

Poisson Processes 2

Sparsity may not be necessary during the derivation of MGF. $G_{N(t)}(z)$.

$$\frac{1}{\Delta t} [G_{N(t+\Delta t)}(z) - G_{N(t)}(z)] = \frac{1}{\Delta t} G_{N(t)}(z) E[z^{N(\Delta t)} - 1]$$

$$E[z^{N(\Delta t)} - 1] = \underbrace{P(N(\Delta t)=0) - 1}_{\exp(-\lambda \Delta t)} + P(N(\Delta t)=1) \cdot z + \sum_{k \geq 2} z^k \cdot P(N(\Delta t)=k)$$

$$M(t) = P(N(t)=1), [0, t] = [0, s] \cup [s, t] \Rightarrow P(N(t)=1) = \underbrace{P(N(s)=1, N(t)-N(s)=0)}_{\text{indep. increment}} + \underbrace{P(N(s)=0, N(t)-N(s)=1)}_{\text{indep. increment}}$$

$$\stackrel{\text{stationary increment}}{=} P(N(s)=1) \cdot \underbrace{P(N(t-s)=0)}_{Z(t-s)} + P(N(s)=0) \cdot \underbrace{P(N(t-s)=1)}_{Z(s)}$$

$$M(t) = M(s) \cdot Z(t-s) + Z(s) M(t-s) \quad Z(t) = P(N(t)=0) = \exp(-\lambda t)$$

$$\Rightarrow M(t+s) = M(t) \cdot Z(s) + M(s) \cdot Z(t) \Rightarrow \frac{M(t+s)}{Z(t+s)} = \frac{M(t)}{Z(t)} + \frac{M(s)}{Z(s)} \quad | \quad Z(t+s) = \exp(-\lambda(t+s)) = Z(t) Z(s)$$

$$\Rightarrow \frac{M(t+s)}{Z(t+s)} = \frac{M(t)}{Z(t)} + \frac{M(s)}{Z(s)} \quad \lambda(s) \Rightarrow \lambda(t+s) = \lambda(t) + \lambda(s) \quad \therefore \lambda(t) = \alpha t$$

$$\therefore M(t) = Z(t) \cdot \lambda(t) = \alpha t \cdot \exp(-\lambda t) \quad P(N(\Delta t)=0) + P(N(\Delta t)=1) + P(N(\Delta t) \geq 2) = 1$$

$$\frac{1 - P(N(\Delta t)=0)}{\Delta t} = \frac{P(N(\Delta t)=1)}{\Delta t} \left(1 + \frac{P(N(\Delta t) \geq 2)}{P(N(\Delta t)=1)} \right)$$

$$\lambda \exp(-\lambda \Delta t) \quad \alpha \exp(-\lambda \Delta t)$$

$$\text{let } \alpha = \lambda \Rightarrow \frac{P(N(\Delta t) \geq 2)}{P(N(\Delta t)=1)} \xrightarrow{\Delta t \rightarrow 0} 0$$

i) Non-Stationary Increment: \Rightarrow Need Another Assumption. $\lim_{\Delta t \rightarrow 0} \frac{1 - P(N(t+\Delta t)-N(t)=0)}{\Delta t} = \lambda(t)$ time-dependent.

$$\begin{cases} \frac{d}{dt} G_{N(t)}(z) = G_{N(t)}(z) \lambda(t)(z-1) \\ G_{N(0)}(z) = E[z^{N(0)}] = E[z^0] = 1 \end{cases} \Rightarrow G_{N(t)}(z) = \exp\left((z-1) \int_0^t \lambda(s) ds\right), \quad P(N(t)=k) = \frac{\int_0^t \lambda(s) ds}{k!} \exp\left(-\int_0^t \lambda(s) ds\right)$$

$$\text{Non-Homogeneous Poisson. } \lambda = \lambda(t) \quad \left| \lim_{\Delta t \rightarrow 0} \frac{1 - P(N(t+\Delta t)-N(t)=0)}{\Delta t} = \lambda(t) \right|$$

