

## Lecture 6: Reading 2.3.2, 2.5.

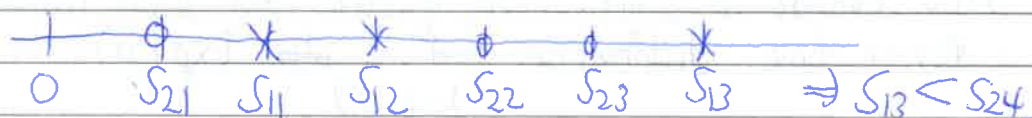
Eg: Let  $S_{1k}$  be the epoch of the  $k^{\text{th}}$  arrival of Poisson process 1.  
Let  $S_{2j}$  be the epoch of the  $j^{\text{th}}$  arrival of Poisson process 2.

PP1 has rate  $\lambda_1$ ; PP2 has rate  $\lambda_2$ .

Qn:  $P(S_{1k} < S_{2j})$ : Prob. that the  $k^{\text{th}}$  epoch of the 1<sup>st</sup> process comes before the  $j^{\text{th}}$  epoch of the 2<sup>nd</sup> process.

Consider a combined/merged process with rate  $\lambda_1 + \lambda_2$ .

Say  $k=3$  (arrivals indicated by x) &  $j=4$  (arrivals indicated by o).



What do we need for  $S_{1k} < S_{2j}$ ? (or  $S_{13} < S_{24}$ ?)

We need, out of the first  $k+j-1$  arrivals of the combined process  $k$  or more of them to be switched to the first process.

Define  $p = \frac{\lambda_1}{\lambda_1 + \lambda_2}$ .

$P(\text{exactly } k \text{ are switched to the first process}) = \binom{k+j-1}{k} p^k (1-p)^{j-1}$ .

$P(\text{exactly } i \text{ are switched to the first process}) = \binom{k+j-1}{i} p^i (1-p)^{k+j-1-i}$ .

$\Rightarrow P(S_{1k} < S_{2j}) = P(\text{more than or equal to } k \text{ are switched to the first process})$

$$= \sum_{i=k}^{k+j-1} \binom{k+j-1}{i} p^i (1-p)^{k+j-1-i}$$

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Ex: M/M/1 Queue

First character (M): M stands for memoryless & mean Poisson arrival process.

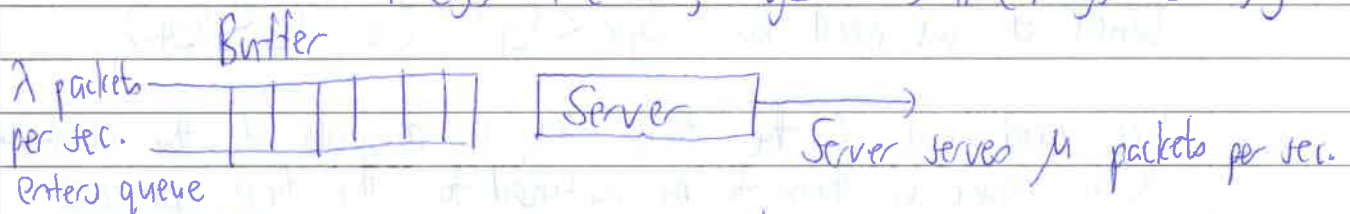
Second character (M): Service process is also memoryless.

Third character (1): Number of servers.

Consider a queuing system with arrival of customers following PP( $\lambda$ ), a Poisson process with rate  $\lambda$ .

Also consider a single server queue that serves customers with a service time distribution that is ~~also~~ Exp( $\mu$ ).

