EP2200 Queueing theory and teletraffic systems

Lecture 8 Semi-Markovian systems The Method of Stages Viktoria Fodor KTH EES/LCN

#### Semi-Markovian system

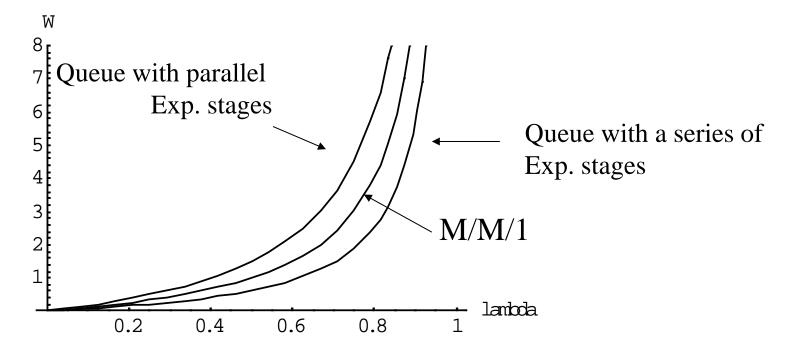
- Advantages with M/M/\*
  - The interarrival time and the service time distribution is memoryless
  - The state can be defined by the number of customers in the system since the time of the next state transition does not depend on the time already spent in the state
- Applicability for real systems
  - The arrival process is often Poisson (large number of potential customers)
  - The service process is often **not** memoryless
    - E.g., packet size distribution on the Internet, file size distribution
    - The future of the system depends on the elapsed service time

# Semi-Markovian system and Method of Stages

- Ways to handle the non-exponential service time
  - 1. Look at distributions consisting of several exponentially distributed stages in series or in parallel Method of stages (this lecture)
  - Describe the system only at specific points of time (e.g., end of service), where the Markovian property is satisfied – Semi-Markovian systems (semi ] half)

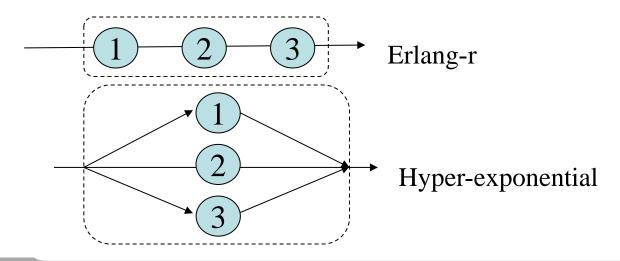
#### Method of Stages

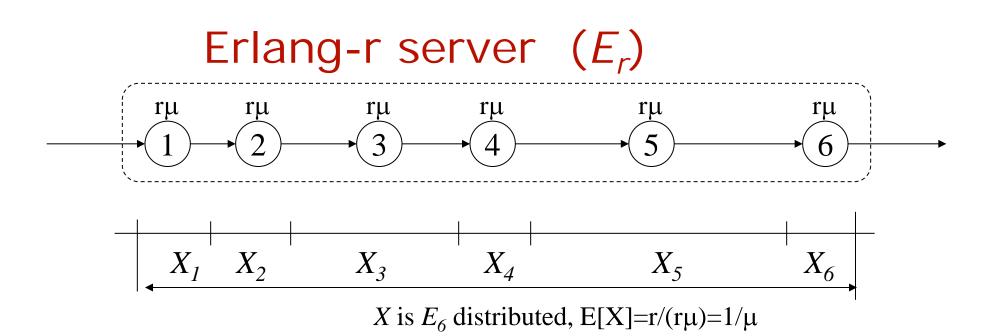
 Use distributions that are composed of Exponential distributions



#### Method of Stages

- Each service stage is Exponential
- Series of stages: the customer has to finish r service stages before the next customer can enter the server  $\rightarrow$  Erlang-r service time distribution
- Parallel (or alternative) stages: the customer selects one server randomly, but only one customer can be in the service unit → Hyper-exponential service time distribution (linear combination of Exp. distributions)





- Goal: service time with average  $E[X] = \bar{x} = 1/\mu$
- Since  $E[X] = \sum E[X_i]$  if  $X = \sum X_i$ , we select:
  - $X_i$  a stochastic variable with Exponential distribution  $b(x_i) = r\mu e^{r\mu x_i}$
  - $X = \sum_{i=1}^{r} X_i$ ,  $X_i$ ,  $X_j$  independent, identically distributed
  - That is, X is Erlang-r distributed

