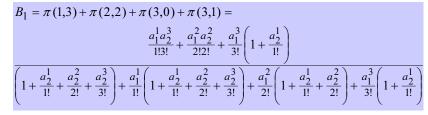


10. Network models

10. Network models

Example

- Consider the example presented in slide 9 (and continued in slide 11)
- The end-to-end blocking probability B_1 for class 1 will be



Product Bound (1)

- link *j* could be modelled as a loss system where new calls arrive according

 $\lambda(j) = \sum_{r \in R(j)} \lambda_r$

- In this case, the blocking probability could be calculated from formula

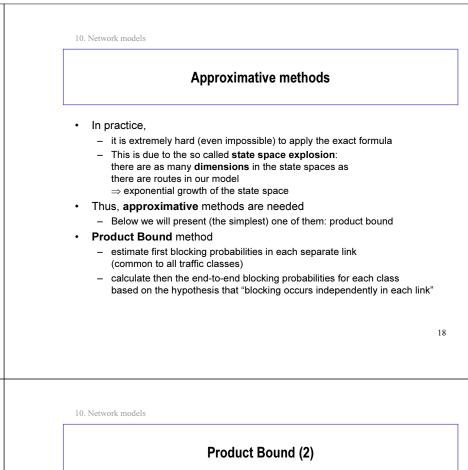
 $B(j) \approx \operatorname{Erl}(n_j, \sum_{r \in R(j)} a_r)$

• Consider first the blocking probability B(i) in an arbitrary link *i*

- Let R(j) denote the set of routes that use link j

to a Poisson process with intensity $\lambda(i)$,

• If the capacities of all the other links (but *j*) were infinite,



- Consider then the end-to-end blocking probability B_r for class r
 - Let J(r) denote the set of the links that belong to route r
 - Note that an arriving call of class r will not be blocked, if it is not blocked in any link $j \in J(r)$
- If blocking occured independently in each link,
 - an arriving call of class *r* would be blocked with probability

$$B_r \approx 1 - \prod_{j \in J(r)} (1 - B(j))$$

- Note that for small values of B(j)'s, we can use the following approximation:

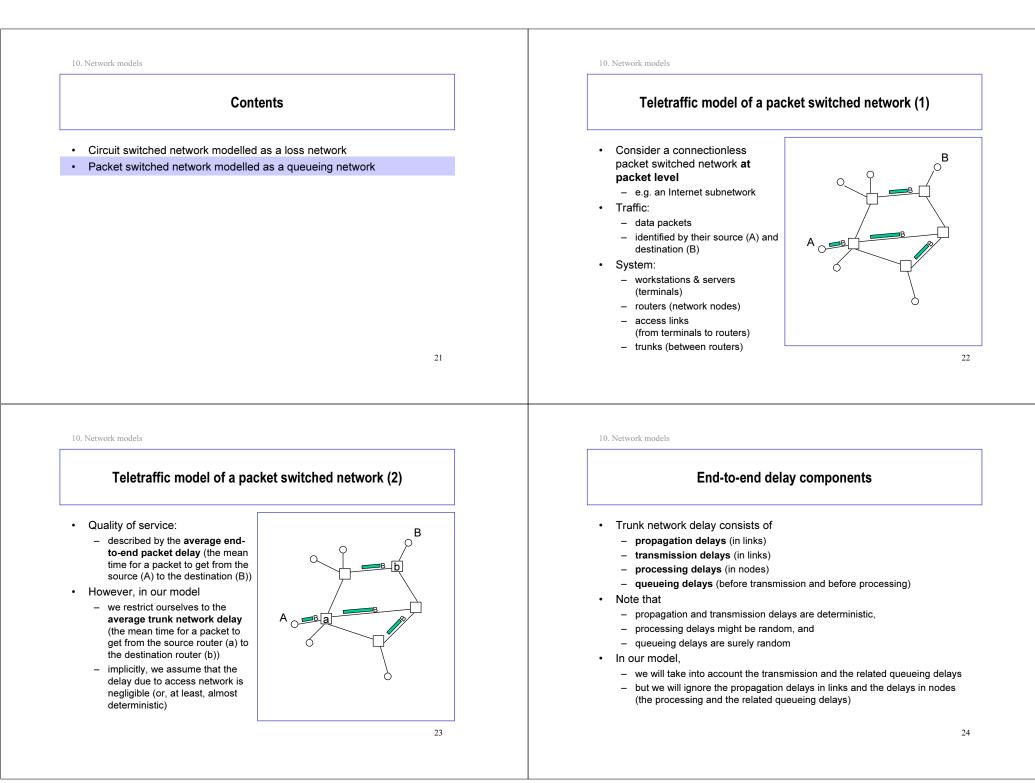
$$B_r \approx \sum_{j \in J(r)} B(j)$$

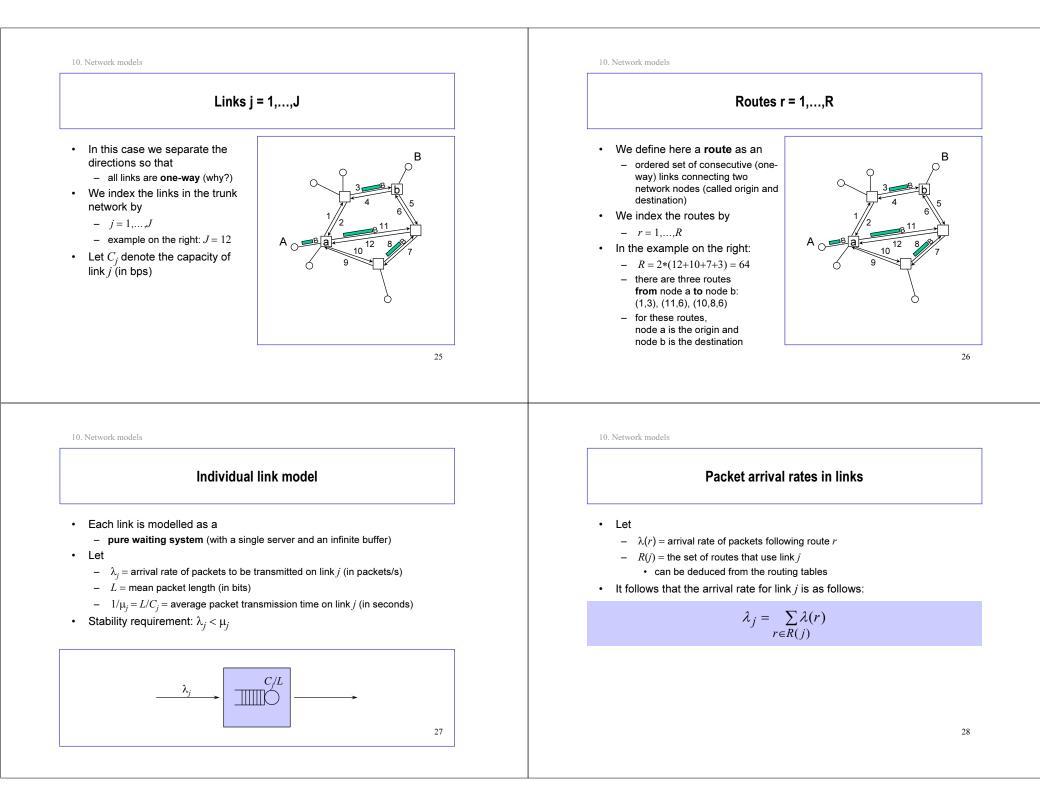
 Note that this is really an approximation, since the traffic offered to link *j* is smaller due to blockings in other links (and not even of Poisson type).

