

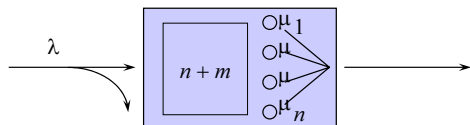
## 8. Queueing systems

### Contents

- Refresher: Simple teletraffic model
- Queueing discipline
- M/M/1 (1 server,  $\infty$  waiting places)
- Application to packet level modelling of data traffic
- M/M/n ( $n$  servers,  $\infty$  waiting places)

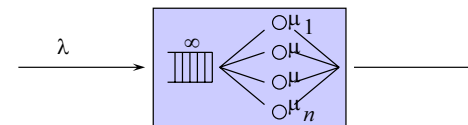
### Simple teletraffic model

- **Customers arrive** at rate  $\lambda$  (customers per time unit)
  - $1/\lambda$  = average inter-arrival time
- Customers are **served** by  $n$  parallel **servers**
- When busy, a server serves at rate  $\mu$  (customers per time unit)
  - $1/\mu$  = average service time of a customer
- There are  $n + m$  **customer places** in the system
  - at least  $n$  **service places** and at most  $m$  **waiting places**
- It is assumed that blocked customers (arriving in a full system) are lost



### Pure queueing system

- Finite number of servers ( $n < \infty$ ),  $n$  service places, infinite number of waiting places ( $m = \infty$ )
  - If all  $n$  servers are occupied when a customer arrives, it occupies one of the waiting places
  - No customers are lost but some of them have to wait before getting served
- From the customer's point of view, it is interesting to know e.g.
  - what is the probability that it has to wait "too long"?



## Contents

- Refresher: Simple teletraffic model
- Queueing discipline
- M/M/1 (1 server,  $\infty$  waiting places)
- Application to packet level modelling of data traffic
- M/M/n ( $n$  servers,  $\infty$  waiting places)

5

## Queueing discipline

- Consider a single server ( $n = 1$ ) queueing system
- **Queueing discipline** determines the way the server serves the customers
  - It tells
    - whether the customers are served one-by-one or simultaneously
  - Furthermore, if the customers are served one-by-one, it tells
    - in which order they are taken into the service
  - And if the customers are served simultaneously, it tells
    - how the service capacity is shared among them
- **Note:** In computer systems the corresponding concept is **scheduling**
- A queueing discipline is called **work-conserving** if customers are served with full service rate  $\mu$  whenever the system is non-empty

6

## Work-conserving queueing disciplines

- First In First Out (**FIFO**) = First Come First Served (FCFS)
  - ordinary queueing discipline (“queue”)
    - arrival order = service order
  - customers served one-by-one (with full service rate  $\mu$ )
  - always serve the customer that has been waiting for the longest time
  - default queueing discipline in this lecture
- Last In First Out (**LIFO**) = Last Come First Served (LCFS)
  - reversed queueing discipline (“stack”)
  - customers served one-by-one (with full service rate  $\mu$ )
  - always serve the customer that has been waiting for the shortest time
- Processor Sharing (**PS**)
  - “fair queueing”
  - customers served simultaneously
  - when  $i$  customers in the system, each of them served with equal rate  $\mu/i$
  - see Lecture 9. Sharing systems

7

## Contents

- Refresher: Simple teletraffic model
- Queueing discipline
- M/M/1 (1 server,  $\infty$  waiting places)
- Application to packet level modelling of data traffic
- M/M/n ( $n$  servers,  $\infty$  waiting places)

8

## M/M/1 queue

- Consider the following simple teletraffic model:
  - Infinite number of independent customers ( $k = \infty$ )
  - Interarrival times are IID and exponentially distributed with mean  $1/\lambda$ 
    - so, customers arrive according to a Poisson process with intensity  $\lambda$
  - One server ( $n = 1$ )
  - Service times are IID and exponentially distributed with mean  $1/\mu$
  - Infinite number of waiting places ( $m = \infty$ )
  - Default queueing discipline: **FIFO**
- Using Kendall's notation, this is an **M/M/1 queue**
  - more precisely: M/M/1-FIFO queue
- Notation:
  - $\rho = \lambda/\mu =$  traffic load

