



Prop: Assume that  $\pi$  satisfies the detailed balance equation.

Then  $\pi_i p_t(i,j) = \pi_j p_t(j,i)$  for all  $i, j$  and  $t$ .

Proof (almost): Detailed balance equation

$$\pi_i q_{ij} = \pi_j q_{ji} \text{ for all } i, j.$$

By Kolmogorov's BE:

$$p_t'(i,j) = \sum_{k \neq i} q_{ik} p_t(k,j) - \lambda_i p_t(i,j) \quad | \quad \pi_i$$

$$\pi_i p_t'(i,j) = \sum_{k \neq i} \pi_i q_{ik} p_t(k,j) - \lambda_i \pi_i p_t(i,j)$$

$$\pi_i p_t'(i,j) = \sum_{k \neq i} \pi_k q_{ki} p_t(k,j) - \lambda_i \pi_i p_t(i,j)$$

$$\rightarrow y_t'(i,j) = \sum_{k \neq i} q_{ki} y_t(k,j) - \lambda_i y_t(i,j)$$

$$\boxed{y_t' = Q^T y_t}$$

Notation:

$$y_t(i,j) = \pi_i p_t(i,j)$$

By Kolmogorov's FE:

$$p_t'(j,i) = \sum_{k \neq j} p_t(j,k) q_{ki} - \lambda_j p_t(j,i) \quad | \quad \pi_j$$

$$\pi_j p_t'(j,i) = \sum_{k \neq j} \pi_j p_t(j,k) q_{ki} - \lambda_j \pi_j p_t(j,i)$$

$$z_t'(i,j) = \sum_{k \neq i} z_t(k,j) q_{ki} - \lambda_i z_t(i,j)$$

$$\boxed{z_t' = Q^T z_t}$$

$$\boxed{z_t(i,j) = \pi_j p_t(j,i)}$$

Hence  $y_t$  and  $z_t$  satisfy the D.E. with the same initial

conditions  $y_0 = \begin{pmatrix} \pi_0 \\ \pi_1 \\ \pi_2 \\ \vdots \end{pmatrix} = z_0$

Since the equation has a unique solution, we have  $y_t = z_t$  for all

Hence

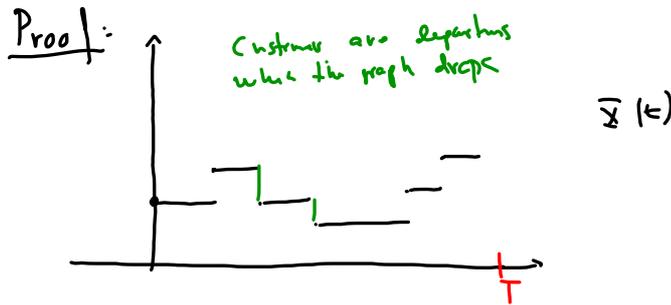
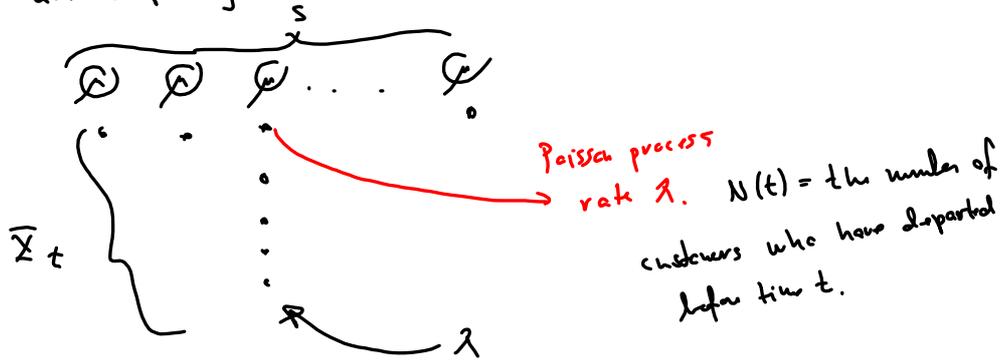
$$\pi_i p_t(i,j) = y_t(i,j) = z_t(i,j) = \pi_j p_t(j,i).$$

Corollary: If  $\pi$  satisfies the detailed balance equation, then the backward process has the same trans. prob. as the forward process: i.e.

$$\hat{p}_t(i,j) = p_t(i,j)$$

Proof:  $\hat{p}_t(i,j) = \frac{\pi_j p_t(j,i)}{\pi_i} \stackrel{DB}{=} \frac{\pi_i p_t(i,j)}{\pi_i} = p_t(i,j)$

Theorem: Consider a M/M/1-queue with  $\lambda < \mu$  and assume that it is started with its stationary distribution. Then customers are departing from the system like a Poisson process with rate  $\lambda$ .

