CS 672 – Solving Queuing Networks

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• Most of the figures in this set of slides come from the book “Performance by Design: computer capacity planning by example,” by Menascé, Almeida, and Dowdy, Prentice Hall, 2004. It is strictly forbidden to copy, post on a Web site, or distribute electronically, in part or entirely, any of the slides in this file.
Solution to QN Models

- Single Class QNs
  - Single class MVA
- Multiple Class QNs
  - Approximate MVA
- Multiclass Open QNs
- Mixed QNs

A Simple Single Class QN Model

Questions:
- How does the throughput vary with N?
- What is the response time when N = 10?
Closed QN Model: Mean Value Analysis (MVA)

\[ R_i(n) = S_i + S_i \times \bar{n}_i^A(n) \]

“My response time is equal to my service time plus my waiting time (i.e., the service time of all those who arrived ahead of me).”

**Notation:**
(n) means “a function of n.”

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Arrival theorem:

\[ \bar{n}_i^A(n) = \bar{n}_i(n-1) \]

“I cannot find myself in the queue, thus the n-1.”

**Notation:**
(n) means “a function of n.”
Closed QN Model: Mean Value Analysis (MVA)

\[ R_i(n) = S_i + S_i \times n_i^A(n) \]

“My response time is equal to my service time plus my waiting time (i.e., the service time of all those who arrived ahead of me).”

Arrival theorem:
\[ n_i^A(n) = n_i(n-1) \]

“A my cannot find myself in the queue, thus the n-1 (Arrival theorem).”

Avg. # people I find in the queue.  Avg. # people in the queue.

So:
\[ R_i(n) = S_i[1 + n_i(n-1)] \]

Notation:
(n) means “a function of n.”

Closed QN Model: Mean Value Analysis (MVA)

But:
\[ R_i(n) = V_i R_i(n) = V_i S_i[1 + n_i(n-1)] \]

“The residence time is equal to the response time per visit times the average number of visits to resource i per transaction.”
Closed QN Model: Mean Value Analysis (MVA)

"The residence time is equal to the response time per visit times the average number of visits to resource i per transaction."

Finally, we get equation (1) of MVA:

\[
R_i'(n) = D_i[1 + \bar{n}_i(n - 1)]
\]

Closed QN Model: Mean Value Analysis (MVA)

Applying Little’s Law to the entire system:

\[
n = X_0(n) \times R_o(n) = X_0(n) \times \sum_{i=1}^{K} R_i'(n)
\]

Remember that the response time is the sum of all residence times?
Closed QN Model: Mean Value Analysis (MVA)

Applying Little’s Law to the entire system:

\[ n = X_0(n) \times R_o(n) = X_0(n) \times \sum_{i=1}^{K} R_i'(n) \]

Remember that the response time is the sum of all residence times?

Finally, we get equation (2) of MVA:

\[ X_0(n) = n / \sum_{i=1}^{K} R_i'(n) \]

Closed QN Model: Mean Value Analysis (MVA)

Applying Little’s Law to resource i:

\[ \bar{n}_i(n) = X_i(n) \times R_i(n) \]

Avg. queue length at resource i
Closed QN Model: Mean Value Analysis (MVA)

Applying Little’s Law to resource i:

\[ \bar{n}_i(n) = X_i(n) \times R_i(n) \]

Average queue length at resource i  
Average # visits to resource i  
Response time at resource i.

Using the Force Flow Law:

\[ \bar{n}_i(n) = X_i(n) \times R_i(n) = V_i X_0(n) \times R_i(n) \]

System throughput  
Residence time at resource i.

