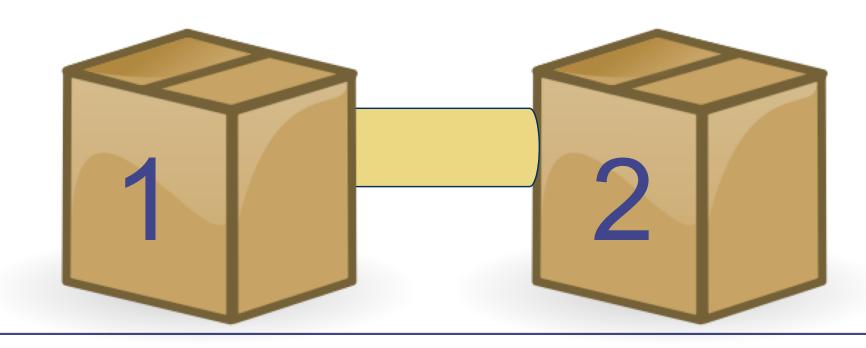
# COMP9334 Capacity Planning for Computer Systems and Networks

Week 2B: Queues with Poisson arrivals

# Pre-lecture exercise: Where is Felix? (Page 1)

- You have two boxes: Box 1 and Box 2, as well as a cat called Felix
- The two boxes are connected by a tunnel
- Felix likes to hide inside these boxes and travels between them using the tunnel.
- Felix is a very fast cat so the probability of finding him in the tunnel is zero
- You know Felix is in one of the boxes but you don't know which one



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# Pre-lecture exercise: Where is Felix? (Page 2)

#### Notation:

- Prob[A] = probability that event A occurs
- Prob[A | B] = probability that event A occurs given event B

#### You do know

- Felix is in one of the boxes at times 0 and 1
- Prob[ Felix is in Box 1 at time 0] = 0.3
- Prob[ Felix will be in Box 2 at time 1| Felix is in Box 1 at time 0] = 0.4
- Prob[ Felix will be in Box 1 at time 1| Felix is in Box 2 at time 0] = 0.2

#### Calculate

- Prob[ Felix is in Box 1 at time 1]
- Prob[ Felix is in Box 2 at time 1]

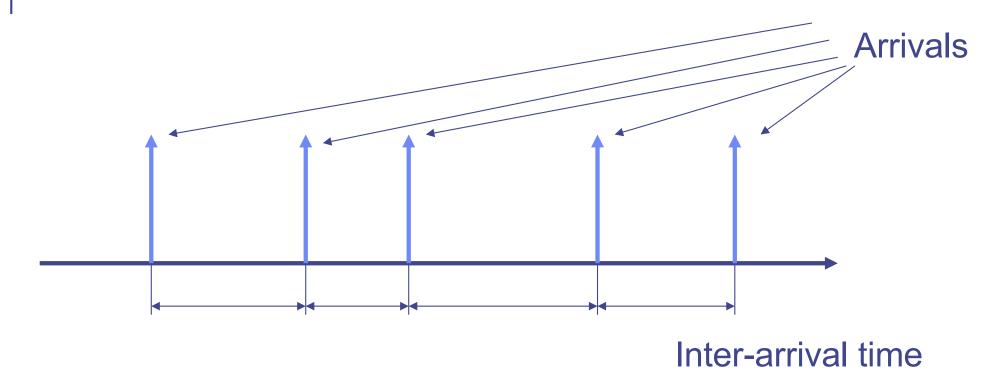


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# Performance analysis

- Modelling a computer system as a network of queues
- Operational analysis
  - Can be used to find performance bound
- What if you want more exact performance?
  - Need to consider
    - Probability distribution of the arrival process
    - Probability distribution of the service time

## Exponential inter-arrival with rate $\lambda$



We assume that successive arrivals are independent

Probability that inter-arrival time is between x and  $x + \delta x$  =  $\lambda \exp(-\lambda x) \delta x$ 

#### Poisson distribution

- The following are equivalent
  - The inter-arrival time is independent and exponentially distributed with parameter  $\boldsymbol{\lambda}$
  - The number of arrivals in an interval T is a Poisson distribution with parameter  $\lambda$

$$Pr[k \text{ arrivals in a time interval } T] = \frac{(\lambda T)^{\kappa} exp(-\lambda T)}{k!}$$

