# A comparative analysis of $(s, Q)$ and $(s, S)$ ordering policies in a queuing-inventory system with stock-dependent arrival and queue-dependent service process 

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#### Abstract

This article deals with a Markovian queuing-inventory system (MQIS) under the stochastic modeling technique. The arrival stream of this system is dependent on the present stock level at an instant. Meanwhile, the system focuses on reducing the waiting time of a unit by assuming a queue-dependent service policy (QDSP). The system consists of an infinite waiting hall to receive an arriving unit. The MQIS assumes that no unit of arrival is allowed when the stock level of the system is empty. The discussion of this MQIS runs over the two types of ordering principles named 1) $(s, Q) 2)(s, S)$. According to both ordering principles, the assumed arrival and service patterns have been considered separately and classified as Model-I (M-I) and Model-II (M-II) respectively. The steady state of the system for both M-I and M-II is analysed and resolved under the Neuts matrix-geometric technique. The system performance measures of the system are also computed. The expected cost function of both M-I and M-II are constructed as well. Further, the necessary numerical illustrations are provided and distinguished for M-I and M-II to explore the proposed model. This paper finds the optimum ordering policy to execute the stock-dependent arrival and queue-dependent service strategies.


Keywords: stock-dependent arrival, queue-dependent service, infinite waiting hall, ordering principles

## 1. Introduction

Successful retail or wholesale businesses almost always have two important aspects: 1) effective management, and 2) inventory control. The observation of an inventory system throughout the whole day provides the knowledge to understand inventory management. In each unit of time, the inventory changes constantly due to sales and service, damage, reordering, and so on. In such a way, effective management
and inventory control are maintained by limiting these factors. On the other hand, customers play a key role in every inventory business. Inventory management does focus on attracting those customers and making them loyal customers. To do such things, management has to introduce and implement new policies in order to make a predictable profit. For the attraction of customers, the business owners start displaying their products in front of the shop and giving advertisements. In real life, one can see that all the organizations give advertisements through Android phones, television, social media, etc. These advertisements make changes psychologically on the customer's mind to buy the product, and naturally, this will increase the number of arrivals into the inventory system. This idea is applied in the proposed MQIS as it assumes that the arriving jobs occur on the basis of the displayed stock level of the system.

The service facility of the management should be smarter and faster than the other competitors. Because it plays an important role in accommodating the customer in the waiting hall, either to wait for the service or to exit the system. A single server service channel faces customer impatient situation problems in the queuing systems. To eradicate them and generate loyal or happy customers, the management must come up with innovative ideas to develop their service facility. Many organisations try to provide a nonhomogeneous service facility in order to avoid customer loss and impatience. When we look into real-life observations, for example, in a single-server fast food restaurant, the server prepares food as fast as the existing queue length. If the queue length decreases, the speed of the server also becomes normal. This phenomenon is to be applied in the proposed MQIS as QDSP. Among these assumptions, the considered MQIS has a detailed discussion based on the two ordering principles: $(s, Q)$ or $(s, S)$.

### 1.1. Literature review

The queuing-inventory has received a lot of attention in recent decades for doing research on stochastic modeling. Many varieties of discussions and analyses that exist in the literature are merely related to our proposed model. Since we are all living in a modern technological world, many companies introduce their products for sale with some innovative features. These features are not easily understandable to all customers. They need an explanation about the corresponding products regarding the handling procedures, guarantee, and warranty of the products, etc. Hence, to obtain such an explanation of the product, a customer requires a service facility from the system. According to the queuing-inventory existing literature, Melikov et al. [21] and Sigman et al. [28] introduced the service facility in order to improve customer satisfaction. Subsequently, many authors developed their research with service facilities (see $[2,3,16,18,27])$. The readers can refer to the enlisted papers $[1,4-6,11,12,20,22,25]$ to learn the service facility related interpretations in the stochastic modeling inventory systems.

When we analyse such kinds of MQIS in this modern scenario, every inventory business needs an innovative idea to increase the birth rate of the system. Many businesses try to display their products in the appropriate places in the store in response to the increase in arrivals. Some companies execute new ideas instead of displaying their products. For example, they do advertisements through television, social media, and so on. When adapting this idea to the inventory business, which can be defined as a stock-dependent arrival process, will increase the birth rate of the system. Datta and Pal [7] discussed the inventory model with an inventory-level-dependent demand rate. Karabi et al. [17] analysed the two different arrival rates, which is called the two-component demand rate (TCDR). In this paper, they classified TCDR as two stages of inventory level: one is zero inventory and the other is positive inventory.

In particular, they assumed that the arrival rate is constant when the inventory is empty and a varying arrival rate if the inventory is positive. Moreover, the varying arrival rate was defined as an increasing function and is controlled by the scaling factor, which lies between $[0,1]$.

Diana Tom Varghese and Dhanya Shajin [31] determined the finite storage inventory system with a variable intensity rate for the arrival process, and it is assumed to be a non-homogeneous Poisson process. Mostly, the traditional inventory systems apply a constant arrival rate to explore their models. However, Sandeepkumar [19] investigated the optimization of an inventory system in which the arriving customer intensity rate is dependent on the current stock level. Recently, Jeganathan et al. [13] presented a comparative study between the $(s, S)$ and $(s, Q)$ ordering policies on the MQIS. This paper explains the stock-dependent arrival rate for those two ordering policies separately. The author makes reference to $[25,29,32,33]$ for the $(s, Q)$ ordering principle as well as [31] for the $(s, S)$ ordering principle. The following listed articles explore the Stock-dependent Arrival Policy (SDAP) [8, 24, 26, 30].

Nowadays, many single-server inventory systems implement a variety of similar practices in order to improve their service facilities. This is because the queue size becomes large in a single server system. When the queue length increases, the waiting time of a customer will also increase, and it will result in the customer's loss. To reduce such losses, some single-server service channels are ready to provide a nonhomogeneous service rate, which is assumed to be independently and identically distributed. Jeganathan et al. [10] used two kinds of non-homogeneous service rates in the MQIS, which are determined by the threshold level of queue length. Recently, Jeganathan et al. [15] worked on an MQIS with a retrial facility in which they applied a non-homogeneous service rate based on a queue-dependent service facility. In this, they assumed that after every completion of the service process, the server observes the queue length and then starts the next service at a different rate as per the size of the queue length. This type of service facility can be seen in fast-food restaurants, supermarkets, and so on. Many papers learn more about QDSP, and a few of them will be provided to the readers [9, 14, 15, 34-36].

To the best of authors' knowledge, no paper has been published with SDAP and QDSP that is currently available with an infinite queue size. This idea would be a research gap in the queuing-inventory literature. In order to fill such a research gap, we proposed a stochastic model with the assumption of SDAP and QDSP. In addition, we investigate the two different types of ordering principles known as: 1) $(s, Q) 2)(s, S)$.

In the end, the remaining part of this paper is partitioned as follows: model description in Section 2, analysis of the model under each ordering principle in Section 3, and Section 4. Furthermore, numerical interpretations are given in Section 5, and finally, the concluded results are stated in Section 6.

## 2. Notations and model description

| 0 | - zero matrix of an appropriate dimension |
| :---: | :---: |
| e | - column vector of convenient size having one in each entry |
| I | - identity matrix |
| $\delta_{i j}$ | $- \begin{cases}1, & \text { if } j=i \\ 0, & \text { otherwise }\end{cases}$ |
| $H(x)$ | $- \begin{cases}1, & \text { if } x \geq 0 \\ 0, & \text { otherwise }\end{cases}$ |
| $\bar{\delta}_{i j}$ | - 1- $\delta_{i j}$ |

The proposed model describes the MQIS with an infinite queue size that can store up to $S$ items in the inventory. The arrival pattern of a unit holds the SDAP. The intensity of an arriving unit is defined as $\lambda_{j}(1 \leq j \leq L)$ where $1 \leq L \leq S$ and the arrival process of a unit follows a non-homogeneous Poisson process. Here, $L$ is said to be the threshold limit to terminate the stock-dependent arrival pattern. That is, if the current inventory level exceeds the threshold limit, then the arrival rate of new customers will follow a homogeneous Poisson process. Each arriving unit joins the infinite queue, which is attached to the MQIS. They approach the service channel on the basis of first come, first served (FCFS). The service channel has a single server to provide the best service to the customer. Every customer can purchase only one unit of a product from the MQIS.

The service channel holds the QDSP in order to give their best service facility to an arriving customer. This QDSP is defined as the service facility that is dependent on the number of customers in the queue at an epoch. The intensity of this service process is denoted as $\mu_{i}(1 \leq i \leq k)$, where $k$ is the threshold limit of the queue length. Due to the practical complications and the assumption of an infinite queue size, the queue-dependent service facility is terminated when the queue size reaches $k$. At this threshold limit, the service rate of the system is assumed to be non-homogeneous. Once the queue length, $i$ crosses the threshold limit $k$, the intensity of a service process is defined as $\mu_{k}$, for all $i \geq k$. At this level, the service rate of the system is assumed to be homogeneous. At the end of service completion, the customer chooses the product with probability $p$ and not with probability $q$ to buy it. The mean service time of the MQIS is assumed to be exponentially distributed. More clearly, the service process of the MQIS does not follow QDSP after the threshold limit point of $k$.

