

Lecture 9

Multiple Class Models



Multiclass MVA
Approximate MVA

Arrival Theorem for Multiple Classes

- With $\bar{N}$ jobs in the system, a job in class r arriving to any server i sees the server as in equilibrium with one job (himself!) removed from the system

$$A_{ir}(\bar{N}) = \bar{n}_i(\bar{N} - \bar{1}_r) = \bar{n}_{ir}^A(\bar{N})$$

$(N_1, N_2, \dots, N_R)$ R job classes
 r^{th} job class
 $(0, 0, \dots, 0, 1, 0, \dots, 0)$

Multiple Class MVA ⁽⁴⁾

starting point: $A_{ir}(\bar{N}) = \bar{n}_i(\bar{N} - \bar{1}_r) = \bar{n}_{ir}^A(\bar{N})$ $\bar{n}_{i,r}(0) = 0$

residence time

$$R'_{i,r}(\bar{N}) = \begin{cases} D_{i,r} [1 + \bar{n}_{i,r}^A(\bar{N})] & \text{queueing device} \\ D_{i,r} & \text{delay device} \end{cases} \quad \forall i, r$$

$$X_{0,r}(\bar{N}) \stackrel{\text{Little}}{\underset{\text{system}}{=}} \frac{N_r}{Z_r + \sum_i R'_{i,r}(\bar{N})} \quad \forall r \quad (\text{response time law})$$

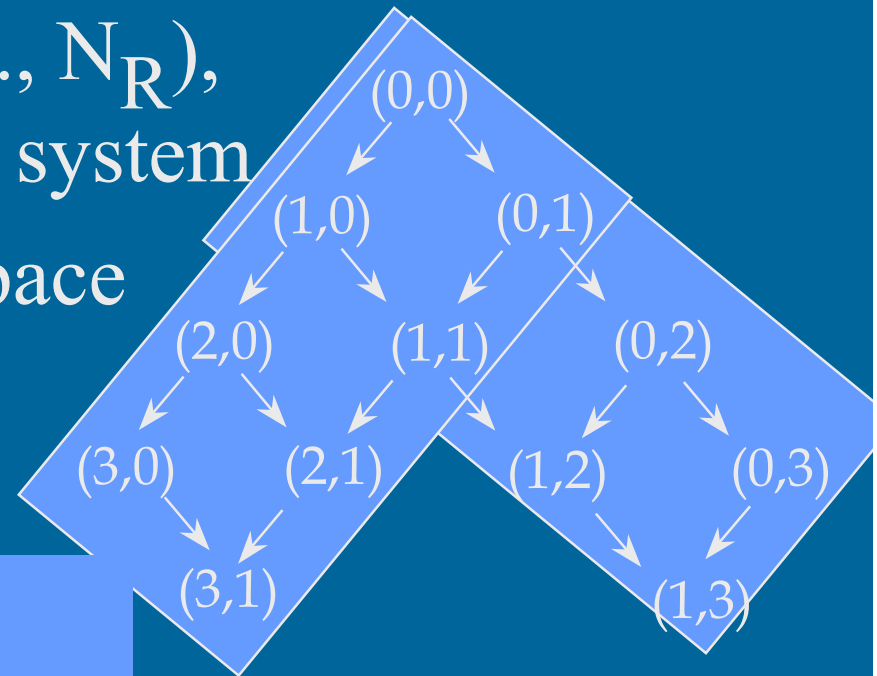
$$\bar{n}_{i,r}(\bar{N}) \stackrel{\text{Little}}{\underset{\text{device}}{=}} X_{i,r}(\bar{N}) R_{i,r}(\bar{N}) = X_{0,r}(\bar{N}) R'_{i,r}(\bar{N}) \quad \forall i, r$$

$$\bar{n}_i(\bar{N}) = \sum_r \bar{n}_{i,r}(\bar{N}) \quad \forall i$$

Multiple Class

Mean Value Analysis (MVA) ⁽²⁾

- Compute solutions through class population space $N=(N_1, N_2, \dots, N_R)$, starting from empty system
- population (state) space can still be large!



two job classes
target population (1,3)?
target population (3,1)?
target population (6, 15, 300)?

