
Lecture 7:

Burke's Theorem and Networks of Queues

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Burke's Theorem

- An interesting property of an M/M/1 queue, which greatly simplifies combining these queues into a network, is the surprising fact that the output of an M/M/1 queue with arrival rate λ is a Poisson process of rate λ
 - This is part of Burke's theorem, which follows from reversibility

- A Markov chain has the property that

- $P[\text{future} \mid \text{present, past}] = P[\text{future} \mid \text{present}]$

Conditional on the present state, future states and past states are independent

$$P[\text{past} \mid \text{present, future}] = P[\text{past} \mid \text{present}]$$

$$\Rightarrow P[X_n=j \mid X_{n+1}=i, X_{n+2}=i_2, \dots] = P[X_n=j \mid X_{n+1}=i] = P^*_{ij}$$

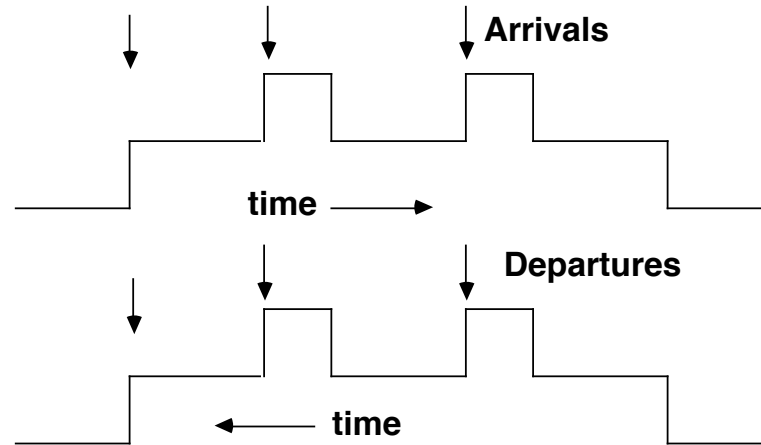
Burke's Theorem (continued)

- The state sequence, run backward in time, in steady state, is a Markov chain again and it can be easily shown that

$$p_i P_{ij}^* = p_j P_{ji} \qquad \text{e.g., M/M/1 } (p_n)\lambda = (p_{n+1})\mu$$

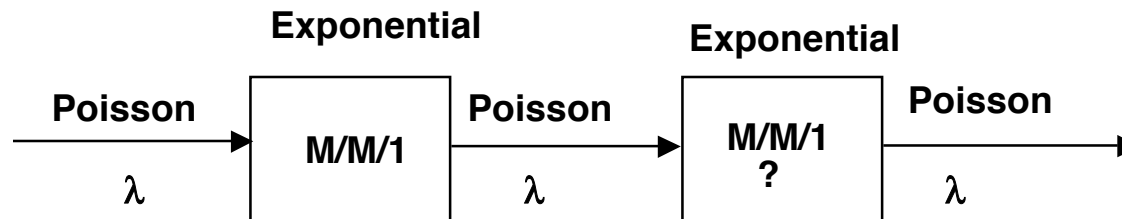
- A Markov chain is reversible if $P_{ij}^* = P_{ji}$
 - Forward transition probabilities are the same as the backward probabilities
 - If reversible, a sequence of states run backwards in time is statistically indistinguishable from a sequence run forward
- A chain is reversible iff $p_i P_{ij} = p_j P_{ji}$
- All birth/death processes are reversible
 - Detailed balance equations must be satisfied

Implications of Burke's Theorem



- **Since the arrivals in forward time form a Poisson process, the departures in backward time form a Poisson process**
- **Since the backward process is statistically the same as the forward process, the (forward) departure process is Poisson**
- **By the same type of argument, the state (packets in system) left by a (forward) departure is independent of the past departures**
 - **In backward process the state is independent of future arrivals**

