

COMP9334

# Capacity Planning for Computer Systems and Networks

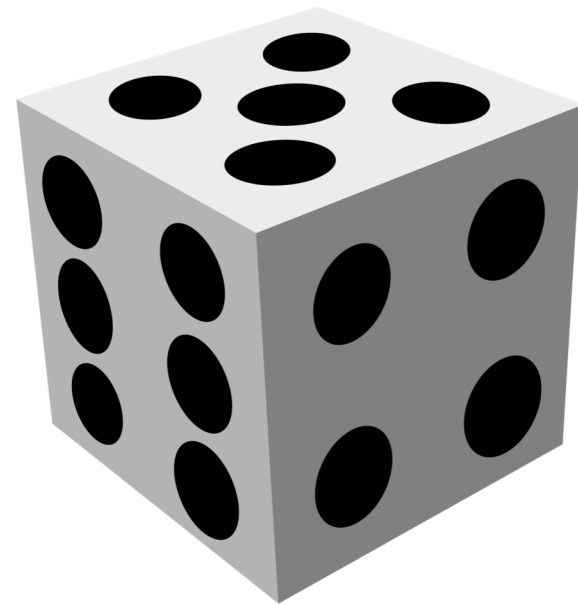
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Week 3A: Queues with Poisson arrivals (2)

## Pre-lecture exercise

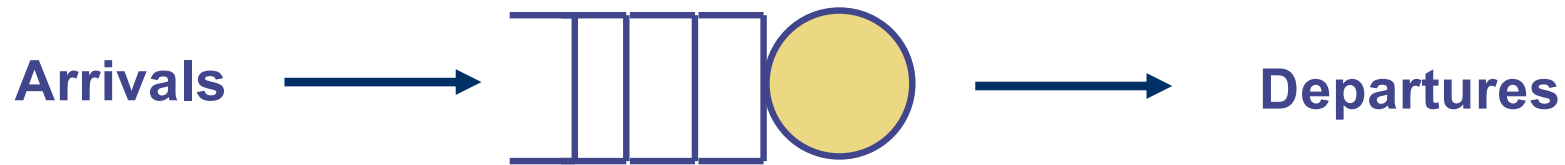
- You have a loaded die with 6 faces with values 1, 2, 3, 4, 5 and 6
- The probability that you can get each face is given in the table below
- What is the mean value that you can get?

Value	Probability
1	0.1
2	0.1
3	0.2
4	0.1
5	0.3
6	0.2



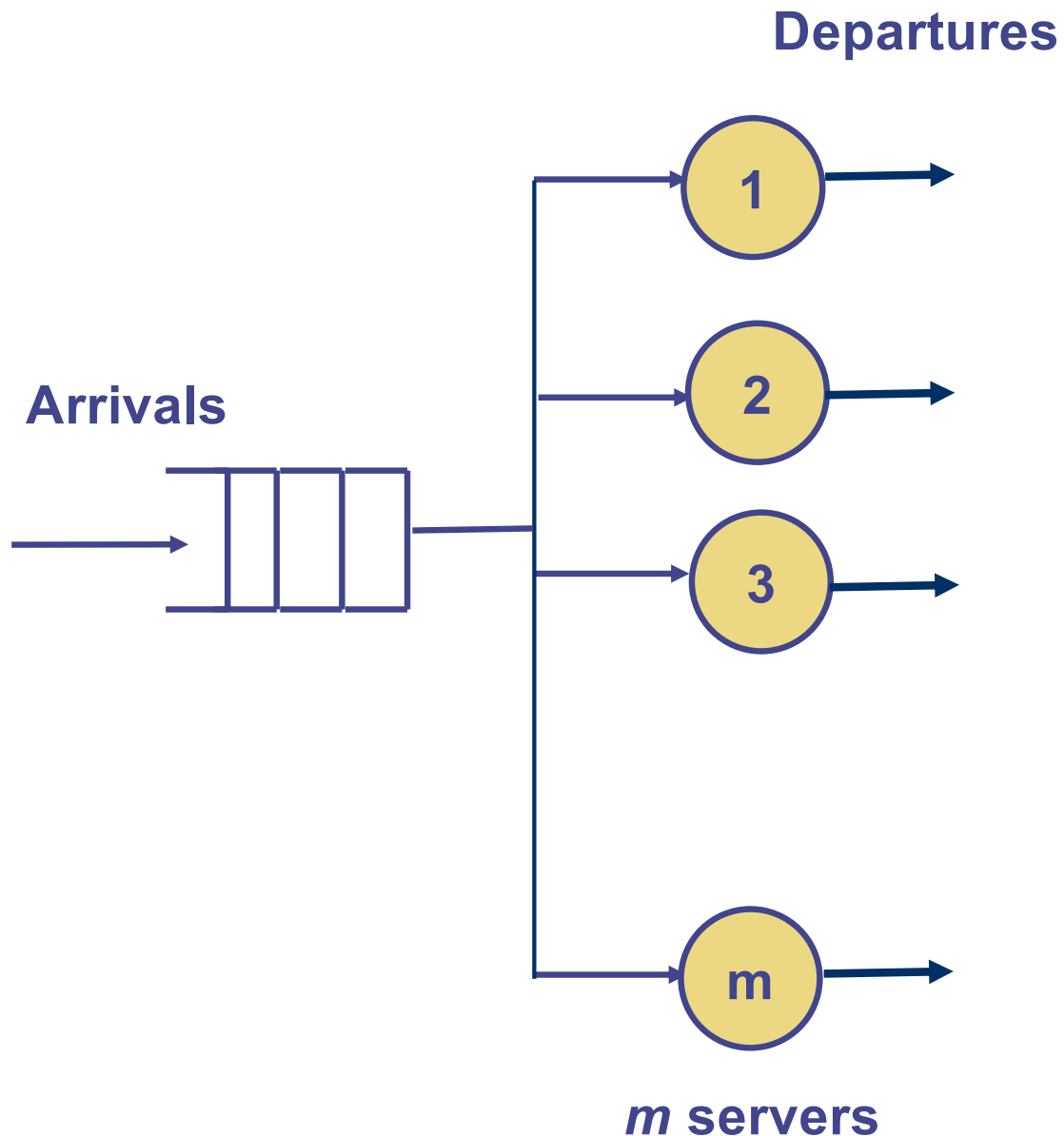
# Single-server queue

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- Open, single server queues
- How to find:
  - Waiting time
  - Response time
  - Mean queue length etc.
- The technique to find waiting time etc. is called *Queueing Theory*

