

ESD.86

Markov Processes and their Application to Queueing



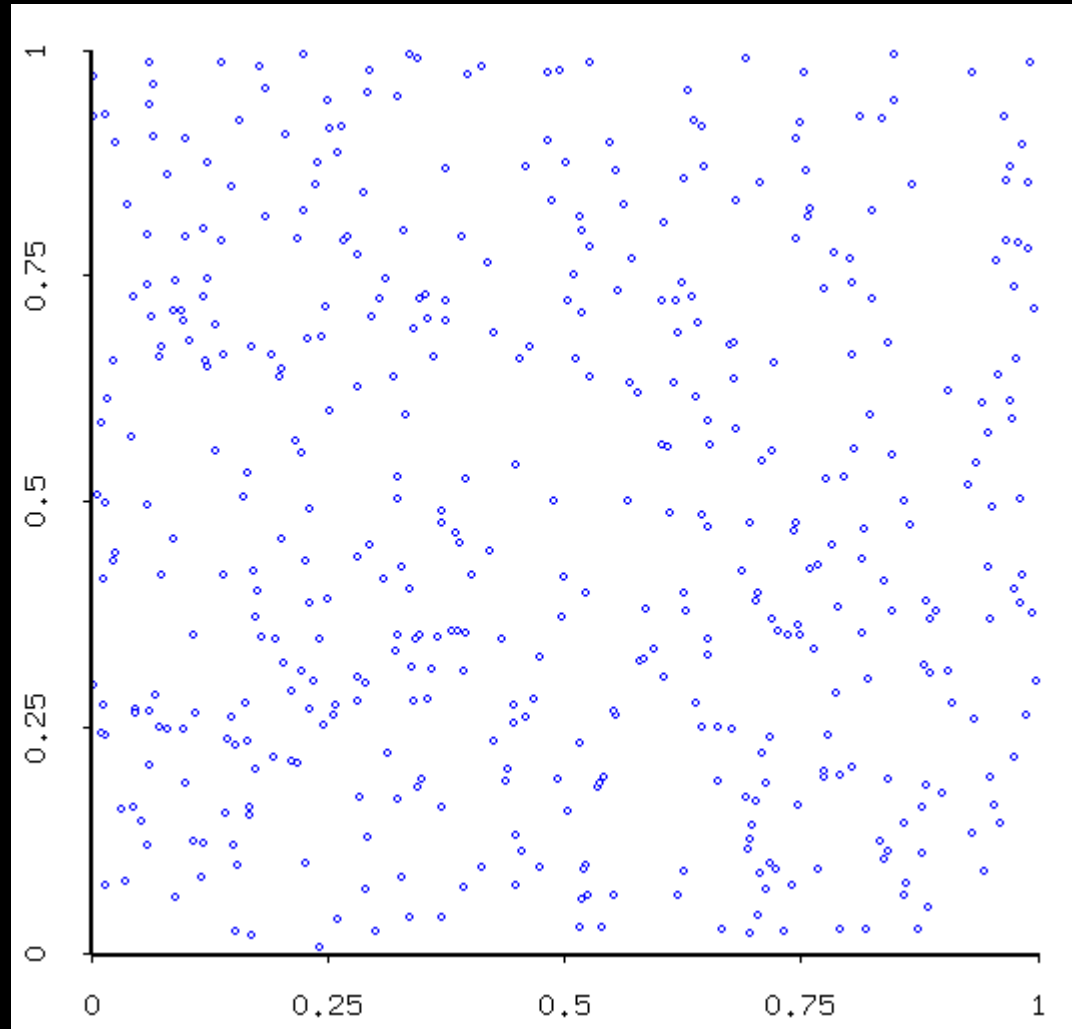
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Outline

- ◆ Spatial Poisson Processes, one more time
- ◆ Introduction to Queueing Systems
- ◆ Little's Law
- ◆ Markov Processes

Spatial Poisson Processes



Courtesy of Andy Long. Used with permission.