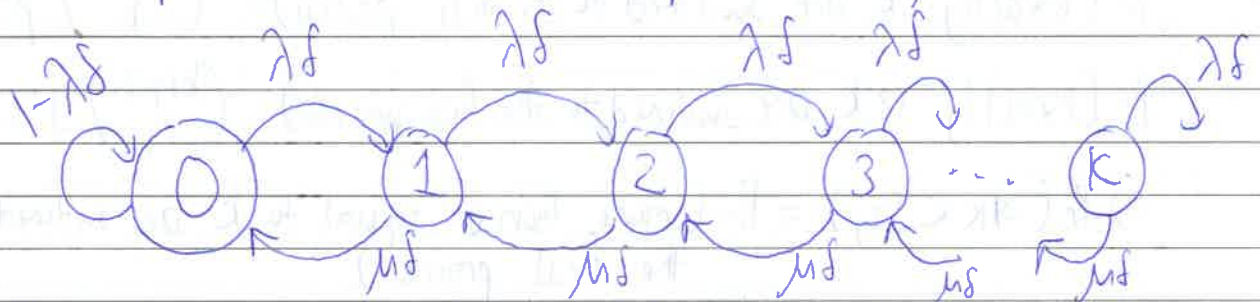
$$F(y) = \frac{1 - e^{-\mu y}}{\mu}, \quad y \geq 0. \Rightarrow P_r(Y > y) = e^{-\mu y}, y \geq 0.$$



Qn:  $P_r(S_{1k} < S_{2j})$  = prob. that  $k$ th packet arrives before  $j$ th departure

Arrival epochs of a PP( $\lambda$ )

Arrival epochs of a PP( $\mu$ )



State  $k$ :  $k$  customers (or packets) in the system.

$P(i, j)$  = Prob. transition from state  $i$  to state  $j$  ( $i, j \in \mathbb{N} \cup \{0\}$ )

Also, let  $\delta$  be an infinitesimally small time interval.  $\delta \rightarrow 0^+$

$$P(0, 0) = 1 - \lambda\delta, \quad P(j, j+1) = \lambda\delta, \quad P(j, j-1) = \mu\delta, \quad P(j, j) = 1 - (\lambda + \mu)\delta.$$

This is an example of a "birth-death" process that we'll study in detail when we talk about Markov chains.

$\lambda_j$ : flow rate from  $j$  to  $j+1$ .

$\mu_j$ : flow rate from  $j$  to  $j-1$ .

$$\rho = \frac{\lambda}{\mu}$$

Let  $P(k)$  be the probability of being in state  $k$ .

At "equilibrium",  $\lambda P(k) = \mu P(k+1) \Rightarrow P(k) = \left(\frac{\lambda}{\mu}\right)^k P(0)$ .

Since  $\sum_{k=0}^{\infty} P(k) = 1 \Rightarrow \sum_{k=0}^{\infty} P(0) \rho^k = 1 \Rightarrow P(0) \frac{1}{1-\rho} = 1 \Rightarrow P(0) = 1-\rho$ .

$\Rightarrow P(k) = \rho^k (1-\rho)$ . (requires that  $\rho < 1$ , i.e., stability).

Average queue size  $E[X] = \sum_{k=0}^{\infty} k P(k) = \sum_{k=0}^{\infty} k \rho^k (1-\rho) = \frac{\rho}{1-\rho} = \frac{\lambda}{\mu - \lambda}$

Note that if  $\lambda < \mu$ , the arrival <sup>rate</sup> of packets to the buffer is smaller than the service rate.

Intuitively, all the packets will be served before they "accumulate" in the buffer, causing an overflow.

Example: # of Poisson Arrivals during an Exponentially distributed Int.

Consider a  $PP(\lambda)$  & an independent rv  $T \sim \text{Exp}(\nu)$ . We would like to find the pmf of the # of Poisson arrivals in  $[0, T]$ , i.e., the pmf of the rv  $N(T)$ , where  $\{N(t); t \geq 0\}$  is the Poisson counting process with rate  $\lambda > 0$ .

Answer: View  $T$  as the first arrival of an indep. Poisson process with rate  $\nu$  and merge this process with the original one.

Each arrival in this merged process comes from the original one with prob.  $\lambda/(\lambda+\nu)$  independent of all other arrivals.



