# Queueing Systems: Lecture 3 

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## Announcements

- PS \#3 due tomorrow by 3 PM
- Office hours - Odoni: Wed, 10/18, 2:30-4:30; next week: Tue, 10/24
- Quiz \#1: October 25, open book, in class; options: 10-12 or 10:30-12:30
- Old quiz problems and solutions: posted Thu evening along with PS \#3 solutions
- Quiz coverage for Chapter 4: Sections 4.0 4.6 (inclusive); Prof. Barnett’s lecture NOT included


## Lecture Outline

- Remarks on Markovian queues
- M/E $E_{2} / 1$ example
- M/G/1: introduction, epochs and transition probabilities
- M/G/1: derivation of important expected values
- Numerical example
- Introduction to M/G/1 systems with priorities

Reference: Section 4.7

## Variations and extensions of birth-and-death queueing systems

- Huge number of extensions on the previous models
- Most common is arrival rates and service rates that depend on state of the system; some lead to closed-form expressions
- Systems which are not birth-and-death, but can be modeled by continuous time, discrete state Markov processes can also be analyzed ["phase systems"]
- State representation is the key (e.g. $\mathrm{M} / \mathrm{E}_{\mathrm{k}} / 1$ or more than one state variables - P.S. \#3)


## M/G/1: Background

- Poisson arrivals; rate $\lambda$
- General service times, $S$; $f_{s}(s) ; E[S]=1 / \mu ; \sigma_{s}$
- Infinite queue capacity
- The system is NOT a continuous time Markov process (most of the time "it has memory")
- We can, however, identify certain instants of time ("epochs") at which all we need to know is the number of customers in the system to determine the probability that at the next epoch there will be $0,1,2, \ldots, n$ customers in the system
- Epochs = instants immediately following the completion of a service


## M/G/1: Transition probabilities for system states at epochs (1)

$N$ = number of customers in the system at a random epoch, i.e., just after a service has been completed
$N^{\prime}=$ number of customers in the system at the immediately following epoch
$R=$ number of new customers arriving during the service time of the first customer to be served after an epoch
$N^{\prime}=N+R-1 \quad$ if $N>0$
$N^{\prime}=R \quad$ if $N=0$

- Note: make sure to understand how $R$ is defined


## Epochs and the value of $R$



M/G/1: Transition probabilities for system states at epochs (2)

- Transition probabilities can be written in terms of the probabilities:
$P$ [number of new arrivals during a service time = r] =

$$
\alpha_{r}=\int_{0}^{\infty} \frac{(\lambda t)^{r} \cdot e^{-\lambda t}}{r!} \cdot f_{S}(t) \cdot d t \quad \text { for } r=0,1,2, \ldots
$$

- Can now draw a state transition diagram at epochs


## A Critical Observation

- The probabilities $P[N=n]$ of having $n$ customers in the system at a random epoch are equal to the steady state probabilities, $P_{n}$, of having $n$ customers in the system at any random time!
- The PASTA property: "Poisson arrivals see time averages"
- Can use simple arguments to obtain (as for M/M/1 systems):

$$
P_{0}=1-(\lambda I \mu)=1-\rho \text { and } E[B]=1 /(\mu-\lambda)=1 /(1-\rho)
$$

- Can also derive closed-form expressions for $L, W$, $L_{q}$ and $W_{q}$


## Probability the Server is Busy

- Suppose we have been watching the system for a long time, $T$.
$\rho$, the utilization ratio, is the long-run fraction of time (= the probability) the server is busy; this means, assuming the system reaches steady state:
$\rho=\frac{\text { amount of time server is busy }}{T}=\frac{\lambda \cdot T \cdot E[S]}{T}=\lambda \cdot E[S]=\frac{\lambda}{\mu}$


## Idle and Busy Periods; $E[B]$

Observe a large number, N , of busy periods:


## Derivation of $L$ and W: M/G/1 (assumes FCFS)

$T=$ total amount of time a randomly arriving customer $j$ will spend in the M/G/1 system
$T_{1}=$ remaining service time of customer currently in service
$T_{2}=$ the time required to serve the customers waiting ahead of $j$ in the queue
$T_{3}=$ the service time of $j$

- Clearly:

$$
\begin{aligned}
& T=T_{1}+T_{2}+T_{3} \\
& W=E[T]=E\left[T_{1}\right]+E\left[T_{2}\right]+E\left[T_{3}\right]
\end{aligned}
$$

## Derivation of $L$ and W: M/G/1 [2]

- $E\left[T_{3}\right]=E[S]$
- Given that there are already $n$ customers in the system when $j$ arrives (and since one customer is being served while $n-1$ are waiting)

$$
\begin{aligned}
& E\left[T_{2} \mid n\right]=(n-1) \cdot E[S], \quad n \geq 1 \\
& E\left[T_{2} \mid n\right]=0, \quad n=0
\end{aligned}
$$

- Thus,

$$
\begin{gathered}
E\left[T_{2}\right]=\sum_{n} E\left[T_{2} \mid n\right] \cdot P_{n}=\sum_{n \geq 1}(n-1) \cdot E[S] \cdot P_{n}=E[S] \cdot\left[\sum_{n \geq 1} n P_{n}-\sum_{n \geq 1} P_{n}\right] \\
E\left[T_{2}\right]=E[S] \cdot L-E[S] \cdot \rho
\end{gathered}
$$

