

#### **OPERATIONS & LOGISTICS MANAGEMENT IN AIR TRANSPORTATION**

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# VARIABILITY IN PROCESSES AND QUEUES





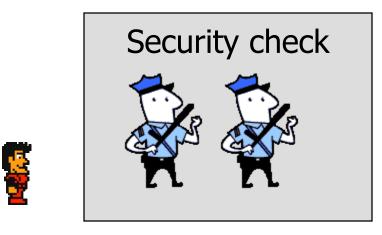
## LEARNING OBJECTIVES

- Variability and Process Analysis
  - What is *variability*?
  - What *impact* does variability have on processes?
  - How can we *quantify* the impact of variability on processes?
  - How can we *manage* variability in processes?





## WHAT IS VARIABILITY?



#### Variability comes from ...







#### TYPES OF VARIABILITY

#### **Predictable Variability**

... refers to "knowable" changes in input and/or capacity rates

Demand of pumpkins will go up during Thanksgiving

#### **Unpredictable Variability**

... refers to "unknowable" changes in input and/or capacity rates

Supply of pumpkins will go down *if* the crop fails

- Both types of variability exist simultaneously
  - Pumpkin sales will go up during Thanksgiving, but we do not know the exact sales of pumpkins





# **Predictable Variability**

#### **Unpredictable Variability**

# Can be *controlled* by making changes to the system

- We could increase or decrease the demand for pumpkins by increasing or decreasing the price
- Restaurants will add staff during peak demand (lunch, dinner, etc.)

Is the result of the *lack* of knowledge or information

- Usually can be expressed with a probability distribution
- E.g., Express the probability that the pumpkin crop will fail using a probability distribution

Can be *reduced* by gaining more knowledge or collection information

• By paying close attention to weather patterns, we could increase the accuracy of our prediction that the pumpkin crop will fail

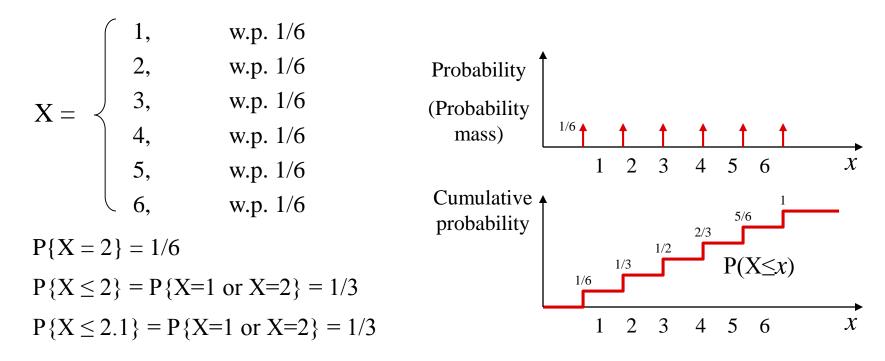




## SHORT REVIEW ON PROBABILITY (1)

#### **Discrete** Random Variable and Probability

Throw a dice; the number you get is a discrete random variable:



 $P{X \le x}$  is a function of *x*, called the **cumulative distribution function (CDF)** 

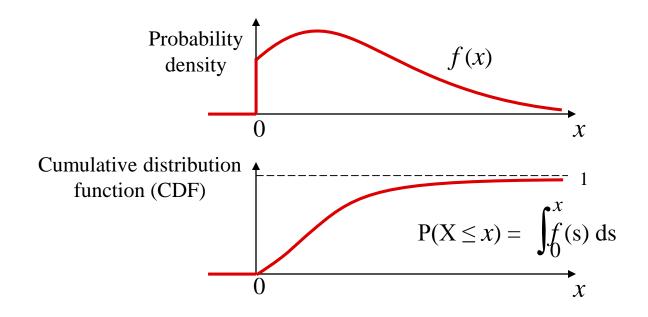




## SHORT REVIEW ON PROBABILITY (2)

**Continuous** Random Variable and Probability

The time between two customers' arrival times is a continuous random variable







# BASIC QUESTIONS

- What are the effects of variability on processes
  - In particular, how does variability affect



• If the effects are negative, how can we deal with it?