Spatial Poisson Processes

- ◆ Entities distributed in space (Examples?)
- ◆ Follow postulates of the (time) Poisson process
 - $\lambda dt = \text{Probability of a Poisson event in } dt$
 - *History not relevant*
 - *What happens in disjoint time intervals is independent, one from the other*
 - *The probability of a two or more Poisson events in dt is second order in dt and can be ignored*
- ◆ Let's fill in the spatial analogue.....

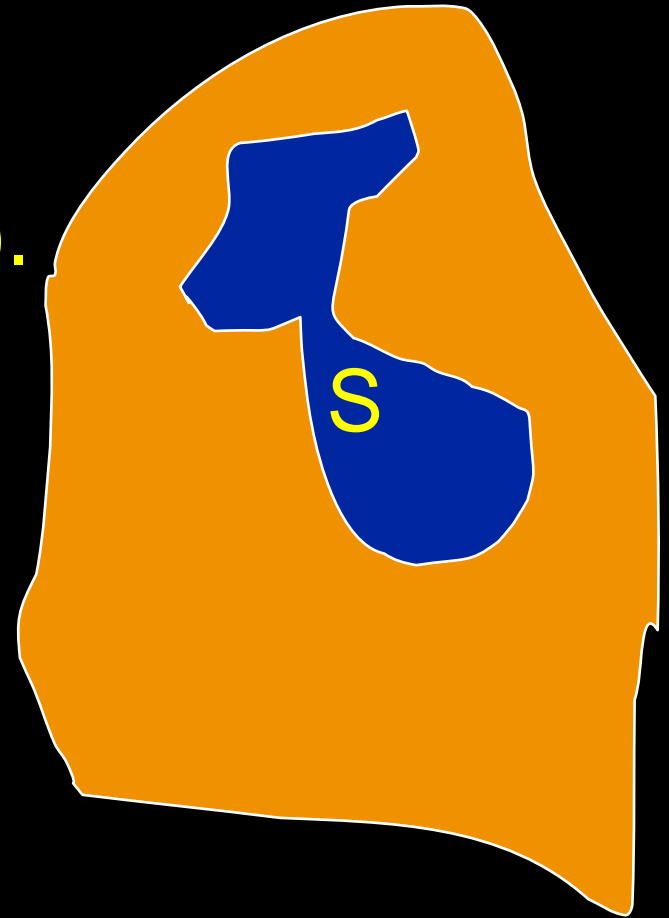
Set S has area $A(S)$.

Poisson intensity is γ

Poisson entities/(unit area).

$X(S)$ is a random variable

$X(S)$ = number of Poisson
entities in S



$$P\{X(S) = k\} = \frac{(\gamma A(S))^k}{k!} e^{-\gamma A(S)}, \quad k = 0, 1, 2, \dots$$

Nearest Neighbors: Euclidean

Define D_2 = distance from a random point to nearest Poisson entity

Want to derive $f_{D_2}(r)$.

Happiness:

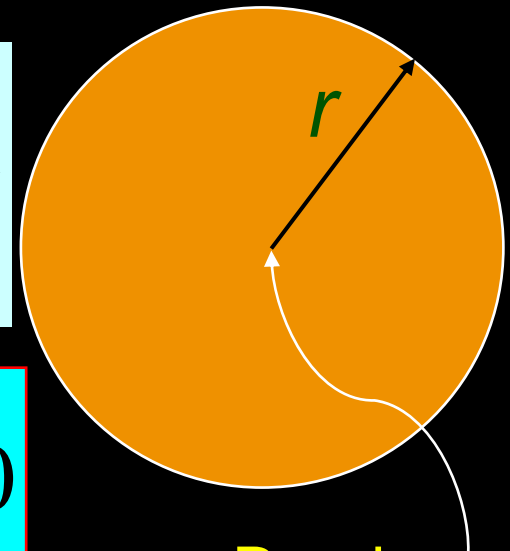
$$F_{D_2}(r) \equiv P\{D_2 \leq r\} = 1 - P\{D_2 > r\}$$