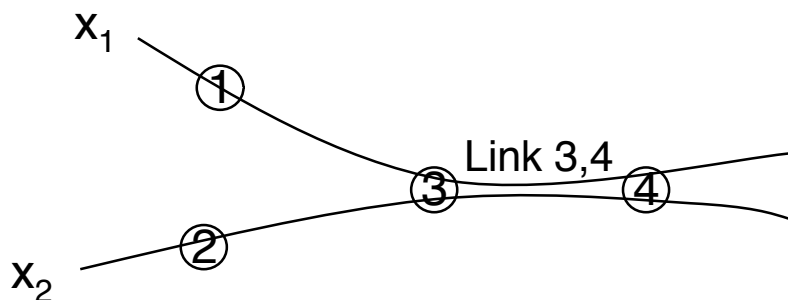
NETWORKS OF QUEUES



- The output process from an M/M/1 queue is a Poisson process of the same rate λ as the input
- Is the second queue M/M/1?

Independence Approximation (Kleinrock)

- Assume that service times are independent from queue to queue
 - Not a realistic assumption: the service time of a packet is determined by its length, which doesn't change from queue to queue



- X_p = arrival rate of packets along path p
- Let λ_{ij} = arrival rate of packets to link (i,j)
- μ_{ij} = service rate on link (i,j)

$$\lambda_{ij} = \sum_{P \text{ traverses link } (i,j)} X_p$$

Kleinrock approximation

- Assume all queues behave as independent M/M/1 queues

- $$N_{ij} = \frac{\lambda_{ij}}{\mu_{ij} - \lambda_{ij}}$$

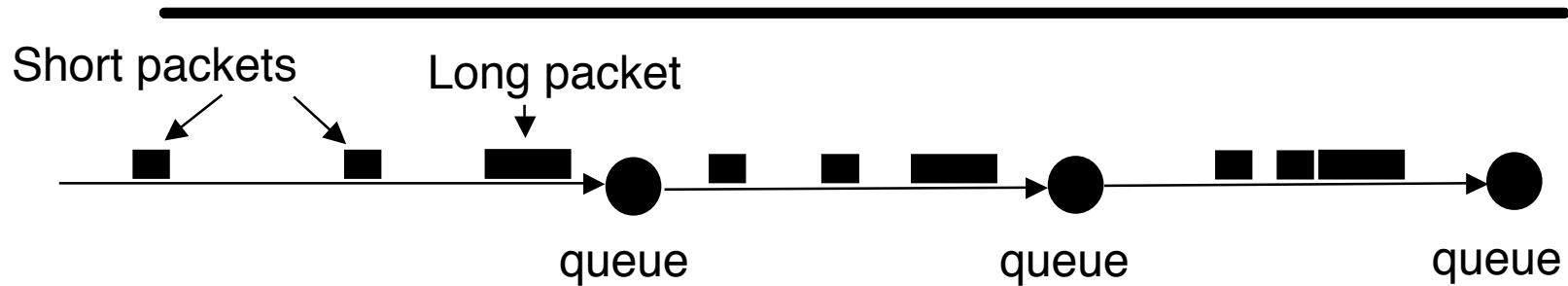
- N = Ave. packets in network, T = Ave. packet delay in network

$$N = \sum_{i,j} N_{ij} = \frac{\lambda_{ij}}{\mu_{ij} - \lambda_{ij}}, \quad T = \frac{N}{\lambda}$$

$$\lambda = \sum_{\text{all paths } p} X_p = \text{total external arrival rate}$$

- Approximation is not always good, but is useful when accuracy of prediction is not critical
 - Relative performance but not actual performance matters
 - E.g., topology design

Slow truck effect



- **Example of bunching from slow truck effect**
 - long packets require long service at each node
 - Shorter packets catch up with the long packets
- **Similar to phenomenon that we experience on the roads**
 - Slow car is followed by many faster cars because they catch up with it

