

Advanced Computer Networks



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Birth and Death Markov Process



**DSSRG: Decentralized
Smart Systems Research
Group**

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The *Birth* and *Death* Markov processes

- Definition:
- A birth/death Markov process is a Markov process where:

$$p_{i,i+1} = b_i \quad (\text{birth})$$

$$p_{i,i-1} = d_i \quad (\text{death})$$

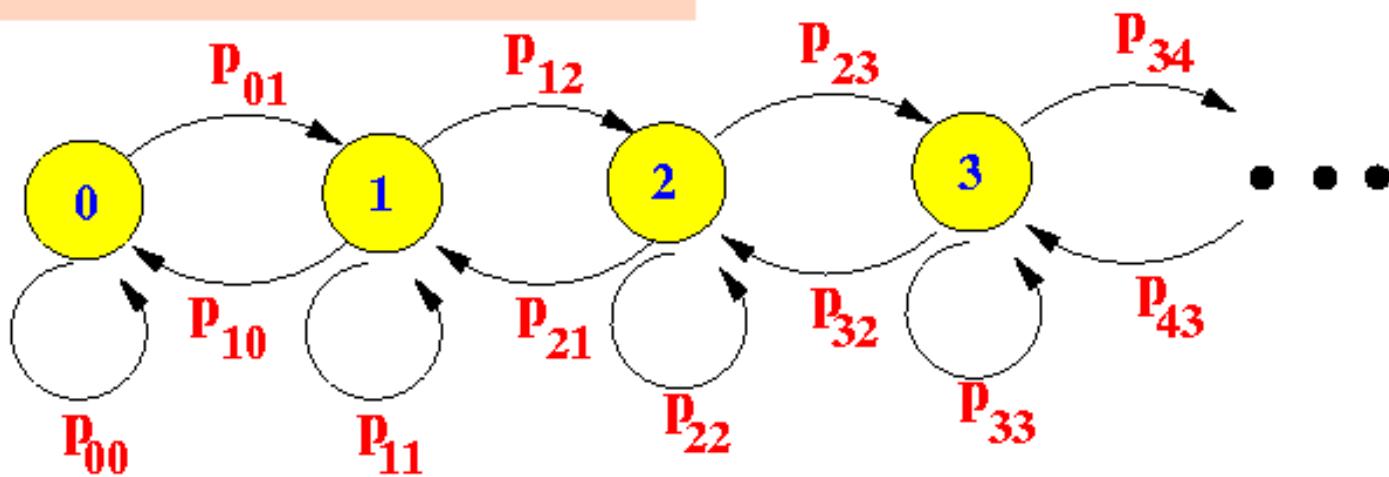
$$p_{i,i} = 1 - b_i - d_i$$

$$p_{i,k} = 0 \quad \text{for } k \leq i-2 \text{ or } k \geq i+2$$



In other words

- A process can either go to $i-1$ or $i+1$
- All other transition probabilities are zero
- The **transition diagram** of a birth/death process looks like this:
- state $k = \text{population size is } k$



The *possible* events in a birth/death Markov process are:

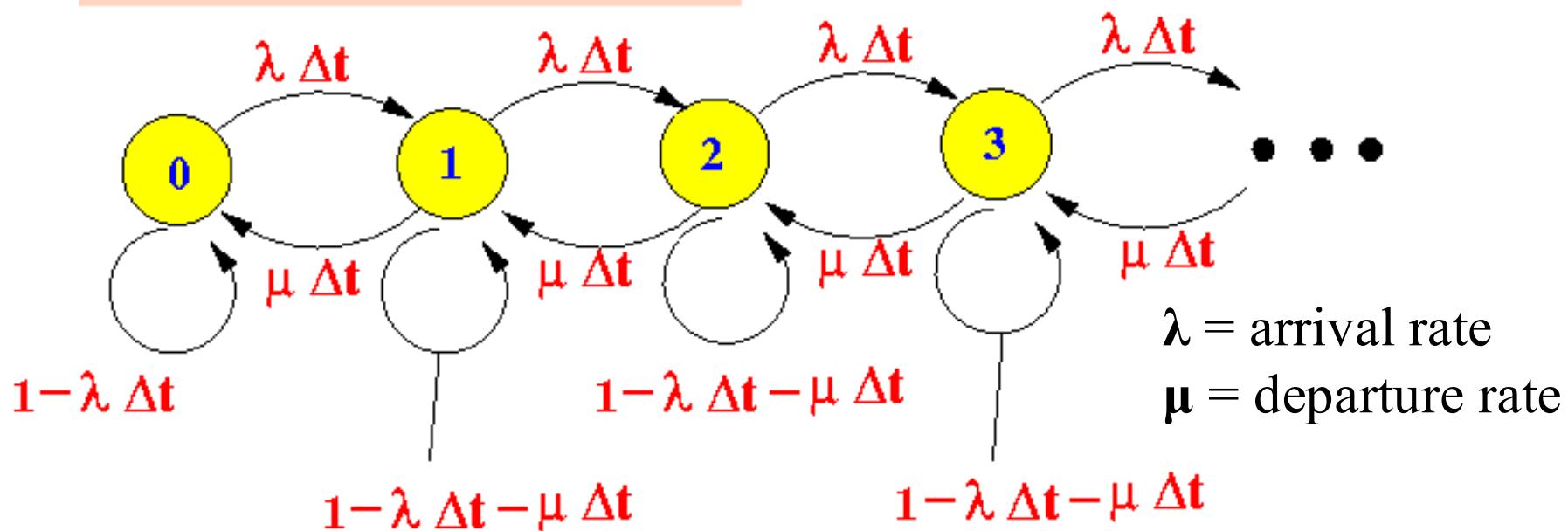
- Exactly one birth
- Exactly one death
- The event that there is **no birth and no death** is the case that **no event occurs**



The *Poisson* Birth/Death process

- $P[\text{an arrival occurs in time interval } \Delta t] = \lambda \times \Delta t$
 - $P[\text{a departure occurs in time interval } \Delta t] = \mu \times \Delta t$

state k = population size is k



Resulting state equations:

$$\begin{aligned} p_0 &= p_0 (1 - \lambda \Delta t) + p_1 \mu \Delta t \\ p_1 &= p_0 \lambda \Delta t + p_1 (1 - \lambda \Delta t - \mu \Delta t) + p_2 \mu \Delta t \\ p_2 &= p_1 \lambda \Delta t + p_2 (1 - \lambda \Delta t - \mu \Delta t) + p_3 \mu \Delta t \\ p_3 &= p_2 \lambda \Delta t + p_3 (1 - \lambda \Delta t - \mu \Delta t) + p_4 \mu \Delta t \end{aligned}$$

...

Or:

$$\begin{aligned} p_0 \lambda \Delta t &= p_1 \mu \Delta t \\ p_1 (\lambda \Delta t + \mu \Delta t) &= p_0 \lambda \Delta t + p_2 \mu \Delta t \\ p_2 (\lambda \Delta t + \mu \Delta t) &= p_1 \lambda \Delta t + p_3 \mu \Delta t \\ p_3 (\lambda \Delta t + \mu \Delta t) &= p_2 \lambda \Delta t + p_4 \mu \Delta t \end{aligned}$$

...