9

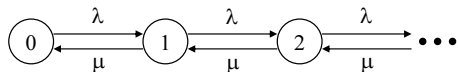
## Related random variables

- $X$  = number of customers in the system at an arbitrary time  
= queue length in equilibrium
- $X^*$  = number of customers in the system at an (typical) arrival time  
= queue length seen by an arriving customer
- $W$  = waiting time of a (typical) customer
- $S$  = service time of a (typical) customer
- $D = W + S$  = total time in the system of a (typical) customer = delay

10

## State transition diagram

- Let  $X(t)$  denote the number of customers in the system at time  $t$ 
  - Assume that  $X(t) = i$  at some time  $t$ , and consider what happens during a short time interval  $(t, t+h]$ :
    - with prob.  $\lambda h + o(h)$ , a new customer arrives (state transition  $i \rightarrow i+1$ )
    - if  $i > 0$ , then, with prob.  $\mu h + o(h)$ , a customer leaves the system (state transition  $i \rightarrow i-1$ )
- Process  $X(t)$  is clearly a Markov process with state transition diagram



- Note that process  $X(t)$  is an irreducible birth-death process with an infinite state space  $S = \{0, 1, 2, \dots\}$

11

## Equilibrium distribution (1)

- Local balance equations (LBE):

$$\pi_i \lambda = \pi_{i+1} \mu \quad (\text{LBE})$$

$$\Rightarrow \pi_{i+1} = \frac{\lambda}{\mu} \pi_i = \rho \pi_i$$

$$\Rightarrow \pi_i = \rho^i \pi_0, \quad i = 0, 1, 2, \dots$$

- Normalizing condition (N):

$$\sum_{i=0}^{\infty} \pi_i = \pi_0 \sum_{i=0}^{\infty} \rho^i = 1 \quad (\text{N})$$

$$\Rightarrow \pi_0 = \left( \sum_{i=0}^{\infty} \rho^i \right)^{-1} = \left( \frac{1}{1-\rho} \right)^{-1} = 1 - \rho, \quad \text{if } \rho < 1$$

12

### Equilibrium distribution (2)

- Thus, for a **stable** system ( $\rho < 1$ ), the equilibrium distribution exists and is a **geometric distribution**:

$$\rho < 1 \Rightarrow X \sim \text{Geom}(\rho)$$

$$P\{X = i\} = \pi_i = (1 - \rho)\rho^i, \quad i = 0, 1, 2, \dots$$

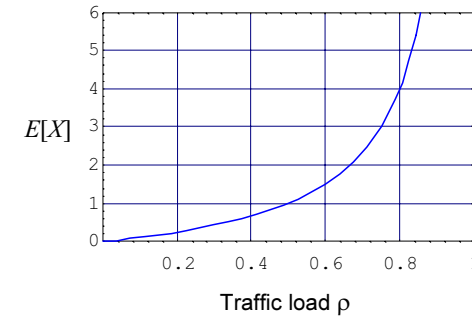
$$E[X] = \frac{\rho}{1 - \rho}, \quad D^2[X] = \frac{\rho}{(1 - \rho)^2}$$

- Remark:**

- This result is valid for any **work-conserving** queueing discipline (FIFO, LIFO, PS, ...)
- This result is **not insensitive** to the service time distribution for **FIFO**
  - even the mean queue length  $E[X]$  depends on the distribution
- However, for any **symmetric** queueing discipline (such as LIFO or PS) the result is, indeed, **insensitive** to the service time distribution

13

### Mean queue length $E[X]$ vs. traffic load $\rho$



14

### Mean delay

- Let  $D$  denote the total time (delay) in the system of a (typical) customer
  - including both the waiting time  $W$  and the service time  $S$ :  $D = W + S$
- Little's formula:  $E[X] = \lambda \cdot E[D]$ . Thus,

