# Lecture 7 : Networks of queues 

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## Open/closed queuing network



## Relative throughputs and utilities

$\square$ Using routing chain, we obtain relative throughputs, visit ratios, relative utilities


$$
\vec{\lambda} P=\vec{\lambda}
$$

$\square \lambda_{j}=\lambda_{1} P_{1 j}+\lambda_{2} P_{2 j}+\ldots+\lambda_{i} P_{i j}+\lambda_{j} P_{i j}+\ldots+\lambda_{m} P_{m j}$ (if there exist $m$ queues)

## Relative throughput, $r_{i}$, and visit ratio, $v_{i}$

ㅁ (ex)



$$
\boldsymbol{V}_{i}=\frac{\lambda_{i}}{\lambda_{1}}=\frac{\boldsymbol{r}_{i}}{\boldsymbol{r}_{\mathbf{1}}} \quad \text { (larger throughput implies more }
$$

## Relative utility, $\boldsymbol{u}_{\boldsymbol{i}}$

$$
\begin{aligned}
& \rho_{i}=\frac{\lambda_{i}}{\mu_{i}}, \lambda_{i}=\mu_{i} \rho_{i} \Leftrightarrow \boldsymbol{r}_{i}=\mu_{i} \boldsymbol{u}_{i} \\
& \sum_{i=1}^{M} \boldsymbol{r}_{i} \boldsymbol{P}_{i j}=\boldsymbol{r}_{j} \Leftrightarrow \sum_{i=1}^{M} \mu_{i} \boldsymbol{u}_{i} \boldsymbol{P}_{i j}=\mu_{j} \boldsymbol{u}_{j} \\
& \left.\boldsymbol{u}_{i}=\frac{\boldsymbol{r}_{i}}{\mu_{i}} \text { (here } \boldsymbol{r}_{i} \text { is solved, } \mu_{i} \text { is given }\right)
\end{aligned}
$$

## M $\rightarrow$ M property

$\square$ To study the output process of a queue
$\square$ Output process $=f($ input process, queuing discipline, service process)
$\square$ A queue has the $M \rightarrow M$ property if when the input is a Poisson process, so is the output. Then timing and order may be different (jobs may be delayed and priority orders become different) but both input and output have same Poisson parameter ( $\lambda$ )

## FCFS M/M/1 queue


(Inter-departure time PDF, $d(t)$ )
$D^{*}(s \mid b u s y)=\frac{\mu}{\mu+s} \quad ;\left(\right.$ ex) $d_{2}=x_{2}$ (when $C_{2}$ arrives, server is busy)
$D^{\prime \prime}(s \mid$ idle $)=\frac{\mu}{\mu+s} \cdot \frac{\lambda}{\lambda+s} \underset{~(\text { ex })}{\mathbf{d}_{3}=\mathbf{r}_{3}+\mathbf{x}_{\mathbf{3}}\left(\text { when } \mathbf{C}_{3} \text { arrives, server is idle) }\right) ~}$
$D^{*}(s)=\rho \frac{\mu}{\mu+s}+(1-\rho)\left[\frac{\mu}{\mu+s} \frac{\lambda}{\lambda+s}\right]=\frac{\lambda}{\lambda+s}$
So $M / \mathrm{M} / 1$ queue has the $M \Rightarrow$ M property

## Easier checking for $\mathbf{M} \rightarrow \mathbf{M}$ property

$\square$ Without deriving the output process
$\square$ At steady-state

$\pi_{\mathrm{n}+1} \mathrm{R}(\mathrm{n}+1 \rightarrow \mathrm{n})=\pi_{\mathrm{n}} \lambda, \quad \forall \mathrm{n}$
( R is the rate of departure)
"No memory in the departure process indicates that even though the departure rate in any state is higher than the arrival rate, it cannot be compensated by a $\qquad$ number state"
$\square$ Checking is done by obtaining the steady-state probability relationship and then substituting it into the state balance equation to check the consistency

## Easier checking for $\mathbf{M} \rightarrow \mathbf{M}$ property

## (ex) $M / M / \infty$ queue

$$
\mathbf{R}(\mathbf{n}+\mathbf{1} \rightarrow \mathbf{n})=(\mathbf{n}+\mathbf{1}) \mu
$$



$$
\pi_{n+1}=\frac{\lambda}{(\mathbf{n}+\mathbf{1}) \mu} \pi_{\mathrm{n}}
$$

$$
\pi_{1}=\frac{\lambda}{\mu} \pi_{0}, \pi_{2}=\frac{\lambda}{2 \mu} \pi_{1}=\frac{\lambda^{2}}{2 \mu^{2}} \pi_{0} \ldots \ldots . . . \pi_{\mathrm{n}}=\left(\frac{\lambda}{\mu}\right)^{\mathrm{n}} \frac{\mathbf{1}}{\mathrm{n}!} \pi_{0}
$$