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$K = \#$  of arrivals until the first success (where success = arrival of the form of ~~a~~ the new process)

$$P_K(k) = \left(\frac{\nu}{\lambda + \nu}\right) \left(\frac{\lambda}{\lambda + \nu}\right)^{k-1}, \quad k=1, 2, 3, \dots$$

$L = \#$  of arrivals of the original Poisson process until the first success.

$$L = N(T) = K-1.$$

$$\Rightarrow P_{N(T)}(l) = P_L(l) = P_K(l+1) = \left(\frac{\nu}{\lambda + \nu}\right) \left(\frac{\lambda}{\lambda + \nu}\right)^l, \quad l=0, 1, 2, \dots$$

## Conditional Arrival Densities & Order Statistics.

Question: Conditioned on the event  $\{N(t)=n\}$ , there are  $n$  arrivals in  $(0, t]$ . What is the distribution of the  $n$  arrival epochs  $S^n = S^{(n)} = (S_1, S_2, \dots, S_n)$ ?

Thm. Let  $f_{S^n|N(t)}(s^n|n)$  be the joint density of  $S^n$  given  $N(t)=n$ . The density is constant over the region  $0 < s_1 < s_2 < \dots < s_n < t$  and has value

$$f_{S^n|N(t)}(s^n|n) = n!/t^n$$

Pf: Recall the joint density of  $S^{n+1} = (S_1, S_2, \dots, S_n, S_{n+1})$

$$f_{S_1, \dots, S_{n+1}}(s_1, s_2, \dots, s_{n+1}) = \lambda^{n+1} \exp(-\lambda s_{n+1}), \quad 0 \leq s_1 \leq s_2 \leq \dots \leq s_n \leq s_{n+1}.$$

By Bayes rule,

$$f_{S^{n+1}|N(t)}(s^{n+1}|n) P_{N(t)}(n) = P_{N(t)}(n|s^{n+1}) f_{S^{n+1}}(s^{n+1})$$

Note that  $N(t)=n$  iff  $t \leq s_n$  &  $s_{n+1} > t$ .

Thus  $P_{N(t)}(n|s^{n+1}) = 1$  if  $s_n \leq t$  &  $s_{n+1} > t$  and 0 otherwise.

Restricting our attention to the case  $N(t)=n$ ,  $s_n \leq t$  &  $s_{n+1} > t$ , we have

$$\begin{aligned} f_{S^{n+1}|N(t)}(s^{n+1}|n) &= \frac{f_{S^{n+1}}(s^{n+1})}{P_{N(t)}(n)} = \frac{\lambda^{n+1} \exp(-\lambda s_{n+1})}{(\lambda t)^n \exp(-\lambda t)/n!} \\ &= \frac{n! \lambda \exp(\lambda(s_{n+1} - t))}{t^n}. \end{aligned}$$

$$\text{Now } f_{S^{n+1}|N(t)}(s^{n+1}|n) = f_{S^n|N(t)}(s^n|n) \underbrace{f_{S_{n+1}|S^n, N(t)}(s_{n+1}|s^n, n)}_{\lambda \exp(-\lambda(s_{n+1} - t))}$$

We thus obtain  $f_{S^n|N(t)}(s^n|n) = n!/t^n$ ,  $0 \leq s_1 \leq s_2 \leq \dots \leq s_n$ .

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How about the joint density of the interarrival times  $X^n = (X_1, \dots, X_n)$  given  $N(t) = n$ ?

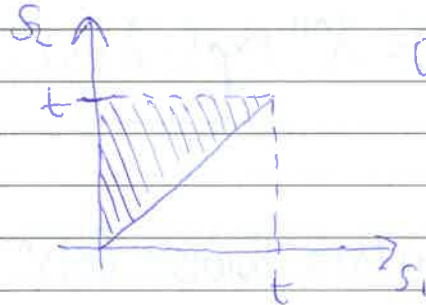
Recall that there is a one-to-one transformation between  $S^n$  &  $X^n$

$$S_1 = X_1, \quad X_i = S_i - S_{i-1}, \quad i \in \{2, \dots, n\}.$$

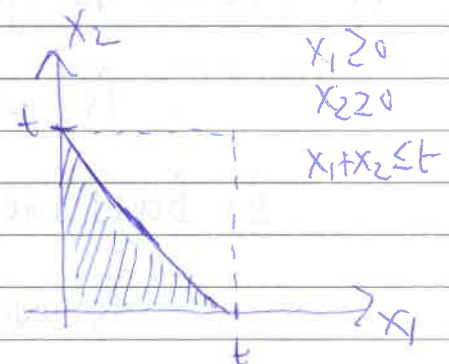
Hence, the value of the density is unchanged but the region in which it is non-zero is changed

$$f_{X^n|N(t)}(x^n|n) = n!/t^n \quad \text{for } x^n > 0 \text{ \& } \sum_{i=1}^n x_i \leq t.$$

For example, for  $n=2$ .



$$0 \leq s_1 \leq s_2 \leq t$$



Connection to Order Statistics.

