

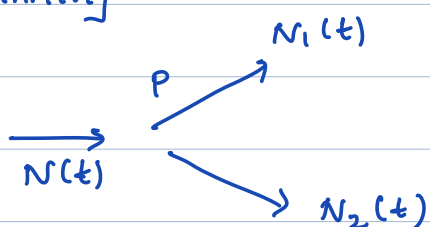
Announcements 1. Lab tmrw : Finish Qns 1, 2, 4

2. ~~OH Proposal: Move to Monday~~

Test 2: 9

Poisson Process

Thinning

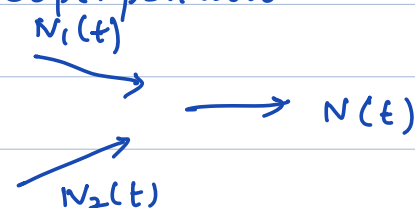


$$N(t) \sim PP(\lambda)$$

$$N_1(t) \sim PP(\lambda p)$$

$$N_2(t) \sim PP(\lambda(1-p))$$

Superposition



$$N_1(t) \sim PP(\lambda_1)$$

$$N_2(t) \sim PP(\lambda_2)$$

$$N(t) = N_1(t) + N_2(t)$$

$$\sim PP(\lambda_1 + \lambda_2)$$

Superposition (Proof)

$$N(t) = N_1(t) + N_2(t)$$

$$N_1(t) \sim PP(\lambda_1)$$

$$N_2(t) \sim PP(\lambda_2)$$

N_1 & N_2 are indep

Independent Increments

$$N(t) - N(s) = N_1(t) + N_2(t) - (N_1(s) + N_2(s))$$

$$= (N_1(t) - N_1(s)) + (N_2(t) - N_2(s))$$

$$\stackrel{||}{=} \underbrace{N_1(r)}_{r \leq s} \quad \text{and} \quad \underbrace{N_2(r)}_{r \leq s}$$

$$= \stackrel{||}{=} N(r) \quad r \leq s$$

Stationary Increment

$$N(t) - N(s) = \underbrace{N_1(t) - N_1(s)}_{\text{Poisson}(\lambda_1(t-s))} + \underbrace{N_2(t) - N_2(s)}_{\text{Poisson}(\lambda_2(t-s))}$$

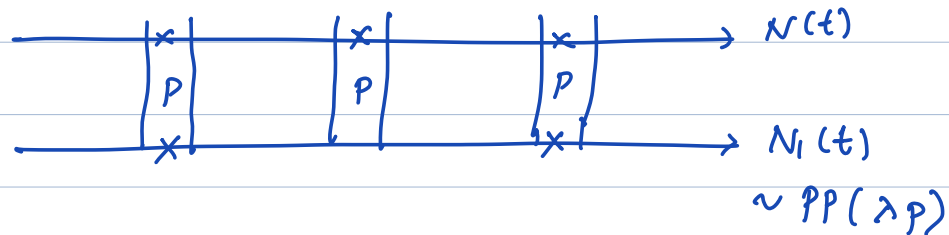
Review : $X \sim \text{Poisson}(\lambda_x)$ $Y \sim \text{Poisson}(\lambda_y)$

$$X+Y \sim \text{Poisson}(\lambda_x + \lambda_y)$$

$$\sim \text{Poisson}((\lambda_1 + \lambda_2)(t-s))$$

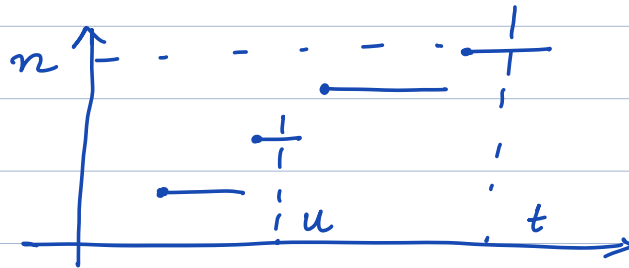
$$N(t) \sim \text{PP}(\lambda_1 + \lambda_2)$$

* Thinning : Exercise !



Properties on Conditional Distributions of PP.

Prop1:



$$P[X(u)=k / X(t)=n] = \binom{n}{k} \left(\frac{u}{t}\right)^k \left(1 - \frac{u}{t}\right)^{n-k}$$

$0 \leq u \leq t$
 $0 \leq k \leq n$

Proof

$$P[X(u)=k | X(t)=n]$$

$$= \frac{P[X(u)=k, X(t)=n]}{P[X(t)=n]}$$

$$= P [X(u) = k, X(t) - X(u) = n - k]$$

$$P [X(t) = n]$$

Poisson($\lambda(t-u)$)

$$P [X(t-u) = n - k]$$

independent increments

$$P [X(u) = k] P [X(t) - X(u) = n - k]$$

$$P [X(t) = n]$$

Poisson(λu)

Poisson(λt)

