies that
 - Probability of an event within Δt : $\lambda \Delta t$
 - Probability of no event within Δt : $1 - \lambda \Delta t$
 - Probability of more than one event within $\Delta t \approx 0$
- ⇒ This can be seen as Bernoulli process

Probability distribution

- Representation with Bernoulli process



- When $\Delta t \rightarrow 0$, $m \rightarrow \infty$
- X : random variable representing the number of events during T
- Since an event occurs at each subinterval according to Bernoulli process with probability $\lambda\Delta t$,

$$\begin{aligned}\Pr\{X = n\} &= \lim_{m \rightarrow \infty} \binom{m}{n} (\lambda\Delta t)^n (1 - \lambda\Delta t)^{m-n} \\ &= \lim_{m \rightarrow \infty} \frac{m!}{n! (m-n)!} \frac{(\lambda T)^n}{m^n} \frac{\left(1 - \lambda \frac{T}{m}\right)^m}{\left(1 - \lambda \frac{T}{m}\right)^n} \\ &= \frac{(\lambda T)^n}{n!} \lim_{m \rightarrow \infty} \frac{m!}{(m-n)! m^n} \frac{\left(1 - \lambda \frac{T}{m}\right)^m}{\left(1 - \lambda \frac{T}{m}\right)^n} \\ &= \frac{(\lambda T)^n}{n!} \lim_{m \rightarrow \infty} \left(1 - \lambda \frac{T}{m}\right)^m = \frac{(\lambda T)^n}{n!} e^{-\lambda T}\end{aligned}$$

⇒ $\Pr\{X = n\} = \frac{(\lambda T)^n}{n!} e^{-\lambda T}$

Exponential Distribution Property (1)

1. The inter-event time is exponentially distributed

<Proof>

- Let $A_c(t)$ denote the probability of no event in $(t_0, t_0 + t]$
- $A_c(t + \Delta t) = Pr\{\text{no event in } (t_0, t_0 + t]\} \times Pr\{\text{no event in } (t_0 + t, t_0 + t + \Delta t]\}$

$$A_c(t + \Delta t) = A_c(t)(1 - \lambda\Delta t + o(\Delta t))$$

$$A_c(t + \Delta t) - A_c(t) = -\lambda\Delta t A_c(t) + o(\Delta t)A_c(t)$$

$$\lim_{\Delta t \rightarrow 0} \frac{A_c(t + \Delta t) - A_c(t)}{\Delta t} = \lim_{\Delta t \rightarrow 0} (-\lambda A_c(t) + \frac{o(\Delta t)A_c(t)}{\Delta t})$$

$$A'_c(t) = -\lambda A_c(t)$$

$$\frac{A'_c(t)}{A_c(t)} = -\lambda \quad \dots (1)$$

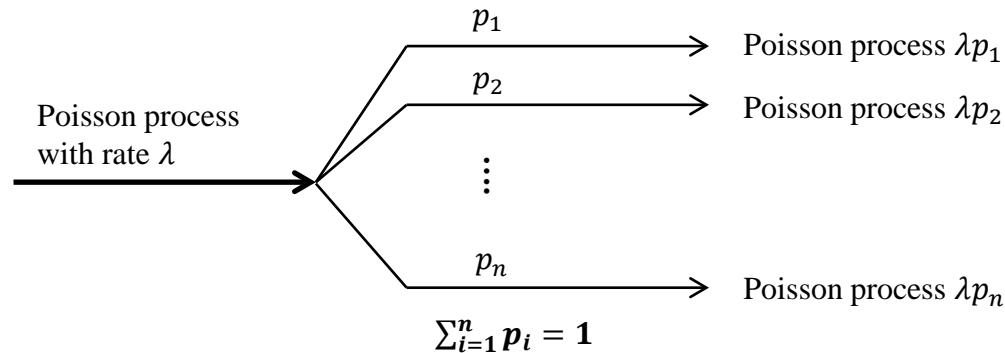
Exponential Distribution Property (2)

- From (1)
 - $\int \frac{A'_c(t)}{A_c(t)} = \ln A_c(t) = -\lambda t + k$
- Since $A_c(0) = 1$
 - $\ln A_c(0) = k = 0$
- Thus, $\ln A_c(t) = -\lambda t$
 - $A_c(t) = e^{-\lambda t}$
- If X is a random variable of inter-event time
 - $A_c(t) = Pr\{X > t\} = e^{-\lambda t}$
 - $Pr\{X \leq t\} = 1 - e^{-\lambda t}$

The last term implies that X is exponentially distributed.

