

# Classics

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 x^{n-1} e^{-\lambda x}}{(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>Poisson Merging and Splitting:</u></p>	<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> <li><math>t_i</math>: length of <math>i</math>th cycle (can be period, time till success)</li> </ul>

<ul style="list-style-type: none"><li>• <math>(N_1(t))_{t \geq 0} \sim PP(\lambda)</math> and <math>(N_2(t))_{t \geq 0} \sim PP(\mu)</math>, then <math>(N_1(t) + N_2(t))_{t \geq 0} \sim PP(\lambda + \mu)</math></li><li>• [Splitting] <math>(Y_i)_{i=1}^\infty</math> discrete, i.i.d independent of <math>(N(t))_{t \geq 0}</math> determines a type</li><li>• <math>N_j(t) = \sum_{i=1}^{N(t)} \mathbb{1}\{Y_i = j\}</math>: arrival process of the <math>j</math>th type. Then <math>(N_j(t))_{j=1}^k \sim PP(\lambda \mathbb{P}[Y = j])</math> and are independent of each other (<b>NOT</b> the parent stream <math>N(t)</math>)</li></ul>	<ul style="list-style-type: none"><li>• <math>r_i</math> some reward associated with <math>i</math>th cycle (can be cost, conditionals, counter)</li><li>• <math>R(t)</math> is the reward collected by time <math>t</math></li><li>• <math>\sum_{i=1}^{N(t)} r_i \leq R(t) \leq \sum_{i=1}^{N(t+1)} r_i</math></li><li>• <math>\lim_{t \rightarrow \infty} \frac{R(t)}{t} = \frac{\mathbb{E}[r_i]}{\mathbb{E}[t_i]}</math> a.s.</li><li>• [Two-state system] For <math>(s_i, t_i)_{i=1}^\infty</math>, time spent in first system is <math>\frac{\mathbb{E}[s_i]}{\mathbb{E}[s_i] + \mathbb{E}[t_i]}</math></li></ul>																									
Inhomogeneous Poisson Process $PP(\lambda(t))$	Probabilistic Toolkit																									
<ul style="list-style-type: none"><li>• Rate changes with time</li><li>• <math>N_s - N_t \sim \text{Poisson}(\int_t^s \lambda(u) du)</math></li><li>• <math>\lambda(u) \equiv \lambda_0</math> reduces to homogeneous</li><li>• <math>\Lambda(t) = \int_0^t \lambda(u) du</math> (rate accumulated)</li><li>• <math>(Y_s)_{s \geq 0}</math> such that <math>Y_s = N_{\Lambda^{-1}(s)}</math></li><li>• <math>(Y_s)_{s \geq 0} \sim PP(1)</math></li><li>• New process lags and extends the time to make sure it is homogeneous</li></ul>	<ul style="list-style-type: none"><li>• [Order statistics]<math display="block">f_{V_r}(x) = \frac{n!}{(r-1)!(n-r)!} f_X(x) (F(x))^{r-1} (1-F(x))^{n-r}</math></li><li>• [Erlang(<math>k, \lambda</math>)]<math display="block">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}</math></li><li>• [Binomial Approximation to Poisson]<math display="block">\lim_{n \rightarrow \infty} \mathbb{P}\left[\text{Binomial}\left(n, \frac{\lambda}{n}\right) = k\right] = \frac{\lambda^k e^{-\lambda}}{k!}</math></li><li>• [Tail Sum] <math>\mathbb{E}[X] = \int_0^\infty \mathbb{P}[X &gt; z] dz</math></li><li>• [Tail Sum] <math>\mathbb{E}[X^n] = \int_0^\infty n z^{n-1} \mathbb{P}[X &gt; z] dz</math></li><li>• [Tail Sum] <math>\mathbb{E}[X] = \sum_{z=0}^\infty \mathbb{P}[X &gt; z]</math></li><li>• [Le Cam] Let <math>\epsilon_i \sim \text{Bernoulli}(p_i)</math> independent, <math>p_i</math> not necessarily equal. <math>S_n = \sum_{i=1}^n \epsilon_i</math> and <math>\mu = p_1 + \dots + p_n</math>. Then <math>\left  \mathbb{P}[S_n = k] - \frac{\mu^k e^{-\mu}}{k!} \right  \leq \sum_{i=1}^n p_i^2</math> (Used to prove that Poisson Process can be constructed by <math>n</math> small intervals of length <math>\frac{1}{n}</math>)</li></ul>																									
