Equilibrium Analysis of the M/G/1 Queue

1. Mean Analysis using Residual Life Arguments (Section 3.1)

2. Analysis using an Imbedded Markov Chain Approach (Section 3.2)

3. Method of Supplementary Variables (done later!) (http://home.iitk.ac.in/~skb/qbook/MG1_SupVar.PDF)

Method of Stages or other exact/approximate analytical methods may also be used
Why is the M/M/1 queue so easy to analyze while the analysis of the M/G/1 queue is substantially more difficult?

- State description for M/M/1 is simple as one needs just one number (i.e. the number in the system) to denote the system state.

- This is possible because the exponential service time distribution is memoryless and service already provided to the customer currently in service need not be considered in the state description.

- This is not true for the M/G/1 queue. Its general state description would require specification of both the number currently in the system and the amount of service already provided to the customer currently being served.

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M/G/1/∞ Queue: Single server, Infinite number of waiting positions

Service discipline assumed to be FCFS unless otherwise specified. Mean results same regardless of the service discipline

Arrival Process: Poisson with average arrival rate λ
Inter-arrival times exponentially distributed with mean 1/λ.

Service Times: Generally distributed with pdf \( b(t) \), cdf \( B(t) \) and L.T.\( L_b(s) \)
Residual Life Approach for Analyzing the M/G/1 Queue (Section 3.1)

Note that this approach can only give the mean results for the performance parameters - state distributions cannot be found.

We will tacitly assume a FCFS queue. However, since only the mean results are being obtained, these will be the same for queues with other service disciplines, such as LCFS, SIRO etc..

Consider a particular arrival of interest entering the M/G/1 queue

Let \( r \) = (random) residual service time of the customer (if any) currently in service

\[ R = E(r) \quad \text{Mean Residual Service Time} \]

Then

\[ W_q = N_q E\{X\} + R = \lambda W_q E\{X\} + R \]

\[ W_q = \frac{R}{1 - \rho} \]

where

\[ \rho = \frac{\lambda E\{X\}}{\lambda \bar{X}} = \frac{\lambda}{\mu} \]

We still need to find \( R \) to find \( W_q \). However, once \( W_q \) is known, the results \( N_q, N \) and \( W \) may be found directly from that.
Residual Service Time $r(t)$ as a Function of $t$

For $t \to \infty$,

$$
\frac{M(t)}{t} \to \lambda
$$

$$
\frac{1}{M(t)} \sum_{i=1}^{M(t)} \frac{1}{2} X_i^2 \to X^2
$$

$$
R = \frac{1}{2} \lambda X^2
$$

$$
W_q = \frac{\lambda X^2}{2(1 - \rho)}
$$

$W = W_q + E(X) = \frac{\lambda X^2}{2(1 - \rho)} + X$

$N_q = \lambda W_q = \frac{\lambda^2 X^2}{2(1 - \rho)}$

$N = \lambda W = \frac{\lambda^2 X^2}{2(1 - \rho)} + \rho$

$R$ may be found as the time average of $r(t)$ using a graphical approach, as shown

$$
R = \lim_{t \to \infty} R_t
$$

where

$$
R_t = \frac{1}{t} \int_0^t r(\tau) d\tau \approx \frac{1}{t} \sum_{i=1}^{M(t)} \frac{1}{2} X_i^2
$$

$$
= \frac{1}{2} \frac{M(t)}{t} \frac{1}{M(t)} \sum_{i=1}^{M(t)} \frac{1}{2} X_i^2
$$

$For the M/M/1 queue

$E(X) = 1/\mu$, $E(X^2) = 2/\mu^2$

For the Pollaczek-Khinchine or P-K Formula

$$
W_q = \frac{\lambda X^2}{2(1 - \rho)}
$$

Substituting these lead to the same results as obtained directly for the M/M/1 queue earlier
\[ R = (1 - \rho)E[r | \text{system found empty on arrival}] + \rho E[r | \text{system found not empty on arrival}] \]

Note that the counter-intuitive nature of the above result, i.e. that it is
\[ \frac{1}{2} \left( \frac{X + \sigma^2}{X} \right) \] rather than \[ \frac{1}{2} \] illustrates the Paradox of Residual Life.

