

Lecture 7

Analytical Solution Method for Complex Models

Multiple Class Markov Chain
Models

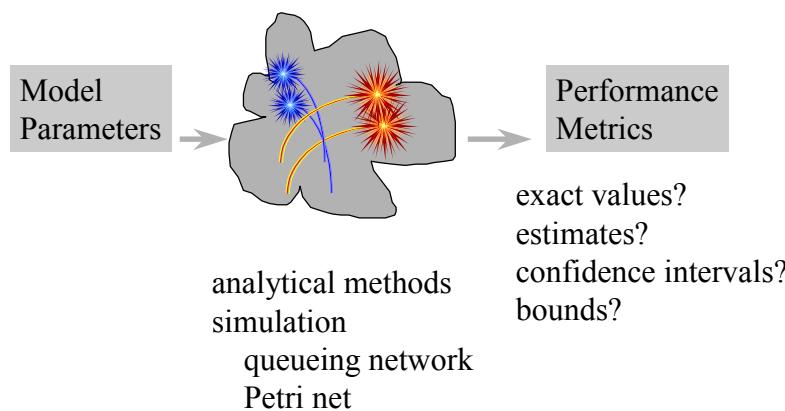
Convolution

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Generic Solution Method



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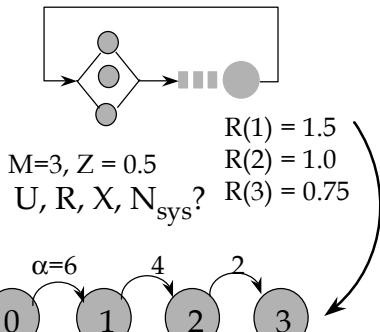
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Markov Chain Solution

- Birth-Death process
- Stochastic process
- Large state space?
- Large normalizing constant?

Birth-Death
Process



$$\begin{aligned}
 & \text{Prob} = P = 0.01 \quad 0.09 \quad 0.36 \quad 0.54 \\
 & U = 1 - P_0 = 0.99, \quad X = \sum \mu_i P_i = 1.14 \text{ tps} \\
 & N_{\text{sys}} = \sum i P_i = 2.43, \quad R = N/X = 2.13 \text{ sec}
 \end{aligned}$$

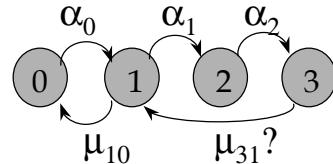
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Markov Chain Solution

- 1. State description
 - finite, infinite? state space?
 - multiple classes? multiple phases?
- 2. State enumeration
- 3. Transition rates
- Balance equations
 - flow in to state = flow out of state, $\sum P_i = 1$
- Solve balance equations
- Use P_i 's to get performance metrics $U = 1 - P_0$



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Markov Example

- Central server model, Fig. 5.1 [Men 94]

- State = (n_c, n_f, n_s) , $\Sigma n_i = 2$
- State enumeration: $n_i \in \{0, 1, 2\}$
 $S = \{(2,0,0), (1,1,0), (1,0,1), (0,2,0), (0,1,1), (0,0,2)\}$
 - state space S can be large (very large)
 - K devices, max mpl=N

$$|S| = \binom{N-K-1}{K-1} = \frac{(N-K-1)!}{N! (K-1)!}$$

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Markov Example (contd)

- State space diagram, Fig. 5.2
- Transition rates

$\text{Rate}((1,1,0) \rightarrow (0,2,0)) = \mu_f = 3$ $\text{Rate}((0,2,0) \rightarrow (1,1,0)) = \mu_c p = 6 * 0.5 = 3$ $\text{Rate}((2,0,0) \rightarrow (1,0,1)) = \mu_c (1-p) = 6 * 0.5 = 3$ $\text{Rate}((1,0,1) \rightarrow (2,0,0)) = \mu_s = 2$ \dots	cpu completes, request fast disk fast disk completes
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Markov Example (contd)

- Notation $P_{011} = \text{Prob } \{ \text{in state } (0,1,1) \}$
- Local balance Fig 5.3 [Men 94]
- Global balance equations

$$\mu_c(1-p)P_{110} + \mu_c p P_{101} = (\mu_s + \mu_f) P_{011}$$

...

“similarly for some other 4 states”

...

$$\sum P_i = 1$$

flow in = flow out

one global
balance
equation is
redundant!