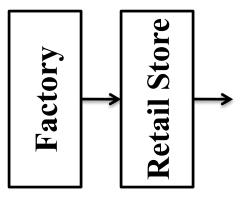




## THE "MAKE-TO-ORDER" DICE GAME

- Retail store will roll dice first to observe demand, which will be communicated to the factory
- Factory will roll dice to observe capacity
- Factory will satisfy retailer demand, but is constrained by realized capacity
  - For example, if demand is 3 and capacity is 4, then factory will give the retail store 3 units
  - But if demand is 5 and capacity is 4, then factory will give the retail store only 4 units
  - No backlog
- Assume 1 roll of demand and capacity

= 1 day

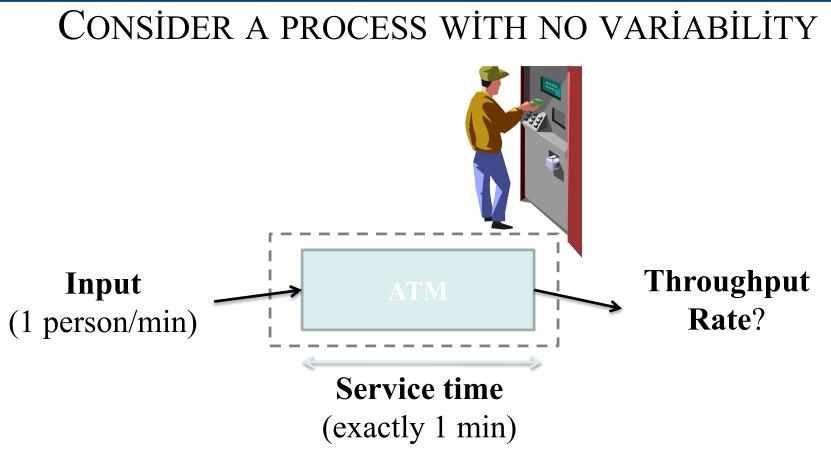


What is the average demand?

What is the average				
capacity?				



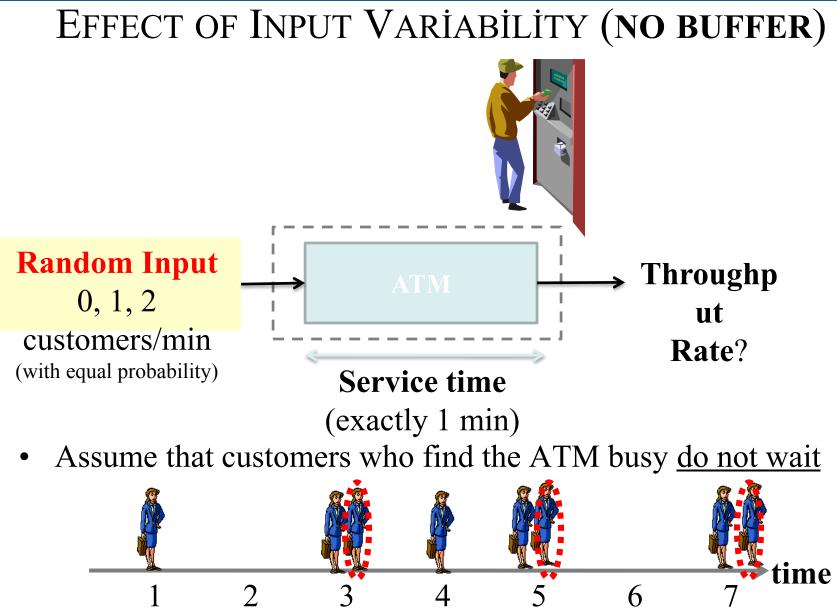




- Assume that all customers are identical
- Customers arrive exactly 1 minute apart
- The service time is exactly 1 minute for all the customers











# EFFECT OF INPUT VARIABILITY (NO BUFFER)

- When a process faces input variability, and a buffer cannot be built, some input may get lost
- Input variability *can* reduce the throughput
- Lower throughput means
  - Lost customers; lost revenue
  - Customer dissatisfaction
  - Less utilization of resources
- Little's Law holds





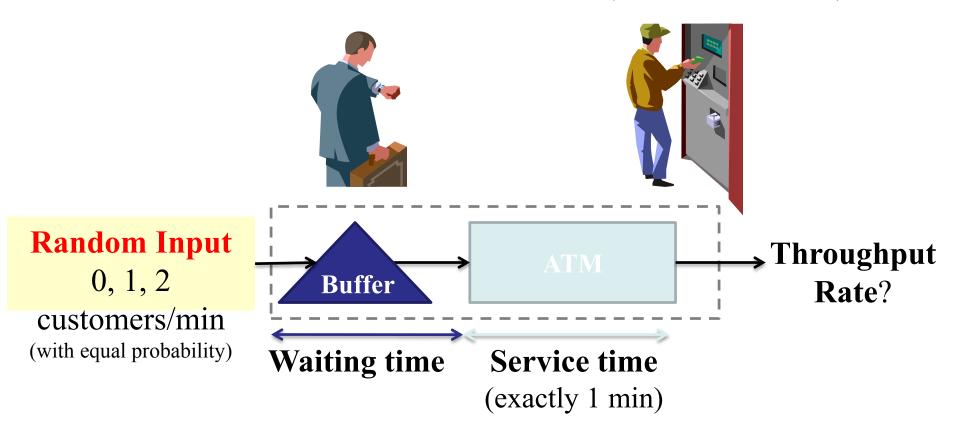
## DEALING WITH VARIABILITY

• When the arrival rate of customers is unpredictable, what could you do to increase throughput?





