

- Each indiv. dies  $\sim \exp(\mu)$
- $\Rightarrow \text{Total death rate: } \boxed{\mu_i = i\mu}$
- Such processes occur naturally in biology  
(biological reproduction, population growth & decay)

## Example 7: Queueing Systems

1) Single server exponential queueing system

2) Multiserver

" " "

Single Server Case

Lec 20 1a

Customers arrive at a single server service station according to a Poisson process with rate  $\lambda$ .

- Upon arrival, customer goes directly to service station if server is free
- If not, customer joins the "queue"  
i.e. waits in line

Successive service times are assumed to be exponential ( $\mu$ ) RVs, i.e. mean =  $\frac{1}{\mu}$

Time of next birth:

minimum of  $i$  independent  $\exp(\lambda)$  RVs

$T_{i,i+1} \sim \exp(i\lambda) \leftarrow \text{linear birth rate}$

Time of next death:

min of  $i$  indep.  $\exp(\mu)$  RVs

$T_{i,i-1} \sim \exp(i\mu) \leftarrow \text{linear death rate}$

$$\text{Thus, } \begin{cases} \lambda_i = i\lambda \\ \mu_i = i\mu \end{cases}$$

Note that  $\lambda_0 = 0 \Rightarrow$  state 0 is absorbing

Example 6: Linear Growth Model with Immigration  $\{X(t) : t \geq 0\}$

Suppose there are  $i$  individuals in the population.

- Each indiv. gives birth  $\sim \exp(\lambda)$

- In addition, there is an exponential rate of increase  $\Theta$  in the population due to an external source, such as immigration.

$\Rightarrow$  Total birth rate:  $\boxed{\lambda_i = i\lambda + \Theta}$

Let  $X(t) = \#$  of customers in the system at time  $t$ .

$\{X(t) : t \geq 0\}$  is a birth  $\notin$  death process with

$$\begin{cases} \lambda_i = \lambda & \forall i \geq 0 \\ \mu_i = \mu & \forall i \geq 1 \end{cases}$$

### Multiserver Case

- Customers arrive according to a Poisson process with rate  $\lambda$ .
- Successive service times are assumed to be indep.  $\exp(\mu)$  RVs.
- Let there be  $s$  servers available, and  
 $X(t) = \#$  of customers in system at time  $t$ .

$\{X(t) : t \geq 0\}$  is a B  $\notin$  D process with

$$\begin{cases} \lambda_i = \lambda & \forall i \geq 0 \\ \mu_i = \begin{cases} i\mu & \text{if } 1 \leq i \leq s \\ s\mu & \text{if } i > s \end{cases} \end{cases}$$

[Example: Bank with  $s$  tellers]

- Suppose  $i$  customers in the system &  $i \leq s$ ,  
then  $i$  servers are busy.

→ Each server works at rate  $\mu$  so  
total departure rate =  $i\mu$

- Suppose  $i$  customers in system and  $i > s$ ,  
then all  $s$  servers are busy.

→ Total departure rate =  $s\mu$

\* These queuing systems are also known as

M/M/1 queuing system ← single server case

M/M/s queuing system ← multiserver

- First 'M' refers to Markovian arrival process  
(since it is a Poisson process)
- Second 'M' refers to the exponential service dist'n  
(∴ hence Markovian)
- '1' or 's' denotes single or multiserver case

There is also an  $M/M/\infty$  queuing system.

Q. How to interpret this?

' $\infty$ ' stands for infinitely many servers

→ there is never a queue so customers can immediately enter service

- assuming we start with a finite # of customers in the system

Q. Absurd model or useful ??

Seems absurd, but actually useful for studying telephone traffic (for instance).

↳ To determine the # of phone lines (land lines) that are needed for UNR employees to call off-campus, we assume there are always enough ( $M/M/\infty$  queue), find the stationary distribution & use this to estimate how many lines are needed s.t. 99% (or whatever accuracy you'd like!) of time there is enough capacity.