Or:

$$p_0 \lambda = p_1 \mu$$

$$p_1(\lambda + \mu) = p_0 \lambda + p_2 \mu$$

$$p_2(\lambda + \mu) = p_1 \lambda + p_3 \mu$$

$$p_3(\lambda + \mu) = p_2 \lambda + p_4 \mu$$

...

$$p_0 \lambda = p_1 \mu$$

$$p_1 \lambda + p_1 \mu = p_0 \lambda + p_2 \mu$$

$$p_2 \lambda + p_2 \mu = p_1 \lambda + p_3 \mu$$

$$p_3 \lambda + p_3 \mu = p_2 \lambda + p_4 \mu$$

...

Or:



Or:

$$P_0 \lambda = P_1 \mu$$

$$P_1 \lambda = P_2 \mu$$

$$P_2 \lambda + P_2 \mu = P_1 \lambda + P_3 \mu$$

$$P_3 \lambda + P_3 \mu = P_2 \lambda + P_4 \mu$$

....

....

$$P_0 \lambda = P_1 \mu$$

$$P_1 \lambda = P_2 \mu$$

$$P_2 \lambda = P_3 \mu$$

$$P_3 \lambda + P_3 \mu = P_2 \lambda + P_4 \mu$$

....

Or:



Final set of equations ...

•

$$p_0 \lambda = p_1 \mu$$

$$p_1 \lambda = p_2 \mu$$

$$p_2 \lambda = p_3 \mu$$

$$p_3 \lambda = p_4 \mu$$

...

Or:

$$p_1 = \lambda/\mu \times p_0$$

$$p_2 = \lambda/\mu \times p_1$$

$$p_3 = \lambda/\mu \times p_2$$

$$p_4 = \lambda/\mu \times p_3$$

...



Or:

$$p_1 = \lambda/\mu \times p_0 \quad \dots \quad (1)$$

$$p_2 = (\lambda/\mu)^2 \times p_0 \quad \dots \quad (2)$$

$$p_3 = (\lambda/\mu)^3 \times p_0 \quad \dots \quad (3)$$

$$p_4 = (\lambda/\mu)^4 \times p_0 \quad \dots \quad (4)$$

...

- We can **express** every p_i , $i = 1, 2, 3, \dots$ in terms on p_0
- **Unfortunately**, we **do not** (yet) know the value of p_0 ...



- We need **one more equation** to solve this system, which is:

$$p_0 + p_1 + p_2 + p_3 + \dots = 1$$

- Substituting p_i in

$$p_0 + (\lambda/\mu)^1 \times p_0 + (\lambda/\mu)^2 \times p_0 + (\lambda/\mu)^3 \times p_0 + \dots = 1$$



$$p_0 \times (1 + (\lambda/\mu)^1 + (\lambda/\mu)^2 + (\lambda/\mu)^3 + \dots) = 1$$

$$\begin{aligned} S &= 1 + x^1 + x^2 + x^3 + x^4 + \dots \\ xS &= \quad x^1 + x^2 + x^3 + x^4 + \dots \\ \hline (1-x)S &= 1 \end{aligned}$$

Therefore:

$$1 + x^1 + x^2 + x^3 + x^4 + \dots = \frac{1}{1 - x}$$



$$p_0 \times \frac{1}{1-(\lambda/\mu)} = 1$$

$$p_0 = 1 - (\lambda/\mu) = 1 - \rho$$

$$\rho = \lambda/\mu$$



- Steady state probability distribution of a Poisson birth/death process with arrival rate λ and departure rate μ :

- $p_0 = (\lambda/\mu)^0 \times (1-(\lambda/\mu))$

- $p_1 = (\lambda/\mu)^1 \times (1-(\lambda/\mu))$

- $p_2 = (\lambda/\mu)^2 \times (1-(\lambda/\mu))$

- $p_3 = (\lambda/\mu)^3 \times (1-(\lambda/\mu))$

- • •



Or:

$$p_0 = \rho^0 \times (1 - \rho)$$

$$p_1 = \rho^1 \times (1 - \rho)$$

$$p_2 = \rho^2 \times (1 - \rho)$$

$$p_3 = \rho^3 \times (1 - \rho)$$

.....



$P[k \text{ customers in system}] = p_k = \rho^k \times (1 - \rho)$



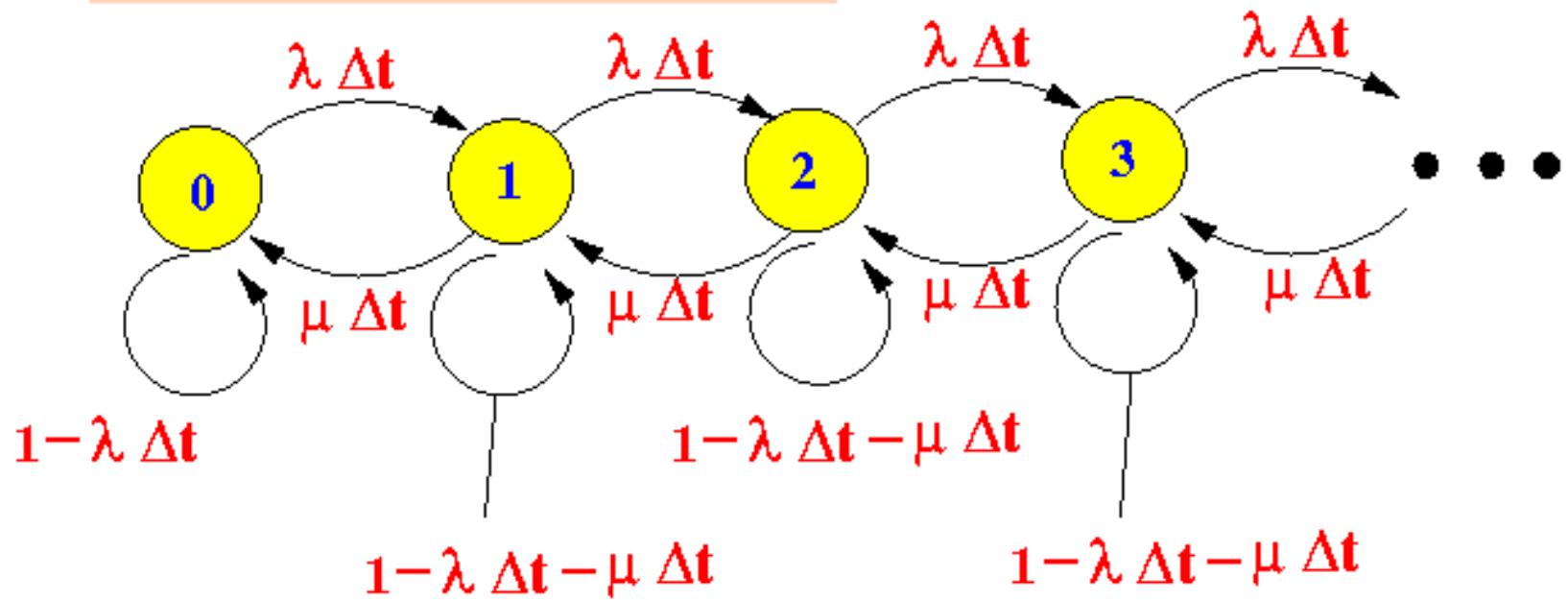
Solving Markov chain using *rate transition* diagram