#### EFFECT OF INPUT VARIABILITY (WITH BUFFER)



• Now assume that <u>customers wait</u> We can build-up an inventory buffer





#### EFFECT OF INPUT VARIABILITY (WITH BUFFER)

- If we can build up an inventory buffer, variability leads to
  - An increase in the average inventory in the process
  - An increase in the average flow time
- Little's Law holds





### THE OM TRIANGLE

**CAPACITY INVENTORY INFORMATION** (Variability **Reduction**)

If a firm is striving to meet the *random* demand, then it can use **capacity**, **inventory** and **information** (variability reduction) as <u>substitutes</u>

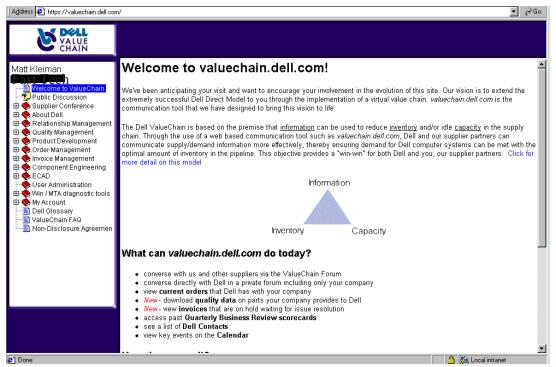
You cannot have low inventory, low capacity, and low information acquisition effort at the same time. This is a <u>trade-off</u>.





## **OPERATIONS AT DELL**

- Inventory as "the physical embodiment of bad information" (a senior exec at Dell)
- Substitute information for inventory
- Less inventory =>higher inventory turns







# QUANTIFYING VARIABILITY

- So far, we focused on **qualitative** effect of variability
  - Without buffer, input may get lost and throughput may decrease
  - With buffer, queue may build up, flow time may increase
- But ...
  - How long is the queue on average?
  - How long does a customer have to wait?





#### WHY IS IT IMPORTANT TO QUANTIFY VARIABILITY AND ITS IMPACT?

These quantitative measures of process performance are important to any functions of a company

Marketing Wants to use the short waiting time as a selling point	Finance Wants to attract investors based on excellent operations performance
Accounting Wants to know how much money is tied up in the queue	Operations Wants to shorten the queue, and wants to quantify the trade-offs between capacity, inventory and variability What is the impact (on inventory and flow time) of increasing/decreasing capacity by 10%?





# First Steps in Quantifying Variability

- Probability Statements P(X=4) $P(20 < T \le 30)$
- Variances and Standard deviation
  - These lead to probability statements
- Coefficient of variation (CV)

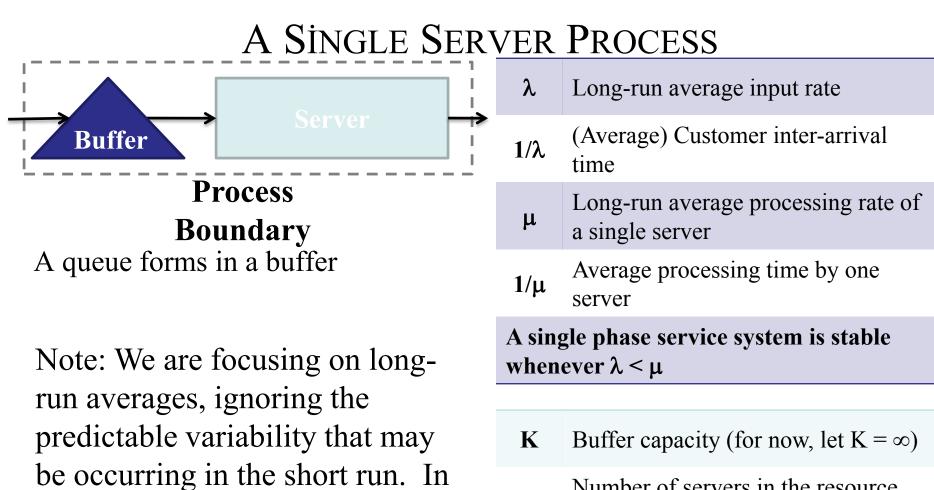
$$CV = \frac{Standard Deviation}{Mean}$$