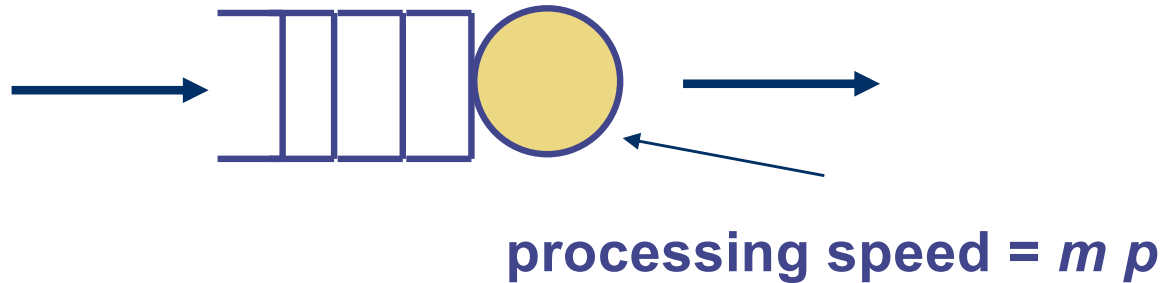
# Multiple server queue



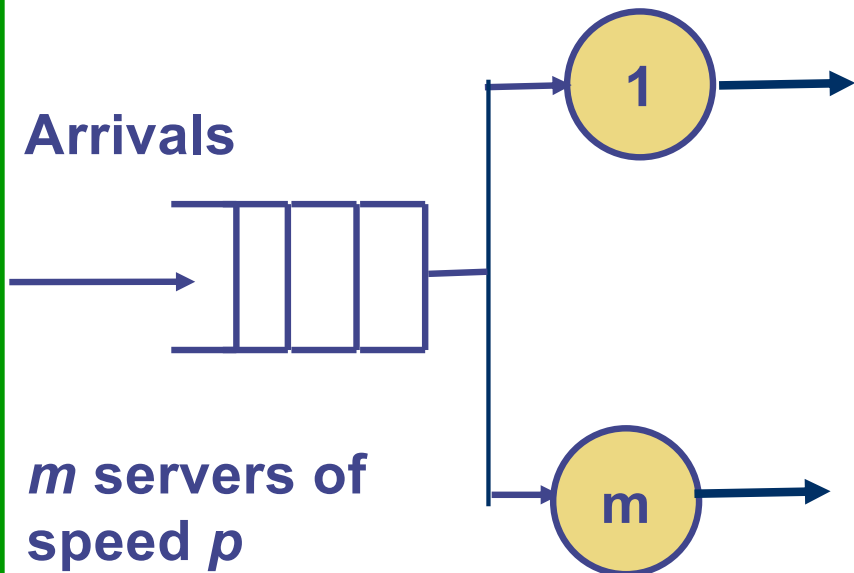
- Open, multi-server queue
- How to find:
  - Waiting time
  - Response time
  - Mean queue length etc.

# What will you be able to do with the results?

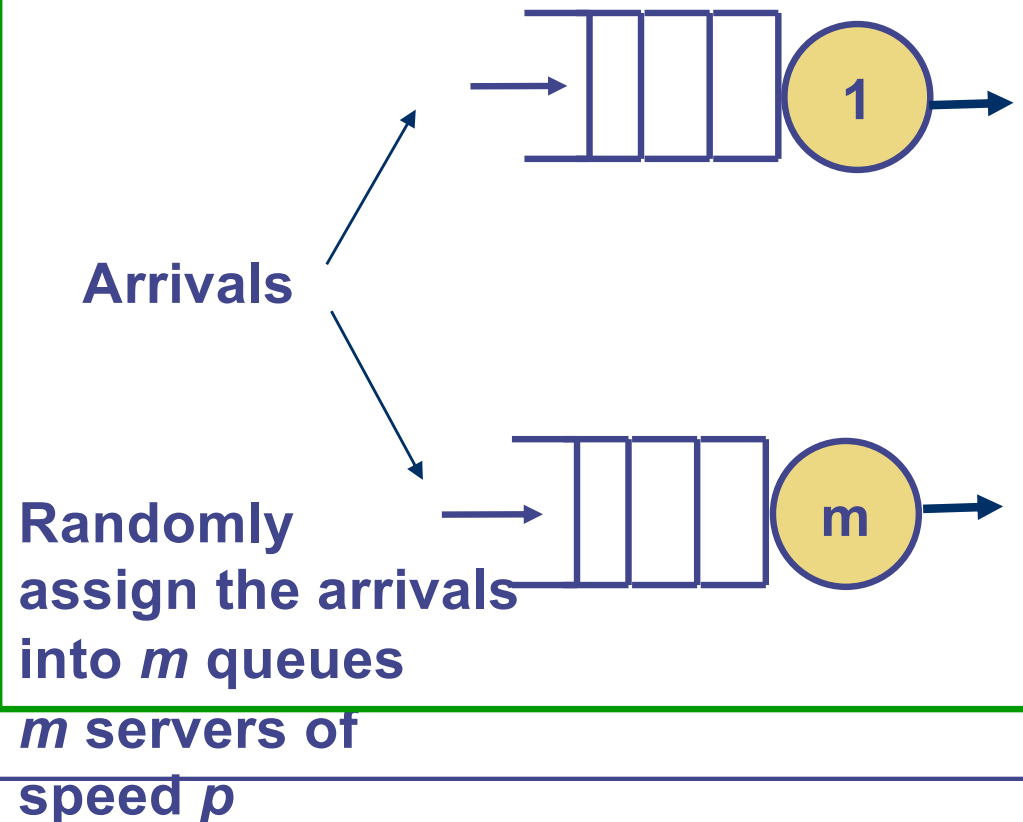
Configuration 1:



Configuration 2:

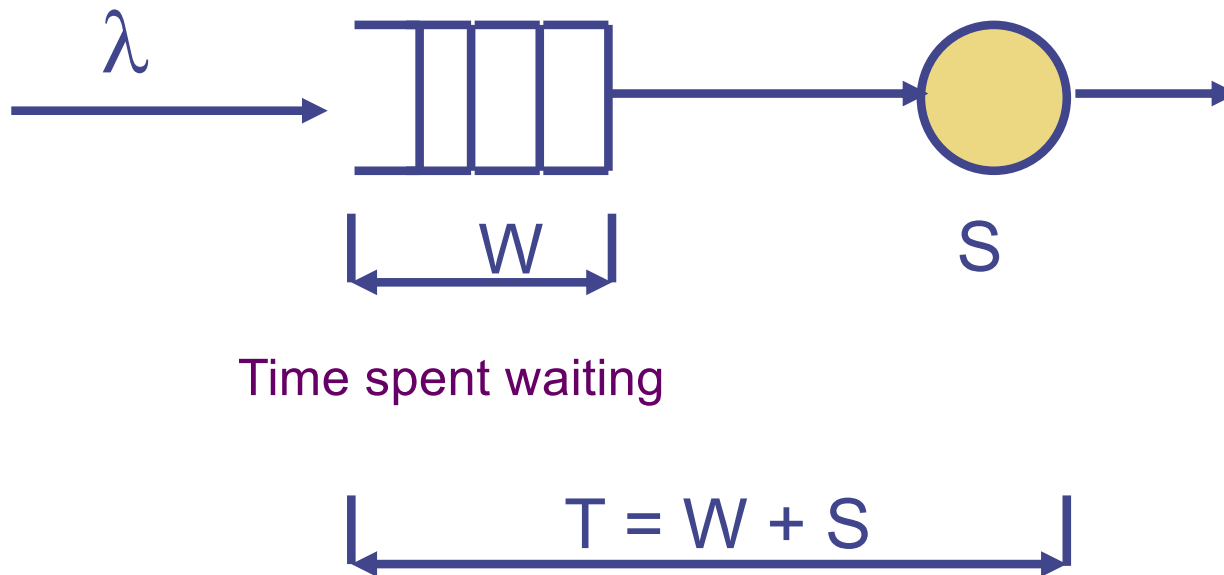


Configuration 3:



Which configuration has the best response time?

# Single Server Queue: Terminology



**Response Time  $T$**

**= Waiting time  $W$  + Service time  $S$**

Note: We use  $T$  for response time because this is the notation in many queueing theory books. For a similar reason, we will use  $\rho$  for utilisation rather than  $U$ .

