

M/M/1

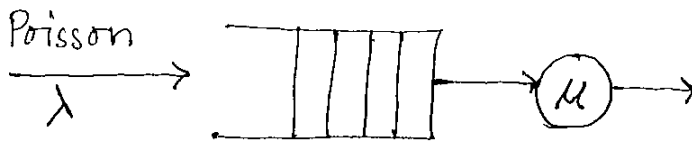
As per the Kendall notation

M: Memoryless, Arrivals, Poisson arrivals

M: Memoryless, Service time, Exponential service-time

1: Single-server

Here, buffer size infinitely large



- Poisson arrivals at the queue and service-time is exponentially distributed with mean μ .
- Suppose, we consider that n th arrival occurs, and it gets serviced for τ_n time. Representing the service time using variable s , we define the quantity $P\{\tau_n \leq s\}$

As service time is exponentially distributed then for the $f(t) = \mu e^{-\mu t}$ pdf

So,

$$\begin{aligned} P\{\tau_n \leq s\} &= \int_0^s f(t) dt = \int_0^s \mu e^{-\mu t} dt \\ &= \mu \int_0^s e^{-\mu t} dt = \mu \left[\frac{1}{-\mu} e^{-\mu t} \right]_0^s \\ &= -[e^{-\mu s} - e^0] = -[e^{-\mu s} - 1] \\ &= 1 - e^{-\mu s}, \text{ where } s \geq 0 \end{aligned}$$

... (7)

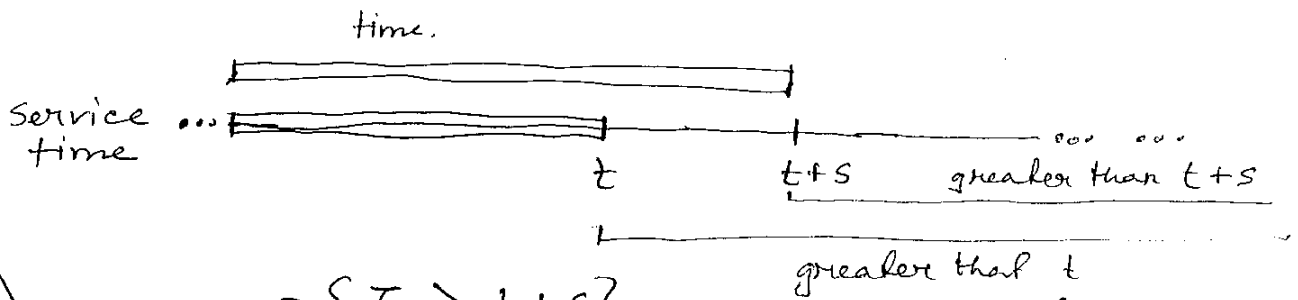
- Given that an arrival (customer/packet/call etc.) is already in service for t seconds.

we might want to know the probability ~~that~~ ~~long~~ ~~with~~ it waits for additional service time s

That is,

$P\{\tau_n > t+s \mid \tau_n > t\}$ is the term that we are interested in. So,

$$P\{\tau_n > t+s \mid \tau_n > t\} = \frac{P\{\tau_n > t+s \cap \tau_n > t\}}{P\{\tau_n > t\}}$$



$$\begin{aligned} \text{So, } &= \frac{P\{\tau_n > t+s\}}{P\{\tau_n > t\}} \\ &= \frac{1 - P\{\tau_n \leq t+s\}}{1 - P\{\tau_n \leq t\}} \\ &= \frac{1 - (1 - e^{-\mu(s+t)})}{1 - (1 - e^{-\mu t})} \quad \text{As per Eq. 1} \end{aligned}$$

$$= \frac{e^{-\mu(s+t)}}{e^{-\mu t}} = e^{-\mu s} \quad \text{What does it say?}$$

Consider

$$\begin{aligned} P\{\tau_n > s\} &= 1 - P\{\tau_n \leq s\} \\ &= 1 - (1 - e^{-\mu s}) \quad \text{Eq. 1} \\ &= e^{-\mu s} \end{aligned}$$

So, $P\{\tau_n > t+s | \tau_n > t\} = P\{\tau_n > s\}$ as the derivation suggests.