Counting Events. \Rightarrow Effect Statistics

Compound Poisson Processes

$$Y(t) = \sum_{k=1}^{N(t)} X_k$$

, $N(t)$: Standard Poisson. X_k i.i.d., indep. of $N(t)$

$$X_k \equiv 1 \Rightarrow Y(t) = N(t)$$

$$\begin{aligned} G_{Y(t)}(z) &= E[z^{Y(t)}] = E\left[z^{\sum_{k=1}^{N(t)} X_k}\right] = E\left[E\left[z^{\sum_{k=1}^{N(t)} X_k} \mid N(t)=n\right]\right] = E\left[E\left[z^{\sum_{k=1}^n X_k} \mid N(t)=n\right]\right] = E\left[E\left[z^{n X_1} \mid N(t)=n\right]\right] = E\left[G_{X_1}^{N(t)}(z)\right] = E\left[z^{N(t)}\right]_{z=G_{X_1}(z)} \\ &= G_{N(t)}(G_{X_1}(z)) \quad \left(G_{N(t)}(z) = \exp(\lambda t(z-1))\right) = \exp(\lambda t(G_{X_1}(z)-1)) \end{aligned}$$

Σx. 1: $N(t), X_k \sim B \begin{pmatrix} M & F \\ p & 1-p \end{pmatrix}$ only focuses on 'M' \Rightarrow still Standard Poisson? : $X_k \stackrel{i.i.d.}{\sim} \begin{pmatrix} 1 & 0 \\ p & 1-p \end{pmatrix}$

$$G_{Y(t)}(z) = G_{N(t)}(G_X(z)) = \exp(\lambda t (G_X(z) - 1)) \Rightarrow G_{Y(t)}(z) = \exp(\lambda t (pz + (1-p) - 1)) = \exp(\lambda t \cdot p(z-1)) = \exp(\underbrace{\lambda p \cdot t}_{\lambda' = \lambda p} \cdot (z-1)) \Rightarrow \text{Poisson}$$

$$G_X(z) = z \cdot p + 1 \cdot (1-p) = pz + 1-p$$

Randomization. (Load Balance, Ethernet (CSMA/CD))

Σx. 2: $N_1(t) : \lambda_1, N_2(t) : \lambda_2, N_1 \text{ indep. } N_2, N = N_1 + N_2$ Poisson?

$$G_{N(t)}(z) = E[z^{N(t)}] = E[z^{N_1 + N_2}] = E[z^{N_1}] \cdot E[z^{N_2}] = \exp(\lambda_1 t (z-1)) \exp(\lambda_2 t (z-1)) = \exp((\lambda_1 + \lambda_2) t (z-1))$$

$$\Rightarrow N_k(t) : \lambda_k, k=1 \sim n, Y(t) = \sum N_k(t) \Rightarrow Y(t) \text{ Poisson, } \lambda = \sum_k \lambda_k \quad | \text{Omin}$$

Σx. 3.

2 counters occupied, serve time $\stackrel{i.i.d.}{\sim} \exp(\lambda)$, comes another customer 3 ($t=0$)

$$P\{\text{customer 3 leaves latest}\} = ? \quad \frac{1}{2}$$

Distribution of 1st leaving customer's leaving time $\sim \exp(\cdot)$

$$T_1 \sim \exp(\lambda), T_2 \sim \exp(\lambda), T = \min(T_1, T_2) \sim \exp(\underbrace{\lambda_1 + \lambda_2}_{\lambda' = \lambda_1 + \lambda_2 = 2\lambda})$$

1st occurrence of a Poisson Process N_1, N_2 indep., 1st occurrence of $N_1 \vee N_2 \Leftrightarrow$ 1st occurrence of Poisson Process N , intensity = $\lambda_1 + \lambda_2$

When one is vacant, and customer 3 filled in \Rightarrow still $\exp(\cdot)$, memoryless?

Memoryless: $P(X > x+y | X > y) = P(X > x)$, if y is r.v. (Y), X still memoryless?

$$\text{Prove: if } X \sim \exp(\lambda_1), Y \sim \exp(\lambda_2), P(X > x+Y | X > Y) = P(X > x)$$

$$P(X > Y) = \int_0^\infty P(Y=y) \cdot P(X > Y | Y=y) dy = \int_0^\infty \lambda_2 \exp(-\lambda_2 y) \cdot \int_y^\infty \lambda_1 \exp(-\lambda_1 x) dx \cdot dy = \int_0^\infty \lambda_2 \exp(-\lambda_2 y) \cdot \exp(-\lambda_1 y) dy = \lambda_2 \cdot \int_0^\infty \exp(-(\lambda_1 + \lambda_2)y) dy = \frac{\lambda_2}{\lambda_1 + \lambda_2}$$

