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$$\begin{aligned} 1. M_X(r) &= E[e^{rx}] = \int_0^{\infty} e^{rx} \cdot f_X(x) dx \\ &= \int_0^{\infty} e^{rx} \cdot \lambda e^{-\lambda x} dx \\ &= \frac{\lambda}{r-\lambda} e^{(r-\lambda)x} \Big|_0^{\infty} \\ &= \frac{\lambda}{\lambda-r} \end{aligned}$$

~~M\_X(r)~~  
 $M_X(r)_{\text{sum}} = E[e^{r \sum_{i=1}^n X_i}] = E[e^{r(X_1 + \dots + X_n)}] = E[e^{rX_1} e^{rX_2} e^{rX_3} \dots e^{rX_n}]$

Because  $X_1, X_2, \dots, X_n$  are iid.

$$M_X(r)_{\text{sum}} = E[e^{rX_1}] \cdot E[e^{rX_2}] \dots E[e^{rX_n}] = \left( \frac{\lambda}{\lambda-r} \right)^n$$

Invert:

$$\begin{aligned} \mathcal{L}^{-1}[M_X(r)_{\text{sum}}] &= \mathcal{L}^{-1}\left[ \left( \frac{\lambda}{\lambda-r} \right)^n \right] = \frac{\lambda^n}{(n-1)!} \cdot \mathcal{L}^{-1}\left[ \frac{(n-1)!}{(\lambda-r)^n} \right] \\ &= \frac{\lambda^n}{(n-1)!} \cdot s_n^{n-1} \cdot e^{-\lambda s_n} \end{aligned}$$

2.

(a)

Mean:

$$\bar{E}[M(t)] = \int_0^{\infty} t \cdot \lambda e^{-\lambda t} dt$$

$$= \int_0^{\infty} \frac{\lambda t^n e^{-\lambda t}}{n!} dt$$

=

$$\begin{aligned} \bar{E}[M(t)] &= \sum_{n=1}^{\infty} n \cdot P_{Me}(n) \\ &= \sum_{n=1}^{\infty} \frac{(\lambda t)^n e^{-\lambda t}}{(n-1)!} \\ &= \lambda t \cdot e^{-\lambda t} \cdot \sum_{n=1}^{\infty} \frac{(\lambda t)^{n-1}}{(n-1)!} \\ &= \lambda t \cdot e^{-\lambda t} \cdot e^{\lambda t} \\ &= \lambda t \end{aligned}$$

$$Var(M(t)) = \bar{E}[M(t)^2] - (\bar{E}[M(t)])^2$$

$$= \sum_{n=1}^{\infty} n^2 \cdot P_{Me}(n) - (\lambda t)^2$$

$$= \lambda t \cdot e^{-\lambda t} \cdot \sum_{n=1}^{\infty} \frac{n \cdot (\lambda t)^{n-1}}{(n-1)!} - (\lambda t)^2$$

$$= \lambda t \cdot e^{-\lambda t} \cdot \left[ \sum_{n=1}^{\infty} \frac{(\lambda t)^{n-1}}{(n-2)!} + \frac{(\lambda t)^{n-1}}{(n-1)!} \right] - (\lambda t)^2$$

$$= (\lambda t)^2 \cdot e^{-\lambda t} \cdot \sum_{n=2}^{\infty} \frac{(\lambda t)^{n-2}}{(n-2)!} + (\lambda t) \cdot e^{-\lambda t} \cdot \sum_{n=1}^{\infty} \frac{(\lambda t)^{n-1}}{(n-1)!} - (\lambda t)^2$$

$$= (\lambda t)^2 + \lambda t - (\lambda t)^2$$

$$= \lambda t$$

MGF:

$$E[e^{rM(t)}] = \sum_{n=0}^{\infty} e^{rn} \cdot \frac{(\lambda t)^n \cdot e^{-\lambda t}}{n!}$$

$$= e^{-\lambda t} \sum_{n=0}^{\infty} \frac{(\lambda t \cdot e^r)^n}{n!}$$

$$= e^{-\lambda t} \cdot e^{\lambda t e^r}$$

$$= e^{\lambda t(e^r - 1)}$$

2.(b)