- A popular method for finding the equilibrium (steady state) probability distribution of a Markov chain is using rate transition diagrams
- Rate transition diagram:
 - A rate transition diagram is obtained by removing the transitions that goes from a state into the same state from a state transition diagram



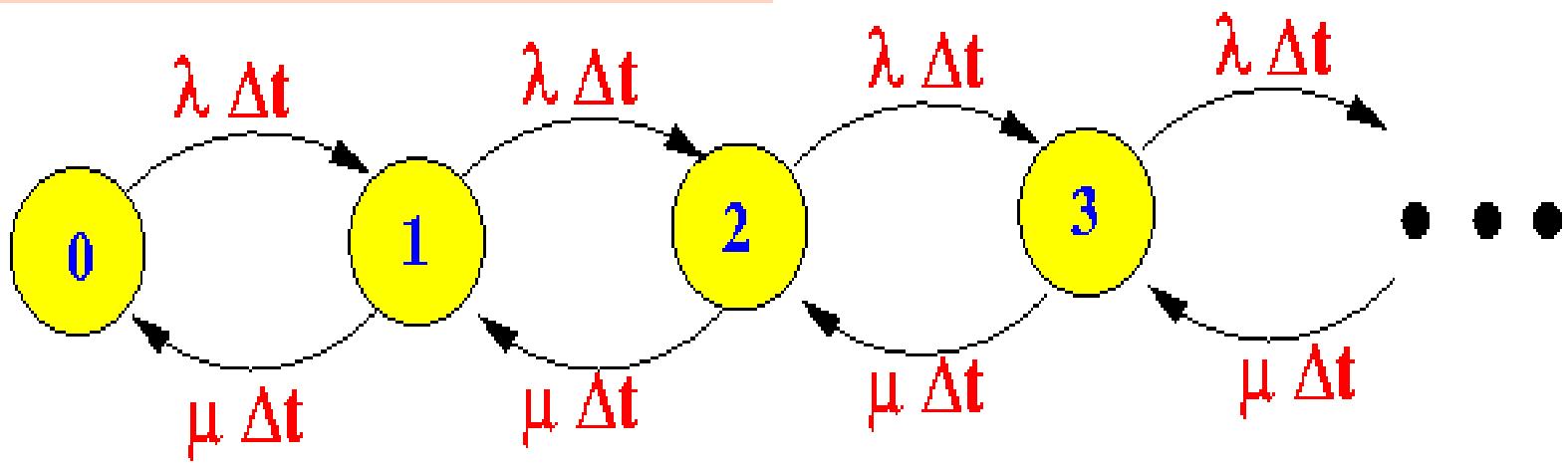
State transition diagram:

state k = population size is k



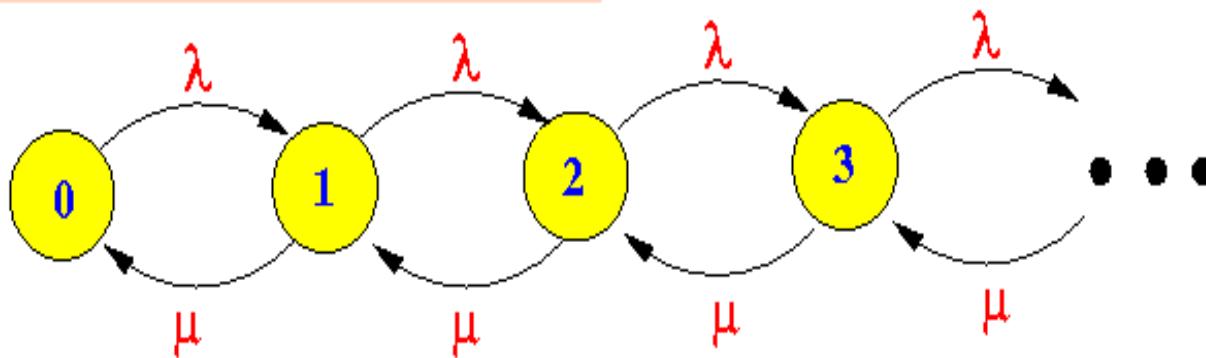
State diagram *without* loops to itself:

state k = population size is k



After *normalization* by dividing by Δt :

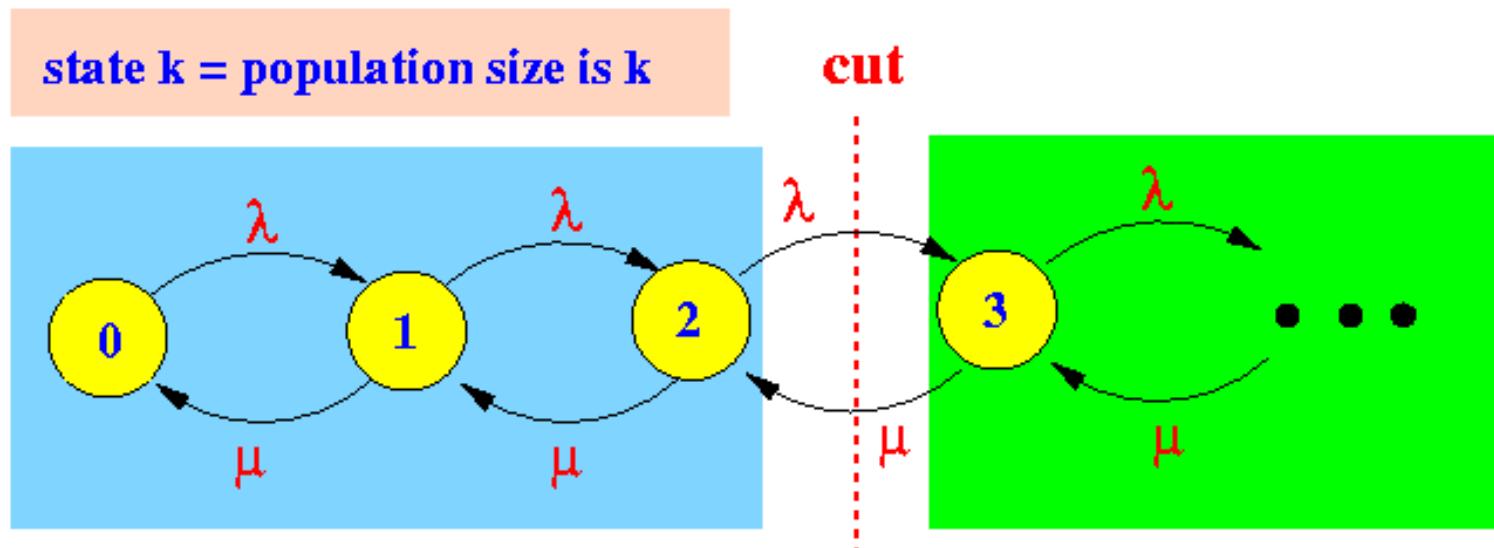
state k = population size is k



- λ = **arrival rate** of clients
- μ = **departure rate** of clients
- The **weights** on the **arcs** in the **diagram** are the **rates** of **arrival** and **departure**
- Hence the **name**: **rate transition diagram**

