

# Lecture 4

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**EE** 336

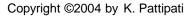
Stochastic Models for the Analysis of Computer Systems
And Communication Networks





#### **Outline**

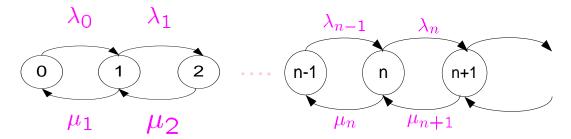
- ☐ Review of Lecture 3
- $\square$  M/M/1/N queue
- $\square$  M/M/m/m queue
- ☐ Engset Model
- $\square$  M/M/1 with feedback
- ☐ Machine Repairman Model
- ☐ Some Application Examples





### **Review of Lecture 3 - 1**

#### ☐ Birth-death Processes:



$$\lambda_{n-1}p_{n-1} + \mu_{n+1}p_{n+1} = (\lambda_n + \mu_n)p_n$$
 global balance Eqns.

$$\lambda_{n-1}p_{n-1}=\mu_np_n$$
 local (detailed) balance Eqns.

$$p_n = \frac{\lambda_{n-1}}{\mu_n} p_{n-1}$$

get 
$$p_0$$
 from

$$(1 + \sum_{n=1}^{\infty} \prod_{i=0}^{n-1} \frac{\lambda_i}{\mu_{i+1}})^{-1} = p_0$$





### **Review of Lecture 3 - 2**

#### ☐ M/M/1 queue:

$$\lambda_n = \lambda$$
  $n \geq 0$   $\mu_n = \mu$   $n \geq 1$  M/M/1 queue with feedback is similar !!

☐ M/M/1 queue (Infinite server queue):

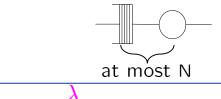
$$\lambda_n = \lambda \quad n \ge 0$$

$$\mu_n = n\mu \quad n \ge 1$$

☐ M/M/m queue:

$$\lambda_n = \lambda$$
$$\mu_n = \min(n, m)\mu$$

☐ M/M/1/N queue: Finite buffer case



$$\lambda_n = \lambda \quad 0 \le n \le N - 1$$
  
 $\mu_n = \mu \quad 1 \le n \le N$ 

$$p_n = \frac{\lambda}{\mu} p_{n-1} = \rho p_{n-1} = \rho^n p_0$$





#### □ M/M/1/N queue

$$\sum_{n=0}^{N} p_n = 1 \Rightarrow (1 + \rho + \rho^2 + \dots + \rho^N) p_0 = 1$$

$$\Rightarrow p_0 = \frac{1 - \rho}{1 - \rho^{N+1}} \text{ (note: as N} \rightarrow \infty, p_0 \rightarrow 1 - \rho \dots M/M/1 queue)$$

$$\therefore p_n = \frac{(1-\rho)\rho^n}{1-\rho^{N+1}} \text{ for } 0 \le n \le N$$

$$p_N=$$
 prob. that the system is full = prob. blocking =  $p_B=\frac{(1-\rho)\rho^N}{1-\rho^{N+1}}$  Truncated modified geometric pmf

$$G(z) = \sum_{n=0}^{N} \frac{\rho^n z^n}{1 - \rho^{N+1}} (1 - \rho) = \frac{(1 - \rho)[1 - (\rho z)^{N+1}]}{(1 - \rho^{N+1})(1 - \rho z)}$$





#### Average queue length

$$Q = \frac{dG(z)}{dz} \Big|_{z=1} = \frac{(1-\rho)\left[-(N+1)\rho^{N+1}z^{N}(1-\rho z) + \rho(1-(\rho z)^{N+1})\right]}{(1-\rho^{N+1})(1-\rho z)^{2}} \Big|_{z=1}$$

$$= \frac{\rho}{(1-\rho)} - \frac{(N+1)\rho^{N+1}}{(1-\rho^{N+1})} = \frac{\rho}{(1-\rho)} [1-(N+1)\frac{\rho^{N}(1-\rho)}{(1-\rho^{N+1})}]$$

$$= \frac{\rho}{(1-\rho)} [1-(N+1)P_{B}] = Q_{M/M/1} [1-(N+1)P_{B}] \rightarrow Q_{M/M/1} \text{ as } N \rightarrow \infty$$