Alg 7.2 [LZGS 84]

Example ⁽⁴⁾

2-class closed model, Fig. 7.2 [LZGS 84]

Job classes A, B

Tbl 7.3 [LZGS 84]

(0,0) $Q_{CPU,A} = Q_{DISK,A} = Q_{CPU,B} = Q_{DISK,B} = 0$

(1,0) $R'_{CPU,A}, R'_{DISK,A}, X_{0,A}, Q_{CPU,A}, Q_{DISK,A}$

(0,1) $R'_{CPU,B}, R'_{DISK,B}, X_{0,B}, Q_{CPU,B}, Q_{DISK,B}$

(1,1) $R'_{CPU,A}, R'_{DISK,A}, R'_{CPU,B}, R'_{DISK,B},$
 $X_{0,A}, X_{0,B},$
 $Q_{CPU,A}, Q_{DISK,A}, Q_{CPU,B}, Q_{DISK,B}$

Another Simple Example

Fig. 6.1 [Men 94]

$D_{i,r}$	query $r=1$	update $r=2$
cpu $i=1$	0.105	0.375
d1 $i=2$	0.180	0.480
d3 $i=3$	0	0.240

Watch out for
index ordering:

$$D_{1,2} = D_{2, \text{query}}$$

Population in
book figures
in order
(Update, Query)

Simple Example ⁽⁵⁾

update
query

(0,0)

$$\bar{n}_i(\bar{0}) = 0 = \bar{n}_{i,r}^A(1_r) \quad \forall i, r$$

(1,0)

$$R'_{1,1}(1,0) = D_{1,1}[1+0] = 0.105 \quad R'_{2,1}(1,0) = 0.180 \quad R'_{3,1}(1,0) = 0$$

$$X_{0,1}(1,0) = \frac{1}{(0.105 + 0.180)} = 3.509$$

$$\bar{n}_{1,1}(1,0) = 3.509 * 0.105 = 0.368$$

$$\bar{n}_{2,1}(1,0) = 3.509 * 0.180 = 0.632 \quad \bar{n}_{3,1}(1,0) = 0$$

$$R'_{1,2}(1,0) = ?$$

$$\bar{n}_{1,2}(1,0) = \bar{n}_{2,2}(1,0) = \bar{n}_{3,2}(1,0) = 0$$

$$\bar{n}_1(1,0) = \bar{n}_{1,1}(1,0) + \bar{n}_{1,2}(1,0) = 0.368$$

$$\bar{n}_2(1,0) = 0.632 \quad \bar{n}_3(1,0) = 0$$

Tbl 6.3

Simple Example (contd) ⁽⁶⁾

(0,1): $\bar{n}_1(0,1) = 0.343 \quad \bar{n}_2(0,1) = 0.438 \quad \bar{n}_3(0,1) = 0.219$

(1,1): $R'_{1,1}(1,1) = D_{1,1} \left[1 + \bar{n}_{1,1}^A(1,1) \right] = D_{1,1} \left[1 + \bar{n}_1(0,1) \right]$
 $= 0.105 [1 + 0.343] = 0.141$

$R'_{2,1}(1,1) = D_{2,1} \left[1 + \bar{n}_2(0,1) \right] = 0.180 [1 + 0.438] = 0.259$

$R'_{3,1}(1,1) = 0$

device
index

$X_{0,1}(1,1) = \frac{1}{(0.141 + 0.258)} = 2.500$

class
index

$\bar{n}_{1,1}(1,1) = 2.5 * 0.141 = 0.352$

$\bar{n}_{2,1}(1,1) = 2.5 * 0.259 = 0.648$

$\bar{n}_{3,1}(1,1) = 0$

$\bar{n}_{1,2}(1,1) = 0.334$

$\bar{n}_{2,2}(1,1) = 0.519$

$\bar{n}_{3,2}(1,1) = 0.156$

...

total popul.

$\bar{n}_1(1,1) = 0.686$

$\bar{n}_2(1,1) = 1.158$

$\bar{n}_3(1,1) = 0.156$

similarly for states (2,0), (3,0), (2,1), (3,1)

Tbl 6.3

Simple Example (contd)

- Baseline solution

Table 6.3 [Men 94]

- external: class *update* response time: 2.444 s
- internal: Disk1 utilization too high

Table 6.4

- Modification: move queries to Disk2

- move query demand from Disk 1 to Disk2

(silly order,
silly notation,
sorry!)