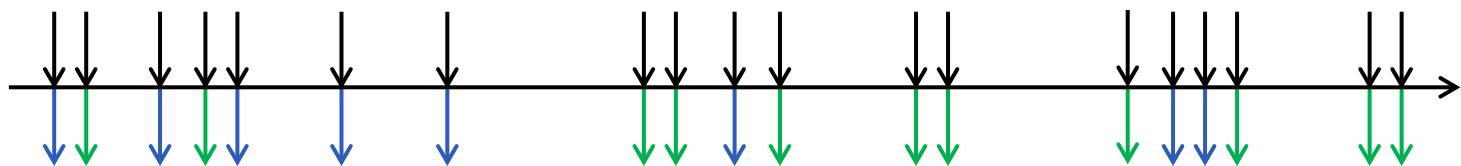
Decomposition Property (1)

2. Decomposition



– Example

- Event arrival follows Poisson process (black arrow)
- Each event is classified with probability
 - ✓ blue arrow with p_1 and green arrow with $(1 - p_1)$
- According to decomposition property, these are also respectively Poisson processes



Decomposition Property (2)

2. Decomposition

< Proof >

- Consider a Poisson process with rate λ
- Each event is categorized into two types
 - Type I with probability p
 - Type II with probability $(1 - p)$
- Let $N_1(t)$ be the number of type I events in $[0,t]$
Let $N_2(t)$ be the number of type II events in $[0,t]$
Let $N(t)$ be the total number of events in $[0,t]$
 - $N(t) = N_1(t) + N_2(t)$

Decomposition Property (3)

< Proof (cont.) >

$$Pr\{N_1(t) = n, N_2(t) = m\}$$

$$= Pr\{N(t) = n + m\} Pr\{N_1(t) = n, N_2(t) = m | N(t) = n + m\}$$

$$= \frac{(\lambda t)^{m+n}}{(m+n)!} e^{-\lambda t} \times \binom{m+n}{n} p^n (1-p)^m$$

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

$$P_r\{N_1(t) = n\} = \sum_{m=0}^{\infty} P_r\{N_1(t) = n, N_2(t) = m\}$$

$$= \sum_{m=0}^{\infty} \frac{(\lambda pt)^n}{n!} \frac{(\lambda(1-p)t)^m}{m!} e^{-\lambda t} = \frac{(\lambda pt)^n}{n!} e^{-\lambda t} \sum_{m=0}^{\infty} \frac{(\lambda(1-p)t)^m}{m!}$$

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

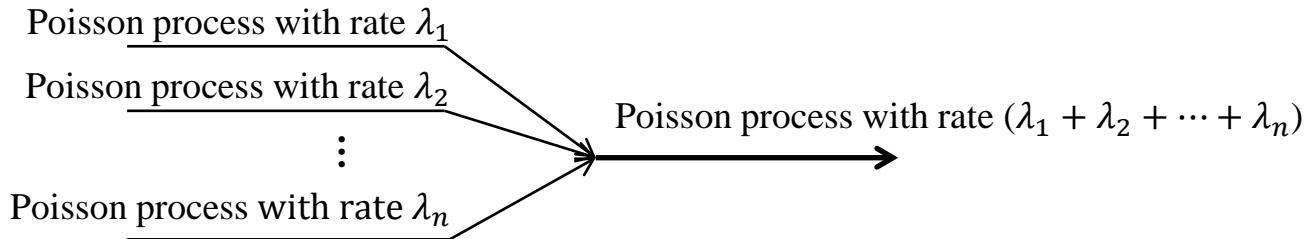
$$\text{Thus, } Pr\{N_1(t) = n\} = \frac{(\lambda pt)^n}{n!} e^{-\lambda pt}$$

$$\text{Similarly, } Pr\{N_2(t) = m\} = \frac{(\lambda(1-p)t)^m}{m!} e^{-\lambda(1-p)t}$$