#### Throughput

$$X = \sum_{n=1}^{N} \mu_n p_n = \mu \sum_{n=1}^{N} p_n = \mu (1 - p_0)$$

$$= \mu \left[1 - \frac{1 - \rho}{1 - \rho^{N+1}}\right] = \rho \mu \left(\frac{1 - \rho^{N}}{1 - \rho^{N+1}}\right) = \lambda \left[1 - P_B\right]$$

$$\lambda = \lambda \left[1 - \frac{\lambda}{1 - \rho^{N+1}}\right] = \lambda \left[1 - \rho^{N}\right]$$





So, in M/M/1/N queue 
$$ho(1-P_B)=1-p_0$$

$$P_B = P_N = \frac{(1 - \rho)\rho^N}{1 - \rho^{N+1}} \approx (1 - \rho)\rho^N \text{ for } N \gg 1 \& \rho < 1$$

$$\Rightarrow \rho^N = \frac{P_B}{1 - \rho} \text{ or } N = \frac{\ln(\frac{P_B}{1 - \rho})}{\ln \rho}$$

Can use this relation to design buffer size.

For example

$$\rho = 0.5, \ P_B = 10^{-3} \Rightarrow \ N = \frac{-2.7}{-0.3} \approx 9$$

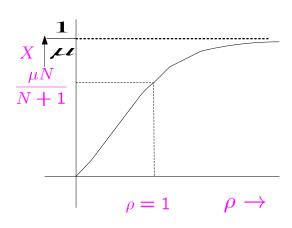
$$\rho = 0.5, \ P_B = 10^{-6} \Rightarrow \ N = \frac{-5.7}{-0.3} \approx 19$$

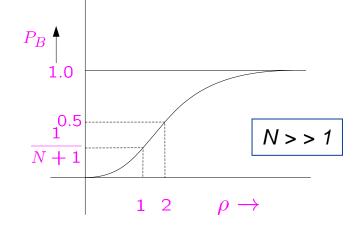
• Note that  $\rho$  can be >1. If  $\rho$  >1, change  $\left| p_0 = \frac{\rho - 1}{\rho^{N+1} - 1} \right|$  and don't

actually have a control over 
$$P_B$$
 since  $P_B = p_N = \frac{(\rho - 1)\rho^N}{\rho^{N+1} - 1} \approx \frac{\rho - 1}{\rho} as \rho \to \infty$ 









Average response time:

$$R = \frac{Q}{X} = \frac{(\frac{\rho}{1-\rho})(1 - P_B) - \frac{N\rho}{1-\rho}P_B}{\lambda(1 - P_B)}$$

$$\therefore R = \frac{1}{\mu} \frac{1}{1 - \rho} - \frac{N}{\mu (1 - \rho)} \frac{P_B}{1 - P_B}$$

(or) 
$$R_{M/M/1/N} = R_{M/M/1} [1 - \frac{NP_B}{1 - P_B}] = \frac{R_{M/M/1}}{1 - P_B} [1 - (N+1)P_B]$$





When 
$$\rho >> 1$$
,  $Q \to N - \frac{1}{\rho - 1} = N - \frac{1}{\rho P_B}$  
$$R \to \frac{N}{\mu} - \frac{1}{\mu(\rho - 1)}$$

Utilization:

$$U = \frac{X}{\mu} = \rho(1 - P_B)$$

Average waiting time

$$W = R - \frac{1}{\mu} = \frac{1}{\mu} \frac{\rho}{1 - \rho} - \frac{NP_B}{\mu(1 - \rho)(1 - P_B)}$$

Average waiting queue length

$$Q_W = WX = \frac{\rho^2}{1 - \rho} (1 - P_B) - \frac{NP_B \cdot \rho}{(1 - \rho)}$$