## Derivation of $L$ and W: M/G/1 [3]

- From our "random incidence" result (2.66):

$$
\begin{aligned}
& E\left[T_{1} \mid n\right]=\frac{\sigma_{S}^{2}+(E[S])^{2}}{2 \cdot E[S]}, \quad n \geq 1 \\
& E\left[T_{1} \mid n\right]=0, \quad n=0
\end{aligned}
$$

- Thus, giving:

$$
E\left[T_{1}\right]=\sum_{n} E\left[T_{1} \mid n\right] \cdot P_{n}=\sum_{n \geq 1} \frac{\sigma_{\mathrm{S}}^{2}+(E[S])^{2}}{2 \cdot E[S]} \cdot P_{n}=\frac{\sigma_{\mathrm{S}}^{2}+(E[S])^{2}}{2 \cdot E[S]} \cdot \rho
$$

## Derivation of $L$ and W: M/G/1 [4]

- So we finally have:
$L=\lambda W \quad$ (Little's theorem)
$W=E[T]=E\left[T_{1}\right]+E\left[T_{2}\right]+E\left[T_{3}\right]$
and solving (1) and (2), we obtain:
$L=\rho+\frac{\rho^{2}+\lambda^{2} \cdot \sigma_{S}^{2}}{2(1-\rho)} \quad(\rho<1)$
$W=\frac{1}{\mu}+\frac{\rho^{2}+\lambda^{2} \cdot \sigma_{S}^{2}}{2 \lambda(1-\rho)}$


## Expected values for M/G/1

$L=\rho+\frac{\rho^{2}+\lambda^{2} \cdot \sigma_{S}^{2}}{2(1-\rho)} \quad(\rho<1)$
$W=\frac{1}{\mu}+\frac{\rho^{2}+\lambda^{2} \cdot \sigma_{S}^{2}}{2 \lambda(1-\rho)}$
$W_{q}=\frac{\rho^{2}+\lambda^{2} \cdot \sigma_{S}^{2}}{2 \lambda(1-\rho)}=\frac{\rho^{2}\left(1+C_{S}^{2}\right)}{2 \lambda(1-\rho)}=\frac{1}{\mu} \cdot \frac{\rho}{(1-\rho)} \cdot \frac{\left(1+C_{S}^{2}\right)}{2}$
$L_{q}=\frac{\rho^{2}+\lambda^{2} \cdot \sigma_{S}^{2}}{2(1-\rho)}$
Note: $\quad C_{S}=\frac{\sigma_{S}}{E[S]}=\mu \cdot \sigma_{S}$

## Dependence on Variability (Variance) of Service Times



## Runway Example

- Single runway, mixed operations
- E[S] = 75 seconds; $\sigma_{s}=25$ seconds

$$
\mu=3600 / 75=48 \text { per hour }
$$

- Assume demand is relatively constant for a sufficiently long period of time to have approximately steady-state conditions
- Assume Poisson process is reasonable approximation for instants when demands occur


## Estimated expected queue length and expected waiting time

| $\lambda$ (per hour) | $\boldsymbol{\rho}$ | $\boldsymbol{L}_{\boldsymbol{q}}$ | $\boldsymbol{L}_{\boldsymbol{q}}$ <br> (\% change) | $\boldsymbol{W}_{\boldsymbol{q}}$ <br> (seconds) | $\boldsymbol{W}_{\boldsymbol{q}}$ <br> (\% change) |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 30 | 0.625 | 0.58 |  | 69 |  |
| 30.3 | 0.63125 | 0.60 | $3.4 \%$ | 71 | $2.9 \%$ |
|  |  |  |  |  |  |
| 36 | 0.75 | 1.25 |  | 125 |  |
| 36.36 | 0.7575 | 1.31 | $4.8 \%$ | 130 | $4 \%$ |
|  |  |  |  |  |  |
| 42 | 0.875 | 3.40 |  | 292 |  |
| 42.42 | 0.88375 | 3.73 | $9.7 \%$ | 317 | $8.6 \%$ |
|  |  |  |  |  |  |
| 45 | 0.9375 | 7.81 |  | 625 |  |
| 45.45 | 0.946875 | 9.38 | $20.1 \%$ | 743 | $18.9 \%$ |

Can also estimate $\mathrm{PHCAP} \cong 40.9$ per hour

## M/G/1 system with non-preemptive priorities: background

- $r$ classes of customers; class 1 is highest priority, class $r$ is lowest
- Poisson arrivals for each class $k$; rate $\lambda_{k}$
- General service times, $S_{k}$, for each class; $f_{S k}(\mathrm{~s}) ; E\left[S_{k}\right]=1 / \mu_{k} ; E\left[S_{k}{ }^{2}\right]$
- FIFO service for each class
- Infinite queue capacity for each class
- Define: $\rho_{k}=\lambda_{k} / \mu_{k}$
- Assume for now that: $\rho=\rho_{1}+\rho_{2}+\ldots+\rho_{r}<1$


