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
- 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
- Ways to handle the non-exponential service time semi-markovian systems (semi-markov: MC to describe prossible state transitions, but the holding times are not exponential)
  - 1. Look at distributions consisting of several exponentially distributed stages in series or in parallel
  - Describe the system only at specific points of time (e.g., end of service) – M/G/1

# Semi-markovian systems – method of stages

 Use distributions that are composed of Exponential distributions



#### The 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)



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- 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:

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

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

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

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

• For the service time:

$$X = X_{1} + X_{2} + \dots X_{r}$$
  

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

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

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

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

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



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

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## 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



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

## 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  $\alpha_i$

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

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

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

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

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

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

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

- For given coefficient of variation 2R-1 free parameters in total
  - *R*-1 of  $\alpha_i$  and *R* of  $\mu_i$

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### 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
  - state probabilities (balance equations)
  - 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: number of customers time r plus 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: number of customers and server used