Arrival to a non-empty queue samples an ongoing service time but would tend to select longer service times more than shorter ones.

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**Some Residual Life Results**

\[ E[\tau_i - \tau_{i-1}] = X \]
\[ E[(\tau_i - \tau_{i-1})^2] = X^2 \]

where \( X \) is a lifetime and \( Y \) is then referred to as the residual lifetime

\[ f_X(x) = \frac{xf_X(x)}{X} \]

where \( \hat{X} \) is the pdf of the selected lifetime

For the distribution of \( Y \), we have the following results

\[ f_Y(y)dy = P(y \leq Y \leq y + dy) = \frac{1}{X} [1 - F_X(y)]dy \quad (3.7) \]
\[ L_Y(s) = LT \left( \frac{1}{X} - \frac{1}{X} \int_0^y f_X(y)dy \right) = \frac{1 - L_X(s)}{sX} \quad (3.8) \]
The Imbedded Markov Chain Approach (M/G/1 Queue)
(Section 3.2)

• Choose *imbedded time instants* $t_i, i=1, 2, 3, \ldots \infty$ as the instants just after the departure of jobs from the system (after completing service)

• At these time instants, we can describe the system state by the number in the system, i.e.

$$n_i = \text{Number left behind in the queue by the } i^{th} \text{ departure}$$

• We can easily see (shown subsequently) that the sequence $n_i$ forms a Markov Chain, which can be solved to obtain the equilibrium state distribution at these specially chosen time instants (“the departure instants”)

Useful Results Applicable to the M/G/1 Queue

**Kleinrock's Result:** For systems where the system state can change at most by +1 or -1, the system distribution as seen by an arriving customer will be the same as that seen by a departing customer

*State Distribution at the Arrival Instants will be the same as the State Distribution at the Departure Instants*

**PASTA:** Poisson Arrival See Time Averages

*State Distributions and Moments seen by an arriving customer will be the same as those observed at an arbitrarily chosen time instant under equilibrium conditions*
\( i^{th} \) departure leaves non-empty system

\[ n_i > 0 \]

\((i+1)^{th}\) departure

\[ n_{i+1} = n_{i} - 1 + a_{i+1} \]

\((i+1)^{th}\) service time

\[ a_{i+1} \text{ arrivals in } (i+1)^{th} \text{ service time} \]

Departure Leaves System Non-empty

\[ n_{i+1} = n_{i} - 1 + a_{i+1} \]

\[ n_{i} = 1, 2, \ldots \]

---

\( i^{th} \) departure leaves empty system

\[ n_i = 0 \]

\((i+1)^{th}\) departure

\[ n_{i+1} = a_{i+1} \]

\((i+1)^{th}\) service time

\[ a_{i+1} \text{ arrivals in } (i+1)^{th} \text{ service time} \]

Departure Leaves System Empty

\[ n_{i+1} = a_{i+1} \]

\[ n_{i} = 1, 2, \ldots \]
\[
\begin{align*}
    n_{i+1} &= a_{i+1} & \text{for } n_i = 0 \\
    &= n_i - 1 + a_{i+1} & \text{for } n_i = 1, 2, 3, \ldots
\end{align*}
\]

or

\[
\begin{align*}
    n_{i+1} &= n_i - U(n_i) + a_{i+1} & \text{for } n_i = 0, 1, 2, 3, \ldots
\end{align*}
\]

Taking expectations of LHS and RHS of (3.11) or (3.12)

\[
E\{U(n_i)\} = E\{a_{i+1}\}
\]

Since \[
\begin{align*}
    E\{U(n_i)\} &= 1 - p_0 \\
    E\{a_{i+1}\} &= \int_0^\infty (\lambda t) b(t) dt = \lambda \bar{X} = \rho
\end{align*}
\]

Therefore \[
p_0 = 1 - \rho \quad \text{P(System Empty)}
\]
\[ P(z) \text{ Generating Function for the Number in the System} \]

\[
P_i(z) = E[z^{n_i}] = \sum_{k=0}^{\infty} z^k P(n_i = k)
\]

\[
P_{r+1}(z) = E[z^{n_{r+1}}] = \sum_{k=0}^{\infty} z^k P(n_{r+1} = k)
\]

\[
P_r(z) = E[z^{n_r-U(n_r)}]E[z^{n_{r+1}}]
\]

