

Random Processes Tute

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1 Tute 1

1.

$$Q = \begin{pmatrix} -a & a \\ b & -b \end{pmatrix}$$

- (a) State space is $\{0, 1\}$
- (b) Eigen values and Eigenvectors:
Eigenvalues: Solve determinant of

$$|Q - \lambda I| = 0$$

$$\begin{aligned} (-a - \lambda)(-b - \lambda) - ab &= 0 \\ ab + (a + b)\lambda + \lambda^2 - ab &= 0 \\ \lambda(a + b + \lambda) &= 0 \\ \implies \lambda &= -a - b \text{ or } 0 \end{aligned}$$

Right eigenvector: find v such that

$$\begin{aligned} Qv &= \lambda v \implies (Q - \lambda I)v = 0 \\ \begin{pmatrix} -a - \lambda & a \\ b & -b - \lambda \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} &= 0 \end{aligned}$$

Plug in the eigenvalues $\lambda = -a - b$:

$$\begin{aligned} \begin{pmatrix} -a - (-a - b) & a \\ b & -b - (-a - b) \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} &= 0 \\ \begin{pmatrix} b & a \\ b & a \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} & \\ bv_1 + av_2 &= 0 \\ \implies v_1 &= \frac{av_2}{b} \\ v &= \begin{pmatrix} a/b \\ 1 \end{pmatrix} \end{aligned}$$

Repeat this for $\lambda = 0$

$$\begin{aligned} \begin{pmatrix} -a & a \\ b & -b \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} &= 0 \\ -av_1 + av_2 &= 0 \\ bv_1 - bv_2 &= 0 \\ \implies v_1 &= v_2 \\ v &= \begin{pmatrix} 1 \\ 1 \end{pmatrix} \end{aligned}$$

Left eigenvectors are a similar problem, and give for $\lambda = 0$:

$$v = (1 \quad a/b)$$

for $\lambda = -a - b$:

$$v = (1 \quad -1)$$

(c) Find trans function

$$\begin{aligned} P(t) &= e^{Qt} \\ &= \sum_{n=0}^{\infty} \frac{(Qt)^n}{n!} \\ &= \sum_{i=1}^m e^{\lambda_i t} M_i \end{aligned}$$

I.e. exp of the eigenvalue, and the matrix based on the eigenvectors. Where $M_i = r'_i l_i$ where r_i is the i^{th} right eigenvector and i^{th} left eigenvector. Where $r'_i l_i = 1$ Let M_1 correspond to $\lambda = 0$

$$M_1 = \begin{pmatrix} 1 & 1 \\ b & a \end{pmatrix} = 1 \implies M_1 = \begin{pmatrix} \frac{b}{b+a} & \frac{a}{b+a} \\ \frac{b}{b+a} & \frac{a}{b+a} \end{pmatrix}$$

Similarly solve M_2

2.

$$Q = \begin{pmatrix} -3 & 3 & 0 \\ 2 & -4 & 2 \\ 0 & 6 & -6 \end{pmatrix}$$

(a) State space $\{0, 1, 2\}$

(b) Equilibrium equations

$$\pi Q = 0$$

$$-3\pi_1 + 2\pi_2 = 0$$

$$3\pi_1 - 4\pi_2 + 6\pi_3 = 0$$

$$2\pi_2 - 6\pi_3 = 0$$

$$\text{such that } \pi_1 + \pi_2 + \pi_3 = 1$$

$$\pi = (1/3 \quad 1/2 \quad 1/6)$$

(c) Find \mathbb{P} using Q solve the equilibrium equations. I.e. solve

$$\pi = \pi \mathbb{P}$$

$P =$

$$\begin{pmatrix} 0 & 1 & 0 \\ 1/2 & 0 & 1/2 \\ 0 & 1 & 0 \end{pmatrix}$$

$$\pi = (1/4, 1/2, 1/4)$$

Solve equilibrium equations for this P basically.

(d) What is the relationship between these two distributions? If we multiply this π based on P by the inverse of the diagonal of Q , and then normalise, we will get the π from the previous question.

