

## Continuous-time Markov Chains.

We will need to generalize the def'n of discrete Markov chains. This requires a few ingredients.

Def-n. Let  $t \in \mathbb{T}$  be a parameter (i.e. time) and  $S(t)$  a random variable for every  $t$ . The family of random variables  $\{S(t)\}_{t \in \mathbb{T}}$  is called a stochastic process.

Def-n. A stochastic process  $\{S(t)\}_{t \in \mathbb{T}}$  is said to satisfy the Markov property (MP) if for any  $n > 0$  and  $t_{i_0} < t_{i_1} < \dots < t_{i_n}$ :

$$\begin{aligned} P(S(t_{i_n}) = s_{i_n} | S(t_{i_0}) = s_{i_0}, \dots, S(t_{i_{n-1}}) = s_{i_{n-1}}) &= \\ &= P(S(t_{i_n}) = s_{i_n} | S(t_{i_{n-1}}) = s_{i_{n-1}}). \end{aligned}$$

Rmk: The property means that 'the future, given the present, does not depend on the past'.

Examples:

1. Let  $(S, P)$  be a Markov chain with  $S = \{s_1, \dots, s_M\}$ .

Then  $P(S(t_{i_n}) = s_{i_n} | S(t_{i_0}) = s_{i_0}, \dots, S(t_{i_{n-1}}) = s_{i_{n-1}}) =$

$$= P(S(t_{i_n}) = s_{i_n} | S(t_{i_{n-1}}) = s_{i_{n-1}}) = p_{i_{n-1} i_n}, \text{ hence}$$

the MC  $(S, P)$  has MP (here  $\mathbb{T} = \mathbb{Z}_{\geq 0}$  is discrete).

2. Consider the stochastic process  $\{S(t)\}$  with  $t \in T = [0; +\infty)$  and states  $S = \{0, 1, 2, 3, \dots\}$  (countably many states), s.t.  $\{S(t) | t \in T\}$  has MP. Then  $S(t)$  is called a continuous-time MC.

Examples:

(1)  $T = \{0, 1, 2, \dots\} = \mathbb{Z}_{\geq 0}$ ,  $S = \mathbb{Z}$ ,  $S(t_0) = 0$ .

$$S(t_i) = \begin{cases} S(t_{i-1}) + 1, & p = \frac{1}{2} \\ S(t_{i-1}) - 1, & p = \frac{1}{2} \end{cases} \quad (\text{random walk}).$$

Exercise. Show that this process has MP.

(2)  $T = \mathbb{Z} = \{\dots, -2, -1, 0, 1, 2, \dots\}$ ,  $S = \{0, 1, \dots, 10\}$ .

$S(t_i)$  is uniform on  $\{0, 1, \dots, 10\}$ , i.e.

$$P(S(t_i) = k) = \frac{1}{10} \text{ for } 0 \leq k \leq 10.$$

Assume, in addition,  $S(t_i) \geq S(t_{i-1})$  (x)

Notice that given  $S(t_i) = k$  for some  $i$ , condition (x) implies  $S(t_j) = k$  for any  $j \in \mathbb{Z}$ . It follows that  $\{S(t) | t \in T\}$  has MP.

(3) Same as (2), but instead of condition (x), impose

$$S(t_i) = S(t_{i-2}) \quad (\text{xx})$$

Let us show that this stochastic process does not

have MP. Choose  $n=2$ ,  $t_0=0$ ,  $t_1=1$  and  $t_2=2$  with  $S(0)=0$  and  $S(1)=1$ . Then  $P(S(2)=0 \mid S(0)=0, S(1)=1) = 1$  (since  $S(2)=S(2-2)=S(0)$ ). However,  $P(S(2)=0 \mid S(1)=1) = 1/2$ .

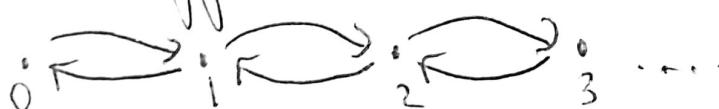
(4)  $T=[0;+\infty)$ ,  $S=\{0, 1, \dots, 10\}$ ,  $S(t)$  is uniform on  $\{0, 1, \dots, 10\}$  and  $S(t)=S(t-1)$  for  $t \geq 1$ .