reality, we should be concerned

with both types of variability



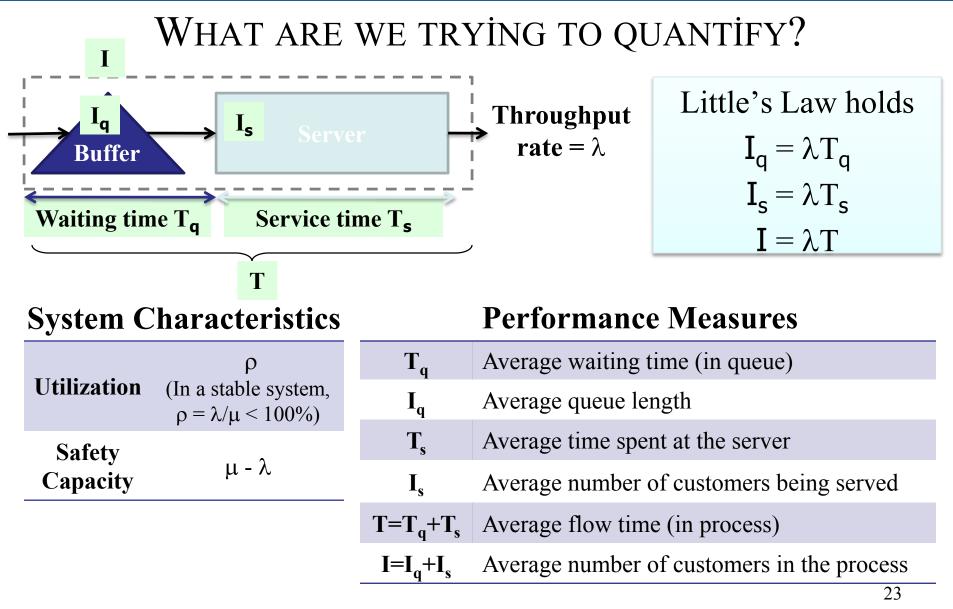


С

Number of servers in the resource pool (for now, let c=1)

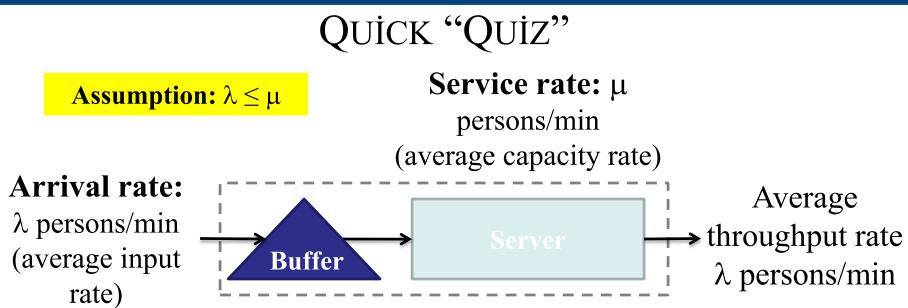












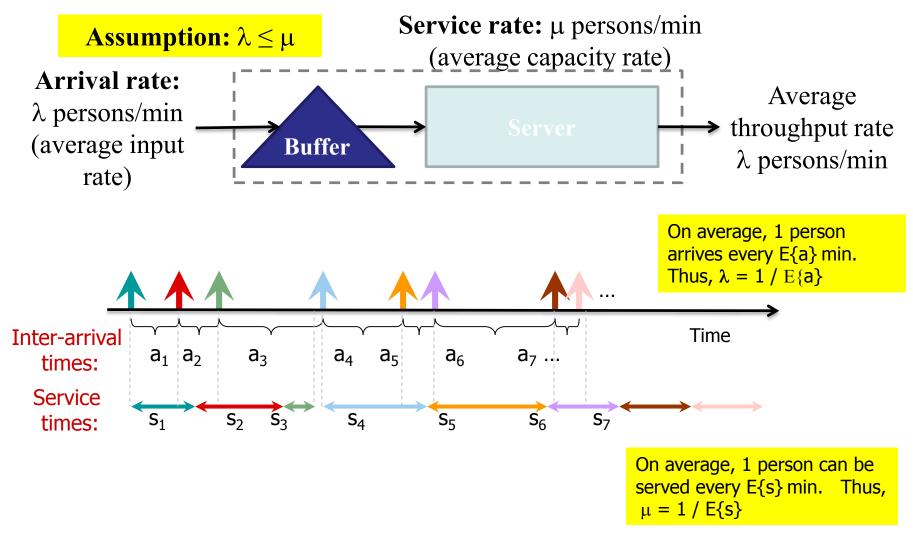
• Average number of persons in the system:  $I = I_q + I_s$ 