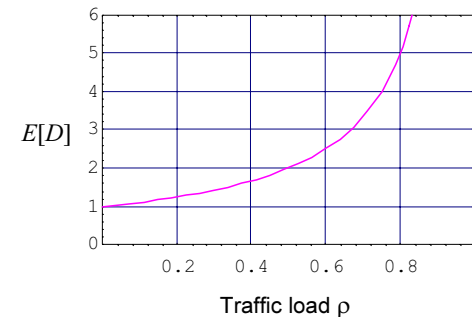
$$E[D] = \frac{E[X]}{\lambda} = \frac{1}{\lambda} \cdot \frac{\rho}{1 - \rho} = \frac{1}{\mu} \cdot \frac{1}{1 - \rho} = \frac{1}{\mu - \lambda}$$

- Remark:**

- The mean delay is the same for all work-conserving queueing disciplines (FIFO, LIFO, PS, ...)
- But the variance and other moments are different!

15

### Mean delay $E[D]$ vs. traffic load $\rho$



16

### Mean waiting time

- Let  $W$  denote the waiting time of a (typical) customer
- Since  $W = D - S$ , we have

$$E[W] = E[D] - E[S] = \frac{1}{\mu} \cdot \frac{1}{1-\rho} - \frac{1}{\mu} = \frac{1}{\mu} \cdot \frac{\rho}{1-\rho}$$

17

### Waiting time distribution (1)

- Let  $W$  denote the waiting time of a (typical) customer
- Let  $X^*$  denote the number of customers in the system at the arrival time
- PASTA:  $P\{X^* = i\} = P\{X = i\} = \pi_i$ .
- Assume now, for a while, that  $X^* = i$ 
  - Service times  $S_2, \dots, S_i$  of the waiting customers are IID and  $\sim \text{Exp}(\mu)$
  - Due to the memoryless property of the exponential distribution, the **remaining** service time  $S_1^*$  of the customer in service also follows  $\text{Exp}(\mu)$ -distribution (and is independent of everything else)
  - Due to the FIFO queueing discipline,  $W = S_1^* + S_2 + \dots + S_i$
  - Construct a Poisson (point) process  $\tau_n$  by defining  $\tau_1 = S_1^*$  and  $\tau_n = S_1^* + S_2 + \dots + S_n$ ,  $n \geq 2$ . Now (since  $X^* = i$ ):  $W > t \Leftrightarrow \tau_i > t$



18

### Waiting time distribution (2)

- Since  $W = 0 \Leftrightarrow X^* = 0$ , we have

$$P\{W = 0\} = P\{X^* = 0\} = \pi_0 = 1 - \rho$$

$$\begin{aligned} P\{W > t\} &= \sum_{i=1}^{\infty} P\{W > t \mid X^* = i\} P\{X^* = i\} \\ &= \sum_{i=1}^{\infty} P\{\tau_i > t\} \pi_i = \sum_{i=1}^{\infty} P\{\tau_i > t\} (1 - \rho) \rho^i \end{aligned}$$

- Denote by  $A(t)$  the Poisson (counter) process corresponding to  $\tau_n$ 
  - It follows that:  $\tau_i > t \Leftrightarrow A(t) \leq i-1$
  - On the other hand, we know that  $A(t) \sim \text{Poisson}(\mu t)$ . Thus,

$$P\{\tau_i > t\} = P\{A(t) \leq i-1\} = \sum_{j=0}^{i-1} \frac{(\mu t)^j}{j!} e^{-\mu t}$$

19

### Waiting time distribution (3)

- By combining the previous formulas, we get

$$\begin{aligned} P\{W > t\} &= \sum_{i=1}^{\infty} P\{\tau_i > t\} (1 - \rho) \rho^i \\ &= \sum_{i=1}^{\infty} \sum_{j=0}^{i-1} \frac{(\mu t)^j}{j!} e^{-\mu t} (1 - \rho) \rho^i \\ &= \rho \sum_{j=0}^{\infty} \frac{(\mu t \rho)^j}{j!} e^{-\mu t} (1 - \rho) \sum_{i=j+1}^{\infty} \rho^{i-(j+1)} \\ &= \rho \sum_{j=0}^{\infty} \frac{(\mu t \rho)^j}{j!} e^{-\mu t} = \rho e^{\mu t \rho} e^{-\mu t} = \rho e^{-\mu(1-\rho)t} \end{aligned}$$

