

Prep: bring ID, water, pen, jacket, watch and this sheet

**You got this!**

Poisson Process $(N_t)_{t \geq 0} \sim PP(\lambda)$	Renewal Process
<p><u>Set-up:</u></p> <ul style="list-style-type: none"> <li><math>N(0) = 0</math>, <math>N([t, t + \Delta t]) \sim \text{Poisson}(\lambda \Delta t)</math></li> <li>Disjoint intervals are independent</li> <li><math>X_i \sim \text{Expo}(\lambda)</math> i.i.d. (interarrival time)</li> </ul> <p><u>Waiting Time <math>W_n</math> Analysis:</u></p> <ul style="list-style-type: none"> <li><math>W_n = \sum_{i=1}^n X_i</math>; <math>W_0 = 0</math> (waiting time)</li> <li><math>W_n \sim \text{Erlang}(n, \lambda)</math> i.e. <math>f_{W_n}(w) = \frac{\lambda^n w^{n-1} e^{-\lambda w}}{(n-1)!}</math></li> <li><math>N(t) = \max\{n : 0 &lt; W_n \leq t\} \sim \text{Poisson}(\lambda t)</math></li> <li><math>N(t) &lt; n \Leftrightarrow W_n &gt; t \Leftrightarrow \sum_{i=1}^n X_i &gt; t</math></li> <li><math>N(t) \geq n \Leftrightarrow W_n \leq t \Leftrightarrow \sum_{i=1}^n X_i \leq t</math></li> <li><math>\mathbb{P}[W_n \leq t] = \mathbb{P}[N(t) \geq n] = \mathbb{P}[\text{Poisson}(\lambda t) \geq n]</math></li> <li><math>N(t)</math> and <math>W_{N(t)+1}</math> are independent</li> </ul> <p><u>Conditioning on Interval <math>N(t) = n</math>:</u></p> <ul style="list-style-type: none"> <li><math>\mathbb{P}[N(s) = k   N(t) = n] \sim \text{Binomial}\left(n, \frac{s}{t}\right)</math> for <math>s \leq t, k \leq n</math></li> <li>Let <math>W_i   N(T) = n</math> denote the arrival of <math>i</math>th customer</li> <li><math>f_{W_1, \dots, W_n   N(T)=n}(t_1, \dots, t_n) = \frac{n!}{T^n}</math> if <math>0 \leq t_1 \leq \dots \leq t_n \leq T</math></li> <li>[Big Theorem] Conditional on <math>N(T) = n</math>, the <math>n</math> arrivals are i.i.d. uniform in <math>[0, T]</math>.</li> <li><math>W_i \sim V_i = i</math>th order statistics of <math>U_1, \dots, U_n</math></li> <li>If <math>g</math> symmetric, <math>g(V_1, \dots, V_n) = g(U_1, \dots, U_n)</math></li> <li>Symmetric functions <math>\mathbb{P}[g(W_1, \dots, W_n) = k   N(t) = n] = \mathbb{P}[g(U_1, \dots, U_n) = k]</math></li> </ul> <p><u>Current Life and Residual Life Analysis:</u></p> <ul style="list-style-type: none"> <li><math>\delta_t</math>: current life, <math>\gamma_t</math>: excess life</li> <li><math>\gamma_t</math> independent of <math>\delta_t</math> (memoryless)</li> <li><math>\delta_t \sim \min(\text{Expo}(\lambda), t)</math>, <math>\gamma_t \sim \text{Expo}(\lambda)</math></li> <li><math>\mathbb{P}[\delta_t \leq x] = \begin{cases} 1 - e^{-\lambda x}, &amp; 0 \leq x &lt; t \\ 1, &amp; t \leq x \end{cases}</math></li> <li><math>\mathbb{P}[\gamma_t \leq x] = 1 - e^{-\lambda x}</math></li> <li><math>\mathbb{P}[\gamma_t &gt; x, \delta_t &gt; y] = \mathbb{P}[\gamma_t &gt; x] \mathbb{P}[\delta_t &gt; y]</math></li> <li><math>\mathbb{E}[\delta_t + \gamma_t] = \frac{2}{\lambda} - \frac{1}{\lambda} e^{-\lambda t}</math> (size-biased)</li> <li><math>M(t) = \mathbb{E}[N(t)] = \lambda t</math></li> </ul> <p><u>Differential Analysis:</u></p> <ul style="list-style-type: none"> <li><math>\mathbb{P}[N(t, t + dt) = 1] = \lambda dt</math></li> <li><math>\mathbb{P}[N(t, t + dt) &gt; 1] = 