It means that additional time an arrival in service waits in the service does not depend on the ~~time~~ amount of time it is already in service.

Some quantities of a queue that we are interested in

① Queue length distribution.

We calculate P_n , that is probability that there are n -arrivals in the queue ~~for~~ waiting for their service to start.

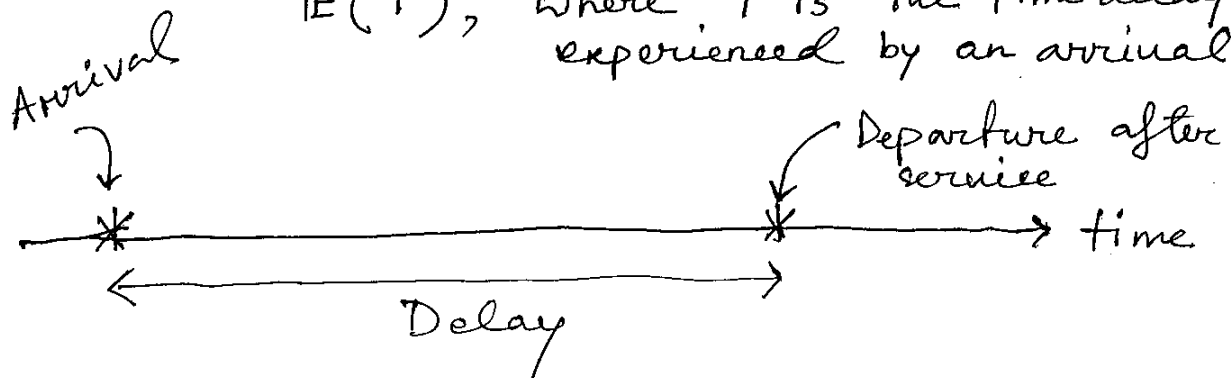
That is, the number of arrivals being waiting in the queue for service.

② Average Queue Length

Simply,
Average number of customers in the system.

It means the average number of customers/arrivals waiting in the system to get the required service from the ~~arrived~~ server.

③ Average Delay Time, also known as Time Delay $IE(T)$, where T is the time delay experienced by an arrival



④ Throughput: mean number of arrivals/customers processed per unit time.

☐ Let's consider that P_n stands for the steady state probability of having n packets/arrivals/calls in the system.

If $P_n(t)$ stands for " n " arrivals in the system at time t , then the steady state probability would be

$$P_n = \lim_{t \rightarrow \infty} P_n(t)$$

Question: Suppose at time $t + \Delta t$, the queue is in state n , then we can ask —

Given that the system is at state n at time $t + \Delta t$, What is the ~~probability~~ possible states of the system at time t ?
↙ queue
at time $t + \Delta t$

☐ Consider that $n \gg 1$, that is there are at least one arrival in the queue. So, the queue state could be any state greater than 1

A few scenarios could happen —

- At time t , the queue could be in state $\geq n+2$ or $\leq n-2$ suggesting at least two new arrivals or two departures.

↓ However, it happens with probability $O(\Delta t)$

□ The queue is in state n at time t , and the following ~~has~~ can happen within the next Δt time

- i) No arrival or departure in $(t, t+\Delta t)$
- ii) 1 arrival and 1 departure in $(t, t+\Delta t)$
- iii) Any other scenario other than i and ii is $O(\Delta t)$