M/M/m/m: the m-server loss system: valid for *M/G/m/m* system also

$$\mu_n = n\mu, \ 1 \le n \le m$$
 
$$p_n = \frac{\lambda}{n\mu} p_{n-1} \Rightarrow p_n = (\frac{\lambda}{\mu})^n \frac{1}{n!} p_0$$
 So, 
$$p_0 = [\sum_{n=0}^{m} (\frac{\lambda}{\mu})^n \frac{1}{n!}]^{-1}$$

 $\lambda_n = \lambda$ , 0 < n < m-1

 $p_m = prob. \{all \ servers \ busy\}$ 

$$=\frac{(\frac{\lambda}{\mu})^m \frac{1}{m!}}{\sum_{n=0}^m (\frac{\lambda}{\mu})^n \frac{1}{n!}} = \frac{\frac{\rho^m}{m!}}{\sum_{n=0}^m \rho^n \frac{1}{n!}} = P_B$$
 similar to truncated Poisson

This is known as Erlang's B-formula

• For  $5 < \rho < 50$ , useful approximations for m for a specified  $P_B$  are:

$$P_B=.01 \ \Rightarrow \ m\approx 5.5+1.17 \rho$$
 
$$P_B=.001 \ \Rightarrow \ m\approx 7.8+1.28 \rho$$
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Average queue length

$$Q = \sum_{n=1}^{m} n p_n = \frac{\sum_{n=1}^{m} \frac{\rho^n}{(n-1)!}}{\sum_{n=0}^{m} \frac{\rho^n}{n!}} = \rho[1 - P_m] = \rho[1 - P_B]$$

= Average number of busy servers

Note that as  $\rho \uparrow P_B \uparrow \& Q \rightarrow m$  as it should!

• Throughput:  $X = \lambda(1 - P_B) = \lambda(1 - P_m)$ 

Alternatively, 
$$X = \sum_{n=1}^{m} \mu_n p_n = \frac{\mu \sum_{n=1}^{m} \frac{\rho^n}{(n-1)!}}{\sum_{n=0}^{m} \frac{\rho^n}{n!}} = \mu Q = \lambda (1 - P_B)$$

- Average response time =  $R = \frac{1}{\mu}$
- Average waiting time = W = 0
- Average waiting queue length =  $Q_W = 0$
- Utilization  $U = \frac{\lambda(1 P_B)}{m\mu} = \frac{\rho(1 P_B)}{m} = \frac{Q}{m}$





### **Engset Model Analysis - 1**

#### **Engset Model**

$$\lambda_{n} = (M - n)\lambda \quad 0 \le n \le N$$

$$\mu_{n} = n\mu \quad N \le M$$

$$\Rightarrow p_{n} = \left(\frac{n - 1}{n}\right)\frac{\lambda}{\mu}p_{n - 1} = \left(\frac{n - 1}{n}\right)\rho \cdot p_{n - 1}$$

$$\text{So,} \quad p_{n} = \binom{M}{n}\rho^{n}p_{0} \quad \text{where} \quad \binom{M}{n} = \frac{M!}{n!(M - n)!}$$

$$p_{0} = \left[\sum_{n = 0}^{N} \binom{M}{n}\rho^{n}\right]^{-1}$$

$$p_n = \frac{\binom{M}{n} \rho^n}{\left[\sum_{m=0}^{N} \binom{M}{m} \rho^m\right]} \quad ; n = 0, 1, 2, ..., N$$
 called Engset distribution





## **Engset Model Analysis - 2**

For M>N, the blocking probability

$$P_B = P_N = \frac{\binom{M}{N} \rho^N}{\left[\sum_{n=0}^{N} \binom{M}{n} \rho^n\right]}$$

Average queue length

$$Q = \sum_{n=1}^{N} np_n \qquad p_0 = p_0(M) \text{ to emphasize its dependence on } M \text{ for a given } N$$

$$= p_0(M) \sum_{n=1}^{N} \frac{M!}{(n-1)!(M-n)!} \rho^n; \quad p_0(M) = \left[\sum_{n=0}^{N} \binom{M}{n} \rho^n\right]^{-1}$$

$$= M\rho \cdot p_0(M) \sum_{n=0}^{N-1} \frac{M-1!}{n!(M-1-n)!} \rho^n$$

$$= \frac{M\rho \cdot p_0(M)}{p_0(M-1)} [1 - P_B(M-1)] = \text{Average # of busy lines}$$