20

### Waiting time distribution (4)

- Waiting time  $W$  can thus be presented as a product  $W = JD$  of two independent random variables  $J \sim \text{Bernoulli}(\rho)$  and  $D \sim \text{Exp}(\mu(1-\rho))$ :

$$P\{W = 0\} = P\{J = 0\} = 1 - \rho$$

$$P\{W > t\} = P\{J = 1, D > t\} = \rho \cdot e^{-\mu(1-\rho)t}, \quad t > 0$$

$$E[W] = E[J]E[D] = \rho \cdot \frac{1}{\mu(1-\rho)} = \frac{1}{\mu} \cdot \frac{\rho}{1-\rho}$$

$$E[W^2] = P\{J = 1\}E[D^2] = \rho \cdot \frac{2}{\mu^2(1-\rho)^2} = \frac{1}{\mu^2} \cdot \frac{2\rho}{(1-\rho)^2}$$

$$D^2[W] = E[W^2] - E[W]^2 = \frac{1}{\mu^2} \cdot \frac{\rho(2-\rho)}{(1-\rho)^2}$$

21

### Contents

- Refresher: Simple teletraffic model
- Queueing discipline
- M/M/1 (1 server,  $\infty$  waiting places)
- Application to packet level modelling of data traffic
- M/M/n ( $n$  servers,  $\infty$  waiting places)

22

### Application to packet level modelling of data traffic

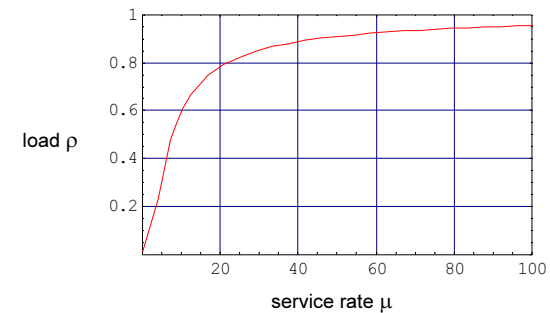
- M/M/1 model may be applied (to some extent) to packet level modelling of data traffic
  - customer = IP packet
  - $\lambda$  = packet arrival rate (packets per time unit)
  - $1/\mu$  = average packet transmission time (aikayks.)
  - $\rho = \lambda/\mu$  = traffic load
- Quality of service is measured e.g. by the packet delay
  - $P_z$  = probability that a packet has to wait "too long", i.e. longer than a given reference value  $z$

$$P_z = P\{W > z\} = \rho e^{-\mu(1-\rho)z}$$

23

### Multiplexing gain

- We determine load  $\rho$  so that prob.  $P_z < 1\%$  for  $z = 1$  (time units)
- Multiplexing gain** is described by the traffic load  $\rho$  as a function of the service rate  $\mu$



24

## Contents

- Refresher: Simple teletraffic model
- Queueing discipline
- M/M/1 (1 server,  $\infty$  waiting places)
- Application to packet level modelling of data traffic
- M/M/n ( $n$  servers,  $\infty$  waiting places)

25

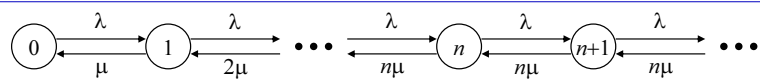
## M/M/n queue

- Consider the following simple teletraffic model:
  - Infinite number of independent customers ( $k = \infty$ )
  - Interarrival times are IID and exponentially distributed with mean  $1/\lambda$ .
    - so, customers arrive according to a Poisson process with intensity  $\lambda$
  - Finite number of servers ( $n < \infty$ )
  - Service times are IID and exponentially distributed with mean  $1/\mu$
  - Infinite number of waiting places ( $m = \infty$ )
  - Default queueing discipline: **FCFS**
- Using Kendall's notation, this is an **M/M/n queue**
  - more precisely: M/M/n-FCFS queue
- Notation:
  - $\rho = \lambda/(n\mu) =$  traffic load

