## M/G/m/m Loss System

We can use our analysis of the *finite capacity* M/G/1/K system, to get results for the M/G/1/1 queue, considered as a special case of the M/G/m/m, *m-server loss system* discussed here.

Consider the M/G/1/K queue for the special case of K=1. For this, we get  $p_{d,0}=1$ , as the departing job will always leave the system empty. This leads to the equilibrium state probabilities at any arbitrary instant to be

$$p_1 = \frac{\mathbf{r}}{1+\mathbf{r}} = P_B$$
  $p_0 = 1 - p_1 = \frac{1}{1+\mathbf{r}}$  for  $\mathbf{r} = \mathbf{I}\overline{X}$  = offered traffic

The throughput (carried traffic) of the M/G/1/1 queue will be

$$\boldsymbol{r}_c = \boldsymbol{r}(1 - P_B) = \frac{\boldsymbol{r}}{1 + \boldsymbol{r}}$$

Note that this,  $\mathbf{r}(1+\mathbf{r})^{-1}$ , will also be the mean number in the system As expected, the mean delay W through this queue will be merely its mean service time  $\overline{X}$ .

The equilibrium state distribution at an arbitrary time instant for the M/G/m/m queue is obtained subsequently. Rather surprisingly, it turns out that the state probabilities of this system are the same as that of the corresponding M/M/m/m system where the service times are exponentially distributed. This is the reason why it is sometimes stated that the state probabilities of a *m*-server loss system are independent of the actual state

1

distribution of its service times. It may be recalled that the M/M/m/m system was actually used to model a telephone exchange and its probability of blocking was given by the *Erlang Blocking* formula to be

$$P_B = B(m, \mathbf{r}) = \frac{\frac{\mathbf{r}^m}{m!}}{\sum_{j=0}^m \frac{\mathbf{r}^j}{j!}} \qquad \mathbf{r} = \mathbf{I}\overline{X}$$
(1)

where r is the load offered to the system. This would then also hold for a M/G/m/m system. Another implication of our earlier statement would then be that the Erlang Blocking formula may still be used to calculate the blocking (i.e. the grade of service) in a telephone system, even when the call duration is not an exponentially distributed random variable.

Consider a M/G/m/m queue, where the average arrival rate is I and where the service time X has a pdf b(x) with cdf B(x). We can define the conditional distribution  $b_c(x)$  as the pdf of the service time X, given that X>x, such that

$$b_{c}(x)dx = P\{x < X < x + dx | X > x\}$$

Using the fact that the cdf  $B(x)=P\{X \mathbf{f} x\}$  and Baye's rule, we get

$$b_c(x) = \frac{b(x)}{1 - B(x)} \tag{2}$$

This conditional distribution is required in the subsequent derivation.

Let *N* be the number in the system, with 0 f N f K. We also define  $p_0$  as the equilibrium probability that the system is empty, i.e.  $p_0 = P\{N=0\}$ . Note that since this queue has no additional buffers other than the servers, the state *N* also represents the number of busy servers in the queue. An arrival is lost (i.e. blocked) if it finds the system in state *N*. Therefore the probability of blocking  $P_B$  under equilibrium conditions will be the same as  $p_N$ .

For a system in state k>0 (i.e. N=k), let the random variables  $X_i$  i=1,...,k represent the elapsed service time for the job at the  $i^{th}$  server. We define the joint probability density  $f_k(x_1,...,x_k)$  as

$$f_{k}(x_{1},...,x_{k})dx_{1}....dx_{k}$$

$$\Delta P\{N=k, x_{1} < X_{1} < x_{1} + dx_{1},...,x_{k} < X_{k} < x_{k} + dx_{k}\} \quad k=1,...,m \quad (3)$$

$$p_{k} = \int_{0}^{\infty} .... \int_{0}^{\infty} f_{k}(x_{1},...,x_{k})dx_{1}....dx_{k}$$