check

$$
\begin{aligned}
& (\lambda+\mathbf{n} \mu) \pi_{\mathrm{n}}=(\mathbf{n}+1) \mu \pi_{\mathrm{n}+1}+\lambda \pi_{\mathrm{n}-1} \\
& (\lambda+\mathbf{n} \mu)\left(\frac{\lambda}{\mu}\right)^{\mathrm{n}} \frac{1}{\mathrm{n}!} \pi_{0}=(\mathbf{n}+1) \mu\left(\frac{\lambda}{\mu}\right)^{\mathrm{n}+1} \frac{\pi_{0}}{(\mathrm{n}+1)!}+\lambda\left(\frac{\lambda}{\mu}\right)^{\mathrm{n}-1} \frac{\pi_{0}}{(\mathrm{n}-1)!} \\
& =\left(\frac{\lambda}{\mu}\right)^{\mathrm{n}} \frac{1}{\mathrm{n}!} \pi_{0}(\lambda+\mathbf{n} \mu)
\end{aligned}
$$

## Feedback


$\square$ Even with Poisson I/O ( $\mathbf{M} \Rightarrow \mathbf{M}$ ), a feedback changes the process to hyperexponential

$$
\lambda=p \lambda^{\prime} \quad \lambda^{\prime}=\frac{\lambda}{p} \quad \rho=\frac{\lambda^{\prime}}{\mu}=\frac{\lambda}{p \mu}
$$

$\square$ For a stable queue, $\rho<1 \rightarrow \lambda<p \mu$

## Local balance

$\square$ State of a network is the $\qquad$ of the states of the individual queues
$\square$ Jackson's theorem: Joint PDF of the number of customers in each queue in the network in steady-state is the $\qquad$ of the marginal PDF's irrespective of the type of input processes to the queues (r.v. of the number of customers are $\qquad$ )
$\square$ Local balance: the property allowing the product form of the solution

## Local balance

## (ex) $\mathbf{2}$ jobs circulating in $M / M / m$ queueing network



## Local balance

$\square$ Analysis using global balance via $\pi \mathrm{Q}=0$

$\square$ Analysis using local balance by cutting 2 edges per state that completely isolates all the states, and check the balances $\rightarrow$ stronger requirement than global balance check

## Local balance



## Local balance

$$
\begin{aligned}
& \pi_{\mathbf{2 0 0}}=\frac{200 \rightarrow \frac{1}{\mu_{1}^{2}}}{\frac{1}{\mu_{1}^{2}}+\frac{1}{\mu_{1} \mu_{3}}+\frac{1}{\mu_{1} \mu_{2}}+\frac{1}{\mu_{2} \mu_{3}}+\frac{1}{\mu_{2}^{2}}+\frac{1}{\mu_{3}^{2}}}=\frac{\frac{1}{\mu_{1}^{2}}}{G(2)} \\
& \pi_{110}=\frac{1}{\mu_{1} \frac{1}{\mu_{2}}, \ldots,}-\quad-\begin{array}{l}
\pi_{200}+\pi_{110}+\pi_{020}+\pi_{101}+\pi_{011}+\pi_{002}=1 \\
\pi_{200}+\left(\mu_{1} / \mu_{2}\right) \pi_{200}+\left(\mu_{3} / \mu_{2}\right)^{2} \pi_{200}+\left(\mu_{1} / \mu_{3}\right) \\
\pi_{200}+\left(\mu_{1}^{2} / \mu_{2} \mu_{3}\right) \pi_{200}+\left(\mu_{1}^{2} / \mu_{2} \mu_{3}\right) \pi_{200}=1
\end{array} \\
& \pi_{\mathbf{n}_{1} \mathbf{n}_{2} \mathbf{n}_{\mathbf{3}}}=\frac{\left(\frac{\mathbf{1}}{\mu_{1}}\right)^{\mathbf{n}_{1}}\left(\frac{\mathbf{1}}{\mu_{2}}\right)^{\mathbf{n}_{\mathbf{2}}}\left(\frac{\mathbf{1}}{\mu_{3}}\right)^{\mathbf{n}_{3}}}{\mathbf{G ( 2 )}}, \mathbf{n}_{\mathbf{i}}=\mathbf{0 , 1 , 2 \&} \sum_{\mathbf{i}} \mathbf{n}_{\mathbf{i}}=\mathbf{2}
\end{aligned}
$$

$\square \pi_{n_{1} n_{2} n_{3}}$ will satisfy the global balance equations
(ex) Check with $\left(\mu_{1}+\mu_{2}\right) \pi_{110}=\mu_{1} \pi_{200}+\mu_{3} \pi_{011}$

## Product form solution

$\square$ Local balance solution technique used in networks is similar to the technique used for birth-death systems
$\square$ General B-D queue: $\pi_{k}=\left(\prod_{j=0}^{k-1} \frac{\lambda_{j}}{\mu_{j+1}}\right) \pi_{0}$
M/M/1 queue: $\pi_{k}=\rho^{k} \pi_{0}=\rho^{k}(1-\rho)$
M/M/ $\infty$ queue: $\pi_{k}=\frac{\rho^{k}}{k!} \pi_{0}=\frac{\rho^{k}}{k!} e^{-\frac{\lambda}{\mu}}$