26

## State transition diagram

- Let  $X(t)$  denote the number of customers in the system at time  $t$ 
  - Assume that  $X(t) = i$  at some time  $t$ , and consider what happens during a short time interval  $(t, t+h]$ :
    - with prob.  $\lambda h + o(h)$ , a new customer arrives (state transition  $i \rightarrow i+1$ )
    - if  $i > 0$ , then, with prob.  $\min\{i, n\} \cdot \mu h + o(h)$ , a customer leaves the system (state transition  $i \rightarrow i-1$ )
- Process  $X(t)$  is clearly a Markov process with state transition diagram



- Note that process  $X(t)$  is an irreducible birth-death process with an infinite state space  $S = \{0, 1, 2, \dots\}$

27

## Equilibrium distribution (1)

- Local balance equations (LBE) for  $i < n$ :

$$\pi_i \lambda = \pi_{i+1} (i+1) \mu \quad (\text{LBE})$$

$$\Rightarrow \pi_{i+1} = \frac{\lambda}{(i+1)\mu} \pi_i = \frac{n\rho}{i+1} \pi_i$$

$$\Rightarrow \pi_i = \frac{(n\rho)^i}{i!} \pi_0, \quad i = 0, 1, \dots, n$$

- Local balance equations (LBE) for  $i \geq n$ :

$$\pi_i \lambda = \pi_{i+1} n \mu \quad (\text{LBE})$$

$$\Rightarrow \pi_{i+1} = \frac{\lambda}{n\mu} \pi_i = \rho \pi_i$$

$$\Rightarrow \pi_i = \rho^{i-n} \pi_n = \rho^{i-n} \frac{(n\rho)^n}{n!} \pi_0 = \frac{n^n \rho^i}{n!} \pi_0, \quad i = n, n+1, \dots$$

28

### Equilibrium distribution (2)

- Normalizing condition (N):

$$\sum_{i=0}^{\infty} \pi_i = \pi_0 \left( \sum_{i=0}^{n-1} \frac{(n\rho)^i}{i!} + \sum_{i=n}^{\infty} \frac{n^n \rho^i}{n!} \right) = 1 \quad (\text{N})$$

$$\begin{aligned} \Rightarrow \pi_0 &= \left( \sum_{i=0}^{n-1} \frac{(n\rho)^i}{i!} + \frac{(n\rho)^n}{n!} \sum_{i=n}^{\infty} \rho^{i-n} \right)^{-1} \\ &= \left( \sum_{i=0}^{n-1} \frac{(n\rho)^i}{i!} + \frac{(n\rho)^n}{n!(1-\rho)} \right)^{-1} = \frac{1}{\alpha + \beta}, \text{ if } \rho < 1 \end{aligned}$$

$$\text{Notation: } \alpha = \sum_{i=0}^{n-1} \frac{(n\rho)^i}{i!}, \quad \beta = \frac{(n\rho)^n}{n!(1-\rho)}$$

29

### Equilibrium distribution (3)

- Thus, for a **stable** system ( $\rho < 1$ , that is:  $\lambda < n\mu$ ), the equilibrium distribution exists and is as follows:

$$\rho < 1 \Rightarrow$$

$$P\{X = i\} = \pi_i = \begin{cases} \frac{(n\rho)^i}{i!} \cdot \frac{1}{\alpha + \beta}, & i = 0, 1, \dots, n \\ \frac{n^n \rho^i}{n!} \cdot \frac{1}{\alpha + \beta}, & i = n, n+1, \dots \end{cases}$$

$$n = 1: \alpha = 1, \quad \beta = \frac{\rho}{1-\rho}, \quad \pi_0 = \frac{1}{\alpha + \beta} = 1 - \rho$$

$$n = 2: \alpha = 1 + 2\rho, \quad \beta = \frac{2\rho^2}{1-\rho}, \quad \pi_0 = \frac{1}{\alpha + \beta} = \frac{1-\rho}{1+\rho}$$

