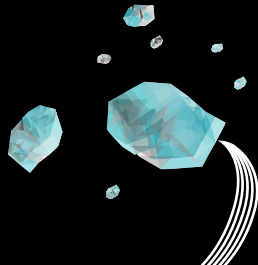
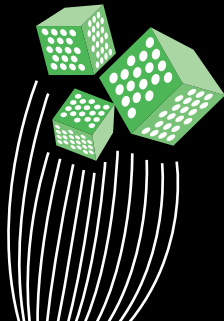


# Markovian Queues and Stochastic Networks

Lecture 2

Richard J. Boucherie

Stochastic Operations Research



## Detailed balance – 1

---

### Definition (2.2.1 Detailed balance)

A Markov chain  $\{N(t)\}$  at state space  $S$  with transition rates  $q(\mathbf{n}, \mathbf{n}')$ ,  $\mathbf{n}, \mathbf{n}' \in S$ , satisfies detailed balance if a distribution  $\pi = (\pi(\mathbf{n}), \mathbf{n} \in S)$  exists that satisfies for all  $\mathbf{n}, \mathbf{n}' \in S$  the **detailed balance equations**:

$$\pi(\mathbf{n})q(\mathbf{n}, \mathbf{n}') - \pi(\mathbf{n}')q(\mathbf{n}', \mathbf{n}) = 0.$$

### Theorem (2.2.2)

*If the distribution  $\pi$  satisfies the detailed balance equations then  $\pi$  is the equilibrium distribution.*

- ▶ The detailed balance equations state that the probability flow between each pair of states is balanced.

## Detailed balance – 2

---

Lemma (2.2.3, 2.2.4 Kolmogorov's criterion)

$\{N(t)\}$  satisfies detailed balance if and only if for all  $r \in \mathbb{N}$  and any finite sequence of states  $\mathbf{n}_1, \mathbf{n}_2, \dots, \mathbf{n}_r \in \mathcal{S}$ ,  $\mathbf{n}_r = \mathbf{n}_1$ ,

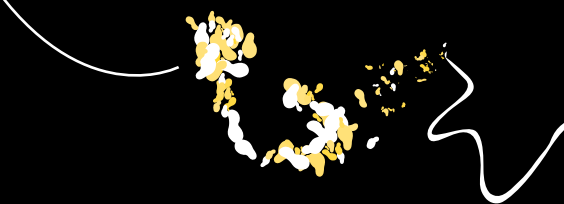
$$\prod_{i=1}^{r-1} q(\mathbf{n}_i, \mathbf{n}_{i+1}) = \prod_{i=1}^{r-1} q(\mathbf{n}_{r-i+1}, \mathbf{n}_{r-i}).$$

If  $\{N(t)\}$  satisfies detailed balance, then

$$\pi(\mathbf{n}) = \pi(\mathbf{n}') \frac{q(\mathbf{n}_1, \mathbf{n}_2)q(\mathbf{n}_2, \mathbf{n}_3) \dots q(\mathbf{n}_{r-1}, \mathbf{n}_r)}{q(\mathbf{n}_2, \mathbf{n}_1)q(\mathbf{n}_3, \mathbf{n}_2) \dots q(\mathbf{n}_r, \mathbf{n}_{r-1})},$$

for arbitrary  $\mathbf{n}' \in \mathcal{S}$  for all  $r \in \mathbb{N}$  and any path  $\mathbf{n}_1, \mathbf{n}_2, \dots, \mathbf{n}_r \in \mathcal{S}$  such that  $\mathbf{n}_1 = \mathbf{n}'$ ,  $\mathbf{n}_r = \mathbf{n}$  for which the denominator is positive.

- Direct generalisation of the result for birth-death process.