- Mean inter-arrival time = 1 / λ
- Mean number of arrivals in time interval  $T = \lambda T$
- Mean arrival rate = λ

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## Sample queueing problems

- Consider a call centre
  - Calls are arriving according to Poisson distribution with rate λ
  - The length of each call is exponentially distributed with parameter μ
    - Mean length of a call is 1/ μ (in, e.g. seconds)

#### Call centre:

#### **Arrivals**

*m* operators

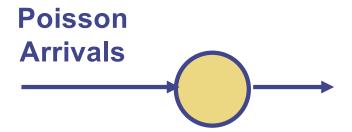
If all operators are busy, the centre can put at most *n* additional calls on hold. If a call arrives when all operators and holding slots are used, the call is rejected.

- Queueing theory will be able to answer these questions:
  - What is the probability that a call is rejected? (This lecture)
  - What is the mean waiting time for a call? (Next lecture)

## Let us start simple

- We will start by looking at a call centre with one operator and no holding slot
  - This may sound unrealistic but we want to show how we can solve a typical queueing network problem





# **Analysis strategy**

- The analysis will consider what happens over a small time interval  $\boldsymbol{\delta}$
- This is so that we can consider only two possibilities in each time interval

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#### Poisson distribution

- Consider a small time interval  $\delta$ 
  - This means  $\delta^n$  (for n >= 2) is negligible
- An interpretation of Poisson arrival:
  - Probability [ no arrival in  $\delta$  ] = 1  $\lambda \delta$
  - Probability [ 1 arrival in  $\delta$  ] =  $\lambda \delta$
  - Probability [ 2 or more arrivals in  $\delta$  ]  $\approx$  0
- This interpretation can be derived from:

$$Pr[k \text{ arrivals in a time interval } T] = \frac{(\lambda T)^{\kappa} exp(-\lambda T)}{k!}$$

#### Service time distribution

- Service time = the amount of processing time a job requires from the server
- We assume that the service time distribution is exponential with parameter  $\boldsymbol{\mu}$ 
  - The probability that the service time is between t and t +  $\delta t$  is:

$$\mu \exp(-\mu t) \delta t$$

- Here:  $\mu$  = service rate = 1 / mean service time
- Another interpretation of exponential service time:
  - Consider a small time interval δ
  - Probability [ a job will finish its service in next  $\delta$  seconds ] =  $\mu$   $\delta$
  - Probability [ a job will **not** finish its service in next  $\delta$  seconds ] =



# Call centre with 1 operator and no holding slots

- Let us see how we can solve the queuing problem for a very simple call centre with 1 operator and no holding slots
- What happens to a call that arrives when the operator is busy?
- What happens to a call that arrives when the operator is idle?
  - •
- We are interested to find the probability that an arriving call is rejected.



# Solution (1)

- There are two possibilities for the operator:
  - Busy or
  - Idle
- Let
  - State 0 = Operator is idle (i.e. #calls in the call centre = ?
  - State 1 = Operator is busy (i.e. #calls in the call centre = ?

 $P_0(t) = \text{Prob. } 0 \text{ call in the call centre at time } t$ 

 $P_1(t) = \text{Prob. 1 call in the call centre at time } t$ 

# Solution (2)

We try to express  $P_0(t + \Delta t)$  in terms of  $P_0(t)$  and  $P_1(t)$ 

- No call at call centre at t + ∆t can be caused by
  - No call at time t and no call arrives in [t, t +  $\Delta$ t], or
  - •

$$P_0(t + \Delta t) = +$$

Question: Why do we NOT have to consider the following possibility: No customer at time t & 1 customer arrives in [t, t +  $\Delta$ t] & the call finishes within [t, t +  $\Delta$ t].