Finally, we get equation (2) of MVA:

\[ \bar{n}_i(n) = X_0(n) R_i'(n) \]
Mean Value Analysis (MVA): putting it all together

Residence Time Equation:

\[ R'_i(n) = D_i[1 + \bar{n}_i(n - 1)] \]

Throughput Equation:

\[ X_0(n) = n / \sum_{i=1}^{K} R'_i(n) \]

Queue length equation:

\[ \bar{n}_i(n) = X_0(n)R'_i(n) \]

A Simple Two Class QN Model

Questions:

- What is the predicted increase in the throughput of query transactions if the load of update transactions is moved to off-peak hours?
- How will the response time change if the total I/O load of query transactions is moved to disk 2?
Notation for Closed Multiclass QNs

- \( R \): number of classes
- \( \tilde{N} = (N_1, ..., N_r, ..., N_R) \): population vector
- \( \tilde{I}_r = (0, ..., 1, ..., 0) \): vector in which all elements except for the \( r \)-th, which is 1, are zero.
- \( R_{i,r}(\tilde{N}) \): avg. residence time at device \( i \) for class \( r \) customers.
- \( R_{i,r}(\tilde{N}) \): avg. response time at device \( i \) for class \( r \) customers.
- \( X_{0,r}(\tilde{N}) \): avg. system throughput for class \( r \) customers.
- \( \tilde{n}_{i,r}(\bar{N}) \): avg. queue length at device \( i \) for class \( r \) customers.
- \( \bar{n}_i(\bar{N}) \): avg. queue length at device \( i \) for all classes.
- \( \bar{n}_{i,r}(\bar{N}) \): avg. queue length at device \( i \) seen by an arriving class \( r \) customers.

The BCMP Theorem

- Network state: \( \tilde{n} = (\tilde{n}_1, ..., \tilde{n}_K) \)
  \[ \tilde{n}_i = (n_{i,1}, ..., n_{i,r}, ..., n_{i,R}) \]
  no. of class \( r \) customers at device \( i \).
- BCM theorem: condition for product form solution:
  \[ p(\tilde{n}) = p(\tilde{n}_1) \times ... \times p(\tilde{n}_i) \times ... \times p(\tilde{n}_K) \]
BCMP Theorem Assumptions

- **FCFS**: service time distribution is exponential with the same mean for all classes. Visit ratios may be different. The service rate can be load dependent but it can only depend on the total number of customers at the device.
- **PS**: processor-sharing discipline. Each class may have a distinct service time distribution.
- **IS**: infinite servers (or ample number of servers or delay server). No queuing.
- **LCFS-PR**: Last-Come First-Served Preemptive Resume. Each class may have a distinct service time distribution.

MVA Formulas for Multiclass QNs

\[
R_{i,r} (\bar{N}) = S_{i,r} \left[ 1 + \bar{n}_{i,r}^A (\bar{N}) \right]
\]

\[
\bar{R}_{i,r} (\bar{N}) = V_{i,r} S_{i,r} \left[ 1 + \bar{n}_{i,r}^A (\bar{N}) \right] = D_{i,r} \left[ 1 + \bar{n}_{i,r}^A (\bar{N}) \right]
\]

\[
X_{0,r} (\bar{N}) = N_r / \sum_{i=1}^{K} R_{i,r} (\bar{N})
\]

\[
\bar{n}_{i,r} (\bar{N}) = X_{i,r} (\bar{N}) R_{i,r} (\bar{N}) = X_{0,r} (\bar{N}) V_{i,r} R_{i,r} (\bar{N}) = X_{0,r} (\bar{N}) R_{i,r} (\bar{N})
\]

\[
\bar{n}_{i} (\bar{N}) = \sum_{r=1}^{R} \bar{n}_{i,r} (\bar{N}) = \sum_{r=1}^{R} X_{0,r} (\bar{N}) R_{i,r} (\bar{N})
\]
MVA Formulas for Multiclass QNs

Arrival Theorem: \( \bar{n}^A_{i,r}(\bar{N}) = \bar{n}_i(\bar{N} - 1_r) \)

So: \( R'_{i,r}(\bar{N}) = D_{i,r} \left[ 1 + \bar{n}_i(\bar{N} - 1_r) \right] \)

\[
X_{0,r}(\bar{N}) = N_r / \sum_{i=1}^{K} R'_{i,r}(\bar{N})
\]

\( \bar{n}_{i,r}(\bar{N}) = X_{0,r}(\bar{N})R'_{i,r}(\bar{N}) \)

\( \bar{n}_i(\bar{N}) = \sum_{r=1}^{R} \bar{n}_{i,r}(\bar{N}) = \sum_{r=1}^{R} X_{0,r}(\bar{N})R'_{i,r}(\bar{N}) \)

MVA Formulas for Multiclass QNs

The term \( \bar{n}_i(\bar{N} - 1_r) \) requires that all \( R \) \( \bar{n}_{i,r}(\bar{N} - 1_r) \) terms be computed.