• Question:  $I_s = ???$  (Express  $I_s$  in terms of  $\lambda$  and  $\mu$ )





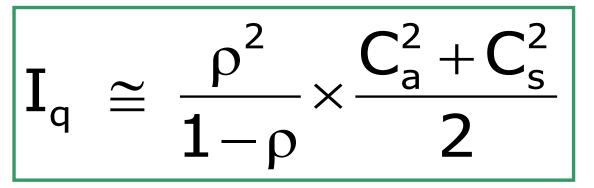
## SINGLE-SERVER QUEUING MODEL







## POLLACZEK-KHINCHIN (PK) FORMULA



"=" for special cases

"≈" in general

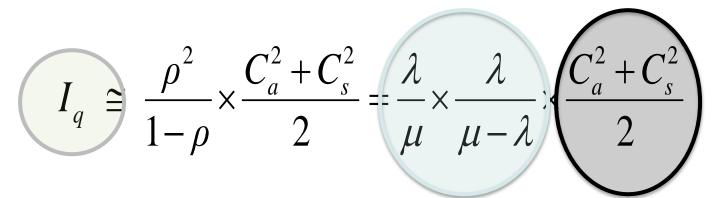
Iq	Average queue length (excl. the one in service)		
ρ	(Long run) Average utilization = Average Throughput / Average Capacity = $\lambda / \mu$		
$C_a = \sigma\{a\}/E\{a\}$	Coefficient of variation (CV) of inter-arrival times		
$C_s = \sigma\{s\}/E\{s\}$	Coefficient of variation (CV) of service times		



**INVENTORY** 



#### PK FORMULA AND OM TRİANGLE



 $\mu = Capacity Rate$  $\lambda = Input Rate$ 

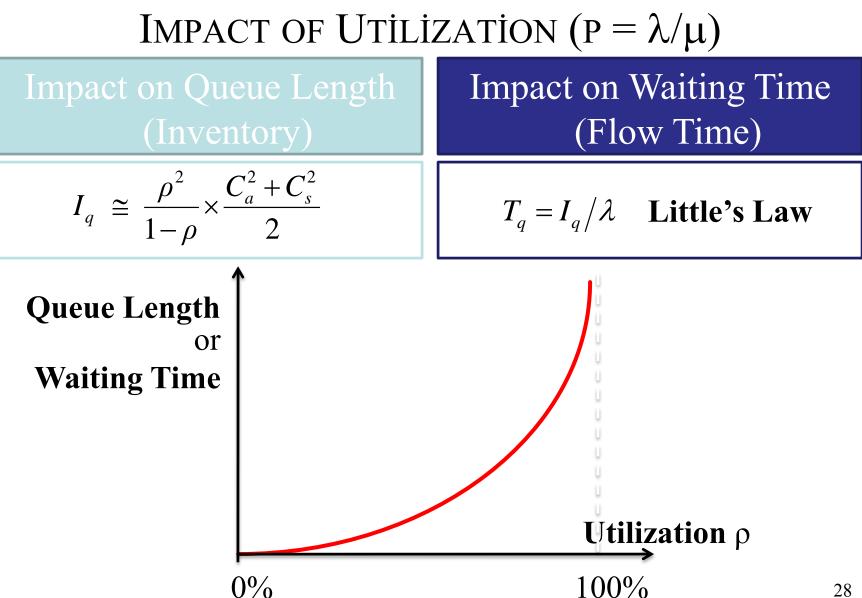
Variability

CAPACITY













# **Utilizatio**n

Utilizatio n =	_ Throughput Rate _	- Actual output rate $< 100%$	-≤100% e
	Capacity Rate	maximum output rate	

- Utilization gives us information about "excess capacity"
- The utilization of each resource in a process can be presented with a *utilization profile*

Resource	Capacity Rate (units/hour)	Input Rate (units/hour)	Utilization
1	6	4	66.67%
2	7	4	57.14%
3	8	4	50.00%
4	6	4	66.67%
5	5	4	80.00%

• What is the optimal utilization of a resource?