$$F_{D_2}(r) = 1 - \text{Prob}\{\text{no Poisson entities in circle of radius } r\}$$

$$F_{D_2}(r) = 1 - e^{-\gamma\pi r^2} \quad r \geq 0$$

$$f_{D_2}(r) = \frac{d}{dr} F_{D_2}(r) = 2r\gamma\pi e^{-\gamma\pi r^2} \quad r \geq 0$$

Rayleigh pdf with parameter $\sqrt{2\gamma\pi}$



Random Point

Nearest Neighbors: Euclidean

Define D_2 = distance from a random point to nearest Poisson entity

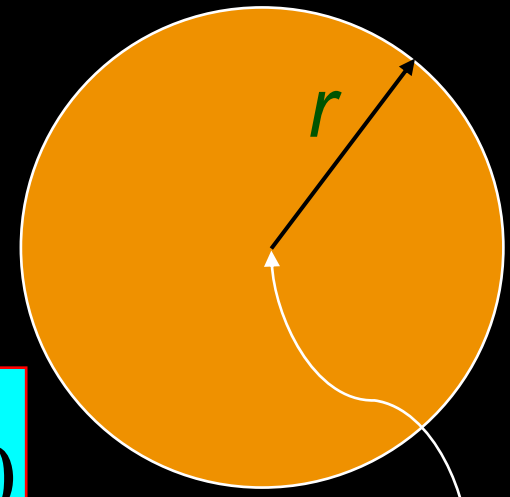
Want to derive $f_{D_2}(r)$.

$$E[D_2] = (1/2) \sqrt{\frac{1}{\gamma}} \quad \text{"Square Root Law"}$$

$$\sigma_{D_2}^2 = (2 - \pi/2) \frac{1}{2\pi\gamma}$$

$$f_{D_2}(r) = \frac{d}{dr} F_{D_2}(r) = 2r\gamma\pi e^{-\gamma\pi r^2} \quad r \geq 0$$