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Markov Example (contd)

- Solve balance equations
 - brute force approach
 - 6 equations, 6 unknowns, OK
 - 92378 equations, 92378 unknowns, ????
 - not always practical,
 - can be very time consuming
- Better method: transform set of equations into simpler form: local balance equations

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Local Balance Equations

- Equations for local balanced transitions between neighboring states
 - Each equation in terms of
 - relative device utilizations
 - normalizing constant
 - needs to be computed
 - can be tricky
 - can be time consuming
 - “the miracle occurs here”
- ← from transition probabilities and service rates

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Relative Utilization ₍₂₎

$$\begin{aligned}\mu_f P_{110} &= \boxed{\mu_c p} P_{200} \\ \mu_s P_{101} &= \mu_c (1-p) P_{200} \\ \mu_f P_{020} &= \mu_c p P_{110} \\ \mu_s P_{011} &= \mu_c (1-p) P_{110} \\ \mu_f P_{011} &= \mu_c p P_{101} \\ \mu_s P_{002} &= \mu_c (1-p) P_{101}\end{aligned}$$

$$\sum P_i = 1$$

Notation:
“Relative Utilization”

$$U_f = \frac{\mu_c p}{\mu_f}$$

$$U_s = \frac{\mu_c (1-p)}{\mu_s}$$

$$U_c = 1$$

(relative to CPU,
serv time & util are
inversely relative to μ)

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Modified Local Balance Equations (1)

$$\begin{aligned}\mu_f P_{110} &= \mu_c p P_{200} \\ \mu_s P_{101} &= \mu_c (1-p) P_{200} \\ \mu_f P_{020} &= \mu_c p P_{110} \\ \mu_s P_{011} &= \mu_c (1-p) P_{110} \\ \mu_f P_{011} &= \mu_c p P_{101} \\ \mu_s P_{002} &= \mu_c (1-p) P_{101}\end{aligned}$$

$$\begin{aligned}P_{110} &= U_f P_{200} \\ P_{101} &= U_s P_{200} \\ P_{020} &= U_f P_{110} = U_f^2 P_{200} \\ P_{011} &= U_s P_{110} = U_s U_f P_{200} \\ P_{011} &= U_f P_{101} = U_f U_s P_{200} \\ P_{002} &= U_s P_{101} = U_s^2 P_{200}\end{aligned}$$

$$\begin{aligned}U_f &= \frac{\mu_c p}{\mu_f} \\ U_s &= \frac{\mu_c (1-p)}{\mu_s} \quad U_c = 1\end{aligned}$$

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Normalizing Constant (2)

$$\begin{aligned}P_{110} &= U_f P_{200} \\ P_{101} &= U_s P_{200} \\ P_{020} &= U_f P_{110} = U_f^2 P_{200} \\ P_{011} &= U_s P_{110} = U_s U_f P_{200} \\ P_{011} &= U_f P_{101} = U_f U_s P_{200} \\ P_{002} &= U_s P_{101} = U_s^2 P_{200}\end{aligned}$$

$$\begin{aligned}P_{110} &= U_c^1 U_f^1 U_s^0 P_{200} \\ P_{101} &= U_c^1 U_f^0 U_s^1 P_{200} \\ P_{020} &= U_c^0 U_f^2 U_s^0 P_{200} \\ P_{011} &= U_c^0 U_f^1 U_s^1 P_{200} \\ P_{002} &= U_c^0 U_f^0 U_s^2 P_{200} \\ \sum P_i &= 1\end{aligned}$$

$$P_{200} = \frac{1}{\sum_{\substack{t \in S \\ t=(i,j,k) \\ i+j+k=2}} U_c^i U_f^j U_s^k} = \frac{1}{G(2)}$$

← normalizing constant
2 jobs

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Local Balance Equations (contd)

$$P_{ijk}(2) = \frac{U_c^i U_f^j U_s^k}{G(2)} \quad \text{when } \text{mpl} = i+j+k=2$$

$$P_{ijk}(N) = \frac{U_c^i U_f^j U_s^k}{G(N)} \quad \text{when } \text{mpl} = i+j+k=N$$

$$G(N) = \sum_{\substack{t \in S \\ t=(i,j,k) \\ i+j+k=N}} U_c^i U_f^j U_s^k$$

product form
sum over all states (complex)