The process  $\{S(t) \mid t \in T\}$  does not have MP. Choose  $n=2$ ,  $t_0=0.5$ ,  $t_1=0.7$  and  $t_2=1.5$  with  $S(0.5)=1$ ,  $S(0.7)=2$ . Then  $P(S(1.5)=1 \mid S(0.5)=1, S(0.7)=2) = 1$ , but (since  $S(1.5)=S(1.5-1)=S(0.5)$ )

$$P(S(1.5)=1 \mid S(0.7)=2) = 1/2.$$

### Birth-and-Death processes.

These are special cases of continuous-time MC where we impose an extra condition that transitions are allowed only between neighboring states. The transition from  $s_i$  to  $s_{i+1}$  is referred to as 'birth' and in the opposite direction 'death':



Let  $\{S(t) \mid t \in [0;+\infty)\}$  be a stochastic process with MP.

Let  $p_{ij}(t) = P(S(t_0+t) = s_j | S(t_0) = s_i)$  be the family of transition probabilities from state  $s_i$  to  $s_j$  and  $P(t) = (p_{ij}(t))$  the corresponding family of transition matrices.

Rmk: we set  $P(0) = I = \begin{pmatrix} 1 & 0 \\ 0 & \ddots \end{pmatrix}$  be the matrix with 1's on diagonal and 0's everywhere else. Notice, that as there are infinitely many states,  $P(t)$  is infinite down ( $\downarrow$ ) and to the right ( $\rightarrow$ ), i.e.

$$P(t) = \begin{pmatrix} p_{00}(t) & p_{01}(t) & \cdots \\ p_{10}(t) & p_{11}(t) & \cdots \\ p_{20}(t) & p_{21}(t) & \cdots \\ \vdots & \vdots & \ddots \end{pmatrix}.$$

Def-n: the matrix  $Q = \lim_{\Delta t \rightarrow 0} \frac{P(\Delta t) - P(0)}{\Delta t} = \lim_{\Delta t \rightarrow 0} \frac{P(\Delta t) - I}{\Delta t}$  is called the transition rate matrix

In case  $S(t)$  is a Birth-and-Death process

$$Q = \begin{pmatrix} -\lambda_0 & \lambda_0 & 0 & 0 & \cdots \\ \mu_0 & -\mu_0 - \lambda_1 & \lambda_1 & 0 & \cdots \\ 0 & \mu_1 & -\mu_1 - \lambda_2 & \lambda_2 & \cdots \\ \vdots & \vdots & \ddots & \ddots & \ddots \end{pmatrix}$$

where  $\lambda_i$  is the  $i^{\text{th}}$  birth rate and  $\mu_i$  is the  $i^{\text{th}}$  death rate.

Def-n. The vector  $\Pi = (\Pi_0, \Pi_1, \Pi_2, \dots)$  with all  $\Pi_n > 0$  and  $\sum_{n=0}^{\infty} \Pi_n = 1$  is called a stationary distr-n if

$\Pi_n = \lim_{t \rightarrow \infty} P_{in}(t)$  (for any  $i$ ). Here we assume that the process has no absorbing states.

Thm 1. Let  $S(t)$  be a Birth-and-Death process with birth rates  $(\lambda_0, \lambda_1, \lambda_2, \dots)$  and death rates  $(\mu_0, \mu_1, \mu_2, \dots)$ . The stationary distr-n is given by

$$\Pi_n = \frac{\lambda_0 \lambda_1 \dots \lambda_{n-1}}{\mu_0 \mu_1 \dots \mu_{n-1}} \Pi_0, \text{ where } \Pi_0 = \frac{1}{1 + \frac{\lambda_0}{\mu_0} + \frac{\lambda_1 \lambda_0}{\mu_0 \mu_1} + \dots}.$$

Proof: we start with the equality

$$P_{in}(t+\Delta t) = P_{in}(t) \cdot P(\Delta t),$$

subtracting  $P_{in}(t)$  from both sides and dividing by  $\Delta t$ , get  $\frac{P_{in}(t+\Delta t) - P_{in}(t)}{\Delta t} = P_{in}(t) \left( \frac{P(\Delta t) - 1}{\Delta t} \right)$ .