Solve for Transient Solution

For Equilibrium State Distribution

1. Drop subscript “ \(i\) “ since equilibrium conditions are considered

2. Use \( A(z)=L_g(\lambda - \lambda z) \) \hspace{1cm} (3.13)

3. Use the following results -

\[
A'(z) = -\lambda L'_g(\lambda - \lambda z) \quad A'(1) = -\lambda L'_g(0) = \lambda \tilde{X} = \rho
\]

\[
A''(z) = \lambda^2 L''_g(\lambda - \lambda z) \quad A''(1) = \lambda^2 L''_g(0) = \lambda^2 X^2
\]

\[
P(z) = A(z)E[z^{n-U(n)}] = A(z)\sum_{k=0}^{\infty} z^{k-U(k)} P(n = k)
\]

\[
= A(z)\left[z^0 p_0 + \sum_{k=1}^{\infty} z^{k-1} p_k\right] = A(z)\left[p_0 + \frac{1}{z} \sum_{k=0}^{\infty} z^k p_k - \frac{1}{z} p_0\right]
\]

\[
= A(z)\left[\frac{1}{z} P(z) - \frac{1}{z} p_0(1-z)\right]
\]

\[
P(z) = \frac{(1-\rho)(1-z)A(z)}{A(z) - z} = \frac{(1-\rho)(1-z)L_g(\lambda - \lambda z)}{L_g(\lambda - \lambda z) - z}
\]

P-K Transform Equation \hspace{1cm} (3.14)
Under equilibrium conditions, \( P(z) \) was derived at the customer departure instants. However -

- It will hold at the customer arrival instants (Kleinrock’s Result)
- It will also hold for the time averages or at an arbitrary time instant under equilibrium conditions

Expressing \[ P(z) = \sum_{i=0}^{\infty} \alpha_i z^i \] (Taylor Series Expansion)

We can obtain \( \alpha_i = P\{i \text{ customers in the system}\} \) under equilibrium conditions

Moments of the system parameters (e.g. number in the system, may be computed directly from \( P(z) \))

For this, use \( A(1) = 1 \quad A'(1) = \lambda \bar{X} = \rho \quad A''(1) = \lambda^2 \bar{X}^2 \)

\[ P(1) = \lim_{z \to 1} P(z) = \lim_{z \to 1} \frac{(1 - \rho)(1 - z)A'(z) - A(z)}{A'(z) - 1} = \frac{(1 - \rho)}{\rho - 1} = 1 \]

This result, i.e. \( P(1) \) must be unity could have been used to obtain \( p_0 \) directly, instead of obtaining it as done earlier

Similarly \[ N = P'(1) = \rho + \frac{\lambda^2 \bar{X}^2}{2(1 - \rho)} \] Mean number in system

Knowing \( N \), the other parameters \( N_p, W, \text{ and } W_q \) may be calculated as before
Delay Distribution in a FCFS M/G/1 Queue

\[ T \quad \text{Total time spent in system (r.v.) by an arrival} \]

\[ Q \quad \text{Total waiting time (r.v.) before service begins for an arrival} \]

\[ T = Q + X \quad Q \perp X \]

\[ L_T(s) = E\{e^{-st}\} = L_Q(s)L_B(s) \]

\[ L_B(s) \text{ is known if the distribution of the service time } X \text{ is given} \]
Consider a particular job arrival and its departure (say the \(n^{th}\) one) in a FCFS M/G/1 queue.

The number of customers that the \(n^{th}\) user will see left behind in the queue when it departs will be the number of arrivals that occurred while it was in the system.

Therefore \(L_T(\lambda - \lambda z) = P(z)\)

Substituting \(s = (\lambda - \lambda z)\) \(\Rightarrow\) \(L_T(s) = \frac{s(1-p)L_B(s)}{s-\lambda + \lambda L_B(s)} \quad (3.15)\)

Substituting \(T = Q + X\), \(Q \perp X\) and \(L_B(s) = E[e^{-sX}]\)

\(L_Q(s) = \frac{L_T(s)}{L_B(s)} = \frac{s(1-p)}{s-\lambda + \lambda L_B(s)} \quad (3.16)\)

\(L_T(s)\) and \(L_Q(s)\) are the L.T.s of the pdfs of the total delay and the queueing delay as seen by an arrival in a FCFS M/G/1 queue.

