

# Inventory System with Postponed Demands and Service Facilities-MAM Approach

Mohammad Ekramol Islam, Department of Business Administration, Northern University Bangladesh, e-mail: [meislam2008@gmail.com](mailto:meislam2008@gmail.com)

## Abstract

*In this paper, we consider an  $(s, S)$  inventory system with random service time. Primary demands occur according to Poisson process with parameter  $\lambda$ . In this system, there is a finite buffer whose capacity varies according to the inventory level at any given time. When the maximum buffer size is reached, further demands join a pool of infinite capacity with probability  $\gamma$  and with probability  $(1 - \gamma)$  it is lost forever. When inventory level is larger than the number of customers in the buffer, an external demand can enter the buffer for service. A pooled customer is transferred to the buffer for service at a service completion epoch with probability  $p$  if the inventory exceeds  $s + 1$  and provided the number of customer in the buffer is less than the number of items held in the inventory. It is assumed that initially the inventory level is  $S$ . When inventory level reaches  $s$  an order for replenishment is placed. The lead time is exponentially distributed with parameter  $\beta$ . We obtain the steady state system size distribution. Some performance measures are obtained and a few numerical illustrations provided.*

**Key words:** Postponed demand, Finite buffer, Replenishment, Service facilities.

## 1. Introduction

In most of the inventory models it is assumed that the inventory deplete at a rate equal to demand rate. However, it becomes unrealistic for the service facilities where the stocked item is delivered to the customers after some service is performed. In this paper we consider an  $(s, S)$  inventory system with service facilities. Arrival of demands form a Poisson process with parameter  $\lambda (> 0)$  to a buffer of finite capacity equal to the inventory level at any given time  $t$ . When the maximum buffer size is reached, further demands join a pool of infinite capacity, with probability  $\gamma$  or with probability  $(1 - \gamma)$  it is lost forever, Pooled customers are taken for service at a service completion epoch if the inventory level is at least  $s + 1$ . The service time is assumed to be exponentially distributed with parameter  $\mu$ . It is also assumed that initially the inventory level is  $S$ . When inventory level reaches  $s$  due to service, an order for replenishment is placed. The lead time is exponentially distributed with parameter  $\beta$ .

The literature available on inventory system with service facilities is too limited. Berman, Kim and Shimshak [2] consider an inventory system with service in which they assume that both the demand and the service rate are deterministic and constant and as such queues can form only during stock out period. They determine optimal order quantity to minimize the total cost per unit time. Later Berman and Kim [3, 4] analyze the non-deterministic inventory model for service facilities. They analyze the system in which customers arrive at a service facility according to Poisson process where service times are exponentially distributed and each customer demands exactly one unit of the item in the inventory both zero and positive lead time cases are discussed. Krishnamoorthy et al [9] analyze a retrieval production inventory system with service time in which primary demands occur according to Markovian Arrival Process (MAP). Using matrix analytic method they carry out the steady state analysis of the system and some performance measures are obtained. Berman and Sapna [5, 6] investigate inventory control at a service facility, which uses one item if inventory for service provided. Assuming Poisson arrival process, arbitrarily distributed service times and zero lead times they analyze the system with the restriction that, waiting space is finite. Under a specific cost structure they derive the optimum ordering quantity that minimizes the long run expected cost rate. Arivarignan, Elango and Arumugam [1] consider a perishable





$$\pi(S - i) = \pi(s)\beta_{s-i}$$

Where

$$\beta_{(s-i)} = \begin{cases} -\beta_{s-i+1}(\tilde{A}_{s-i+1,s-i} + \tilde{C}_{s-i+1,s-i})(\tilde{A}_{s-i,s-i} + \tilde{B}_{s-i,s-i} + \tilde{C}_{s-i,s-i}) & \text{if } i = 1, 2, \dots, s \\ I & \text{if } i = 0 \end{cases}$$

$$\beta_{(s-i)} = \begin{cases} [-\beta_{s-1+1}\tilde{A}_{s-i+1,s-i} - \beta_{s-i}\tilde{A}_{s-i,s-i}](\tilde{A}_{s-i,s-i} + \tilde{B}_{s-i,s-i} + \tilde{C}_{s-i,s-i})^{-1} & \text{if } i = 1, 2, \dots, s \\ -\tilde{A}_{s,s}(\tilde{A}_{s,s} + \tilde{B}_{s,s} + \tilde{C}_{s,s})^{-1} & \text{if } i = 0 \end{cases}$$

$$\pi(i) = \beta_{i+1}(A_{i+1,i} + C_{i+1,i})(A_{i,i} + B_{i,i} + C_{i,i})^{-1} \text{ if } i = Q - 1, Q - 2, \dots, s + 1$$

We have the following result on system stability

Lemma 2.1.

$$\sum_{i=s+2}^s \sum_{j=1}^{i-1} \pi(i, j) > \left(\frac{\lambda\gamma}{p\mu}\right) \sum_{i=0}^s \pi(i, i)$$

Proof. From the well-known result [Neuts (1981) theorem 1.7.1] on the positive recurrence of  $Q$ , which states that-

$$\pi A_0 e < \pi A_2 e$$

and by exploiting the structure of the matrices  $A_0$  and  $A_2$ , the stated result follows.