Next, taking the limits when  $t \rightarrow 0$ , come up with

$$\lim_{t \rightarrow 0} \frac{P_{in}(t+\Delta t) - P_{in}(t)}{\Delta t} = P'_{in}(t) = P_{in}(t) \cdot Q = \lim_{t \rightarrow 0} P_{in}(t) \frac{P(\Delta t) - 1}{\Delta t}$$

Rmk: the equations  $P'_{in}(t) = P_{in}(t) \cdot Q$  are called Kolmogorov forward equations.

Finally, taking limits when  $t \rightarrow \infty$ , obtain

$\lim_{t \rightarrow \infty} P'_{in}(t) = 0 = \Pi \cdot Q$ , where the first limit is zero since  $\lim_{t \rightarrow \infty} P_{in}(t) = \Pi_n$  is a constant.

Finally,  $\lim_{t \rightarrow \infty} \Pi(t) \cdot Q = \Pi \cdot Q$  must be equal to 0.

Recalling that  $Q = \begin{pmatrix} -\lambda_0 & \lambda_0 & 0 & 0 & \dots \\ \mu_0 & -\mu_0 - \lambda_1 & \lambda_1 & 0 & \dots \\ 0 & \mu_1 & -\mu_1 - \lambda_2 & \lambda_2 & \dots \\ \vdots & \vdots & \vdots & \ddots & \ddots \end{pmatrix}$ , get

$$\Pi Q = 0 \quad (\Rightarrow) \quad \left\{ \begin{array}{l} \lambda_0 \Pi_0 = \mu_0 \Pi_1 \\ (\mu_0 + \lambda_1) \Pi_1 = \lambda_0 \Pi_0 + \mu_1 \Pi_2 \\ (\mu_1 + \lambda_2) \Pi_2 = \lambda_1 \Pi_1 + \mu_2 \Pi_3 \\ \dots \end{array} \right.$$

Rmk: the system of equations above is called system of balance eq-ns.

Now, from the first balance eq-n  $\Pi_1 = \frac{\lambda_0}{\mu_0} \Pi_0$  from the second  $\Pi_2 = \frac{\lambda_1 + \lambda_0}{\mu_1 + \mu_0} \Pi_0$ , etc.

Exercise. Show that  $\Pi_n = \frac{\lambda_{n-1} + \dots + \lambda_0}{\mu_{n-1} + \dots + \mu_0} \Pi_0$  (use induction)

Finally, as  $\Pi_0 + \Pi_1 + \Pi_2 + \dots = \Pi_0 \left( 1 + \frac{\lambda_0}{\mu_0} + \frac{\lambda_1 + \lambda_0}{\mu_1 + \mu_0} + \dots \right) = 1$ ,

we have  $\Pi_0 = \frac{1}{\frac{\lambda_0}{\mu_0} + \frac{\lambda_1 + \lambda_0}{\mu_1 + \mu_0} + \dots} \quad (*)$ .

Rmk: For the expression in (8) to converge, we need

$$1 + \frac{\lambda_0}{\mu_0} + \frac{\lambda_1 \lambda_0}{\mu_1 \mu_0} + \dots < \infty.$$

For example it is sufficient to have  $\frac{\lambda_i}{\mu_i} < q < 1$ , then

$$1 + \frac{\lambda_0}{\mu_0} + \frac{\lambda_1 \lambda_0}{\mu_1 \mu_0} + \dots < 1 + q + q^2 = \frac{1}{1-q}$$
 is convergent geometric series.

## Queues.

A basic queue model consists of three entities:

- arriving objects;
- a queue of arrived objects;
- (a) processing unit(s)

## The Kendall notation

The standard notation for a queue model is

$A/S/c/b$ , where

$A$  is the pdf for arrival times

$S$  is the pdf for service times

$c$  is the number of parallel service channels

$b$  is the system capacity restriction (max. number of people in queue).

## The M/M/1 model ( $M/M/1/\infty$ )

Customers are served one at a time, the length of the line is unbounded. The arrivals are according to a Poisson process and the service time distribution is exponential (with parameters  $\lambda$  and  $\mu$ ).

Using Thm 1 we get that the fixed prob. vector is  $w = (1-p, p(1-p), p^2(1-p), \dots)$ , where  $p = \frac{\lambda}{\mu}$ , and one needs  $p < 1$  for the expression in (8) to converge. Rmk. The distribution  $\pi_n = p^n(1-p)$  is called geometric with param.  $p$ .

