Redes de Computadores

Delay Models in Computer Networks

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- » What are the common multiplexing strategies?
- *» What is a Poisson process?*
- » What is the Little theorem?
- » What is a queue?
- » What is the meaning of service time $1/\mu$ in a queue of packets?
- » What is the meaning of traffic intensity ρ in a queue model?
- » What is the probability of a M/M/1 queue being in a given state n?
- » What is the mean number of clients in a M/M/1 queue? What is the mean waiting time in a M/M/1 queue? What is the relationship between N and ρ in a M/M/1 queue?
- » What are the differences between M/M/1 and M/G/1 queues? How to estimate mean number of packets and mean delay in a M/G/1 queue?
- » How to model a network of transmission lines? How to calculate the mean number of packets and mean delay in this case?
- » What is a Jackson Network? Why is it important?

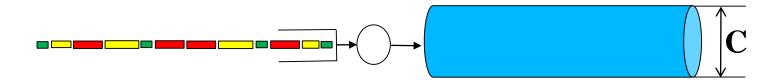
Multiplexing Traffic on a Link



- Communication link
 - » Bit pipe with a given capacity C (bit/s)
 - » Link capacity → rate at which bits are transmitted to the link
 - » Link may transport multiplexed traffic streams
- Multiplexing strategies
 - » Statistical Multiplexing
 - » Frequency Division Multiplexing
 - » Time Division Multiplexing
- Multiplexing strategy affects traffic delay

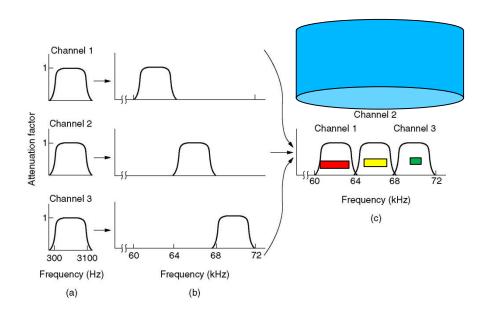
Statistical Multiplexing

- Packets of all traffic streams merged in a single queue
- Packets transmitted on a first-come first-served basis
- Time required to transmit a packet of length $L \rightarrow T_{frame} = L/C$



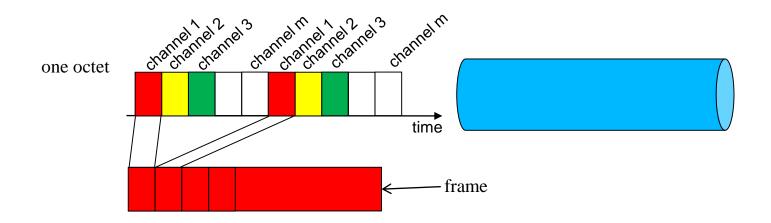
FDM – Frequency Division Multiplexing

- Link capacity C subdivided into *m* portions
- Channel bandwidth W subdivided into m channels of W/m Hz
- Capacity of each channel → C/m
- Time required to transmit a packet of length $L \rightarrow T_{frame} = Lm/C$

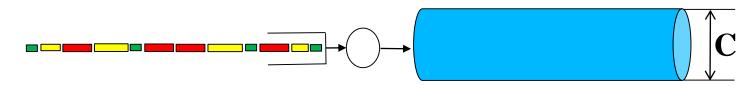


TDM – Time Division Multiplexing

- ◆ Time axis divided into *m* slots of fixed length (usually one octet long)
- ◆ Communication → m channels with capacity C/m
- Time required to transmit a packet of length $L \rightarrow T_{frame} = Lm/C$