Rayleigh pdf with parameter $\sqrt{2\gamma\pi}$

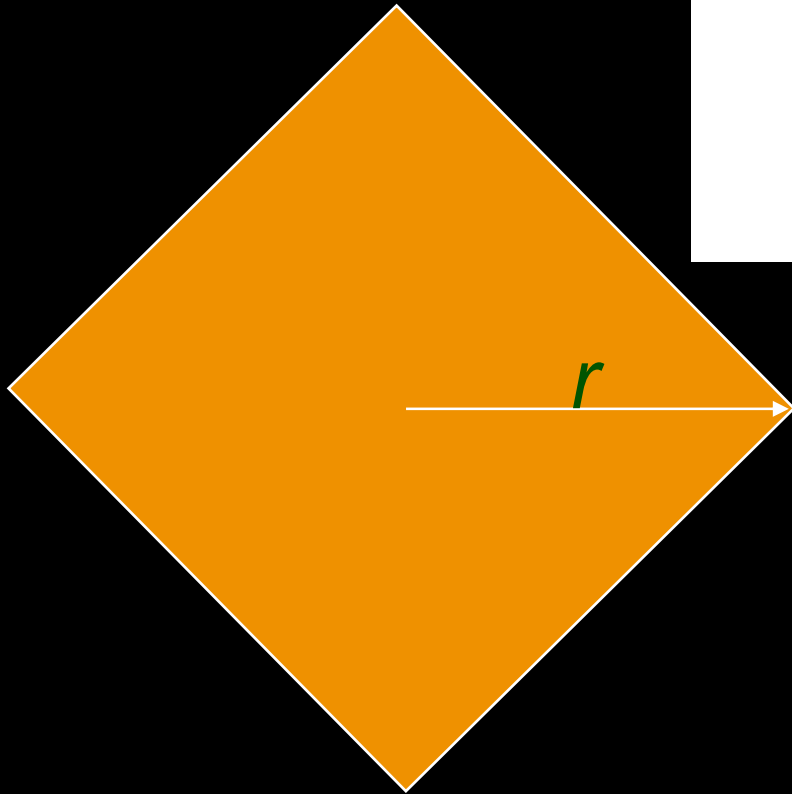


Random Point

Nearest Neighbor: Taxi Metric

$$F_{D_1}(r) \equiv P\{D_1 \leq r\}$$

$$F_{D_1}(r) = 1 - \Pr\{\text{no Poisson entities in diamond}\}$$



How Might you Derive the PDF
fo the k^{th} Nearest Neighbor?

Blackboard exercise!

To Queue or Not to Queue,
That May be a Question!



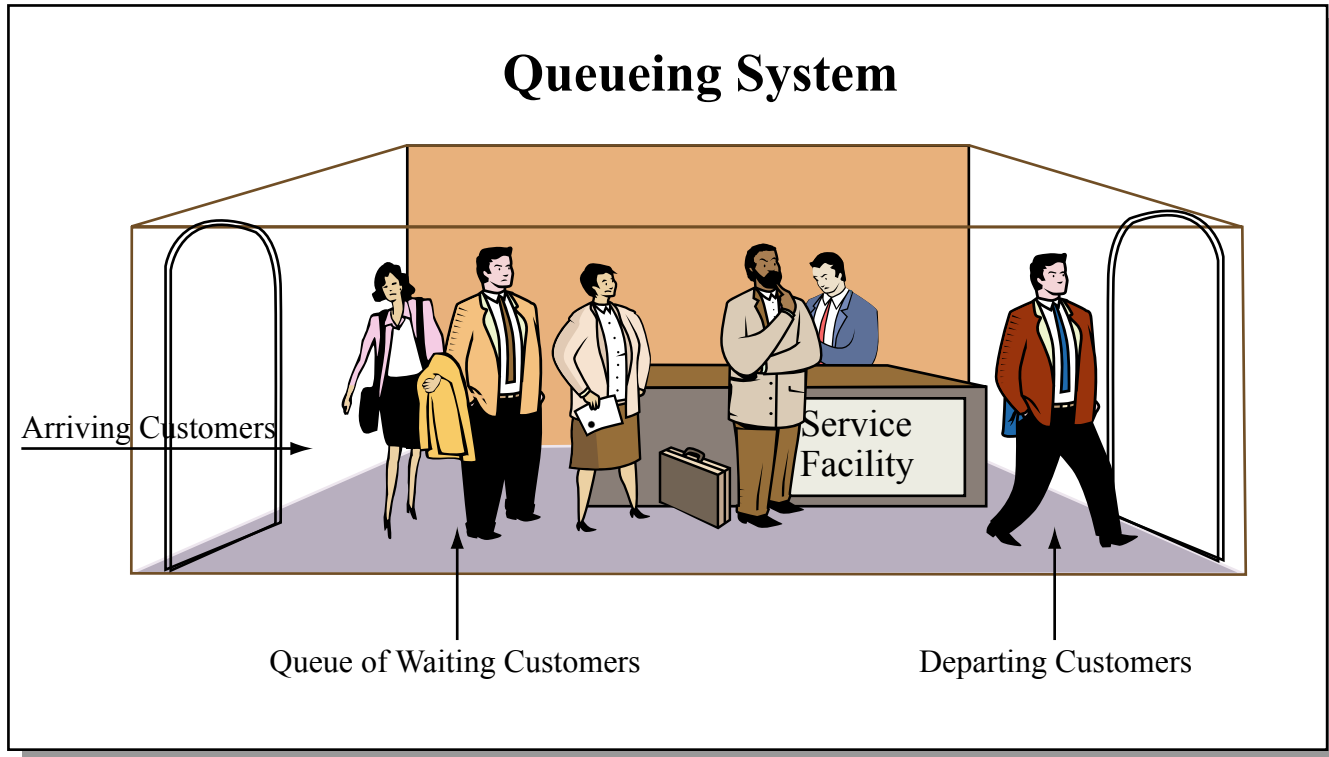
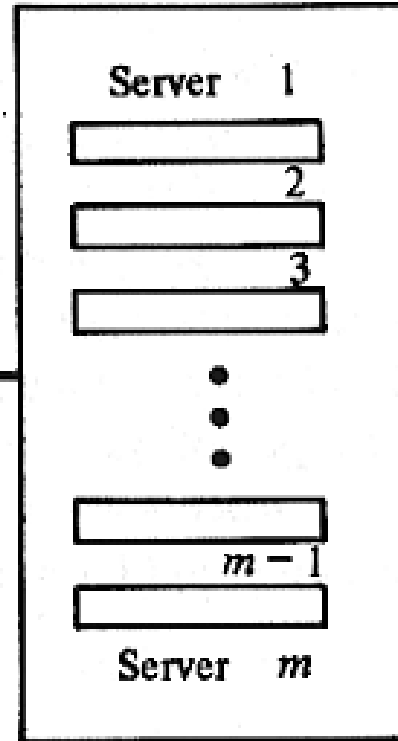
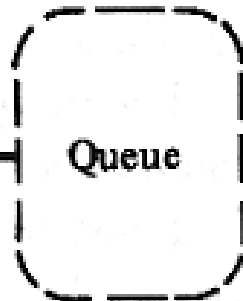