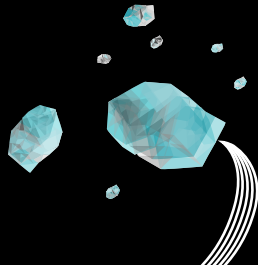
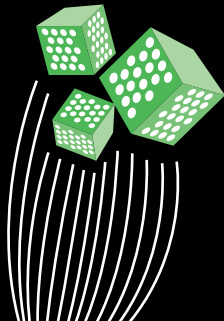


# Markovian Queues and Stochastic Networks

Lecture 2

Richard J. Boucherie

Stochastic Operations Research



# Reversibility – 1



## Definition (Stationary process)

A stochastic process  $\{N(t), t \in \mathbb{R}\}$  is **stationary** if  $(N(t_1), N(t_2), \dots, N(t_k))$  has the same distribution as  $(N(t_1 + \tau), N(t_2 + \tau), \dots, N(t_k + \tau))$  for all  $k \in \mathbb{N}$ ,  $t_1, t_2, \dots, t_k \in T$ ,  $\tau \in T$

## Definition (2.4.1 Reversibility)

A stochastic process  $\{N(t), t \in \mathbb{R}\}$  is **reversible** if  $(N(t_1), N(t_2), \dots, N(t_k))$  has the same distribution as  $(N(\tau - t_1), N(\tau - t_2), \dots, N(\tau - t_k))$  for all  $k \in \mathbb{N}$ ,  $t_1, t_2, \dots, t_k \in \mathbb{R}$ ,  $\tau \in \mathbb{R}$ .

## Theorem (2.4.2)

*If  $\{N(t)\}$  is reversible then  $\{N(t)\}$  is stationary.*

## Reversibility – 2

---

### Theorem (2.4.3 Reversibility and detailed balance)

*Let  $\{N(t), t \in \mathbb{R}\}$  be a stationary Markov chain with transition rates  $q(\mathbf{n}, \mathbf{n}')$ ,  $\mathbf{n}, \mathbf{n}' \in S$ .  $\{N(t)\}$  is reversible if and only if there exists a distribution  $\pi = (\pi(\mathbf{n}), \mathbf{n} \in S)$  that satisfies the detailed balance equations. When there exists such a distribution  $\pi$ , then  $\pi$  is the equilibrium distribution of  $\{N(t)\}$ .*

## Example: Departures from the $M|M|1$ queue

---

- ▶ Arrival process to the  $M|M|1$  queue is a Poisson process with rate  $\lambda$ .
- ▶ If  $\lambda < \mu$  departure process from  $M|M|1$  queue has rate  $\lambda$ .
- ▶  $\{N(t)\}$  recording the number of customers in  $M|M|1$  with arrival rate  $\lambda$  and service rate  $\mu$  satisfies detailed balance.
- ▶ Markov chain  $\{N^r(t)\}$  in reversed time has Poisson arrivals at rate  $\lambda$  and service rate  $\mu$ .
- ▶ Therefore  $\{N^r(t)\}$  is the Markov chain of an  $M|M|1$  queue with Poisson arrivals at rate  $\lambda$  and negative-exponential service at rate  $\mu$ .
- ▶ Epochs of the arrival process for the reversed queue coincide with the epochs of the arrival process for the original queue, it must be that the departure process from the  $M|M|1$  queue is a Poisson *process* with rate  $\lambda$ .

## Reversibility – 3

---

### Theorem (2.4.3 Reversibility and detailed balance)

*Let  $\{N(t), t \in \mathbb{R}\}$  be a stationary Markov chain with transition rates  $q(\mathbf{n}, \mathbf{n}')$ ,  $\mathbf{n}, \mathbf{n}' \in S$ .  $\{N(t)\}$  is reversible if and only if there exists a distribution  $\pi = (\pi(\mathbf{n}), \mathbf{n} \in S)$  that satisfies the detailed balance equations. When there exists such a distribution  $\pi$ , then  $\pi$  is the equilibrium distribution of  $\{N(t)\}$ .*

**Proof.** If  $\{N(t)\}$  is reversible, then for all  $t, h \in \mathbb{R}$ ,  $\mathbf{n}, \mathbf{n}' \in S$ :

$$\mathbb{P}(N(t+h) = \mathbf{n}', N(t) = \mathbf{n}) = \mathbb{P}(N(t) = \mathbf{n}', N(t+h) = \mathbf{n}).$$

$\{N(t), t \in \mathbb{R}\}$  is stationary. Let  $\pi(\mathbf{n}) = \mathbb{P}(N(t) = \mathbf{n}), t \in \mathbb{R}$ .

$$\frac{\mathbb{P}(N(t+h) = \mathbf{n}' | N(t) = \mathbf{n})}{h} \pi(\mathbf{n}) = \frac{\mathbb{P}(N(t+h) = \mathbf{n} | N(t) = \mathbf{n}')}{h} \pi(\mathbf{n}')$$

Letting  $h \rightarrow 0$  yields the detailed balance equations.