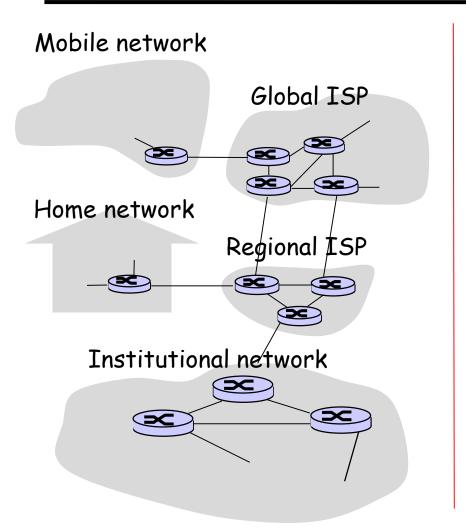


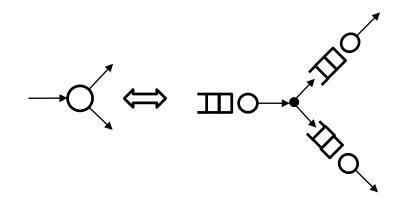
Delay on Computer Networks

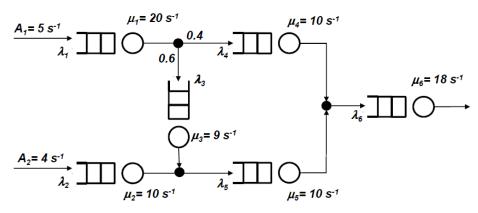


- Delay
 - » Important performance parameter in computer networks
 - » Characterized using queue models
- Queue model
 - » Customers arrive at random times to obtain service
 - » Customer → packet to be transmitted through a link
 - » Serve a packet = transmit a packet
 - » Service time \rightarrow packet transmission time = $T_{pac(frame)}$ = L/C
- Queue models enable the quantification of
 - » Average number of customers/packets in the network
 - » Average delay per packet → waiting plus service times

Computer Networks Modeled as Queue Networks







Poisson Distribution and Poisson Process

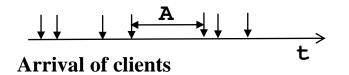
◆ Poisson distribution with parameter m

$$P[N=n] = p_n = \frac{m^n e^{-m}}{n!}, \quad n = 0,1,... \qquad E[N] = Var[N] = m$$

- Poisson process
 - $\rightarrow \lambda T = m$, (e.g. $\lambda \rightarrow arrivals/s$)
 - » P[n arrivals in interval T] = $p_n(T) = p_n = \frac{(\lambda T)^n e^{-\lambda T}}{n!}$ $E[N] = Var[N] = \lambda T$

 $p_{n}(T) = \frac{(\lambda T)^{n} e^{-\lambda T}}{n!}$ 0.08 0.08 0.06 0.04 0.02 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00

Inter-Arrival Interval A — Statistical Characterization



A - time interval between the arrival of consecutive clients

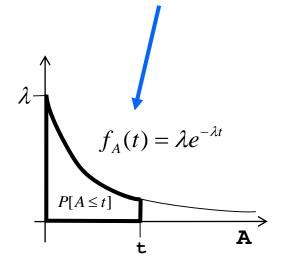
$$F_A(t) = P[A \le t] = 1 - P[A > t] = 1 - p_0(t) = 1 - e^{-\lambda t}$$

$$f_A(t) = pdf = \frac{\partial F_A(t)}{\partial t} = \lambda e^{-\lambda t}$$

$$E[A] = \frac{1}{\lambda}$$

$$Var[A] = \frac{1}{\lambda^2}$$

Exponential distribution



• What is the difference between

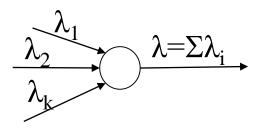
Deterministic arrivals and

Poisson arrivals?