$$\Pr(X = X_1 + X_2) = \sum_{k=0}^{\infty} \Pr(X_1 = k) \cdot \Pr(X_2 = k)$$

$$\sum_{k=0}^{\infty}$$

Let  $X_1 + X_2 = n$ ,

$$\begin{aligned} \Pr(X_1 + X_2 = n) &= \sum_{i=0}^n \Pr(X_1 = i) \Pr(X_2 = n-i) \\ &= \sum_{i=0}^n \frac{\lambda_1^i e^{-\lambda_1}}{i!} \cdot \frac{\lambda_2^{n-i} e^{-\lambda_2}}{(n-i)!} \\ &= e^{-(\lambda_1 + \lambda_2)} \cdot \sum_{i=0}^n \frac{\lambda_1^i \cdot \lambda_2^{n-i}}{i! (n-i)!} \\ &= \frac{e^{-(\lambda_1 + \lambda_2)}}{n!} \cdot \sum_{i=0}^n \binom{n}{i} \cdot \lambda_1^i \lambda_2^{n-i} \\ &= \frac{e^{-(\lambda_1 + \lambda_2)}}{n!} \cdot (\lambda_1 + \lambda_2)^n. \end{aligned}$$

$$X_1 + X_2 \sim \text{Poi}(\lambda_1 + \lambda_2)$$

3. (a)

$$P_{N(t)}(n) = \frac{(\lambda t)^n \cdot e^{-\lambda t}}{n!}$$

$$P_r\{X_1 > x\} = P_r\{N(t)=0\} = e^{-\lambda t}.$$

(b) Because  $N(t)$  has IIP and SLP.

$N(t, t+x)$  and  $N(t)$  are independent.

$$P_r\{X_n > x | S_n = t\} = P_r\{\tilde{N}(t, t+x) = n-1 | N(t) = n-1\}.$$

$$= P_r\{N(t)=0\} = e^{-\lambda x}.$$

$$(c) P_r\{X_n > x\} = P_r\{\tilde{N}(t, t+x) = n-1\} = P_r\{N(t)=0\} = e^{-\lambda x}$$

Because  $P_r\{X_n > x\} = P_r\{X_n > x | S_{n-1} = t\}$ ,  $X_n$  is independent of  $S_{n-1}$ .

(d). Let  $(X_1 = t_1, X_2 = t_2, \dots, X_{n-1} = t_{n-1})$ ,

$$P_r\{X_n > x | X_1 = t_1, X_2 = t_2, \dots, X_{n-1} = t_{n-1}\}$$

$$= P_r\{X_n > x | S_{n-1} = (t_1 + \dots + t_{n-1})\}$$

$$= P_r(X_n > x).$$

So  $X_n$  is independent of  $X_1, X_2, \dots, X_{n-1}$ .

4.

$$\begin{aligned}
 (a) \frac{dF_0(\tau)}{d\tau} &= \lim_{\delta \rightarrow 0} \frac{F_0(\tau+\delta) - F_0(\tau)}{\delta} \\
 &= \lim_{\delta \rightarrow 0} \frac{P_r\{N(\tau+\delta)=0\} - P_r\{N(\tau)=0\}}{\delta} \\
 &= \lim_{\delta \rightarrow 0} \frac{P_r\{N(\tau)=0\} \cdot P_r\{\tilde{N}(\tau, \tau+\delta)=0\} - P_r\{N(\tau)=0\}}{\delta} \\
 &= \lim_{\delta \rightarrow 0} \frac{P_r\{N(\tau)=0\} \cdot [P_r\{\tilde{N}(\tau, \tau+\delta)=0\} - 1]}{\delta} \\
 &= P_r\{N(\tau)=0\} \cdot \lim_{\delta \rightarrow 0} \frac{1 - \lambda \tau_0(\delta) - 1}{\delta} \\
 &= -\lambda \cdot F_0(\tau)
 \end{aligned}$$

$$(b) P_r\{X_1 > x\} = P_r\{N(t)=0\}$$

$$\frac{dF^c(x_1)}{dx} = -\lambda F^c(x_1) = -\lambda F_0(\tau)$$

$$\Rightarrow F^c(x_1) = e^{-\lambda x} \quad x > 0$$

$$\therefore F(x_1) = 1 - e^{-\lambda x}$$

$$\therefore f_{x_1}(x) = \begin{cases} \lambda e^{-\lambda x} & x > 0 \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned}
 (c) \cancel{F_n^c(t) = P_r\{N(t)=0\}} \quad F_n^c(t) &= P_r\{\tilde{N}(t, t+\tau)=0 \mid S_{n-1}=t\} \\
 &= P_r\{\tilde{N}(t, t+\tau)=0 \mid N(t)=n-1\}
 \end{aligned}$$