- $D_{21} = D_{d1,query} = 0, D_{31} = D_{d2,query} = 0.180$

- solve again

Table 6.4 [Men 94]

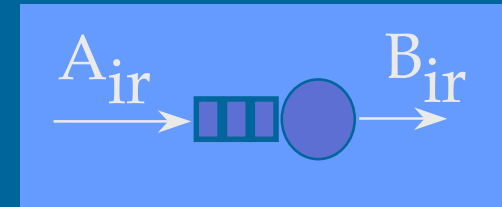
- class *update* response time: 1.934 s

Product Form Solution Exists: NW is Analytically Solvable

- BCMP-networks
 - Baskett, Chandy, Muntz & Palacios (1975)
- Service discipline
 - FCFS, PS, IS, LCFS-PR
- Job classes, class switching
- Service time distributions, interarrival times
 - Exponential interarrivals times for FCFS servers or for open job class
 - more generic for others (rational Laplace transformation exists)
- Load-dependent servers (LD-servers)
 - $S_{ir} = f(n_i)$ for FCFS (I.e., same for each class)
 - $S_{ir} = f(n_{ir})$ for others

Product Form Solution Exists?

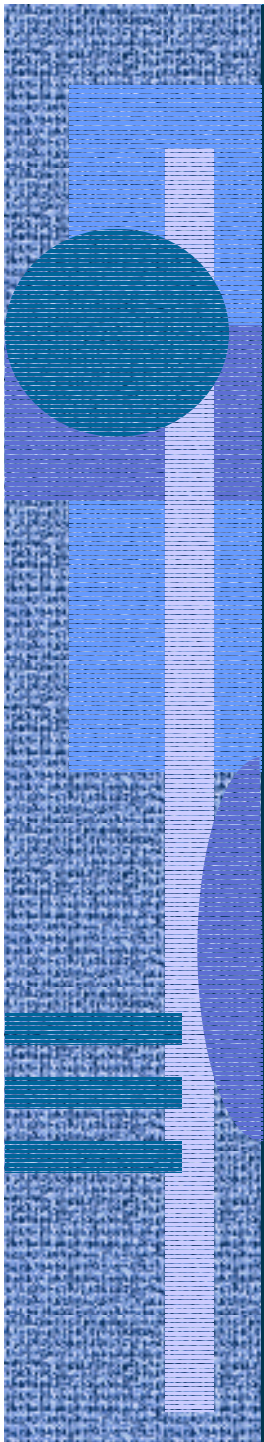
- Job Flow balance
 - flow in = flow out
 - per device, per system
- One step behavior
- Device homogeneity
 - single resource possession
 - no blocking
 - independent job behavior
 - local information
 - fair service
 - routing homogeneity



$$\begin{aligned}\mu_{ir} &= f(n_{ir}) \\ \mu_i &= f(n_i)\end{aligned}$$

$$p_{ikr} \text{ constant}$$

(prob for class change
is load independent)



8.4.2002

Copyright Teemu Kerola 2002

12

How Useful are Exact Solution Methods?

- MVA & Convolution based algorithms
- Very good for single class cases
- Might be too time consuming for multiple class cases if nr of classes or class populations are (very) large

$$O(\text{MVA}) = O(K R \prod_r (1 + N_r))$$

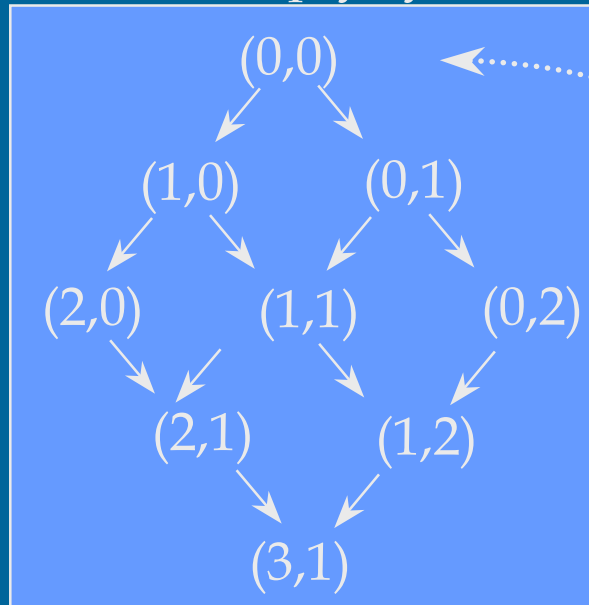


K	R	N _r	#ops
3	2	2	54
5	20	1	100M
20	5	1	3200
20	5	9	10M

Approximate MVA ⁽¹⁾

- Helps to solve MVA state space problem
- Based on Schweitzer-approximation:

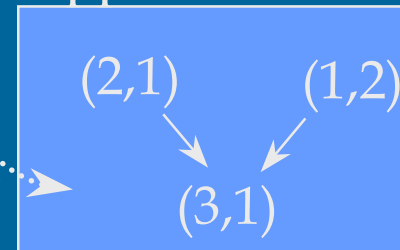
MVA empty system



start
here

$$\frac{n_{ir}(N)}{N_r} \approx \text{constant}$$

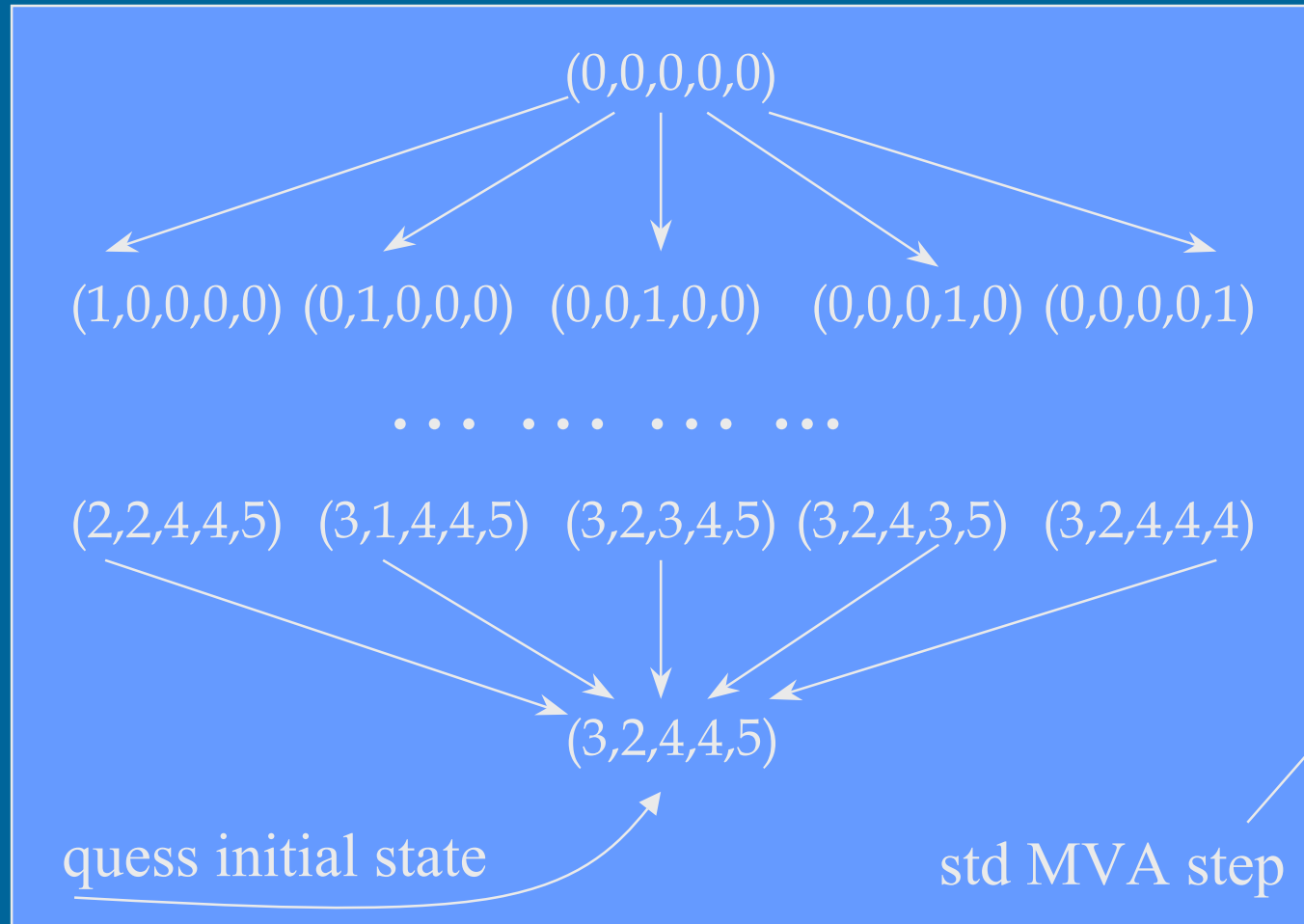
Appr. MVA



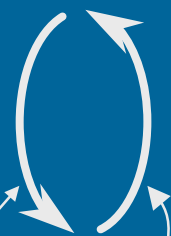
guess first,
iterate

MVA vs Approximate MVA

MVA



Appr
MVA



Schweitzer appr

Single Class Schweitzer Approximation

$$\frac{\bar{n}_i(N)}{N} \approx \text{constant}$$
$$\therefore \frac{\bar{n}_i(N)}{N} = \frac{\bar{n}_i(N-1)}{N-1} \quad \Rightarrow \bar{n}_i(N-1) = \frac{N-1}{N} \bar{n}_i(N)$$