Figure by MIT OCW.

Servers:
Statistical Clones?

Finite or
Infinite?

Finite or
Infinite?



Queue
Discipline:
How queuers
Are selected
for service

Point of "arrival"
at the system

Departure
from the system

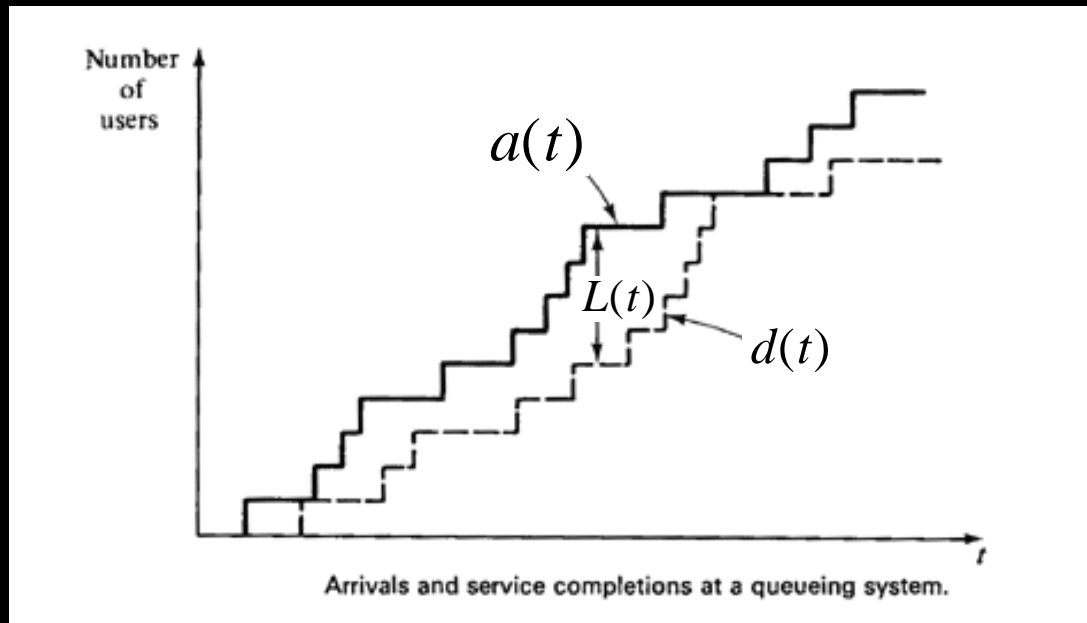
Generic queueing system.