### Erlang-r server $(E_r)$

• For each exponential stage:

$$b(x_{i}) = r\mu e^{-r\mu x_{i}}$$

$$E[X_{i}] = \frac{1}{r\mu}$$

$$V[X_{i}] = \left(\frac{1}{r\mu}\right)^{2}$$

$$C_{x_{i}}^{2} = \frac{V[X_{i}]}{E[X_{i}]^{2}} = 1$$

X = X + X + X - X - X *iid* 

• For the service time:

$$K = R_{1} + R_{2} + \dots + R_{r}, \quad R_{i}, R_{j} + u.u.$$

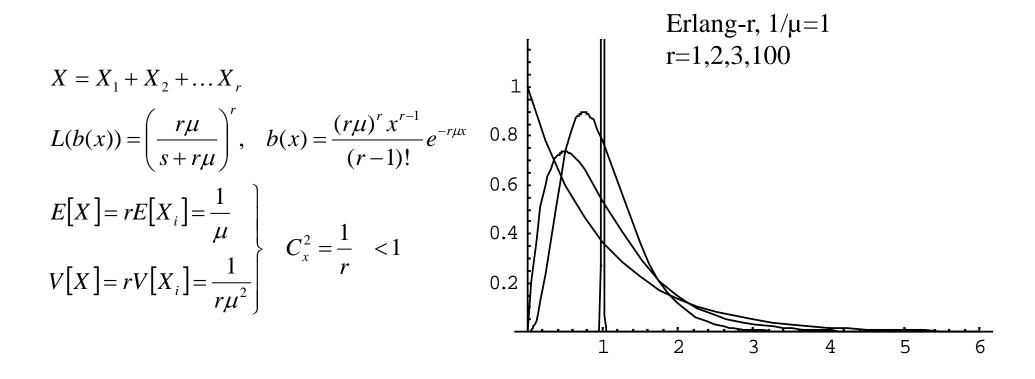
$$b(x) = \frac{(r\mu)^{r} x^{r-1}}{(r-1)!} e^{-r\mu x} \quad (Erlang - r)$$

$$E[X] = rE[X_{i}] = \frac{1}{\mu}$$

$$V[X] = rV[X_{i}] = \frac{1}{r\mu^{2}}$$

$$C_{x}^{2} = \frac{1}{r} < 1$$

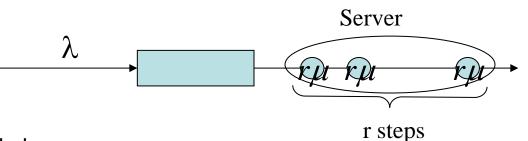
#### Erlang-r server $(E_r)$



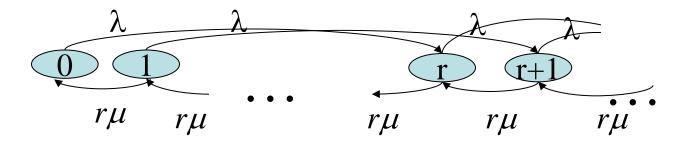
As r→∞, V[X]→0, which means deterministic service time!

## The M/E<sub>r</sub>/1- queue

• If the system to be modeled has serial service or the service distribution has  $C_x^2 < 1 - approximate$  with Erlang-r

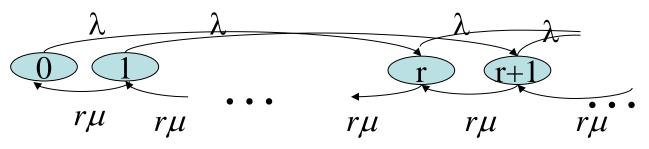


- System state:
  - {number of remaining service stages, number of customers}, or
  - number of remaining service stages + r\*number of waiting customers
- The system can be modeled as a Markov chain



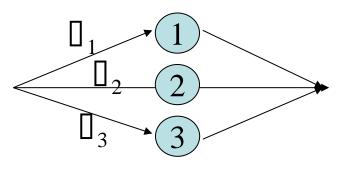
## The $M/E_r/1$ - queue

- System state:
  - number of remaining service stages + r\*number of waiting customers
- Number of customers in the system in state i:  $N_i = \lfloor i/r \rfloor$
- State probability distribution with z-transforms (Kleinrock p.127-128)
  - (not exam material)
- But, the followings hold:
  - PASTA
  - Little:  $N_s = \lambda x = \lambda / \mu = Utilization$
  - For r=1: M/M/1, for r=∞: M/D/1
  - You will have to be able to calculate state probabilities and performance measures for limited buffer systems (e.g.,  $M/E_2/1/3$ )!
  - Average performance for  $M/E_r/1$  with general forms of M/G/1