Setting up equilibrium equations using a *rate* transition diagram

- Find a **cut** in the Markov chain that divide the Markov chain into **2 disjoint pieces**
- In the **equilibrium state**, the **number of transitions** from one side the **cut** to the **other side** must be **equal to the reverse direction**
-



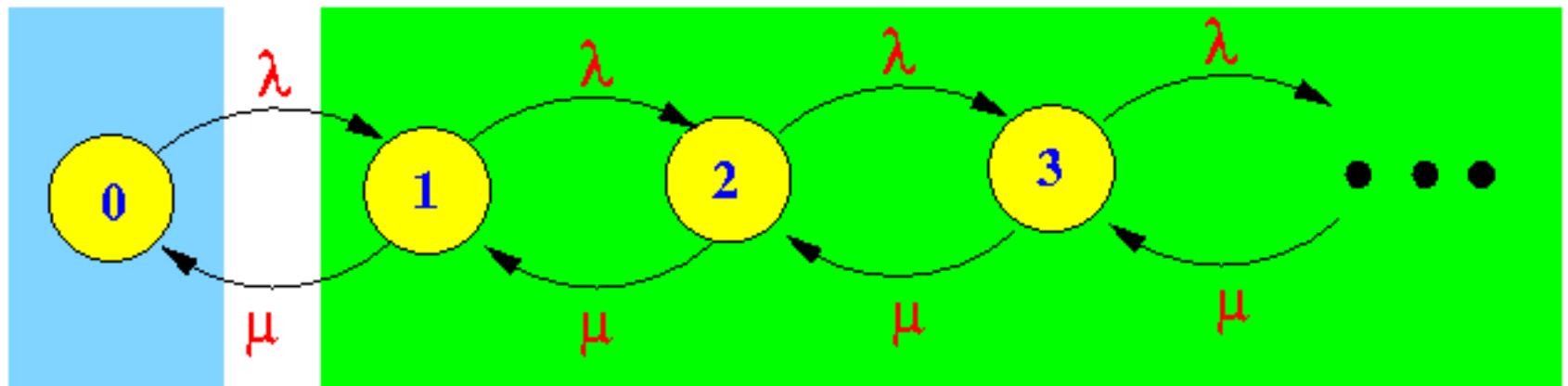
Equilibrium equation:

- Flow from left to right: $p_2 \times \lambda$
- Flow from left to right: $p_3 \times \mu$
- Equilibrium: $p_2 \times \lambda = p_3 \times \mu$
-



Complete example: equilibrium equations for the M/M/1 queuing system

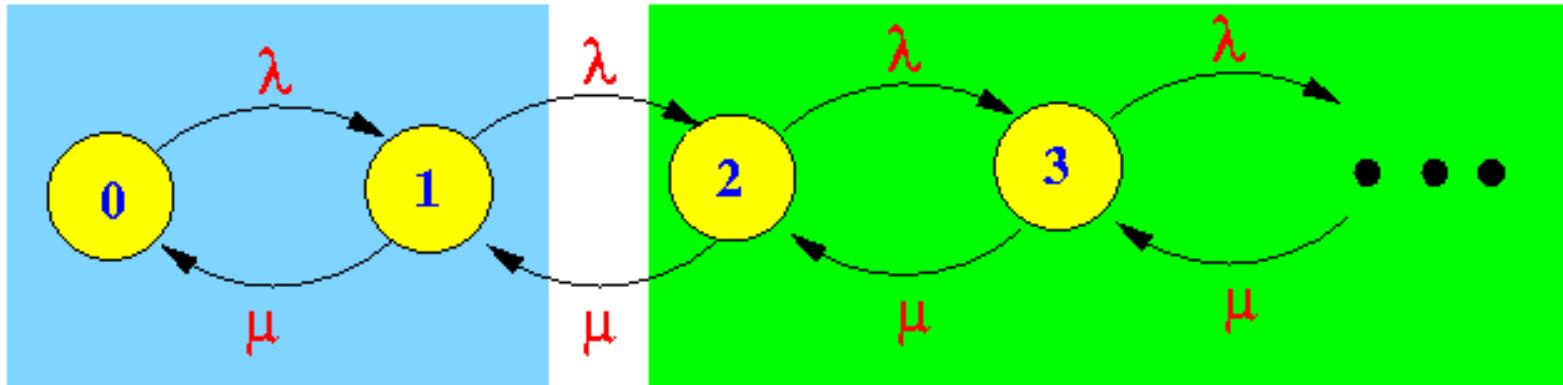
Cut 1:



Equilibrium equation:

Flow from left to right: $\lambda \times p_0$ Flow from right to left: $\mu \times p_1$ Equilibrium: $\lambda \times p_0 = \mu \times p_1$

- Cut 2:



Equilibrium equation:

Flow from left to right: $\lambda \times p_1$

Flow from right to left: $\mu \times p_2$

Equilibrium: $\lambda \times p_1 = \mu \times p_2$

- Resulting set of equation for the equilibrium state:
-
- $p_0 \lambda = p_1 \mu$
- $p_1 \lambda = p_2 \mu$
- $p_2 \lambda = p_3 \mu$
- $p_3 \lambda = p_4 \mu$
- This is the **same** set of equation as before...



Introduction to queueing theory

- **Queue**
- A **queue** is a **waiting line**...
- The **behaviour** of a **queue** is **characterized** by following parameters:
 - the **arrival process** (commonly used: **Poisson process**)
 - the **service (departure) process** (commonly used: **Poisson process**)
 - the **number of servers** in the **system**
 - The **queueing discipline** (often **FIFO**)
 - The **capacity** of the **queue** (buffer space)
 - The **size of the client population** (commonly used value: **infinite**)