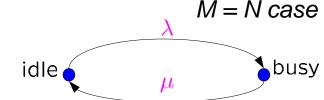




## **Engset Model Analysis - 3**

• Throughput 
$$X = \sum_{n=1}^{N} n\mu p_n = \mu Q = \frac{M\lambda \cdot p_0(M)}{p_0(M-1)} [1 - P_B(M-1)]$$

- Response time R =
- Utilization  $U = \frac{X}{N\mu} = \frac{Q}{N}$



Note: 1) As M!1, &  $\lambda!03M\lambda!\lambda'$ , the measures reduce to those  $Q = \frac{\lambda}{1 - P_B} \quad X = \lambda' (1 - P_B)$ for *M/M/m/m* system:

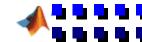
2) 
$$M=N$$

$$\Rightarrow p_n = \frac{\binom{N}{n} \rho^n}{\left[\sum_{m=0}^{N} \binom{N}{m} \rho^m\right]} = \binom{N}{n} a^n (1-a)^{N-n}; \ a = \frac{\rho}{1+\rho} = \frac{\lambda}{\lambda+\mu}$$

$$\Rightarrow \text{ each server is either busy or idle \& independent of others}$$

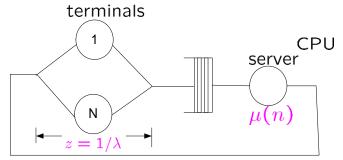
- ⇒ each server is either busy or idle & independent of others
- ⇒ performance measures

$$Q = Na = \frac{N\rho}{1+\rho}; \ X = \frac{N\lambda}{1+\rho} = \mu Q; \ R = \frac{1}{\mu}; \ U = \frac{Q}{N} = a = \frac{\rho}{1+\rho}$$





Machine Repairman Model (simplest form of closed network)



When *n* user requests are at the CPU, (*N-n*) are *potential* user requests, so that

$$\lambda_{n} = \begin{cases} (N-n)\lambda & 0 \leq n \leq N-1 \\ 0 & for \ n \geq N \end{cases}$$
showed earlier that:
$$\frac{N}{z + \frac{N}{\mu}} \leq X(N) \leq min \left\{ \mu, \frac{N}{z + 1/\mu} \right\}$$

$$\mu(n) = \mu \left\{ \text{constant} \right\}$$

$$max[N/\mu, \ z + 1/\mu] \leq R(N) \leq z + N/\mu$$

$$\frac{N}{z + \frac{N}{\mu}} \le X(N) \le \min\left\{\mu, \frac{N}{z + 1/\mu}\right\}$$

In general: 
$$p_n = \frac{(N-n+1)\lambda}{\mu(n)} p_{n-1}; \quad n = 1, 2, ..., N$$

For later notation, we write it as:

$$p(n/N) = \frac{(N-n+1)\lambda}{\mu(n)} p(n-1/N); \quad n=1,2,...,N$$
 
$$= \frac{N!}{(N-n)!} \cdot \frac{\lambda^n}{\prod_{i=1}^n \mu(i)} p(o/N)$$
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$$p(o/N) = \left[\sum_{n=0}^{N} \frac{N!}{(N-n)!} \cdot \frac{\lambda^n}{\prod_{i=1}^n \mu(i)}\right]^{-1}$$
Single server:  $\mu(n) = \mu \Rightarrow p(o/N) = \left[\sum_{n=0}^{N} \frac{N!}{(N-n)!} \cdot \rho^n\right]^{-1} = [G(N)]^{-1}$ 

$$p(n/N) = \frac{N!}{(N-n)!} \cdot \rho^n \cdot p(o/N)$$

Infinite server: 
$$\mu(n) = n\mu \Rightarrow p(o/N) = \left[\sum_{n=0}^{N} \frac{N!}{(N-n)!n!} \rho^n\right]^{-1} = [1+\rho]^{-N}$$

Binomial Distribution 
$$p(n/N) = \binom{N}{n} a^n (1-a)^{N-n}; \quad a = \frac{\rho}{1+\rho}$$
 Engset Model  $M = N$  case

Let us consider the single server case in detail.