Moreover, the MQIS does not allow the arriving unit to enter the waiting hall. During the stock-out situation, the already-arrived customer has to wait for the commencement of reordered products. Once the replenishment products are received, the service starts immediately. To perform such replenishment of the products, the proposed MQIS has an analysis of two types of ordering principles along with the above-mentioned assumptions separately.

Definition 1. $(s, Q)$ ordering principle. This principle states that when a reorder is triggered, the replenishment quantity of a $Q=S-s$ number of products is always fixed. Such reorder is to be done if the present stock level falls to the reorder limit $s$.

Definition 2. $(s, S)$ ordering principle. This principle states that the replenishment quantity varies in order to fill the maximum capacity of the system when the reorder is triggered. Such reorder is to be done if the present stock level falls to the reorder limit $s$.

These two ordering principles are to be discussed as Model-I and Model-II, respectively. For each ordering principle, the intensity of the replenishment process is identified as $\beta$. The mean reorder time of each principle follows an exponential distribution. Furthermore, the MQIS will consist of defective products. An item in the inventory may become imperfect. So, we use $\gamma$ to denote a defective rate of an inventory at any time $t$. The defective rate of a current inventory is defined as $j \gamma$, where $1 \leq j \leq S$. The mean lifetime of a product is assumed to be exponentially distributed.

State space. Let $N(t)$ denote the number of customers in the system at time $t$ and $S(t)$ indicate the present stock level of the system at time $t$. A stochastic process is formed by the doublet $\{X(t), t \geq 0\}$
$=\{(N(t), S(t)), t \geq 0\}$. It also generates a quasi birth-and-death ( QBD ) process. The state space of the system is defined by $E=\{(i, j): i=0,1,2, \ldots$ and $j=0,1, \ldots, S\}$. Since the state space is discrete, we say that the proposed system comes under the classification of a discrete state and a continuous-time stochastic process. Also, the process $\{X(t), t \geq 0\}$ is said to be a continuous time Markov chain (CTMC).

## 3. Model-I

This section describes the MQIS with the $(s, Q)$ ordering principle.

### 3.1. Construction of matrices

The CTMC has an infinitesimal generator matrix $P$ as follows
where

$$
A_{i, i}= \begin{cases}j \gamma, & i^{\prime}=i, i=0,1,2, \ldots, k \\ & j^{\prime}=j-1, j=1,2, \ldots, S \\ -\beta & i^{\prime}=i, i=0,1,2, \ldots, k \\ \beta & j^{\prime}=j, j=0 \\ & i^{\prime}=i, i=0,1,2, \ldots, k \\ -\left(\bar{\delta}_{i 0}\left(\mu_{i}+j \gamma\right)+H(s-j) \beta+\lambda_{j}\right) & i^{\prime}=Q+j, j=0,1,2, \ldots, s \\ & i^{\prime}=i=0,1,2, \ldots, k \\ -\left(\bar{\delta}_{i 0}\left(\mu_{i}+j \gamma\right)+\lambda_{L}\right) & i^{\prime}=i, j=1,2, \ldots, L \\ & j^{\prime}=j, j=L+1, L+\ldots, k \\ 0 & \text { otherwise }\end{cases}
$$

$$
A_{i, i}= \begin{cases}h_{j}^{\prime}, & i^{\prime}=i, i=k, k+1, \ldots \\ & j^{\prime}=j-1, j=1,2, \ldots, S \\ \beta & i^{\prime}=i, i=k, k+1, \ldots \\ j^{\prime}=Q+j, j=0,1,2, \ldots, s \\ f_{j}^{\prime} & i^{\prime}=i, i=k, k+1, \ldots \\ & j^{\prime}=j, j=0,1,2, \ldots, S \\ 0 \quad & \text { otherwise }\end{cases}
$$

where $f_{j}^{\prime}=\left\{\begin{array}{ll}-\left(\bar{\delta}_{j 0}\left(\mu_{k}+j \gamma+\lambda_{j}\right)+H(s-j) \beta\right), & \text { if } j=0,1, \ldots, L \\ -\left(\mu_{k}+j \gamma+\lambda_{L}+H(s-j) \beta\right), & \text { if } j=L+1, L+2, \ldots, S\end{array}\right.$ and $h_{j}^{\prime}=j \gamma$.

$$
A_{K}= \begin{cases}h_{j}, & i^{\prime}=i-1, i=k \\ & j^{\prime}=j-1, j=1,2, \ldots, S \\ \beta & i^{\prime}=i, i=k \\ f^{\prime}=Q+j, j=0,1,2, \ldots, s \\ f_{j} & i^{\prime}=i, i=k \\ & j^{\prime}=j, j=0,1, \ldots, S \\ 0 & \text { otherwise }\end{cases}
$$

where $f_{j}=-\left(\bar{\delta}_{j 0}\left(\mu_{i}+j \gamma\right)+H(s-j) \beta\right)$ and $h_{j}=p \mu_{k}+j \gamma$.

$$
\begin{aligned}
& A_{i, i-1}= \begin{cases}p \mu_{i}, & i^{\prime}=i-1, i=1,2, \ldots, k-1 \\
& j^{\prime}=j-1, j=1,2, \ldots, S \\
q \mu_{i}, & i^{\prime}=i-1, i=1,2, \ldots, k-1 \\
& j^{\prime}=j, j=1,2, \ldots, S \\
0, & \text { otherwise }\end{cases} \\
& A_{i, i-1}= \begin{cases}p \mu_{k}, & i^{\prime}=i-1, i=k, k+1, \ldots \\
& j^{\prime}=j-1, j=1,2, \ldots, S \\
q \mu_{k}, & i^{\prime}=i-1, i=k, k+1, \ldots \\
& j^{\prime}=j, j=1,2, \ldots, S \\
0, & \text { otherwise }\end{cases} \\
& A_{0,1}= \begin{cases}\lambda_{j} & i^{\prime}=i+1, i=0,1, \ldots \\
& j^{\prime}=j, j=1,2, \ldots, L \\
\lambda_{L} & i^{\prime}=i+1, i=0,1, \ldots \\
& j^{\prime}=j, j=L+1, L+2, \ldots, S \\
0 & \text { otherwise }\end{cases}
\end{aligned}
$$

Lemma 1. The stationary probability vector $\Pi_{1}=\left(\pi_{1}^{(0)}, \pi_{1}^{(1)}, \ldots, \pi_{1}^{(S)}\right)$ to the generator matrix, $A_{K}$ is determined by

$$
\pi_{1}^{(j)}=\pi_{1}^{(0)} \Omega_{j}, \quad j=0,1,2, \ldots, S
$$

where

$$
\Omega_{j}= \begin{cases}1 & j=0 \\ \frac{(-1)^{j} \prod_{z=0}^{j-1} f_{z}}{\prod_{z=1}^{j} h_{z}}, & j=1,2, \ldots, Q \\ -\left[\frac{\Omega_{j-Q+1} \beta-\Omega_{j-1} f_{j-1}}{h_{j}}\right], & j=Q+1, Q+2, \ldots, S-1 \\ -\frac{\Omega_{s} \beta}{f_{S}}, & j=S\end{cases}
$$

Proof. Let $A_{K}=A_{k k-1}+A_{k k}+A_{01}$ and solving $\Pi A_{K}=\mathbf{0}$, we get

$$
\begin{align*}
\pi_{1}^{(j)} f_{j}+\pi_{1}^{(j+1)} h_{j+1} & =0, \quad j=0,1, \ldots, Q-1  \tag{1}\\
\pi_{1}^{(Q-j)} \beta+\pi_{1}^{(j)} f_{j}+\pi_{1}^{(j+1)} h_{j+1} & =0, \quad j=Q, Q+1, \ldots, S-1  \tag{2}\\
\pi_{1}^{(Q-j)} \beta+\pi_{1}^{(j)} f_{j} & =0, \quad j=S \tag{3}
\end{align*}
$$

By solving the above system of equations recursively, we get the stated result.