Queueing Theory	Problem Solving																									
<ul style="list-style-type: none"><li>• [GI/G/1] Interarrival <math>t_i \sim \text{Expo}(\lambda)</math>, service time <math>s_i \sim \text{Expo}(\mu)</math>. If <math>\lambda &lt; \mu</math>, then queue clears with probability 1 and long run average proportion of time spent working = <math>\frac{\lambda}{\mu} &lt; 1</math> a.s.<ul style="list-style-type: none"><li>◦ <math>\lambda &lt; \mu</math>: positive recurrent MC</li><li>◦ <math>\lambda = \mu</math>: null recurrent MC</li><li>◦ <math>\lambda &gt; \mu</math>: transient MC (<math>\equiv</math> branching process with replacement <math>&gt; 1</math>)</li></ul></li><li>• [M/G/1] Only assumption is <math>t_i \sim \text{Expo}(\lambda)</math></li><li>• Customers arriving during the <math>n</math>th service time<ul style="list-style-type: none"><li>◦ <math>\mathbb{P}[k \text{ arrivals}   S_n = s] = \frac{(\lambda s)^k e^{-\lambda s}}{k!}</math></li><li>◦ <math>\mathbb{P}[k \text{ arrivals}] = \mathbb{E}\left[\frac{(\lambda S_n)^k e^{-\lambda S_n}}{k!}\right]</math></li><li>◦ <math>X_{n+1} = \max(0, X_n - 1 + S_n)</math></li><li>◦ <math>(X_n)_{n=1}^\infty</math> is a Markov chain</li></ul></li></ul>	<ul style="list-style-type: none"><li>• Break into <u>disjoint</u> intervals</li><li>• Convert condition <math>N(t) &lt; k</math> and <math>W_k &gt; t</math></li><li>• Convert to indicators on each individual arrival<math display="block">\mathbb{P}[M(t) = m   N(t) = n] = \mathbb{P}\left[\sum_{k=1}^n \mathbb{1}\{W_k + Y_k \geq m\}   N(t) = n\right].</math></li><li>• <math>\mathbb{E}\left[\sum_{i=1}^{N(t)+1} X_i\right] = \mathbb{E}\left[\sum_{i=1}^\infty X_i \mathbb{1}\{i \leq N(t) + 1\}\right] = \mathbb{E}\left[\sum_{i=1}^\infty X_i \mathbb{1}\{W_{i-1} \leq t\}\right]</math></li><li>• Total probability on <math>N(t) = k</math> then condition<math display="block">\mathbb{P}[\delta_t &gt; x, \gamma_t &gt; y] = \sum_{k=0}^\infty \mathbb{P}[N(t) = k, \delta_t &gt; x, \gamma_t &gt; y]</math></li><li>• Nest conditions <math>t_i = \mathbb{1}\{U_i &lt; S_i\}(U_i + V_i \mathbb{1}\{U_i &gt; 1\}) + \mathbb{1}\{U_i &gt; S_i\} S_i</math></li><li>• Use graphical method for order statistics</li><li>• Refine your reward and period. Shift the burden through conditioning and indicators.</li><li>• Bash renewal with limit theorems</li></ul>																									
<table><tr><th><math>X_n \backslash X_{n+1}</math></th><th>0</th><th>1</th><th>2</th><th>3</th></tr><tr><th>0</th><td><math>p_0 + p_1</math></td><td><math>p_2</math></td><td><math>p_3</math></td><td><math>p_4</math></td></tr><tr><th>1</th><td><math>p_0</math></td><td><math>p_1</math></td><td><math>p_2</math></td><td><math>p_3</math></td></tr><tr><th>2</th><td>0</td><td><math>p_0</math></td><td><math>p_1</math></td><td><math>p_2</math></td></tr><tr><th>3</th><td>0</td><td>0</td><td><math>p_0</math></td><td><math>p_1</math></td></tr></table>	$X_n \backslash X_{n+1}$	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	$p_0$	$p_1$	
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3	0	0	$p_0$	$p_1$																						
Checks																										
<ul style="list-style-type: none"><li>• Consider edge cases i.e. length 0 intervals, edge effects for <math>\delta_t</math>; justify symmetry</li></ul>																										