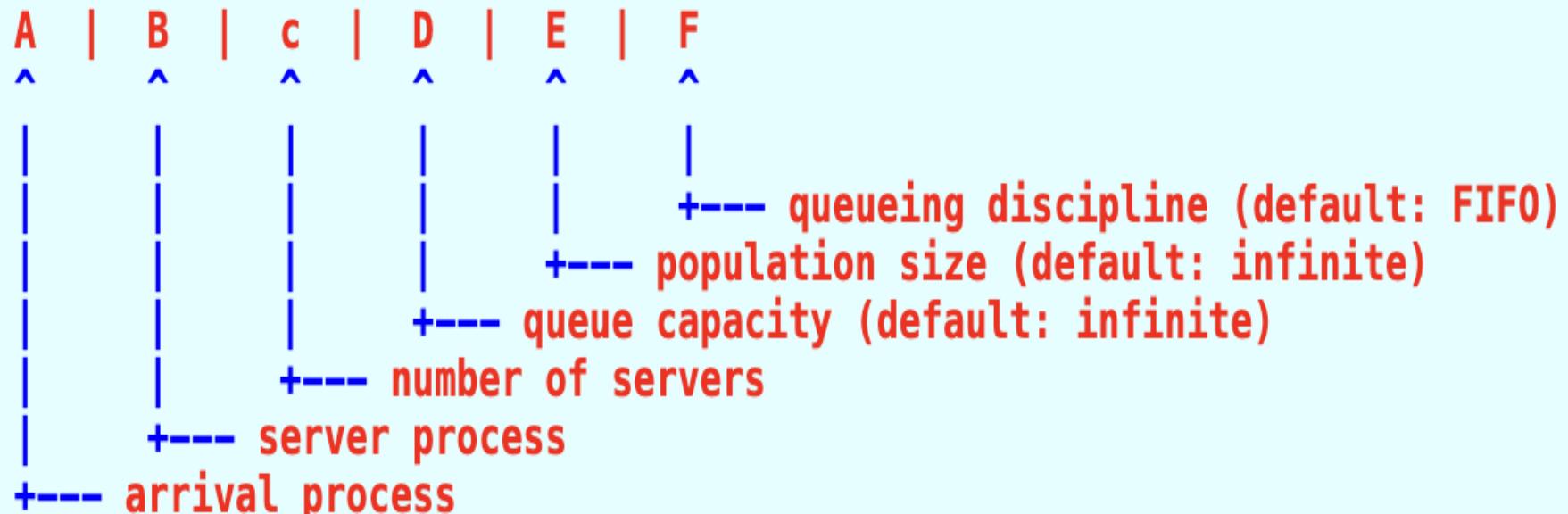


Kendall notation for a queueing system

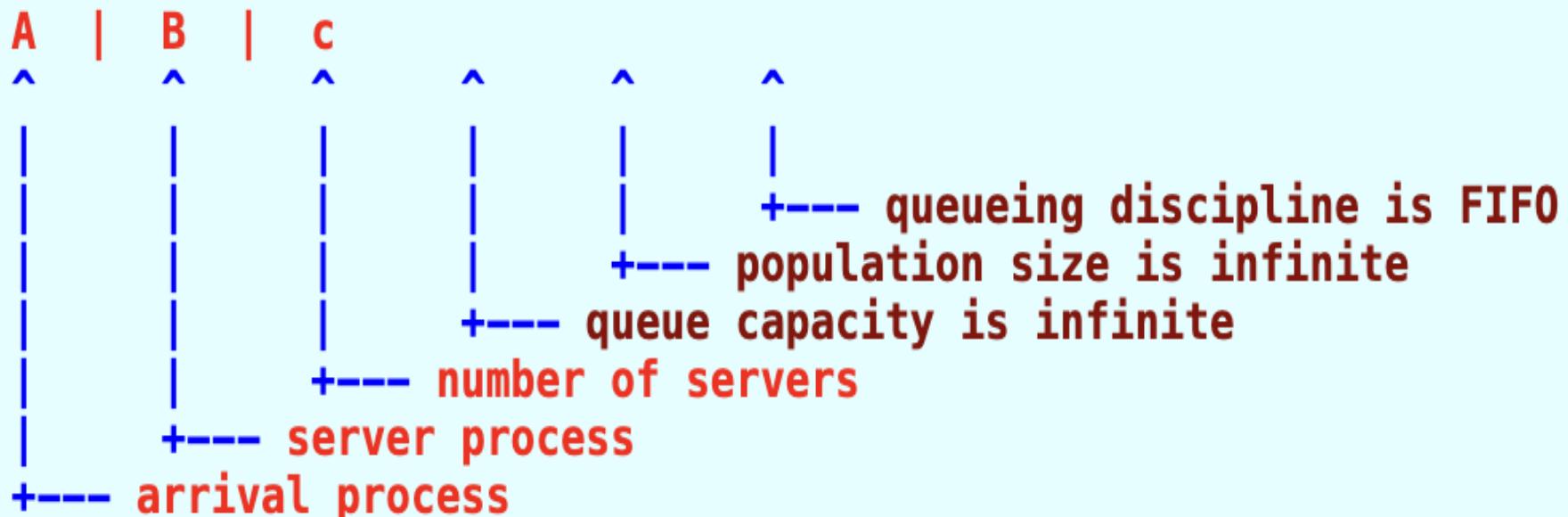
- Kendall's notation (or sometimes Kendall notation) the standard system used to describe and classify the queueing model that a queueing system corresponds to.
- The notation was first suggested by **D. G. Kendall** in **1953**
-



Long Kendall notation:



Abbreviated Kendall notation:



Further notations:

- Abbreviations for arrival and service processes:

- **M = Poisson process**
- **D = Deterministic process** (fixed time between 2 consecutive events)
- **G = general process**

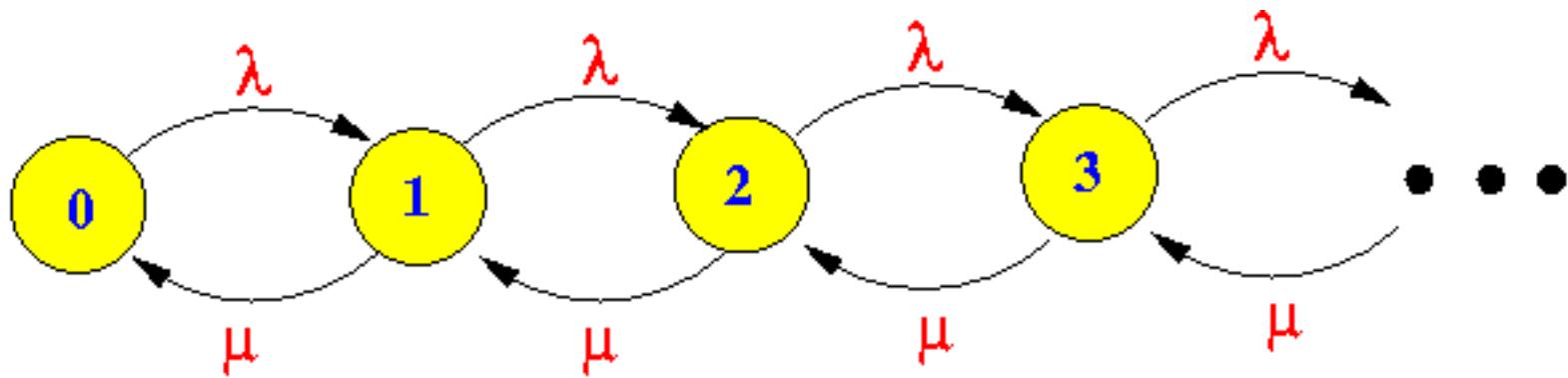


The *classic* M/M/1 queue

- The M/M/1 queue is a **short hand** notation for the **M/M/1/ $\infty/\infty/\text{FIFO}$** queue:
- **M** = **arrival process** is **Poisson** (with some parameter **lambda**);
- **M** = **service (departure) process** is **Poisson** (with some parameter **mu**);
- **1** = there is **1 server** in system
- **∞** = **infinite queue capacity**; **no arriving client will be rejected**
- **∞** = **infinite population size** (the **arrival process** will be **unaffected** by the number of clients already in the system)
- **FIFO** = first in first out service
-