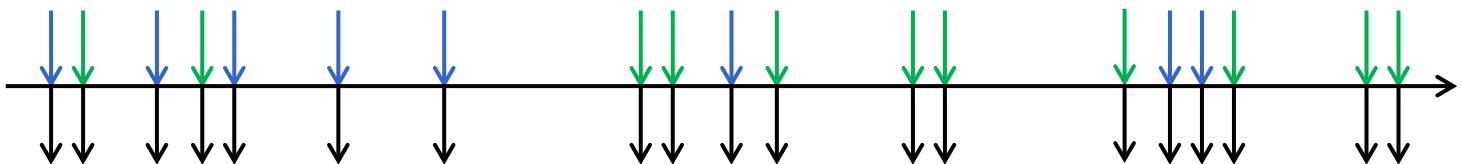
Poisson process

Superposition Property (1)

3. Superposition



- Example
 - There are two types of events (blue and green), which respectively arrive according to Poisson process
 - The combination of those events (black arrows) also follows Poisson process



Superposition Property (2)

<Proof>

- Consider two Poisson processes with rate λ_1 and λ_2

$$\begin{aligned} Pr\{N(t) = n\} &= \sum_{k=0}^n Pr\{N_1(t) = k\} \times Pr\{N(t) = n | N_1(t) = k\} \\ &= \sum_{k=0}^n \frac{(\lambda_1 t)^k}{k!} e^{-\lambda_1 t} \times \frac{(\lambda_2 t)^{n-k}}{(n-k)!} e^{-\lambda_2 t} \\ &= \frac{e^{-(\lambda_1 + \lambda_2)t}}{n!} \sum_{k=0}^n \frac{n!}{k! (n-k)!} (\lambda_1 t)^k (\lambda_2 t)^{n-k} \\ &= \frac{((\lambda_1 + \lambda_2)t)^n}{n!} e^{-(\lambda_1 + \lambda_2)t} \end{aligned}$$

- Therefore, $\{N(t), t \geq 0\}$ is a Poisson process with rate $(\lambda_1 + \lambda_2)$

Non-Homogeneous Poisson Process (NHPP)

- Definition
 - The counting process $\{N(t), t \geq 0\}$ is NHPP with rate $\lambda(t), t \geq 0$ if
 - $N(0) = 0$
 - $\Pr\{N(t + \Delta t) - N(t) = 0\} = 1 - \lambda(t)\Delta t + o(\Delta t)$
 - $\Pr\{N(t + \Delta t) - N(t) = 1\} = \lambda(t)\Delta t + o(\Delta t)$
 - $\Pr\{N(t + \Delta t) - N(t) \geq 2\} = o(\Delta t)$
- Derivation of pdf

$$\begin{aligned}\Pr\{N(t + \Delta t) = 0\} \\ = \Pr\{N(t) = 0\} \times \Pr\{N(t + \Delta t) - N(t) = 0\}\end{aligned}$$

Let $P_n(t) := \Pr\{N(t) = n\}$. Then

$$P_0(t + \Delta t) = P_0(t)(1 - \lambda(t)\Delta t + o(\Delta t)) \quad \dots \quad (1)$$

$$P_n(t + \Delta t) = \sum_{k=0}^n P_{n-k}(t) \Pr\{N(t + \Delta t) - N(t) = k\} \quad \dots \quad (2)$$

NHPP (2)

- From (1)

$$\frac{P_0(t + \Delta t) - P_0(t)}{\Delta t} = -P_0(t)\lambda(t) + \frac{P_0(t)o(\Delta t)}{\Delta t}$$

$$\lim_{\Delta t \rightarrow 0} \frac{P_0(t + \Delta t) - P_0(t)}{\Delta t} = -P_0(t)\lambda(t) + \lim_{\Delta t \rightarrow 0} \frac{o(\Delta t)}{\Delta t} P_0(t)$$

$$\Rightarrow \quad \frac{dP_0(t)}{dt} = -P_0(t)\lambda(t) \quad \dots (3)$$

- From (2)