Example (HW, page 23, #1). Suppose that 2 people arrive per hour and it takes an average of 10 minutes to serve a customer. Find the fixed probability vector for this queue.

Sol-n. This is an M/M/1 queue. Here  $\lambda = 2$  and

$\mu = 6$  (since  $1 \text{ customer}/10 \text{ mins} = 6 \text{ customers/hour}$ ), so  $p = \frac{\lambda}{\mu} = \frac{1}{3} < 1$ .

$$\text{Hence, } \pi_n = p^n(1-p) = \frac{1}{3^n} \cdot \frac{2}{3} = \frac{2}{3^{n+1}}.$$

(2) HW, page 23, #2. Find the expected number of customers in the M/M/1 queue.

$$\text{Sol-n. } E(N) = \sum_{n=0}^{\infty} n \pi_n = \sum_{n=0}^{\infty} n \cdot p^n(1-p) = (1-p) \sum_{n=0}^{\infty} np^n =$$

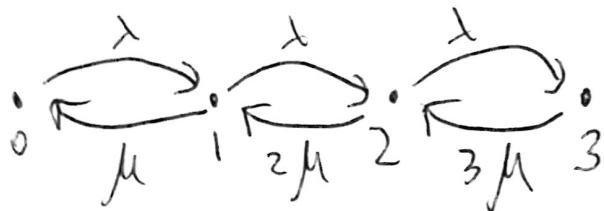
$$= (1-p) \cdot p \frac{d}{dp} \sum_{n=0}^{\infty} p^n = (1-p) \cdot p \frac{d}{dp} \left( \frac{1}{1-p} \right) = \frac{p(1-p)}{(1-p)^2} = \frac{p}{1-p} =$$

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

Example. (HW, p.24, #3).

M/M/3/3 queue, find  $\pi_i$ 's.

Sol-n:



$$Q = \begin{pmatrix} -\lambda & \lambda & 0 & 0 \\ \mu & -\mu-\lambda & \lambda & 0 \\ 0 & 2\mu & -\lambda-2\mu & \lambda \\ 0 & 0 & 3\mu & -3\mu \end{pmatrix}$$

$$\pi Q = 0 \Leftrightarrow \begin{cases} \lambda \pi_0 = \mu \pi_1 \\ (\mu + \lambda) \pi_1 = \lambda \pi_0 + 2\mu \pi_2 \\ 3\mu \pi_3 = \lambda \pi_2 \end{cases}$$

$$\pi_1 = \frac{\lambda}{\mu} \pi_0, \pi_2 = \frac{\lambda^2}{2\mu^2} \pi_0 = \frac{\rho^2}{f_2} \pi_0, \pi_3 = \frac{\lambda}{3\mu} \pi_2 = \frac{\rho^3}{f_6} \pi_0$$

$\rho \pi_0$

$$\pi_0 \left( 1 + \rho + \frac{\rho^2}{f_2} + \frac{\rho^3}{f_6} \right) = 1, \text{ so } \pi_0 = \frac{1}{1 + \rho + \frac{\rho^2}{f_2} + \frac{\rho^3}{f_6}}.$$

Thm 2 (Little's formula). Let  $L$  be the average number of customers in the system and  $W$  the average amount of time (including waiting and service) a customer spends in the system. Then

$$L = \lambda \cdot W$$

Let us give the idea of the proof. It is based on the cost identity. Imagine that each entering customer is forced to pay money (according to some rule) to the system. Then we would have the following identity:

$$(xx) \quad \text{average rate at which the system earns money} = \lambda \cdot \begin{matrix} \downarrow \\ \text{arrival rate} \end{matrix} \cdot \begin{matrix} \text{average amount of} \\ \text{money an entering} \\ \text{customer pays} \end{matrix}$$

Proof of (xx): let  $R(t)$  be the amount of money the system earned by time  $t$ , then

$$\begin{aligned} \text{l.h.s of (xx)} &= \lim_{t \rightarrow \infty} \frac{R(t)}{t} = \lim_{t \rightarrow \infty} \frac{N(t)}{t} \cdot \frac{R(t)}{N(t)} = \\ &= \lambda \lim_{t \rightarrow \infty} \frac{R(t)}{N(t)} = \text{r.h.s of (xx), where } N(t) \text{ is the number} \\ &\text{of customers who entered the system by time } t \text{ (inclusive)} \end{aligned}$$

