

# Multi-class Queues and Stochastic Networks

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# Multi-class Queues and Stochastic Networks

Detailed content:

1. reversibility, stationarity, basic queues, output theorem, feedforward networks
2. partial balance, Jackson network, Kelly-Whittle network, arrival theorem
3. quasi-reversibility, customer types, BCMP networks, bandwidth sharing networks
4. blocking, aggregation, decomposition
5. loss networks, insensitivity via supplementary variables
6. sojourn time distribution in networks
7. [MVA](#), [AMVA](#), [QNA](#)
8. fluid queues, basic models
9. feedback fluid queues, networks of fluid queues

# Multi-class Queues and Stochastic Networks

## Today (lecture 6): MVA, AMVA, QNA

Handbook, chapter 13 – published handbook on website MQSN

- MVA for single queues
- MVA for closed networks

## Preliminaries

- Product form equilibrium distribution not enough for performance analysis:

closed network : equilibrium distribution

- Theorem: The equilibrium distribution for the closed Jackson network containing  $N$  jobs is

$$\pi(n) = B_N \prod_{j=1}^J \rho_j^{n_j} \quad \rho_j = \gamma_j / \mu_j \quad n \in \mathcal{S}_N = \{n : \sum_j n_j = N\}$$

and satisfies partial balance

$$\sum_{k=1}^J \pi(n) q(n, n - e_j + e_k) = \sum_{k=1}^J \pi(n - e_j + e_k) q(n - e_j + e_k, n)$$

traffic equations 
$$\sum_k \gamma_j P_{jk} = \sum_k \gamma_k P_{kj}$$

$$q(n, n - e_j + e_k) = \mu_j P_{jk}$$

$$q(n, n - e_j) = \mu_j P_{j0}$$

$$q(n, n + e_k) = \lambda_k$$

# Open and closed networks

$$\lambda_i = \gamma_i + \sum_{j=1}^N \lambda_j \cdot r_{ji} \quad i = 1, \dots, N$$

- Theorem
  - steady state distribution (if open: provided  $\rho_i = \lambda_i / \mu_i < 1$ ):

$$P(k_1, \dots, k_N) = C \prod_{i=1}^N \rho_i^{k_i}$$

Intermezzo: Normalising constant for open networks

- Determining the normalising constant easy for open networks

$$P(k_1, \dots, k_N) = C \prod_{i=1}^N \rho_i^{k_i} = \prod_{i=1}^N (1 - \rho_i) \cdot \rho_i^{k_i}$$

$$C^{-1} = \sum_{(k_1, \dots, k_N): k_i \geq 0} \prod_{i=1}^N \rho_i^{k_i} = \prod_{i=1}^N \sum_{k_i \geq 0} \rho_i^{k_i} = \prod_{i=1}^N (1 - \rho_i)^{-1}$$

## Determining the normalising constant: Buzen's algorithm

- Recursion for summation

$$C_m^{-1} = \sum_{k_1 + \dots + k_N = m} \prod_{i=1}^N \rho_i^{k_i}$$

Define

$$G_i(k), \quad i = 0, 1, 2, \dots, N \quad k = 0, \dots, m$$

with initial values

$$G_1(k) = \rho_1^k \quad k = 0, \dots, m$$

$$G_i(0) = 1 \quad i = 0, 1, 2, \dots, N$$

recursion

$$G_i(k) = G_{i-1}(k) + \rho_i G_i(k-1)$$

then

$$C_m^{-1} = G_N(m)$$



# Correctness of Buzen's algorithm

$$C_m^{-1} = \sum_{k_1 + \dots + k_N = m} \prod_{i=1}^N \rho_i^{k_i}$$

$$\begin{aligned} \sum_{k_1 + k_2 = m} \prod_{i=1}^2 \rho_i^{k_i} &= \sum_{k_1 + k_2 = m} \rho_1^{k_1} \rho_2^{k_2} = \sum_{k_1=0}^m \rho_1^{k_1} \rho_2^{m-k_1} \\ &= \rho_1^m + \sum_{k_1=0}^{m-1} \rho_1^{k_1} \rho_2^{m-k_1} = \rho_1^m + \rho_2 \sum_{k_1=0}^{m-1} \rho_1^{k_1} \rho_2^{m-1-k_1} \end{aligned}$$

$$G_1(k) = \rho_1^k \quad k = 0, \dots, m$$

$$G_i(k) = G_{i-1}(k) + \rho_i G_i(k-1)$$

## Preliminaries

- Product form equilibrium distribution not enough for performance analysis
- But what about the normalising constant?
- Use algorithms that focus on the mean performance measures
- Little's formula  $L = \lambda S, Q = \lambda W \quad \rho = \lambda b.$
- PASTA, MUSTA (ASTA)

# Arrival theorem

- PASTA:  
The distribution of the number of customers in the system seen by a customer arriving to a system according to a Poisson process (i.e., at an arrival epoch) equals the distribution of the number of customers at an arbitrary epoch.
- Arrival theorem (open Jackson network):  
In an open network in equilibrium, a customer arriving to queue  $j$  observes the equilibrium distribution.
- Arrival theorem (closed Jackson network):  
In a closed network in equilibrium, a customer arriving to queue  $j$  observes the equilibrium distribution of the network containing one customer less.