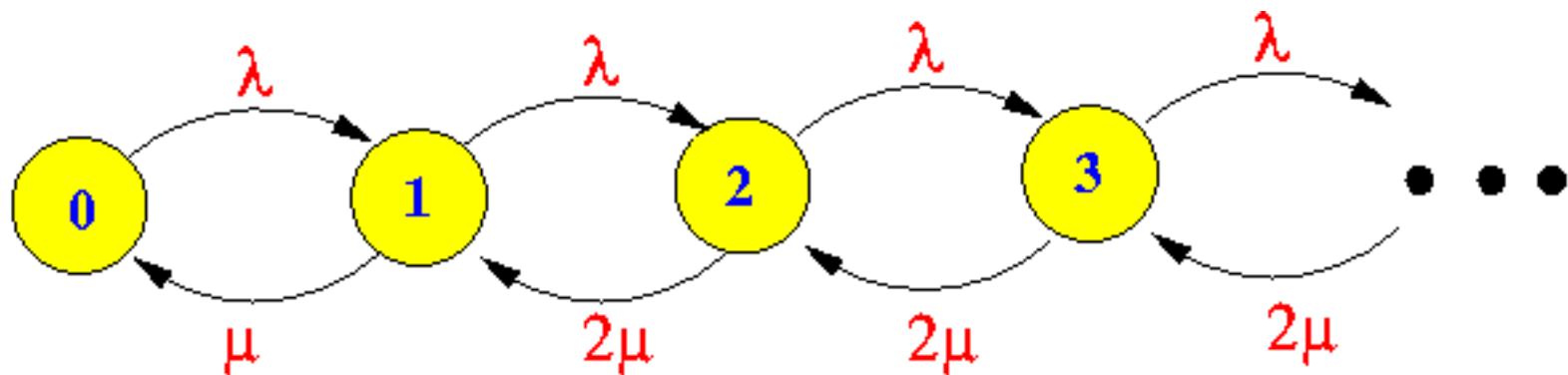
The rate transition diagram for the M/M/1 queueing system is:



- We have already studied this system...

One more servers: the M/M/2 system

The **rate transition diagram** for the **M/M/2** system is:



Note: A single service system where the maximum departure rate is now 2μ .

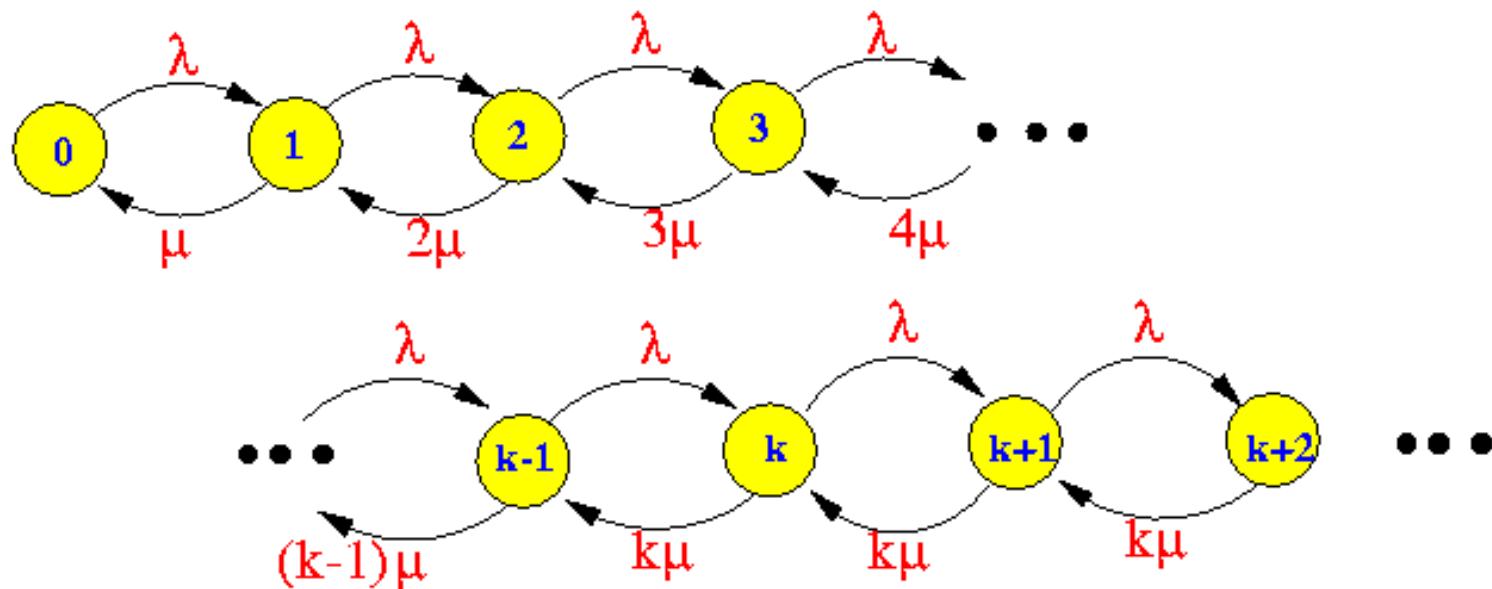
Equilibrium equations for M/M/2:

- $p_0 \times \lambda = p_1 \times \mu$
- $p_1 \times \lambda = p_2 \times 2\mu$
- $p_2 \times \lambda = p_3 \times 2\mu$
- $p_3 \times \lambda = p_4 \times 2\mu$
-



More servers in general: M/M/k

- The rate transition diagram for the M/M/k system is:
-



Equilibrium equations for M/M/k:

- $p_0 \times \lambda = p_1 \times \mu$
- $p_1 \times \lambda = p_2 \times 2\mu$
- $p_2 \times \lambda = p_3 \times 3\mu$
-
- $p_{k-2} \times \lambda = p_{k-1} \times (k-1)\mu$
- $p_{k-1} \times \lambda = p_k \times k\mu$
- $p_k \times \lambda = p_{k+1} \times k\mu$
- $p_{k+1} \times \lambda = p_{k+2} \times k\mu$
- $p_{k+2} \times \lambda = p_{k+3} \times k\mu$