Markov Process - Properties

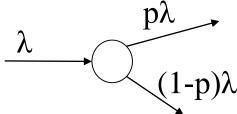
Merging Property

- » $A_1, A_2, \dots A_k$ are independent Poisson Processes with rates $\lambda_1, \lambda_2, \dots \lambda_k$
- » $A=\Sigma A_i$ still is a Poisson process, with rate $\lambda=\Sigma\lambda_i$



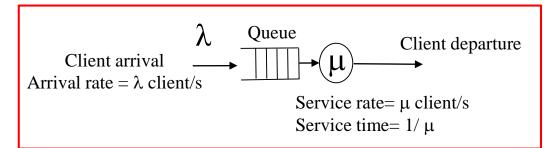
Splitting property

- » Packets arrive to a router according to a Poisson Process (A,λ)
- » They are routed randomly to two output lines with probabilities **p** and **1-p**
- » Packets leaving the router still are Poisson Processes, characterized by $(A,p\lambda)$ and $(A,(1-p)\lambda)$



Queue Model

- ◆ Queue model used for
 - » Customers waiting in line
 - » Packets in a network



- Used to determine
 - » Average number of clients in the system \rightarrow N
 - » Average delay experienced by a client \rightarrow T
- Queue characterized in terms of
 - » λ arrival rate of client (average number of clients per time unit)
 - » μ service rate (average number of clients the server processes per time unit)
 - » $\rho = \lambda / \mu$ traffic intensity (occupation of the server)
- ★ Kendall notation → A/S/s/K
 - → A arrival statistical process
 - » S − service statistical process

 - \times K capacity of the system in buffers

Little's Theorem

\bullet N= λ T

- » N- average number of clients in a system
- \rightarrow T average amount of time a client spends in the system
- » λ arrival rate of clients to the system
- \bullet T=T_w+T_s
 - » T_w time a client waits in the queue for being served
 - T_s service time
- \bullet N=N_w+N_s
 - \sim N_w number of clients waiting in the queue for being served
 - » N_s number of clients being served
- $N_w = \lambda T_w$

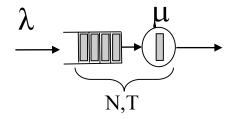
$$N_w = \lambda T_w \rightarrow T_w = N_w / \lambda$$

• The (mean) time a client has to wait before being served (T_w) depends on the number of clients waiting (N_w) and on the arrival rate of clients (λ)

- No dependence on the service rate?!
- Can you explain it?

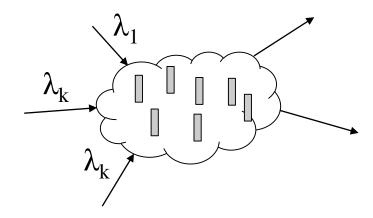
Little's Theorem

• Can be applied to a single Queue



- Can be applied to a complex system
 - » For each stream $i \rightarrow N_i = \lambda_i T_i$
 - » For the system:

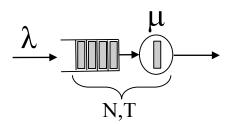
$$\lambda = \Sigma \lambda_i$$
 $N = \Sigma N_i$
 $T = (\Sigma N_i) / (\Sigma \lambda_i) \rightarrow T = N/\lambda$



M/M/1 Queue

♦ M/M/1

» Poisson arrival, exponential service time



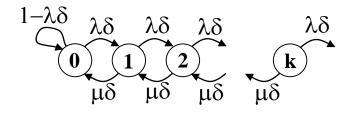
Modeled by a Markov Chain

- » State **k** k clients in the queue
- p(i,j) probability of transition from state i to state j
- » When $\delta \rightarrow 0$

$$p(i, i+1) = \lambda \delta \qquad p(i, i-1) = \mu \delta$$

$$p(i, i) = 1 - \lambda \delta - \mu \delta \qquad p(0, 0) = 1 - \lambda \delta$$

$$p(i, j) = 0 \text{ for other values } i, j$$



» Birth-death chain

- Transitions between adjacent states
- $-\lambda\delta$ and $\mu\delta$ become flow rates between states

$$p(i,i+1) = p_1(\delta) = (\lambda \delta)e^{-\lambda \delta} \approx \lambda \delta$$
$$p(0,0) = p_0(\delta) = e^{-\lambda \delta} \approx 1 - \lambda \delta$$