## Proof continued

---

Assume  $\pi = (\pi(\mathbf{n}), \mathbf{n} \in S)$  satisfies detailed balance.

Consider  $\{N(t)\}$  for  $t \in [-H, H]$ . Suppose  $\{N(t)\}$  moves along the sequence of states  $\mathbf{n}_1, \dots, \mathbf{n}_k$  and has sojourn time  $h_i$  in  $\mathbf{n}_i$ ,  $i = 1, \dots, k-1$ , and remains in  $\mathbf{n}_k$  for at least  $h_k$  until time  $H$ .

With probability  $\pi(\mathbf{n}_1) = \mathbb{P}(N(-H) = \mathbf{n}_1)$   $\{N(t)\}$  starts in  $\mathbf{n}_1$ .

Probability density with respect to  $h_1, \dots, h_k$  for this sequence

$$\pi(\mathbf{n}_1)q(\mathbf{n}_1)e^{-q(\mathbf{n}_1)h_1}p(\mathbf{n}_1, \mathbf{n}_2) \cdots q(\mathbf{n}_{k-1})e^{-q(\mathbf{n}_{k-1})h_{k-1}}p(\mathbf{n}_{k-1}, \mathbf{n}_k)e^{-q(\mathbf{n}_k)h_k},$$

Kolmogorov's criterion implies that

$$\pi(\mathbf{n}_1)q(\mathbf{n}_1, \mathbf{n}_2) \cdots q(\mathbf{n}_{k-1}, \mathbf{n}_k) = \pi(\mathbf{n}_k)q(\mathbf{n}_k, \mathbf{n}_{k-1}) \cdots q(\mathbf{n}_2, \mathbf{n}_1),$$

probability density equals the probability density for the reversed path that starts in  $\mathbf{n}_k$  at time  $H$ .

Thus  $(N(t_1), N(t_2), \dots, N(t_k)) \sim (N(-t_1), N(t_2), \dots, N(-t_k))$

Stationarity completes the proof,.

# Burke's theorem and feedforward networks – 1

---

## Theorem (2.5.1 Burke's theorem)

*Let  $\{N(t)\}$  record the number of customers in the  $M|M|1$  queue with arrival rate  $\lambda$  and service rate  $\mu$ ,  $\lambda < \mu$ . Let  $\{D(t)\}$  record the customers' departure process from the queue. In equilibrium the departure process  $\{D(t)\}$  is a Poisson process with rate  $\lambda$ , and  $N(t)$  is independent of  $\{D(s), s < t\}$ .*

**Proof.**  $M|M|1$  reversible: epochs at which  $\{N(-t)\}$  jumps up form Poisson process with rate  $\lambda$ .

If  $\{N(-t)\}$  jumps up at time  $t^*$  then  $\{N(t)\}$  jumps down at  $t^*$ .

Departure process forms a Poisson process with rate  $\lambda$ .

$\{N(t)\}$  reversible: departure process up to  $t^*$  and  $N(t^*)$  have same distribution as arrival process after  $-t^*$  and  $N(-t^*)$ .

Arrival process is Poisson process: arrival process after  $-t^*$  independent of  $N(-t^*)$ .

Hence, the departure process up to  $t^*$  independent of  $N(t^*)$ .

## Burke's theorem and feedforward networks – 2



- ▶ **Tandem network** of two  $M|M|1$  queues
- ▶ Poisson  $\lambda$  arrival process to queue 1, service rates  $\mu_j$ .
- ▶ Provided  $\rho_j = \lambda/\mu_j < 1$ , marginal distributions  $\pi_j(n_j) = (1 - \rho_j)\rho_j^{n_j}$ ,  $n_j \in \mathbb{N}_0$ .
- ▶ Burke's theorem: departure process from queue 1 before  $t^*$  and  $N_1(t^*)$ , are independent.
- ▶ Hence, in equilibrium, the at time  $t^*$  the random variables  $N_1(t^*)$  and  $N_2(t^*)$  are independent:

$$\pi(\mathbf{n}) = \prod_{i=1}^2 \pi_i(n_i), \quad \mathbf{n} \in \mathcal{S} = \mathbb{N}_0^2.$$