Using these, the following balance equations may be written

$$f_1(0)dx = \mathbf{I}p_0 dx$$

$$f_1(0) = \mathbf{I}p_0$$
(4)

$$f_{k+1}(x_1,...,x_k,0)(k+1)dx = \mathbf{I}f_k(x_1,...,x_k)dx$$
  
$$f_{k+1}(x_1,...,x_k,0) = \frac{\mathbf{I}}{k+1}f_k(x_1,...,x_k)$$
  
$$k=1,...,m-1$$
 (5)

$$f_{k}(x_{1} + \Delta x, \dots, x_{k} + \Delta x) = f_{k}(x_{1}, \dots, x_{k})(1 - I\Delta x)\prod_{n=1}^{k} [1 - b_{c}(x_{n})]\Delta x$$
$$+ (k + 1)\int_{0}^{\infty} f_{k+1}(x_{1}, \dots, x_{k}, x)b_{c}(x)dx\Delta x \qquad k = 1, \dots, m - 1$$
$$f_{m}(x_{1} + \Delta x, \dots, x_{m} + \Delta x) = f_{m}(x_{1}, \dots, x_{m})\prod_{n=1}^{m} [1 - b_{c}(x_{n})]\Delta x \qquad k = m$$
(6)

It should be noted that for  $\Delta x$  small, i.e. when  $\Delta x \rightarrow 0$ 

$$f_{k}(x_{1} + \Delta x, \dots, x_{k} + \Delta x) = f_{k}(x_{1}, \dots, x_{k}) + \Delta x \sum_{n=1}^{k} \frac{\partial f_{k}(x_{1}, \dots, x_{k})}{\partial x}$$
(7)

Using (6) and (7), we get that for  $\Delta x \rightarrow 0$ 

$$\sum_{n=1}^{k} \frac{\partial f_{k}(x_{1},\dots,x_{k})}{\partial x_{n}} + \left[ \mathbf{I} + \sum_{n=1}^{k} b_{c}(x_{n}) \right] f_{k}(x_{1},\dots,x_{k})$$
$$= (k+1) \int_{0}^{\infty} f_{k+1}(x_{1},\dots,x_{k},x) b_{c}(x) dx \qquad k = 1,\dots,m-1$$
(8)
$$\sum_{n=1}^{m} \frac{\partial f_{m}(x_{1},\dots,x_{m})}{\partial x_{n}} + \left[ \sum_{n=1}^{k} b_{c}(x_{n}) \right] f_{m}(x_{1},\dots,x_{m}) = 0$$

The conditional distributions  $f_k(x_1,...,x_k)$  may be obtained by solving (8) subject to the boundary conditions given by (4) and (5). This solution will be

$$f_k(x_1,...,x_k) = f_k(0,...,0) \prod_{n=1}^k [1 - B(x_n)] \qquad k = 1,...,m$$
 (9)

subject to  $If_k(0,...,0) = (k+1)f_{k+1}(0,...,0,0)$  k = 1,...,m-1 (10)

Using (5) and (10) gives

$$f_k(0,...,0) = p_0 \frac{\mathbf{l}^k}{k!} \qquad k = 1,...,m$$
 (11)

Note that 
$$\int_{0}^{\infty} [1 - B(x)] dx = [1 - B(x)] x \Big|_{0}^{\infty} - \int_{0}^{\infty} \frac{d[1 - B(x)]}{dx} dx = \int_{0}^{\infty} x b(x) dx = \overline{X}$$

We use this and apply (3), (9) and (11) to the normalization condition, to get

$$1 = \sum_{k=0}^{m} p_{k} = p_{0} + \sum_{k=1}^{m} f_{k} (0, \dots, 0) (\overline{X})^{k} = p_{0} \sum_{k=0}^{m} \frac{r^{k}}{k!} \quad r = l\overline{X}$$
(12)

This gives the equilibrium state probabilities as

$$p_{k} = \frac{\frac{\boldsymbol{r}^{k}}{k!}}{\sum_{n=0}^{m} \frac{\boldsymbol{r}^{n}}{n!}} \quad \text{for } k=0,1,\dots,m$$
(13)

and the blocking probability  $P_B = p_m$  leading to the same equation as (1), mentioned earlier.