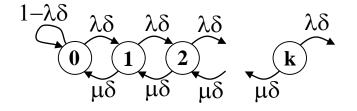
M/M/1 Queue – Equilibrium Analysis

- P(j) probability of the Markov chain be in state j
- Markov Chain global balance equations

$$P(j)\sum_{i=0}^{\infty} p(j,i) = \sum_{i=0}^{\infty} P(i)p(i,j)$$

$$i \neq j$$

$$i \neq j$$



◆ In the case of M/M/1

$$P(0)\lambda\delta = P(1)\mu\delta \implies P(1) = \rho P(0)$$

$$\sum_{i=0}^{\infty} P(i) = 1$$

$$P(2) = \rho P(1) = \rho^{2} P(0)$$

$$\sum_{i=0}^{\infty} \rho^{i} P(0) = \frac{P(0)}{1 - \rho} = 1$$

$$P(n) = \rho^{n} P(0)$$

$$P(0) = 1 - \rho$$

M/M/1 Queue

Average Queue size N

$$N = \sum_{n=0}^{\infty} nP(n) = \sum_{n=0}^{\infty} n\rho^{n} (1-\rho) = \frac{\rho}{1-\rho} \qquad N = \sum_{n=0}^{\infty} nP(n) = \frac{\rho}{1-\rho} = \frac{\lambda/\mu}{1-\lambda/\mu} = \frac{\lambda}{\mu-\lambda}$$

- Average amount of time the client spends in the system, T
 - » Little's formula, $T=N/\lambda$ \rightarrow $T=\frac{1}{\mu-\lambda}$
- Average waiting time $T_w \rightarrow T_w = T T_s = \frac{1}{\mu \lambda} \frac{1}{\mu} = \frac{\rho}{\mu(1 \rho)}$
- Average number of clients waiting in the queue, N_w

$$N_w = T_w \lambda = \frac{\lambda}{\mu - \lambda} - \frac{\lambda}{\mu} = N - \rho$$

M/M/1 Queue $-N=f(\rho)$

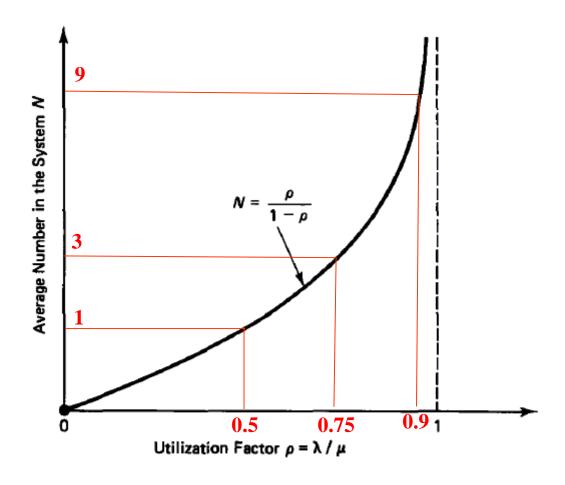


Figure 3.6 The average number in the system versus the utilization factor in the M/M/1 system. As $\rho \to 1$, $N \to \infty$.

• M/M/1: $\rho = 0.9 \rightarrow N = 9$

• Why have clients to wait if the server is busy only 90% of his time?

• What would happen for D/D/1, ρ =0.9?

Packet Length, Service Time, Speed

- » 100 packet/s are required to be transmitted through a link
- » Packets arrive according to a Poisson process
- » Packet lengths are exponentially distributed \rightarrow E[L]=10⁴ bit/packet
- » Link has capacity C=10 Mbit/s

• Then

- » Arrival rate: $\lambda = 100$ packet/s
- » Service rate: $\mu = C/E[L] = 10^7/10^4 = 10^3 \text{ packet/s}$
- » $\rho = \lambda/\mu = 0.1$, $N = \rho/(1-\rho) = 1/9$, $T = N/\lambda = 1/900$ s
- Assume now: $\lambda'=10\lambda$ and C'=10C $\rightarrow \mu'=10C/E[L]=10\mu$
 - » Then $\rho'=\rho$ and N'=N but T'=N'/ λ '=T/10 The speed of the system increases!