## Product form solution

$\square$ General queueing network of single server queues satisfying local balance

$$
\pi_{n_{1} n_{2} \ldots n_{M}}=\frac{\rho_{1}^{n_{1}} \rho_{2}^{n_{2}} \ldots \rho_{M}^{n_{M}}}{C}(\mathbb{C} \text { is normalization constant for making it a PDF })
$$

$\square$

$$
C=\left\{\begin{array}{l}
\left.\frac{1}{\pi_{00 . . .0}} \text { (for open } \mathrm{NWs}\right) \\
P[\mathbf{N} \text { customers in } \mathrm{NW}](\text { for closed } \mathrm{NWs})
\end{array}\right.
$$

## Product form solution

$\square$ Open NW of M/M/1 queues
$\pi_{n_{1} n_{2} \ldots \mathrm{n}_{\mathrm{M}}}=\underbrace{\left(1-\rho_{1}\right) \rho_{1}^{\mathrm{n}_{1}}(\overbrace{1-\rho_{2}}) \rho_{2}^{\mathbf{n}_{2}} \ldots\left(1-\rho_{M}\right)}_{-}) \rho_{\mathrm{M}}^{\mathrm{n}_{\mathrm{M}}}$
(calculated from $\qquad$ , as $\lambda_{M} / \mu_{M}$ )
$\square$ Open NW of M/M/S $\mathbf{S}_{\mathbf{i}}$ queues
$S_{i}$ : no. of servers in queue-i

$$
\mathbf{f}_{\mathrm{i}}\left(\mathbf{n}_{\mathbf{i}}\right) \equiv\left\{\begin{array}{l}
\frac{\mathbf{n}_{i}!}{\mathbf{S}_{\mathrm{i}}^{n_{i}}} \mathbf{n}_{\mathrm{i}}<\mathbf{S}_{\mathrm{i}} \\
\frac{\mathbf{S}_{i}!}{\mathbf{S}_{\mathrm{i}}!} \mathbf{n}_{\mathrm{i}} \geq \mathbf{S}_{\mathrm{i}}
\end{array} \quad \pi_{n_{1} \mathbf{n}_{2} . . \mathbf{n}_{\mathrm{M}}}=\frac{\rho_{1}^{\mathbf{n}_{1}} \rho_{2}^{\mathbf{n}_{2}} \ldots . . \rho_{\mathrm{M}}^{\mathbf{n}_{\mathrm{M}}}}{\mathbf{f}_{1}\left(\mathbf{n}_{1}\right) \mathbf{f}_{2}\left(\mathbf{n}_{2}\right) . . \mathbf{f}_{\mathrm{M}}\left(\mathbf{n}_{\mathrm{M}}\right) \mathbf{C}}\right.
$$

## Product form solution summary

$\square$ Short cut solutions (product form) possible with local balance NWs
$\square$ Any NW composed of queues, each having the $M \Rightarrow M$ property, will satisfy local balance
$\square$ NWs constructed from $M \Rightarrow$ M queues permit product form solutions

The queue types yielding product form NWs
$\square$ FCFS with exp. dist. service time.
$\square$ LCFS preemptive-resume, processor sharing, infinite server with Coxian distribution service time

## Solving open networks

(ex)


Avg. no. of customers in the queue

$$
\mathbf{E}[\mathbf{n}]=\mathbf{N}_{\mathbf{1}}=\frac{\rho_{1}}{1-\rho_{1}}=\frac{\frac{\lambda_{1}}{\mu_{1}}}{1-\frac{\lambda_{1}}{\mu_{1}}}=\frac{\frac{\lambda}{\mu_{1}(1-p)}}{1-\frac{\lambda}{\mu_{1}(1-p)}}=\frac{\lambda}{\mu_{1}(1-p)-\lambda}
$$

## Solving open networks

$T_{1}=$ Avg. time in the queue considering only the queue and server

$$
=\frac{\mathbf{N}_{1}}{\lambda_{1}}=\frac{\lambda}{\frac{\lambda}{1-\mathbf{p}}\left(\mu_{1}(\mathbf{1}-\mathbf{p})-\lambda\right)}=\frac{\mathbf{1}}{\mu_{1}-\frac{\lambda}{1-\mathbf{p}}}=\frac{\mathbf{1}-\mathbf{p}}{\mu_{1}(\mathbf{1}-\mathbf{p})-\lambda}
$$

$T_{1}=$ Avg. time in the queue considering the feedback also

$$
=\frac{\mathbf{N}_{1}}{\lambda}=\frac{\mathbf{1}}{\mu_{1}(\mathbf{1}-\mathbf{p})-\lambda}
$$

## Solving open networks

$T_{1}=$ Avg. time in the queue considering only the queue and server

$$
=\frac{\mathbf{N}_{1}}{\lambda_{1}}=\frac{\lambda}{\frac{\lambda}{1-\mathbf{p}}\left(\mu_{1}(\mathbf{1}-\mathbf{p})-\lambda\right)}=\frac{\mathbf{1}}{\mu_{1}-\frac{\lambda}{1-\mathbf{p}}}=\frac{\mathbf{1}-\mathbf{p}}{\mu_{1}(\mathbf{1}-\mathbf{p})-\lambda}
$$