□ Queue is in state $n-1$ in time t

- i) One arrival and no departure
- ii) Any other scenario $O(\Delta t)$

□ Queue is in state $n+1$ at time t

- i) One departure and no arrival
- ii) Any other scenario $O(\Delta t)$

So,

$$\begin{aligned}
 P_n(t+\Delta t) &= O(\Delta t) + P_n(t) [\mu \Delta t \lambda \Delta t + (1-\lambda \Delta t)(1-\mu \Delta t) + O(\Delta t)] \\
 &\quad + P_{n-1}(t) [\lambda \Delta t (1-\mu \Delta t) + O(\Delta t)] \\
 &\quad + P_{n+1}(t) [\mu \Delta t (1-\lambda \Delta t) + O(\Delta t)] \\
 &= O(\Delta t) + P_n(t) [\mu \lambda (\Delta t)^2 + (1-\mu \Delta t - \lambda \Delta t + \mu \lambda (\Delta t)^2) + O(\Delta t)] \\
 &\quad + P_{n-1}(t) [\lambda \Delta t - \mu \lambda (\Delta t)^2 + O(\Delta t)] + P_{n+1}(t) [\mu \Delta t - \mu \lambda (\Delta t)^2 + O(\Delta t)]
 \end{aligned}$$

Considering that $(\Delta t)^2$ is very small, we obtain $P_n(t+\Delta t)$

$$P_n(t+\Delta t) = O(\Delta t) + P_n(t) [(1-\mu \Delta t - \lambda \Delta t)] + P_{n-1}(t) [\lambda \Delta t] + P_{n+1}(t) [\mu \Delta t]$$

$$\textcircled{1} \dots = P_n(t) [1 - (\mu + \lambda) \Delta t] + P_{n-1}(t) [\lambda \Delta t] + P_{n+1}(t) [\mu \Delta t] + O(\Delta t)$$

Applying Taylor's expansion to $P_n(t+\Delta t)$ and ignoring Δt^2 terms and beyond

$$P_n(t+\Delta t) = P_n(t) + \frac{d}{dt} P_n(t) \cdot \Delta t + O(\Delta t) \dots \textcircled{2}$$

Equating ① and ②

$$P_n(t) + \frac{d}{dt} P_n(t) \Delta t + o(\Delta t) = P_n(t) [1 - (\mu + \lambda) \Delta t] + P_{n-1}(t) [\lambda \Delta t] + P_{n+1}(t) [\mu \Delta t] + o(\Delta t)$$

$$\Rightarrow \frac{d}{dt} P_n(t) \Delta t + o(\Delta t) = -P_n(t) (\mu + \lambda) \Delta t + P_{n-1}(t) \lambda \Delta t + P_{n+1}(t) [\mu \Delta t] + o(\Delta t)$$

By dividing Δt , we obtain

$$\frac{\Delta t}{\Delta t} \frac{d}{dt} P_n(t) + \frac{o(\Delta t)}{\Delta t} = -\frac{P_n(t) (\mu + \lambda) \Delta t}{\Delta t} + \frac{P_{n-1}(t) \lambda \Delta t}{\Delta t} + \frac{P_{n+1}(t) \mu \Delta t}{\Delta t} + \frac{o(\Delta t)}{\Delta t}$$

$$\Rightarrow \frac{d}{dt} P_n(t) + \frac{o(\Delta t)}{\Delta t} = -P_n(t) (\mu + \lambda) + P_{n-1}(t) \lambda + P_{n+1}(t) \mu + \frac{o(\Delta t)}{\Delta t}$$

$$\Rightarrow \frac{d}{dt} P_n(t) = -P_n(t) (\mu + \lambda) + P_{n-1}(t) \cdot \lambda + P_{n+1}(t) \cdot \mu$$

↳ Rate of change of Probability that the system state is at n in time t

For stationary probability, $\frac{d}{dt} P_n(t) = 0$

$$0 = -P_n(t) (\mu + \lambda) + P_{n-1}(t) \cdot \lambda + P_{n+1}(t) \cdot \mu$$

$$\Rightarrow P_n(t) (\mu + \lambda) = P_{n-1}(t) \cdot \lambda + P_{n+1}(t) \cdot \mu$$