Computational complexity:

\[
K \times R \times \prod_{r=1}^{R} (1 + N_r)
\]
Approximate MVA for Multiclass Closed QNs

Bard-Schweitzer Approximation:

\[ \tilde{n}_{i,r}(\tilde{N} - \tilde{1}_r) = \frac{N_r - 1}{N_r} \tilde{n}_{i,r}(\tilde{N}) \]

So,

\[ \tilde{n}_{i}(\tilde{N} - \tilde{1}_r) = \frac{N_r - 1}{N_r} \tilde{n}_{i,r}(\tilde{N}) + \sum_{s=1 \& s \neq r}^{R} \tilde{n}_{i,s}(\tilde{N}) = \]

Therefore, need \( \tilde{n}_{i,r}(\tilde{N}) \) to compute \( \tilde{n}_{i}(\tilde{N} - \tilde{1}_r) \).

Solution: start with initial value for \( \tilde{n}_{i,r}(\tilde{N}) \) as:

\[ \tilde{n}_{i,r}^{e}(\tilde{N}) = \frac{N_r}{K_r} \text{ where } K_r = \left\{ i \mid D_{i,r} \neq 0 \right\} \]

Approximate MVA for Multiclass QNs

Step 1: \( \tilde{n}_{i,r}^{e}(\tilde{N}) = \frac{N_r}{K_r} \quad \forall \quad i, r \)

Step 2: \( \tilde{n}_{i}(\tilde{N} - \tilde{1}_r) = \frac{N_r - 1}{N_r} \tilde{n}_{i,r}^{e}(\tilde{N}) + \sum_{s=1 \& s \neq r}^{R} \tilde{n}_{i,s}^{e}(\tilde{N}) \quad \forall \quad i, r \)

Step 3: \( R_{i,r}^{e}(\tilde{N}) = D_{i,r} \left[ 1 + \tilde{n}_{i}(\tilde{N} - \tilde{1}_r) \right] \quad \forall \quad i, r \)

Step 4: \( X_{0,r}(\tilde{N}) = \frac{N_r}{K_r} \sum_{i=1}^{K} R_{i,r}^{e}(\tilde{N}) \quad \forall \quad r \)

Step 5: \( \tilde{n}_{i,r}(\tilde{N}) = X_{0,r}(\tilde{N}) R_{i,r}^{e}(\tilde{N}) \quad \forall \quad i, r \)

If \( \max_{i,r} \left\{ \text{abs} \left[ \frac{\tilde{n}_{i,r}^{e}(\tilde{N}) - \tilde{n}_{i,r}(\tilde{N})}{\tilde{n}_{i,r}(\tilde{N})} \right] \right\} > \varepsilon \)

then \( \tilde{n}_{i,r}^{e}(\tilde{N}) = \tilde{n}_{i,r}(\tilde{N}) \quad \forall \quad i, r \) and go to step 2.
Notation for Open Multiclass QNs

\( R \) : number of classes
\( \vec{\lambda} = (\lambda_1, \ldots, \lambda_r, \ldots, \lambda_R) \) : arrival rate vector
\( R_{i,r}'(\vec{\lambda}) \) : avg. residence time at device \( i \) for class \( r \) customers.
\( R_{i,r}(\vec{\lambda}) \) : avg. response time at device \( i \) for class \( r \) customers.
\( X_{0,r}(\vec{\lambda}) \) : avg. system throughput for class \( r \) customers.
\( \bar{n}_{i,r}(\vec{\lambda}) \) : avg. queue length at device \( i \) for class \( r \) customers.
\( \bar{n}_r(\vec{\lambda}) \) : avg. queue length at device \( i \) for all classes.
\( \bar{n}_{i,r}^\Lambda(\vec{\lambda}) \) : avg. queue length at device \( i \) seen by an arriving class \( r \) customers.