An alternate approach for deriving \(L_T(s)\) and \(L_Q(s)\) may be found in Section 3.7.
Busy Period Analysis of a M/G/1 Queue (Section 3.4)

Unfinished Work $U(t)$ in a M/G/1 Queue

Idle Period

Exponentially distributed with mean $1/\lambda$.

$$f_{IP}(t) = \mu e^{-\mu t} \quad t \geq 0$$

$$L_{IP}(s) = \frac{\lambda}{s + \lambda}$$

This will have the same distribution as an *inter-arrival time*
Busy Period

Consider a busy period that starts with the arrival of customer \( A_1 \).

Let \( X_1 \) be the service time for \( A_1 \).

Let there be \( n^* \) arrivals (\( A_2, \ldots, A_{n^*+1} \)) that arrive during the service time \( X_1 \), in the sequence \( A_2, \ldots, A_{n^*+1} \).

Note that the busy period \( BP \) will consist of the sum of \( X_1 \) and \( n^* \) sub-busy periods.

Each of the sub-busy periods are i.i.d. random variables with the same distribution as that of the busy period \( BP \) (to be found)

\[
BP = X_1 + BP_2 + \ldots + BP_{n^*+1} \quad BP_j \perp BP_k \quad BP_j \perp X_1 \quad \forall j, k
\]

Solve (3.19) to obtain \( L_{BP}(s) \)

The moments of \( BP \) may be obtained directly from (3.19) using the moment generating properties of the L.T. \( L_{BP}(s) \). See Section 3.4 for the mean and some higher moments of \( BP \).
Delay Distribution in a LCFS M/G/1 Queue

Customer arrival/departure instants and delays in a LCFS M/G/1 Queue

Queueing Delay $Q = D_0 + D_1$, waiting time in queue before service

$D_0 = $ Residual service time of job during whose service $A$ arrives

$D_0 = 0$ if $A$ arrives to an empty queue (probability = $1-p$)

\[ f_{D_0}(t) = \frac{1 - B(t)}{X} \quad (3.20) \]

\[ L_{D_0}(s) = \frac{1 - L_p(s)}{sX} \quad (3.21) \]

$D_1$ will consist of *sub busy periods*, one associated with each of the customer arrivals in $D_0$

Note that $D_0$ and $D_1$ are not independent of each other

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For the case where the arrival \( A \) comes to a non-empty queue

\[
L_Q(s) = E[e^{-sQ}] = (1 - \rho) + pE[e^{-sD_0}] \text{ arrival to non-empty queue }
\]

\[
= (1 - \rho) + pE[\exp(-s(D_0 + D_t))|\text{arrival to non-empty queue}]
\]

\[
= (1 - \rho) + pE[e^{-sD_0} \exp(-sD_t)|\text{arrival to non-empty queue}]
\]

\[
E[e^{-sD_0} | D_0 = y, N_0 = n] = \text{[\text{L}_{\text{arr}}(s)]}^n
\]

\[
E[e^{-sD_0} | D_0 = y, N_0 = n] = e^{-\lambda y} \text{[\text{L}_{\text{arr}}(s)]}^n
\]

which leads to

\[
E[e^{-sD_0} | D_0 = y] = \sum_{n=0}^{\infty} \frac{(\lambda y)^n}{n!} e^{-\lambda y} \text{[\text{L}_{\text{arr}}(s)]}^n = \exp[-y(\lambda - \lambda \text{L}_{\text{arr}}(s))]
\]

\[
E[e^{-sD_0} | D_0 = y] = \exp[-y(s + \lambda - \lambda \text{L}_{\text{arr}}(s))]
\]

Using (3.21), we then get

\[
E[e^{-sD_0}] = \frac{1 - L_{D_0}(\lambda - \lambda \text{L}_{\text{arr}}(s))}{X(\lambda - \lambda \text{L}_{\text{arr}}(s))} \tag{3.22}
\]
Similarly

\[
E[e^{-s\gamma}] = \int_0^\infty E[e^{-s\gamma} | D_A = y] f_{D_A}(y) dy
\]

\[
= \int_0^\infty \exp[-y(s + \lambda L_{BP}(s))] f_{D_A}(y) dy
\]

\[
= L_{D_A}(s + \lambda - s L_{BP}(s))
\]