0</math></li> <li><math>\mathbb{P}[N(t, t + dt) = 0] = 1 - \lambda dt</math></li> </ul>	<p><u>Set-up:</u></p> <ul style="list-style-type: none"> <li><math>N(0) = 0</math>, <math>N(t) = \max\{n   W_n \leq t\}</math> is the number of replacements by time <math>t</math></li> </ul> <p><u>Current Life and Residual Life Analysis:</u></p> <ul style="list-style-type: none"> <li><math>\delta_t = t - W_{N(t)}</math>: current life</li> <li><math>\gamma_t = W_{N(t)+1} - t</math>: excess life</li> <li><math>\delta_t + \gamma_t = W_{N(t)+1} - W_{N(t)}</math>: total life</li> <li><math>\mathbb{E}[\gamma_t] = \mathbb{E}[X_i] \mathbb{E}[N(t) + 1] - t</math></li> <li><math>\mathbb{P}[\gamma_t &gt; x] = \mathbb{P}[\text{no renewal in } (t, t + x)]</math></li> <li><math>\mathbb{P}[\delta_t &gt; x] = \mathbb{P}[\text{no renewal in } (t - x, t)]</math></li> <li><math>\lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t \mathbb{1}\{\delta_s &gt; x, \gamma_s &gt; y\} ds = \frac{\int_{x+y}^{\infty} \mathbb{P}[X_i &gt; z] dz}{\mathbb{E}[X_i]}</math> i.e. proportion of time up until <math>t</math> where <math>\delta_s &gt; x, \gamma_s &gt; y</math> <ul style="list-style-type: none"> <li><math>r_i = \max(0, X_i - (x + y))</math></li> <li><math>t_i = X_i</math></li> </ul> </li> <li><math>\lim_{t \rightarrow \infty} \mathbb{P}[\delta_t &gt; x, \gamma_t &gt; y] = \frac{\int_{x+y}^{\infty} \mathbb{P}[X_i &gt; z] dz}{\mathbb{E}[X_i]}</math></li> <li><math>f_\delta(x) = \frac{\mathbb{P}[X_i &gt; x]}{\mathbb{E}[X_i]}</math>, <math>f_\gamma(x) = \frac{\mathbb{P}[X_i &gt; x]}{\mathbb{E}[X_i]}</math> (same)</li> <li><math>f_{\gamma, \delta}(x, y) = \frac{\mathbb{P}[X_i &gt; x+y]}{\mathbb{E}[X_i]^2}</math></li> <li>Define <math>L(s) = \delta_s + \gamma_s</math></li> <li><math>\lim_{s \rightarrow \infty} \mathbb{E}[L(s)] = 2 \lim_{s \rightarrow \infty} \mathbb{E}[\gamma] = \frac{\mathbb{E}[X_i^2]}{\mathbb{E}[X_i]} \geq \mathbb{E}[X_i]</math> (size-biased sampling)</li> <li><math>\lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t \mathbb{E}[L(s)] ds = \frac{\mathbb{E}[X_i^2]}{\mathbb{E}[X_i]}</math></li> <li><math>\lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t L(s) ds = \frac{\mathbb{E}[X_i^2]}{\mathbb{E}[X_i]}</math> <ul style="list-style-type: none"> <li><math>t_i = X_i</math></li> <li><math>r_i = X_i^2</math></li> </ul> </li> </ul> <p><u>Expected Number of Replacements by Time <math>t</math>:</u></p> <ul style="list-style-type: none"> <li><math>M(t) = \mathbb{E}[N(t)]</math></li> <li><math>M(t) = \sum_{k=1}^{\infty} \mathbb{P}[N(t) \geq k] = \sum_{k=1}^{\infty} \mathbb{P}[W_k \leq t] = \sum_{k=1}^{\infty} F_{W_k}(t)</math> where <math>F_{W_k}</math> is the <math>k</math>-fold convolution of <math>X_i</math></li> <li><math>\mathbb{E}[W_{N(t)+1}] = \mathbb{E}[X_i] \mathbb{E}[N(t) + 1]</math></li> <li><math>M(t) = F(t) + \int_0^t M(t-x) dF(x)</math></li> </ul> <p><u>Renewal Theorem:</u></p> <ul style="list-style-type: none"> <li><math>\lim_{t \rightarrow \infty} \frac{t}{N(t)} = \lim_{n \rightarrow \infty} \frac{X_1 + \dots + X_n}{n} = \mu</math> a.s.</li> <li>Consider i.i.d. pairs <math>(r_i, t_i)_{i=1}^{\infty}</math>; <math>r_i, t_i</math> could be dependent</li> </ul>

