

## Lecture 19 Chapter 8 Queueing Models

Let

$X(t)$  = # of customers in the system at time  $t$

$Q(t)$  = # of customers waiting in queue at time  $t$

Assume that  $\{X(t), t \geq 0\}$  and  $\{Q(t), t \geq 0\}$  has a stationary distribution.

$$L = \lim_{t \rightarrow \infty} \frac{\int_0^t X(t) dt}{t} = \text{the average \# of customers in the system}$$

$$L_Q = \lim_{t \rightarrow \infty} \frac{\int_0^t Q(t) dt}{t} = \text{the average \# of customers waiting in queue}$$

$W$  = the average amount of time, including waiting time and service time, a customer spends in the system;

$W_Q$  = the average amount of time a customer waiting in queue.

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### Cost Identity

Many interesting and useful relationships between quantities in Queueing models can be obtained by using the **cost identity**.

Imagine that entering customers are forced to pay money (according to some rule) to the system. We would then have the following basic cost identity:

$$\begin{aligned} & \text{average rate at which the system earns} \\ &= \lambda_a \times \text{average amount an entering customer pays} \end{aligned}$$

*Proof.* Let  $R(t)$  be the amount of money the system has earned by time  $t$ . Then we have

$$\begin{aligned} & \text{average rate at which the system earns} \\ &= \lim_{t \rightarrow \infty} \frac{R(t)}{t} = \lim_{t \rightarrow \infty} \frac{N(t) R(t)}{t N(t)} = \lambda_a \lim_{t \rightarrow \infty} \frac{R(t)}{N(t)} \\ &= \lambda_a \times \text{average amount an entering customer pays,} \end{aligned}$$

provided that the limits exist.

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### 8.3.1 M/M/1 Model

Let  $X(t)$  be number of customers in the system at time  $t$ .  $\{X(t), t \geq 0\}$  is a birth and death process with

$$\text{birth rates } \lambda_j \equiv \lambda, \quad \text{and death rates } \mu_j \equiv \mu.$$

Recall in Example 6.14 we have showed that the stationary distribution exists when  $\lambda < \mu$ , and the stationary distribution is

$$P_n = \lim_{t \rightarrow \infty} P(X(t) = n) = \left(1 - \frac{\lambda}{\mu}\right) \left(\frac{\lambda}{\mu}\right)^n, \quad n = 0, 1, \dots$$

Thus

$$\begin{aligned} L &= \lim_{t \rightarrow \infty} \sum_{n=1}^{\infty} n P(X(t) = n) = \sum_{n=1}^{\infty} n P_n = \frac{\lambda}{\mu - \lambda} \\ &= \frac{1/\mu}{1/\lambda - 1/\mu} \end{aligned}$$

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## Little's Formula

Let

$N(t)$  = # of customers enter the system at or before time  $t$ .

We define  $\lambda_a$  be the arrival rate of entering customers,

$$\lambda_a = \lim_{t \rightarrow \infty} \frac{N(t)}{t}$$

**Little's Formula:**

$$L = \lambda_a W$$

$$L_Q = \lambda_a W_Q$$

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### Proof of Little's Formula

To prove  $L = \lambda_a W$ :

- ▶ we use the payment rule:

each customer pays \$1 per unit time while in the system.

- ▶ the average amount a customer pay =  $W$ , the average waiting time of customers.
- ▶ the amount of money the system earns during the time interval  $(t, t + \Delta t)$  is  $X(t)\Delta t$ , where  $X(t)$  is the number of customers in the system at time  $t$ ,
- ▶ and the rate the system earns is thus  $\lim_{t \rightarrow \infty} \frac{\int_0^t X(s) ds}{t} = L$ , the formula follows from the cost identity.

To prove  $L_Q = \lambda_a W_Q$ , we use the payment rule:

each customer pays \$1 per unit time while in queue.

The argument is similar.

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### 8.3.1 M/M/1 Model (Cont'd)

Let  $T$  be the time of a customer spend in the system.

If there are  $n$  customers in the system while this customer arrives, then  $T$  is the sum of the service times of the  $n + 1$  customers  $\sim \text{Gamma}(n + 1, \mu)$ . That is,

$$\begin{aligned} P(T \leq t) &= \sum_{n=0}^{\infty} P_n \int_0^t \frac{\mu^{n+1}}{n!} s^n e^{-\mu s} ds \\ &= \sum_{n=0}^{\infty} \left(1 - \frac{\lambda}{\mu}\right) \left(\frac{\lambda}{\mu}\right)^n \int_0^t \frac{\mu^{n+1}}{n!} s^n e^{-\mu s} ds \\ &= (\mu - \lambda) \int_0^t \left(\underbrace{\sum_{n=0}^{\infty} \frac{(\lambda s)^n}{n!}}_{=e^{\lambda s}}\right) e^{-\mu s} ds \\ &= (\mu - \lambda) \int_0^t e^{-(\mu - \lambda)s} ds = 1 - e^{-(\mu - \lambda)t} \end{aligned}$$

Therefore,  $T \sim \text{Exp}(\mu - \lambda) \Rightarrow W = \mathbb{E}[T] = \frac{1}{\mu - \lambda}$ .

This verifies Little's formula,  $L = \lambda W$ .

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### 8.3.1 M/M/1 Model (Cont'd)

$$W_Q = W - \mathbb{E}[\text{service time}] = W - 1/\mu = \frac{\lambda}{\mu(\mu - \lambda)}$$

$$L_Q = \lambda W_Q = \frac{\lambda^2}{\mu(\mu - \lambda)}$$

**Example 8.2** Suppose customers arrive at a Poisson rate of 1 in 12 minutes, and that the service time is exponential at a rate of one service per 8 minutes. What are  $L$  and  $W$ ?