## Burke's theorem and **feedforward networks** – 3

---

- ▶ Customer leaving queue  $j$  can route to any of the queues  $j + 1, \dots, J$ , or may leave the network.
- ▶  $p_{ij}$  fraction of customers from queue  $i$  to queue  $j > i$ ,  $p_{i0}$  fraction leaving the network.
- ▶ Arrival process is Poisson process with rate  $\mu_0$ .
- ▶ Fraction  $p_{0j}$  of these customers is routed to queue  $j$ .
- ▶ The service rate at queue  $j$  is  $\mu_j$ .
- ▶ Burke's theorem implies that all flows of customers among the queues are Poisson flows.
- ▶ Arrival rate  $\lambda_j$  of customers to queue  $j$  is obtained from superposition and random splitting of Poisson processes:

$$\lambda_j = \mu_0 p_{0j} + \sum_{i=1}^{j-1} \lambda_i p_{ij}, \quad j = 1, \dots, J,$$

- ▶ **traffic equations**: the mean flow of customers.

## Burke's theorem and feedforward networks – 4

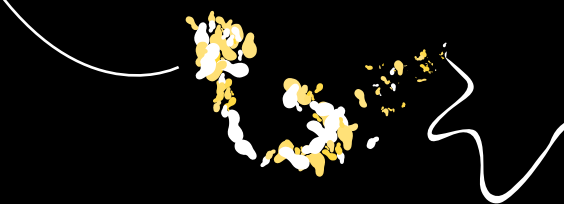
---

Theorem (2.5.4 Equilibrium distribution)

Let  $\{N(t) = (N_1(t), \dots, N_J(t))\}$  at state space  $S = \mathbb{N}_0^J$ , where  $\mathbf{n} = (n_1, \dots, n_J)$  and  $n_j$  the number of customers in queue  $j$ ,  $j = 1, \dots, J$ , record the number of customers in the feedforward network of  $J$   $M|M|1$  queues described above. If  $\rho_j = \lambda_j/\mu_j < 1$ , with  $\lambda_j$  the solution of the traffic equations,  $j = 1, \dots, J$ , then the equilibrium distribution is the product of the marginal distributions of the queues:

$$\pi(\mathbf{n}) = \prod_{j=1}^J (1 - \rho_j) \rho_j^{n_j}, \quad n_j \in \mathbb{N}_0, \quad j = 1, \dots, J. \quad (1)$$

► Next: networks of  $M|M|1$  queues.

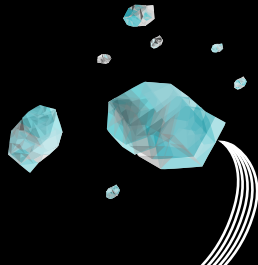
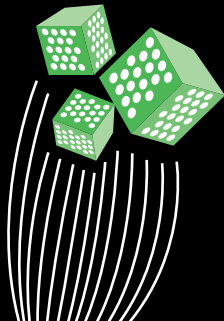


# Markovian Queues and Stochastic Networks

Lecture 2

Richard J. Boucherie

Stochastic Operations Research



## Open network of $M|M|1$ queues – 1

---

- ▶ Customer leaving queue  $j$  can route to any of the queues  $1, \dots, J$ , or may leave the network.
- ▶  $p_{ij}$  fraction of customers from queue  $i$  to queue  $j$ ,  
 $p_{i0}$  fraction leaving the network.
- ▶ Arrival process is Poisson process with rate  $\mu_0$ .
- ▶ Fraction  $p_{0j}$  of these customers is routed to queue  $j$ .
- ▶ The service rate at queue  $j$  is  $\mu_j$ .
- ▶ Arrival rate  $\lambda_j$  of customers to queue  $j$  is obtained from the **traffic equations**

$$\lambda_j = \mu_0 p_{0j} + \sum_{i=1}^J \lambda_i p_{ij}, \quad j = 1, \dots, J,$$