T = Avg. time in the queue considering the feedback also

$$
=\frac{\mathbf{N}_{1}}{\lambda}=\frac{1}{\mu_{1}(1-\mathbf{p})-\lambda}
$$

## Solving open networks



Clerk 2
$\square$ FCFS queue with exp. dist. service

$$
\begin{aligned}
& \mu_{1}=\frac{1}{20 \mathrm{sec}}, \mu_{2}=\frac{1}{10 \mathrm{~min}}=\frac{1}{600 \mathrm{sec}} \\
& \mu_{3}=\frac{1}{5 \mathrm{~min}}=\frac{1}{300 \mathrm{sec}}, \mu_{4}=\frac{1}{1 \mathrm{~min}}=\frac{1}{60 \mathrm{sec}}
\end{aligned}
$$

## Solving open networks

$$
\begin{aligned}
& \lambda_{1}=\lambda, \rho_{1}=\frac{\lambda}{\mu_{1}}=20 \lambda \\
& \lambda_{2}=0.3 \lambda+0.1 \lambda_{3} \quad \lambda_{2}=0.38 \lambda, \rho_{2}=\frac{0.38 \lambda}{\frac{1}{600}}=228 \lambda \\
& \lambda_{3}=0.7 \lambda+0.2 \lambda_{2} \int \lambda_{3}=0.78 \lambda, \rho_{3}=\frac{0.78 \lambda}{\frac{1}{300}}=234 \lambda \\
& \lambda_{4}=0.8 \lambda_{2}+0.9 \lambda_{3}=\lambda, \rho_{4}=\frac{\lambda}{\frac{1}{60}}=60 \lambda
\end{aligned}
$$

## Solving open networks

$$
\begin{aligned}
& \rho_{\mathrm{n}}<1 \rightarrow \lambda<\frac{1}{234}<15.38 / \text { hour } \\
& \pi_{n_{1} n_{2} n_{3} n_{4}}=\left(1-\rho_{1}\right) \rho_{1}^{n_{1}}\left(1-\rho_{2}\right) \rho_{2}^{n_{2}}\left(1-\rho_{3}\right) \rho_{3}^{n_{3}}\left(1-\rho_{4}\right) \rho_{4}^{n_{4}}
\end{aligned}
$$

$E\left[n_{1}\right]=\frac{\rho_{1}}{1-\rho_{1}}=\frac{20 \lambda}{1-20 \lambda} \leq 0.093$ for $\lambda \leq 0.00427 / \mathrm{sec}$

$$
E\left[n_{2}\right]=\frac{\rho_{2}}{1-\rho_{2}}=\frac{228 \lambda}{1-228 \lambda} \leq 36.82
$$

## Solving open networks

$$
\begin{aligned}
& E\left[n_{3}\right]=\frac{\rho_{3}}{1-\rho_{3}}=\frac{234 \lambda}{1-234 \lambda} \leq \infty \\
& E\left[n_{4}\right]=\frac{\rho_{4}}{1-\rho_{4}}=\frac{60 \lambda}{1-60 \lambda} \leq 0.34 \quad N=\sum_{j=1}^{4} E\left[n_{j}\right]
\end{aligned}
$$

$\square$ Avg. time to process,

$$
T=\frac{N}{\lambda} \quad(=23 \min \text { for } \lambda=10 / \text { hour })
$$

$\square$ Avg. time to visit queue 1,3, and then 4,

$$
\mathrm{T}_{\mathrm{R}}=\frac{\mathbf{E}\left[\mathbf{n}_{1}\right]}{\lambda_{1}}+\frac{\mathbf{E}\left[\mathbf{n}_{3}\right]}{\lambda_{3}}+\frac{\mathbf{E}\left[\mathbf{n}_{4}\right]}{\lambda_{4}}
$$

## Solving closed networks

$\square$ Need utilization and normalization constant (sum of all combination of utilization)
$\square$ Use relative utilization

- $r_{m}=u_{m} \mu_{m}=\alpha \lambda_{m}$
$\square$ Normalization constant (with N customers in the network)



## Solving closed networks

$\square$ Prob. that queue-m has $n$ or more customers
$\square \mathrm{P}\left[\mathrm{n}_{\mathrm{m}} \geq \mathrm{n}\right]=\mathbf{u}_{\mathrm{m}}^{\mathrm{n}} \frac{\mathbf{G}(\mathbf{N}-\mathbf{n})}{\mathbf{G}(\mathbf{N})}$
( $n$ customers are already in queue- $m$, while the remaining ( $N-n$ ) are $\qquad$ )
$\square$ Solution involves
$\square$ Relative utility ( $\mathrm{u}_{\mathrm{m}}$ )
$\square$ Relative throughput ( $\mathrm{r}_{\mathrm{m}}$ )

- Normalization constant (G(N))