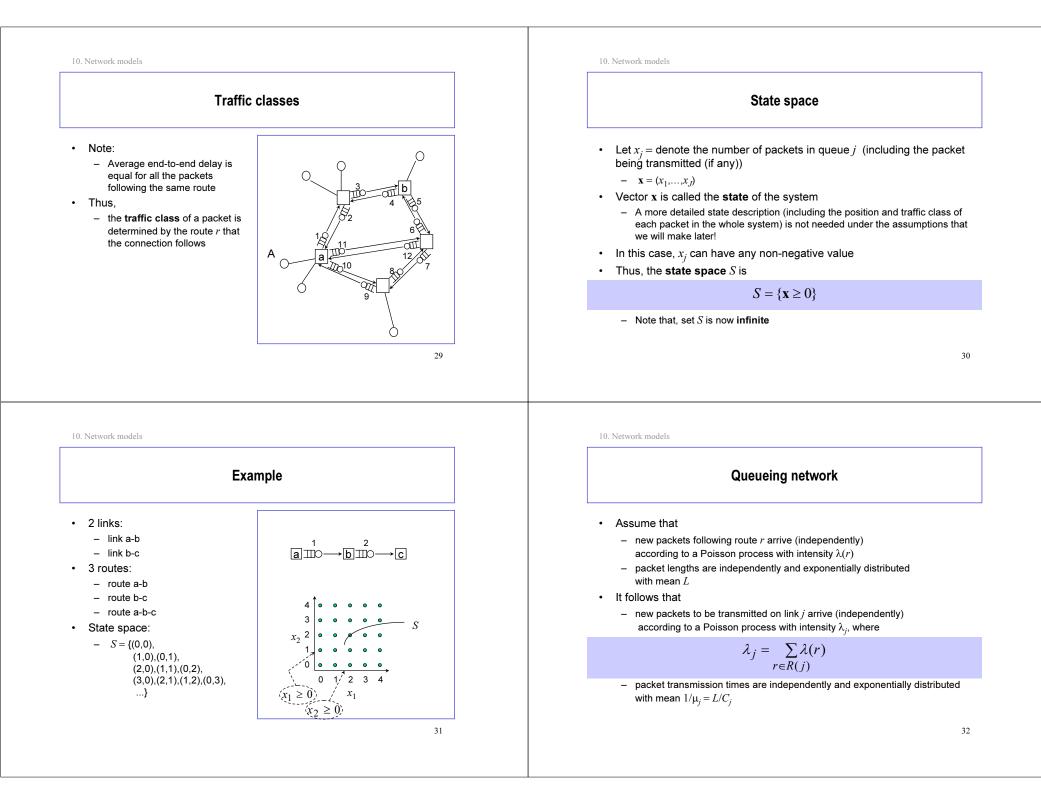
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Equilibrium distribution (1)

- Assume further that
 - the system is **stable**: $\lambda_j < \mu_j$ for all j
 - packet length is independently redrawn (from the same distribution) every time the packet moves from one link to another
 - This is so called Kleinrock's independence assumption
- Under these assumptions, it is possible to show that

 $\rho_j =$

- the stationary state probability $\pi(\mathbf{x})$ for any state $\mathbf{x} \in S$ is as follows:

$$\pi(\mathbf{x}) = \prod_{i=1}^{J} (1 - \rho_j) \rho_j^{x_j}$$

- where ρ_i denotes the traffic load of link *j*:

$$\frac{\lambda_j}{\mu_j} = \frac{\lambda_j L}{C_j} < 1$$

10. Network models

Mean end-to-end delay

- Consider then the mean end-to-end delay for class r
 - Let J(r) denote the set of the links that belong to route r
- In our model, the mean end-to-end delay will be
 - the sum of mean delays experienced in the links along the route (including **both** the transmission delay **and** the queueing delay)
- By Little's formula, the mean link delay is

$$\overline{T}_{j} = \frac{\overline{X}_{j}}{\lambda_{j}} = \frac{1}{\lambda_{j}} \cdot \frac{\rho_{j}}{1 - \rho_{j}} = \frac{1}{\mu_{j}} \cdot \frac{1}{1 - \rho_{j}} = \frac{1}{\mu_{j} - \lambda_{j}}$$

• Thus, the mean end-to-end delay for class r is

$$\overline{T}(r) = \sum_{j \in J(r)} \overline{T_j} = \sum_{j \in J(r)} \frac{1}{\mu_j (1 - \rho_j)} = \sum_{j \in J(r)} \frac{1}{\mu_j - \lambda_j}$$

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Equilibrium distribution (2)

- Probability π(x) is again said to be of product-form
 Now, the number of packets in different queues are independent (why?)
- Each individual queue *j* behaves as an M/M/1 queue

10. Network models

- Number of packets in queue *j* follows a geometric distribution with mean



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