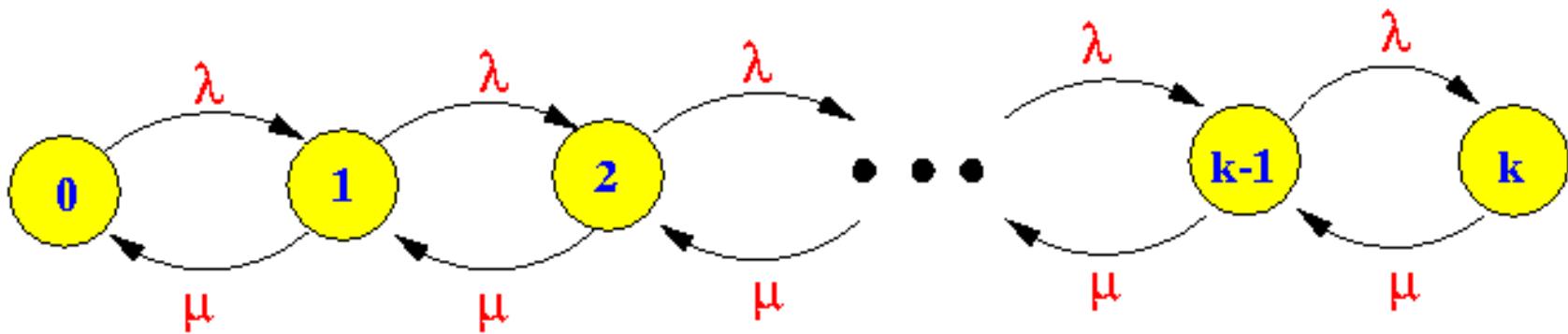


Finite buffer capacity: M/M/1/k

- The M/M/1/k queue is a **short hand** notation for the M/M/1/k/ ∞ /FIFO queue:
- - M = arrival process is Poisson (with some parameter lambda;)
 - M = service (departure) process is Poisson (with some parameter lambda;)
 - 1 = there is 1 server in system
 - k = queue capacity; the (k+1)th arriving client will be *rejected*
 - ∞ = infinite population size (the arrival process will be unaffected by the number of clients already in the system)
 - FIFO = first in first out service



The rate transition diagram for the M/M/1/k queueing system is:



Notice that:

The **number of customers** in the system is **at most k**

Hence, the states of this **Markov chain** are **0, 1, 2, ..., k**

Equilibrium equations for M/M/1/k:

$$p_0 \times \lambda = p_1 \times \mu$$

$$p_1 \times \lambda = p_2 \times \mu$$

$$p_2 \times \lambda = p_3 \times \mu$$

$$p_3 \times \lambda = p_4 \times \mu$$

....

$$p_{k-1} \times \lambda = p_k \times \mu$$



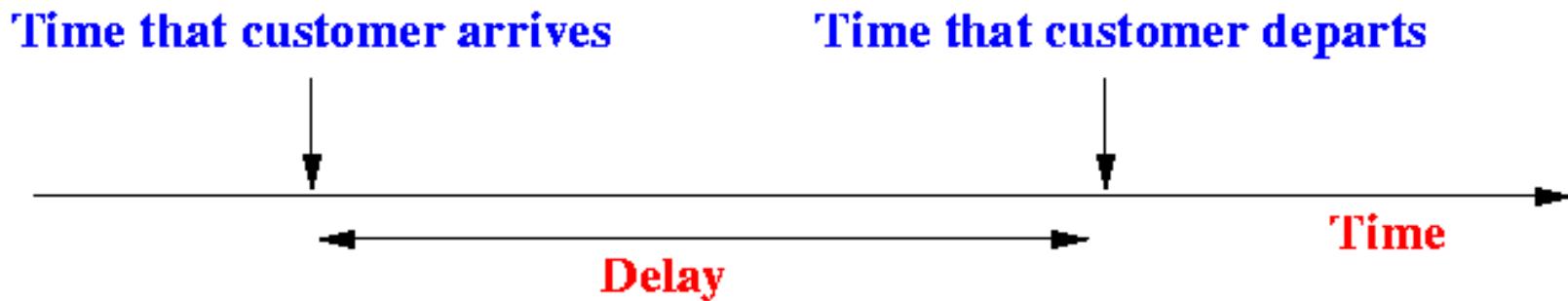
Performance analysis: Average queue length (average number of customers in system)

- Purpose of queueing models:
- Analysis of queueing models can provide answers to performance related questions through mathematical analysis
- Commonly used performance measures
 - **Average queue length N**
 - This is the **average number of customers** in the system.
 - (Better: This is the **average number of customers waiting** in the system to get service....)



Average delay time \bar{T}

- The **delay** is defined as:



- The **average delay time** is the **average amount of time** that a customer spends in the system.

Computing the *average queue length* (in general)

- Consider the following simple example:
- **50% of the time**, the system is **empty**
- **50% of the time**, the system has **1 customer**
- Then the **average number of customers** in the system is the **weighted sum**
- $\underline{N} = 0.5 \times 0 + 0.5 \times 1 = 0.5$



Average queue length

- In general, the average queue length (or the average number of customers in system) is equal to:

$$\begin{aligned} \underline{N} &= \text{Mean (expected) number of customer} \\ &= 0 \times P[0 \text{ customers in system}] \\ &+ 1 \times P[1 \text{ customer in system}] \\ &+ 2 \times P[2 \text{ customers in system}] + \dots \end{aligned}$$

$$= \sum_{\{k = 0, 1, \dots, \infty\}} k \times P[k \text{ customers in system}]$$

(Definition of "expected value")



Example: Average queue length in the M/M/1 queue

Recall that the state probability i of the M/M/1 queue is
 $P[k \text{ customers in system}] = \rho^k (1 - \rho)$

- What is the the **average (expected) queue length \bar{N}** of the M/M/1 queue ?
- Let us derive this ...



Contd...

$$\begin{aligned}\underline{N} &= \sum_{k=0}^{\infty} k \times p_k \\&= \sum_{k=0}^{\infty} k \times \rho^k (1 - \rho) \\&= (1 - \rho) \times \sum_{k=0}^{\infty} k \times \rho^k \quad \dots \quad (2)\end{aligned}$$

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

Or:

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



Thus:

$$\frac{dU}{dp} = \sum_{\{k = 0, 1, \dots, \infty\}} k \rho^{k-1} = (1 - \rho)^{-2}$$

Therefore:

$$\sum_{\{k = 0, 1, \dots, \infty\}} k \rho^k = \rho \times (1 - \rho)^{-2}$$

..... (3)



Finally...

Substitute (3) in (2):

$$\begin{aligned} \underline{N} &= (1 - \rho) \times \sum_{\{k = 0, 1, \dots, \infty\}} k \times \rho^k \\ &= (1 - \rho) \times \rho \times (1 - \rho)^{-2} \\ &= \rho \times (1 - \rho)^{-1} \end{aligned}$$



So...