M/M/1/B Queue

- ♦ M/M/1 queue has limited capacity (B buffers)
 - » Packets can be lost
 - » Probability of packet being lost = P(B) \rightarrow Queue is full
- Analysis similar to M/M/1

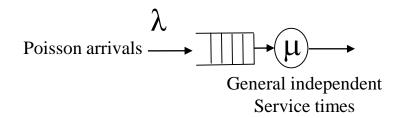
$$\sum_{i=0}^{B} P(i) = 1 P(n) = \rho^{n} P(0)$$

$$P(0) = \frac{1 - \rho}{1 - \rho^{B+1}} \qquad P(B) = \frac{(1 - \rho)\rho^{B}}{1 - \rho^{B+1}}$$

Particular cases

$$\rho = 1, \quad P(B) = \frac{1}{B+1}$$
 $\rho >> 1, \quad P(B) \approx \frac{\rho - 1}{\rho} = \frac{\lambda - \mu}{\lambda}$

M/G/1 Queue



- Poisson arrivals at rate λ
- Service time X has arbitrary distribution with given E[X] and $E[X^2]$
 - » Service times Independent and Identically Distributed (IID)
 - » Independent of arrival times
 - » $E[service time] = E[X] = 1/\mu$
 - » Single Server queue

M/G/1 Queue — Pollaczek-Khinchin (P-K) Formula

$$T_{w} = \frac{\lambda E[X^{2}]}{2(1-\rho)}$$

- where $\rho = \lambda/\mu = \lambda E[X] = line utilization$
- From Little's Theorem

$$\begin{array}{ll} \gg & N_w = \lambda T_w \\ \\ \gg & T = T_w + E[X] = T_w + 1/\mu \end{array}$$

$$\rightarrow$$
 $\mathbf{N} = \lambda \mathbf{T} = \lambda (\mathbf{T}_{w} + 1/\mu) = \mathbf{N}_{w} + \boldsymbol{\rho}$

M/G/1 Queue – Proof of (P-K) Formula

$$T_{w} = \frac{\lambda E[X^{2}]}{2(1-\rho)}$$

• Let

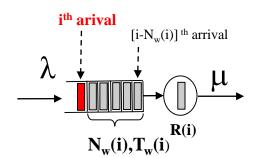
- T_w(i) waiting time in queue of ith arrival
- R(i) residual service time seen by the ith arrival
- $-N_{\rm w}(i)$ number of clients found in queue by the ith arrival
- -X(i) service time of the i^{th} arrival

$$T_{w}(i) = \sum_{j=i-N_{w}(i)}^{i-1} X(j) + R(i)$$

$$E[T_{w}(i)] = T_{w} = E[N_{w}(i)] \times E[X(i)] + E[R(i)] = \frac{N_{w}}{\mu} + E[R(i)]$$

» Using Little's formula

$$T_{w} = \frac{\lambda T_{w}}{\mu} + E[R(i)] \qquad T_{w} = \frac{E[R(i)]}{1 - \rho}$$



M/G/1 Queue – Proof of (P-K) Formula

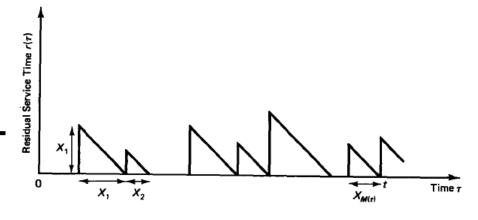


Figure 3.10 Derivation of the mean residual service time. During period [0,t], the time average of the residual service time $r(\tau)$ is