2 Tute 3

1. $P_{Nj}(t)$

(a) Write down Q . In general:

$$q_{i,j} = \begin{cases} \mu & j = i - 1 \\ -\mu & j = i \\ 0 & \text{otherwise} \end{cases}$$

In this case we only have 1 individual

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

- (b) Write and solve the KFDEs for $P_{10}(t)$ and $P_{11}(t)$

$$\frac{\partial P_{ij}(t)}{\partial t} = \sum_{k \in S} P_{ik}(t) q_{kj}$$

$$P' = PQ = \begin{pmatrix} P_{00} & P_{10} \\ P_{01} & P_{11} \end{pmatrix} \begin{pmatrix} 0 & 0 \\ \mu & -\mu \end{pmatrix}$$

$$\begin{pmatrix} \frac{\partial P_{00}}{\partial t} & \frac{\partial P_{01}}{\partial t} \\ \frac{\partial P_{10}}{\partial t} & \frac{\partial P_{11}}{\partial t} \end{pmatrix} = \begin{pmatrix} P_{10}\mu & -\mu P_{10} \\ P_{11}\mu & -\mu P_{11} \end{pmatrix}$$

Start with $\frac{\partial P_{11}}{\partial t} = -\mu P_{11}$

$$\implies P_{11}(t) = Ae^{-\mu t}$$

$$P_{11}(0) = 1 \implies A = 1 \implies P_{11}(t) = e^{-\mu t}$$

Now find $P_{10}(t)$

$$\frac{\partial P_{10}(t)}{\partial t} = \mu P_{11}(t) = \mu e^{-\mu t}$$

$$P_{10}(t) = B - e^{-\mu t}$$

But $P_{10}(t) = 1 - P_{11}(t) \implies B = 1$

$$P_{10}(t) = 1 - e^{-\mu t}$$

- (c) Physical meaning:

$P_{10}(t)$ is the probability of being absorbed into 0 by time t . As $t \rightarrow \infty$ we find $P_{10}(t) = 1$ i.e. we will be absorbed into 0. Similarly, $P_{11}(t)$ is the probability of remaining in state 1 by time t . And this goes to 0 as $t \rightarrow \infty$

- (d) What if we had N components? Find $P_{Nj}(t)$ We could consider the components as iid bernoulli trials

$$P_{Nj}(t) = \binom{N}{j} P_{11}(t)^j P_{10}(t)^{N-j}$$

$$= \binom{N}{j} (e^{-\mu t})^j (1 - e^{-\mu t})^{N-j}$$

2. (a)

- (b)

3. Population of constant size N as a CTMC where $X(t)$ is the number of infectious people at time t . Assume infections occur at rate

$$q_{i,i+1} = \beta i(N-i)/(N-1)$$

And recoveries occur at

$$q_{i,i-1} = \gamma i$$

For $i \in S = \{0, 1, \dots, N\}$

- (a) What are the communicating classes?

0 is a communicating class, and $\{1, \dots, N\}$ is a communicating class 0 is an absorbing state. Irreducible. Finite means recurrent, 0 is recurrent, and $\{1, \dots, N\}$ is transient. It has a steady state solution

$$\pi = P\pi$$

- (b) System of linear equations which allow us to calculate probability of having at least j infectious individuals. at some stage over the course of the epidemic, having started with $i \leq j$ infectious individuals. Evaluate this for $i = 1$ and $j = 1, 2, 3$

$$-q_{ii}f_{ij} = \sum_{k \neq i} q_{ik}f_{kj}$$

$$-q_{11}f_{1j} = \sum_{k \neq 1} q_{1k}f_{kj}$$

$$(\beta 1(N-1)/(N-1) + \gamma 1)f_{1j} = q_{10}f_{0j} + q_{12}f_{2j}$$

$$(\beta + \gamma)f_{1j} = q_{12}f_{2j}$$

$$= \beta(N-1)/(N-1)f_{21} = \beta f_{2j}$$

$$(\beta + \gamma)f_{1j} = \beta f_{2j}$$

For $j = 1$ we get $f_{11} = 1$ because trivially it is 1. For $j = 2$:

$$(\beta + \gamma)f_{12} = \beta f_{22}$$

$$f_{12} = \frac{\beta}{\beta + \gamma}$$

For $j = 3$:

$$(\beta + \gamma)f_{13} = \beta f_{23}$$

$$f_{13} = \frac{\beta}{\beta + \gamma} f_{23}$$

$$f_{23} = \frac{q_{21}}{-q_{22}} f_{13} + \frac{q_{23}}{-q_{22}} f_{33}$$

$$= \frac{q_{21}}{-q_{22}} f_{13} + \frac{q_{23}}{-q_{22}}$$

its fucking grotty but we end up with

$$\implies f_{13} = \frac{\beta^2(N-2)}{\beta^2(N-2) + \beta\gamma(N-2) + \gamma^2/(N-1)}$$

3 Tute 4

1. Infinite buffered switch. Poisson arrival rate λ , processed at rate $\mu < \lambda$. If the queue increases past K , the process rate becomes $2\mu > \lambda$. This happens until it hits 0, and then it goes back to μ .

- (a) Use a CTMC model to show that if the number of packets is less than K , it will hit K with probability 1. Starting in j then

$$f_j = \frac{\lambda}{\lambda + \mu} f_{j+1} + \frac{\mu}{\lambda + \mu} f_{j-1}$$

With boundaries

$$f_k = 1, \quad f_0 = f_1$$

$$f_j = A \left(\frac{\mu}{\lambda} \right)^j + B$$

Using boundaries:

$$f_k = 1 = A \left(\frac{\mu}{\lambda} \right)^k + B \implies B = 1 - A \left(\frac{\mu}{\lambda} \right)^k$$

Meaning

$$f_j = 1 + A \left(\left(\frac{\mu}{\lambda} \right)^j - \left(\frac{\mu}{\lambda} \right)^k \right)$$

$$f_0 = f_1 \implies 1 + A \left(\left(\frac{\mu}{\lambda} \right)^1 - \left(\frac{\mu}{\lambda} \right)^k \right) = 1 + A \left(\left(\frac{\mu}{\lambda} \right)^0 - \left(\frac{\mu}{\lambda} \right)^k \right)$$

$$\implies A = 0$$

Which then gives

$$f_j = 1 \text{ always}$$

- (b) Show that it will hit 0 after hitting K Similar problem, only with $f_0 = 1$ and no other boundary condition. This has to use the minimal non-negative thing. So fun! Let f_j be hitting prob for 0.

$$f_j = \frac{\lambda}{\lambda + 2\mu} f_{j+1} + \frac{2\mu}{\lambda + 2\mu} f_{j-1}$$

Again we get

$$f_j = A \left(\frac{2\mu}{\lambda} \right)^j + B$$

$$f_0 = 1 \implies B = 1 - A$$

$$f_j = 1 + \left(A \left(\frac{2\mu}{\lambda} \right)^j - 1 \right)$$

We need the minimal non-negative solution. Since $2\mu > \lambda$ then $A = 0$. Giving:

$$f_j = 1$$

(c) Calculate the expected time between reaching K and zero again.

Let $t_k^{(0)}$ be the expected hitting time for state k to reach 0

$$t_k = -\frac{1}{q_{ii}} + \sum_{i \neq k} \frac{q_{ki}}{-q_{kk}} t_i$$

$$t_0 = 0.$$

$$t_k = \frac{1}{\lambda + 2\mu} + \frac{\lambda}{\lambda + 2\mu} t_{k+1} + \frac{2\mu}{\lambda + 2\mu} t_{k-1}$$

guess : $t_k = Ck$

$$Ck = \frac{1}{\lambda + 2\mu} + \frac{\lambda}{\lambda + 2\mu} C(k+1) + \frac{2\mu}{\lambda + 2\mu} C(k-1)$$

$$Ck = Ck + \frac{\lambda}{\lambda + 2\mu} - C \frac{2\mu}{\lambda + 2\mu} + \frac{1}{\lambda + 2\mu}$$

$$\implies C = \frac{1}{2\mu - \lambda}$$

Homogeneous Solution:

$$t_k = A \left(\frac{2\mu}{\lambda} \right)^k + B + \frac{1}{2\mu - \lambda} k$$