Single Class Approximate MVA

guess initial : $\bar{n}_i(N)$

compute : $\bar{n}_i(N-1) = \frac{N-1}{N} \bar{n}_i(N)$

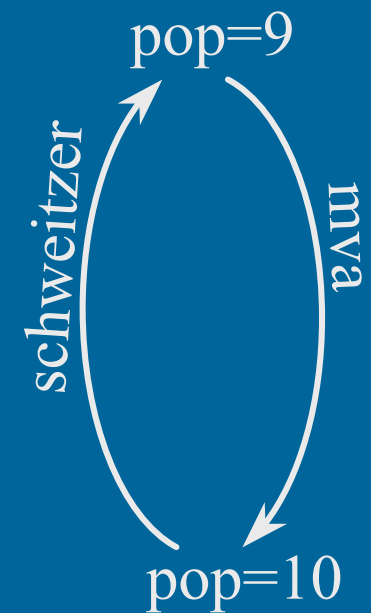
std MVA step :

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

$$X_0(N) = \frac{N}{\left(Z + \sum R'_i(N)\right)}$$

$$\bar{n}_i(N) = X_0(N) * R'_i(N)$$

repeat
until
conver-
gence



Example ⁽⁴⁾ $D_i = (10, 10, 15), K=3, N=2$ [Fig. 5.1]

step 0 $\bar{n}_i(2) = 2 / 3 = 0.667 \quad \forall i$ initial guess: even distribution

step 1

$$R'_1 = D_1 \left[1 + \frac{2-1}{2} \bar{n}_1 \right] = 10 * 1.333 = 13.333 = R'_2 \quad R'_3 = 15 * 1.333 = 20$$

$$X_0 = \frac{2}{46.67} = 0.04286$$

$$\bar{n}_1 = 0.5714 = \bar{n}_2 \quad \bar{n}_3 = 0.8572 \quad \left(\sum \bar{n}_i = 2 \right)$$

step 2

$$R'_1 = 10 \left[1 + \frac{0.5714}{2} \right] = 12.857 = R'_2 \quad R'_3 = 15 * \left[1 + \frac{0.8572}{2} \right] = 21.429$$

$$X_0 = \frac{2}{47.143} = 0.042424$$

$$\bar{n}_1 = 0.5454 = \bar{n}_2 \quad \bar{n}_3 = 0.9092 \quad \left(\sum \bar{n}_i = 2 \right)$$

why stop here?

Questions ⁽³⁾

- Q. How to guess initial job distribution

A: $\bar{n}_i(N) = \frac{N}{K}$ (even distribution on all devices)

- Q. Does it always converge?

A. No. Almost always. No guarantee.

- Q. If it converges, does it converge to the right value?

A. We hope so. It seems to do it

Questions (contd)

- Q. What is good measure of convergence?

A. E.g., max relative change in $\bar{n}_i(N)$ must be less than 1%

General Stationary Iterative Method ⁽¹⁾

(from Numerical Linear Algebra)

Fixed point equation $\bar{n} = B\bar{n} + \bar{c}$

converges from arbitrary initial $\bar{n}^{(0)}$

if $\rho(B) = \max |\lambda_i(B)| < 1$

↑ spectral radius of B ↑ i^{th} eigenvalue of B

- Q. Do we check this before using approximate MVA?
A. No.