**Poisson Merging and Splitting:**

- $(N_1(t))_{t \geq 0} \sim PP(\lambda)$  and  $(N_2(t))_{t \geq 0} \sim PP(\mu)$ , then  $(N_1(t) + N_2(t))_{t \geq 0} \sim PP(\lambda + \mu)$
- [Splitting]  $(Y_i)_{i=1}^\infty$  discrete, i.i.d independent of  $(N(t))_{t \geq 0}$  determines a type
- $N_j(t) = \sum_{i=1}^{N(t)} \mathbb{1}\{Y_i = j\}$ : arrival process of the  $j$ th type. Then  $(N_j(t))_{j=1}^k \sim PP(\lambda \mathbb{P}[Y = j])$  and are independent of each other (**NOT** the parent stream  $N(t)$ )

- $t_i$ : length of  $i$ th cycle (can be period, time till success)
- $r_i$  some reward associated with  $i$ th cycle (can be cost, conditionals, counter)
- $R(t)$  is the reward collected by time  $t$
- $\sum_{i=1}^{N(t)} r_i \leq R(t) \leq \sum_{i=1}^{N(t)+1} r_i$
- $\lim_{t \rightarrow \infty} \frac{R(t)}{t} = \frac{\mathbb{E}[r_i]}{\mathbb{E}[t_i]}$  a.s.
- [Two-state system] For  $(s_i, t_i)_{i=1}^\infty$ , time spent in first system is  $\frac{\mathbb{E}[s_i]}{\mathbb{E}[s_i] + \mathbb{E}[t_i]}$

**Inhomogeneous Poisson Process  $PP(\lambda(t))$** 

- Rate changes with time
- $N_s - N_t \sim \text{Poisson}(\int_t^s \lambda(u) du)$
- $\lambda(u) \equiv \lambda_0$  reduces to homogeneous
- $\Lambda(t) = \int_0^t \lambda(u) du$  (rate accumulated)
- $(Y_s)_{s \geq 0}$  such that  $Y_s = N_{\Lambda^{-1}(s)}$
- $(Y_s)_{s \geq 0} \sim PP(1)$
- New process lags and extends the time to make sure it is homogeneous

**Queueing Theory**

- [GI/G/1] Interarrival  $t_i \sim \text{Expo}(\lambda)$ , service time  $s_i \sim \text{Expo}(\mu)$ . If  $\lambda < \mu$ , then queue clears with probability 1 and long run average proportion of time spent working =  $\frac{\lambda}{\mu} < 1$  a.s.
  - $\lambda < \mu$ : positive recurrent MC
  - $\lambda = \mu$ : null recurrent MC
  - $\lambda > \mu$ : transient MC ( $\equiv$  branching process with replacement  $> 1$ )
- [M/G/1] Only assumption is  $t_i \sim \text{Expo}(\lambda)$
- Customers arriving during the  $n$ th service time
  - $\mathbb{P}[k \text{ arrivals} | S_n = s] = \frac{(\lambda s)^k e^{-\lambda s}}{k!}$
  - $\mathbb{P}[k \text{ arrivals}] = \mathbb{E} \left[ \frac{(\lambda S_n)^k e^{-\lambda S_n}}{k!} \right]$
  - $X_{n+1} = \max(0, X_n - 1 + S_n)$
  - $(X_n)_{n=1}^\infty$  is a Markov chain