### Transition Rate Matrix $Q$

In the  $M/M/\infty$  queue,

$X(t) = \# \text{ of customers in the system at time } t$

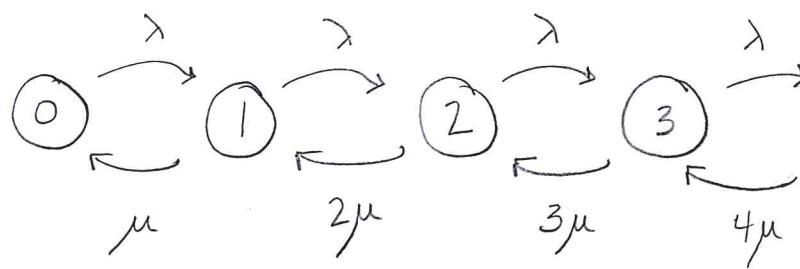
State Space  $S = \{0, 1, 2, \dots\}$

jump rates

Matrix  $Q$  has entries  $q_{ij} = v_i P_{ij}$  for  $i \neq j$

$$Q = \begin{matrix} & 0 & 1 & 2 & 3 & 4 & \dots \\ 0 & -\lambda & \lambda & & & & \\ 1 & \mu - (\mu + \lambda) & \lambda & & & & \\ 2 & & 2\mu - (2\mu + \lambda) & \lambda & & & \\ 3 & & & 3\mu - (3\mu + \lambda) & \lambda & & \\ \vdots & & & & \ddots & & \end{matrix}$$

check:  
Rows sum  
to 0 ✓



$i \rightarrow i+1$   
Poisson Process w/rate λ  
(arrivals)

Exponential ( $\mu$ ) service  
times, i.e.,  
 $i \rightarrow i-1$  transitions  
occur at rate  
 $i\mu$

In other words, the entries in  $Q$  are

$$Q_{ij} = \begin{cases} q_{ij} = v_i P_{ij} & \text{if } i \neq j \\ -\sum_{j \neq i} q_{ij} = -v_i & \text{if } i = j \end{cases}$$

so that the  
rows of  $Q$   
sum to 0

$$P_{ij} = \begin{cases} 1 & \text{if } j = i+1 \text{ or } j = i-1 \\ 0 & \text{o.w.} \end{cases}$$

$$q_{01} = v_0 P_{01} = \lambda \cdot 1 = \lambda$$

$$q_{10} = v_1 P_{10} = \mu \cdot 1 = \mu$$

In general  $q_{i,i+1} = v_i P_{i,i+1} = \lambda \cdot 1 = \lambda$

$$q_{i,i-1} = v_i P_{i,i-1} = i\mu \cdot 1 = i\mu$$

CTMC Constructions (skipped in Lecture)

- ①  $\{X(t) : t \geq 0\}$  is a CTMC if it can be constructed from an embedded MC  $\{X_n\}$  with transition (aka discrete time) matrix  $P$  with the duration of a visit to state  $i$  having an exponential ( $v_i$ ) distribution.
- ② When  $X(t)$  arrives at state  $i$ , generate i.i.d. exponential RVs  $Y_j \sim \exp(q_{ij})$  where  $q_{ij} = v_i P_{ij}$  for  $j \neq i$ .  
Choose the next state to be  $k = \arg \min_j Y_j$  and the time until the transition to be  $\min_j Y_j$ .  
(i.e. duration spent in  $i$ )  
Note that  $\min_j Y_j \sim \exp(v_i)$

Example: A household consists of 4 individuals, some are healthy & some are infected with the flu. Suppose that infected individuals get well at rate  $\mu$ , while each pair of indivs. has encounters at rate  $\lambda$  & the disease is transmitted if 1 is infected & the other is susceptible. (a) Formulate a continuous-time MC model for this system. (b) Simulate & estimate how long it might take until all members are infected for the case  $\mu=0$ .

$$(a) \quad S = \{0, 1, 2, 3, 4\}$$

$X(t)$  = # of individuals infected at time  $t$

$$Q = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \left[ \begin{matrix} 0 & 0 & 0 & 0 & 0 \\ \mu & -(\mu+3\lambda) & 3\lambda & & \\ 0 & 2\mu & -(2\mu+4\lambda) & 4\lambda & \\ 0 & & 3\mu & -(3\mu+3\lambda) & 3\lambda \\ 0 & & & 4\mu & -4\mu \end{matrix} \right] \end{matrix}$$

↖ infections  
↖ diag makes rows  $\sum$  to 0

Assuming  
 $X(0) \geq 1$

1 infected  $\Rightarrow \binom{3}{2} = 3$  pairs  
3 suscep. of indivs. encounter at rate  $\lambda$

↖ recovery, linear