General Theory of Iteration ⁽¹⁾

$$\bar{n} = f(\bar{n})$$

Thm:

If $\bar{n} = f(\bar{n})$ has root $\bar{\alpha}$

and $f(\bar{n})$ exists close to α , i.e., in $J = \{\bar{n} : |\bar{n} - \bar{\alpha}| < \rho\}$

and $|f'(\bar{n})| < 1 \quad \forall \bar{n} \in J$

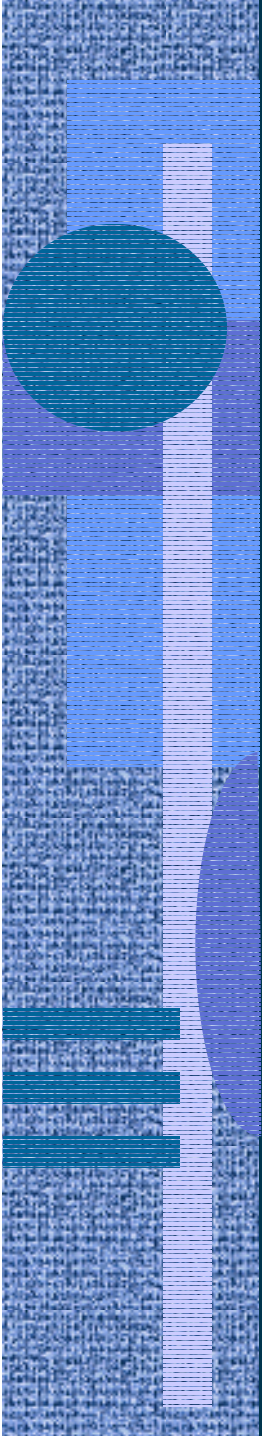
Then (a) $\bar{n} \in J$ in each iteration

(b) $\bar{n}$ converges to $\bar{\alpha}$

(c) $\bar{\alpha}$ is the only root in J

- Q. Do we check this for approximate MVA?

A. No.



How Good is Approximate MVA

- Pretty good for throughput and response time
- Not so good for queue lengths at heavy loads

Figs 34.1-34.3 [Jain 91]

Multiple Class Schweitzer Approximation ⁽¹⁾

single class: $\frac{\bar{n}_i(N)}{N} \approx \text{constant}$

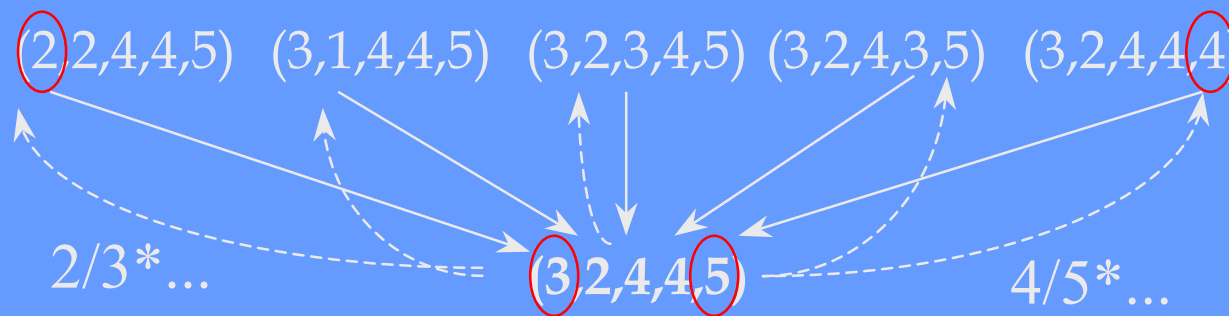
$$\therefore \frac{\bar{n}_i(N)}{N} = \frac{\bar{n}_i(N-1)}{N-1} \Rightarrow \bar{n}_i(N-1) = \frac{N-1}{N} \bar{n}_i(N)$$

multiple class: $\frac{\bar{n}_{ir}(\bar{N})}{N_r} \approx \text{constant}$

$$\therefore \frac{\bar{n}_{ir}(\bar{N})}{N_r} = \frac{\bar{n}_{ir}(\bar{N}-1_r)}{N_r-1}$$
$$\Rightarrow \bar{n}_{ir}(\bar{N}-1_r) = \frac{N_r-1}{N_r} \bar{n}_{ir}(\bar{N})$$

Multiple Class Approximate MVA

$$\bar{n}_{ir}(\bar{N} - 1_r) = \frac{N_r - 1}{N_r} \bar{n}_{ir}(\bar{N})$$



Multiple Class Approximate MVA

- (1) guess initial $n_{ir}(\mathbf{N}) = \mathbf{N}_r / \mathbf{K}_r$ at target state
- (2) compute back $n_{ir}(\mathbf{N}-\mathbf{1}_r) = (\mathbf{N}_r-1)/\mathbf{N}_r n_{ir}(\mathbf{N})$
- (3) use standard MVA step to compute new estimate of $n_i(\mathbf{N})$
- iterate steps (2) and (3) until convergence
 - usually 4-6 iterations enough
 - std output from last iteration