30

### Probability of waiting

- Let  $p_W$  denote the probability that an arriving customer has to wait
- Let  $X^*$  denote the number of customers in the system at an arrival time
- An arriving customer has to wait whenever all the servers are occupied at her arrival time. Thus,

$$p_W = P\{X^* \geq n\}$$

- PASTA:  $P\{X^* = i\} = P\{X = i\} = \pi_i$ . Thus,

$$p_W = P\{X^* \geq n\} = \sum_{i=n}^{\infty} \pi_i = \sum_{i=n}^{\infty} \pi_0 \cdot \frac{n^n \rho^i}{n!} = \pi_0 \cdot \frac{(n\rho)^n}{n!(1-\rho)} = \frac{\beta}{\alpha + \beta}$$

$$n = 1: p_W = \rho$$

$$n = 2: p_W = \frac{2\rho^2}{1+\rho}$$

31

### Mean number of waiting customers

- Let  $X_W$  denote the number of waiting customers in equilibrium
- Then

$$\begin{aligned} E[X_W] &= \sum_{i=n}^{\infty} (i-n) \pi_i = \pi_0 \frac{(n\rho)^n}{n!(1-\rho)} \sum_{i=n}^{\infty} (i-n) \cdot (1-\rho) \rho^{i-n} \\ &= p_W \cdot \frac{\rho}{1-\rho} \end{aligned}$$

$$n = 1: E[X_W] = p_W \cdot \frac{\rho}{1-\rho} = \frac{\rho^2}{1-\rho}$$

$$n = 2: E[X_W] = p_W \cdot \frac{\rho}{1-\rho} = \frac{2\rho^2}{1+\rho} \cdot \frac{\rho}{1-\rho} = \frac{2\rho^3}{1-\rho^2}$$

32



### Mean waiting time

- Let  $W$  denote the waiting time of a (typical) customer
- Little's formula:  $E[X_W] = \lambda \cdot E[W]$ . Thus,

$$E[W] = \frac{E[X_W]}{\lambda} = \frac{1}{\lambda} \cdot p_W \cdot \frac{\rho}{1-\rho} = \frac{1}{\mu} \cdot \frac{p_W}{n(1-\rho)} = p_W \cdot \frac{1}{n\mu - \lambda}$$

$$n = 1: E[W] = \frac{1}{\mu} \cdot \frac{p_W}{1-\rho} = \frac{1}{\mu} \cdot \frac{\rho}{1-\rho}$$

$$n = 2: E[W] = \frac{1}{\mu} \cdot \frac{p_W}{2(1-\rho)} = \frac{1}{\mu} \cdot \frac{\rho^2}{1-\rho^2}$$

33

### Mean delay

- Let  $D$  denote the total time (delay) in the system of a (typical) customer
  - including both the waiting time  $W$  and the service time  $S$ :  $D = W + S$
- Then,

$$E[D] = E[W] + E[S] = \frac{1}{\mu} \cdot \left( \frac{p_W}{n(1-\rho)} + 1 \right) = p_W \cdot \frac{1}{n\mu - \lambda} + \frac{1}{\mu}$$

$$n = 1: E[D] = \frac{1}{\mu} \cdot \left( \frac{p_W}{1-\rho} + 1 \right) = \frac{1}{\mu} \cdot \left( \frac{\rho}{1-\rho} + 1 \right) = \frac{1}{\mu} \cdot \frac{1}{1-\rho}$$

$$n = 2: E[D] = \frac{1}{\mu} \cdot \frac{p_W}{2(1-\rho)} + \frac{1}{\mu} \cdot \left( \frac{\rho^2}{1-\rho^2} + 1 \right) = \frac{1}{\mu} \cdot \frac{1}{1-\rho^2}$$

34

### Mean queue length

- Let  $X$  denote the number of customers in the system (queue length) in equilibrium
- Little's formula:  $E[X] = \lambda \cdot E[D]$ . Thus,

$$E[X] = \lambda \cdot E[D] = p_W \cdot \frac{\lambda}{n\mu - \lambda} + \frac{\lambda}{\mu} = p_W \cdot \frac{\rho}{1-\rho} + n\rho$$

$$n = 1: E[X] = p_W \cdot \frac{\rho}{1-\rho} + \rho = \rho \cdot \frac{\rho}{1-\rho} + \rho = \frac{\rho}{1-\rho}$$

$$n = 2: E[X] = p_W \cdot \frac{\rho}{1-\rho} + 2\rho = \frac{2\rho^2}{1+\rho} \cdot \frac{\rho}{1-\rho} + 2\rho = \frac{2\rho}{1-\rho^2}$$