For the case where the arrival \( A \) comes to a non-empty queue

Using (3.21), we then get

\[
E[e^{-s\gamma}] = \frac{1 - L_{BP}(s + \lambda - \lambda L_{BP}(s))}{X(s + \lambda - \lambda L_{BP}(s))}
\]

\[
= \frac{1 - L_{BP}(s)}{X(s + \lambda - \lambda L_{BP}(s))}
\]

(3.23)

Therefore, considering both the cases where Customer A finds the queue empty and non-empty -

\[
L_Q(s) = (1 - \rho) + \rho \frac{1 - L_{BP}(s)}{X(s + \lambda - \lambda L_{BP}(s))}
\]

(3.24)

and

\[
L_Q(s) = L_Q(s) L_A(s)
\]

(3.25)

\( L_Q(s) \) and \( L_A(s) \) are the L.T.s of the pdfs of the total delay and the queueing delay as seen by an arrival in a LCFS M/G/1 queue.

The results obtained for the M/G/1 queue may be used to obtain the delay distributions for the M/D/1 queue as well. This is given in Section 3.6.
An Elapsed Time Approach for the M/G/1 Queue

Allows us to show that the state distribution at the customer departure instants will be the same as the equilibrium state distribution without using either Kleinrock’s Principle or PASTA.

\[ r(t) \]

For \( i=0,1,2,\ldots,\infty \)

\[ q_i = P\{i \text{ jobs left in system as seen by a departing job}\} \]

\[ p_i = P\{i \text{ jobs in system as seen at an arbitrarily time instant between successive imbedded points}\} \]

Note that -

• \( p_i, i=0,1,2,\ldots,\infty \) is the equilibrium state distribution of the system

• We want to prove that \( p_i=q_i, i=0,1,2,\ldots,\infty \) without using either Kleinrock’s Principle or PASTA.
Arrival Process: Poisson with rate \( \lambda \).

Service Time: pdf \( b(t) \), cdf \( B(t) \), L.T. \( L_d(s)=LT[b(t)] \)

Mean \( E[X] = \bar{X} = \frac{1}{\mu} \)

\[ \alpha_k = P\{k \text{ arrivals in a service time}\} = \int_{0}^{\infty} \frac{(\lambda x)^k}{k!} e^{-\lambda x} b(x) dx \quad k = 0, 1, \ldots, \infty \]

with generating function \( A(z) = \sum_{k=0}^{\infty} \alpha_k z^k = L_d(\lambda - \lambda z) \)

We now focus on the Markov Chain at only the imbedded points corresponding to departures from the system.

For this imbedded chain, we had obtained the generating function earlier (as \( P(z) \) in Eq. (3.14)).

We write this again as \( Q(z) \)

\[ Q(z) = \sum_{j=0}^{\infty} q_j z^j = \frac{(1 - \rho)(1 - z)A(z)}{A(z) - z} \quad \rho = \lambda \bar{X} \]

By expanding this, we can get

\[ q_0 = 1 - \rho \]

\[ q_1 = \frac{1}{\alpha_0} \left[ q_0 (1 - \alpha_0) \right] \quad k = 1 \]

\[ q_k = \frac{1}{\alpha_0} \left[ q_{k-1} - \sum_{j=1}^{k-1} q_j \alpha_{k-j} - q_0 \alpha_{k-1} \right] \quad k = 2, \ldots, \infty \]
Alternatively, we may note that this imbedded Markov Chain has the following state transition probabilities -

\[ q_{jk} = \alpha_k \]
\[ = \alpha_{k-j+1} \]
\[ j = 0 \]
\[ j = 1, 2, \ldots, \infty \]

Its equilibrium state probabilities \( \{q_j\} \) may be obtained by solving

\[ q_k = \sum_{j=0}^{\infty} q_j q_{jk} \]
\[ k = 0, 1, \ldots, \infty \]

along with the normalization condition \( \sum_{k=0}^{\infty} q_k = 1 \)

This solution method, which is used to directly obtain \( Q(z) \), has been given in more detail in the notes.