### 3.2. Stability condition

Lemma 2. The stability condition of the system is

$$
\begin{equation*}
\mu_{k} \sum_{j=1}^{S} \Omega_{j}>\sum_{j=1}^{L} \Omega_{j} \lambda_{j}+\lambda_{L} \sum_{j=L+1}^{S} \Omega_{j} \tag{4}
\end{equation*}
$$

Proof. By the Neuts result for the stability condition,

$$
\begin{equation*}
\Pi_{1} A_{k k-1} \mathbf{e}>\Pi_{1} A_{01} \mathbf{e} \tag{5}
\end{equation*}
$$

and writing it explicitly, we get L.H.S as $\pi^{(0)} \sum_{j=1}^{S} \Omega_{j} \mu_{k}$ and R.H.S as $\pi^{(0)}\left(\sum_{j=1}^{L} \Omega_{j} \lambda_{j}+\lambda_{L} \sum_{j=L+1}^{S} \Omega_{j}\right)$.
Substituting the obtained L.H.S and R.H.S in (5), we obtain

$$
\mu_{k} \sum_{j=1}^{S} \Omega_{j}>\sum_{j=1}^{L} \Omega_{j} \lambda_{j}+\lambda_{L} \sum_{j=L+1}^{S} \Omega_{j}
$$

### 3.3. Computation of $R_{1}$ matrix

Due to the structure of the generator matrix and stationary probability vector, the $R_{1}$ matrix can be determined by the matrix equation

$$
\begin{equation*}
R_{1}^{2} A_{k k-1}+R_{1} A_{k k}+A_{01}=\mathbf{0} \tag{6}
\end{equation*}
$$

where

$$
R_{1}=\begin{gathered}
0 \\
0 \\
1 \\
2 \\
\vdots \\
S
\end{gathered}\left(\begin{array}{cccc}
0 & 1 & \ldots & S \\
0 & 0 & \ldots & 0 \\
r_{10} & r_{11} & \ldots & r_{1 S} \\
r_{20} & r_{21} & \ldots & r_{2 S} \\
\vdots & \vdots & \ldots & \vdots \\
r_{S 0} & r_{S 1} & \ldots & r_{S S}
\end{array}\right)
$$

Substituting $R_{1}$ in (6) and writing it explicitly, we get the following set of equations. For $j=$ $1,2, \ldots, S$

$$
\begin{aligned}
x_{j j^{\prime}} p \mu_{k}+r_{j j^{\prime}} f_{j}^{\prime}+r_{j j^{\prime}+1} h_{j+1}^{\prime}=0, & \text { if } j^{\prime}=0 \\
x_{j j^{\prime}} q \mu_{k}+x_{j j^{\prime}+1} p \mu_{k}+r_{j j^{\prime}} f_{j}^{\prime}+r_{j j^{\prime}+1} h_{j+1}^{\prime}+\lambda_{j} \delta_{j j^{\prime}}=0, & \text { if } j^{\prime}=1,2, \ldots, L \\
x_{j j^{\prime}} q \mu_{k}+x_{j j^{\prime}+1} p \mu_{k}+r_{j j^{\prime}} f_{j}^{\prime}+r_{j j^{\prime}+1} h_{j+1}^{\prime}+\lambda_{L} \delta_{j j^{\prime}}=0, & \text { if } j^{\prime}=L+1, L+2, \ldots, Q-1 \\
x_{j j^{\prime}} q \mu_{k}+x_{j j^{\prime}+1} p \mu_{k}+r_{j j^{\prime}-Q} \beta+r_{j j^{\prime}} f_{j}^{\prime}+r_{j j^{\prime}+1} h_{j+1}^{\prime}+\lambda_{L} \delta_{j j^{\prime}}=0, & \text { if } j^{\prime}=Q, Q+1, \ldots, S-1 \\
x_{j j^{\prime}} q \mu_{k}+r_{j j^{\prime}-Q} \beta+r_{j j^{\prime}} f_{j}^{\prime}+\lambda_{L} \delta_{j j^{\prime}}=0, & \text { if } j^{\prime}=S
\end{aligned}
$$

Solving the above system of non-linear equations by the Gauss-Seidal iterative process, we will obtain the $R_{1}$ matrix.

### 3.4. Partition of the steady state vector

The partition of the steady-state probability vector of the system is defined as follows:

$$
\begin{gathered}
\phi_{1}=\left(\phi_{1}^{(0)}, \phi_{1}^{(1)}, \phi_{1}^{(2)}, \ldots\right) \\
\phi_{1}^{(i)}=\left(\phi_{1}^{(i, 0)}, \phi_{1}^{(i, 1)}, \phi_{1}^{(i, 2)}, \ldots, \phi_{1}^{(i, S)}\right), i=0,1,2, \ldots
\end{gathered}
$$

### 3.5. Computation of steady state probability vector

The entire probability vector of all system states is $\phi_{1}=\left(\phi_{1}^{(0)}, \phi_{1}^{(1)}, \phi_{1}^{(2)}, \ldots\right)$. The system balance equations are given by $\phi_{1} P=\mathbf{0}$ and $\phi_{1} \mathbf{e}=1$. Then, the steady-state probabilities of the queuing-inventory system are calculated [23] by $\phi_{1}^{(i)}=\phi_{1}^{(k)} R_{1}^{(i-k)}$ where $i=k+1, k+2, \ldots$ and the initial conditions are represented by the vectors $\phi_{1}^{(i)}=0,1, \ldots, k$ are obtained by solving part of the balance equations

$$
\begin{aligned}
\phi_{1}^{(0)} A_{00}+\phi_{1}^{(1)} A_{10} & =\mathbf{0} \\
\phi_{1}^{(0)} A_{01}+\phi_{1}^{(1)} A_{11}+\phi_{1}^{(2)} A_{21} & =\mathbf{0} \\
\phi_{1}^{(i-1)} A_{01}+\phi_{1}^{(i)} A_{i i}+\phi_{1}^{(i+1)} A_{i+1, i} & =\mathbf{0}, \quad i=2,3, \ldots, k-1 \\
\phi_{1}^{(k-1)} A_{01}+\phi_{1}^{(k)}\left(A_{1}+R_{1} A_{2}\right) & =\mathbf{0} \\
\sum_{n=0}^{k-1} \phi_{1}^{(n)} \mathbf{e}+\phi_{1}^{(k)}\left[I-R_{1}\right] \mathbf{e} & =1
\end{aligned}
$$

### 3.6. System performance measures

The expected system performance of the model under the $(s, Q)$ ordering principle is determined by the following measures:
Expected inventory level. In the MQIS, the expected inventory level of the system is defined as the sum of the product value of the current inventory level and the stationary probability vector

$$
\Theta_{1}=\sum_{i=0}^{\infty} \sum_{j=1}^{S} j \phi_{1}^{(i, j)}
$$

Expected reorder rate. when the present inventory level reduces to $s+1$, there can either be a service completion happened or an item becomes defective. In the case of either of these two occurrences, the replenishment process is immediately triggered

$$
\Theta_{2}=\sum_{i=1}^{k} p \mu_{i} \phi_{1}^{(i, s+1)}+\sum_{i=k+1}^{\infty} p \mu_{k} \phi_{1}^{(i, s+1)}+\sum_{i=0}^{\infty}(s+1) \gamma \phi_{1}^{(i, s+1)}
$$

Expected perishable rate. Since the system may have imperfect items in the storage space, we require an expected perishable rate of the system. This could be done using the sum of the product $j \gamma$ and the stationary probability vector, where $j$ is the current inventory level

$$
\Theta_{3}=\sum_{i=0}^{\infty} \sum_{j=1}^{S} j \gamma \phi_{1}^{(i, j)}
$$

Expected number of customers in the system. All the customers in the system purchase an item through the first come first serve discipline. The expected number of customers in the system is the sum of the product value of the number of customers present in the system and the stationary probability vector

$$
\Theta_{4}=\sum_{i=1}^{\infty} \sum_{j=0}^{S} i \phi_{1}^{(i, j)}
$$

Expected arrival rate of a customer in the system. The sum of the product of the average arrival rate of a customer and the stationary probability vector defines the expected arrival rate of a customer

$$
\Theta_{5}=\sum_{i=0}^{\infty} \sum_{j=1}^{L} \lambda_{j} \phi_{1}^{(i, j)}+\sum_{i=0}^{\infty} \sum_{j=L+1}^{S} \lambda_{L} \phi_{1}^{(i, j)}
$$

Expected waiting time. The expected waiting time of a customer is obtained by Little's formula

$$
\Theta_{6}=\frac{\Theta_{4}}{\Theta_{5}}
$$

Expected number of customers lost. The customer loss in the system occurs only at the time of zero inventory level. It is defined by

$$
\Theta_{7}=\sum_{i=0}^{\infty} \lambda_{0} \phi_{1}^{(i, 0)}
$$

### 3.7. Construction of cost function

The expected total cost of the proposed model under the $(s, Q)$ ordering principle is constructed by the cost function,

$$
T c=a_{1} \Theta_{1}+a_{2} \Theta_{2}+a_{3} \Theta_{3}+a_{4} \Theta_{4}+a_{5} \Theta_{7}
$$

where $a_{1}$ refers to holding cost per item in the system, $a_{2}$ refers to set up cost per order, $a_{3}$ denotes perishable cost per item, $a_{4}$ indicates waiting cost per customer in the system, and $a_{5}$ refers to lost cost per customer in the system.

## 4. Model-II

This section describes the MQIS with the $(s, S)$ ordering principle.

