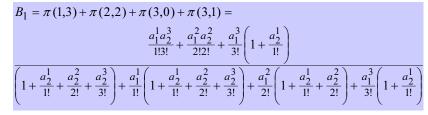


10. Network models

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



10. Network models
 Approximative methods
 In practice,

 it is extremely hard (even impossible) to apply the exact formula
 This is due to the so called state space explosion: there are as many dimensions in the state spaces as there are routes in our model
 ⇒ exponential growth of the state space

 Thus, approximative methods are needed

 Below we will present (the simplest) one of them: product bound
 Product Bound method

- estimate first blocking probabilities in each separate link (common to all traffic classes)
- calculate then the end-to-end blocking probabilities for each class based on the hypothesis that "blocking occurs independently in each link"

10. Network models

Product Bound (1)

- Consider first the blocking probability *B*(*j*) in an arbitrary link *j*
 - Let R(j) denote the set of routes that use link j
- If the capacities of all the other links (but *j*) were infinite,
 - link *j* could be modelled as a loss system where new calls arrive according to a Poisson process with intensity λ(*j*),

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

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

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

 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).

19

17

- Product Bound (2)
- 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,

10. Network models

- an arriving call of class r would be blocked with probability

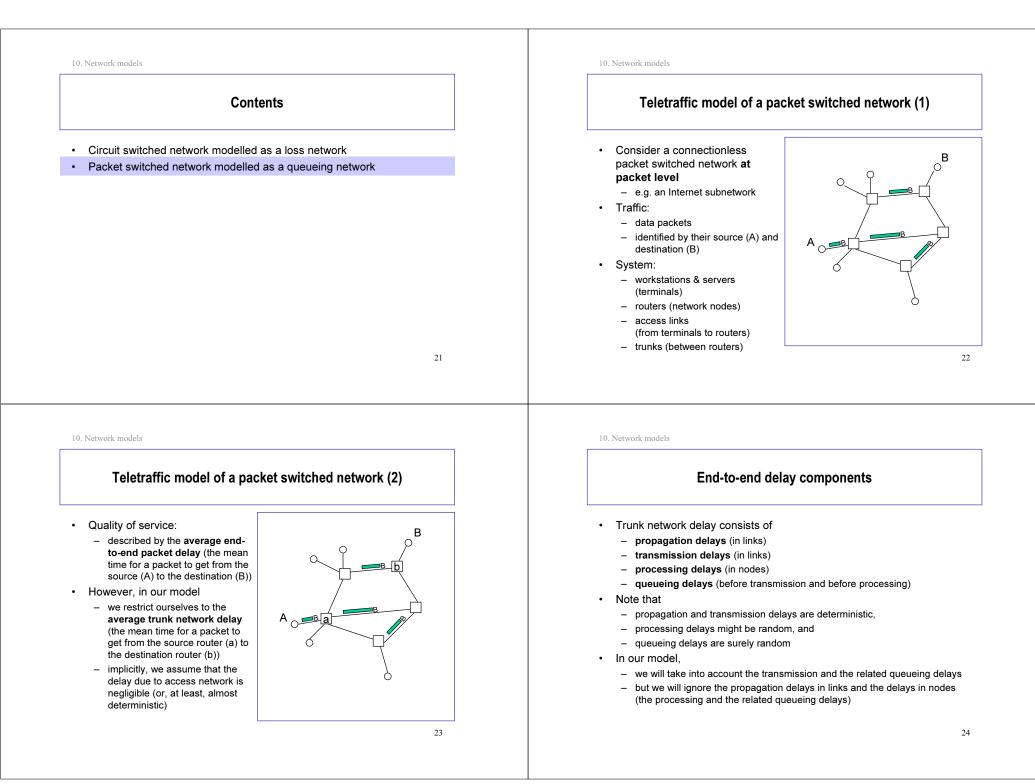
$$B_r \approx 1 - \prod_{j \in J(r)} (1 - B(j))$$