# Call centre analogy from Week 2B

- Consider a call centre
  - Calls are arriving according to Poisson distribution with rate  $\lambda$
  - The length of each call is exponentially distributed with parameter  $\mu$ 
    - Mean length of a call is  $1/\mu$

**Call centre:**

**Arrivals**



**$m$  operators**

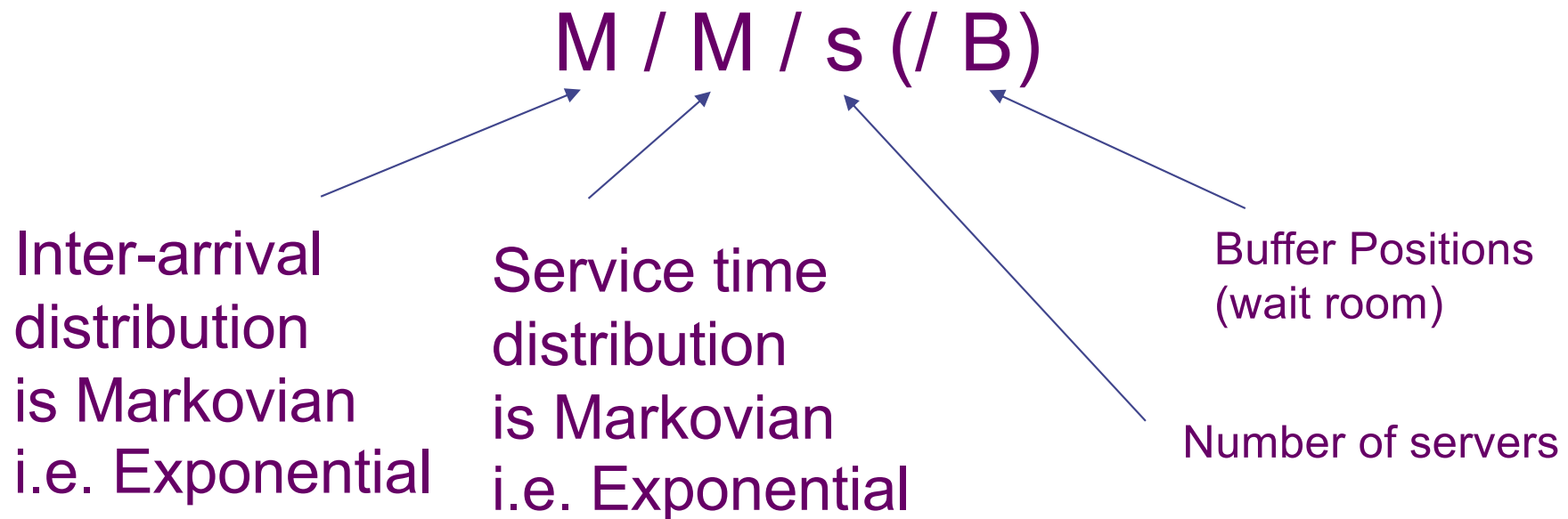
**If all operators are busy, the centre can put at most  $n$  additional calls on hold.**

**If a call arrives when all operators and holding slots are used, the call is rejected.**

- We solved the problems for
  - $(m = 1 \text{ and } n = 0)$ , and  $(m = 1 \text{ and } n = 1)$
- How about other values of  $m$  and  $n$ ? What about response time?

# Kendall's notation

- To represent different types of queues, queueing theorists use the Kendall's notation
- The call centre example on the previous page can be represented as:



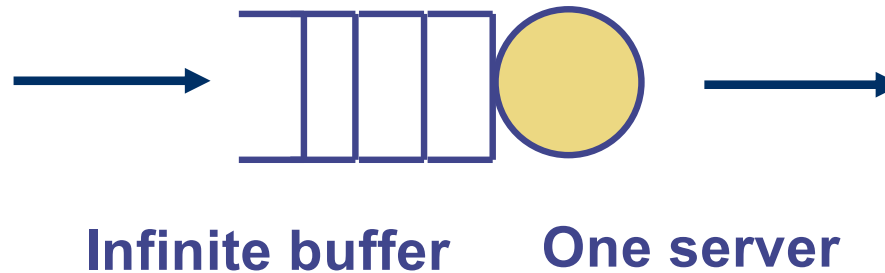
The call centre example on the last page is a  $M/M/m/(m+n)$  queue  
If  $n = \infty$ , we simply write  $M/M/m$



# M/M/1 queue

Exponential  
Inter-arrivals ( $\lambda$ )

Exponential  
Service time ( $\mu$ )



- Consider a call centre analogy
  - Calls are arriving according to Poisson distribution with rate  $\lambda$
  - The length of each call is exponentially distributed with parameter  $\mu$ 
    - Mean length of a call is  $1/\mu$

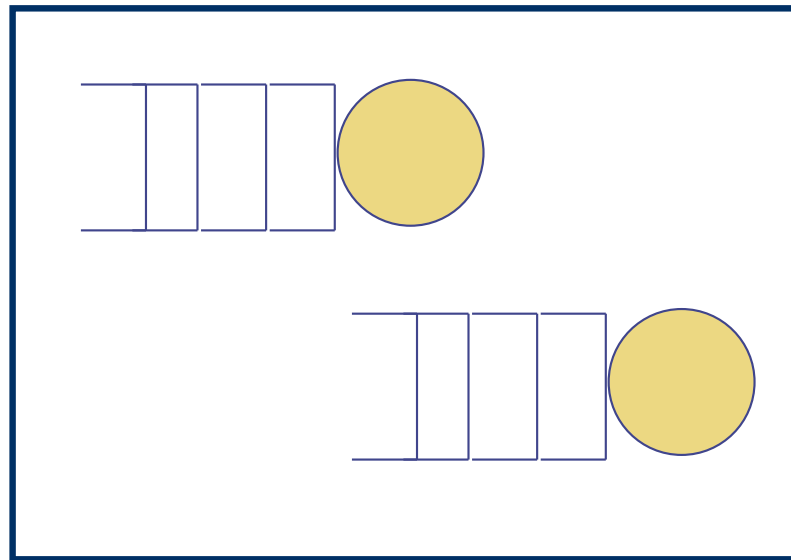
Arrivals  
→

**Call centre with 1 operator**  
**If the operator is busy, the centre will put the call on hold.**  
**A customer will wait until his call is answered.**

- Queueing theory will be able to answer these questions:
  - What are the mean waiting time, mean response time for a call?

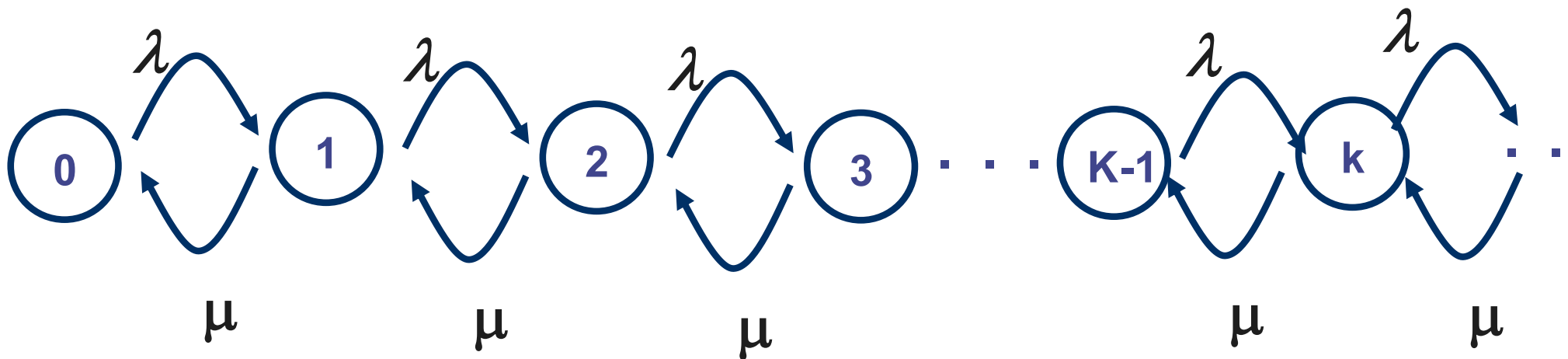
# Little's Law

- Applicable to any “box” that contains some queues or servers
- Mean number of jobs in the “box” =  
Mean response time x Throughput
- We will use Little's Law in this lecture to derive the mean response time
  - We first compute the mean number of jobs in the “box” and throughput