# Solution (3)

Similarly, we can show that

$$P_1(t + \Delta t) = P_0(t)\lambda \Delta t + P_1(t)(1 - \mu \Delta t)$$

• If we let  $\Delta t \rightarrow 0$ , we have

$$\frac{dP_0(t)}{dt} = -P_0(t)\lambda + P_1(t)\mu$$

$$\frac{dP_1(t)}{dt} = P_0(t)\lambda - P_1(t)\mu$$

## Solution (4)

We can solve these equations to get

$$P_0(t) = \frac{\mu}{\lambda + \mu} - \frac{\mu}{\lambda + \mu} e^{-(\mu + \lambda)t}$$

$$P_1(t) = \frac{\lambda}{\lambda + \mu} + \frac{\mu}{\lambda + \mu} e^{-(\mu + \lambda)t}$$

This is too complicated, let us look at steady state solution

$$P_0 = P_0(\infty) = \frac{\mu}{\lambda + \mu}$$

$$P_1 = P_1(\infty) = \frac{\lambda}{\lambda + \mu}$$

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# Solution (5)

- From the steady state solution, we have
  - The probability that an arriving call is rejected
  - = The probability that the operator is busy

$$P_1 = \frac{\lambda}{\lambda + \mu}$$

- Let us check whether it makes sense
  - For a constant  $\mu$ , if the arrival rate rate  $\lambda$  increases, will the probability that the operator is busy go up or down?
  - Does the formula give the same prediction?

# An alternative interpretation

We have derived the following equation:

$$P_0(t + \Delta t) = P_0(t)(1 - \lambda \Delta t) + P_1(t)\mu \Delta t$$

Which can be rewritten as:

$$P_0(t + \Delta t) - P_0(t) = -P_0(t)\lambda \Delta t + P_1(t)\mu \Delta t$$

At steady state:

Change in Prob in State 0 = 0

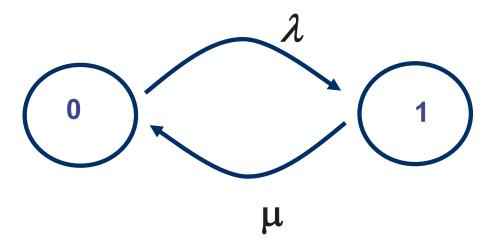
$$\Rightarrow 0 = -P_0 \lambda \Delta t + P_1 \mu \Delta t$$

Rate of leaving state 0

Rate of entering state 0

# Faster way to obtain steady state solution (1)

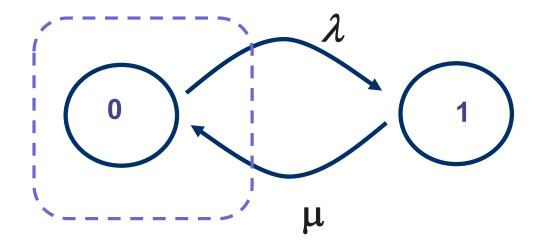
- Transition from State 0 to State 1
  - Caused by an arrival, the rate is λ
- Transition from State 1 to State 0
  - Caused by a completed service, the rate is  $\mu$
- State diagram representation
  - Each circle is a state
  - Label the arc between the states with transition rate



# Faster way to obtain steady state solution (2)

- Steady state means
  - rate of transition out of a state = Rate of transition into a state
- We have for state 0:

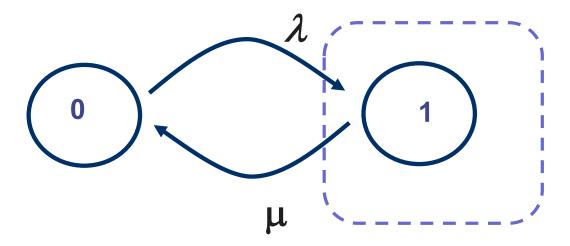
$$\lambda P_0 = \mu P_1$$



# Faster way to obtain steady state solution (3)

- We can do the same for State 1:
- Steady state means
  - Rate of transition into a state = rate of transition out of a state
- We have for state 1:

$$\lambda P_0 = \mu P_1$$



# Faster way to obtain steady state solution (4)

- We have one equation  $~\lambda P_0 = \mu P_1$
- We have 2 unknowns and we need one more equation.
- Since we must be either one of the two states:

$$P_0 + P_1 = 1$$

 Solving these two equations, we get the same steady state solution as before

$$P_0 = \frac{\mu}{\lambda + \mu} \qquad P_1 = \frac{\lambda}{\lambda + \mu}$$

# Summary

- Solving a queueing problem is not simple
- It is harder to find how a queue evolves with time
- It is simpler to find how a queue behaves at steady state
  - Procedure:
    - Draw a diagram with the states
    - Add arcs between states with transition rates
    - Derive flow balance equation for each state, i.e.
      - Rate of entering a state = Rate of leaving a state
    - Solve the equation for steady state probability