## Solving closed networks

$\square \mathbf{r}_{\mathrm{m}}=\mathbf{u}_{\mathrm{m}} \mu_{\mathrm{m}}=\alpha \lambda_{\mathrm{m}}$
$\square \alpha$ : scalar required to change relative to absolute
$\square \rho_{m}=u_{m} \frac{G(N-1)}{G(N)} \quad \lambda_{m}=r_{m} \frac{G(N-1)}{G(N)}=\mu_{m} u_{m} \frac{G(N-1)}{\mathbf{G}(\mathbf{N})}$
$\mathbf{E}[\mathbf{n}]=\sum_{\mathrm{n}=1}^{\infty} \mathbf{n P [ n ]}=\sum_{\mathrm{n}=1}^{\infty} \sum_{k=\mathrm{n}}^{\infty} \mathbf{P}[k]=\sum_{\mathrm{n}=1}^{\infty} \mathbf{P}[k \geq \mathbf{n}]$
$\mathbf{E}\left[\mathbf{n}_{\mathrm{m}}\right]=\sum_{\mathrm{n}=1}^{\mathrm{N}} \mathbf{u}_{\mathrm{m}}^{\mathrm{n}} \frac{\mathbf{G}(\mathbf{N}-\mathbf{n})}{\mathbf{G}(\mathbf{N})}$

## Generating the normalization constant

$\square$ Generating the normalization constant - (Convolution method)

- By Buzen
$\square$ Partial sum including the terms for $\mathrm{n} \leq \mathrm{N}$ customers in $\mathrm{m} \leq \mathrm{M}$ of the queues

$$
\begin{aligned}
& \mathbf{g}(\mathbf{n}, \mathbf{m})=\sum_{\sum_{i=1}^{m} n_{i} n_{i}=n} \prod_{i=1}^{m} \mathbf{u}_{i}^{n_{i}} \\
& =\left(\mathbf{u}_{1}^{\mathrm{n}}+\mathbf{u}_{1}^{\mathrm{n}-1} \mathbf{u}_{2}+\ldots+\mathbf{u}_{1}^{\mathrm{n}-1} \mathbf{u}_{\mathrm{m}}+\mathbf{u}_{1}^{\mathrm{n}-2} \mathbf{u}_{2}^{2}+\ldots+\mathbf{u}_{\mathrm{m}}^{\mathrm{n}}\right) \\
& \mathbf{g}(\mathbf{n}-1, \mathbf{m})=\left(\mathbf{u}_{1}^{\mathrm{n}-1}+\mathbf{u}_{1}^{\mathrm{n}-2} \mathbf{u}_{2}+\ldots+\mathbf{u}_{1}^{\mathrm{n}-2} \mathbf{u}_{\mathrm{m}}+\mathbf{u}_{1}^{\mathrm{n}-3} \mathbf{u}_{2}^{2}+\ldots+\mathbf{u}_{\mathrm{m}}^{\mathrm{n}-1}\right) \\
& \mathbf{g}(\mathbf{n}, \mathbf{m}-1)=\left(\mathbf{u}_{1}^{\mathrm{n}}+\mathbf{u}_{1}^{\mathrm{n}-1} \mathbf{u}_{2}+\ldots+\mathbf{u}_{1}^{\mathrm{n}-1} \mathbf{u}_{\mathrm{m}-1}+\mathbf{u}_{1}^{\mathrm{n}-2} \mathbf{u}_{2}^{2}+\ldots+\mathbf{u}_{\mathrm{m}-1}^{\mathrm{n}-1}\right) \\
& g(n, m)=g(n, m-1)+u_{m} g(n-1, m) \\
& \mathbf{g}(\mathbf{n}, \mathbf{0})=\mathbf{0}, \mathbf{g}(\mathbf{0}, \mathrm{m})=\mathbf{1}, \mathbf{g}(\mathbf{n}, \mathbf{1})=\mathbf{u}_{\mathbf{1}}{ }^{\mathbf{n}}, \mathbf{G}(\mathbf{N})=\mathbf{g}(\mathbf{N}, \mathbf{M})
\end{aligned}
$$

## Generating the normalization constant

| n\m | 0 |  | 1 | 2 |  |  |  | m |  |  |  | M |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0 | 0 |  | 1 | 1 | - |  | - | 1 | - |  | . | 1 |
| 1 | 0 |  | g(1,1) | $\mathrm{g}(1,2)$ | - | - |  | g(1,m) | - |  | - | G(1) |
| 2 | 0 |  | $\mathrm{g}(2,1)$ | $\mathrm{g}(2,2)$ | - | - | - | g(2,m) | - | - | - | G(2) |
| . |  |  |  |  |  |  |  |  |  |  |  |  |
| - |  |  |  |  |  |  |  |  |  |  |  |  |
| - |  |  |  |  |  |  |  |  |  |  |  |  |
| n | 0 |  | g(n,1) | $\mathrm{g}(\mathrm{n}, 2)$ | - | - | - | g(n,m) | - |  | - | G(n) |
| . |  |  |  |  |  |  |  |  |  |  |  |  |
| - |  |  |  |  |  |  |  |  |  |  |  |  |
| - |  |  |  |  |  |  |  |  |  |  |  |  |
| N | 0 |  | $\mathrm{g}(\mathrm{N}, 1)$ | $\mathrm{g}(\mathrm{N}, 2)$ | . | - | . | $\mathbf{g}(\mathbf{N}, \mathrm{m})$ | . | . | . | G(N) |