# MVA for single-station systems: MM1

$$W = \rho \frac{1}{\mu} + Q \frac{1}{\mu} \quad \text{Service upon arrival: PASTA}$$

$$Q = \lambda W$$

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

# MVA for single-station systems: MG1

$$W = \rho R + Qb$$

Residual service time upon arrival:  
PASTA

$$Q = \lambda W$$

$$R = \frac{b^{(2)}}{2b} = \frac{b}{2}(1 + c^2) \quad c^2 = (b^{(2)} - b^2)/b^2$$

$$W = \frac{\rho R}{1 - \rho}$$

MVA for single-station systems: MG1 w priorities

$$W_i = \frac{\sum_j \rho_j R_j}{(1 - \sum_{j \leq i} \rho_j)(1 - \sum_{j < i} \rho_j)}, \quad i = 1, 2, \dots, C.$$

# Arrival theorem

- PASTA:  
If customers arrive according to a Poisson process, then the probability distribution of the number of customers at an arrival time is equal to the probability distribution of the number of customers at an arbitrary instant.
- Arrival theorem (open network):  
In an open network that is in steady state, a customer arriving at station  $j$  observes the system as if he observes it at an arbitrary instant.
- Arrival theorem (closed network):  
In a closed network with  $m$  customers that is in steady state, a customer arriving at station  $j$  observes the system as if he observes the network with  $(m-1)$  customers at an arbitrary instant.

# CLOSED NETWORKS OF $M/M/1$ QUEUES.

## MVA

Mean queue length, mean response time?

$\lambda_m(i)$  arrival rate at station  $i$ ,

$F_m(i)$  expected response time at station  $i$ ,

$L_m(i)$  expected queue length at station  $i$ , if  $m$  customers in system

$$F_m(j) = \frac{1}{\mu_j} + L_{m-1}(j) \frac{1}{\mu_j} \text{Arrival theorem, FCFS}$$

$$L_m(j) = \lambda_m(j) F_m(j)$$

Little's law

Another relation needed to determine  $\lambda_m(j)$



# Closed networks of M/M/1 queues:

## • MVA Notation

- $m$  : number of customers in the system
- $\pi$  steady state distribution of Markov chain with transition matrix  $R=(r_{ij})$ :

$$\pi_i = \sum_{j=1}^N \pi_j \cdot r_{ji} \quad , \quad \sum_{i=1}^N \pi_i = 1$$

- $\lambda_m(i)$  is the arrival rate at station  $i$ , if  $m$  customers are circulating in the system

## • Flow equations

$$\lambda_m(i) = \sum_{j=1}^N \lambda_m(j) \cdot r_{ji}$$

- Such that

$$\lambda_m(j) = \lambda_m \pi_j \quad \lambda_m = \sum_{j=1}^N \lambda_m(j)$$

- $\lambda_m$  # customers that are served in the system per time unit

# CLOSED NETWORKS OF $M/M/1$ QUEUES.

## MVA

- $\lambda_m(i)$  arrival rate at station  $i$ ,
- $F_m(i)$  expected response time at station  $i$ ,
- $L_m(i)$  expected queue length at station  $i$ , if  $m$  customers in system

$$\pi_i = \sum_{j=1}^N \pi_j \cdot r_{ji} \quad , \quad \sum_{i=1}^N \pi_i = 1$$

$$\lambda_m(i) = \sum_{j=1}^N \lambda_m(j) \cdot r_{ji} \quad , \quad \lambda_m(i) = \lambda_m \cdot \pi_i \quad , \quad \lambda_m = \sum_{i=1}^N \lambda_m(i)$$

- Little's law

$$m = \sum_{i=1}^N L_m(i) = \sum_{i=1}^N \lambda_m(i) \cdot F_m(i) = \lambda_m \cdot \sum_{i=1}^N \pi_i \cdot F_m(i)$$

- Consequence  $\lambda_m = m \cdot \left\{ \sum_{i=1}^N \pi_i \cdot F_m(i) \right\}^{-1}$

# CLOSED NETWORKS OF $M/M/1$ QUEUES.

## MVA

- $\lambda_m(i)$  arrival rate at station  $i$ ,
- $F_m(i)$  expected response time at station  $i$ ,
- $L_m(i)$  expected queue length at station  $i$ , if  $m$  customers in system

$$F_m(j) = \frac{1}{\mu_j} + L_{m-1}(j) \frac{1}{\mu_j},$$

$$L_m(j) = \lambda_m(j) F_m(j)$$

$$\lambda_m(i) = \lambda_m \cdot \pi_i = m \cdot \left\{ \sum_{j=1}^N \pi_j \cdot F_m(j) \right\}^{-1} \cdot \pi_i$$

# Closed networks of M/M/1 queues: MVA

calculates  $\lambda_m(i)$ ,  $F_m(i)$  and  $L_m(i)$  for all  $m, i$  through recursion

## Algorithm

- Determine solution  $\pi$  of the flow equations
- $m:=1$
- $F_m(i)=1/\mu_i$  for all  $i$

R: recursive step

$m:=m+1$  and go to R

$$\lambda_m(i) = \lambda_m \cdot \pi_i = m \cdot \left\{ \sum_{j=1}^N \pi_j \cdot F_m(j) \right\}^{-1} \cdot \pi_i$$

$$L_m(i) = \lambda_m(i) \cdot F_m(i)$$

$$F_{m+1}(i) = \frac{1 + L_m(i)}{\mu_i}$$

# Generalisations

- Approximate MVA
- QNA:  
W. Whitt (1983) The queueing network analyzer. Bell. Syst. Tech J 62,  
pp 2779 - 2815