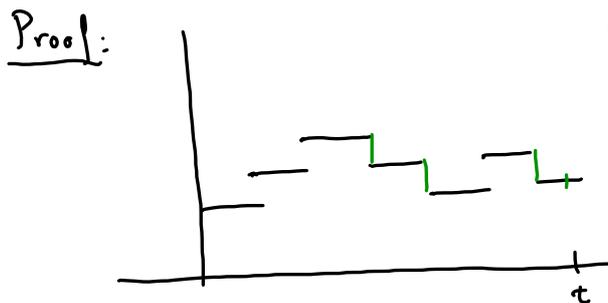


Customers leave when the backward process increases  
 = as often as the forward process increases  
 = according to Poisson rate  $\lambda$ .

Observation: The argument still goes through even when the servers are not Markovian, provided a stationary distribution exists.

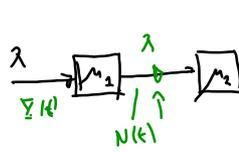
Theorem:  $X(t)$  and  $N(t)$  are independent

↑ customers currently in queue  
 ↑ customer who has left



Look at the backward process. Then  $N(t)$  is counting how many customers that will arrive in the future, and that is independent of  $X_t$  since customer will arrive at rate  $\lambda$  independent of what  $X_t$  is.

Systems of queues (sect. 4.6)

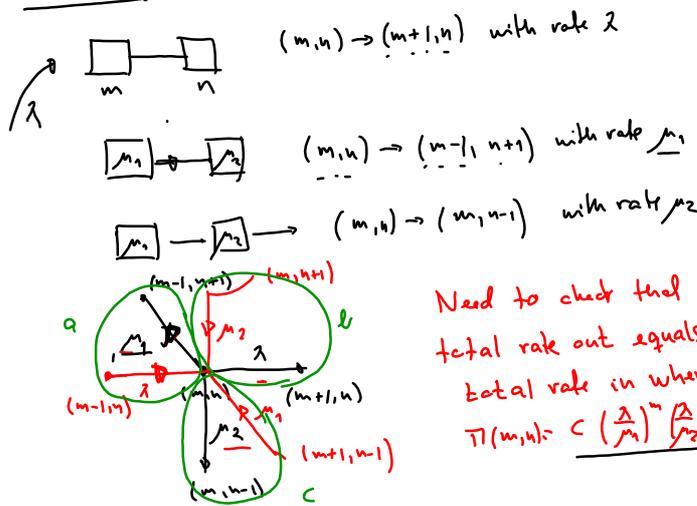


equilibrium distribution  
Need  $\lambda < \mu_1$   
First and second waits  
independently.

Guess for stationary distribution: Server 1:  $\pi_1(m) = \frac{\mu_2 - \lambda}{\lambda} \left(\frac{\lambda}{\mu_1}\right)^m$   
Server 2:  $\pi_2(n) = \frac{\mu_2 - \lambda}{\lambda} \left(\frac{\lambda}{\mu_2}\right)^n$

$$\pi(m, n) = \frac{\mu_1 - \lambda}{\lambda} \left(\frac{\lambda}{\mu_1}\right)^m \frac{\mu_2 - \lambda}{\lambda} \left(\frac{\lambda}{\mu_2}\right)^n = c \left(\frac{\lambda}{\mu_1}\right)^m \left(\frac{\lambda}{\mu_2}\right)^n$$

Want to check that this really is a stationary distribution:  
kind of transition:



Need to check that  
total rate out equals  
total rate in when  
 $\pi(m, n) = c \left(\frac{\lambda}{\mu_1}\right)^m \left(\frac{\lambda}{\mu_2}\right)^n$

Look at a: Out:  $\mu_1 \pi(m, n) = \mu_1 c \left(\frac{\lambda}{\mu_1}\right)^m \left(\frac{\lambda}{\mu_2}\right)^n$   
In:  $\lambda \pi(m-1, n) = \lambda c \left(\frac{\lambda}{\mu_1}\right)^{m-1} \left(\frac{\lambda}{\mu_2}\right)^n$   
 $= c \left(\frac{\lambda}{\mu_1}\right)^m \left(\frac{\lambda}{\mu_2}\right)^n$

Look at b: Out:  $\lambda c \left(\frac{\lambda}{\mu_1}\right)^m \left(\frac{\lambda}{\mu_2}\right)^n$   
In:  $\mu_2 c \left(\frac{\lambda}{\mu_1}\right)^m \left(\frac{\lambda}{\mu_2}\right)^{n+1} = \lambda c \left(\frac{\lambda}{\mu_1}\right)^m \left(\frac{\lambda}{\mu_2}\right)^n$

Look at c: Out:  $\mu_2 c \left(\frac{\lambda}{\mu_1}\right)^m \left(\frac{\lambda}{\mu_2}\right)^n$   
In:  $\mu_1 c \left(\frac{\lambda}{\mu_1}\right)^{m+1} \left(\frac{\lambda}{\mu_2}\right)^{n-1} = \lambda c \left(\frac{\lambda}{\mu_1}\right)^m \left(\frac{\lambda}{\mu_2}\right)^n$