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

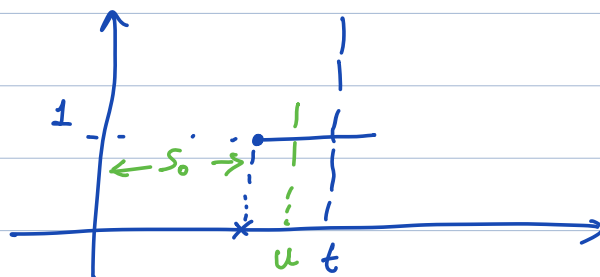
$$\frac{e^{-\lambda t} (\lambda t)^n}{n!}$$

$$= \frac{n!}{k! (n-k)!}$$

$$\frac{u^k (t-u)^{n-k}}{t^n}$$

$$= \binom{n}{k} \left(\frac{u}{t} \right)^k \left(1 - \frac{u}{t} \right)^{n-k} \sim \text{Bin}(n, \frac{u}{t})$$

Prop 2:



$$P [S_0 \leq u | X(t) = 1] = \frac{u}{t}$$

"Uniform arrival in $[0, t]$ "

Proof:

$$\rightarrow \frac{P(S_0 \leq u, X(t) = 1)}{P(X(t) = 1)}$$

$$X(u) = 1$$

$$= \frac{P(X(u) = 1, X(t) = 1)}{P(X(t) = 1)}$$

$$= \frac{P(X(u) = 1, X(t) - X(u) = 0)}{P(X(t) = 1)}$$

$$= \frac{P(X(u) = 1) P(X(t) - X(u) = 0)}{P(X(t) = 1)}$$

$$= \frac{(\cancel{\lambda u}) \cancel{e^{-\lambda u}} \cdot \cancel{e^{-\lambda(t-u)}}}{(\cancel{\lambda t}) \cancel{e^{-\lambda t}}}$$

$$= \frac{u}{t} //$$

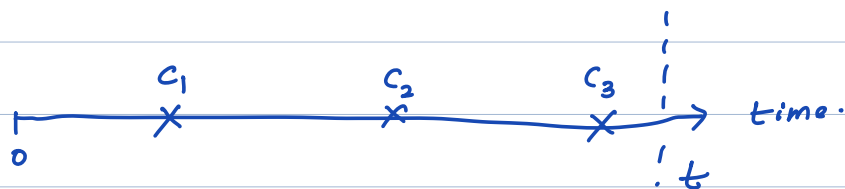
Generalization



Property*: Given $X(t) = n$, the n inter-arrival times are ^{independent} uniformly distributed in $(0, t)$
 [Order these independent & uniform distributed in $(0, t)$]

Proof: Text book.

Example: Insurance claims are made at times distributed according to a $PP(\lambda)$



Each claim value is an independent random variable with mean μ . (independent of claim arrival)
 $C_1, C_2, C_3 \sim G$ (mean μ)

Insurance company wants to figure out the total expected discounted cost upto time t (Discount rate is α)

$$D(t) = \sum_{i=1}^{X(t)} C_i e^{-\alpha S_i}$$

↑
discounted cost

$$\mathbb{E}[D(t)] = \mathbb{E}\left[\sum_{i=1}^{X(t)} C_i e^{-\alpha S_i}\right]$$

random variable random sum

Random sum = $\mathbb{E}[\mathbb{E}[D(t) | X(t) = n]]$
trick

$$= \mathbb{E}\left[\mathbb{E}\left[\sum_{i=1}^n c_i e^{-\alpha S_i}\right]\right]$$

$$= \sum_{n=0}^{\infty} \mathbb{E}\left[\sum_{i=1}^n c_i e^{-\alpha S_i} \mid X(t)=n\right] P[X(t)=n]$$

$\mathbb{E}\left[\sum_{i=1}^n c_i e^{-\alpha S_i} \mid X(t)=n\right]$ $X(t) \sim \text{Poisson}(\lambda t)$

$$= \sum_{i=1}^n \mathbb{E}[c_i e^{-\alpha S_i} \mid X(t)=n]$$

independent $= \sum_{i=1}^n \underbrace{\mathbb{E}[c_i | X(t)]}_{\mu} \mathbb{E}[e^{-\alpha S_i} \mid X(t)=n]$

$$= \mu \cdot \sum_{i=1}^n \mathbb{E}[e^{-\alpha S_i} \mid X(t)=n]$$

$$= n \cdot \mu \int_0^t e^{-\alpha \tau} \underbrace{f_{S_i}(\tau)}_{\frac{1}{t}} d\tau$$

$$= \frac{n\mu}{t} (1 - e^{-\alpha t})$$

$$= \sum_{n=0}^{\infty} \frac{n\mu}{t} (1 - e^{-\alpha t}) e^{-\lambda t} \frac{(\lambda t)^n}{n!}$$

$$E_{\text{expec}} = \frac{\mu\lambda}{\alpha} (1 - e^{-\alpha t}) //$$