As it is stationary, we can get rid of t notation

$$P_n (\mu + \lambda) = P_{n-1} \cdot \lambda + P_{n+1} \cdot \mu, \quad n \geq 1 \quad \dots \quad \textcircled{3}$$

For $n=0$, a similar approach yield

$$P_0 \lambda = P_1 \mu \quad \dots \quad \textcircled{4}$$

By rearranging Eq.3, we obtain

$$\begin{aligned} \lambda P_n &= P_{n-1} \cdot \lambda + P_{n+1} \cdot \mu - P_n \cdot \mu \\ \Rightarrow \lambda P_n &= P_{n+1} \cdot \mu + [\lambda P_{n-1} - \mu P_n] \quad \dots \quad \textcircled{5} \end{aligned}$$

M/M/1: Interpretations of the Balance Equations

The ~~balance~~ stationary P_n equation as obtained for $n \geq 1$.

$$\underbrace{(\lambda + \mu) P_n}_{\text{LHS}} = \lambda P_{n-1} + \mu P_{n+1}, \quad n \geq 1$$

The LHS denotes that the rate at which the system moves out of state n

→ Can happen because of
1 arrival and no departure] a)
→ Can happen because of
no arrival and one departure] b)

For a) Probability that the system moves up (from n to $n+1$) is

$$\begin{aligned} & \underbrace{\lambda \Delta T (1 - \mu \Delta T)}_{\substack{\text{1 arrival} \\ \text{no departure}}} + o(\Delta T) \\ &= \lambda \Delta T - \mu \lambda (\Delta T)^2 + o(\Delta T) \quad \left| \begin{array}{l} (\Delta T)^2 \text{ falls} \\ \text{under } o(\Delta T) \end{array} \right. \\ &= \lambda \Delta T + o(\Delta T) \end{aligned}$$

b) Probability that the system moves down (from n to $n-1$) by one state during ΔT .

$$\begin{aligned} & \underbrace{(1 - \lambda \Delta T)}_{\text{No arrival}} \underbrace{\mu \Delta T}_{\text{1 departure}} \\ &= \mu \Delta T - \lambda \Delta T \mu \Delta T + o(\Delta T) \\ &= \mu \Delta T - \mu \lambda (\Delta T)^2 + o(\Delta T) \\ &= \mu \Delta T + o(\Delta T) \end{aligned}$$

So, by considering Eq.4 and Eq.5 together, and considering different values of n , we can write

$$\lambda P_0 = \mu P_1 \dots \textcircled{6}$$

$$\Rightarrow P_1 = \frac{\lambda}{\mu} P_0$$

$$\lambda P_1 = \mu P_2 + [\lambda P_0 - \mu P_1] \quad \text{Considering } \lambda P_0 = \mu P_1$$

$$\Rightarrow \lambda P_1 = \mu P_2 \dots \textcircled{7}$$

$$\Rightarrow P_2 = \frac{\lambda}{\mu} P_1$$

$$\Rightarrow P_2 = \frac{\lambda}{\mu} \cdot \frac{\lambda}{\mu} P_0$$

$$\Rightarrow P_2 = \left(\frac{\lambda}{\mu}\right)^2 P_0$$

Again, for $n=2$

$$\lambda P_2 = \mu P_3 + [\lambda P_1 - \mu P_2]$$

$$\Rightarrow \lambda P_2 = \mu P_3$$

$$\Rightarrow P_3 = \frac{\lambda}{\mu} P_2$$

$$\Rightarrow P_3 = \frac{\lambda}{\mu} \cdot \left(\frac{\lambda}{\mu}\right)^2 P_0$$

$$\Rightarrow P_3 = \left(\frac{\lambda}{\mu}\right)^3 P_0$$

... following the recursive derivation, we finally obtain

$$P_n = \left(\frac{\lambda}{\mu}\right)^n P_0$$

Consider $\frac{\lambda}{\mu} = \rho$ and apply the normalization of the probabilities

$$\sum_{n=0}^{\infty} P_n = 1$$

$$\Rightarrow \sum_{n=0}^{\infty} \left(\frac{\lambda}{\mu}\right)^n P_0 = 1 \dots \textcircled{8}$$