To prove Little's formula, we use the following paying rule: each customer pays \$1 per unit time he/she spends in the system. Then the average amount of money payed by a customer = average amount of time spent by a customer in the system ( $W$ ). On the other hand, we can show that  $L$  = average rate at which the system earns money. Indeed, the amount of money earned by the system on the time interval  $(t, t+\Delta t)$  is given by  $X(t)\Delta t$ , where  $X(t)$  is the number of customers in the system at time  $t$ . Hence, the rate in question is  $\lim_{t \rightarrow \infty} \frac{\int X(s)ds}{t} = L$  (by def'n).

Now the result follows from the cost identity.

Rmk. Let  $L_Q$  be the average number of customers waiting in queue and  $W_Q$  the average amount of time a customer spends in queue. Then  $L_Q = \lambda \cdot W_Q$ . The proof is completely analogous.

Exercise. Show the required modification to the paying rule.

Example (HW, page 25, #1).

M/M/1 queue with  $\lambda=2$  and  $\mu=6$ . Find  $L, W, L_Q, W_Q$ .

Sol-n. We have shown that  $L = \frac{\lambda}{\mu-\lambda} = \frac{2}{6-2} = \frac{1}{2}$  (page 7).

Using Little's f-la, we get  $W = \frac{L}{\lambda} = \frac{1}{4}$ .

Next,  $W_Q = W - \frac{1}{\mu}$   $= \frac{1}{4} - \frac{1}{6} = \frac{1}{12}$ .

↑ mean service time

Using Little's f-la again,  $L_Q = \lambda W_Q = \frac{2}{12} = \frac{1}{6}$ .

General f-las for the M/M/1 model:

$$L = \frac{\lambda}{\mu-\lambda}$$

$$W = \frac{1}{\mu-\lambda}$$

$$W_Q = \frac{\lambda}{\mu(\mu-\lambda)}$$

$$L_Q = \frac{\lambda^2}{\mu(\mu-\lambda)}$$

## Time Spent in the System.

Let  $T$  be the time a customer spends in the system. If there are  $n$  customers at the moment he/she arrives then  $T$  is the sum of the service times of  $n$  customers.

Recall that the service time in M/M/1 queue has the exponential distr-n with pdf

$$f(x) = \begin{cases} \mu e^{-\mu x}, & x \geq 0 \\ 0, & x < 0 \end{cases}$$

This distribution also has the memorylessness property:  $P(T > s+t | T > s) = P(T > t)$ ,  $s, t \geq 0$ . (x)

We would like to find the pdf of r.v.  $T_n$ , the amount of time the customer spends in system given there are  $n$  customers in front.

It follows from (x) that  $f_{T_n}(t) = \frac{(\lambda t)^n}{n!} e^{-\lambda t}$ ,  $t \geq 0$ .

Exercise. Show that for a collection of i.i.d. random variables  $X_1, \dots, X_n$  with pdf  $f_{X_i}(t) = \lambda e^{-\lambda t}$  (exp. distr-ns with param.  $\lambda$ ), one has  $f_X(t) = \frac{(\lambda t)^n}{n!} e^{-\lambda t}$ , where  $X = X_1 + \dots + X_n$ .

To obtain the distribution for  $T$  (time the customer spends in the system), we compute:

$$f_T(t) = \sum_{n=0}^{\infty} \frac{(\mu t)^n}{n!} \mu e^{-\mu t} \cdot \text{term} = \sum_{n=0}^{\infty} \frac{(\mu t)^n}{n!} \mu e^{-\mu t} p^n (1-p)$$

$$= \mu e^{-\mu t} (1-p) \sum_{n=0}^{\infty} \frac{(\lambda t)^n}{n!} = (\mu - \lambda) e^{(\lambda - \mu)t}, \quad t > 0,$$

$$p = \frac{\lambda}{\mu} \Rightarrow p\mu t = \lambda t$$

which is the exponential distribution with parameter  $\lambda - \mu$ . It has the cdf  $F_T(t) = P(T \leq t) = \int_0^t (\mu - \lambda) e^{-(\mu - \lambda)s} ds$

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

Rmk. In particular,  $W = E(T) = \frac{1}{\mu - \lambda}$ , as we have established before using Little's formula.