• Recursive expression for G(N)

$$G(N) = \sum_{n=0}^{N} \frac{N!}{(N-n)!} \rho^n = 1 + N\rho + N(N-1)\rho^2 + \dots + N!\rho^N$$
$$= 1 + N\rho[1 + (N-1)\rho + \dots + (N-1)!\rho^{N-1}]$$
$$= 1 + N\rho G(N-1): G(0) = 1$$

Later, we will see that this kind of recursion extends to networks also and forms the basis of the so called "Convolution Algorithm" Copyright ©2004 by K. Pattipati





• Throughput 
$$X(N) = \sum_{n=1}^{N} \mu(n)p(n/N)$$
  
=  $\mu(1 - p(o/N))$   
=  $\mu U(N)$ 

• Utilization,  $U(N) = \frac{\text{Throughput}}{\text{Max service rate}} = \frac{X(N)}{\mu}$ 

Also 
$$X(N) = \mu \left[ 1 - \frac{1}{G(N)} \right] = \mu \cdot \frac{G(N) - 1}{G(N)} = N\lambda \cdot \frac{G(N - 1)}{G(N)} = \lambda \frac{\tilde{G}(N - 1)}{\tilde{G}(N)}$$

$$U(N) = \frac{G(N) - 1}{G(N)} = \rho \frac{\tilde{G}(N - 1)}{\tilde{G}(N)}$$

$$\tilde{G}(N) = \frac{G(N)}{N!}$$

Mean value analysis (MVA) recursion:

Know 
$$p(n/N) = \frac{N!}{(N-n)!} \rho^n \cdot \frac{1}{G(N)} = \frac{\rho^n}{(N-n)!} \cdot \frac{1}{\tilde{G}(N)}$$
$$p(n-1/N-1) = \frac{(N-1)!}{(N-n)!} \rho^{n-1} \cdot \frac{1}{G(N-1)} = \frac{\rho^{n-1}}{(N-n)!} \cdot \frac{1}{\tilde{G}(N-1)}$$
$$\frac{p(n/N)}{p(n-1/N-1)} = \frac{N\rho \cdot G(N-1)}{G(N)} = \frac{G(N)-1}{G(N)} = \rho \frac{\tilde{G}(N-1)}{\tilde{G}(N)} = U(N)$$





$$p(n/N) = U(N) \cdot p(n-1|N-1)$$
$$= \frac{X(N)}{\mu} \cdot p(n-1|N-1)$$

• In words, the probability distribution with *N* customers is related to the probability distribution of the <u>same</u> system with (*N-1*) customers. This observation is valid in a more general context involving multiple nodes and forms the basis of Reiser and Lavenberg's Mean Value Analysis (*MVA*) Algorithm.

#### Pursuing the recursion further,

Average Queue length

$$Q(N) = \sum_{n=1}^{N} np(n/N)$$

$$= \frac{X(N)}{\mu} \sum_{n=1}^{N} np(n-1/N-1)$$

$$= \frac{X(N)}{\mu} [\sum_{n=1}^{N} (n-1)p(n-1/N-1) + \sum_{n=1}^{N} p(n-1/N-1)]$$

$$= \frac{X(N)}{\mu} [1 + Q(N-1)]$$





Or

$$R_c(N) = \frac{1}{\mu} [1 + Q(N-1)]$$

valid for networks also

- In words, the mean response times of an arriving customer is equal to his own service time  $1/\mu$  plus the waiting time due to customers in a system with (*N*-1) customers  $\Rightarrow$  an arriving customer in a system with *N* terminals "sees" Q(N-1) in equilibrium. We will see presently that in an M/M/1 system (or so called open systems) an arriving and departing customers "see" the same # of customers (so called Burke's theorem). Not so in a closed system. In a closed system, an arrival sees a network with one less <u>customer</u> (himself removed)... so called "arrival theorem of closed systems".
  - Implementation (no closed form expression ⇒ need to evaluate via a <u>recursive</u> algorithm)