\[ D = \text{Mean time interval between successive embedded points} \]

\[ D = q_0 \left( \frac{1}{\lambda} + X \right) + \left( 1 - q_0 \right) X \]
\[ = \bar{X} + q_0 \frac{1}{\lambda} \]

Using this, \( p_0 \) (of the equilibrium state distribution) may also be obtained as the fraction of time the system stays idle, in the time interval between successive imbedded points

\[ p_0 = \frac{q_0 \frac{1}{\lambda}}{\bar{X} + q_0 \frac{1}{\lambda}} = \frac{q_0}{q_0 + \rho} = 1 - \rho \]

Same as obtained from earlier analysis.
The other equilibrium state probabilities \( p_k, k \geq 1 \) are obtained as the probability of the event of examining the system at an arbitrary time instant and finding \( k \) jobs in the system, where \( k \geq 1 \).

(Since \( k=0 \) is not being considered, this arbitrarily chosen time instant will not be one where the system is empty. So if the system became empty at the last imbedded point, the time instant chosen will have to fall after the arrival of the first customer coming subsequent to the imbedded point where the system became empty.)

Case (a) Time instant falls in a service time following an imbedded point where the system became empty.

Case (b) Time instant falls in a service time following an imbedded point where the system was not empty

The probability of occurrence of Case (a) will be \( \frac{q_0 \bar{X}}{\bar{X} + q_0 \frac{1}{\lambda}} \)

The probability of occurrence of Case (b) will be \( \frac{q_j \bar{X}}{\bar{X} + q_0 \frac{1}{\lambda}} \) when the system state at the earlier imbedded point (seen left behind by the departing customer) is \( j \) for \( j=1,2,\ldots,k \).
For both Cases (a) & (b), the pdf of the elapsed service time \( x \) for the job currently in service when the system is examined will be given by \( [1 - B(x)]/X \) using residual life arguments.

Therefore

\[
\begin{align*}
P_k &= \left( \frac{q_0 X}{X + q_0} \right) \int \left( \frac{(\lambda x)^{k-1}}{(k-1)!} e^{-\lambda x} \frac{1 - B(x)}{X} \right) dx \\
&\quad + \sum_{j=0}^{k} \left( \frac{q_j X}{X + q_0} \right) \int \left( \frac{(\lambda x)^{j-1}}{(k-j)!} e^{-\lambda x} \frac{1 - B(x)}{X} \right) dx
\end{align*}
\]

Case (a)

Case (b)

\[k = 1, \ldots, \infty\]

Let

\[
A_k = P\{k \text{ or more job arrivals in a service time} \} \quad k=0,1,2,\ldots,\infty
\]

From definition of \( A_k \)

\[
\begin{align*}
A_0 &= 1 \\
A_k &= \sum_{i} \alpha_i \\
\alpha_i &= A_k - A_{k+1}
\end{align*}
\]

\[k = 0,1,\ldots,\infty\]
We can also show that

\[ \sum_{k=1}^{\infty} A_k = \lambda \bar{x} = \rho \]

\[ A_k = \sum_{j=k}^{\infty} \alpha_j = \sum_{j=k}^{\infty} \left[ \frac{(\lambda x)^j}{j!} e^{-\lambda x} b(x) dx \right] \]

for \( k=1,2,\ldots,\infty \)

Applying these to the expression for \( p_k \) given earlier, we get

\[ p_k = \left( \frac{q_0}{q_0 + \rho} \right) \left[ A_k + \sum_{j=1}^{k} \left( \frac{q_j}{q_0} \right) A_{k-j+1} \right] \]

\[ = p_0 \left[ A_k + \sum_{j=1}^{k} \left( \frac{q_j}{q_0} \right) A_{k-j+1} \right] \]

for \( k=1,2,\ldots,\infty \)

\[ = q_0 \left[ A_k + \sum_{j=1}^{k} \left( \frac{q_j}{q_0} \right) A_{k-j+1} \right] \]
Using

\[
\left[ A_k + \sum_{j=1}^{k} \left( \frac{q_j}{q_0} \right) A_{k-j+1} \right] = \frac{q_k}{q_0} \quad \text{for } k=1,2,\ldots,\infty
\]

we then get the desired result

\[ p_k = q_k \quad \text{for } k=1,2,\ldots,\infty \]