Because  $N(t)$  is IID & SLP.

$$\text{LHS} = P_r\{\tilde{N}(t, t+\tau)=0\}$$

$$= P_r\{N(\tau)=0\} \quad \text{Q.E.D.}$$

$$\text{From 4.(a), we can get } \frac{dF_n^c(\tau)}{d\tau} = -\lambda F_n^c(\tau)$$

4. (d)

From 4. (c).

$$Pr\{\tilde{N}(t, t+\tau)=0 | S_{n-1}=t\}$$

$$= Pr\{X_n > \tau | S_{n-1}=t\}$$

$$= Pr\{X_n > \tau | X_1=\tau_1, X_2=\tau_2, \dots, X_{n-1}=\tau_{n-1}\}$$

$$= Pr(X_n > \tau) = e^{-\lambda\tau} = F(X_n). \quad [4. (b) \text{ and } 4. (c)]$$

$$\Rightarrow X_n \text{ is independent of } X_1, X_2, \dots, X_{n-1}$$

5.

$W = A + B$ . Because  $A, B$  are Poisson process,  $W$  is also a Poisson Process

$$(2. (b)) \quad W \sim \text{Poi}(\lambda_A + \lambda_B)$$

$$(a) Pr\{N(t)=q\} = \frac{(\lambda_A + \lambda_B)^q}{q!} \cdot e^{-(\lambda_A + \lambda_B) \cdot t}$$

$$(b) E[N] = E[E[N | M(t)=n]]$$

For 1 message, the expected of words is:

$$E[N | M(t)=1] = \frac{2}{6} \times 1 + \frac{3}{6} \times 2 + \frac{1}{6} \times 3 = \frac{11}{6}$$

Because the transimite process is independent

$$E[N | M(t)=n] = \frac{11}{6} n.$$

$$E[N] = E\left[\frac{11}{6} n\right] = \frac{11}{6} E[M(t)=n] = \frac{11}{6} \cdot (\lambda_A + \lambda_B) \cdot t$$



5.

$$(c) \Pr\{W=3 | M(t)=8\} = \left(\frac{1}{7}\right)^8$$

$$\begin{aligned} \Pr\{W=3, M(t)=8\} &= \Pr\{W=3 | M(t)=8\} \cdot \Pr\{M(t)=8\} \\ &= \left(\frac{1}{6}\right)^8 \cdot \frac{\lambda^8 \cdot e^{-\lambda t}}{8!} \end{aligned}$$

5.  
(d)  ~~$\Pr\{A=8 | A+B=12\}$~~   $\Pr\{A=8, A+B=12\}$

~~$\Pr\{A=8 | B=4\}$~~

$$= \Pr\{A=8, B=4\}$$

$$= \Pr\{A=8\} \cdot \Pr\{B=4\}$$

(A and B are independent)

$$= \frac{\lambda^8 e^{-\lambda t}}{8!} \cdot \frac{\lambda^4 e^{-\lambda t}}{4!}$$

6.

(i) Because Poisson Process have both the stationary increment and independent increment properties.

(ii) exponential distribution.

(iii)  $\lambda \cdot X$

(iv)  ~~$P\{t^* - U \leq x\} = P\{t^* \leq U + x\}$~~   
 ~~$= P\{t^* \leq U + x\}$~~

$$P\{t^* - U \leq x\} = e^{-\lambda x}.$$

So,  $t^* - U$  is an exponential distribution

$$f_{t^* - U}(x) = \begin{cases} \lambda e^{-\lambda x}, & x > 0 \\ 0, & \text{o.w.} \end{cases}$$

(v)  $P\{L \leq l\}$

$$= P\{t^* - U + L - (t^* - U) \leq l\}.$$

$$= P\{t^* - U \leq l_1\} \cdot P\{L - (t^* - U) \leq l - l_1\}.$$

$$= e^{-\lambda l_1} \cdot e^{-\lambda(l-l_1)}$$

$$= e^{-\lambda l}$$

$$\therefore f_L(l) = \begin{cases} \lambda e^{-\lambda l}, & l > 0 \\ 0, & \text{o.w.} \end{cases}$$

(vi) It's an exponential distribution with rate  $\lambda$ .