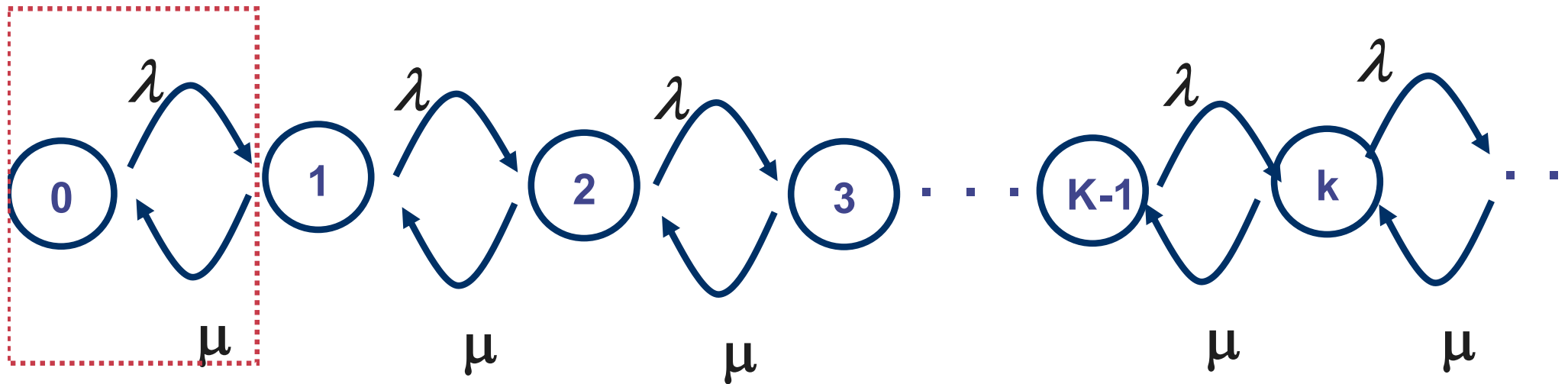
# M/M/1: State and transition diagram

- We will solve for the steady state response
- Define the states of the queue
  - State 0 = There is zero job in the system (= The server is idle)
  - State 1 = There is 1 job in the system (= 1 job at the server, no job queueing)
  - State 2 = There are 2 jobs in the system (= 1 job at the server, 1 job queueing)
  - State  $k$  = There are  $k$  jobs in the system (= 1 job at the server,  $k-1$  job queueing)
- The state transition diagram



## M/M/1 state balance:

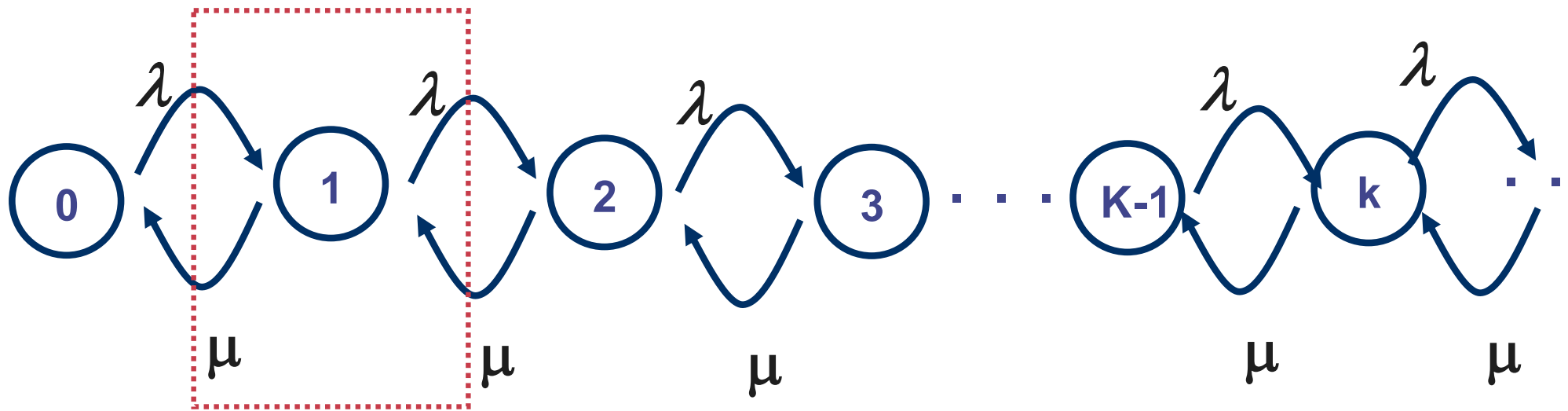
$P_k$  = Prob.  $k$  jobs in system



$$\lambda P_0 = \mu P_1$$

$$\Rightarrow P_1 = \frac{\lambda}{\mu} P_0$$

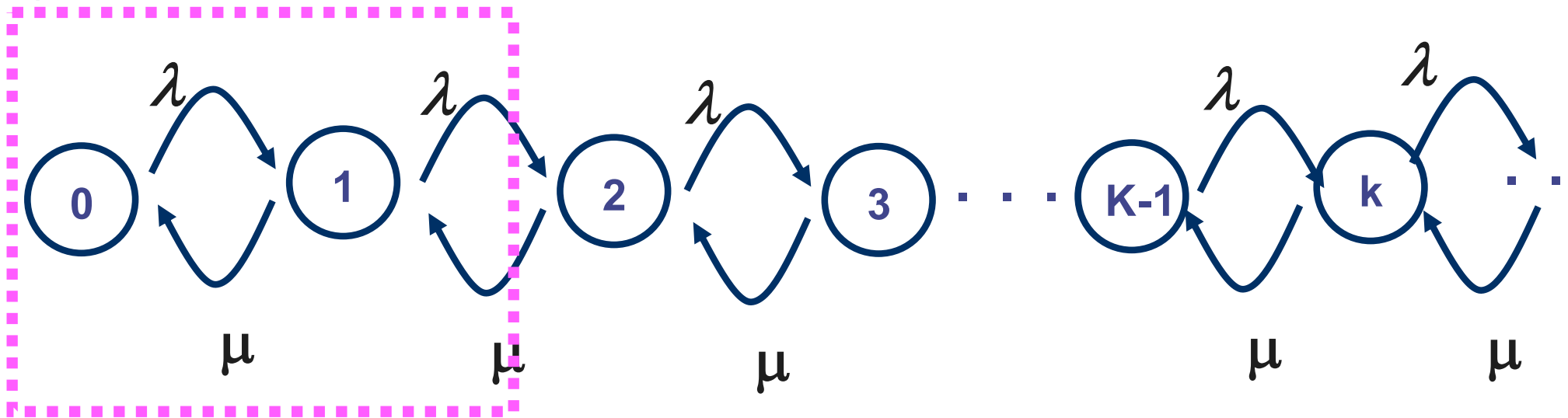
## M/M/1 state balance: Exercise 1



- Exercise: Write the state balance equation for State 1

$$\lambda P_0 + \mu P_2 = (\lambda + \mu) P_1$$

## M/M/1 state balance: Exercise 2

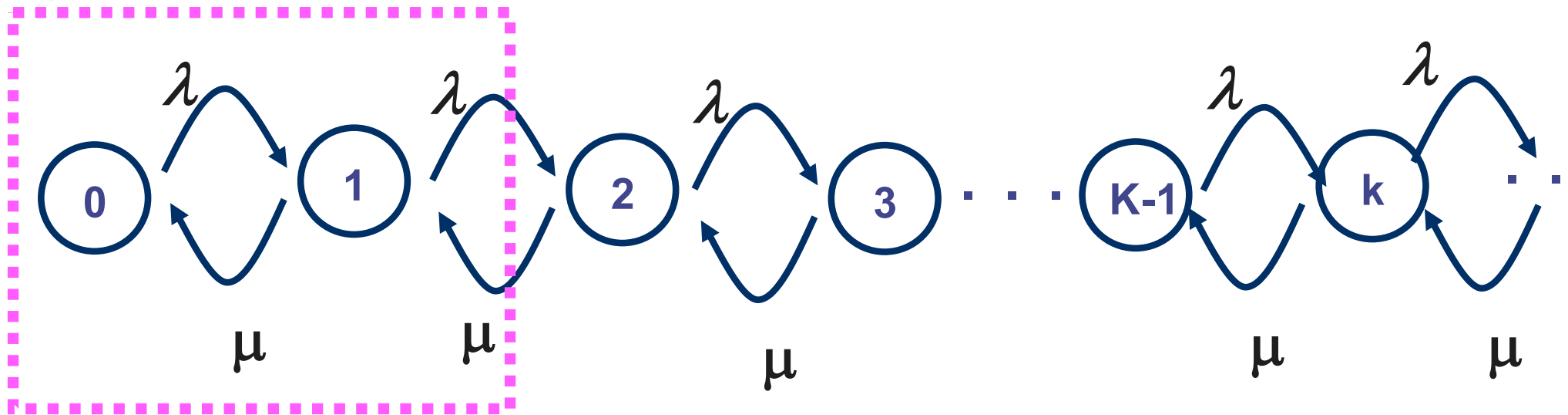


- Exercise: Write the state balance equation for magenta box, i.e.