35

### Waiting time distribution (1)

- Let  $W$  denote the waiting time of a (typical) customer
- Let  $X^*$  denote the number of customers in the system at the arrival time
- The customer has to wait only if  $X^* \geq n$ . This happens with prob.  $p_W$ .
- Under the assumption that  $X^* = i \geq n$ , the system, however, looks like an ordinary M/M/1 queue with arrival rate  $\lambda$  and service rate  $n\mu$ .
  - Let  $W'$  denote the waiting time of a (typical) customer in this M/M/1 queue
  - Let  $X^{**}$  denote the number of customers in the system at the arrival time
- It follows that

$$P\{W = 0\} = 1 - p_W$$

$$P\{W > t\} = P\{X^* \geq n\} P\{W > t \mid X^* \geq n\}$$

$$= p_W \cdot P\{W' > t \mid X^{**} \geq 1\} = p_W \cdot e^{-n\mu(1-\rho)t}, \quad t > 0$$

36

### Waiting time distribution (2)

- Waiting time  $W$  can thus be presented as a product  $W = JD'$  of two indep. random variables  $J \sim \text{Bernoulli}(p_W)$  and  $D' \sim \text{Exp}(n\mu(1-\rho))$ :

$$P\{W = 0\} = P\{J = 0\} = 1 - p_W$$

$$P\{W > t\} = P\{J = 1, D' > t\} = p_W \cdot e^{-n\mu(1-\rho)t}, \quad t > 0$$

$$E[W] = E[J]E[D'] = p_W \cdot \frac{1}{n\mu(1-\rho)} = \frac{1}{\mu} \cdot \frac{p_W}{n(1-\rho)}$$

$$E[W^2] = P\{J = 1\}E[D'^2] = p_W \cdot \frac{2}{n^2\mu^2(1-\rho)^2} = \frac{1}{\mu^2} \cdot \frac{2p_W}{n^2(1-\rho)^2}$$

$$D^2[W] = E[W^2] - E[W]^2 = \frac{1}{\mu^2} \cdot \frac{p_W(2-p_W)}{n^2(1-\rho)^2}$$

37

### Example (1)

- Printer problem
  - Consider the following two different configurations:
    - One rapid printer (IID printing times  $\sim \text{Exp}(2\mu)$ )
    - Two slower parallel printers (IID printing times  $\sim \text{Exp}(\mu)$ )
  - Criterion: minimize mean delay  $E[D]$ 
    - One rapid printer (M/M/1 model with  $\rho = \lambda/(2\mu)$ ):

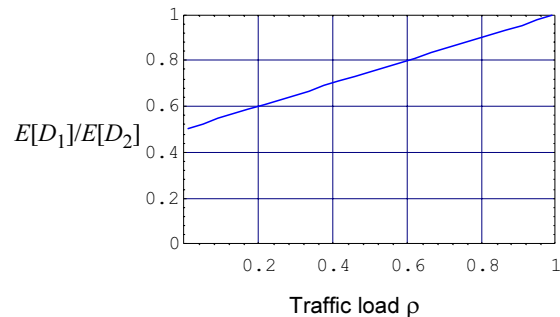
$$E[D_1] = \frac{1}{2\mu} \cdot \frac{1}{1-\rho}$$

- Two slower printers (M/M/2 model with  $\rho = \lambda/(2\mu)$ ):

$$E[D_2] = \frac{1}{\mu} \cdot \frac{1}{1-\rho^2} = \frac{1}{2\mu} \cdot \frac{2}{(1-\rho)(1+\rho)} = E[D_1] \cdot \frac{2}{1+\rho} > E[D_1]$$

38

### Example (2)



39