$$Q \leftarrow 0$$

$$DO n = 1, 2, ..., N$$

$$R_c \leftarrow \frac{1}{\mu}(1+Q)$$

$$X \leftarrow \frac{n}{R_c + \frac{1}{\lambda}}$$

$$Q \leftarrow XR_c$$

$$END DO$$

$$U(N) = \frac{X}{\mu}$$

$$Q(N)$$

$$X(N)$$

#### Questions:

- 1. How to compute response time distributions? ... we will do it only for M/M/1
- 2. What does it all mean? ... B-D processes are <u>"time-reversible Markov</u> chains". We will pursue this in next class.





## Response-time Density of M/M/1 Queue -1

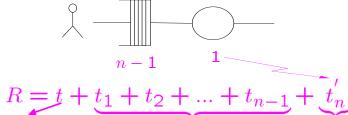
- ☐ In discussing *M/M/1* queue, we made no assumption regarding queuing discipline
- ⇒ All queue disciplines yield <u>identical</u> mean response times and mean queue lengths However, the distribution of response times does depend on the queue discipline.

  <u>We will compute the distributions for FCFS discipline.</u>

Let us denote

 $F_R(r/n) \triangleq P\{\text{the response time of tagged customer} \leq r|$   $n \text{ customers are found in the system upon arrival}\}$ 

tagged customer



service time service time of tagged customer customers in queue

service time of remaining (or residual)
ner customers service time of customer

in service

• Since service time is exponentially distributed  $t_n^{'}$  has the <u>same</u> density (recall memory-less property)





## Response-time Density of M/M/1 Queue -2

 $\Rightarrow t, t_1, t_2, ..., t_{n-1}, t'_n$  are i.i.d. random variables

$$L_R(s/n) = \left(\frac{\mu}{s+\mu}\right)^{n+1}$$

$$L_R(s) = \sum_{n=0}^{\infty} L_R(s/n)p_n$$

$$= \sum_{n=0}^{\infty} \left(\frac{\mu}{s+\mu}\right)^{n+1} (1-\rho)\rho^n$$

$$= \left(\frac{\mu}{s+\mu}\right) (1-\rho) \frac{1}{1-\frac{\rho\mu}{s+\mu}} = \frac{\mu(1-\rho)}{[s+\mu(1-\rho)]}$$

 $\Rightarrow$  Response time is exponentially distributed with mean  $\frac{1}{\mu(1-\rho)}$ 

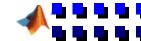
$$f_R(t) = \mu(1-\rho)e^{-\mu(1-\rho)t}$$

#### **Examples**:

1). Suppose have M/M/1 queue and increase  $\lambda$  and  $\mu$  by a factor k>1

$$\Rightarrow \rho = \frac{k\lambda}{k\mu} \text{ same}$$

$$\Rightarrow Q = \frac{\rho}{1-\rho} \text{ same}$$

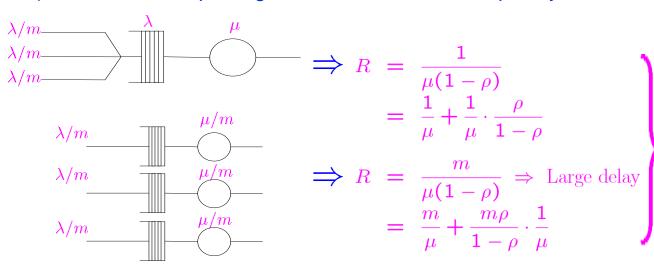




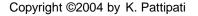
But 
$$R = \frac{1}{k\mu(1-\rho)} \Rightarrow$$
 reduced by a factor of k.

An arriving customer sees the same # of customers in the steady state, but customers will be moving *k* times faster.