### 4.1. Construction of matrices

The CTMC has an infinitesimal generator matrix $P^{\prime}$ as follows:
where

$$
\begin{aligned}
& A_{i, i}^{\prime}= \begin{cases}j \gamma, & i^{\prime}=i, i=0,1,2, \ldots, k \\
& j^{\prime}=j-1, j=1,2, \ldots, S \\
\beta & i^{\prime}=i, i=0,1,2, \ldots, k \\
& j^{\prime}=S, j=0,1,2, \ldots, s \\
-\beta & i^{\prime}=i, i=0,1,2, \ldots, k \\
& j^{\prime}=j, j=0 \\
-\left(\bar{\delta}_{i 0}\left(\mu_{i}+j \gamma\right)+H(s-j) \beta+\lambda_{j}\right) & i^{\prime}=i, i=0,1,2, \ldots, k \\
& j^{\prime}=j, j=1,2, \ldots, L \\
-\left(\bar{\delta}_{i 0}\left(\mu_{i}+j \gamma\right)+\lambda_{L}\right) & i^{\prime}=i, i=0,1,2, \ldots, k \\
0 & j^{\prime}=j, j=L+1, L+2, \ldots, S \\
0 & \text { otherwise }\end{cases} \\
& A_{i, i}^{\prime}= \begin{cases}h_{j}^{\prime}, & i^{\prime}=i, i=k, k+1, \ldots \\
& j^{\prime}=j-1, j=1,2, \ldots, S \\
\beta \quad & i^{\prime}=i, i=k, k+1, \ldots \\
j^{\prime}=S, j=0,1,2, \ldots, s \\
f_{j}^{\prime} \quad & i^{\prime}=i, i=k, k+1, \ldots \\
& j^{\prime}=j, j=0,1,2, \ldots, S \\
0 \quad & \text { otherwise }\end{cases}
\end{aligned}
$$

where $f_{j}^{\prime}=\left\{\begin{array}{ll}-\left(\bar{\delta}_{j 0}\left(\mu_{k}+j \gamma+\lambda_{j}\right)+H(s-j) \beta\right), & \text { if } j=0,1, \ldots, L \\ -\left(\mu_{k}+j \gamma+\lambda_{L}+H(s-j) \beta\right), & \text { if } j=L+1, L+2, \ldots, S\end{array}\right.$ and $h_{j}^{\prime}=j \gamma$.

$$
A_{H}= \begin{cases}h_{j}, & i^{\prime}=i-1, i=k \\ & j^{\prime}=j-1, j=1,2, \ldots, S \\ \beta & i^{\prime}=i, i=k \\ j^{\prime}=S, j=0,1,2, \ldots, s \\ f_{j} & i^{\prime}=i, i=k \\ & j^{\prime}=j, j=0,1, \ldots, S \\ 0 & \text { otherwise }\end{cases}
$$

where $f_{j}=-\left(\bar{\delta}_{j 0}\left(\mu_{i}+j \gamma\right)+H(s-j) \beta\right)$ and $h_{j}=p \mu_{k}+j \gamma$.

Lemma 3. The stationary probability vector $\Pi_{2}=\left(\pi_{2}^{(0)}, \pi_{2}^{(1)}, \ldots, \pi_{2}^{(S)}\right)$ to the generator matrix, $A_{H}$ is determined by

$$
\pi_{2}^{(j)}=\pi_{2}^{(0)} \Lambda_{j}, \quad j=0,1,2, \ldots, S
$$

where

$$
\Lambda_{j}= \begin{cases}1 & j=0 \\ \frac{(-1)^{j} \prod_{z=0}^{j-1} f_{z}}{\prod_{\substack{z=1}}^{j} h_{z}}, & j=1,2, \ldots, S-1 \\ -\frac{\sum_{z=0}^{s} \lambda_{z} \beta}{f_{j}}, & j=S\end{cases}
$$

Proof. Let $A_{H}=A_{k k-1}+A_{k k}^{\prime}+A_{01}$ and solving $\Pi A_{H}=\mathbf{0}$, we get

$$
\begin{align*}
\pi_{2}^{(j)} f_{j}+\pi_{2}^{(j+1)} h_{j+1}=0, \quad j=0,1, \ldots, S-1  \tag{7}\\
\sum_{z=0}^{s} \pi_{2}^{(z)} \beta+\pi_{2}^{(j)} f_{j}=0, \quad j=S \tag{8}
\end{align*}
$$

By solving the above system of equations recursively, we get the stated result.

Lemma 4. The stability condition of the system is

$$
\begin{equation*}
\mu_{k} \sum_{j=1}^{S} \Lambda_{j}>\sum_{j=1}^{L} \Lambda_{j} \lambda_{j}+\lambda_{L} \sum_{j=L+1}^{S} \Lambda_{j} \tag{9}
\end{equation*}
$$

Proof. Using the Neuts result for the stability condition on

$$
\begin{equation*}
\pi_{2} A_{k k-1} \mathbf{e}>\pi_{2} A_{01} \mathbf{e} \tag{10}
\end{equation*}
$$

writing it explicitly we get L.H.S as $\pi_{2}^{(0)} \sum_{j=1}^{S} \Lambda_{j} \mu_{k}$ and R.H.S as $\pi_{2}^{(0)}\left(\sum_{j=1}^{L} \Lambda_{j} \lambda_{j}+\sum_{j=L+1}^{S} \Lambda_{j} \lambda_{L}\right)$.
Substituting the L.H.S and R.H.S on (10), we obtain

$$
\mu_{k} \sum_{j=1}^{S} \Lambda_{j}>\sum_{j=1}^{L} \Lambda_{j} \lambda_{j}+\lambda_{L} \sum_{j=L+1}^{S} \Lambda_{j}
$$

### 4.2. Computation of $R_{2}$ matrix

Due to the structure of the generator matrix and stationary probability vector, $R_{2}$ matrix can be determined which satisfies the matrix equation

$$
\begin{equation*}
R_{2}^{2} A_{k k-1}+R_{2} A_{k k}^{\prime}+A_{01}=\mathbf{0} \tag{11}
\end{equation*}
$$

where

$$
R_{2}=\begin{gathered}
\\
0 \\
1 \\
2 \\
\vdots \\
S
\end{gathered}\left(\begin{array}{cccc}
0 & 1 & \ldots & S \\
0 & 0 & \ldots & 0 \\
y_{10} & y_{11} & \ldots & y_{1 S} \\
y_{20} & y_{20} & \ldots & y_{2 S} \\
\vdots & \vdots & \ldots & \vdots \\
y_{S 0} & y_{S 1} & \ldots & y_{S S}
\end{array}\right)
$$

Substituting $R_{2}$ in (11), we get the following set of equations. For $j=1,2, \ldots, S$

$$
\begin{aligned}
x_{j j^{\prime}} p \mu_{k}+y_{j j^{\prime}} f_{j}^{\prime}+y_{j j^{\prime}+1} h_{j+1}^{\prime}=0, & \text { if } j^{\prime}=0 \\
x_{j j^{\prime}} q \mu_{k}+x_{j j^{\prime}+1} p \mu_{k}+y_{j j^{\prime}} f_{j}^{\prime}+y_{j j^{\prime}+1} h_{j+1}^{\prime}+\lambda_{j} \delta_{j j^{\prime}}=0, & \text { if } j^{\prime}=1,2, \ldots, L \\
x_{j j^{\prime}} q \mu_{k}+x_{j j^{\prime}+1} p \mu_{k}+y_{j j^{\prime}} f_{j}^{\prime}+y_{j j^{\prime}+1} h_{j+1}^{\prime}+\lambda_{L} \delta_{j j^{\prime}}=0, & \text { if } j^{\prime}=L+1, L+2, \ldots, S-1 \\
x_{j j^{\prime}} q \mu_{k}+\sum_{z=0}^{s} y_{j z} \beta+y_{j j^{\prime}} f_{j}^{\prime}+\lambda_{L} \delta_{j j^{\prime}}=0, & \text { if } j^{\prime}=S .
\end{aligned}
$$

Solving the above system of non-linear equations by the Gauss-Seidal iterative process, we will obtain the $R_{2}$ matrix.

### 4.3. Partition of the steady state vector

The partition of the steady-state probability vector of the system is defined as follows:

$$
\begin{gathered}
\phi_{2}=\left(\phi_{2}^{(0)}, \phi_{2}^{(1)}, \phi_{2}^{(2)}, \ldots\right) \\
\phi_{2}^{(i)}=\left(\phi_{2}^{(i, 0)}, \phi_{2}^{(i, 1)}, \phi_{2}^{(i, 2)}, \ldots, \phi_{2}^{(i, S)}\right), \quad i=0,1,2, \ldots
\end{gathered}
$$

### 4.4. Computation of steady state probability vector

The entire probability vector of all system states is $\phi_{2}=\left(\phi_{2}^{(0)}, \phi_{2}^{(1)}, \phi_{2}^{(2)}, \ldots\right)$. The system balance equations are given by $\phi_{2} P^{\prime}=\mathbf{0}$ and $\phi_{2} \mathbf{e}=1$. Then, the steady-state probabilities of the queuing-inventory system are calculated [23]) by $\phi_{2}^{(i)}=\phi_{2}^{(k)} R_{2}^{(i-k)}$ where $i=k+1, k+2, \ldots$ and the initial conditions are represented by the vectors $\phi_{2}^{(i)}=0,1, \ldots, k$ are obtained by solving part of the balance equations

$$
\begin{aligned}
\phi_{2}^{(0)} A_{00}^{\prime}+\phi_{2}^{(1)} A_{10} & =\mathbf{0} \\
\phi_{2}^{(0)} A_{01}+\phi_{2}^{(1)} A_{11}^{\prime}+\phi_{2}^{(2)} A_{21} & =\mathbf{0} \\
\phi_{2}^{(i-1)} A_{01}+\phi_{2}^{(i)} A_{i i}^{\prime}+\phi_{2}^{(i+1)} A_{i+1, i} & =\mathbf{0} \quad i=2,3, \ldots, k-1, \\
\phi_{2}^{(k-1)} A_{01}+\phi_{2}^{(k)}\left(A_{1}^{\prime}+R_{2} A_{2}\right) & =\mathbf{0} \\
\sum_{n=0}^{k-1} \phi_{2}^{(n)} \mathbf{e}+\phi_{2}^{(k)}\left[I-R_{2}\right] \mathbf{e} & =1 .
\end{aligned}
$$