Now plug in boundary:

$$t_0 = 0 = A + B \implies B = -A$$

$$\implies t_k = A \left(\left(\frac{2\mu}{\lambda} \right)^k - 1 \right) + \frac{1}{2\mu - \lambda} k$$

Minimal non-negative:

$$\implies A = 0$$

$$\therefore t_k = \frac{j}{2\mu - \lambda}$$

2. Expected hitting times and cost

$$c_i = \mathbb{E} [\text{cost until reaching } j | X(0) = i]$$

$$= \int_0^\infty E[\$_{X(t)} | X(0) = i] dt$$

$$c_i = E[E[\text{cost until reaching } j | X(0) = i, T_1 = s, X(T_1) = k]]$$

$$= \int_0^\infty \sum_{\substack{k \in S \\ k \neq i}} E[\text{cost until } j | X(0) = i, T_1 = s, X(T_1) = k] \frac{q_{ik}}{-q_{ii}} (-q_{ii}) e^{q_{ii}s} ds$$

$$= \int_0^\infty \sum_{\substack{k \in S \\ k \neq i}} E[\text{cost until } j | X(0) = i, T_1 = s, X(T_1) = k] q_{ik} e^{q_{ii}s} ds \text{ come back...}$$

$$\begin{aligned} E[\text{cost until } j | X(0) = i, T_1 = s, X(T_1) = k] &= \int_0^\infty E[\$_{X(t)} | X(0) = i, T_i = S, X(T_i) = k] dt \\ &= \int_0^S \$_i dt + \int_S^\infty E[\$_{X(t)} | X(S) = k] dt \\ &= \$_i S + c_k \end{aligned}$$

Back to the come back.. line

$$\begin{aligned}
 c_i &= \int_0^\infty \sum_{\substack{k \in S \\ k \neq i}} (\$iS + c_k) q_{ik} e^{q_{ii}S} dS \\
 &= \sum_{\substack{k \in S \\ k \neq i}} \left[\$i \int_0^\infty S e^{q_{ii}S} dt + c_k \int_0^\infty e^{q_{ii}S} dS \right] \\
 &= \frac{\$i}{-q_{ii}} + \sum_{\substack{k \in S \\ k \neq i}} \frac{q_{ik}}{-q_{ii}} c_k
 \end{aligned}$$

3. 2D birth death process. Room 1 has R_1 spots for customers of type 1 for queue 1, similarly R_2 for type 2 for queue 2 and R_3 holds both 1 and 2 as overflow. These customers then move to their rooms when they can.

Let λ_1, λ_2 be the poisson arrivals and μ_1, μ_2 be the service rates.

- (a) Restrict the state space:

Let n_1, n_2 be the number of people in queue 1 and queue 2 respectively.

$$A = \{(n_1, n_2) \in \mathbb{N}^2 : [n_1 - (R_1 + 1)]^+ + [n_2 - (R_2 + 1)]^+ \leq R_3\}$$

- (b) Write down the joint equilibrium

Since all the rates are independent, we just use the formula. For when we un-restrict it:

$$\pi(n) = C \left(\frac{\lambda_1}{\mu_1} \right)^{n_1} \left(\frac{\lambda_2}{\mu_2} \right)^{n_2}$$

$$C = (1 - \frac{\lambda_1}{\mu_1})(1 - \frac{\lambda_2}{\mu_2})$$

Truncated version;

$$\pi(n) = C_{trunc} \left(\frac{\lambda_1}{\mu_1} \right)^{n_1} \left(\frac{\lambda_2}{\mu_2} \right)^{n_2}$$

$$C_{trunc} = \sum_{n \in A} \left(\frac{\lambda_1}{\mu_1} \right)^{n_1} \left(\frac{\lambda_2}{\mu_2} \right)^{n_2}$$