Rate of transiting out of the magenta box  
= Rate of transiting into the magenta box

$$\lambda P_1 = \mu P_2$$

## Which state balance is easier to work with?



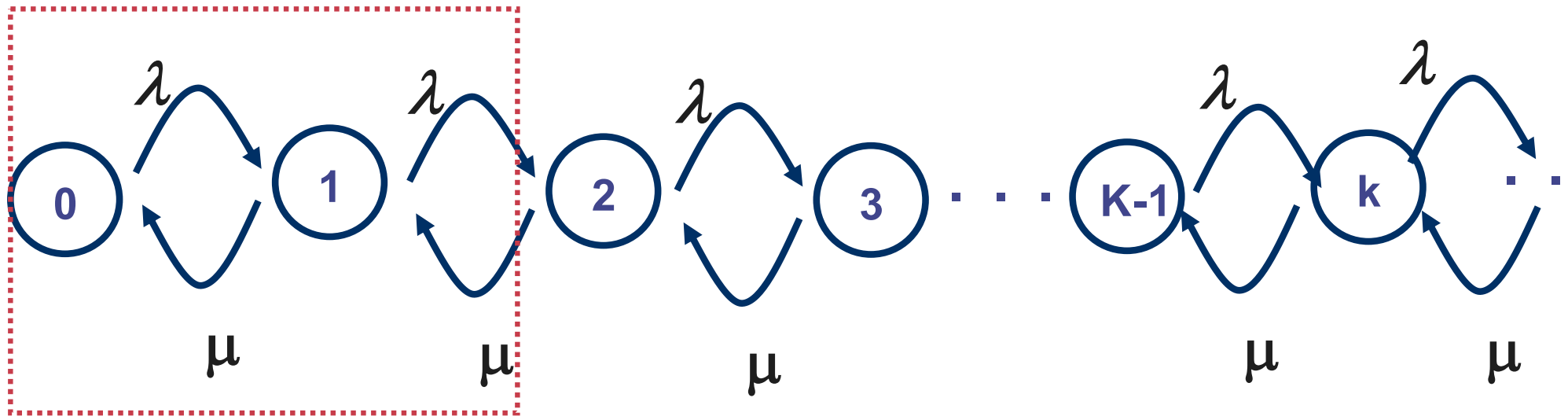
State balance for State 1

$$\lambda P_0 + \mu P_2 = (\lambda + \mu) P_1$$

State balance for State 1  
and State 2 combined

$$\lambda P_1 = \mu P_2$$

## M/M/1 state balance: Relating $P_2$ and $P_0$

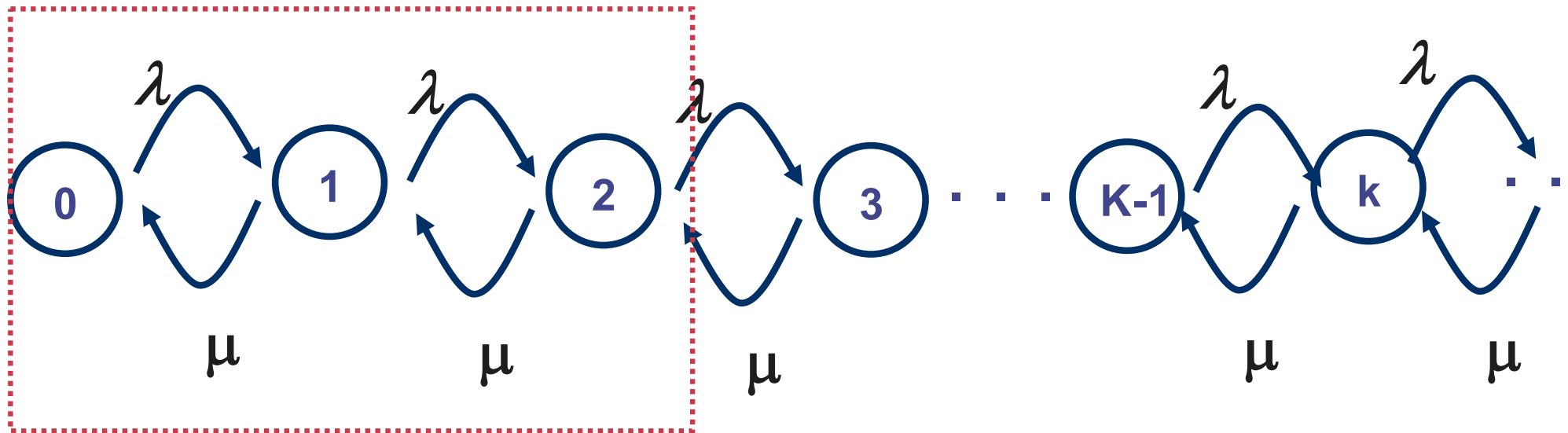


$$\lambda P_0 = \mu P_1 \quad \lambda P_1 = \mu P_2$$

$$\Rightarrow P_2 = \frac{\lambda}{\mu} P_1 \quad \Rightarrow P_2 = \left( \frac{\lambda}{\mu} \right)^2 P_0$$



## M/M/1 state balance: Relating $P_3$ and $P_0$



$$\lambda P_2 = \mu P_3$$

$$\Rightarrow P_3 = \frac{\lambda}{\mu} P_2 \Rightarrow P_3 = \left( \frac{\lambda}{\mu} \right)^3 P_0$$

## M/M/1 state balance: Relating $P_k$ and $P_0$

**In general** 
$$P_k = \left( \frac{\lambda}{\mu} \right)^k P_0$$

**Let** 
$$\rho = \frac{\lambda}{\mu}$$

**We have** 
$$P_k = \rho^k P_0$$

## Solving for $P_k$

With  $P_k = \rho^k P_0$  and

$$P_0 + P_1 + P_2 + P_3 + \dots = 1$$

$$\Rightarrow (1 + \rho + \rho^2 + \dots)P_0 = 1$$

$$\Rightarrow P_0 = 1 - \rho \text{ if } \rho < 1$$

$$\Rightarrow P_k = (1 - \rho)\rho^k$$

Since  $\rho = \frac{\lambda}{\mu}$ ,  $\rho < 1 \Rightarrow \lambda < \mu$

$\rho$  = utilisation  
= Prob server is busy  
=  $1 - P_0$   
= 1 - Prob server is idle

Arrival rate < service rate

## Exercise: Mean number of jobs

Recall that  $P_k$  = Prob.  $k$  jobs in system

and we have calculated that  $P_k = (1 - \rho)\rho^k$

Determine the mean number of jobs in the system.

Hint 1: Look at pre-lecture exercise 1.

You can use the following formula to help you.