Let  $U^n = (U_1, \dots, U_n)$  be iid rvs in  $(0, t] = U_i \sim \text{Unif}(0, t)$ .

For any point  $u^n \in (0, t]^n$ ,  $f_{U^n}(u^n) = \frac{1}{t^n}$ ,  $0 < u_i \leq t$ ,  $1 \leq i \leq n$ .

Fact: Both  $f_{S^n|N(t)}(s^n|n)$  and  $f_{U^n}(u^n)$  [as well as  $f_{X^n|N(t)}(x^n|n)$  and  $f_{U^n}(u^n)$ ] are uniform over the space they are defined on and are non-zero but the value of the latter is  $n!$  smaller than that of the former.

$$\text{Indeed } f_{S^n|N(t)}(s^n|n) = \frac{n!}{t^n}, \quad 0 \leq s_1 \leq s_2 \leq \dots \leq s_{n-1} \leq s_n$$

$$f_{U^n}(u^n) = \frac{1}{t^n}, \quad 0 < u_i \leq t, \quad 1 \leq i \leq n.$$



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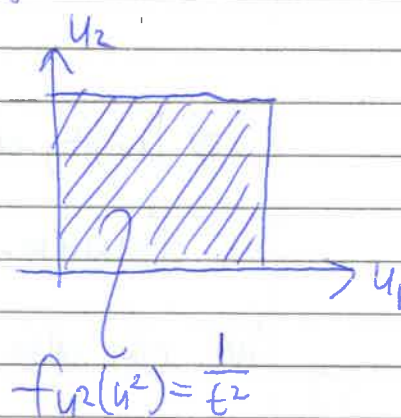
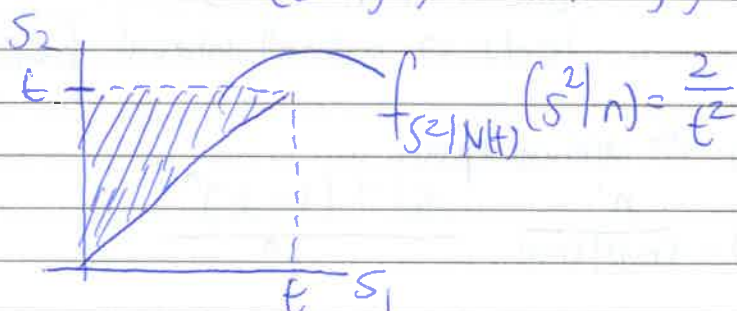
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To see this more explicitly, let  $S^n = (S_1, S_2, \dots, S_n)$  be defined, not as the arrivals ~~of~~ epochs of a Poisson process, but as the order statistics of the  $U^n$  iid random process, i.e.,

$$S_1 = \min\{U_1, \dots, U_n\}, \quad S_2 = \min\{U_1, \dots, U_n\} \setminus S_1$$

$\Rightarrow S_2$  is the 2nd min of the  $U^n$  process

$$S_k = \min\{U_1, \dots, U_n\} \setminus \{S_1, \dots, S_{k-1}\}$$



Note that the  $n$ -cube (cube of side length  $t$  in dim  $n$ ) is partitioned into  $n!$  regions, one in which  $0 < u_1 < u_2 < \dots < u_n$ .

For each permutation  $\pi: \{1, \dots, n\} \rightarrow \{1, \dots, n\}$ , there is a region  $0 < u_{\pi(1)} < u_{\pi(2)} < \dots < u_{\pi(n-1)} < u_{\pi(n)}$ .