$$
(\mathrm{ex}) g(2,2)=g(2,1)+u_{2} g(1,2)
$$

## Generating the normalization constant

$\square(e x)$


Queue-1: processor

Queue-2: 5.25' HD

Queue-3: 8' HD
$28 \mathrm{msec} \quad \mu_{1}=\frac{1}{0.028 \mathrm{sec}}$
$40 \mathrm{msec} \quad \mu_{2}=\frac{1}{0.04 \mathrm{sec}}$
$280 \mathrm{msec} \mu_{3}=\frac{1}{0.28 \mathrm{sec}}$

$$
\begin{array}{ll}
\mu_{2} \mathbf{u}_{2}=0.7 \mu_{1} \mathbf{u}_{1} & \mu_{3} u_{3}=0.2 \mu_{1} \mathbf{u}_{1} \\
\frac{1}{40} \mathbf{u}_{2}=0.7 \frac{1}{28} \mathbf{u}_{1} & \frac{1}{280} \mathbf{u}_{3}=0.2 \frac{1}{28} \mathbf{u}_{1} \\
\mathbf{u}_{2}=\mathbf{u}_{1} & \mathbf{u}_{3}=2 \mathbf{u}_{1}
\end{array}
$$

## Generating the normalization constant

$\square \mathbf{g}(\mathbf{n}, \mathrm{m})=\mathbf{g}(\mathbf{n}, \mathbf{m}-1)+\mathbf{u}_{\mathrm{m}} \mathbf{g}(\mathbf{n}-\mathbf{1}, \mathbf{m})$

$$
\mathbf{u}_{1} \rightarrow 1 ; \mathbf{u}_{2}=\mathbf{u}_{1}=1 ; \mathbf{u}_{3}=2 \mathbf{u}_{1}=2
$$

| $\mathbf{n} \backslash \mathrm{m}$ | 0 | 1 | 2 | 3 |
| :--- | :--- | :--- | :--- | :--- |
| 0 | 0 | 1 | 1 | $1=\mathbf{G}(0)$ |
| 1 | 0 | 1 | 2 | 4 |
| 2 | 0 | 1 | 3 | 11 |
| 3 | 0 | 1 | 4 | 26 |
| 4 | 0 | 1 | 5 | 57 |
| 5 | 0 | 1 | 6 | 120 |
| 6 | 0 | 1 | 7 | 247 |
| 7 | 0 | 1 | 8 | $502=\mathbf{G ( 7 )}$ |

$$
\begin{aligned}
& \mathrm{g}(1,1)=0+\mathrm{u}_{1} * 1=\mathrm{u}_{1}=1 \\
& \mathrm{~g}(1,2)=\mathrm{u}_{1}+\mathrm{u}_{2} * 1=2 \mathrm{u}_{1}=2 \\
& \mathrm{~g}(1,3)=\mathrm{u}_{1}+\mathrm{u}_{2}+\mathrm{u}_{3} * 1=4 \mathrm{u}_{1}=4 \\
& \mathrm{~g}(2,1)=0+\mathrm{u}_{1} * 1=1 \\
& \mathrm{~g}(2,2)=1+\mathrm{u}_{2} * 2=3 \\
& \mathrm{~g}(2,3)=3+\mathrm{u}_{3} * 4=11 \\
& \mathrm{~g}(3,1)=0+\mathrm{u}_{1} * 1=1 \\
& \mathrm{~g}(3,2)=1+\mathrm{u}_{2}^{*} * 3=4 \\
& \mathrm{~g}(3,3)=4+\mathrm{u}_{3} * 11=26
\end{aligned}
$$

## Generating the normalization constant

$$
\begin{aligned}
& \mathrm{N}=7 \\
& \rho_{1}=u_{1} \frac{G(6)}{G(7)}=1 * \frac{\mathbf{2 4 7}}{\mathbf{5 0 2}}=0.492, \rho_{2}=\rho_{1}, \rho_{3}=u_{3} \frac{\mathbf{2 4 7}}{502}=0.984 \\
& \lambda_{1}=\rho_{1} \mu_{1}=0.492 * \frac{1}{0.028}=17.57 / \mathrm{sec} \\
& \lambda_{2}=12.3 / \mathrm{sec}, \lambda_{3}=3.51 / \mathrm{sec} \\
& \mathbf{E}\left[\mathrm{n}_{\mathrm{m}}\right]=\sum_{\mathrm{n}=1}^{\mathrm{N}} \mathbf{u}_{\mathrm{m}}^{\mathrm{n}} \frac{\mathbf{G ( N - n )}}{\mathbf{G ( N )}}, \mathbf{u}_{1}^{\mathrm{n}}=\mathbf{1} \\
& E\left[n_{1}\right]=E\left[n_{2}\right]=\frac{1+4+\ldots+247}{502}=0.928 \\
& E\left[n_{3}\right]=\frac{\mathbf{2}^{7} \times 1+2^{6} \times 4+\bullet \bullet \bullet 2^{1} \times 247}{502}=5.143 \\
& T_{1}=\frac{E\left[n_{1}\right]}{\lambda_{1}}=\frac{0.928}{17.57}=0.0528 \mathrm{sec}, T_{2}=\frac{\mathrm{E}\left[\mathrm{n}_{2}\right]}{\lambda_{2}}=0.0755, \\
& \mathrm{~T}_{3}=1.465 \\
& \pi_{\mathrm{n} 1, \mathrm{n} 2,7-\mathrm{n} 1-\mathrm{n} 2}=(1 / 502) \mathbf{1}^{\mathrm{n} 1} 1^{\mathrm{n} 2} \mathbf{2}^{7-\mathrm{n} 1-\mathrm{n} 2}
\end{aligned}
$$