# Don't forget the probabilistic interpretation

Change in probability in State 0

$$P_0(t + \Delta t) - P_0(t) = -P_0(t)\lambda \Delta t + P_1(t)\mu \Delta t$$

$$\Rightarrow 0 = -P_0 \lambda \Delta t + P_1 \mu \Delta t$$

Rate of leaving state 0

Rate of entering state 0

$$\Rightarrow 0 = -P_0 \lambda \Delta t + P_1 \mu \Delta t$$

Prob[Leaving State 0 | State 0]

Prob[Entering State 0 | State 1]

# A call centre with 1 operator and 1 holding slot

 We want to determine the probability that an arriving call will be rejected



# Analysing the queueing problem

- The system can be in one of the following three states
  - State 0 = 0 call in the system (= the operator is idle)
  - State 1 = 1 call in the system (= Operator busy. Holding slot empty.)
  - State 2 = 2 calls in the system (= Operator busy. Holding slot occupied.)
- Define the probability that a certain state occurs

$$P_0 = \text{Probability in State 0}$$

$$P_1 = \text{Probability in State 1}$$

$$P_2 = Probability in State 2$$

# The transition probabilities

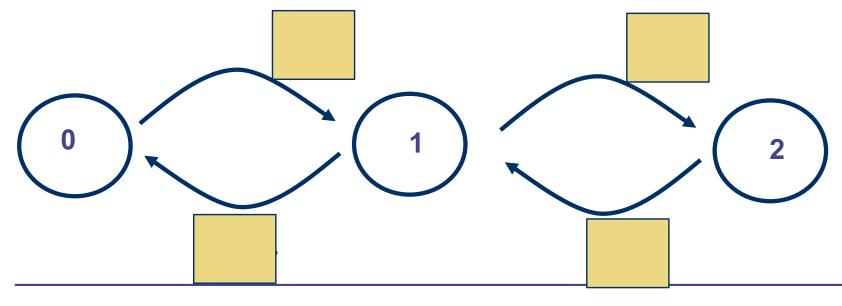
- Consider a small time interval  $\delta$ 
  - Given the system is in State 1
    - What is the probability that it will move to State 0?
    - What is the probability that it will move to State 2?
- Transiting from State 1 → State 0
  - This can only occur when
  - Conditional probability for this to occur = \_\_\_\_\_\_
- Transiting from State 1 → State 2
  - This can only occur when
  - Conditional probability for this to occur = \_\_\_\_\_\_
- Prob [State 1 → State 0 | State 1] = \_\_\_\_\_\_

## Exercise: The transition probabilities

- Can you work out the following transition probabilities
  - Prob [State 0 → State 1 | State 0] =
  - Prob [State 0 → State 2 | State 0] =
  - Prob [State 2 → State 0 | State 2] =
  - Prob [State 2 → State 1 | State 2] =

# The state transition diagram

- Given the following transition probabilities (over a small time interval  $\delta$ )
  - Prob [State 0 → State 1 | State 0] =
  - Prob [State 0 → State 2 | State 0] =
  - Prob [State 1 → State 0 | State 1] =
  - Prob [State 1 → State 2 | State 1] = [
  - Prob [State 2 → State 0 | State 2] =
  - Prob [State 2 → State 1 | State 2] =
- We draw the following state transition diagram
  - Note 1: We label the arc with transition rate = transition probability /  $\delta$
  - Note 2: Arcs with zero rate are not drawn