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Performance Measures ₍₁₎

CPU utilization for $\text{mpl}=2 \neq \text{CPU relative utilization}$

$$U_c(2) = \sum_{i \geq 1} P_{ijk} = P_{110} + P_{101} + P_{200}$$

$$\sum_{\substack{(i,j,k) \\ i+j+k=2}} U_c^i U_f^j U_s^k = \frac{\sum_{\substack{(i,j,k) \\ i+j+k=2}} U_c^i U_f^j U_s^k}{G(2)} = \frac{\sum_{\substack{(i,j,k) \\ i+j+k=1}} U_c^i U_f^j U_s^k}{G(2)} = \frac{U_c G(1)}{G(2)}$$

$$U_c(N) = U_c \frac{G(N-1)}{G(N)}$$

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Performance Measures (contd) (3)

$$U_c(N) = U_c \frac{G(N-1)}{G(N)}$$

$$X_c(N) \stackrel{\text{Little}}{=} U_c(N) \mu_c = \mu_c U_c \frac{G(N-1)}{G(N)}$$

$$\bar{n}_c(2) = \sum_{\substack{i,j,k \\ i>0}} i P_{ijk} = \sum_{i=1}^N i P_{ijk} + \sum_{i=2}^N i P_{ijk} \stackrel{\text{homework}}{=} U_c^1 \frac{G(1)}{G(2)} + U_c^2 \frac{G(0)}{G(2)} \quad (7-1)$$

In general, $\bar{n}_c(N) = \sum_{m=1}^N U_c^m \frac{G(N-m)}{G(N)}$

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Performance Measures (contd) (3)

mean device response time:

$$R_c(N) \stackrel{\text{Little}}{=} \frac{\bar{n}_c(N)}{X_c(N)} = \frac{\sum_{m=1}^N U_c^{m-1} G(N-m)}{\mu_c G(N-1)}$$

mean device residence time:

$$R'_c(N) = V_c R_c(N)$$

mean system response time:

$$R(N) = \sum_c R'_c(N) = \sum_c V_c R_c(N)$$

mean system throughput:

$$X_0(N) \stackrel{\text{FFL}}{=} \frac{X_c(N)}{V_c(N)}$$

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How to Compute Relative U_c 's and G (2)

in steady state:

$$\text{vector } X_i \quad \text{matrix } P_{ij} = \text{vector } X_i$$

transition matrix

relative flow through each device

Solve for X , use Little to get U_c 's:

$$U_c^{\text{Little}} = X_c S_c = \frac{X_c}{\mu_c}$$

(many solutions for relative X 's)

And, finally get

$$G(N) = \sum_{\substack{t \in S \\ t=(i,j,k) \\ i+j+k=N}} U_c^i U_f^j U_s^k$$

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$$(X_c, X_f, X_s) \begin{bmatrix} 0 & p & 1-p \\ 1 & 0 & 0 \\ 1 & 0 & 0 \end{bmatrix} = (X_c, X_f, X_s)$$

$X_f + X_s = X_c$	$X_c = \mu_c$	Example (4) (relative X_i , relative U_i)
$pX_c = X_f$	$X_f = \mu_c p$	
$(1-p)X_c = X_s$	$X_s = \mu_c(1-p)$	

$U_c = X_c / \mu_c = 1$	$U_f = \mu_c p / \mu_f = 0.5 \mu_c / \mu_f = 1$ $U_s = \mu_c(1-p) / \mu_s = 0.5 \mu_c / \mu_s = 1.5$
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$$G(2) = \sum_{\substack{i,j,k \\ N=2}} 1^i 1^j 1.5^k = 1.5 + 1.5 + 2.25 = 5.25$$

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State Space Explosion

K	N	$ S $
3	2	$\binom{4}{2} = \frac{4!}{2!2!} = 6$
10	3	$\binom{12}{9} = 220$
3	10	$\binom{12}{2} = 66$
10	10	$\binom{19}{9} = 92378$

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Convolution Algorithm

- How to compute $G(N)$?
 - problem: sum over all states? many states...
 - Buzen: Convolution algorithm

$$G(N) = g(N, K) \quad \begin{matrix} \text{for population } N \\ \text{for } K \text{ devices} \end{matrix}$$

$$\begin{aligned} g(2,3) &= \sum_{\substack{t=(i,j,k) \\ |t|=2}} U_c^i U_f^j U_s^k = \sum_{\substack{t=(i,j,k) \\ |t|=2 \\ i=0}} U_c^i U_f^j U_s^k + \sum_{\substack{t=(i,j,k) \\ |t|=2 \\ i>0}} U_c^i U_f^j U_s^k \\ &= \sum_{\substack{t=(0,j,k) \\ |t|=2}} U_c^i U_f^j U_s^k + U_c \sum_{\substack{t=(i,j,k) \\ |t|=1}} U_c^i U_f^j U_s^k = g(2,2) + U_c g(1,3) \end{aligned}$$