### 4.5. System performance measures

The system performance of the model under the $(s, S)$ ordering principle is determined as follows:
Expected inventory level. In the MQIS, the expected inventory level of the system is obtained by the sum of the product value of the current inventory level and the stationary probability vector defined by

$$
\aleph_{1}=\sum_{i=0}^{\infty} \sum_{j=1}^{S} j \phi_{2}^{(i, j)}
$$

Expected reorder rate. When the present inventory level reduces to $s+1$, there can either be a service completion happened or an item become defective. In the case of either of these two occurrences, the replenishment process is immediately triggered. It is defined by

$$
\aleph_{2}=\sum_{i=1}^{k} p \mu_{i} \phi_{2}^{(i, s+1)}+\sum_{i=k+1}^{\infty} p \mu_{k} \phi_{2}^{(i, s+1)}+\sum_{i=0}^{\infty}(s+1) \gamma \phi_{2}^{(i, s+1)}
$$

Expected perishable rate. Since the system may have imperfect items in the storage space, we require an expected perishable rate of the system. This could be done using the sum of the product $j \gamma$ and the stationary probability vector where $j$ is the current inventory level. It is defined by

$$
\aleph_{3}=\sum_{i=0}^{\infty} \sum_{j=1}^{S} j \gamma \phi_{2}^{(i, j)}
$$

Expected number of customers in the system. All customers in the system purchase an item through the first come first serve discipline. The expected number of customers in the system is the sum of the product value of the number of customers present and the stationary probability vector. It is defined by

$$
\aleph_{4}=\sum_{i=1}^{\infty} \sum_{j=0}^{S} i \phi_{2}^{(i, j)}
$$

The expected arrival rate of a customer enters into the system. The sum of the product of the average arrival rate of a customer and the stationary probability vector is used to define the expected arrival rate of a customer in the system

$$
\aleph_{5}=\sum_{i=0}^{\infty} \sum_{j=1}^{L} \lambda_{j} \phi_{2}^{(i, j)}+\sum_{i=0}^{\infty} \sum_{j=L+1}^{S} \lambda_{L} \phi_{2}^{(i, j)}
$$

Expected waiting time. The expected waiting time of a customer is obtained by Little's formula

$$
\aleph_{6}=\frac{\aleph_{4}}{\aleph_{5}}
$$

Expected number of customers lost. The customer loss in the system occurs only at the time of zero inventory level. It is defined by

$$
\aleph_{7}=\sum_{i=0}^{\infty} \lambda_{0} \phi_{2}^{(i, 0)}
$$

Construction of cost function. The expected total cost of the proposed model under the $(s, S)$ ordering principle is constructed by the cost function

$$
T c=a_{1} \aleph_{1}+a_{2} \aleph_{2}+a_{3} \aleph_{3}+a_{4} \aleph_{4}+a_{5} \aleph_{7}
$$

## 5. Numerical discussion

In this section, we explore the proposed system with the $(s, Q)$ and $(s, S)$ ordering principles by numerical discussion. In such a way, this section explains the total expected cost of the system, the mean number of customers in the waiting hall, the mean number of customers lost, and the mean waiting time of customers in the waiting hall, which is to be discussed for M-I and M-II along with the scaling factors $k_{1}$ and $k_{2}$. We use the scaling factor $k_{1}$ as the controlling factor of a non-homogeneous arrival rate. Similarly, we use $k_{2}$ as the controlling factor on the non-homogeneous service rate. On applying those $k_{1}$ and $k_{2}$, we have given the generalized model. This is because, if we assume $k_{1}=k_{2}=0$, this model becomes a purely homogeneous arrival and service rate of the system. Each illustration is explained for homogeneous and non-homogeneous arrival/service rates. The major objective of this section is to find the best ordering principle for the queuing-inventory setup. For this numerical work, we fixed the cost rates as $S=30, s=6, a_{1}=0.001, a_{2}=1, a_{3}=0.02, a_{4}=1.1, a_{5}=3$, as well as the rate of parameter $\lambda_{0}=0.4, k_{1}=0.2, k_{2}=0.2, \gamma=3, \beta=4.9, \lambda=1.3, \mu=12.8, k=11, L=7, p=0.8, q=0.2, r=k$.

The monotonicity of the parameters is to be assumed for the following examples as follows:

- $\lambda$ and $k_{1}$ increase; the average number of customers entering the system is increased.
- $\beta$ increases; the average replenishment time is reduced.
- $\gamma$ increases; the average lifetime of a product is reduced.
- $\mu$ and $k_{2}$ increase; the average service time per customer is reduced.


### 5.1. Example I

In this example, we investigate the expected total cost by varying the parameters $\gamma, \beta, \lambda, \mu$ and also varying the two distinct scaling factors $k_{1}$ and $k_{2}$ (Tables 1 and 2). In Table 1 , we mainly present the scaling factor $k_{1}$, and how it impacts the expected total cost by varying different parameters. Similarly, Table 2 shows the effect of the scaling factor $k_{2}$ on the expected total cost by varying parameters.

- The expected total cost for M-I and M-II decreases in proportion to the increase in $\mu$ for every increment in $k_{1}$. This is because the average service time is reduced.
- Both $\lambda$ and $k_{1}$ increase, the expected total cost is increased and also when we compared M-II to M-I (Table 1), M-II gave the minimum ETC. This is because M-II's ordering policy depends on the current stock level which could mean that the mean inventory level is much higher in M-II.
- When the value of $\gamma$ is increasing, the total cost rate increases, and the total cost rate decrease as $\beta$ increases. In this case, the ETC of the M-II is lower than that of the M-I. As $\gamma$ increases, the defective items in the inventory also increase. So that total cost will increase. When $\beta$ increases, the average time for replenishment decreases. Hence the total cost of the system reduces.

[^0]- The total expected cost rate decreases when the scaling vector $k_{2}$ increases. And compared to M-I and M-II, M-II has a minimum expected cost value in Table 2. The scaling factor $k_{2}$ reduces the service completion time per customer when it increases. Therefore total cost rate is decreased.
- As $\mu$ and $\beta$ increase, the total cost decreases as $k_{2}$ increases. When $\lambda$ and $\gamma$ increase, the total cost also increases (Table 2). The increase of $\lambda$ will cause an increase in customers in the system. The waiting cost of each customer in the system reflects the increase in total cost. In this case, M-II has a minimum value of expected total cost compared to M-I.
- From Table 11, we conclude that the expected total cost of the system is high for M-I and low for M-II under the parameter variation $S, s$, and $L$. When we increase the inventory level of a system, the expected total cost is increased due to the holding cost. Similarly, the increment in $L$ causes the number of customers in the system. Thus the expected total cost is increased.
This example suggests that the firm has to maintain enough service speed or reorder time to increase profit as well as decrease TC while the number of arriving customers increases and the life of products decreases.


### 5.2. Example II

In this example, the mean number of customers in the waiting hall is shown by different parameters $\lambda, \mu, \beta, \gamma$ with the scaling factor for arrival $\left(k_{1}\right)$ in Table 3 and the scaling factor for the service $\left(k_{2}\right)$ in Table 4. In addition, we compared the M-I and M-II in Tables 3 and 4.

- The mean number of customers in the system is increased when $k_{1}$ increases because $k_{1}$ causes the increase of arriving customers in the system. The mean number of customers in the waiting hall for M-I is minimum as compared to M-II in Table 3.
- When the service rate $\mu_{2}$ and the lead time rate $\beta$ with scaling factor $k_{1}$ increases then the mean number of customers in the waiting hall is decreased. And the mean number of customers in the waiting hall increases when the arrival rate $\lambda$ and the perishable rate $\gamma$ increases (Table 3). We observe that the service time and lead time will reduce the number of customers in the system when they reduce. In the system, a defective item causes a shortage in the inventory. So customers will face insufficient stock levels in the system. Hence the number of customers in the system will increase.
- Table 4 displayed when the scaling factor for service $k_{2}$ increases the mean number of customers in the system decreases. Therefore, the number of customers in the system occurs low for M-II because it helps to reduce the average service time per customer. So that customer in the system quickly reduces when $k_{2}$ increases.
- The mean number of customers in the system decreases when $\mu$ and $\beta$ increase. And when $\lambda$ and $\gamma$ increase, the mean number of customers in the system increases (Table 4). When we control the defective items in the inventory, we can give service to the customers quickly. If this is to happen in the system, the mean number of customers in the system can be reduced.

From this example, the queuing-inventory-based seller can control the crowd in the queue easily by increasing the service time or reorder time or else increasing the life of the products, so that customers will get service.