By symmetry, each of these regions has the same volume, which ~~means that~~ then must be  $1/n!$  of the volume of the cube which is  $t^n$ .

Hence, what is the density  $f_{S_1|N(t)}(s_1|n)$ , the prob. density of the  $i$ th arrival epoch given that there are  $n$  in  $(0, t]$ ?

Consider  $S_1 = \min\{U_1, \dots, U_n\}$ .

$$P(S_1 > \tau | N(t) = n) = \left(\frac{t-\tau}{t}\right)^n \quad \text{for } 0 < \tau \leq t$$

$\Downarrow$  Prob. of all  $U_i$  exceeds  $\tau$ .

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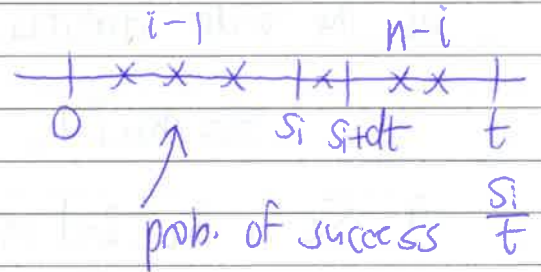
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To derive the density of  $S_i$ ,  $i \in \{1, \dots, n\}$ , given  $\{N(t) = n\}$ , we consider

$$f_{S_i | N(t)=n}(s_i) dt$$

$$= \binom{n-1}{i-1} \left(\frac{s_i}{t}\right)^{i-1} \left(1 - \frac{s_i}{t}\right)^{n-i} \frac{n dt}{t}$$



One of the  $n$  arrivals lands in a small interval  $[s_i, s_i + dt]$ .

$\Rightarrow$  Density of the  $i$ th arrival epoch is

$$f_{S_i | N(t)=n}(s_i) = \frac{n!}{(n-i)!(i-1)!} \frac{s_i^{i-1} (t-s_i)^{n-i}}{t^n}$$

We can also find the expectations of  $S_i$  given  $N(t)=n$ .

$$E[S_i | N(t)=n] = \int_0^t P_r(S_i > \tau | N(t)=n) d\tau$$

$$= \int_0^t \left(\frac{t-\tau}{t}\right)^n d\tau = \frac{1}{t^n} \left[ \frac{(t-\tau)^{n+1}}{n+1} \right]_0^t$$

$$= \frac{t}{n+1}$$

For the other  $E[S_i | N(t)=n]$ , recall that  $f_{X_i | N(t)=n}(x_i | n) \propto \frac{1}{t^n}$  for all  $x_i > 0$  &  $\sum_{i=1}^n x_i < t$ .

This density is symmetric in the  $x_i$ 's, thus all the densities  $f_{X_i | N(t)=n}(x_i | n)$  must be the same by symmetry.

$$\text{Thus } P_r(X_i > \tau | N(t)=n) = P_r(X_1 > \tau | N(t)=n) = P_r(S_1 > \tau | N(t)=n)$$

$$= \left(\frac{t-\tau}{t}\right)^n, \quad 1 \leq i \leq n, \quad 0 < \tau \leq t.$$

$$\Rightarrow E[X_i | N(t)=n] = E[X_1 | N(t)=n] = \frac{t}{n+1}.$$

$$\Rightarrow E[S_i | N(t)=n] = E\left[\sum_{j=1}^i X_j \mid N(t)=n\right] = \frac{it}{n+1}$$



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Counting Process with a Random Rate (Gallager Ex 2.18).

Consider a Poisson process with a random rate  $\Lambda$

$$f_{\Lambda}(\lambda) = \alpha \exp(-\alpha\lambda) \quad \text{for } \lambda \geq 0$$

That is,  $\Lambda$  is an exponential random variable with mean  $1/\alpha$ .

a) What is  $P(N(t)=n \mid \Lambda=\lambda)$ ?

$$P(N(t)=n \mid \Lambda=\lambda) = \frac{(\lambda t)^n e^{-\lambda t}}{n!}, \quad n = 0, 1, 2, 3, \dots$$

b) What is  $P(N(t)=n)$ ?