Simplifying further, we obtain the below from equation $\textcircled{8}$

$$\sum_{n=0}^{\infty} (\rho)^n P_0 = 1$$

$$\Rightarrow P_0 \sum_{n=0}^{\infty} (\rho)^n = 1$$

$$\Rightarrow P_0 \cdot \frac{1}{1-\rho} = 1$$

$$\Rightarrow P_0 = 1-\rho$$

So, stationary probability that the system is at state n is

$$P_n = \rho^n (1-\rho)$$

For a stable queue,

$$\rho < 1 \quad \text{That is.}$$

arrival rate must be less than the service rate

$$\begin{aligned}
&\text{So, Probability of leaving state } n \text{ in } \Delta T \text{ time} \\
&= P\{\text{leaving state } n \text{ in } \Delta T\} \\
&= P\{\text{~~Being~~ staying in state } n \text{ and moving up or down in } \Delta T\} \\
&= P_n (\lambda \Delta T + \mu \Delta T) + o(\Delta T)
\end{aligned}$$

So, Rate of leaving state n is

$$\begin{aligned}
&= \lim_{\Delta T \rightarrow 0} \frac{P\{\text{leaving state } n \text{ in } \Delta T\}}{\Delta T} \\
&= \lim_{\Delta T \rightarrow 0} \frac{P_n (\lambda \Delta T + \mu \Delta T) + o(\Delta T)}{\Delta T} \\
&= \lim_{\Delta T \rightarrow 0} \left[\frac{P_n (\lambda + \mu) \Delta T}{\Delta T} + \frac{o(\Delta T)}{\Delta T} \right] \\
&= \lim_{\Delta T \rightarrow 0} \left[P_n (\lambda + \mu) + \frac{o(\Delta T)}{\Delta T} \right] \\
&= P_n (\lambda + \mu) + \lim_{\Delta T \rightarrow 0} \frac{o(\Delta T)}{\Delta T} \\
&= P_n (\lambda + \mu)
\end{aligned}$$

↳ Rate of leaving state n

Similarly, $\lambda P_{n-1} + \mu P_{n+1}$ stands for the transition rate into P_n .

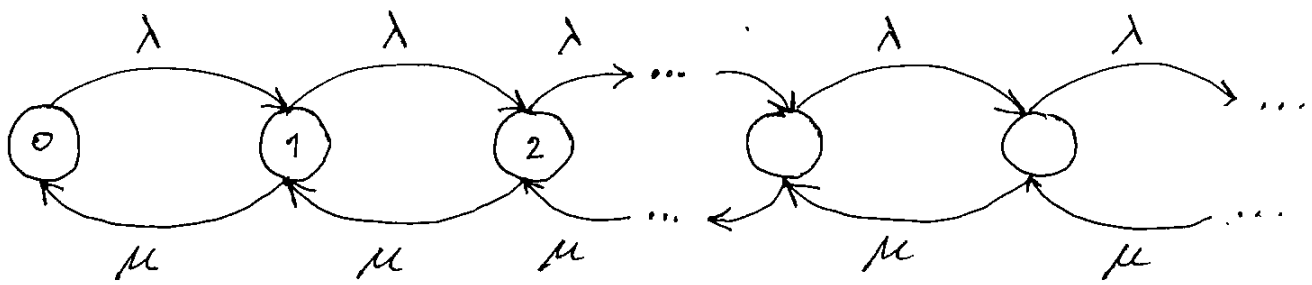
For stationary probability

Rate of leaving = Rate of transition

$$\Rightarrow (\lambda + \mu) P_n = \lambda P_{n-1} + \mu P_{n+1}$$

conservation of rates

State Transition Diagram (STD)



In the STD, state i represents the number of arrivals/packets/calls/customers in the queue.