What Kinds of Queues Occur in Systems of Interest to ESD?



ESD
Queues?

Little's Law for Queues



Source: Larson and Odoni, *Urban Operations Research*

$a(t)$ = cumulative # arrivals to system in $(0, t]$

$d(t)$ = cumulative # departures from system in $(0, t]$

$L(t) = a(t) - d(t)$

$L(t)$ = number of customers in the system
(in queue and in service) at time t

Little's Law for Queues

$$\gamma(t) = \int_0^t [a(\tau) - d(\tau)] d\tau = \int_0^t L(\tau) d\tau$$

$\gamma(t)$ = total number of customer minutes spent in the system

$a(t)$ = cumulative # arrivals to system in $(0, t]$

$d(t)$ = cumulative # departures from system in $(0, t]$

$$L(t) = a(t) - d(t)$$

$L(t)$ = number of customers in the system

(in queue and in service) at time t

Let's Get an expression for Each of 3 Quantities

$\lambda_t \equiv$ average customer arrival rate = $a(t)/t$

$$L_t = \frac{\gamma(t)}{t} = \frac{a(t)}{t} \frac{\gamma(t)}{a(t)} = \lambda_t W_t$$

In the limit,

$$L = \lambda W, \quad \text{Little's Law}$$

Key Issues

$$L = \lambda W$$

- ◆ L is a time-average. Explain
- ◆ λ is average of arrival rate of customers who actually enter the system
- ◆ W is average time in system (in queue and in service) for actual customers who enter the system

More Issues

$$L = \lambda W$$

- ◆ *Little's Law is general. It does not depend on*
 - Arrival process
 - Service process
 - # servers
 - Queue discipline
 - Renewal assumptions, etc.
- ◆ *It just requires that the 3 limits exist.*