$$\begin{aligned} \text{Pf 1: } P(N(t)=n) &= \int_0^{\infty} P(N(t)=n \mid \Lambda=\lambda) f_{\Lambda}(\lambda) d\lambda \\ &= \int_0^{\infty} \frac{(\lambda t)^n e^{-\lambda t}}{n!} \alpha \exp(-\alpha\lambda) d\lambda \\ &= \frac{\alpha t^n}{(t+\alpha)^n} \int_0^{\infty} \frac{[\lambda(t+\alpha)]^n \exp(-\lambda(t+\alpha))}{n!} d\lambda \\ &= \frac{\alpha t^n}{(t+\alpha)^{n+1}} \int_0^{\infty} \frac{x^n e^{-x}}{n!} dx = \frac{\alpha t^n}{(t+\alpha)^{n+1}} \end{aligned}$$

$\uparrow$   
 $x = \lambda(t+\alpha)$       Erlang of order  $n+1$

Pf 2: Smarter & less mechanical proof.

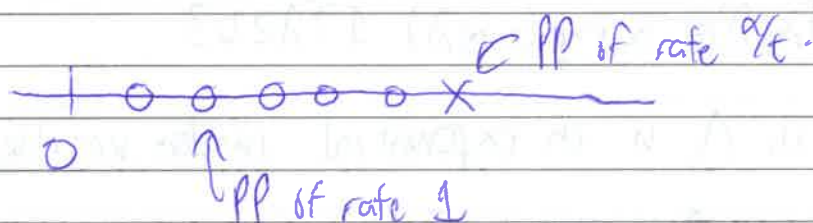
Observe that the solution can be written as  $pq^n$  where  $p = \frac{\alpha}{\alpha+t}$  and  $q = \frac{t}{\alpha+t}$ . function

Note that  $P(N(t)=n)$  for a PP of <sup>fixed deterministic</sup> rate  $\lambda$  is a  $p^n$  of  $\lambda t$  and  $n$  only ↙

$N(t)$  for a Poisson Process of rate  $\lambda \stackrel{d}{=} N(\lambda t)$  for a PP with rate 1.

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Since  $\Lambda t$  is an exponential rv with parameter  $\alpha/t$ ,  $N(\Lambda t)$  represents the # of arrivals of a PP of unit rate before the first arrival of an indep PP of rate  $\alpha/t$ .



The two PP are independent & the combined process has rate  $1 + \alpha/t$ .

Event  $\{N(t)=n\}$  is the prob. of  $n$  arrivals from the unit rate PP followed by 1 arrival of the PP with rate  $\alpha/t$ .

This yields the prob  $q^n p$  (a geometric dist<sup>n</sup>),  $n=0,1,2,\dots$

$$c) f_{\Lambda}(\lambda | N(t)=n) = f_{\Lambda|N(t)}(\lambda | n)$$

$$\text{By Bayes rule } f_{\Lambda|N(t)}(\lambda | n) = \frac{P_r(N(t)=n | \Lambda=\lambda) f_{\Lambda}(\lambda)}{P_r(N(t)=n)}$$

$$= \frac{(\lambda t)^n e^{-\lambda t}}{n!} \alpha e^{-\alpha \lambda} / \frac{\alpha t^n}{(t+\alpha)^{n+1}}$$

$$= \frac{\lambda^n e^{-\lambda(\alpha+t)} (\alpha+t)^{n+1}}{n!} \quad \lambda \geq 0.$$

This is an Erlang pdf of order  $n+1$  with rate  $\alpha+t$ .

$$d) E[\Lambda | N(t)=n]$$

Since conditioned on  $N(t)=n$ ,  $\Lambda$  is Erlang of order  $n+1$ , it is the sum of  $n+1$  independent  $\text{Exp}(\alpha+t)$  rvs. Each of these rvs has mean  $(\alpha+t)^{-1}$ .

$$E[\Lambda | N(t)=n] = \frac{n+1}{t+\alpha}.$$

For  $N(t)=0$  &  $t \ll \alpha$ , this is close to  $1/\alpha = E\Lambda$ . Intuitive since now the data has little effect on  $\Lambda$ , so we revert to prior knowledge.