*Solution.* Since  $\lambda = 1/12$ ,  $\mu = 1/8$ , we have

$$L = \frac{1/\mu}{1/\lambda - 1/\mu} = \frac{8}{12 - 8} = 2, \quad W = \frac{1}{\mu - \lambda} = 24$$

Observe if the arrival rate increases 20% to  $\lambda = 1/10$ , then

$$L = 4, \quad W = 40$$

When  $\lambda/\mu \approx 1$ , a slight increase in  $\lambda/\mu$  will lead to a large increase in  $L$  and  $W$ .

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### M/M/∞ Model

In this case, customers will be served immediately upon arrival. Nobody will be in queue. We have

$$W_Q = L_Q = 0, \quad W = \text{average service time} = 1/\mu,$$

and hence  $L = \lambda W = \lambda/\mu$ .

As a verification, observe that  $\{X(t), t \geq 0\}$  is a birth and death process with

$$\text{birth rates } \lambda_j \equiv \lambda, \quad \text{and death rates } \mu_j \equiv j\mu.$$

The stationary distribution is

$$P_n = \frac{\lambda^n}{n! \mu^n} P_0 = \frac{\lambda^n}{n! \mu^n} \frac{1}{\sum_{n=0}^{\infty} \frac{\lambda^n}{n! \mu^n}} = e^{-\lambda/\mu} \frac{(\lambda/\mu)^n}{n!}, \quad n = 0, 1, \dots$$

Therefore  $X(t) \sim \text{Poisson}(\lambda/\mu)$  as  $t \rightarrow \infty$ ,

$$L = \mathbb{E}[X(t)] = \lambda/\mu.$$

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### Birth & Death Queueing Models

In addition to  $M/M/1$  and  $M/M/\infty$  models, a more general family of birth & death queueing models is the following:

#### M/M/k Queueing System with Balking

Consider the  $M/M/k$  system, but suppose that a customer finds  $n$  others in the system upon his arrival will only join the system with probability  $\alpha_n$ , i.e., he balks (walks away) w/ prob.  $1 - \alpha_n$ . This system is a birth and death process with

$$\lambda_n = \lambda \alpha_n, \quad n \geq 0$$

$$\mu_n = \min(n, k)\mu, \quad n \geq 1$$

A special case of  $M/M/k$  queueing system with balking is the  $M/M/k$  system with finite capacity  $N$ , where

$$\alpha_n = \begin{cases} 1 & \text{if } n < N \\ 0 & \text{if } n \geq N \end{cases}$$

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### Birth & Death Queueing Models

For a birth & death queueing model, the stationary distribution of the number of customers in the system is given by

$$P_k = \lim_{t \rightarrow \infty} P(X(t) = k) = \frac{\lambda_0 \lambda_1 \cdots \lambda_{k-1} / (\mu_1 \mu_2 \cdots \mu_k)}{1 + \sum_{n=1}^{\infty} \frac{\lambda_0 \lambda_1 \cdots \lambda_{n-1}}{\mu_1 \mu_2 \cdots \mu_n}}, \quad k \geq 1$$

The necessary and sufficient condition for such a stationary distribution to exist is that

$$\sum_{n=1}^{\infty} \frac{\lambda_0 \lambda_1 \cdots \lambda_{n-1}}{\mu_1 \mu_2 \cdots \mu_n} < \infty.$$

With  $\{P_n\}$ , the average number of customers in the system is simply

$$L = \sum_{n=0}^{\infty} n P_n.$$

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### Birth & Death Queueing Models (Cont'd)

With balking, the rate that customers enter the system is not  $\lambda$  (since not all customers enter the system), but

$$\lambda_a = \sum_{n=0}^{\infty} \lambda_n P_n.$$

Consequently, the average waiting time is

$$W = L/\lambda_a = \frac{\sum_{n=0}^{\infty} n P_n}{\sum_{n=0}^{\infty} \lambda_n P_n},$$

and the average amount of time waiting in queue ( $W_Q$ ) and average number of customers in queue ( $L_Q$ ) are respectively

$$W_Q = W - \mathbb{E}[\text{service time}] = W - (1/\mu),$$

$$L_Q = \lambda_a W_Q$$

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### Busy Period in a Birth & Death Queueing Model

There is a alternating renewal process embedded in a birth & death queueing model.

We say a renewal occurs if the system become empty.

Using the alternating renewal theory, the long-run proportion of time that the system is empty is  $\frac{\mathbb{E}[\text{Idle}]}{\mathbb{E}[\text{Idle}] + \mathbb{E}[\text{Busy}]}$ , where

$\mathbb{E}[\text{Idle}]$  = expected length of an idle period

$\mathbb{E}[\text{Busy}]$  = expected length of a busy period

Also note that the long-run proportion of time that the system is empty is simply  $P_0 = \lim_{t \rightarrow \infty} P(X(t) = 0)$ . Since the length of an idle period  $\sim \text{Exp}(\lambda_0)$ , we have  $\mathbb{E}[\text{Idle}] = 1/\lambda_0$ . In summary, we have that

$$P_0 = \frac{1/\lambda_0}{(1/\lambda_0) + \mathbb{E}[\text{Busy}]}$$

or

$$\mathbb{E}[\text{Busy}] = \frac{1 - P_0}{\lambda_0 P_0}$$

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