4 Tute 5

1. Open Jackson network with 3 queues. Each queue has a single server which serves at rate μ . Calls arrive to the first queue from outside the network according to a Poisson process at rate λ and when their service is complete, are routed according to:

$$[\gamma_{ij}] = \begin{pmatrix} 0 & 1/8 & 7/8 \\ 3/4 & 0 & 1/4 \\ 1/2 & 1/4 & 0 \end{pmatrix}$$

- (a) Solve the traffic equations to determine the average arrival rate at each queue

Note that the $[\gamma_{ij}]$ is a probability matrix.

$$\begin{aligned}
 \mathbf{y} &= \lambda + \mathbf{y}\gamma \\
 y_1 &= \lambda + \frac{3}{4}y_2 + \frac{1}{2}y_3 \\
 y_2 &= \frac{1}{8}y_1 + \frac{1}{4}y_3 \\
 y_3 &= \frac{7}{8}y_1 + \frac{1}{4}y_2
 \end{aligned}$$

Solve this system of linear equations to get

$$\begin{aligned}
 y_1 &= \frac{120}{29}\lambda \\
 y_2 &= \frac{44}{29}\lambda \\
 y_3 &= 4\lambda
 \end{aligned}$$

- (b) Write down an expression for the equilibrium distribution for the network

The invariant measures are:

$$Q_i(n_i) = \prod_{l=1}^{n_i} \frac{y_l}{\mu_i(l)}$$

But in this case all the $\mu_i(l) = \mu$

$$Q_i(n_i) = \left(\frac{y_1}{\mu}\right)^{n_i}$$

The equilibrium distribution is then (given C exists)

$$\begin{aligned} \pi(n_1, n_2, n_3) &= \left(\frac{y_1}{\mu}\right)^{n_1} \left(\frac{y_2}{\mu}\right)^{n_2} \left(\frac{y_3}{\mu}\right)^{n_3} C \\ &= \left(1 - \frac{120\lambda}{29\mu}\right) \left(\frac{120\lambda}{29\mu}\right)^{n_1} \left(1 - \frac{44\lambda}{29\mu}\right) \left(\frac{44\lambda}{29\mu}\right)^{n_2} \left(1 - \frac{4\lambda}{\mu}\right) \left(\frac{4\lambda}{\mu}\right)^{n_3} \end{aligned}$$

Now note we need

$$\frac{120\lambda}{29\mu} < 1, \quad \lambda < \frac{\mu 29}{120}, \quad \lambda < \frac{29}{44\mu}, \quad \lambda < \frac{1}{4}\mu$$

So we need

$$\lambda < \min\left\{\frac{29}{120}\mu, \frac{29}{44}\mu, \frac{1}{4}\mu\right\} = \frac{29}{120}\mu$$

- (c) If $\lambda = 9/2$ and $\mu = 10$. Does the equilibrium exist? And how stable is the network

The equilibrium won't exist as $\frac{9}{2} > \frac{29}{12}$. The network is not stable.

2. Consider a queue with poisson arrival with rate λ . Exponentially distributed service time with parameter μ . Let the residence time be the total of the waiting time and the service time

- (a) If the queue is $M/M/\infty$ what is the distribution of the residence time?

$Exp(\mu)$

- (b) If it is $M/M/N$ what is the mean waiting time? So what is the mean residence time?

$$E[W] = \frac{C(N, \frac{\lambda}{\mu})}{N\mu - \lambda}$$

$$E[\text{residence time}] = \frac{C(N, \frac{\lambda}{\mu})}{N\mu - \lambda} + \frac{1}{\mu}$$

(I.e. the expected waiting time plus the service time)

- (c) If the queue is a $M/M/1$ queue, what is the mean waiting time? So what is the mean residence time?

