

Lecture 7

Analytical Solution Method for Complex Models

Multiple Class Markov Chain
Models
Convolution

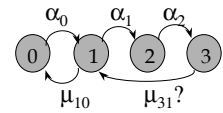
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Markov Chain Solution

- 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 $U = 1 - P_0$

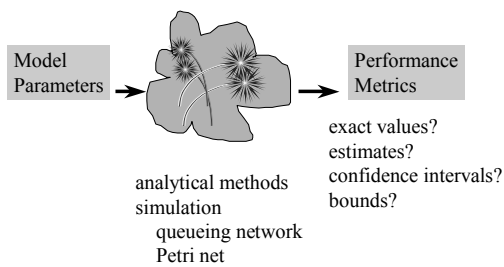


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Generic Solution Method



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Markov Example

- Central server model, Fig. 5.1 [Men 94]
- State = (n_c, n_f, n_s) , $\sum n_i = 2$ (number of jobs in CPU)
- 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

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

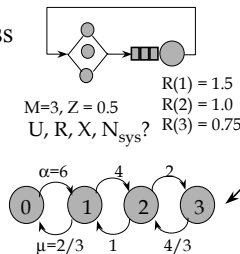
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Markov Chain Solution

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



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

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Markov Example (contd)

- State space diagram, Fig. 5.2
- Transition rates
 - Rate((1,1,0) -> (0,2,0)) = $\mu_f = 3$ (cpu completes, request fast disk)
 - Rate((0,2,0) -> (1,1,0)) = $\mu_c p = 6 * 0.5 = 3$ (fast disk completes)
 - Rate((2,0,0) -> (1,0,1)) = $\mu_c (1-p) = 6 * 0.5 = 3$
 - Rate((1,0,1) -> (2,0,0)) = $\mu_s = 2$
 -

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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

$$\begin{aligned} \mu_c(1-p)P_{110} + \mu_c p P_{101} &= (\mu_s + \mu_f) P_{011} \\ \dots \\ \text{"similarly for some other 4 states"} \\ \dots \\ \sum P_i &= 1 \end{aligned}$$

flow in = flow out

one global balance equation is redundant!

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Relative Utilization (2)

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

Notation:
"Relative Utilization"

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

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

?

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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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Modified Local Balance Eqs (1)

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

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

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

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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
 - can be tricky
 - can be time consuming
 - "the miracle occurs here"
- from transition probabilities and service rates

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Normalizing Constant (2)

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

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

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

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 (contd) ⁽³⁾

mean device response time: $R_c(N) = \frac{\bar{n}_c(N)}{X_c(N)} = \frac{\sum_{m=1}^N U_c^{m-1} G(N-m)}{\mu_c G(N-1)}$

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

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Performance Measures ⁽¹⁾CPU utilization for mpl=2 $\neq$ CPU relative utilization

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

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

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

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

in steady state: $\text{vector } \underline{X}_i \quad \text{matrix } P_{ij} = \text{vector } \underline{X}_i$

transition matrix $\underline{U}_c = \underline{X}_c S_c = \frac{X_c}{\mu_c}$

Solve for $\underline{X}$, use Little to get U_c 's:

(many solutions for relative $\underline{X}$'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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Performance Measures (contd) ⁽³⁾

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

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

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

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

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

Example ⁽⁴⁾

(relative X_p , relative U_i)

$$\begin{aligned} X_f + X_s &= X_c & X_c &= \mu_c \\ p X_c &= X_f & X_f &= \mu_c p \\ (1-p) X_c &= X_s & X_s &= \mu_c (1-p) \end{aligned}$$

$$\begin{aligned} U_c &= X_c / \mu_c = 1 \\ U_f &= \mu_c p / \mu_f = 0.5 \mu_c / \mu_f = 1 \\ U_s &= \mu_c (1-p) / \mu_s = 0.5 \mu_c / \mu_s = 1.5 \end{aligned}$$

$$G(2) = \sum_{\substack{i,j,k \\ N=2}} 1^i 1^j 1.5^k = 1.5 + 1.5 + 2.25 = 5.25$$

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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 Example

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

	V_i	S_i	$U_i=V_iS_i$
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	$\leftarrow G(0)$
1	10	20	35	$\leftarrow G(1)=20+1*15$
2				
3				
...				

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Convolution Algorithm

- How to compute $G(N)$?
 - problem: sum over all states? many states...
 - Buzen: Convolution algorithm
- $G(N)=g(N,K)$ for population N for K devices

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

– Fig 5.4 [Men 94]

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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	$\leftarrow G(2)=300+15*35$
3				
...				

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Convolution Example ^(3/3)

g	CPU 10	Fast 10	Slow 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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Average Number of Jobs at Each Server (in Queue and in Service)

Job distribution in network

$$\begin{aligned}\bar{n}_c(2) &= U_c \frac{G(1)}{G(2)} + U_c^2 \frac{G(0)}{G(2)} \\ &= 10 * \frac{35}{825} + 100 * \frac{1}{825} = \frac{450}{825} = 0.545 \\ \bar{n}_f(2) &= 0.545 \\ \bar{n}_s(2) &= 15 * \frac{35}{825} + 225 * \frac{1}{825} = \frac{750}{825} = 0.91\end{aligned}$$

total = 2 OK

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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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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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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)

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

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

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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
 - Qi, Ni, Ui, Ri, Xi
 - Qir, Nir, Uir, Rir, Xir
 - Xsystem, Rsystem
- Practical computational problem
 - G can overflow or underflow!

(Multiple class
version exists!)

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