## Generating the normalization constant

$\square$ (Ex 7.3) What is the utilization of queue- 1 with 4 customers?

$$
\rho_{1}=u_{1}\left(G(3) / G\left(\__{-}\right)\right)=1 \times 26 / 57=0.456
$$

$\square$ (Ex 7.4) What is the utilization of queue- 3 with 4 customers?

$$
\rho_{3}=\ldots(G(3) / G(4))=2 \times 26 / 57=0.912
$$

$\square$ (Ex 7.5) What is the normalization constant with 8 customers?

| 7 | 1 | 8 | 502 |
| :--- | :--- | :--- | :--- |

$8 \quad 1 \quad 9 \quad=1013$
$\square$ (Ex 7.6) What is the utilization of queue- 1 with 8 customers?

$$
\rho_{1}=u_{1}(G(-) / G(8))=1 \times 502 / 1013=0.496
$$

$\square$ What is the prob. for queue- 1 to have six or more customers with 7 customers in the network?

$$
\begin{aligned}
& \begin{aligned}
\mathrm{P}\left[\mathrm{n}_{1} \geq 6\right]=\ldots & (\mathrm{G}(1) / \mathrm{G}(7))= \\
\mathrm{P}\left[\mathrm{n}_{1} \geq 6\right]=\pi_{610}+\pi_{601}+\ldots & =\left(1^{6} 1^{1} 2^{0}+1^{6} 1^{0} 2^{1}+1^{7} 1^{0} 2^{0}\right) / 502=4 / 502 \\
& \left(=1^{6}\left(1^{0} 1^{1} 2^{0}+1^{0} 1^{0} 2^{1}+1^{1} 1^{0} 2^{0}\right) / 502\right. \\
& \left.=u_{1}^{6}(\mathrm{G}(1) / \mathrm{G}(7))\right)
\end{aligned}
\end{aligned}
$$

## Mean Value Analysis (MVA)

$\square$ The normalization constant method may cause overflow or underflow when automated

$$
\rho_{m}=\mathbf{u}_{\mathrm{m}}(\mathbf{G}(\mathbf{N}-1) / \mathbf{G}(\mathbf{N})) \text {; if } \rho_{\mathrm{m}}>\mathbf{u}_{\mathrm{m}}, \mathbf{G}(\mathbf{N}) \text { gets very ___ ; if } \rho_{\mathrm{m}}<\mathbf{u}_{\mathrm{m}} \text {, very large }
$$

$\square$ One solution is scaling the relative utilization
$\square$ MVA avoids the problem by computing averages. It also computes system parameters for all populations less than the target level as well
$\square$ But it cannot provide specific prob. dist. :
$\mathbf{P}\left[\mathbf{n}_{\mathrm{m}} \geq \mathbf{n}\right]=\mathrm{u}_{\mathrm{m}}^{\mathrm{m}} \frac{\mathbf{G}(\mathbf{N}-\mathrm{n})}{\mathbf{G}(\mathbf{N})}$
$\square$ Little's law is applied to each queue and network as a whole.

## The MV theorem

$\square$ The MV theorem (by Lavenberg and Reiser)
"A customer arriving at a queue sees the same dist. of customers as does an observer outside of the network if one less customer was circulating." (since itself needs to be $\qquad$ )
(Good for product form NWs, approximation for others)

## MVA procedure

$\square$ For n := 1 to N
(* Iterate through NW population to target, $\mathbf{N}^{*}$ )

Form := 1 to M
(* Iterate through all queues *)
(average

time)
$\mathrm{T}_{\mathrm{m}}=$ (* Application of MV theorem *)
$\mathbf{n}_{\mathrm{m}}(\mathbf{0})=0 \quad$ (just $\qquad$
$\lambda=\frac{\mathbf{n}}{\sum_{m=1}^{M} \mathbf{T}_{\mathrm{m}} \mathbf{V}_{\mathrm{m}}}$
$\lambda_{m}=\lambda \mathbf{V}_{\mathrm{m}}$
$\mathbf{n}_{\mathrm{m}}(\mathbf{n})=\lambda_{\mathrm{m}} \mathbf{T}_{\mathrm{m}}$