Theorem 2.2. When the stability condition holds the steady state probability vector  $x$  of  $Q$  which satisfies  $xQ = 0$ ,  $x e = 1$  exists.

The steady state probability vector

$$x = (x(0), x(1), x(2), \dots)$$

where components are given by

$$x(i) = x(0) R^i, i \geq 0$$

Where  $R$  is the minimal non-negative solution of the matrix quadratic equation:-

$$R^2 A_2 + R A_1 + A_0 = 0$$

The vector  $x(0)$  can be calculated using the equation

$$x(0)[B_0 + R A_2] = 0$$

together with the normalizing condition

$$x(0)(1 - R)^{-1} e = 1$$

Proof. Follows immediately from the well-known result on matrix-geometric methods (see Neuts (1981)).

For calculating the rate matrix  $R$  one can use Logarithmic Reduction Algorithm

### 3. System Performance Measures

Steady state probability vector  $x = (x(0), x(1), x(2), \dots)$  can be partitioned as

$$x(i) = (y(i, j, k)); i \geq 0, 0 \leq j \leq S, 0 \leq k \leq j$$

Some of the system performance measures are given below:

- (1) The probability mass function of number of customer in the pool: The probability that there are  $I$  customers in the pool is given by

$$P_i = x(i)e = x(0)R^i e; i \geq 0$$

- (2) Expected Inventory level in the system: Expected inventory level in the system is given by-

$$\alpha_1 = \sum_{i=0}^{\infty} \left[ \sum_{j=1}^S j \sum_{k=0}^j y(i, j, k) \right] e$$

- (3) Expected number of customers in the buffer is,

$$\alpha_2 = \sum_{i=0}^{\infty} \left[ \sum_{j=1}^S j \sum_{k=1}^j ky(i, j, k) \right] e$$

- (4) Expected number of customers in the pool,

$$\alpha_3 = \sum_{i=0}^{\infty} ix(i)e = x(0)R(I - R)^{-2} e$$

- (5) Average customers lost to the system is,

$$\alpha_4 = \lambda(1 - \gamma) \sum_{i=0}^{\infty} \left[ \sum_{j=k=0}^S y(i, j, k) \right] e$$

- (6) Expected rate that a customer will enter the pool is,

$$\alpha_5 = \lambda\gamma \sum_{i=0}^{\infty} \left[ \sum_{j=k=0}^S y(i, j, k) \right] e$$

- (7) The average rate at which the pooling customers will enter the buffer is given by

$$\alpha_6 = p\mu \sum_{i=0}^{\infty} \left[ \sum_{j=s+1}^S \sum_{k=1}^j y(i, j, k) \right] e$$

#### 4. Cost Function

Define

$C_1$  = Inventory holding cost of the system,  $C_2$  = Cost for buffer customers in the system

$C_3$  = Cost for pool customers in the system,  $C_4$  = Cost of customers lost to the system, So the total expected cost of the system is

$$\begin{aligned} E(TC(S, s, P)) &= C_1\alpha_1 + C_2\alpha_2 + C_3\alpha_3 + C_4\alpha_4 \\ E(TC(S, s, P)) &= C_1 \sum_{i=0}^{\infty} \left[ \sum_{j=1}^S j \sum_{k=0}^j y(i, j, k) \right] e + C_2 \sum_{i=0}^{\infty} \left[ \sum_{j=1}^S j \sum_{k=1}^j ky(i, j, k) \right] e \\ &\quad + C_3 \sum_{i=0}^{\infty} ix(i)e + C_4 \sum_{i=0}^{\infty} ix(i)e \end{aligned}$$

By using the above cost function, we can exploit a lot of interesting feature and can make a sensitivity analysis.

#### 4. Numerical Illustration

We provide a simple numerical illustration based on our performance measure by fixing Fixed  $S = 5, s = 2, Q = 3, \lambda = 0.5, \mu = 0.7, \beta = 0.6, p = 0.6, \gamma = 0.6, C_1 = 1, C_2 = 2, C_3 = 1, C_4 = 2$ . In table 1-5, we provide measures of the system performance by fixing the parameter's values involved in

the system, we vary over the parameters  $\lambda, \mu, \beta, p$  and  $\gamma$ . For different values of these parameters corresponding values of the system measures are provided.