## Open network of $M|M|1$ queues – 2

---

- ▶ Evolution number of customers in the queues recorded by Markov chain  $\{N(t) = (N_1(t), \dots, N_J(t)), t \in \mathbb{R}\}$
- ▶ State space  $S \subseteq \mathbb{N}_0^J$ , states  $\mathbf{n} = (n_1, \dots, n_J)$ .
- ▶ If  $\{N(t)\}$  is in state  $\mathbf{n}$  and a customer routes from queue  $i$  to queue  $j$  then the next state is  $\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j$ ,  $i, j = 0, \dots, J$ .
- ▶ Queue 0 is introduced to represent the outside.
- ▶ If a customer routes from queue  $i$  to queue 0 then this customer leaves the network
- ▶ and if a customer routes from queue 0 to queue  $j$  then this customer enters the network at queue  $j$ ,  $j = 1, \dots, J$ .
- ▶ State space  $S = \mathbb{N}_0^J$ .
- ▶ The transition rates of  $\{N(t)\}$  for an open network are, for  $\mathbf{n} \neq \mathbf{n}'$ ,  $\mathbf{n}, \mathbf{n}' \in S$ ,

$$q(\mathbf{n}, \mathbf{n}') = \begin{cases} \mu_i p_{ij}, & \text{if } \mathbf{n}' = \mathbf{n} - \mathbf{e}_i + \mathbf{e}_j, i, j = 0, \dots, J, \\ 0, & \text{otherwise.} \end{cases}$$



## Closed network of $M|M|1$ queues – 1

---

- ▶ Queueing network is **closed** if arrivals to the network and departures from the network are not possible.
- ▶ Closed network by setting  $\mu_0 = 0$  and  $p_{j0} = 0, j = 1, \dots, J$ .
- ▶ Number of customers in a closed network is constant:  
 $S = S_M = \{\mathbf{n} : \sum_{j=1}^J n_j = M\}$  for some  $M$ , the number of customers in the network.
- ▶ The transition rates of  $\{N(t)\}$  for a closed network are, for  $\mathbf{n} \neq \mathbf{n}', \mathbf{n}, \mathbf{n}' \in S$ ,

$$q(\mathbf{n}, \mathbf{n}') = \begin{cases} \mu_i p_{ij}, & \text{if } \mathbf{n}' = \mathbf{n} - \mathbf{e}_i + \mathbf{e}_j, i, j = 1, \dots, J, \\ 0, & \text{otherwise.} \end{cases}$$

## Open network of $M|M|1$ queues – 3

---

Lemma (3.1.1 Traffic equations: open network)

*Consider an open network. Assume that the routing matrix  $P = (p_{ij}, i, j = 0, \dots, J)$  is irreducible. The traffic equations*

$$\lambda_j = \mu_0 p_{0j} + \sum_{i=1}^J \lambda_i p_{ij}, \quad j = 1, \dots, J,$$

*have a unique non-negative solution  $\{\lambda_j, j = 1, \dots, J\}$ .*

**Proof.** Let  $\lambda_0 = \mu_0$ . Observe that the traffic equations also imply a traffic equation for queue 0:  $\mu_0 = \sum_{j=1}^J \lambda_j p_{j0}$ . Then the traffic equations for the open network read

$$\sum_{i=0}^J \lambda_i p_{ji} = \sum_{i=0}^J \lambda_i p_{ij}, \quad j = 0, \dots, J.$$

## Open network of $M|M|1$ queues – 4

---

Theorem (3.1.4 Equilibrium distribution)

Consider the Markov chain  $\{N(t)\}$  at state space  $S = \mathbb{N}_0^J$  for the open network of  $M|M|1$  queues. Assume the routing matrix  $P = (p_{ij})$  is irreducible and let  $\{\lambda_j\}$  be the unique solution of the traffic equations. If  $\rho_j := \lambda_j/\mu_j < 1$ ,  $j = 1, \dots, J$ , then  $\{N(t)\}$  has unique **product-form** equilibrium distribution

$$\pi(\mathbf{n}) = \prod_{j=1}^J (1 - \rho_j) \rho_j^{n_j} = \prod_{j=1}^J \pi_j(n_j), \quad \mathbf{n} \in S.$$