An arc from i to j represents the rate at which the transition occurs

An arc from j to i represents the rate at which the departure occurs.

Given the above, we can evaluate P_n , the stationary probability that the state is n in two ways:

① Global Balance Equations

② Local Balance Equations

Global Balance Approach:

We obtained $\lambda P_0 = \mu P_1$

For state $n=2$: From eqⁿ. $(\lambda + \mu) P_n = \lambda P_{n-1} + \mu P_{n+1}$

$$\text{For } n=1, (\lambda + \mu) P_1 = \lambda P_0 + \mu P_2$$

$$\Rightarrow \lambda P_1 + \mu P_1 = \lambda P_0 + \mu P_2$$

$$\Rightarrow \lambda P_1 = \mu P_2 + (\lambda P_0 - \mu P_1)$$

$$\Rightarrow \lambda P_1 = \mu P_2 \dots \textcircled{1}$$

$$\Rightarrow (\lambda + \mu) P_2 = \lambda P_1 + \mu P_3$$

$$\Rightarrow \lambda P_2 + \mu P_2 = \lambda P_1 + \mu P_3$$

$$\Rightarrow \lambda P_2 = \lambda P_1 + \mu P_3 - \mu P_2$$

$$\Rightarrow \lambda P_2 = \mu P_3 + \underbrace{(\lambda P_1 - \mu P_2)}_{\text{From } \textcircled{1}}$$

$$\Rightarrow \lambda P_2 = \mu P_3$$

So, we obtain following the steps in a recursive way

$$\begin{aligned}\lambda P_0 &= \mu P_1 \Rightarrow P_1 = \frac{\lambda}{\mu} P_0 = \rho P_0 \\ \lambda P_1 &= \mu P_2 \Rightarrow P_2 = \frac{\lambda}{\mu} P_1 = \frac{\lambda}{\mu} \cdot \frac{\lambda}{\mu} P_0 = \left(\frac{\lambda}{\mu}\right)^2 P_0 \\ \lambda P_2 &= \mu P_3 \Rightarrow P_3 = \frac{\lambda}{\mu} P_2 = \frac{\lambda}{\mu} \left(\frac{\lambda}{\mu}\right)^2 P_0 = \left(\frac{\lambda}{\mu}\right)^3 P_0 \\ &\vdots \\ \lambda P_n &= \mu P_{n+1} \quad P_n = \left(\frac{\lambda}{\mu}\right)^n P_0\end{aligned}$$

Considering $\left(\frac{\lambda}{\mu}\right) = \rho = \text{traffic intensity}$, we can write
 $P_n = \rho^n P_0$. By applying normalization of probability $\sum_{n=0}^{\infty} P_n = 1$, we obtain.

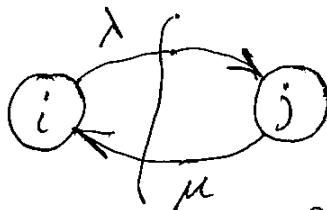
$$\begin{aligned}P_0 + P_1 + P_2 + \dots + P_n + \dots &= \sum_{n=0}^{\infty} P_n = 1 \\ \Rightarrow P_0 \sum_{n=0}^{\infty} (\rho)^n &= 1 \Rightarrow P_0 (1 - \rho) = 1 \\ &\Rightarrow P_0 = \frac{1}{1/(1-\rho)} = 1 - \rho\end{aligned}$$

Finally, $P_n = \rho^n (1 - \rho)$

Local Balance Equation:

We separate the transition from state i to j

as



$$P_0 \lambda = P_1 \mu$$

rate from
left to
right

rate from
right to
left.

So, $P_0 \lambda = P_1 \mu$

$$\Rightarrow P_1 = \frac{\lambda}{\mu} P_0$$

$$\Rightarrow P_2 = \left(\frac{\lambda}{\mu}\right)^2 P_0$$

Following previous explanations & $\sum_{n=0}^{\infty} P_n = 1$

We obtain

$$P_n = (1 - \rho) \rho^n$$