(distribute evenly to every node visited)

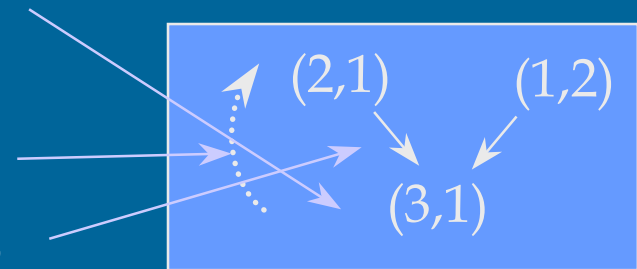


Fig. 6.5 [Men 94]

Multiple Class Approximate MVA

- Does it converge?
 - Almost always. No guarantee.
- If it converges, does it converge to the right value?
 - It seems to do good work....
- What is good measure of convergence?
 - e.g., max relative change in $\bar{n}_{ir}(\bar{N}) < 1\%$

Example ₍₂₎

Model: Fig. 6.1 [Men 94]

step 0

$$\begin{array}{ll} \bar{n}_{11}(3,1) = 1.5 & \bar{n}_{12}(3,1) = 1/3 = 0.333 \\ \bar{n}_{21}(3,1) = 1.5 & \bar{n}_{22}(3,1) = 0.333 \\ \bar{n}_{31}(3,1) = 0 & \bar{n}_{32}(3,1) = 0.333 \end{array}$$

step 1

$$R'_{11}(3,1) = D_{11} \left[1 + \bar{n}_{11}(2,1) + \bar{n}_{12}(3,0) \right]$$

$$\stackrel{\text{Schweitzer}}{=} 0.105 \left[1 + \frac{2}{3} \bar{n}_{11}(3,1) + \frac{0}{1} \bar{n}_{12}(3,1) \right] = 0.105 \left[1 + \frac{2}{3} 1.5 + 0 \right] = 0.210$$

$$R'_{21}(3,1) = 0.180 \left[1 + \frac{2}{3} \bar{n}_{21}(3,1) + \frac{0}{1} \bar{n}_{22}(3,1) \right] = 0.180 * 2 = 0.360$$