## UTILIZATION: AN IMPORTANT INSIGHT

#### With No Variability

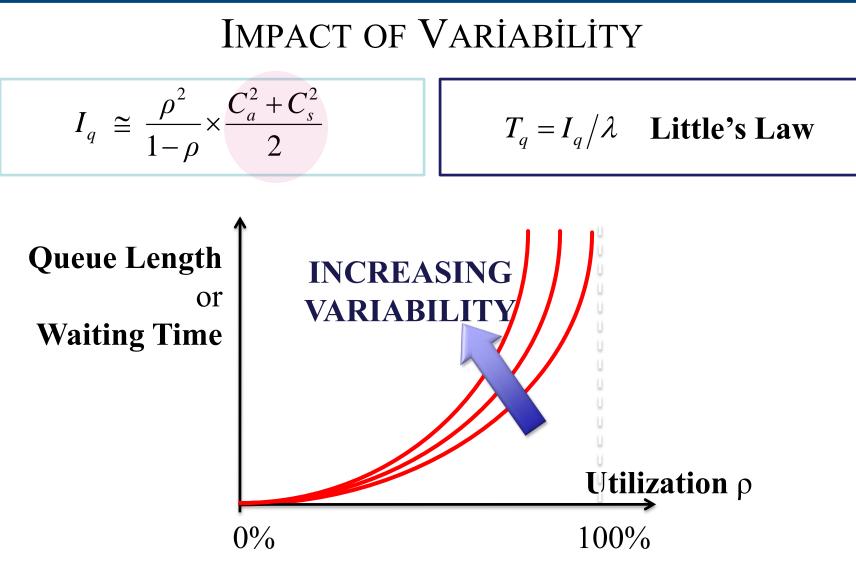
• Maximizing utilization is a good idea in a process with no variability

#### With Variability

- Maximizing utilization is a very bad idea in a process with variability
- What is the correct utilization for a resource when variability is present?
- It depends ... on the amount of variability, the sensitivity to delay, etc.











## QUEUING THEORY

The PK formula given above comes from "queuing theory", the study of queues

The version of PK formula we used above makes the following assumptions

#### Single server

Single queue

No limit on queue length

All units that arrive enter the queue system stays in the queue till served

(No units "balk" at the length of the queue)

First-in-first-out (FIFO)

All units arrive independently of each other





# QUEUING NOTATION: G/G/1 QUEUE

• The queue we studied above is called a

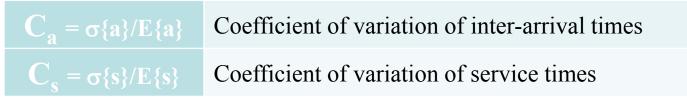
#### G/G/1 queue

The first **"G"** refers to the fact that the "**arrivals**" follows a "general" (probability) distribution

The second **"G"** refers to the fact that the "**service time**" follows a "general" (probability) distribution

The "1" refers to the fact that there is a **single server** 

• Using observed data, get estimates for C<sub>a</sub> and C<sub>a</sub>







#### SIMPLE EXAMPLE

- Customers arrive at rate 4/hour, and mean service time is 10 minutes
- Assume that standard deviation of inter-arrival times equals 5 minutes, and the standard deviation of service time equals 3 minutes
- What is the average size of the queue? What is the average time that a flow unit spends in the queue?

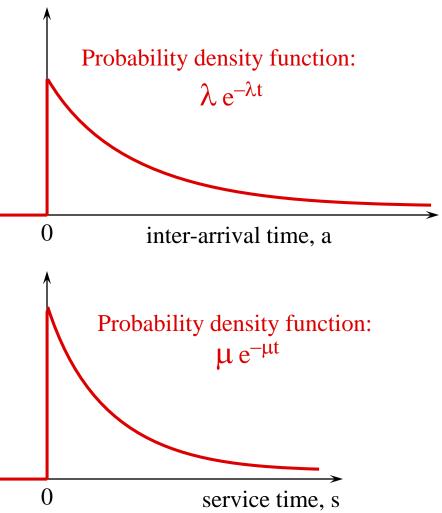
 $\lambda = 4$  E[a] = 1/4 hour  $\mu = 6$  E[s] = 1/6 hour  $\rho = \lambda/\mu = 4/6 = 2/3$  $\sigma[a] = 1/12 \text{ hour}$   $C_a = \frac{\sigma[a]}{E[a]} = \frac{1/12}{1/4} = \frac{1}{3}$  $\sigma[s] = 1/20 \text{ hour}$   $C_s = \frac{\sigma[s]}{E[s]} = \frac{1/20}{1/6} = \frac{3}{10}$  $I_q \cong \frac{\rho^2}{1-\rho} \times \frac{C_a^2 + C_s^2}{2} = \frac{(2/3)^2}{1/3} \times \frac{(1/3)^2 + (3/10)^2}{2}$  $T_a = I_a / \lambda = I_a / 4$ 





#### WHAT IF WE DON'T HAVE DATA ABOUT THE PROCESS?

- Suppose you start a service business. You haven't seen the actual customers arrival process, but you want to have some idea about the queue you will be facing.
- Need to make some assumptions about the customer arrival process, and service time distribution
- A mostly commonly used distribution is the exponential distribution







#### WHY USE THESE ASSUMPTIONS?

- In many situations, the exponential distribution assumption is a good approximation for what really happens
- Easy to analyze because coefficient of variation (CV) is 1 for exponential distributions

$$CV = \frac{Standard Deviation}{Mean}$$

• Recall the P-K formula

$$I_q \cong \frac{\rho^2}{1-\rho} \times \frac{C_a^2 + C_s^2}{2} =$$
 ????