Still More Issues

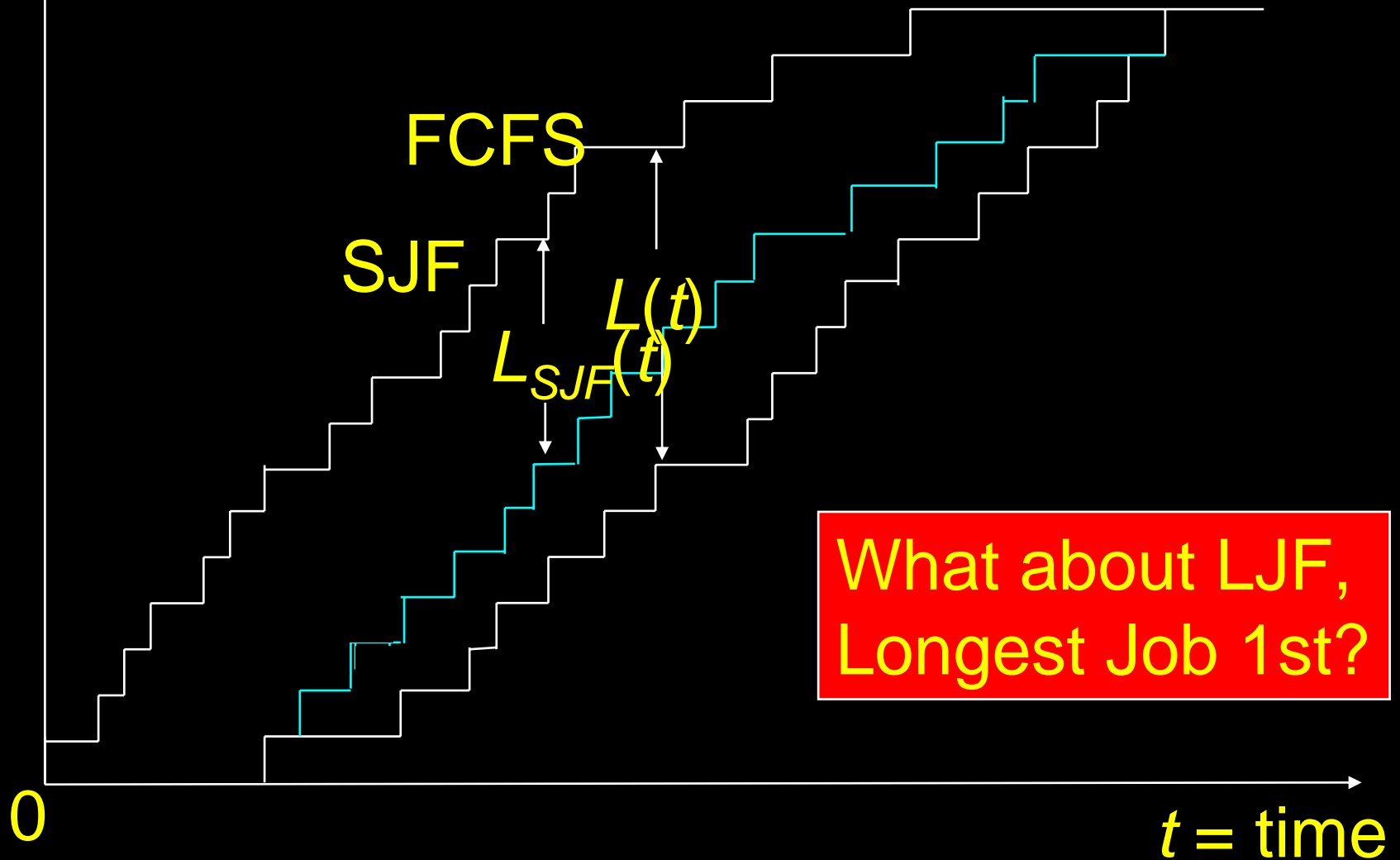
$$L = \lambda W$$

- ◆ *What about balking? Reneging? Finite capacity?*
- ◆ *Do we need iid service times? Iid inter-arrival times?*
- ◆ *Do we need each busy period to behave statistically identically?*
- ◆ *Look at role of $\gamma(t)$. Can change queue statistics by changing queue discipline.*

Cumulative # of Arrivals

FCFS=First Come, First Served

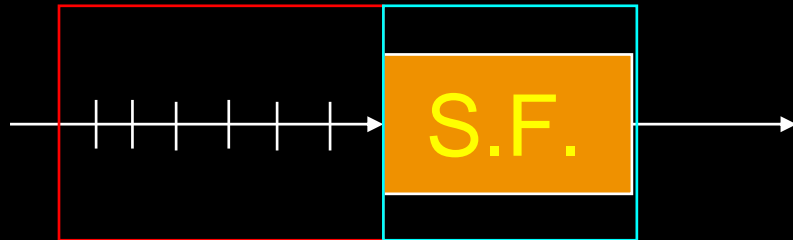
SJF=Shortest Job First



“System” is
General

$$L = \lambda W$$

- ◆ *Our results apply to entire queue system, queue plus service facility*
- ◆ *But they could apply to queue only!*



$$L_q = \lambda W_q$$

- ◆ *Or to service facility only!*

$$L_{SF} = \lambda W_{SF} = \lambda / \mu$$

$1 / \mu =$ mean service time

All of this means,
“You buy one, you get the other 3 for free!”