Moreover, the equilibrium distribution satisfies **partial balance**, for all  $\mathbf{n} \in S$ ,  $i = 0, \dots, J$ ,

$$\sum_{j=0}^J \{ \pi(\mathbf{n}) q(\mathbf{n}, \mathbf{n} - \mathbf{e}_i + \mathbf{e}_j) - \pi(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j) q(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j, \mathbf{n}) \} = 0.$$

## Proof of Theorem 3.1.4

---

$$\begin{aligned}
 & \sum_{j=0}^J \{m(\mathbf{n})q(\mathbf{n}, \mathbf{n} - \mathbf{e}_i + \mathbf{e}_j) - m(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j)q(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j, \mathbf{n})\} \\
 &= \sum_{j=0}^J \left\{ \prod_{k=1}^J \rho_k^{n_k} \mu_i p_{ij} \mathbb{1}(\mathbf{n} - \mathbf{e}_i \in \mathbb{N}_0^J) - \prod_{k=1}^J \rho_k^{n_k - \delta_{ki} + \delta_{kj}} \mu_j p_{ji} \mathbb{1}(\mathbf{n} - \mathbf{e}_i \in \mathbb{N}_0^J) \right\} \\
 & \sum_{j=0}^J \{m(\mathbf{n})q(\mathbf{n}, \mathbf{n} - \mathbf{e}_i + \mathbf{e}_j) - m(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j)q(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j, \mathbf{n})\} \mathbb{1}(i = 0) \\
 &= \left\{ \mu_0 - \sum_{j=1}^J \lambda_j p_{j0} \right\} \prod_{k=1}^J \rho_k^{n_k} \mathbb{1}(\mathbf{n} \in \mathbb{N}_0^J) = 0, \\
 & \sum_{j=0}^J \{m(\mathbf{n})q(\mathbf{n}, \mathbf{n} - \mathbf{e}_i + \mathbf{e}_j) - m(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j)q(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j, \mathbf{n})\} \mathbb{1}(i \neq 0) \\
 &= \left\{ \sum_{j=0}^J \lambda_j p_{ij} - \mu_0 p_{0i} - \sum_{j=1}^J \lambda_j p_{ji} \right\} \prod_{k=1}^J \rho_k^{n_k - \delta_{ki}} \mathbb{1}(\mathbf{n} - \mathbf{e}_i \in \mathbb{N}_0^J) \mathbb{1}(i \neq 0) = 0.
 \end{aligned}$$

# Traffic equations

---

$$\sum_{j=0}^J \{m(\mathbf{n})q(\mathbf{n}, \mathbf{n} - \mathbf{e}_i + \mathbf{e}_j) - m(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j)q(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j, \mathbf{n})\}$$

$$= \sum_{j=0}^J \left\{ \prod_{k=1}^J \rho_k^{n_k} \mu_i p_{ij} \mathbb{1}(\mathbf{n} - \mathbf{e}_i \in \mathbb{N}_0^J) - \prod_{k=1}^J \rho_k^{n_k - \delta_{ki} + \delta_{kj}} \mu_j p_{ji} \mathbb{1}(\mathbf{n} - \mathbf{e}_i \in \mathbb{N}_0^J) \right\}$$

$$\sum_{j=0}^J \{m(\mathbf{n})q(\mathbf{n}, \mathbf{n} - \mathbf{e}_i + \mathbf{e}_j) - m(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j)q(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j, \mathbf{n})\} \mathbb{1}(i = 0)$$

$$= \left\{ \mu_0 - \sum_{j=1}^J \lambda_j \rho_{j0} \right\} \prod_{k=1}^J \rho_k^{n_k} \mathbb{1}(\mathbf{n} \in \mathbb{N}_0^J) = 0,$$

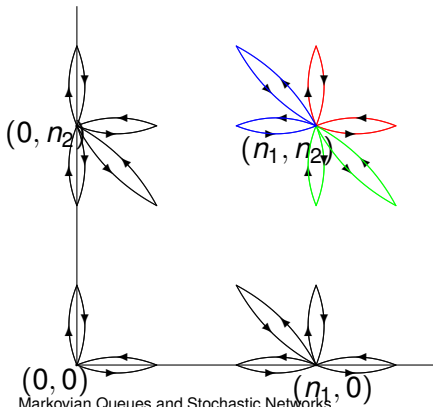
$$\sum_{j=0}^J \{m(\mathbf{n})q(\mathbf{n}, \mathbf{n} - \mathbf{e}_i + \mathbf{e}_j) - m(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j)q(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j, \mathbf{n})\} \mathbb{1}(i \neq 0)$$

$$= \left\{ \lambda_i - \mu_0 \rho_{0i} - \sum_{j=1}^J \lambda_j \rho_{ji} \right\} \prod_{k=1}^J \rho_k^{n_k - \delta_{ki}} \mathbb{1}(\mathbf{n} - \mathbf{e}_i \in \mathbb{N}_0^J) \mathbb{1}(i \neq 0) = 0.$$