## MVA procedure

$\square$ (ex) same example as for the convolution method

$$
\mathrm{n}=1
$$

$$
\mathbf{n}_{\mathrm{m}}(0)=0, \mathrm{~T}_{1}=0.028\{1+0\}=0.028 \mathrm{sec}
$$

$$
T_{2}=0.04, T_{3}=0.28
$$

$$
\lambda=\frac{\mathrm{n}}{\sum_{\mathrm{m}=1}^{\mathrm{M}} \mathrm{~T}_{\mathrm{m}} \mathrm{~V}_{\mathrm{m}}}=\frac{1}{0.028+0.04 * 0.7+0.28 * 0.2}=\frac{1}{0.112}=8.93 / \mathrm{sec}
$$

$$
\lambda_{m}=\lambda V_{m} ; \lambda_{1}=8.93, \lambda_{2}=6.25, \lambda_{3}=1.786
$$

$$
\mathbf{n}_{\mathrm{m}}(\mathbf{n})=\lambda_{\mathrm{m}} \mathbf{T}_{\mathrm{m}}
$$

$$
n_{1}(1)=0.25, n_{2}(1)=0.25, n_{3}(1)=0.5
$$

| $\mathrm{n} \backslash \mathrm{m}$ | 0 | 1 | 2 | 3 |
| :--- | :--- | :--- | :--- | :--- |
| 0 | 0 | 1 | 1 | $1=\mathrm{G}(0)$ |
| 1 | 0 | 1 | 2 | 4 |
| 2 | 0 | 1 | 3 | 11 |
| 3 | 0 | 1 | 4 | 26 |
| 4 | 0 | 1 | 5 | 57 |
| 5 | 0 | 1 | 6 | 120 |
| 6 | 0 | 1 | 7 | 247 |
| 7 | 0 | 1 | 8 | $502=\mathbf{G}(7)$ |

(From convolution, $\pi_{100}=(1 / 4) 1^{1} 1^{0} 2^{0}=0.25 ; \pi_{010}=(1 / 4) 1^{0} 1^{1} 2^{0}=0.25 ; \pi_{001}=(1 / 4)$ $\qquad$ $=0.5$ )

## MVA procedure

$$
\begin{aligned}
& \mathrm{n}=2 ; \\
& \mathrm{T}_{1}=0.28\{1+0.25\}=0.035 \mathrm{sec} \\
& \mathrm{~T}_{2}=0.05, \mathrm{~T}_{3}=0.42 \\
& \lambda=\frac{\mathrm{n}}{\sum_{\mathrm{m}=1}^{\mathrm{M}} \mathrm{~T}_{\mathrm{m}} \mathrm{~V}_{\mathrm{m}}}=\frac{2}{0.035+0.035+0.084}=\frac{2}{0.154}=12.99 / \mathrm{sec} \\
& \lambda_{m}=\lambda V_{m} ; \lambda_{1}=12.99, \lambda_{2}=9.09, \lambda_{3}=2.60 \\
& n_{1}(2)=0.455, n_{2}(2)=0.455, n_{3}(2)=1.092
\end{aligned}
$$

| $n \backslash m$ | 0 | 1 | 2 | 3 |
| :--- | :--- | :--- | :--- | :--- |
| 0 | 0 | 1 | 1 | $1=\mathbf{G}(0)$ |
| 1 | 0 | 1 | 2 | 4 |
| 2 | 0 | 1 | 3 | 11 |
| 3 | 0 | 1 | 4 | 26 |
| 4 | 0 | 1 | 5 | 57 |
| 5 | 0 | 1 | 6 | 120 |
| 6 | 0 | 1 | 7 | 247 |
| 7 | 0 | 1 | 8 | $502=\mathbf{G}(7)$ |

From convolution, $E\left[n_{1}\right]=\mathbf{u}_{1}{ }^{1}(\mathbf{G}(1) / G(2))+$ $\qquad$ $=4 / 11+1 / 11=0.455$

## MVA procedure

$$
\begin{aligned}
& \mathrm{n}=3 ; \\
& \mathrm{T}_{1}=0.028\{1+0.455\}=0.041 \mathrm{sec} \\
& \mathrm{~T}_{2}=0.058, \mathrm{~T}_{3}=0.586 \\
& \lambda=\frac{\mathrm{n}}{\sum_{\mathrm{m}=1}^{\mathrm{M}} \mathrm{~T}_{\mathrm{m}} \mathrm{~V}_{\mathrm{m}}}=\frac{3}{0.041+0.041+0.117}=\frac{3}{0.199}=15.1 / \mathrm{sec} \\
& \lambda_{\mathrm{m}}=\lambda \mathrm{V}_{\mathrm{m}} ; \\
& \lambda_{1}=15.11, \lambda_{2}=10.6, \lambda_{3}=3.02 \\
& \mathbf{n}_{1}(3)=0.615, n_{2}(3)=0.615, n_{3}(3)=1.77
\end{aligned}
$$

## Verification with convolution method

$$
\begin{aligned}
& \rho_{1}=\frac{\lambda_{1}}{\mu_{1}}=\frac{15.11}{\frac{1}{0.028}}=0.423 \\
& \rho_{1}=u_{1} \frac{G(2)}{G(3)}=1 * \frac{11}{26}=0.423
\end{aligned}
$$