$$P(X > Y+x) = \int_0^\infty P(Y=y) \cdot P(X > Y+x | Y=y) dy = \int_0^\infty \lambda_2 \exp(-\lambda_2 y) \cdot \int_{x+y}^\infty \lambda_1 \exp(-\lambda_1 x) dx \cdot dy = \int_0^\infty \lambda_2 \exp(-\lambda_2 y) \exp(-\lambda_1(x+y)) dy = \lambda_2 \cdot \exp(-\lambda_1 x) \cdot \frac{1}{\lambda_1 + \lambda_2}$$

$$= \frac{\lambda_2}{\lambda_1 + \lambda_2} \exp(-\lambda_1 x)$$

$$P(X > x) = \int_x^\infty \lambda_1 \exp(-\lambda_1 s) ds = \exp(-\lambda_1 x) \quad \therefore P(X > x+Y | X > Y) = \frac{P(X > x+Y)}{P(X > Y)} = \exp(-\lambda_1 x) = P(X > x)$$

\Rightarrow Customer 3's leaving time $\sim \exp(\lambda)$, $P(\dots) = \frac{1}{2}$

Σx. 4. $N(t) = N_1(t) - N_2(t)$ still Poisson? \times $P(N_1(t) - N_2(t) < 0) > 0$

Compound Poisson.

Compound Poisson: $\exp(\lambda t (G_X(z) - 1))$

$$G_{N(t)}(z) = E[z^{N(t)}] = E[z^{N_1 - N_2}] = E[z^{N_1}] \cdot E[z^{-N_2}] = G_{N_1}(z) \cdot G_{N_2}(\frac{1}{z}) = \exp(\lambda_1 t (z-1)) \exp(\lambda_2 t (z^{-1}-1)) = \exp((\lambda_1 + \lambda_2) t (\frac{\lambda_1 z + \lambda_2 z^{-1}}{\lambda_1 + \lambda_2} - 1))$$

$$G_N(z) = \exp\left((\lambda_1 + \lambda_2)t \left(\frac{\lambda_1 z + \lambda_2 z^2}{\lambda_1 + \lambda_2} - 1\right)\right) \Rightarrow \text{Compound Poisson}$$

intensity

 $X \sim \begin{pmatrix} \lambda_1 & -1 \\ \lambda_1 + \lambda_2 & \lambda_1 + \lambda_2 \end{pmatrix}$

Ex. 5: 2 counters $\sim \exp(\lambda_1), \exp(\lambda_2)$ are occupied. customer 3 comes in ($t=0$)

Expected leaving time of customer 3 = ?

Waiting + Serving:

Waiting: $\sim \exp(\lambda_1 + \lambda_2)$, $E[\text{Waiting}] = \frac{1}{\lambda_1 + \lambda_2}$

Serving:

- #1 $P(T_1 < T_2) E[T_1 | T_1 < T_2] = P(T_1 < T_2) \cdot \frac{1}{\lambda_1} = \frac{\lambda_1}{\lambda_1 + \lambda_2} \cdot \frac{1}{\lambda_1} \searrow \frac{2}{\lambda_1 + \lambda_2}$
- #2 $P(T_2 < T_1) E[T_2 | T_2 < T_1] = P(T_1 > T_2) \cdot \frac{1}{\lambda_2} = \frac{\lambda_2}{\lambda_1 + \lambda_2} \cdot \frac{1}{\lambda_2} \nearrow \frac{2}{\lambda_1 + \lambda_2}$

$$\Rightarrow E[T_3] = \frac{3}{\lambda_1 + \lambda_2}$$

My Solution:

3 \rightarrow #1 $P(T_1 < T_2) \Rightarrow t_1 + E[T_1]$ $P(T_1 < T_2) = P(T_1 = t_1) \cdot P(T_2 > T_1 | T_1 = t_1) \Rightarrow P(T_1 = t_1) P(T_2 > t_1) \cdot (t_1 + \frac{1}{\lambda_1}) \int_0^\infty dt_1$

#2 $P(T_1 > T_2) \Rightarrow t_2 + E[T_2]$ $P(T_1 > T_2) \Rightarrow P(T_2 = t_2) P(T_1 > t_2) \cdot (t_2 + \frac{1}{\lambda_2}) \cdot \int_0^\infty dt_2$

$$\int_0^\infty \lambda_1 \exp(-\lambda_1 t_1) \cdot \int_{t_1}^\infty \lambda_2 \exp(-\lambda_2 s) ds (t_1 + \frac{1}{\lambda_1}) dt_1$$

$$= \int_0^\infty \lambda_1 \exp(-\lambda_1 t_1) \cdot \exp(-\lambda_2 t_1) (t_1 + \frac{1}{\lambda_1}) dt_1$$

$$= \int_0^\infty \lambda_1 \exp(-(\lambda_1 + \lambda_2) t_1) (t_1 + \frac{1}{\lambda_1}) dt_1$$