Therefore, the mean number of customers
in an M/M/1 queue is equal to:

$$\underline{N(M/M/1)} = \frac{\rho}{1 - \rho} \quad \dots \dots (4)$$



Performance analysis: Average delay

- Unlike *average queue length*, the average delay cannot be directly derived
- [We don't have the distribution]
- Fortunately, there is a simple indirect way to compute the **average delay**
- **Little's law (Little's formula)**
- John Little proved the following *famous* formula
- The **long-term average number of customers** in a **stable system** is equal to: the long-term average arrival rate multiplied by the long-term average time a customer spends in the system



In other words:

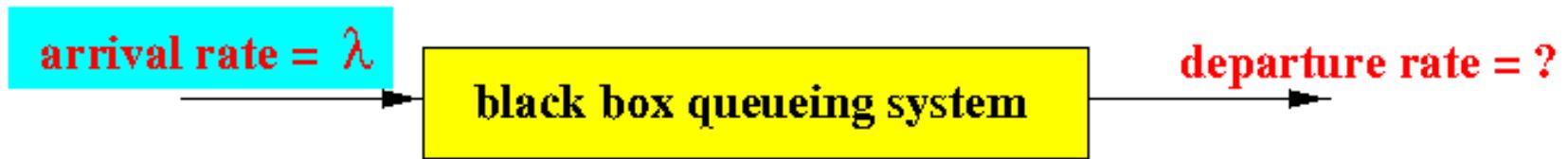
- Avg. # customers in system =
 $\lambda \times$ Avg. delay of customers
- **More intuitive form -**

$$\bullet = \frac{\text{Avg. # customers in system}}{\text{Avg. delay of customers}} = \lambda \text{ (Little's formula)}$$



Proof:

- Consider a **stable system** that has an **average arrival rate** of λ :



Claim: The **long term departure rate** is **also** equals to λ

Reason:

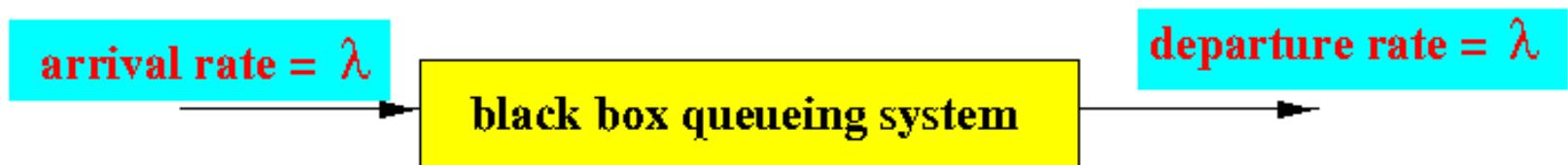
If the **long term departure rate** is less ($<$) the **long term arrival rate** λ , then the **number of customers** in the system will keep **growing to infinity** and the **system will not be stable**

On the other hand, if the **long term departure rate** is greater ($>$) the **long term arrival rate** λ , then the **number of customers** in the system will **becomes equal to 0**. Such a system is **not be stable**

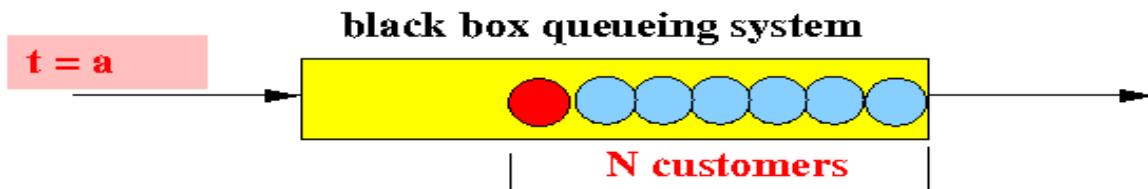


So...

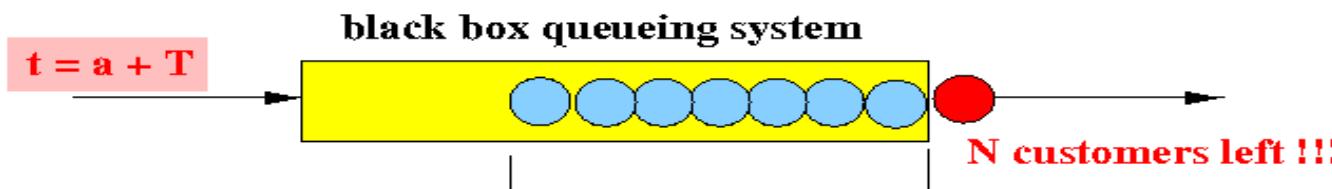
- Conclusion: in a *stable* system, *long term* departure rate = *long term* arrival rate



We can calculate the *long term* departure rate as follows:



When red customer arrives,
he finds an average of N customers
in system (including himself !)



The red customer spends avg. T sec in system

Hence, the *long term* departure rate is equal to:

which is the Little's formula !!!

$$\text{long term departure rate} = \frac{\text{Avg. # customers in system}}{\text{Avg. time a customer spends in system}}$$

Example: Average service time of a customer in the M/M/1 queue

- According to a previous analysis the **average queue length** in the M/M/1 queueing system is:
-

$$\underline{N(M/M/1)} = \frac{\rho}{1 - \rho}$$

According to the Little's formula:

$$T = N / \lambda$$

(Little's formula in another form)



So...

- Therefore, the **average service time** of the M/M/1 queue is

$$T = \frac{N}{\lambda}$$

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

$$= \frac{1}{\lambda} * \frac{\lambda/\mu}{1 - \rho}$$

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

..... (3)



Other performance measures: Idle time/busy time

How often do you waste system resources:

- **Idle system:** The system is **idle** when there are **zero (0) customers** in the system
- The **fraction of time** that the **system is idle** is equal to the **probability that the system is *empty***
- **Therefore:**
- Fraction of time that system is idle =

$$P[\text{0 customer in system}] = p_0$$



The fraction of time that an M/M/1 queue is idle is equal to:

- fraction of time that M/M/1 queue is idle = $1 - \rho$
- **Busy system: (busy = not idle)**
- The system is **idle** when there are **not zero (0) customers** in the system
- The **fraction of time that the system is idle** is equal to the **probability that the system is *not* empty**
- **Therefore:** fraction of time that system is idle = $1 - P[0 \text{ customer in system}] = 1 - p_0$



So...

- The **fraction of time** that an M/M/1 queue is **busy** is equal to:
- **fraction of time that M/M/1 queue is idle = ρ**



Other performance measure: Loaded system

- The **M/M/*/k** queueing model is used to model system with **finite waiting capacity** (the **maximum number of customers in the system is at most k**)
- We can ask the following **capacity question** on system with **finite waiting capacity**:
- **How often** is the system running **at system capacity** i.e., **how often** is the **number of customers** in the system **equal to its maximum**
- The **fraction of time** that an **M/M/*/k** queueing system is running **at maximum capacity** can be computed as: **fraction of time that M/M/*/k system is at max. cap.**

 $= P[k \text{ customers in system}] = p_k$



Summary: Visualizing the relations

Inputs:

Arrival Rate

Departure Rate

No of Servers

Queue Capacity

...

Outputs:

Probability of staying at a state (Steady state probability)

Average length of the queue

Average delay of a customer

Idle time

Busy time

...