Example. (page 26, #9) For the M/M/1 system, where 3 people arrive each minute (on average) and it takes 15 sec. to serve a customer: ...

(a) find  $L, L_Q, W, W_Q$ .

$$\text{Answer: } \mu = 4 \text{ customers/min, so } L = \frac{\lambda}{\mu - \lambda} = 3,$$

$$L_Q = \frac{\lambda^2}{\mu(\mu - \lambda)} = \frac{9}{4}, \quad W = \frac{1}{\mu - \lambda} = 1, \quad W_Q = \frac{\lambda}{\mu(\mu - \lambda)} = \frac{3}{4}.$$

(b) give  $F_T(t)$  and use it to find  $P(T>1)$ .

Answer:  $f_T(t) = (\mu - \lambda) e^{-(\mu - \lambda)t} = e^{-t}$

$$P(T>1) = 1 - P(T \leq 1) = 1 - F_T(1) = 1 - \int_0^1 e^{-t} dt =$$
$$= 1 + e^{-t} \Big|_0^1 = 1 + (e^{-1} - 1) = e^{-1} = \frac{1}{e}.$$

(c)  $0.5 = \int_0^{\text{median}} e^{-t} dt = -e^{-t} \Big|_0^{\text{median}} = 1 - e^{-\text{median}}$

$$e^{-\text{median}} = 0.5$$

$$-\text{median} = \ln(0.5).$$

$$\text{median} = -\ln(0.5).$$

#5. For M/M/1 queue. (page 24)

Practice problems

(ii) find  $p_n$



$$\begin{cases} \lambda p_0 = \mu p_1 \\ (\lambda + \mu) p_1 = \lambda p_0 + 2\mu p_2 \\ (\lambda + 2\mu) p_2 = \lambda p_1 + 2\mu p_3 \\ \vdots \end{cases}$$

$$p_1 = f_1 p_0$$

$$p_2 = f_2 p_0$$

$$p_3 = f_3 p_0$$

$$(f_1 + \lambda) f_2 p_0 = f_1^2 p_0 + 2p_3$$

Claim:  $p_n = f_{2^{n-1}}^n p_0$

Proof:  $(\lambda + 2\mu) p_{n-1} = \lambda p_{n-2} + 2\mu p_n \quad | : \mu \quad (\text{ind-n assumption})$

$$(f_1 + \lambda) f_{2^{n-2}}^{n-1} p_0 = f_{2^{n-3}}^{n-1} p_0 + 2p_n$$

$$p_n = f_{2^{n-1}}^n p_0$$

$$p_0 \underbrace{(1 + f_1 + f_2 + f_3 + \dots)}_{\text{II}} = 1$$

$$1 + \frac{f_1}{1 - f_2} = \frac{p_0}{1 + p_1/2} = \frac{\frac{1}{1 + p_1/2}}{1 - p_1/2} = \frac{1}{1 + p_1/2}$$

$$(b) \lambda = 5/\text{hour}, M = 6/\text{hour} \Rightarrow p = \frac{5}{6}.$$

$$p_n = \frac{p^n}{2^{n-1}} \cdot \frac{1-p/2}{1+p/2} = \frac{5^n}{2^{n-1} \cdot 6^n} \cdot \frac{7/12}{17/12} = \frac{7}{17 \cdot 2^{n-1}} \cdot \left(\frac{5}{6}\right)^n.$$

$$(c) L = \sum_{n=0}^{\infty} np_n = \sum_{n=0}^{\infty} n \frac{p^n}{2^{n-1}} \cdot \frac{1-p/2}{1+p/2} = \frac{2-p}{2+p} \cdot p \sum_{n=0}^{\infty} n \cdot \left(\frac{p}{2}\right)^{n-1} =$$

$$= \frac{p(2-p)}{2+p} \cdot \left(\frac{1}{1-p/2}\right)' = \frac{p(2-p)}{2+p} \cdot \frac{1}{2(1-p/2)^2} = \frac{p(2-p)}{(2+p)(2-p)(1-p/2)}$$

$$= \frac{p}{(2+p)(1-p/2)} = \frac{p}{2(1+p/2)(1-p/2)} = \frac{p}{2(1-p^2/4)}$$

Example. (HW, p.24, #3).