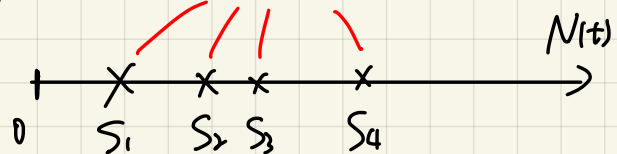
$$= \frac{1}{\lambda_1 + \lambda_2} + \lambda_1 \int_0^\infty t_1 \exp(-(\lambda_1 + \lambda_2) t_1) dt_1$$

$$= \frac{1}{\lambda_1 + \lambda_2} + \frac{\lambda_1}{(\lambda_1 + \lambda_2)^2}$$

$$\therefore \int_0^\infty \lambda_2 \exp(-\lambda_2 t_2) \int_{t_2}^\infty \lambda_1 \exp(-\lambda_1 s) ds (t_2 + \frac{1}{\lambda_2}) dt_2 = \frac{1}{\lambda_1 + \lambda_2} + \frac{\lambda_2}{(\lambda_1 + \lambda_2)^2}$$

$$E[T_3] = \frac{1}{\lambda_1 + \lambda_2} + \frac{\lambda_1}{(\lambda_1 + \lambda_2)^2} + \frac{1}{\lambda_1 + \lambda_2} + \frac{\lambda_2}{(\lambda_1 + \lambda_2)^2} = \frac{3}{\lambda_1 + \lambda_2}$$

Ex. 6: Event Occurrence



$$E[S_4 | N(t)=2] = ?$$

$$1 - P(N(t)=0) - P(N(t)=1) = 1 - \exp(-\lambda(t-1)) \left[1 + \frac{\lambda(t-1)}{1!}\right]$$

$$f_{S_4}(t | N(t)=2) = ? \Rightarrow F_{S_4}(t | N(t)=2) = P(S_4 \leq t | N(t)=2) = \frac{P(S_4 \leq t, N(t)=2)}{P(N(t)=2)} = \frac{P(N(t)-N(t-1) \geq 2, N(t)=2)}{P(N(t)=2)} = \frac{P(N(t)-N(t-1) \geq 2) P(N(t)=2)}{P(N(t)=2)}$$

indep increment.

$$\stackrel{\text{Stationary increment.}}{=} \frac{P(N(t-1) \geq 2) P(N(t)=2)}{P(N(t)=2)} = 1 - (\lambda(t-1) + 1) \exp(-\lambda(t-1))$$

$$f_{S_4}(t | N(t)=2) = \frac{dF_{S_4}(t | N(t)=2)}{dt} = -\lambda \exp(-\lambda(t-1)) + \lambda(\lambda(t-1) + 1) \exp(-\lambda(t-1)) = \lambda^2(t-1) \exp(-\lambda(t-1))$$

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

$$E[S_4 | N(t)=2] = \int_0^\infty t \cdot f_{S_4}(t | N(t)=2) dt = \int_0^\infty t \lambda^2(t-1) \exp(-\lambda(t-1)) dt = \int_{-1}^\infty (t+1) \lambda^2 t \exp(-\lambda t) dt = \int_{-1}^\infty (\lambda^2 t^2 + \lambda^2 t) \exp(-\lambda t) dt$$

$N(t)=2$. 2 occurrences afterwards. each $\sim \exp(\lambda) \Rightarrow t = 1 + 2 \cdot E[T(t)] = 1 + 2 \cdot \frac{1}{\lambda}$

$$2^{\text{nd}} - 3^{\text{rd}} : \frac{1}{\lambda}$$

$$1 - 3^{\text{rd}} < \frac{1}{\lambda}$$

However \rightarrow 3rd occurrence should be less than $E[T(t)] = \frac{1}{\lambda}$, expectation should be $1 + \text{portion of } \frac{1}{\lambda} + \frac{1}{\lambda}$ not $1 + \frac{1}{\lambda} + \frac{1}{\lambda}$