$$\begin{aligned} P_n(t + \Delta t) &= P_n(t) \Pr\{N(t + \Delta t) - N(t) = 0\} + P_{n-1}(t) \Pr\{N(t + \Delta t) - N(t) = 1\} \\ &\quad + \sum_{k=2}^n P_{n-k}(t) \Pr\{N(t + \Delta t) - N(t) = k\} \\ &= P_n(t)(1 - \lambda(t)\Delta t + o(\Delta t)) + P_{n-1}(t)(\lambda(t)\Delta t + o(\Delta t)) + o(\Delta t) \sum_{k=2}^n P_{n-k}(t) \end{aligned}$$

$$\begin{aligned} \lim_{\Delta t \rightarrow 0} \frac{P_n(t + \Delta t) - P_n(t)}{\Delta t} &= -\lambda(t)P_n(t) + \lambda(t)P_{n-1}(t) + \lim_{\Delta t \rightarrow 0} \frac{o(\Delta t)}{\Delta t} \sum_{k=2}^n P_{n-k}(t) \\ \Rightarrow \quad \frac{dP_n(t)}{dt} &= -\lambda(t)P_n(t) + \lambda(t)P_{n-1}(t) \quad \dots (4) \end{aligned}$$

NHPP (3)

- Define PGF for random variable $N(t)$: $\mathbb{P}(z, t) := \sum_{n=0}^{\infty} P_n(t) z^n$
- Differentiate: $\frac{d\mathbb{P}(Z, t)}{dt} = \frac{dP_0(t)}{dt} + z \frac{dP_1(t)}{dt} + z^2 \frac{dP_2(t)}{dt} + \dots$
- From (3), (4)

$$\begin{aligned}\frac{d\mathbb{P}(Z, t)}{dt} &= -\lambda(t)P_0(t) \\ &\quad - z\lambda(t)P_1(t) + z\lambda(t)P_0(t) \\ &\quad - z^2\lambda(t)P_2(t) + z^2\lambda(t)P_1(t) \\ &\quad \dots \\ &= -\lambda(t)(P_0(t) + zP_1(t) + z^2P_2(t) + \dots) \\ &\quad + z\lambda(t)(P_0(t) + zP_1(t) + z^2P_2(t) + \dots) \\ &= -\lambda(t)\mathbb{P}(z, t) + z\lambda(t)\mathbb{P}(z, t)\end{aligned}$$

$$\frac{d\mathbb{P}(z, t)}{dt} = -\lambda(t)(1 - z)\mathbb{P}(z, t)$$

$$-\left. \frac{d\mathbb{P}(z, t)}{dt} \right|_{\mathbb{P}(z, t)} = -\lambda(t)(1 - z) \quad \dots (5)$$

NHPP (4)

- Integrating (5)

$$\int_0^t \frac{d\mathbb{P}(z, u)}{du} \Big/ \mathbb{P}(z, u) du = \int_0^t -\lambda(u)(1 - z) du + K(z)$$

$$\ln \mathbb{P}(z, t) = \int_0^t -\lambda(u)(1 - z) du + K(z)$$

- Since $N(0) = 0$, $P_0(0) = 1$, $P_n(0) = 0$ for $n \geq 1 \Rightarrow \mathbb{P}(z, 0) = 1$

$$\ln \mathbb{P}(z, 0) = \int_0^0 -\lambda(u)(1 - z) du + K(z) = 0 \Rightarrow K(z) = 0$$

- $\ln \mathbb{P}(z, t) = - \int_0^t \lambda(u) du + z \int_0^t \lambda(u) du$

$$\mathbb{P}(z, t) = e^{- \int_0^t \lambda(u) du} e^{z \int_0^t \lambda(u) du}$$

NHPP (5)

- Since $e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!}$, $\mathbb{P}(z, t) = e^{-\int_0^t \lambda(u)du} e^{z \int_0^t \lambda(u)du}$

$$\sum_{n=0}^{\infty} P_n(t) z^n = e^{-\int_0^t \lambda(u)du} \sum_{n=0}^{\infty} \frac{(z \int_0^t \lambda(u)du)^n}{n!}$$

$$= \sum_{n=0}^{\infty} \frac{(\int_0^t \lambda(u)du)^n}{n!} e^{-\int_0^t \lambda(u)du} z^n$$

- Therefore,

$$P_n(t) = \frac{(\int_0^t \lambda(u) du)^n}{n!} e^{-\int_0^t \lambda(u)du}$$