For  $0 \leq x < 1$ ,

$$p = 0, x = \rho, q = 1$$

$$p + x(p + q) + x^2(p + 2q) + x^3(p + 3q) + \dots = \frac{p}{1 - x} + \frac{xq}{(1 - x)^2}$$

## Mean number of jobs

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$P_k$  = Prob.  $k$  jobs in system

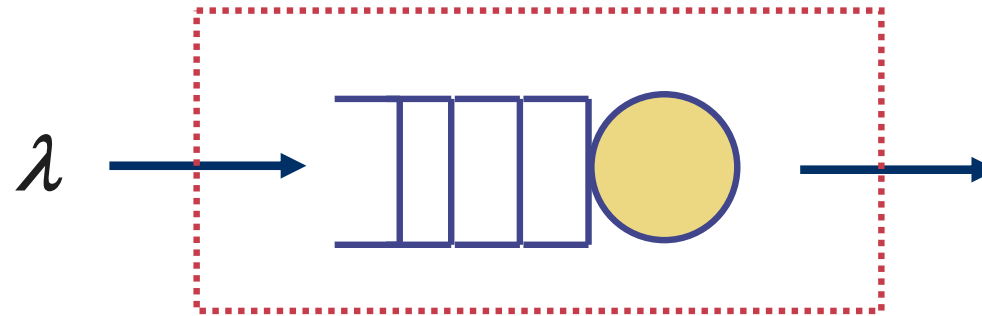
$$P_k = (1 - \rho)\rho^k$$

The mean number of jobs in the system =

$$\sum_{k=0}^{\infty} k P_k = \sum_{k=0}^{\infty} k (1 - \rho) \rho^k$$

$$= \frac{\rho}{1 - \rho}$$

## M/M/1: mean response time



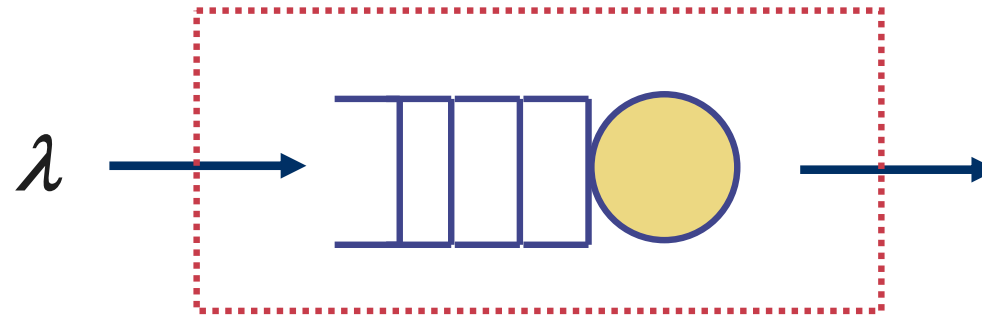
Little's law:

mean number of customers = throughput x response time

Throughput is  $\lambda$  (*why?*)

$$\text{Response time } T = \frac{\rho}{\lambda(1 - \rho)} = \frac{1}{\mu - \lambda}$$

## Exercise: M/M/1 mean waiting time



What is the mean waiting time at the queue?

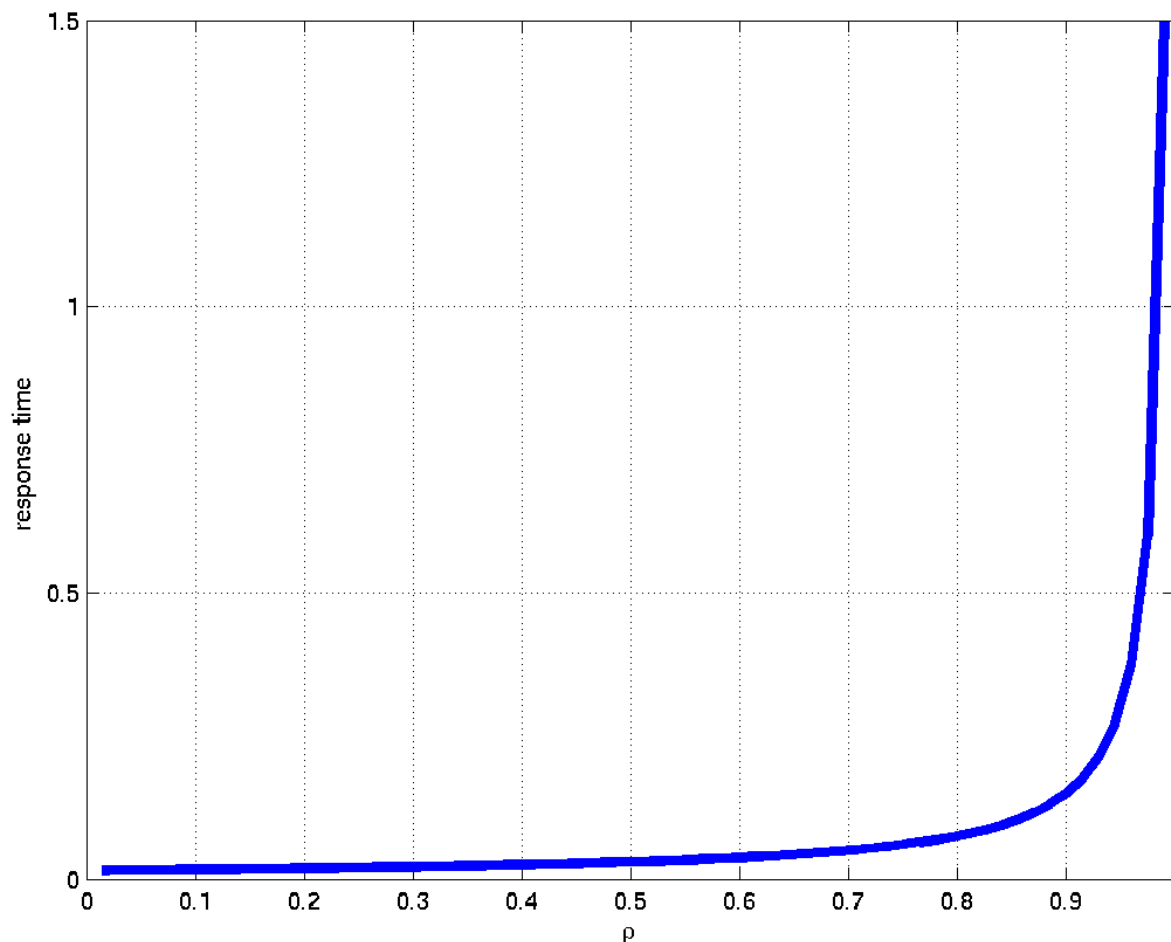
Mean waiting time = mean response time - mean service time

We know mean response time (from last slide)

Mean service time is  $= 1 / \mu$

Using the service time parameter ( $1/\mu = 15\text{ms}$ ) in the example, let us see how response time  $T$  varies with  $\lambda$

$$T = \frac{1}{\mu(1 - \rho)}$$



Observation:  
Response time increases sharply when  $\rho$  gets close to 1

Infinite queue assumption means  $\rho \rightarrow 1$ ,  $T \rightarrow \infty$



## Non-linear effect on response time

- The response time of an M/M/1 queue 
$$= \frac{1}{\mu - \lambda}$$
- Assuming the mean arrival rate is 10 requests/s
- We will calculate the effect of service rate on response time
- Complete the following table and see what you can conclude