### 5.3. Example III

In this example, we investigate the influence of parameters $\beta$ and $\gamma$ with the scaling factors $k_{1}$ and $k_{2}$ on mean waiting time for M-I and M-II.

- Figure 1 shows that the mean waiting time decreases when $\beta$ and the mean waiting time increases when $k_{1}$ increases for both M-I and M-II. Figure 2 demonstrates that the mean waiting time decreases when both $\beta$ and $k_{2}$ increase for both M-I and M-II.


Figure 1. $\Theta_{6}$ of M-I on $k_{1}$ vs. $\beta$ (a), and $\aleph_{6}$ of M-II on $k_{1}$ vs. $\beta$ (b)


Figure 2. $\Theta_{6}$ of M-I on $k_{2}$ vs. $\beta$ (a), and $\aleph_{6}$ of M-II on $k_{2}$ vs. $\beta$ (b)

- Figure 3 shows that the mean waiting time increases when $\gamma$ and $k_{1}$ increase for both M-I and M-II. Furthermore, when the mean waiting time decreases (increases) when $k_{2}(\gamma)$ increases for both M-I and M-II (Figure 4).
- Figure 5 demonstrates that when $\lambda$ increases, the mean waiting time for both M-I and M-II increases. And the mean waiting time for both M-I and M-II increases when $k_{1}$ increases. Next, the mean waiting time for both M-I and M-II decreases when $k_{2}$ increases (Figure 6).
- Figure 7 shows that the mean waiting time for both M-I and M-II decreases when $\mu$ increases and when $k_{1}$ increases the mean waiting time for both M-I and M-II increases. Figure 8 shows that the mean waiting time for both $\mathrm{M}-\mathrm{I}$ and $\mathrm{M}-\mathrm{II}$ decreases when $k_{2}$ increases.
- Table 11 shows that the mean waiting time per customer increases for M-II rather than M-I under the parameter variation $S, s$, and $L$.


Figure 3. $\Theta_{6}$ of M-I on $k_{1}$ vs. $\gamma(\mathrm{a})$, and $\aleph$ of M-II on $k_{1}$ vs. $\gamma(\mathrm{b})$



Figure 4. $\Theta_{6}$ of M-I on $k_{2}$ vs. $\gamma(\mathrm{a})$, and $\aleph_{6}$ of M-II on $k_{2}$ vs. $\gamma(\mathrm{b})$
We are witnessing from this example that if a seller provides good service or maintains a good ordering time, the customer's waiting time will decrease, which tends them to visit the shop more times.

### 5.4. Example IV

This example explores the mean number of customers lost for both M-I and M-II using different parameters $\lambda, \mu, \beta, \gamma$ with the scaling factors $k_{1}$ and $k_{2}$. Additionally, we also compare the M-I and M-II.


Figure 5. $\Theta_{6}$ of M-I on $k_{1}$ vs. $\lambda$ (a), and $\aleph_{6}$ of M-II on $k_{1}$ vs. $\lambda$ (b)



Figure 6. $\Theta_{6}$ of M-I on $k_{2}$ vs. $\lambda(\mathrm{a})$, and $\aleph_{6}$ of M-II on $k_{2}$ vs. $\lambda$ (b)

- When $k_{1}$ increases, the mean number of lost customers increases. In addition, M-II has a lower mean number of customers lost than M-I in Tables 5 and 6.
- When $\mu$ and $\beta$ increase then the mean number of customers lost decreases and the mean number of customers lost increases when $\lambda$ and $\gamma$ increases with $k_{1}$ increases ( $k_{2}$ increases) in Table 5 (6).
- Tables 7-10 show the expected inventory and expected perishable for both M-I and M-II decreases when $\mu$ increases with $k_{1}$ and $k_{2}$ respectively. Table 7 and 8 demonstrate the expected inventory for both M-I and M-II decreases when $\gamma, \lambda, k_{1}$ and $k_{2}$ increases. But the expected inventory for both M-I and M-II increases when $\beta$ increases in Table 7 and 8.
- The expected perishable for both M-I and M-II increases when $\beta$ and $\gamma$ increase, but the expected perishable for both M-I and M-II decreases when $\lambda, k_{1}$ and $k_{2}$ increase in Table 9 and 10. Furthermore, M-I is minimum compared to M-II for Tables 7-10.
- In Table 11, we show how the parameters $S, s$, and $L$ influence the customer loss of the system. The customer loss is higher in M-II and lower in M-I for the increment of $S, s$, and $L$.


Figure 7. $\Theta_{6}$ of M-I on $k_{1}$ vs. $\mu(\mathrm{a})$, and $\aleph_{6}$ of M-II on $k_{1}$ vs. $\mu$ (b)



Figure 8. $\Theta_{6}$ of M-I on $k_{2}$ vs. $\mu(\mathrm{a})$, and $\aleph_{6}$ of M-II on $k_{2}$ vs. $\mu(\mathrm{b})$
The loss of customers crucially affects the profit of the business. To increase the profit, it is essential to decrease the number of lost customers. It is possible by increasing the reorder time or speed of the service facility or storage capacity or by decreasing the queue size or perishable rate.

## 6. Conclusion

The single server service channel of the MQIS investigated the SDAP and QDSP with an infinite queue size. The proposed system is a generalised version of the homogeneous and non-homogeneous arrival and service rates, respectively. Mostly, in the existing literature, an inventory system, the discussion with SDAP and QDSP is separate. However, this paper fills such a research gap with an infinite queue in an MQIS. Among these policies, the proposed MQIS deals with two different types of ordering principles. Efficient attention is given to each ordering principle in order to explore and bring managerial inputs to the inventory business. The impact of the parameter variation and the assumption of the ordering
principles gave significant results in the numerical section. The output factors will enhance the proposed MQIS economically for every business tycoon. In the future, this model will be extended to a multi-server MQIS. In each of the numerical outputs, we gave the comparison results for both models. In addition, the results are obtained for both homogeneous and non-homogeneous arrival/service rates. According to the limitations of the scaling factor, homogeneous and non-homogeneous classifications are made in each discussion. When we observe the total expected cost of the system, the optimum cost is obtained in M-II. That is, the $(s, S)$ ordering principle set up in a queuing-inventory system produces the minimum total cost rather than the $(s, Q)$ ordering principle. Due to the stock-dependent arrival policy, the mean number of customers in the waiting hall increases in M-II. In the case of the mean number of customers lost in the system, this is mostly reduced for M-II.

In the future, this proposed work will be discussed using the multi-server service facility.

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Table 1. Effect of various parameters with $k_{1}$ on the total expected cost


Table 2. Different parameter effect with $k_{2}$ on Total Expected Cost


Table 3. Different parameter effect with $k_{1}$ on Mean Number of Customers in the Waiting hall


Table 4. Different parameter effect with $k_{2}$ on Mean Number of Customers in the Waiting hall


Table 5. Mean number of customers lost for different parameters with $k_{1}$


Table 6. Mean number of customers lost for different parameters with $k_{2}$


Table 7. Effect of different parameters with $k_{1}$ on $\Theta_{1}$ and $\aleph_{1}$

| $k_{1}$ | $\gamma$ | $\beta$ | $\mu$$\lambda$ | 12.5 |  | 13 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  | M-I | M-II | M-I | M-II |
| 0 | 3 | 4.9 | 1.3 | 11.643509 | 11.978372 | 11.631575 | 11.953363 |
|  |  |  | 1.4 | 11.641260 | 11.958263 | 11.632533 | 11.930972 |
|  |  |  | 1.5 | 11.635265 | 11.935409 | 11.629857 | 11.905908 |
|  |  | 5 | 1.3 | 11.677983 | 12.013947 | 11.668481 | 11.988261 |
|  |  |  | 1.4 | 11.672968 | 11.993086 | 11.666719 | 11.965172 |
|  |  |  | 1.5 | 11.664200 | 11.969527 | 11.661320 | 11.939454 |
|  | 3.5 | 4.9 | 1.3 | 11.407928 | 11.769787 | 11.404296 | 11.759722 |
|  |  |  | 1.4 | 11.397780 | 11.766476 | 11.397137 | 11.753861 |
|  |  |  | 1.5 | 11.384333 | 11.760468 | 11.386783 | 11.745413 |
|  |  | 5 | 1.3 | 11.445804 | 11.784400 | 11.444268 | 11.769062 |
|  |  |  | 1.4 | 11.433539 | 11.775281 | 11.435059 | 11.757684 |
|  |  |  | 1.5 | 11.417910 | 11.763716 | 11.422601 | 11.743936 |
| 0.5 | 3 | 4.9 | 1.3 | 11.587238 | 11.958756 | 11.583968 | 11.944571 |
|  |  |  | 1.4 | 11.573839 | 11.941275 | 11.563769 | 11.924666 |
|  |  |  | 1.5 | 11.556629 | 11.910899 | 11.549830 | 11.901913 |
|  |  | 5 | 1.3 | 11.634680 | 11.937981 | 11.627600 | 11.883069 |
|  |  |  | 1.4 | 11.626055 | 11.921611 | 11.622047 | 11.872333 |
|  |  |  | 1.5 | 11.613700 | 11.918399 | 11.612827 | 11.858802 |
|  | 3.5 | 4.9 | 1.3 | 11.329030 | 11.699264 | 11.324573 | 11.694649 |
|  |  |  | 1.4 | 11.311321 | 11.693515 | 11.303943 | 11.691329 |
|  |  |  | 1.5 | 11.290350 | 11.685035 | 11.279981 | 11.684356 |
|  |  | 5 | 1.3 | 11.351214 | 11.740148 | 11.344664 | 11.734722 |
|  |  |  | 1.4 | 11.331290 | 11.735449 | 11.321812 | 11.732386 |
|  |  |  | 1.5 | 11.308125 | 11.728070 | 11.295647 | 11.727448 |
| 1 | 3 | 4.9 | 1.3 | 11.565393 | 11.712361 | 11.544733 | 11.608410 |
|  |  |  | 1.4 | 11.547972 | 11.692887 | 11.525011 | 11.586774 |
|  |  |  | 1.5 | 11.497765 | 11.570832 | 11.472479 | 11.563130 |
|  |  | 5 | 1.3 | 11.600881 | 11.901692 | 11.597011 | 11.827743 |
|  |  |  | 1.4 | 11.590723 | 11.852619 | 11.578604 | 11.807160 |
|  |  |  | 1.5 | 11.551815 | 11.751607 | 11.527422 | 11.744611 |
|  | 3.5 | 4.9 | 1.3 | 11.495393 | 11.112961 | 11.474733 | 11.108410 |
|  |  |  | 1.4 | 11.447972 | 11.092887 | 11.425011 | 11.086774 |
|  |  |  | 1.5 | 11.397765 | 11.070832 | 11.372479 | 11.063130 |
|  |  | 5 | 1.3 | 11.556811 | 11.800256 | 11.531353 | 11.790498 |
|  |  |  | 1.4 | 11.550566 | 11.774972 | 11.478097 | 11.763755 |
|  |  |  | 1.5 | 11.452093 | 11.747920 | 11.422409 | 11.735215 |