M(t) – number of clients served by time t

$$E[R(i)] = R_t = \frac{1}{t} \int_{0}^{t} r(\tau) \partial \tau = \frac{1}{t} \sum_{i=1}^{M(t)} \frac{X_i^2}{2} = \frac{M(t)}{2t} \sum_{i=1}^{M(t)} \frac{X_i^2}{M(t)}$$

$$t \to \infty$$
, $\frac{M(t)}{t} = \lambda$ = arrival rate= departure rate

$$E[R(i)] = \frac{\lambda}{2} \sum_{i=1}^{M(t)} \frac{X_i^2}{M(t)} = \frac{\lambda}{2} \times E[X^2]$$

$$T_w = \frac{E[R(i)]}{1 - \rho}$$

$$T_{w} = \frac{\lambda E[X^{2}]}{2(1-\rho)}$$

M/G/1 Examples

◆ Case M/M/1

»
$$E[X]=1/\mu$$
; $E[X^2]=2/\mu^2$

$$T_{w} = \frac{\lambda}{\mu^{2}(1-\rho)} = \frac{\rho}{\mu(1-\rho)}$$

- ◆ Case M/D/1
 - » Deterministic, constant service time $1/\mu$
 - » $E[X] = 1/\mu$; $E[X^2] = 1/\mu^2$

$$T_{w} = \frac{\lambda}{2\mu^{2}(1-\rho)} = \frac{\rho}{2\mu(1-\rho)}$$

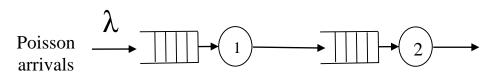
Poisson arrivals
$$\lambda$$

◆ Assume Queue 1 is M/D/1.

• Can the arrival of packets to Queue 2 be described as a Poisson process?

Networks of Transmission Lines - Problems

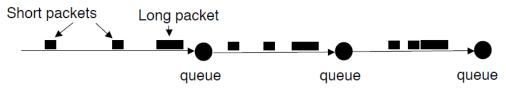
• Case 1



- » Arrival to $Q_1 \rightarrow Poisson, \lambda$
- » Assume contant packet length → $Q_1 = M/D/1$
- » Arrival to Q_2 is not Poisson; $\lambda_2 < \mu_2 \rightarrow 1/\lambda_2 > 1/\mu_2$
 - \rightarrow no waiting at Q_2

• Case 2

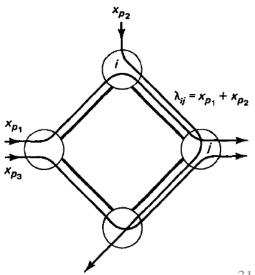
 \rightarrow $Q_1 = M/M/1$



- \rightarrow arrival to Q_2 strongly related to packet length
- » long packets require long service at each node
- » shorter packets will catch up long packets → interarrival times change
- \rightarrow Q₂ cannot be modeled as M/M/1

Kleinrock Independence Approximation

- Merging several packet streams on a transmission line restores independence of interarrival times and packet lengths
- M/M/1 can be used to model each communication link
- Approximation good for
 - » systems involving Poisson stream arrivals at the entry points
 - » packet lengths nearly exponentially distributed
 - » densely connected networks
 - » Moderate to heavy traffic loads



Kleinrock Independence Approximation

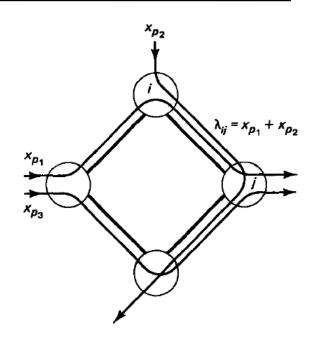
• Let

- $x_p =$ arrival rate of packets along path p
- » λ_{ij} = arrival rate of packets to link (i,j)
- » μ_{ii} = service rate on link (i,j)
- Link queues → <u>independent M/M/1 queues</u>

$$\lambda_{ij} = \sum_{\substack{\text{all p traversing} \\ \text{link(ij)}}} x_p \qquad \rho_{ij} = \frac{\lambda_{ij}}{\mu_{ij}} \qquad N_{ij} = \frac{\rho_{ij}}{1 - \rho_{ij}}$$

$$\rho_{ij} = \frac{\lambda_{ij}}{\mu_{ij}}$$

$$N_{ij} = \frac{\rho_{ij}}{1 - \rho_{ij}}$$



And

- N= Average number of packets in network
- T Average packet delay in network

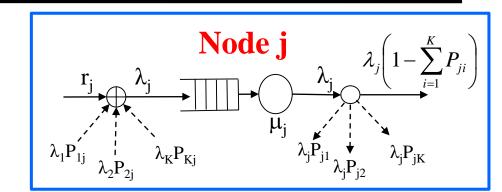
$$N = \sum_{i,j} N_{ij}$$
 $\lambda = \sum_{\text{all pathsp}} x_p = \text{total external arrival rate}$