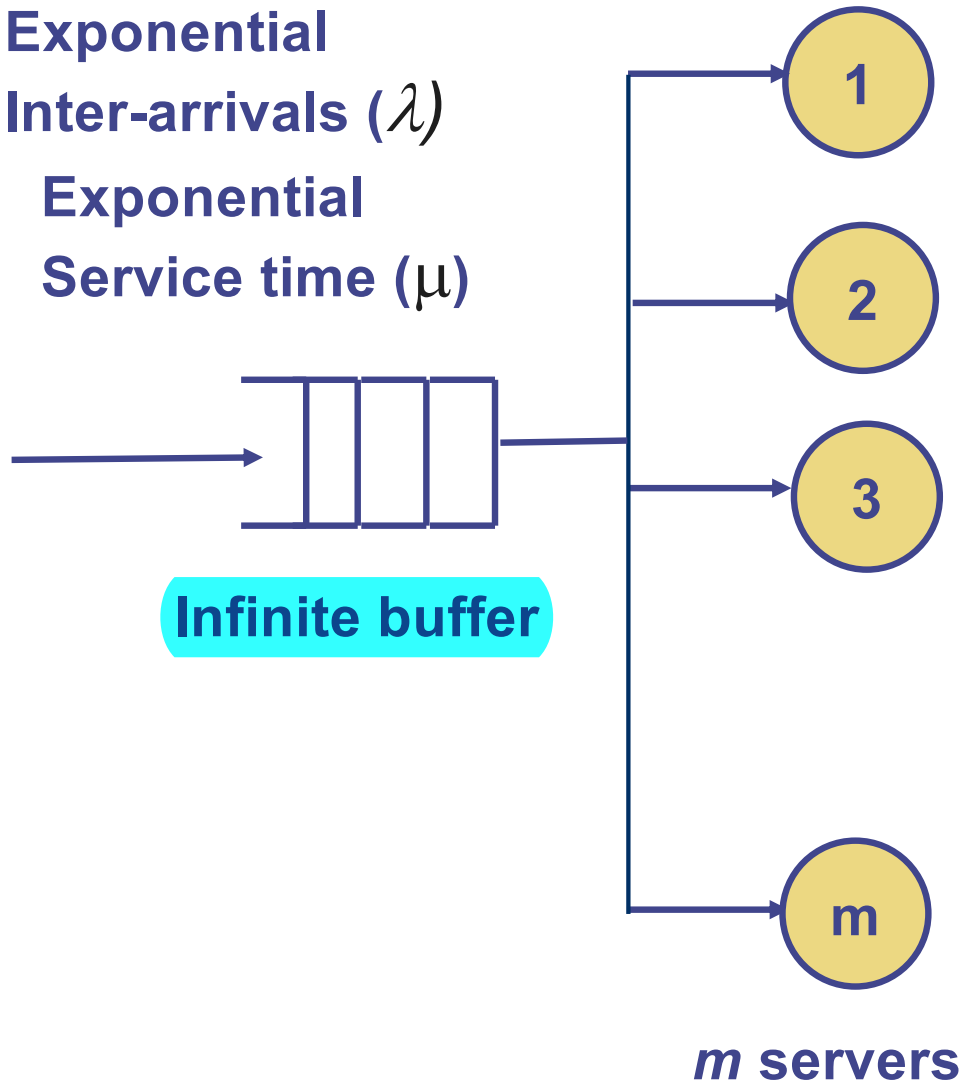
Service rate	Utilisation $\lambda/\mu$	Response time
11	$10/11 = 0.909$	1
22	$10/22 = 0.454$	0.08

- Doubling the service rate can sometimes reduce response time by a factor more than 2.

# Multi-server queues M/M/m

Exponential  
Inter-arrivals ( $\lambda$ )

Exponential  
Service time ( $\mu$ )



All arrivals go into one queue.

Customers can be served by any one of the  $m$  servers.

When a customer arrives

- If all servers are busy, it will join the queue
- Otherwise, it will be served by one of the available servers

# A call centre analogy of M/M/m queue

- Consider a call centre analogy
  - Calls are arriving according to Poisson distribution with rate  $\lambda$
  - The length of each call is exponentially distributed with parameter  $\mu$ 
    - Mean length of a call is  $1/\mu$

Arrivals

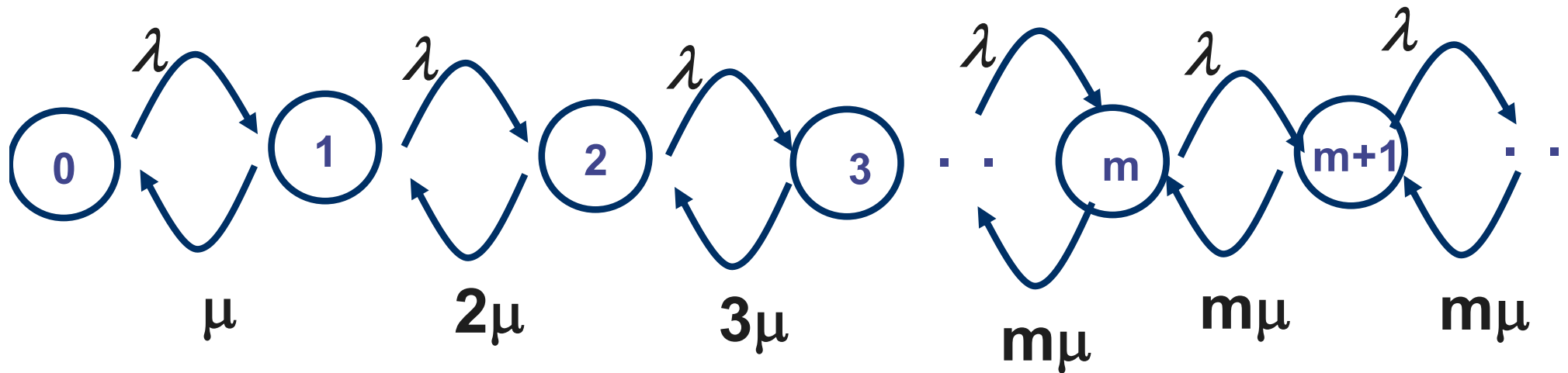


**Call centre with  $m$  operators**

**If all  $m$  operators are busy, the centre will put the call on hold.**

**A customer will wait until his call is answered.**

## State transition for M/M/m



- Following the same method, we have mean response time  $T$  is

$$T = \frac{C(\rho, m)}{m\mu(1 - \rho)} + \frac{1}{\mu}$$

where

$$\rho = \frac{\lambda}{m\mu}$$

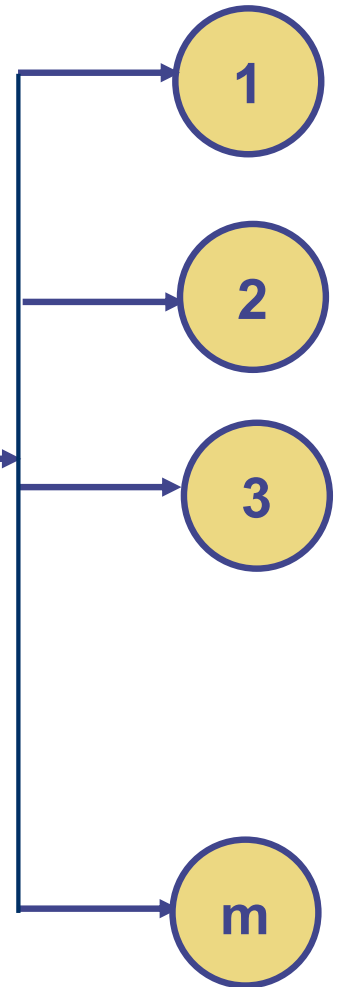
$$C(\rho, m) = \frac{\frac{(m\rho)^m}{m!}}{(1 - \rho) \sum_{k=0}^{m-1} \frac{(m\rho)^k}{k!} + \frac{(m\rho)^m}{m!}}$$

# Multi-server queues M/M/m/m with no waiting room

Exponential  
Inter-arrivals ( $\lambda$ )

Exponential  
Service time ( $\mu$ )

No waiting  
Room or  
No buffer



**$m$  servers**

An arrival can be served by any one of the  $m$  servers.