Table 8. Effect of different parameters with $k_{2}$ on $\Theta_{1}$ and $\aleph_{1}$

| $k_{2}$ | $\gamma$ | $\beta$ $\begin{gathered}\mu \\ \lambda\end{gathered}$ |  | 12.5 |  | 13 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  | M-I | M-II | M-I | M-II |
| 0 | 3 | 4.9 | 1.3 | 12.273132 | 12.729638 | 12.223139 | 12.667039 |
|  |  |  | 1.4 | 12.265706 | 12.717481 | 12.215035 | 12.654215 |
|  |  |  | 1.5 | 12.258121 | 12.705189 | 12.206772 | 12.641263 |
|  |  | 5 | 1.3 | 12.286323 | 12.737340 | 12.236624 | 12.675195 |
|  |  |  | 1.4 | 12.278776 | 12.725092 | 12.228395 | 12.662276 |
|  |  |  | 1.5 | 12.271071 | 12.712707 | 12.220006 | 12.649227 |
|  | 3.5 | 4.9 | 1.3 | 11.963606 | 12.398156 | 11.920386 | 12.341148 |
|  |  |  | 1.4 | 11.955403 | 12.385777 | 11.911545 | 12.328108 |
|  |  |  | 1.5 | 11.947072 | 12.373305 | 11.902571 | 12.314976 |
|  |  | 5 | 1.3 | 11.977397 | 12.406890 | 11.934491 | 12.350366 |
|  |  |  | 1.4 | 11.969082 | 12.394423 | 11.925534 | 12.337234 |
|  |  |  | 1.5 | 11.960637 | 12.381862 | 11.916442 | 12.324009 |
| 0.5 | 3 | 4.9 | 1.3 | 11.538341 | 11.989785 | 11.518496 | 11.986791 |
|  |  |  | 1.4 | 11.514712 | 11.968154 | 11.494465 | 11.964678 |
|  |  |  | 1.5 | 11.489563 | 11.945538 | 11.469089 | 11.941561 |
|  |  | 5 | 1.3 | 11.564089 | 12.053259 | 11.543576 | 12.050267 |
|  |  |  | 1.4 | 11.540045 | 12.032114 | 11.519138 | 12.028639 |
|  |  |  | 1.5 | 11.514489 | 12.009973 | 11.493361 | 12.005994 |
|  | 3.5 | 4.9 | 1.3 | 11.370195 | 11.518305 | 11.351242 | 11.515628 |
|  |  |  | 1.4 | 11.352890 | 11.496735 | 11.332821 | 11.493677 |
|  |  |  | 1.5 | 11.333995 | 11.474427 | 11.312933 | 11.470974 |
|  |  | 5 | 1.3 | 11.397107 | 11.586001 | 11.377529 | 11.583320 |
|  |  |  | 1.4 | 11.379399 | 11.564838 | 11.358721 | 11.561774 |
|  |  |  | 1.5 | 11.360115 | 11.542926 | 11.338456 | 11.539465 |
| 1 | 3 | 4.9 | 1.3 | 11.212146 | 11.761392 | 11.206852 | 11.741896 |
|  |  |  | 1.4 | 11.177145 | 11.728519 | 11.171214 | 11.709574 |
|  |  |  | 1.5 | 11.141009 | 11.694759 | 11.134432 | 11.676530 |
|  |  | 5 | 1.3 | 11.282098 | 11.785546 | 11.276803 | 11.765213 |
|  |  |  | 1.4 | 11.247558 | 11.752286 | 11.241624 | 11.732502 |
|  |  |  | 1.5 | 11.211868 | 11.718138 | 11.205283 | 11.699064 |
|  | 3.5 | 4.9 | 1.3 | 10.710175 | 11.589218 | 10.705675 | 11.567176 |
|  |  |  | 1.4 | 10.677162 | 11.561620 | 10.672166 | 11.539326 |
|  |  |  | 1.5 | 10.643320 | 11.532969 | 10.637824 | 11.510539 |
|  |  | 5 | 1.3 | 10.783777 | 11.614952 | 10.779267 | 11.592156 |
|  |  |  | 1.4 | 10.751126 | 11.587001 | 10.746118 | 11.563956 |
|  |  |  | 1.5 | 10.717631 | 11.558001 | 10.712118 | 11.534822 |

Table 9. Effect of different parameters with $k_{1}$ on $\Theta_{3}$ and $\aleph_{3}$

| $k_{1}$ | $\gamma$ | $\beta$ $\begin{gathered}\mu \\ \lambda\end{gathered}$ |  | 12.5 |  | 13 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  | M-I | M-II | M-I | M-II |
| 0 | 3 | 4.9 | 1.3 | 34.930526 | 35.935117 | 34.894725 | 35.860089 |
|  |  |  | 1.4 | 34.923779 | 35.874790 | 34.897599 | 35.792915 |
|  |  |  | 1.5 | 34.905795 | 35.806226 | 34.889572 | 35.717724 |
|  |  | 5 | 1.3 | 35.033950 | 36.041842 | 35.005442 | 35.964784 |
|  |  |  | 1.4 | 35.018904 | 35.979257 | 35.000156 | 35.895515 |
|  |  |  | 1.5 | 34.992601 | 35.908580 | 34.983960 | 35.818362 |
|  | 3.5 | 4.9 | 1.3 | 39.951966 | 41.194254 | 39.929131 | 41.159026 |
|  |  |  | 1.4 | 39.927749 | 41.182665 | 39.915034 | 41.138514 |
|  |  |  | 1.5 | 39.892229 | 41.161638 | 39.889979 | 41.108946 |
|  |  | 5 | 1.3 | 40.054939 | 41.245400 | 40.039669 | 41.191716 |
|  |  |  | 1.4 | 40.022708 | 41.213485 | 40.017686 | 41.151894 |
|  |  |  | 1.5 | 39.979103 | 41.173006 | 39.984687 | 41.103775 |
| 0.5 | 3 | 4.9 | 1.3 | 34.090550 | 35.176269 | 34.071333 | 35.133712 |
|  |  |  | 1.4 | 34.061714 | 35.153826 | 34.051904 | 35.103997 |
|  |  |  | 1.5 | 34.021517 | 35.122698 | 34.021307 | 35.065738 |
|  |  | 5 | 1.3 | 34.178164 | 35.293943 | 34.166141 | 35.249206 |
|  |  |  | 1.4 | 34.141100 | 35.268832 | 34.138482 | 35.217000 |
|  |  |  | 1.5 | 34.092773 | 35.235196 | 34.089768 | 35.176405 |
|  | 3.5 | 4.9 | 1.3 | 39.946004 | 41.097425 | 39.911606 | 41.081272 |
|  |  |  | 1.4 | 39.913801 | 41.077304 | 39.900624 | 41.069653 |
|  |  |  | 1.5 | 39.829933 | 41.047622 | 39.866224 | 41.048746 |
|  |  | 5 | 1.3 | 40.039248 | 41.225041 | 40.026325 | 41.161473 |
|  |  |  | 1.4 | 40.009516 | 41.194902 | 39.976341 | 41.129161 |
|  |  |  | 1.5 | 39.928438 | 41.145881 | 39.884764 | 41.101157 |
| 1 | 3 | 4.9 | 1.3 | 33.736179 | 34.238483 | 33.628200 | 34.125229 |
|  |  |  | 1.4 | 33.613917 | 34.179661 | 33.475034 | 34.060322 |
|  |  |  | 1.5 | 33.433294 | 34.119496 | 33.312437 | 34.009391 |
|  |  | 5 | 1.3 | 33.940644 | 34.335075 | 33.881034 | 34.273230 |
|  |  |  | 1.4 | 33.802168 | 34.201856 | 33.735811 | 34.111481 |
|  |  |  | 1.5 | 33.655445 | 34.188648 | 33.582265 | 34.043834 |
|  | 3.5 | 4.9 | 1.3 | 33.786179 | 34.338883 | 33.724200 | 34.325229 |
|  |  |  | 1.4 | 33.643917 | 34.278661 | 33.575034 | 34.260322 |
|  |  |  | 1.5 | 33.493294 | 34.212496 | 33.417437 | 34.189391 |
|  |  | 5 | 1.3 | 33.998839 | 34.380896 | 33.909735 | 34.566743 |
|  |  |  | 1.4 | 33.819815 | 34.312402 | 33.623341 | 34.373144 |
|  |  |  | 1.5 | 33.732326 | 34.310718 | 33.528430 | 34.273253 |