# M/M/1 QUEUE

The first **"M"** indicates that the **inter-arrival times** are **exponentially** distributed

The second **"M"** indicates that the **service times** are **exponentially** distributed

The "1" refers to the fact that there is a **single server** 

- Assume First-Come First-Serve (FCFS) rule
- For M/M/1 queue, the P-K formula is *exact* (=, not  $\approx$ )  $I_q = \frac{\rho^2}{1-\rho} = \frac{\lambda^2}{\mu(\mu-\lambda)}$

$$I = I_q + I_s = ???$$
$$T = T_q + T_s = ???$$

• Average waiting time in queue (Little's Law)  $T_q = I_q / \lambda$ 





### SIMPLE EXAMPLE

• Customers arrive at rate 4/hour, and mean service time is 10 minutes

• What is the average size of the queue? What is the average time that a flow unit spends in the queue?  $\lambda = 4 \quad E[a] = 1/4 \text{ hour}$  $\mu = 6 \quad E[s] = 1/6 \text{ hour}$  $\rho = \lambda/\mu = 4/6 = 2/3$ 





## PRACTICE PROBLEM

- Professor Longhair holds office hours everyday to answer students' questions.
- Students arrive at an average rate of 50 per hour.
- Professor Longhair can process students at an average rate of 60 per hour.
- What is the average number of students waiting outside Professor Longhair's office, and how long do they wait on average?
- Assume the inter-arrival time and the service time are both exponentially distributed

(We can also say that the arrival rate follows a **Poisson** distribution)

$$\lambda = 50$$
  

$$\mu = 60$$
  

$$\rho = \lambda/\mu = 50/60 = 5/6$$
  

$$I_q = \frac{\lambda^2}{\mu(\mu - \lambda)} = \frac{50^2}{60(60 - 50)} = \frac{25}{6}$$
  

$$T_q = \frac{I_q}{\lambda} = \frac{25/6}{50} = \frac{1}{12}$$





# M/D/1 Queue

The first **"M"** indicates that the **inter-arrival times** are **exponentially** distributed

The second **"D"** indicates that the **service times** are a **constant** 

The "1" refers to the fact that there is a single server

- Assume First-Come First-Serve (FCFS) rule
- For M/D/1 queue, the P-K formula gives

$$I = I_q + I_s = ???$$

$$I_q = \frac{\rho^2}{1-\rho} \times \frac{1}{2} = \frac{\lambda^2}{2\mu(\mu-\lambda)}$$

• Average waiting time in queue

(Little's Law)  $T_q = I_q / \lambda$ 

 $T = T_q + T_s = ???$ 





### SIMPLE EXAMPLE

• Customers arrive at rate 4/hour, and mean service time is *exactly* 10 minutes

• What is the average size of the queue? What is the average time that a flow unit spends in the queue?

$$\lambda = 4 \quad E[a] = 1/4 \text{ hour}$$

$$\mu = 6 \quad E[s] = 1/6 \text{ hour}$$

$$\rho = \lambda/\mu = 4/6 = 2/3$$

$$I_q = \frac{\lambda^2}{2\mu(\mu - \lambda)} = \frac{4^2}{2 \times 6(6 - 4)} = \frac{2}{3}$$

$$T_q = \frac{I_q}{2} = \frac{2/3}{4} = \frac{1}{6}$$

4

6

λ





## OTHER TYPES OF QUEUES

- Multiple servers
- Limited buffer size

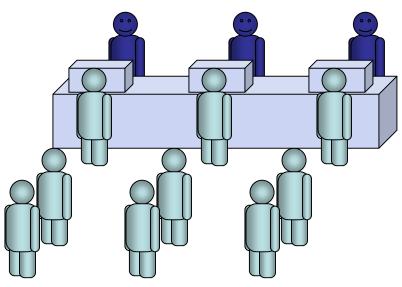




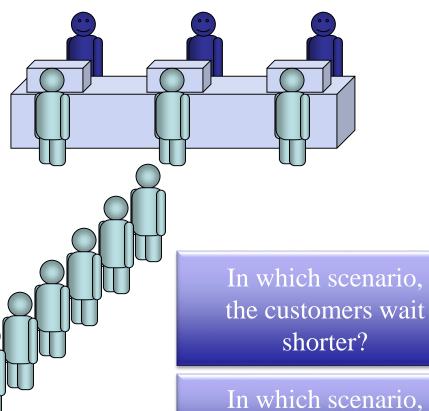
### WHICH TYPE OF QUEUE DO YOU PREFER?