$X_n \backslash X_{n+1}$	0	1	2	3
0	$p_0 + p_1$	$p_2$	$p_3$	$p_4$
1	$p_0$	$p_1$	$p_2$	$p_3$
2	0	$p_0$	$p_1$	$p_2$
3	0	0	$p_0$	$p_1$

**Final Checks**

- Consider edge cases i.e. length 0 intervals, edge effects for  $\delta_t$ ; justify symmetry
- Read problem and understand the model properly; check with intuition

**Probabilistic Toolkit**

- [Order statistics]
 
$$f_{V_r}(x) = \frac{n!}{(r-1)!(n-r)!} f_X(x) (F(x))^{r-1} (1-F(x))^{n-r}$$
- [Erlang( $k, \lambda$ )]
 
$$f(x) = \frac{\lambda^k x^{k-1} e^{-\lambda x}}{(k-1)!}; \mathbb{E}[X] = \frac{k}{\lambda}; \text{Var}[X] = \frac{k}{\lambda^2}$$
- [Binomial Approximation to Poisson]
 
$$\lim_{n \rightarrow \infty} \mathbb{P} \left[ \text{Binomial} \left( n, \frac{\lambda}{n} \right) = k \right] = \frac{\lambda^k e^{-\lambda}}{k!}$$
- [Tail Sum]  $\mathbb{E}[X] = \int_0^\infty \mathbb{P}[X > z] dz$
- [Tail Sum]  $\mathbb{E}[X^n] = \int_0^\infty n z^{n-1} \mathbb{P}[X > z] dz$
- [Tail Sum]  $\mathbb{E}[X] = \sum_{z=0}^\infty \mathbb{P}[X > z]$
- [Le Cam] Let  $\epsilon_i \sim \text{Bernoulli}(p_i)$  independent,  $p_i$  not necessarily equal.  $S_n = \sum_{i=1}^n \epsilon_i$  and  $\mu = p_1 + \dots + p_n$ . Then  $\left| \mathbb{P}[S_n = k] - \frac{\mu^k e^{-\mu}}{k!} \right| \leq \sum_{i=1}^n p_i^2$  (Used to prove that Poisson Process can be constructed by  $n$  small intervals of length  $\frac{1}{n}$ )

**Last Resort**

- Break into disjoint intervals
- Convert condition  $N(t) < k$  and  $W_k > t$
- Convert to indicators on each individual arrival
 
$$\mathbb{P}[M(t) = m | N(t) = n] = \mathbb{P}[\sum_{k=1}^n \mathbb{1}\{W_k + Y_k \geq m\} | N(t) = n].$$
- $\mathbb{E}[\sum_{i=1}^{N(t)+1} X_i] = \mathbb{E}[\sum_{i=1}^\infty X_i \mathbb{1}\{i \leq N(t) + 1\}] = \mathbb{E}[\sum_{i=1}^\infty X_i \mathbb{1}\{W_{i-1} \leq t\}]$
- Total probability on  $N(t) = k$  then condition
 
$$\mathbb{P}[\delta_t > x, \gamma_t > y] = \sum_{k=0}^\infty \mathbb{P}[N(t) = k, \delta_t > x, \gamma_t > y]$$
- Nest conditions  $t_i = \mathbb{1}\{U_i < S_i\}(U_i + V_i \mathbb{1}\{U_i > 1\}) + \mathbb{1}\{U_i > S_i\} S_i$
- Use graphical method for order statistics
- Refine your reward and period. Shift the burden through conditioning and indicators.
- Bash renewal with limit theorems