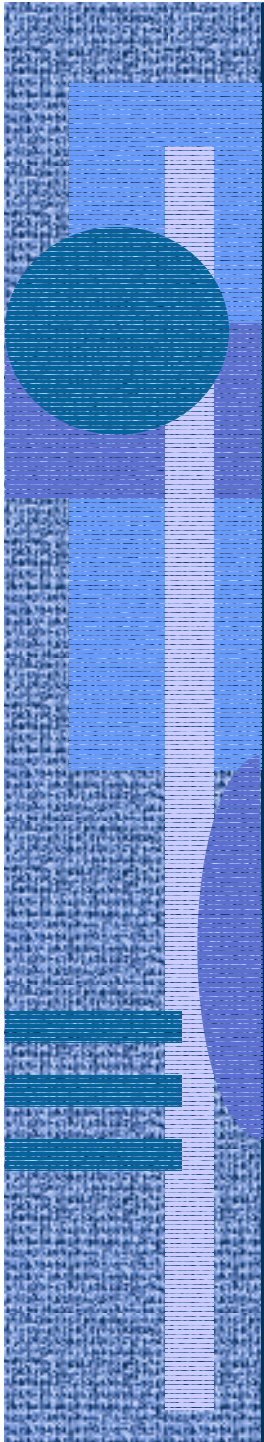
$$R'_{31}(3,1) = 0$$

$$R'_{12}(3,1) = \dots \quad R'_{22}(3,1) = \dots \quad R'_{32}(3,1) = \dots$$



Result: Tbl 6.6, Qsolver/1 output, PMVA output

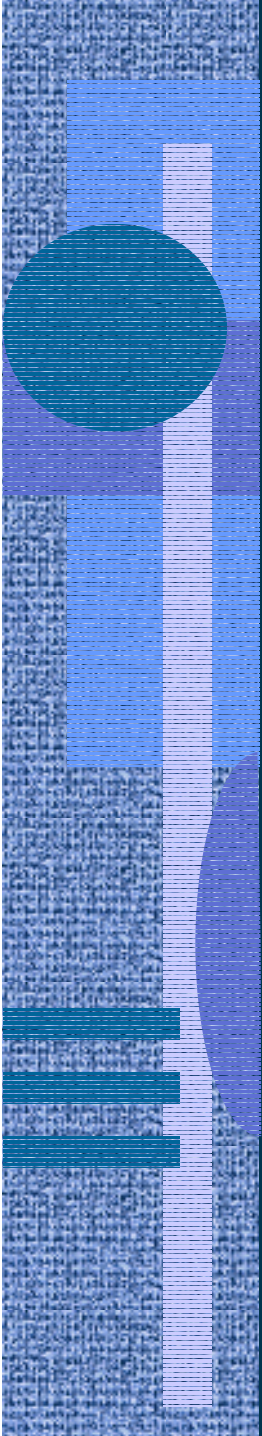
= ?



8.4.2002

Copyright Teemu Kerola 2002

29



Problems with Analytical Solutions

- Load dependent servers
- Memory
- Some part of system not product form
 - however, solution method is robust
- Computational problems
 - accuracy, overflow, underflow

Figs 8.1, 8.3 [Men 94]

Solution Packages (Software)

- Commercial

- BGS, Boston, MA
- AT&T, Holmdale, NJ
- SES, Austin, TX

Best/1

Q+

PAWS

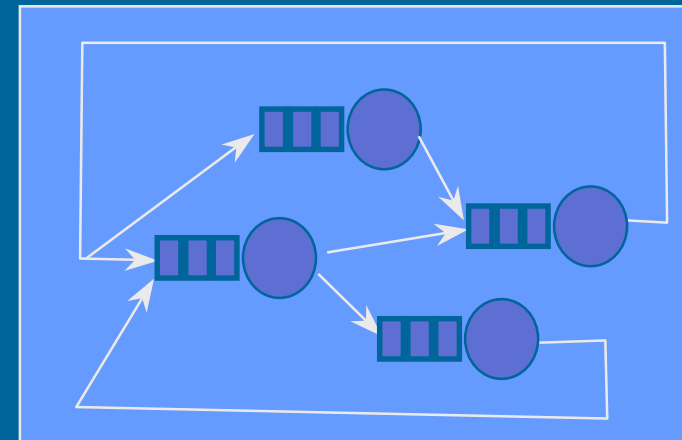
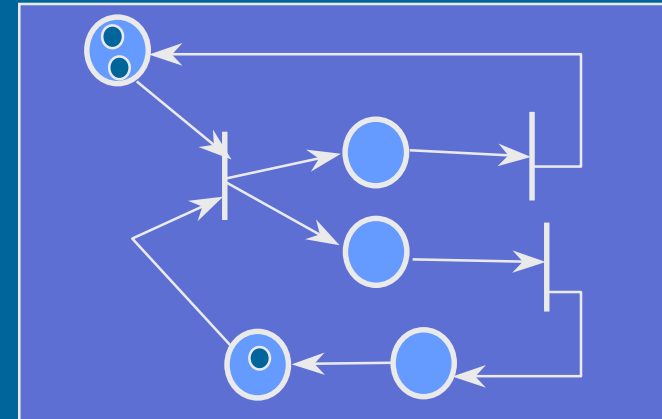
- University Projects

- Jeff Brumfield, U of Austin, TX
- ...

PMVA

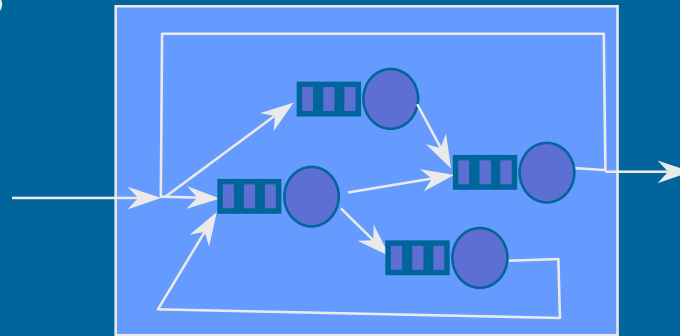
Petri Nets vs. Queueing Networks

- Petri Nets
 - good for concurrency
 - Stochastic Petri Nets (SPN)
 - Timed Petri Nets
 - simulators
- Queueing Networks
 - good for queuing
 - simple bounds
 - Bottleneck Bounds (ABA)
 - Balanced Job Bounds (BJB)
 - exact solution methods
 - approximate solution methods
 - simulators



Exact Solution Methods for Queuing Networks

- Open Networks



- Closed Networks

- Convolution
- MVA
- Appr. MVA