M/M/3/3 queue, find  $\bar{\pi}_i$ 's.

Sol-n:



$$Q = \begin{pmatrix} -\lambda & \lambda & 0 & 0 \\ \mu & -\mu-\lambda & \lambda & 0 \\ 0 & 2\mu & -\lambda-2\mu & \lambda \\ 0 & 0 & 3\mu & -3\mu \end{pmatrix}$$

$$\Pi Q = 0 \Leftrightarrow \begin{cases} * \lambda \bar{\pi}_0 = \mu \bar{\pi}_1 \\ (\mu + \lambda) \bar{\pi}_1 = \lambda \bar{\pi}_0 + 2\mu \bar{\pi}_2 \\ 3\mu \bar{\pi}_3 = \lambda \bar{\pi}_2 \end{cases}$$

$$\bar{\pi}_1 = \frac{\lambda}{\mu} \bar{\pi}_0, \quad \bar{\pi}_2 = \frac{\lambda^2}{2\mu^2} \bar{\pi}_0 = \frac{p^2}{2!} \bar{\pi}_0, \quad \bar{\pi}_3 = \frac{\lambda}{3\mu} \bar{\pi}_2 = \frac{p^3}{6!} \bar{\pi}_0$$

$p \bar{\pi}_0$

$$\bar{\pi}_0 \left( 1 + p + \frac{p^2}{2!} + \frac{p^3}{6!} \right) = 1, \text{ so } \bar{\pi}_0 = \frac{1}{1 + p + \frac{p^2}{2!} + \frac{p^3}{6!}}.$$

Problem. M/M/2/3 system,  $\lambda=3$ ,  $\mu=1$ .

(a) Find  $p_0, p_1, p_2, p_3$ .



Balance eqns:

$$\begin{cases} 3p_0 = p_1 \\ (3+1)p_1 = 3p_0 + 2p_2 \\ (2+2)p_2 = 3p_1 + 2p_3 \end{cases}$$

$$p_1 = 3p_0, \quad p_2 = \frac{9}{2}p_0, \quad p_3 = \frac{27}{4}p_0$$

$$p_0(1 + 3 + \frac{9}{2} + \frac{27}{4}) = 1.$$

$$\frac{61}{4}p_0 = 1 \Rightarrow p_0 = \frac{4}{61}, \quad p_1 = \frac{12}{61}, \quad p_2 = \frac{18}{61}, \quad p_3 = \frac{27}{61}.$$

(b) Find  $L, L_Q, W, W_Q$ .

$$L = 0 \cdot p_0 + 1 \cdot p_1 + 2 \cdot p_2 + 3 \cdot p_3 = \frac{12 + 2 \cdot 18 + 3 \cdot 27}{61} = \frac{129}{61}$$

$$\text{Little's f-la, } W = \frac{129}{61 \cdot 3} = \frac{129}{183} = \frac{43}{61}.$$

$$W_Q = W - \frac{1}{\mu} = \frac{43}{61} - \frac{61}{102} = 0.107$$

mean waiting time

$$\mu' = 1 \cdot p_1 + 2 \cdot p_2 + 2 \cdot p_3 = \frac{12 + 36 + 54}{61} = \frac{102}{61}$$

$$\text{Little's f-la, } L_Q = \lambda W_Q = 3 \cdot 0.107 = 0.321.$$

Problem 6, page 26. Show that the exponential distribution  $f_X(x) = \lambda e^{-\lambda x}$  for  $x \geq 0$ , has the memoryless property:

$$P(X > a | X > b) = P(X > a - b).$$

Proof:  $P(X > a | X > b) = \frac{P((X > a) \cap (X > b))}{P(X > b)} \stackrel{a > b}{=} \frac{P(X > a)}{P(X > b)}$

$$P(X > a) = 1 - P(X \leq a) = 1 - \int_0^a \lambda e^{-\lambda x} dx = 1 + e^{-\lambda x} \Big|_0^a = e^{-\lambda a}.$$

Hence,  $\frac{P(X > a)}{P(X > b)} = \frac{e^{-\lambda a}}{e^{-\lambda b}} = e^{-\lambda(a-b)}.$