Table: 1. Arrival rate Vs Different performance measures

	$\lambda = 0.2$	$\lambda = 0.3$	$\lambda = 0.4$	$\lambda = 0.5$	$\lambda = 0.6$
$\alpha_1$	3.54122	3.44974	3.27038	3.1189	3.00847
$\alpha_2$	0.362165	0.593453	0.85437	1.14709	1.47747
$\alpha_3$	0.0427584	0.176074	0.551752	1.71793	9.46571
$\alpha_4$	0.00222627	0.00937859	0.0246719	0.050412	0.0882317
$\alpha_5$	0.3394	0.0140679	0.0370079	0.075618	0.132348
$\alpha_6$	0.103909	0.143281	0.175928	0.203971	0.229517

Table: 2. Service rate Vs Different performance measures

	$\mu = 0.6$	$\mu = 0.7$	$\mu = 0.8$	$\mu = 0.9$	$\mu = 1.0$
$\alpha_1$	3.15381	3.1189	3.10037	3.09042	3.0852
$\alpha_2$	1.44111	1.14709	0.945552	0.800566	0.692144
$\alpha_3$	3.87896	1.71793	1.07129	0.778464	0.617492
$\alpha_4$	0.0634823	0.050412	0.0419241	0.0361165	0.0319699
$\alpha_5$	0.0952251	0.075618	0.0628862	0.0541748	0.0479548
$\alpha_6$	0.125296	0.146179	0.167062	0.207452	0.208827

Table: 3. Replenishment rate Vs Different performance measures

	$\beta = 0.4$	$\beta = 0.5$	$\beta = 0.6$	$\beta = 0.7$	$\beta = 0.8$
$\alpha_1$	2.7194	2.9549	3.1189	3.24018	3.33281
$\alpha_2$	1.10445	1.12528	1.14709	1.16721	1.18511
$\alpha_3$	6.50534	2.68521	1.17793	1.29464	1.06335
$\alpha_4$	0.0695216	0.0576941	0.050412	0.0456062	0.042261
$\alpha_5$	0.104282	0.0865412	0.075618	0.0684092	0.0633917
$\alpha_6$	0.178756	0.193112	0.203971	0.212473	0.219313

Table: 4. Probability of transferring the pool customer to buffer Vs Different performance measures

	$p = 0.4$	$p = 0.5$	$p = 0.6$	$p = 0.7$	$p = 0.8$
$\alpha_1$	3.1357800	3.12578	3.11890	3.11374	3.06936
$\alpha_2$	1.1719900	1.15660	1.14709	1.14089	1.00008
$\alpha_3$	4.2286200	2.41191	1.17793	1.35453	0.68010
$\alpha_4$	0.0505280	0.05041	0.05041	0.05048	0.04668
$\alpha_5$	0.0757921	0.07562	0.07562	0.07571	0.04669
$\alpha_6$	0.1381150	0.171072	0.20397	0.23680	0.25003

Table: 5. Joining probability to pool Vs Different performance measures

	$\gamma = 0.4$	$\gamma = 0.5$	$\gamma = 0.6$	$\gamma = 0.7$	$\gamma = 0.8$
$\alpha_1$	3.14825	3.13416	3.1189	3.10209	3.08329
$\alpha_2$	1.05942	1.10032	1.14709	1.20083	1.26299
$\alpha_3$	0.65343	1.05776	1.17793	2.90313	5.41881
$\alpha_4$	0.06727	0.059313	0.050412	0.04035	0.02885
$\alpha_5$	0.04484	0.059312	0.075618	0.094147	0.11539
$\alpha_6$	0.196428	0.199945	0.203971	0.208600	0.213959

## 5.Sensitivity Analysis

Table: 6 . Arrival rate Vs Different performance measures

Arrival Rate	Total Cost	Service rate	Total cost	Replenishment	Total cost
0.2	4.3127609	0.6	9.92768646	0.4	11.5726832
0.3	4.83147718	0.7	7.30193400	0.5	8.0052482
0.4	5.5802158	0.8	6.1466122	0.6	6.9618340
0.5	7.231834	0.9	5.5422490	0.7	6.9604524
0.6	9.856552	1.0	5.1509190	0.8	6.8586410

Form the table 6, It is observed that the arrival rate is has a vital impact to the system. The result is obvious as the rate is increased it has impact on higher reordering, lost sales and also increased the cost of carrying pool and Buffer customers. The service rate increase indicates cost decrease as because of less cost of customers lost and less waiting time to pool and buffer customers. Finally we can see that replenishment rate increase results cost decrease.This is obvious as more replenish will support more service, less the cost of lost sales as well as less cost for pool and buffer.

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