Using (ii)

$$E[W] = \frac{C(1, \frac{\lambda}{\mu})}{\mu - \lambda} = \frac{\lambda}{\mu(\mu - \lambda)}$$

$$E[\text{residence time}] = \frac{\lambda}{\mu(\mu - \lambda)} + \frac{1}{\mu} = \frac{1}{\mu - \lambda}$$

- (d) How does the answer to (iii) relate to the conditional expected waiting time in such a queue, given you have to wait?

Can you explain the relationship?

The conditional expected waiting time given you have to wait:

$$E[W_Q | W_Q > 0] = \frac{1}{\mu - \lambda}$$

(Using the thing from lectures)

These are the same!

$$E(\text{residence time}) = E[E[\text{residence time} | j \text{ people in the queue}]]$$

$$= \sum_{j=0}^{\infty} \pi_j (E[W_Q | \text{see } j \text{ in queue}] + \frac{1}{\mu})$$

$$= \sum_{j=0}^{\infty} \pi_j \left(\frac{j}{\mu} + \frac{1}{\mu}\right)$$

$$= \sum_{j=0}^{\infty} \pi_j \frac{j+1}{\mu}$$

$$\pi_j = (1 - \frac{\lambda}{\mu})(\frac{\lambda}{\mu})^j, \quad j \geq 0$$

$$\hat{\pi}_k = (1 - \frac{\lambda}{\mu})(\frac{\lambda}{\mu})^{k-1}, \quad k \geq 1$$

$$\pi_j = \hat{\pi}_{j+1}$$

$$\begin{aligned} E[\text{res time}] &= \sum_{j=0}^{\infty} \pi_j \left(\frac{j+1}{\mu} \right) \\ &= \sum_{j=0}^{\infty} \hat{\pi}_{j+1} \frac{j+1}{\mu} \\ &= \sum_{k=1}^{\infty} \hat{\pi}_k \frac{k}{\mu} \\ &= \sum_{k=1}^{\infty} \hat{\pi}_k E[Q_Q | j \text{ customers}] \\ &= E[W_Q | W_Q > 0] \end{aligned}$$

Basically showed that the two things are the same holy shit.

3. A hospital operates like a complex queueing system with patients moving between areas and waiting for services

- (a) Consider the hospital operates at 90% occupancy, and admits on average 100 people per day. Assuming it has 500 beds, use Little's Law to determine the average stay of a patient.

We need average queue length and entry rate. So $\bar{Q} = (90\%) * 500 = 450$

$\lambda = 100$ per day

$$\bar{Q} = \lambda \bar{W} \implies \bar{W} = \frac{\bar{Q}}{\lambda} = \frac{450}{100} = 4.5$$

- (b) Now consider the ED of the hospital. Here, the waiting time between arrival and the start of treatment is recorded. Assuming the average waiting time is 3 hours and that the ED starts to treat on average 10 new patients per hour, determine the average number of patients waiting for service. List any assumptions made.

$$\bar{W} = 3 \text{ hours} \quad \lambda = 10 \text{ people / hour} \quad \bar{Q} = \lambda \bar{W} = 3 * 10 = 30$$

- (c) Consider a single resuscitation room within the ED. This operates like a M/G/1/1 queue. Show that the probability that the resuscitation room is occupied is given by $\frac{\lambda}{\mu + \lambda}$, where λ is the arrival rate of patients requiring the resuscitation room, and $1/\mu$ is the mean time that a patient occupied the room.

$$\bar{W} = \frac{1}{\mu} \quad \bar{\lambda} = \lambda(1 - P(\text{occupied}))$$

$$\bar{Q} = 0 * (1 - P(\text{occupied})) + 1 * P(\text{occupied})$$

$$\bar{Q} = P(\text{occupied})$$

Now use little's law

$$\bar{Q} = \bar{\lambda} \bar{W}$$

$$P(\text{occupied}) = \frac{1}{\mu} (\lambda(1 - P(\text{occupied}))) \implies P(\text{occupied}) = \frac{\lambda}{\mu + \lambda}$$

Little's Law:

$$\bar{Q} = \lambda \bar{W}$$

I.e. Average queue length = entry rate times average waiting time.