# Partial balance -1

Moreover, the equilibrium distribution satisfies **partial balance**,  
for all  $\mathbf{n} \in \mathcal{S}$ ,  $i = 0, \dots, J$ ,

$$\sum_{j=0}^J \{ \pi(\mathbf{n})q(\mathbf{n}, \mathbf{n} - \mathbf{e}_i + \mathbf{e}_j) - \pi(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j)q(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j, \mathbf{n}) \} = 0.$$



## Closed network of $M|M|1$ queues – 2

Theorem (3.1.5 Equilibrium distribution)

Consider Markov chain  $\{N(t)\}$  at state space

$S = S_M = \{\mathbf{n} : \sum_{j=1}^J n_j = M\}$  for the closed network of  $M|M|1$  queues containing  $M$  customers. Assume  $P = (p_{ij})$  is irreducible and let  $\{\lambda_j\}$  be the unique solution of the traffic equations such that  $\sum_{j=1}^J \lambda_j = 1$ . Let  $\rho_j := \lambda_j / \mu_j$ . Then  $\{N(t)\}$  has unique **product-form** equilibrium distribution

$$\pi(\mathbf{n}) = G_M \prod_{j=1}^J \rho_j^{n_j}, \quad \mathbf{n} \in S, \quad G_M = \left[ \sum_{\mathbf{n} \in S} \prod_{j=1}^J \rho_j^{n_j} \right]^{-1}.$$

Moreover, the equilibrium distribution satisfies **partial balance**, for all  $\mathbf{n} \in S$ ,  $i = 1, \dots, J$ ,

$$\sum_{j=1}^J \{ \pi(\mathbf{n}) q(\mathbf{n}, \mathbf{n} - \mathbf{e}_i + \mathbf{e}_j) - \pi(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j) q(\mathbf{n} - \mathbf{e}_i + \mathbf{e}_j, \mathbf{n}) \} = 0.$$

## Closed network of $M|M|1$ queues – 3

---

Algorithm (3.1.8 Buzen's Algorithm)

Define  $G(m, j)$ ,  $m = 0, \dots, M$ ,  $j = 1, \dots, J$ . Set

$$\begin{aligned}G(0, j) &= 1, \quad j = 1, \dots, J, \\G(m, 1) &= \rho_1^m, \quad m = 0, \dots, M.\end{aligned}$$

For  $j = 2, \dots, J$ ,  $m = 1, \dots, M$ , do

$$G(m, j) = G(m, j-1) + \rho_j G(m-1, j).$$

Then  $G_M = G(M, J)^{-1}$ .

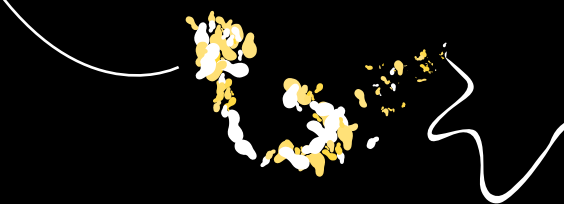
- ▶ Buzen's algorithm yields  $G_m$ ,  $m = 1, \dots, M$ , and marginals and means:

$$\pi_j(n_j) = G_M \rho_j^{n_j} [G_{M-n_j}^{-1} - \rho_j G_{M-n_j-1}^{-1}], \quad n_j = 0, \dots, M-1,$$

$$\pi_j(M) = G_M \rho_j^{n_j},$$

$$\mathbb{E}[N_j] = \sum_{m=1}^M \rho_j^m \frac{G_M}{G_{M-m}}.$$





# Markovian Queues and Stochastic Networks

Lecture 2

Richard J. Boucherie

Stochastic Operations Research