When a customer arrives

- If all servers are busy, it will *depart* from the system

- Otherwise, it will be served by one of the available servers

# A call centre analogy of M/M/m/m queue

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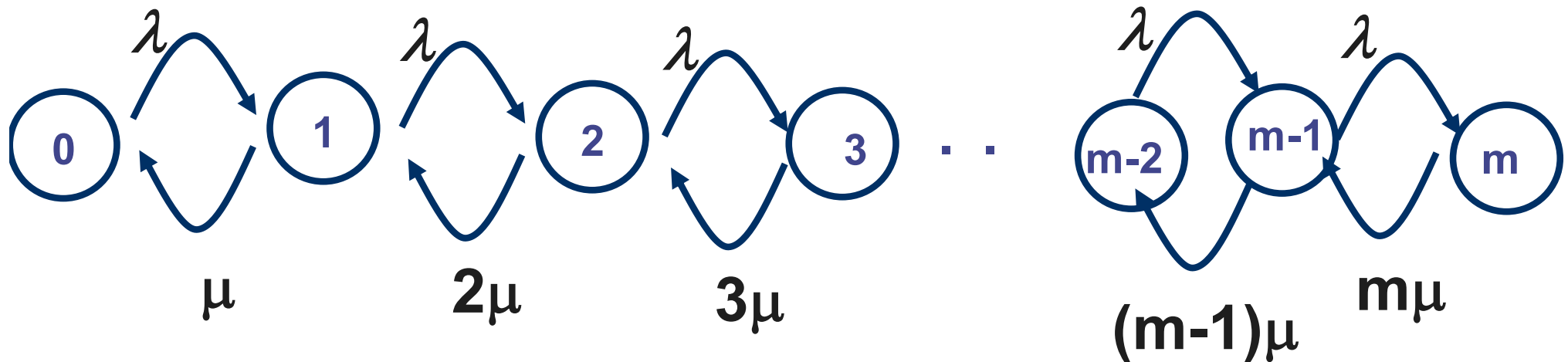
- Consider a call centre analogy
  - Calls are arriving according to Poisson distribution with rate  $\lambda$
  - The length of each call is exponentially distributed with parameter  $\mu$ 
    - Mean length of a call is  $1/\mu$

Arrivals



**Call centre with  $m$  operators**  
**If all  $m$  operators are busy, the call is dropped.**

## State transition for M/M/m/m



**Probability that an arrival is blocked**  
**= Probability that there are m customers in the system**

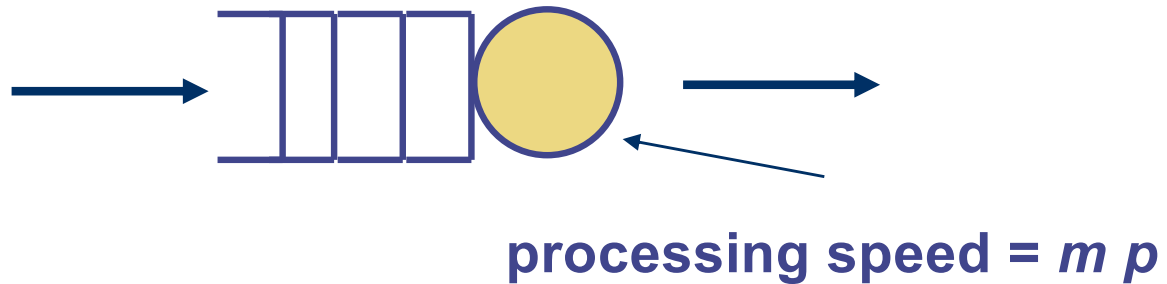
$$P_m = \frac{\frac{\rho^m}{m!}}{\sum_{k=0}^m \frac{\rho^k}{k!}} \quad \text{where} \quad \rho = \frac{\lambda}{\mu}$$

“Erlang B formula”

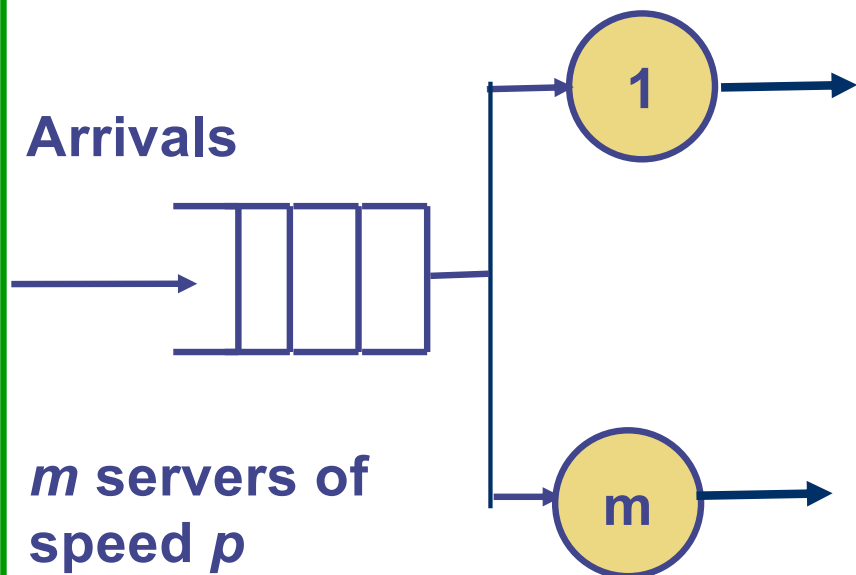


# What configuration has the best response time?

Configuration 1:

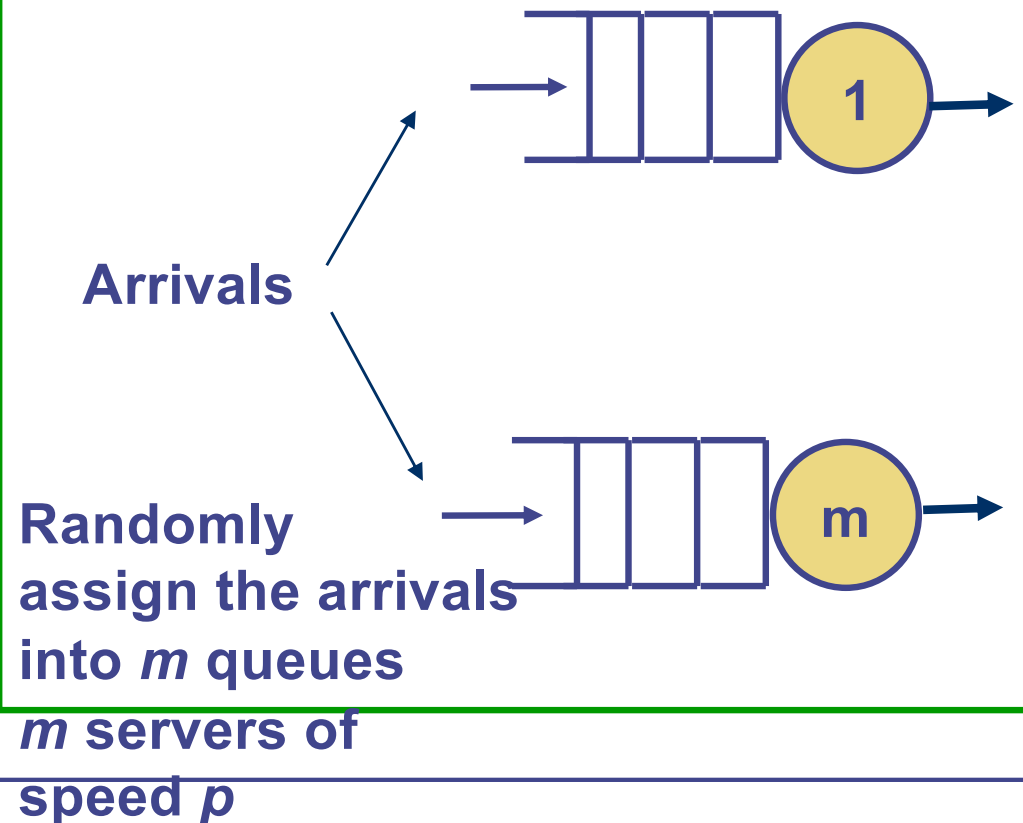


Configuration 2:



Try out the tutorial question!

Configuration 3:



# References

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- Recommended reading
  - Queues with Poisson arrival are discussed in
  - Bertsekas and Gallager, *Data Networks*, Sections 3.3 to 3.4.3
  - Note: I derived the formulas here using continuous Markov chain but Bertsekas and Gallager used discrete Markov chain
  - Mor Harchal-Balter. Chapters 13 and 14