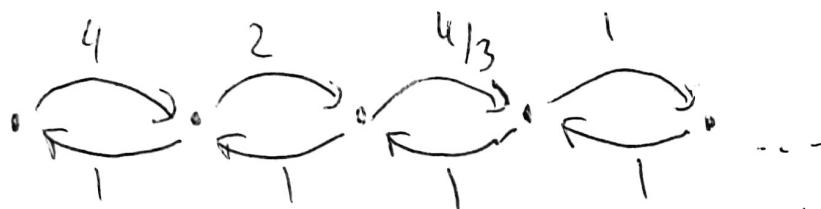
$$P(X > a - b) = e^{-\lambda(a-b)}, \checkmark.$$

Problem 7, page 24. Suppose a food truck has one server and serves on average one person per 2 minutes. The arrival rate is  $\frac{2}{n+1}$  people per minute, where  $n$  is the number of people in line.

(a) Find  $p_i$ 's.

(b) Find  $L$ .

Sol-n:  $M=1, \lambda_n = \frac{4}{n+1}$  people / 2 mins



(a) Notice that the balance eq-n at vertex  $n$  is

$$(\lambda_n + \mu_n) p_n = \lambda_{n-1} p_{n-1} + \mu_{n+1} p_{n+1} \text{ or}$$

$$\left(\frac{4}{n+1} + 1\right) p_n = \frac{4}{n} p_{n-1} + p_{n+1}, \text{ i.e.}$$

$$p_{n+1} = \left(\frac{4}{n+1} + 1\right) p_n - \frac{4}{n} p_{n-1} \quad (\text{a})$$

Let  $P(s) = \sum_{n=0}^{\infty} p_n \cdot s^n$  be the generating function

for  $(p_0, p_1, p_2, \dots)$ .

We want to see which functional eq-n on  $P(s)$  corresponds to (a).

$$\text{Notice that } \int p(s)ds = \sum_{n=0}^{\infty} p_n \int s^n ds = \sum_{n=0}^{\infty} \frac{1}{n+1} p_n \cdot s^{n+1}.$$

It follows that the functional eq-n we are looking for is  $p(s) - p_0 + p_0 \cdot s = 4 \int p(s)ds + s p(s) - 4s \int p(s)ds$ .  
Check: the coefficients of  $s^{n+1}$ :

$$s^0: 0 = 0. \checkmark$$

$$s^1: p_0 + p_1 = 4p_0 + p_0, \text{ i.e. } p_1 = 4p_0 \checkmark$$

$$s^{n+1}, n \geq 1: p_{n+1} = 4 \cdot \frac{1}{n+1} p_n + p_n - 4 \cdot \frac{1}{n} p_{n-1} \hookrightarrow (\star) \checkmark$$

Next we find the sol-n of functional eq-n  $(\star)$ :

$$p(s) - s p(s) - p_0 + p_0 s = (4 - 4s) \int p(s)ds \mid : (1-s)$$

$$p(s) - p_0 = 4 \int p(s)ds$$

It is not hard to see that  $p(s) = C \cdot e^{4s}$ , where  $C$  is a constant. Since we know that  $p(0) = p_0$ , get

$$p(0) = C = p_0, \text{ hence } p(s) = p_0 \cdot e^{4s}.$$

$$\text{So } p(s) = \sum_{n=0}^{\infty} p_0 \cdot \frac{4^n}{n!} s^n \text{ and } p_n = \frac{4^n}{n!} p_0.$$

Finally, as  $p_0 \cdot \sum_{n=0}^{\infty} \frac{4^n}{n!} = 1$ , we get  $p_0 = e^{-4}$  and  $p_n = \frac{4^n}{n!} e^{-4}$ .

$$(b) L = \sum_{n=0}^{\infty} \frac{4^n}{n!} e^{-4} \cdot n = e^{-4} \sum_{n=0}^{\infty} \frac{4^n}{n!} \cdot n \stackrel{(!)}{=} e^{-4} \cdot 4 \cdot e^4 = 4.$$

$$(!) \text{ Use that } x \frac{d}{dx}(e^x) = \sum_{n=0}^{\infty} \frac{x^n}{n!} \cdot n \text{ and } x \frac{d}{dx}(e^x) = x e^x.$$