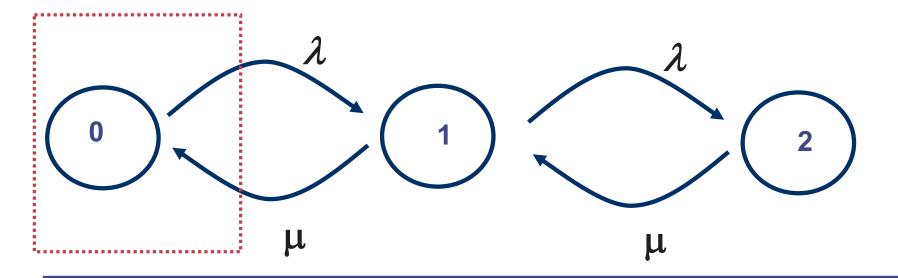


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# Setting up the balance equations (1)

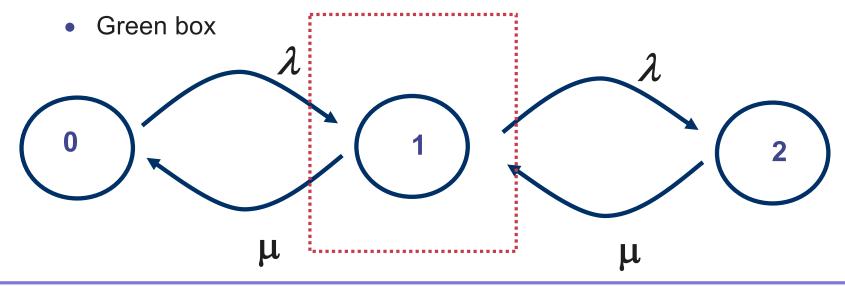
- For steady state, we have
  - Prob of transiting into a "box" = Prob of transiting out of a "box"
  - Rate of transiting into a "box" = Rate of transiting out of a "box"
- Note a "box" can include one or more state
- The "box" is the dotted square shown below

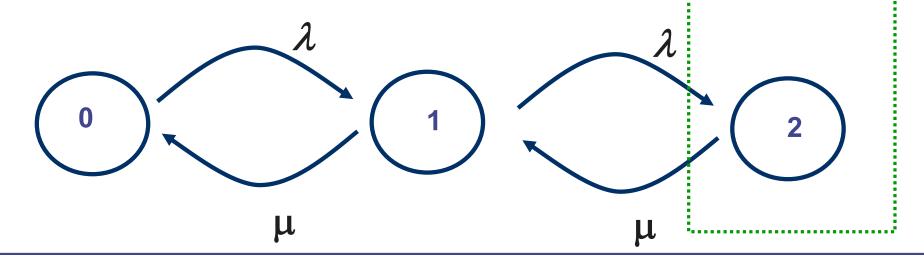
Prob out of "box" = 
$$P_0\lambda\delta$$
  
Prob into "box" =  $P_1\mu\delta$   $\Rightarrow \lambda P_0 = \mu P_1$ 



# Exercise: Setting up the balance equations (2)

- Set up the balance equations for the
  - Red box





# The balance equations

There are three balance equations

- Note that these three equations are not linearly independent
  - First equation + Third equation = Second equation
- There are 3 unknowns (P<sub>0</sub>, P<sub>1</sub>, P<sub>2</sub>) but we have only 2 equations
- We need 1 more equation. What is it?

# Solving for the steady state probabilities

- An addition equation: Sum( Probabilities ) = 1
- Solve the following equations for the steady state probabilities P<sub>0</sub>, P<sub>1</sub>, P<sub>2</sub>:

$$\lambda P_0 = \mu P_1$$

$$\mu P_2 = \lambda P_1$$

By solving these 3 equations, we have

# Steady state probabilities

 By solving the equations on the previous slide, we have the steady state probabilities are:

$$P_0 = \frac{1}{1 + \frac{\lambda}{\mu} + \left(\frac{\lambda}{\mu}\right)^2}$$

$$P_1 = \frac{\frac{\lambda}{\mu}}{1 + \frac{\lambda}{\mu} + \left(\frac{\lambda}{\mu}\right)^2}$$

$$P_2 = \frac{\left(\frac{\lambda}{\mu}\right)^2}{1 + \frac{\lambda}{\mu} + \left(\frac{\lambda}{\mu}\right)^2}$$

- If we know the values of λ
   and μ, we can find the
   numerical values of
   these probabilities
- Do the expressions make sense?

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## Summary and References

#### Summary

- Poisson queues with 1 server + (0 or 1) holding slot
- How to solve the steady state solution

#### 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