Formulas for Open Multiclass QNs

Arrival Theorem: \( \pi_{i,r}^\Lambda(\vec{\lambda}) = \pi_i(\vec{\lambda}) \)

So: \( R_{i,r}'(\vec{\lambda}) = D_{i,r} \left[ 1 + \pi_i(\vec{\lambda}) \right] \)
\( X_{0,r}(\vec{\lambda}) = \lambda_r \)
Little’s Law: \( \pi_{i,r}(\vec{\lambda}) = X_{0,r}(\vec{\lambda}) R_{i,r}(\vec{\lambda}) \)

\[ \Rightarrow \pi_{i,r}(\vec{\lambda}) = \lambda_r R_{i,r}(\vec{\lambda}) = \lambda_r D_{i,r} \left[ 1 + \pi_i(\vec{\lambda}) \right] \]

\[ = U_{i,r}(\vec{\lambda}) \left[ 1 + \pi_i(\vec{\lambda}) \right] \]

For any two classes \( i \) and \( j \):

\[ \frac{\bar{n}_{i,r}(\vec{\lambda})}{\bar{n}_r(\vec{\lambda})} = \frac{U_{i,r}(\vec{\lambda})}{U_{i,j}(\vec{\lambda})} \]

Since \( \pi_i(\vec{\lambda}) = \sum_{r=1}^R \pi_{i,r}(\vec{\lambda}) \)

\[ \Rightarrow \bar{n}_{i,r}(\vec{\lambda}) = \frac{U_{i,r}(\vec{\lambda})}{1 - U_i(\vec{\lambda})} \]

Little’s Law: \( R_{i,r}(\vec{\lambda}) = \frac{D_{i,r}}{1 - U_i(\vec{\lambda})} \)
Formulas for Open QN Models

\[ U_{i,r} = \lambda_r \times D_{i,r} \]

\[ U_i = \sum_{r=1}^{R} U_{i,r} \]

\[ R_{i,r} = \frac{D_{i,r}}{1-U_i} \]

\[ R_r = \sum_{i=1}^{K} R_{i,r} \]

Mixed Class Models

• Classes 1, …, O are open and classes O+1, O+2, …, O+C are closed.

• The O open classes are characterized by the vector of arrival rates \( \vec{\lambda} = (\lambda_1, ..., \lambda_O) \)

• The C closed classes are characterized by the population vector \( \vec{N} = (N_{O+1}, ..., N_{O+C}) \)
Solution of Mixed Class Models

Step 1: Solve the open submodels and obtain:
\[ U_{i,r}(\bar{\lambda}) = \lambda_r D_{i,r} \quad \forall \; i, r \]

Step 2: Find the total utilization of all open classes
\[ U_{\text{open}} = \sum_{r=1}^{C} U_{i,r}(\bar{\lambda}) \quad \forall \; i \]

Step 3: Elongate service demands of closed classes:
\[ D'_{i,r} = \frac{D_{i,r}}{1 - U_{\text{open}}} \quad \forall \; i, r \]

Step 4: Use MVA and solve closed model to find:
\[ R_{i,r}(\bar{N}), \pi_{i,r}(\bar{N}), x_{i,r}(\bar{N}) \quad \forall \; r = O + 1,..., O + C \]

Step 5: Find \[ \pi_{\text{close}}(\bar{N}) = \sum_{r=O+1}^{O+C} \pi_{i,r}(\bar{N}) \quad \forall \; i \]

Step 6: Compute metrics for open submodel:
\[ R_{i,r}(\bar{\lambda}) = \frac{D_{i,r}[1 + \pi_{\text{close}}(\bar{N})]}{1 - U_{\text{open}}} \]
\[ \pi_{i,\text{open}}(\bar{\lambda}) = \sum_{r=1}^{C} \lambda_r R_{i,r}(\bar{\lambda}) \]