Type 1

#### Type 2



Same arrival processes and the service capacities

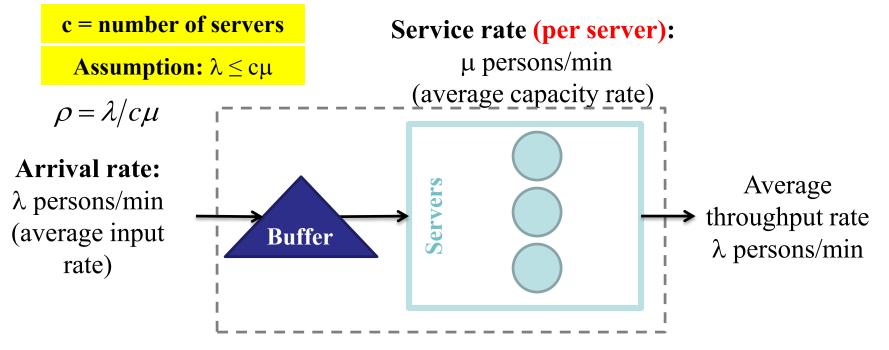


the queue is shorter?









- Customers only form one queue
- The first customer in the queue will be served by the next empty server



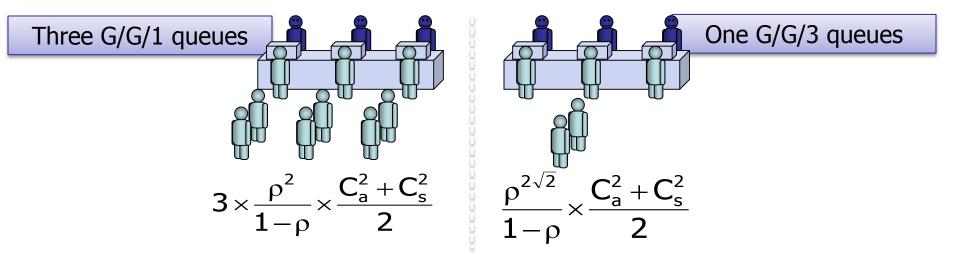


MULTI-SERVER QUEUE: P-K FORMULA

$$I_{q} \cong \frac{\rho^{\sqrt{2(c+1)}}}{1-\rho} \times \frac{C_{a}^{2} + C_{s}^{2}}{2}$$

$$ho = \lambda / c \mu$$

 All other things being equal, if the number of servers c increases, then I<sub>q</sub> decreases



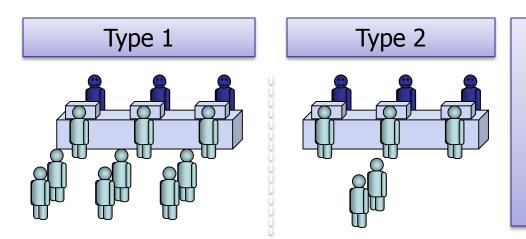
• "Risk Pooling" decreases queue length dramatically

Recommended reading: "A long line for a shorter wait at the supermarket" New York Times, June 23, 2007.





### **Risk pooling** or Demand Aggregation



The inventory in queue and wait time is reduced in an G/G/c queue (as compared to c number of G/G/1 queues)

Why does it make sense?

Independent demand streams impose greater variability when compared to a "pooled" demand stream

Approach: Adding independent random variables

**Example Applications:** 

Component commonality in product design Portfolio effects in finance Safety stock





## DICE GAME REVISITED





# M/M/C QUEUE

The first **"M"** indicates that the **inter-arrival times** are **exponentially** distributed

The second **"M"** indicates that the **service times** are **exponentially** distributed

The last "c" indicates c servers

- Assume First-Come First-Serve (FCFS) rule
- For M/M/1 queue, the P-K formula is

$$I_q \cong \frac{\rho^{\sqrt{2(c+1)}}}{1-\rho}$$

• Note:  $C_q$  and  $C_s$  are equal to 1 because of the exponential distribution assumption





### SUMMARY

- In systems with variability, averages do not tell the whole story
- Unpredictable variability can cause loss of throughput rate
- Inventory buffers or increased capacity may be needed to deal with variety
- In variable systems, inventory and flow time increase nonlinearly with utilization (see the P-K formula)
- The impact of variability (on inventory and flow time) can be quantified using the P-K formula, Little's Law, and assumptions about the probability distributions of variability
- "Risk pooling" reduces queue length and wait times