2). Statistical multiplexing vs. Time division or Frequency division multiplexing



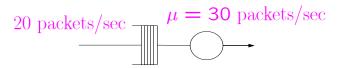
True only for Poisson streams If arrivals are regular, time & frequency division multiplexing can be good





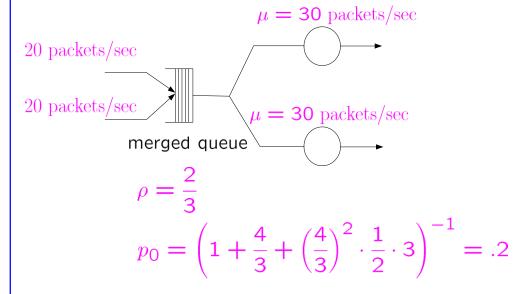


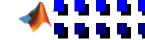
3). Single queue versus separate queues (OR) why banks have single queues?



20 packets/sec 
$$\mu = 30$$
 packets/sec

seperate 
$$\Rightarrow R_{S_1} = R_{S_2} = \frac{1}{30(1 - \frac{2}{3})} = .1 \text{ sec}$$



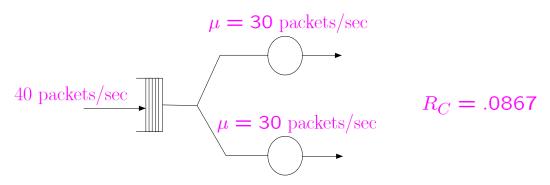




$$P_Q = .2 \cdot \frac{16}{9} \cdot \frac{1}{2} \cdot 3 = \frac{1.6}{3} = .533$$

$$R_C = \frac{1}{30} + \frac{.533}{30 \cdot \frac{1}{3}} = \frac{1}{30} (1 + 1.6) = .0867 < R_S \quad \text{(merged queue is better)}$$

4). Should I buy several small severs or one large one?



40 packets/sec 
$$\mu = 60$$
 packets/sec  $R_{LS} = \frac{1}{60 - 40} = .05 < R_c$ 

What happens to waiting time?





$$W_C = \frac{1}{60 - 40} \times .53 = .0265$$
 $W_{LS} = \frac{2/3}{60/3} = .0334 \implies W_{LS} > W_C \text{ but } R_{LS} < R_C$ 

Buy the largest !!

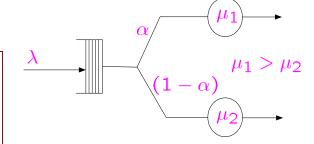
But, from reliability viewpoint, several small ones may be better!!

#### 5) Optimal Routing

$$R_{1} = \frac{1}{\mu_{1} - \lambda \alpha}; R_{2} = \frac{1}{\mu_{2} - \lambda(1 - \alpha)}$$

$$R = \alpha R_{1} + (1 - \alpha)R_{2} = \frac{\alpha}{\mu_{1} - \lambda \alpha} + \frac{(1 - \alpha)}{\mu_{2} - \lambda(1 - \alpha)}$$

$$Optimum \Rightarrow \frac{dR}{d\alpha} = 0 \Rightarrow \alpha^{*} = \frac{\sqrt{\frac{\mu_{1}}{\mu_{2}} + \frac{\lambda}{\mu_{2}} - 1}}{\frac{\lambda}{\sqrt{\mu_{1}\mu_{2}}} + \frac{\lambda}{\mu_{2}}}$$



$$\lambda = 4; \ \mu_1 = 4; \ \mu_2 = 1$$

$$\alpha^* = \frac{\sqrt{\frac{\mu_1}{\mu_2} + \frac{\lambda}{\mu_2} - 1}}{\frac{\lambda}{\sqrt{\mu_1 \mu_2}} + \frac{\lambda}{\mu_2}} = \frac{5}{6}$$





## **Summary**

- $\square$  M/M/1/N queue
- □ M/M/m/m queue
- ☐ Engset Model
- $\square$  M/M/1 with feedback
- ☐ Machine Repairman Model
- ☐ Some Application Examples