$$W = \frac{1}{\mu} + W_q$$

$$L = L_q + L_{SF} = L_q + \frac{\lambda}{\mu}$$

$$L = \lambda W$$

Utilization Factor ρ

- ◆ Single Server. Set $Y = \begin{cases} 1 & \text{if server is busy} \\ 0 & \text{if server is idle} \end{cases}$

$$E[Y] = 1 * P\{\text{server is busy}\} + 0 * P\{\text{server is idle}\}$$

$$E[Y] = 1 * \rho + 0 = \rho = E[\# \text{ customers in SF}] = ?$$

- ◆ $E[Y]$ is time-average number of customers in the SF
- ◆ Buy Little's Law,

$$\rho = \lambda / \mu < 1$$

Utilization Factor ρ

- ◆ Similar logic for N identical parallel servers gives

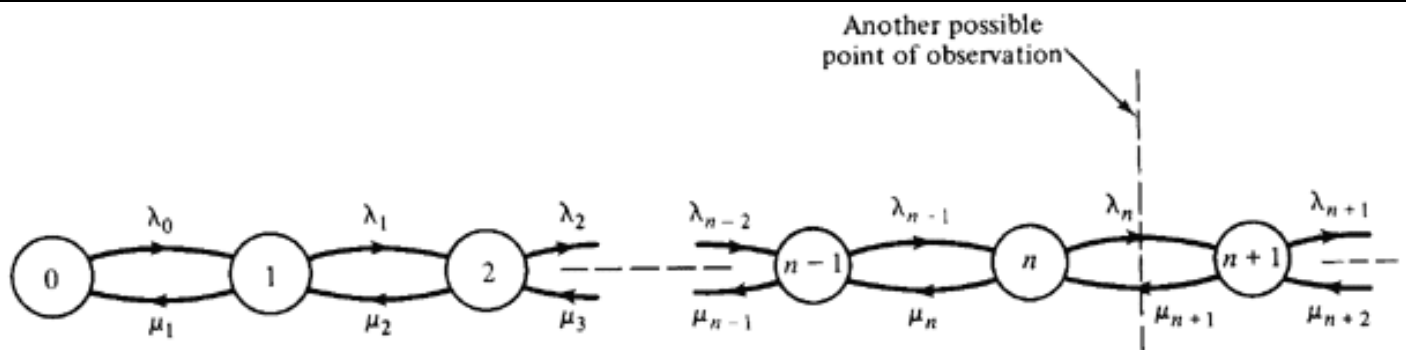
$$\rho = \left(\frac{\lambda}{N}\right) \frac{1}{\mu} = \frac{\lambda}{N\mu} < 1$$

- ◆ Here, λ/μ corresponds to the time-average number of servers busy

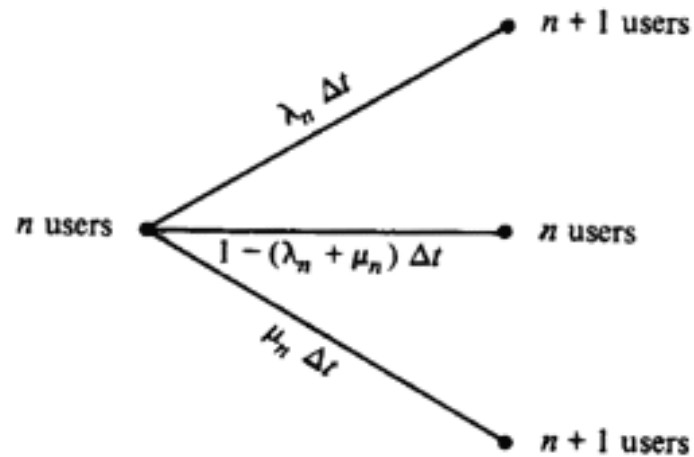
Markov Queues

Markov here means, “No Memory”





State-transition diagram for the fundamental birth-and-death model.



Time: t

Time: $t + \Delta t$

Probabilities of transitions for birth-and-death model in time Δt .

Balance of Flow Equations

$$\lambda_0 P_0 = \mu_1 P_1$$

$$(\lambda_n + \mu_n) P_n = \lambda_{n-1} P_{n-1} + \mu_{n+1} P_{n+1} \text{ for } n = 1, 2, 3, \dots$$

To be continued.....