

STAT253/317 Winter 2022 Lecture 19

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Chapter 8 Queueing Models

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A queueing model models “customers” arriving to receive some service and and departing. The mechanisms involved are

- ▶ input mechanism: the arrival pattern of customers in time
- ▶ queueing mechanism: the number of servers, order of the service
- ▶ service mechanism: the time to serve one or a batch of customers

We consider queueing models that follow the most common rule of service: **first come, first served**.

Common Queueing Processes

It is often reasonable to assume

- ▶ the interarrival times of customers are i.i.d. (the arrival of customers follows a renewal process),
- ▶ the service times for customers are i.i.d. and are independent of the arrival of customers.

Kendall's notation

M = memoryless, or Markov, G = General

- ▶ $M/M/1$: Poisson arrival, service time $\sim \text{Exp}(\mu)$, 1 server
= a birth and death process with birth rates $\lambda_j \equiv \lambda$, and death rates $\mu_j \equiv \mu$
- ▶ $M/M/\infty$: Poisson arrival, service time $\sim \text{Exp}(\mu)$, ∞ servers
= a birth and death process with birth rates $\lambda_j \equiv \lambda$, and death rates $\mu_j \equiv j\mu$
- ▶ $M/M/k$: Poisson arrival, service time $\sim \text{Exp}(\mu)$, k servers
= a birth and death process with birth rates $\lambda_j \equiv \lambda$, and death rates $\mu_j \equiv \min(j, k)\mu$

Common Queueing Processes (Cont'd)

- ▶ $M/G/1$: Poisson arrival, General service times $\sim G$, 1 server
- ▶ $M/G/\infty$: Poisson arrival, General service time $\sim G$, ∞ servers
- ▶ $M/G/k$: Poisson arrival, General service times $\sim G$, k servers
- ▶ $G/M/1$: General interarrival times, service times $\sim \text{Exp}(\mu)$, 1 server
- ▶ $G/G/k$: General interarrival times $\sim F$, General service times $\sim G$, k servers
- ▶ ...

Quantities of Interest for 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 $\{Q(t), t \geq 0\}$ have a stationary distribution.

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

$L_Q = \lim_{t \rightarrow \infty} \frac{\int_0^t Q(t) dt}{t}$ = 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.

Little's Formula

Let

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

We define λ_a be the arrival rate of entering customers

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

Little's Formula:

$$L = \lambda_a W$$

$$L_Q = \lambda_a W_Q$$

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

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

provided that the limits exist.

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.

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

birth rates $\lambda_j \equiv \lambda$, and death rates $\mu_j \equiv \mu$.

Recall that (see Example 6.14 in the book) 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

$$\begin{aligned} L = \lim_{t \rightarrow \infty} \mathbb{E}[X(t)] &= \sum_{n=1}^{\infty} n P_n = \frac{\lambda}{\mu - \lambda} = \frac{1/\mu}{1/\lambda - 1/\mu} \\ &= \frac{\mathbb{E}[\text{service time}]}{\mathbb{E}[\text{interarrival time}] - \mathbb{E}[\text{service time}]} \end{aligned}$$

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, $L = \lambda W$.

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

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

Note that

of customers in queue = $\max(0, \text{# of customers in system} - 1)$.

So

$$\begin{aligned} L_Q &= \sum_{n=1}^{\infty} (n-1)P_n = \underbrace{\sum_{n=1}^{\infty} nP_n}_L - \underbrace{\left(\sum_{n=1}^{\infty} P_n\right)}_{1-P_0} \\ &= L - 1 + P_0 \\ &= \frac{\lambda}{\mu - \lambda} - 1 + \left(1 - \frac{\lambda}{\mu}\right) \\ &= \frac{\lambda^2}{\mu(\mu - \lambda)} = \lambda W_Q \end{aligned}$$

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

$$L = 4, W = 40$$

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

$M/M/\infty$ 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

birth rates $\lambda_j \equiv \lambda$, 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 $t \rightarrow \infty$,

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

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 a $M/M/k$ system, but suppose a customer arrives finding n others in the system will only join the system with probability α_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

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

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

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

Birth & Death Queueing Models (Cont'd)

With balking, the rate that customers enter the system is not λ (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

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

$$L_Q = \lambda_a W_Q$$

Busy Period in a Birth & Death Queueing Model

There is an 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

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