– Fig 5.4 [Men 94]

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Convolution Example

Use D_i 's as (relative) U_i 's:

	V_i	S_i	$U_i = V_i S_i$
CPU	1	10	10
Fast	0.5	20	10
Slow	0.5	30	15

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Convolution Example (1/3)

g	CPU	Fast	Slow	
	10	10	15	
0	1	1	1	$G(0)$
1	10	20	35	$G(1) = 20 + 1 * 15$
2				
3				
...				

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Convolution Example (2/3)

g	CPU	Fast	Slow	
	10	10	15	
0	1	1	1	
1	10	20	35	
2	100	300	825	$G(2) = 300 + 15 * 35$
3				
...				

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Convolution Example (3/3)

g	CPU	Fast	Slow	
	10	10	15	
0	1	1	1	
1	10	20	35	
2	100	300	825	
3	1000	4000	16375	G(3)= 4000+ 15*825
...				

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Response Time

mean device response time

$$R_c(2) = \frac{U_c^0 G(1) + U_c^1 G(0)}{\mu_c G(1)} = \frac{35 + 10}{\frac{1}{10} 35} = \frac{45 * 10}{35} = 12.9 \text{ sec}$$

$$R_f(2) = \frac{U_f^0 G(1) + U_f^1 G(0)}{\mu_f G(1)} = \frac{35 + 10}{\frac{1}{20} 35} = \frac{45 * 20}{35} = 25.7 \text{ sec}$$

$$R_s(2) = \frac{35 + 15}{\frac{1}{30} 35} = \frac{50 * 30}{35} = 42.9 \text{ sec}$$

$$R(2) = \sum R_i(2) = \sum V_i R_i(2) = 47.2 \text{ sec}$$

mean system response time

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Utilization & Throughput ₍₂₎

$$U_c(2) = U_c \frac{G(1)}{G(2)} = 10 \frac{35}{825} = 0.424$$

$$U_f(2) = 10 \frac{35}{825} = 0.424$$

$$U_s(2) = 15 \frac{35}{825} = 0.636$$

true utilization
(not relative utilization)

$$X_c(2) = \mu_c U_c \frac{G(1)}{G(2)} = V_c \frac{G(1)}{G(2)} = 1 * \frac{35}{825} = 0.042$$

$$\text{Little: } X_0(2) R(2) = \frac{X_c(2)}{V_c} R(2) = \frac{0.042}{1} 47.2 = 1.98 \approx 2 \quad \text{OK}$$

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Average Number of Jobs at Each Server (in Queue and in Service)

Job distribution in network

$$\bar{n}_c(2) = U_c \frac{G(1)}{G(2)} + U_c^2 \frac{G(0)}{G(2)}$$

$$= 10 * \frac{35}{825} + 100 * \frac{1}{825} = \frac{450}{825} = 0.545$$

$$\bar{n}_f(2) = 0.545$$

$$\bar{n}_s(2) = 15 * \frac{35}{825} + 225 * \frac{1}{825} = \frac{750}{825} = 0.91$$

total = 2 OK

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Device Queue Lengths

Queue = aver. population – utilization:

$$\begin{aligned}\bar{n}_c(2) - U_c(2) &= 0.545 - 0.424 = 0.125 \\ \bar{n}_f(2) - U_f(2) &= 0.545 - 0.424 = 0.125 \\ \bar{n}_s(2) - U_s(2) &= 0.91 - 0.636 = 0.27\end{aligned}$$

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Convolution Summary

- Jeff Buzen
 - founded BGS Corporation with two friends
 - BGS merged to BMC Software in 1998
- Novel practical way to compute normalizing constant for closed queueing networks
- Get all standard measures
 - Q_i, N_i, U_i, R_i, X_i
 - $Q_{ir}, N_{ir}, U_{ir}, R_{ir}, X_{ir}$
 - X_{system}, R_{system}
- Practical computational problem
 - G can overflow or underflow!

(Multiple class
version exists!)

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