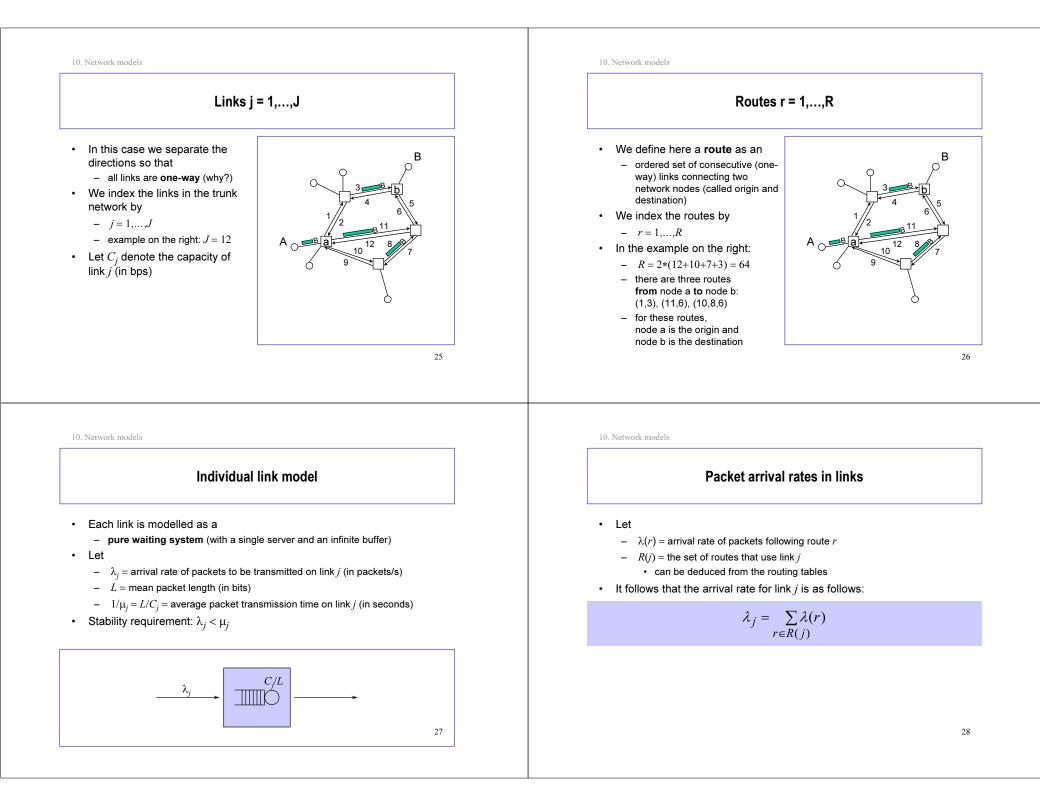
- Note that for small values of B(j)'s, we can use the following approximation:

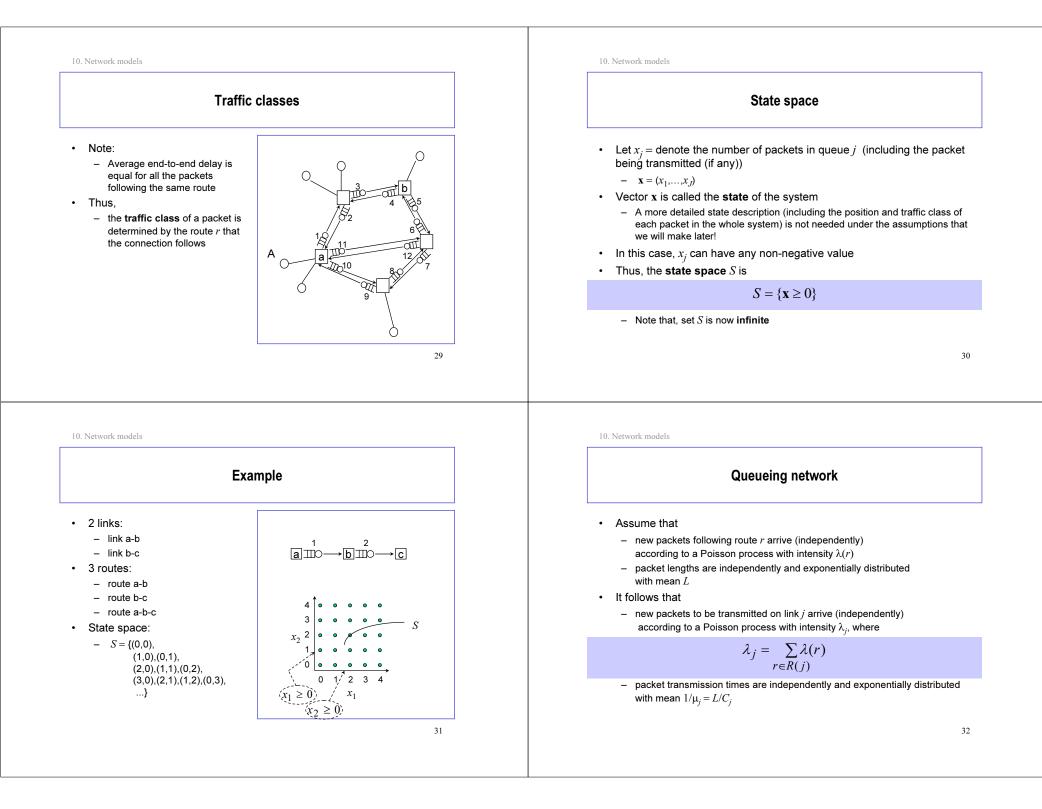
$$B_r \approx \sum_{j \in J(r)} B(j)$$

20

18









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*:

$$\frac{i}{j} = \frac{\lambda_j L}{C_j} < 1$$

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

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

33

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



34