Jackson Networks

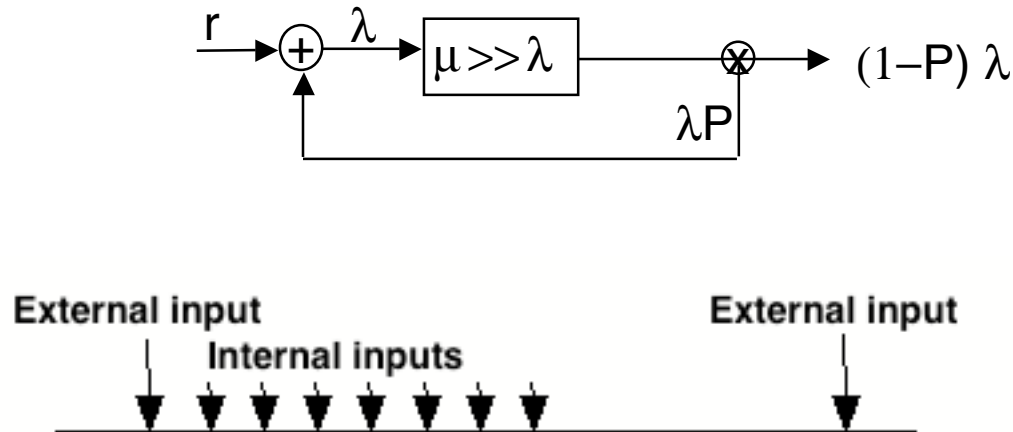
- **Independent external Poisson arrivals**
- **Independent Exponential service times**
 - Same job has independent service time at different queues
- **Independent routing of packets**
 - When a packet leaves node i it goes to node j with probability P_{ij}
 - Packet leaves system with probability $1 - \sum_j P_{ij}$
 - Packets can loop inside network
- **Arrival rate at node i ,**

$$\lambda_i = r_i + \sum_k \lambda_k P_{ki}$$

External arrivals Internal arrivals from Other nodes

- Set of equations can be solve to obtain unique λ_i 's
- Service rate at node $i = \mu_i$

Jackson Network (continued)



- **Customers are processed fast ($\mu \gg \lambda$)**
- **Customers exit with probability $(1-P)$**
 - Customers return to queue with probability P
 - $\lambda = r + P\lambda \Rightarrow \lambda = r/(1-P)$
- **When P is large, each external arrival is followed by a burst of internal arrivals**
 - Arrivals to queues are not Poisson

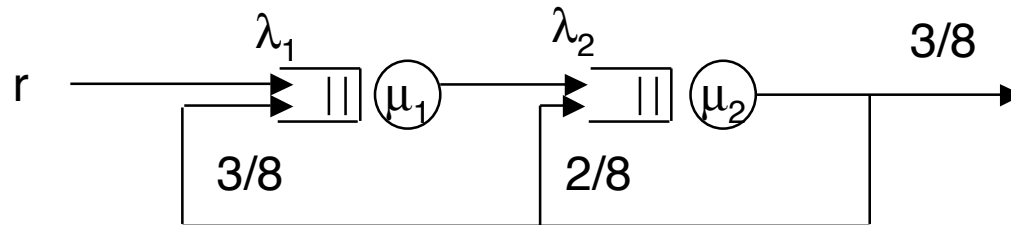
Jackson's Theorem

- We define the state of the system to be $\vec{n} = (n_1, n_2 \cdots n_k)$ where n_i is the number of customers at node i
- Jackson's theorem:

$$P(\vec{n}) = \prod_{i=1}^{i=k} P_i(n_i) = \prod_{i=1}^{i=k} \rho_i^{n_i} (1 - \rho_i), \quad \text{where } \rho_i = \frac{\lambda_i}{\mu_i}$$

- That is, in steady state the state of node i (n_i) is independent of the states of all other nodes (at a given time)
 - Independent M/M/1 queues
 - Surprising result given that arrivals to each queue are neither Poisson nor independent
 - Similar to Kleinrock's independence approximation
 - Reversibility
 - Exogenous outputs are independent and Poisson
 - The state of the entire system is independent of past exogenous departures

Example



$$\lambda_1 = ?$$

$$\lambda_2 = ?$$

$$P(n_1, n_2) = ?$$