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

Jackson Networks

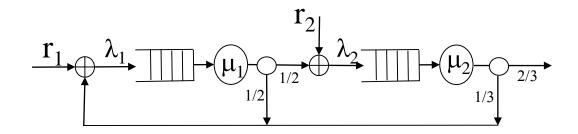
• Arrival rate at node j

$$\lambda_{j} = r_{j} + \sum_{i=1}^{K} \lambda_{i} P_{ij}$$
 , $j = 1, 2, ..., K$



- Independent routing of packets
 - » When a packet leaves node i it comes to node j with probability P_{ij}
 - » Packets can loop inside network
 - » Packet leaves the system at node j with probability

$$P = 1 - \sum_{i=1}^{K} P_{ji}$$

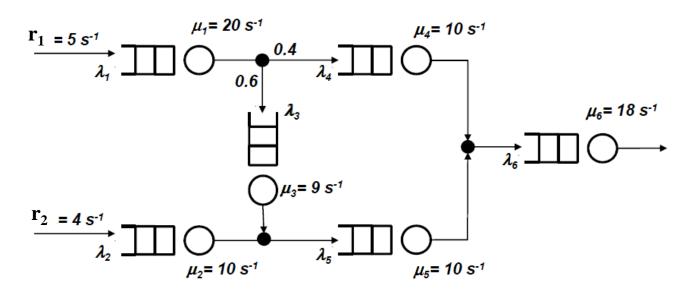


Jackson Networks

- Let the state of the system be defined by $\vec{n} = (n_1, n_2, ..., n_K)$ n_j – number of clients in Q_j
- Jackson's theorem: $P(\vec{n}) = \prod_{j=1}^{K} P_j(n_j) = \prod_{j=1}^{K} \rho_j^{n_j} (1 \rho_j)$, where $\rho_j = \frac{\lambda_j}{\mu_j}$
 - » State of $Q_j(n_j)$ is independent $\left(\prod_{j=1}^{\kappa}\right)$ of state of other queues
 - » Similar to independent M/M/1 queues!
 - » Similar to Kleinrock's independence
- Again, by Little's theorem

$$N_{j} = \frac{\rho_{j}}{1 - \rho_{i}} \qquad N = \sum_{i=1}^{K} N_{j} \qquad \lambda = \sum_{j=1}^{K} r_{j} \qquad T = \frac{N}{\lambda}$$

Jackson Network - Example



$$\lambda = \sum_{i=1}^{6} r_i = 9 \, \mathrm{s}^{-1}$$

$$N = \sum_{i=1}^{6} N_i = 5.08$$

$$T = \frac{N}{\lambda} = \frac{5.08}{9} = 0.56 \,\mathrm{s}$$

Queue i	$\mathbf{r_i} \ \left(s^{-1}\right)$	$\lambda_i \ \left(s^{-1}\right)$	μ_i (s^{-1})	$\rho_i = \lambda_i/\mu_i$	$\mathbf{N_i} = \rho_i/(1-\rho_i)$
1	5	5	20	0.25	0.33
2	4	4	10	0.40	0.67
3	-	3	9	0.33	0.50
4	-	2	10	0.20	0.25
5	-	7	10	0.70	2.33
6	-	9	18	0.50	1

Homework

- 1. Review slides
- 2. Read Bertsekas&Gallager
 - » Sections 3.1, 3.2, 3.3, 3.5, 3.6, 3.8
- 3. Answer questions at moodle