## Hyper-exponential server $(H_r)$

- r exponential servers with different µ<sub>i</sub>-s
- Server *i* is chosen with the probability  $\alpha_i$ 
  - E.g., different types of packets intermixed
  - service time distribution is the linear
     combination (mixture) of Exp distributions



$$\begin{aligned} & = \exp[\mu_1] \operatorname{Exp}[\mu_2] & \operatorname{Exp}[\mu_1] & \operatorname{Exp}[\mu_3] & \operatorname{Exp}[\mu_3] & \operatorname{Exp}[\mu_2] \\ & \text{a possible sequence of service of 6 customers} \\ & b(x_i) = \mu_i e^{-\mu_i x} \\ & b(x) = \alpha_1 \mu_1 e^{-\mu_1 x} + \ldots + \alpha_R \mu_R e^{-\mu_R x}, \quad \sum \alpha_i = 1 \end{aligned}$$

## The hyper-exponential server $(H_r)$

- r exponential servers with different  $\mu$ -s  $B(x_i) = 1 e^{-\mu_i x}$
- Server *i* is chosen with the probability  $\alpha_i$

$$b(x) = \alpha_1 \mu_1 e^{-\mu_1 x} + \ldots + \alpha_R \mu_R e^{-\mu_R x}, \quad \sum \alpha_i = 1$$

$$L(b(x)) = \sum_{i=1}^{r} \alpha_i \frac{\mu_i}{s + \mu_i}$$

$$\begin{array}{c}
 \begin{bmatrix}
 1 \\
 \end{bmatrix}_{2} \\
 \end{bmatrix}_{3} \\
 \end{bmatrix}_{3} \\
 \end{bmatrix}$$

$$E[X] = \sum_{i} \frac{\alpha_{i}}{\mu_{i}}$$

$$E[X^{2}] = \sum_{i} \alpha_{i} \frac{2}{\mu_{i}^{2}}$$

$$V[X] = E[X^{2}] - E[X]^{2}$$

$$C_{x_{i}}^{2} = \frac{V[X_{i}]}{E[X_{i}]^{2}} = \frac{E[X_{i}^{2}] - E[X_{i}]^{2}}{E[X_{i}]^{2}} = \frac{E[X_{i}^{2}]}{E[X_{i}]^{2}} - 1$$

• For given coefficient of variation 2R-1 free parameters in total

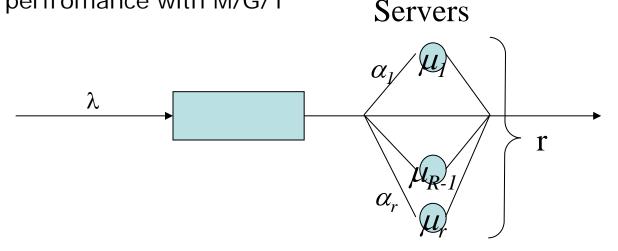
- *R*-1 of  $\alpha_i$  and *R* of  $\mu_i$ 

### The $M/H_r/1$ queue

- If there are different service needs randomy intermixed
  - E.g., packet size distribution

or if the service time distribution has  $C_x^2 > 1 - approximate$  with  $H_r$ 

- The state represents the number of customers in the system and the actual server used (only one server used at a time!)
  - complicated Markov-chain (see notes from class)
  - you have to be able to handle it for limited buffer systems
  - Little, PASTA holds
  - Average perfromance with M/G/1



# The $M/H_r/1$ queue

 Example problem: Packets of two types arrive to a multiplexer intermixed. The total arrival intensity is λ.

Packet of type 1 arrives with probability  $\alpha_{l}$ , its transmission time is exponential with parameter  $\mu_{l}$ .

Packet of type 2 arrives with probability  $\alpha_2$ , its transmission time is exponential with parameter  $\mu_2$ .

There is no buffer.

- Give:
  - Kendall, Markov-chain
  - Balance equations (no need to calculate the sate probabilities...)
  - P(packet type 1 under transmission)
  - P(packet blocked)
  - Utilization

#### Method of stages for the arrival process

- Non-exponential inter-arrival times can be modeled similarly
- E.g., round-robin customer spreading: E<sub>r</sub>/M/1

#### Semi-markovian system Method of stages - Summary

- Ways to handle the non-exponential service / inter-arrival time
  - Method of stages: look at distributions consisting of several exponentially distributed stages in series or in parallel
  - Describe the system in specific points of time (end of service) M/G/1, embedded Markov-chains
- Erlang-r service / inter-arrival times
  - series of stages in the real system, or
  - has distribution with  $C_x^2 < 1$
  - can be modeled with Markov-chain
     state: r\*number of customers + number of stages left from service
- Hyper-exponential service /inter-arrival times
  - parallel stages in the real system
  - has distribution with  $C_x^2 > 1$
  - can be modeled with Markov chain state has 2 parameters: number of customers and server used