Table 10. Effect of different parameters with $k_{2}$ on $\Theta_{3}$ and $\aleph_{3}$

| $k_{2}$ | $\gamma$ | $\beta$ | $\begin{aligned} & \hline \mu \\ & \lambda \end{aligned}$ | 12.5 |  | 13 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  | M-I | M-II | M-I | M-II |
| 0 | 3 | 4.9 | 1.3 | 36.819397 | 38.188913 | 36.669418 | 38.001116 |
|  |  |  | 1.4 | 36.797117 | 38.152442 | 36.645105 | 37.962646 |
|  |  |  | 1.5 | 36.774364 | 38.115567 | 36.620316 | 37.923790 |
|  |  | 5 | 1.3 | 36.858969 | 38.212021 | 36.709871 | 38.025586 |
|  |  |  | 1.4 | 36.836328 | 38.175275 | 36.685184 | 37.986828 |
|  |  |  | 1.5 | 36.813212 | 38.138122 | 36.660019 | 37.947681 |
|  | 3.5 | 4.9 | 1.3 | 41.872622 | 43.393544 | 41.721350 | 43.194018 |
|  |  |  | 1.4 | 41.843912 | 43.350219 | 41.690409 | 43.148378 |
|  |  |  | 1.5 | 41.814750 | 43.306567 | 41.658997 | 43.102416 |
|  |  | 5 | 1.3 | 41.920891 | 43.424115 | 41.770719 | 43.226282 |
|  |  |  | 1.4 | 41.891787 | 43.380481 | 41.739370 | 43.180320 |
|  |  |  | 1.5 | 41.862229 | 43.336516 | 41.707547 | 43.134033 |
| 0.5 | 3 | 4.9 | 1.3 | 34.615022 | 35.969356 | 34.555489 | 35.960373 |
|  |  |  | 1.4 | 34.544135 | 35.904461 | 34.483394 | 35.894035 |
|  |  |  | 1.5 | 34.468688 | 35.836613 | 34.407267 | 35.824682 |
|  |  | 5 | 1.3 | 34.692266 | 36.159776 | 34.630728 | 36.150801 |
|  |  |  | 1.4 | 34.620135 | 36.096342 | 34.557414 | 36.085918 |
|  |  |  | 1.5 | 34.543466 | 36.029919 | 34.480083 | 36.017983 |
|  | 3.5 | 4.9 | 1.3 | 39.795683 | 40.562262 | 39.729348 | 40.485116 |
|  |  |  | 1.4 | 39.735114 | 40.465670 | 39.664874 | 40.387640 |
|  |  |  | 1.5 | 39.668983 | 40.365393 | 39.595265 | 40.286888 |
|  |  | 5 | 1.3 | 39.889874 | 40.652333 | 39.821350 | 40.572546 |
|  |  |  | 1.4 | 39.827897 | 40.554503 | 39.755523 | 40.473845 |
|  |  |  | 1.5 | 39.760402 | 40.453005 | 39.684597 | 40.371876 |
| 1 | 3 | 4.9 | 1.3 | 33.636437 | 35.284177 | 33.620556 | 35.225687 |
|  |  |  | 1.4 | 33.531434 | 35.185557 | 33.513642 | 35.128722 |
|  |  |  | 1.5 | 33.423028 | 35.084276 | 33.403295 | 35.029589 |
|  |  | 5 | 1.3 | 33.846294 | 35.356637 | 33.830410 | 35.295639 |
|  |  |  | 1.4 | 33.742675 | 35.256859 | 33.724873 | 35.197505 |
|  |  |  | 1.5 | 33.635603 | 35.154413 | 33.615848 | 35.097191 |
|  | 3.5 | 4.9 | 1.3 | 37.485612 | 40.314068 | 37.469861 | 40.304697 |
|  |  |  | 1.4 | 37.370067 | 40.238572 | 37.352582 | 40.227868 |
|  |  |  | 1.5 | 37.251619 | 40.160493 | 37.232382 | 40.148410 |
|  |  | 5 | 1.3 | 37.743220 | 40.551003 | 37.727434 | 40.541620 |
|  |  |  | 1.4 | 37.628942 | 40.476932 | 37.611412 | 40.466209 |
|  |  |  | 1.5 | 37.511708 | 40.400240 | 37.492413 | 40.368129 |

Table 11. Effect of parameters $s, S$ and $L$ on $T c, \Theta_{6}, \aleph_{6}, \Theta_{7}, \aleph_{7}$ for M-I and M-II

| $S$ | $s$ | $L$ | M-I |  |  | M-II |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  | Tc | $\Theta_{6}$ | $\Theta_{7}$ | Tc | $\aleph_{6}$ | $\aleph_{7}$ |
| 25 | 3 | 9 | 3.238036 | 0.142861 | 0.015800 | 3.208242 | 0.146133 | 0.015974 |
|  |  | 11 | 3.490347 | 0.157464 | 0.016857 | 3.418891 | 0.163172 | 0.016888 |
|  |  | 13 | 3.708020 | 0.167833 | 0.017450 | 3.582784 | 0.176616 | 0.017614 |
|  | 5 | 9 | 4.071632 | 0.151859 | 0.012341 | 4.038042 | 0.157782 | 0.012591 |
|  |  | 11 | 4.366458 | 0.167377 | 0.013698 | 4.254299 | 0.178028 | 0.013765 |
|  |  | 13 | 4.623548 | 0.177405 | 0.014476 | 4.406793 | 0.194053 | 0.014683 |
|  | 7 | 9 | 4.787175 | 0.160186 | 0.010461 | 4.768930 | 0.168921 | 0.010820 |
|  |  | 11 | 5.164389 | 0.176973 | 0.012051 | 5.016830 | 0.193346 | 0.012174 |
|  |  | 13 | 5.492868 | 0.186689 | 0.012974 | 5.167740 | 0.212878 | 0.013234 |
|  | 3 | 9 | 3.321759 | 0.151609 | 0.015270 | 3.294398 | 0.154038 | 0.015420 |
| 30 |  | 11 | 3.691844 | 0.172875 | 0.016775 | 3.628219 | 0.177302 | 0.016817 |
|  |  | 13 | 4.069249 | 0.191854 | 0.018026 | 3.953855 | 0.199015 | 0.018154 |
|  | 5 | 9 | 4.113452 | 0.161801 | 0.011844 | 4.080144 | 0.166092 | 0.012060 |
|  |  | 11 | 4.544993 | 0.185377 | 0.013587 | 4.445382 | 0.193458 | 0.013679 |
|  |  | 13 | 4.996804 | 0.205877 | 0.015026 | 4.798062 | 0.219202 | 0.015156 |
|  |  | 9 | 4.783348 | 0.171360 | 0.009929 | 4.755632 | 0.177518 | 0.010230 |
|  | 7 | 11 | 5.320347 | 0.197952 | 0.011865 | 5.187014 | 0.210097 | 0.012026 |
|  |  | 13 | 5.889404 | 0.220523 | 0.013498 | 5.590835 | 0.241137 | 0.013619 |
|  | 3 | 9 | 3.402898 | 0.158250 | 0.014815 | 3.377587 | 0.160140 | 0.014946 |
|  |  | 11 | 3.874466 | 0.185037 | 0.016650 | 3.817350 | 0.188600 | 0.016699 |
|  |  | 13 | 4.400359 | 0.211664 | 0.018436 | 4.294597 | 0.217653 | 0.018529 |
|  |  | 9 | 4.166496 | 0.169140 | 0.011421 | 4.134227 | 0.172420 | 0.011609 |
| 35 | 5 | 11 | 4.713830 | 0.199246 | 0.013428 | 4.624679 | 0.205639 | 0.013529 |
|  |  | 13 | 5.344847 | 0.228925 | 0.015362 | 5.163771 | 0.239888 | 0.015438 |
|  |  | 9 | 4.806328 | 0.179338 | 0.009493 | 4.774310 | 0.183953 | 0.009747 |
|  | 7 | 11 | 5.475793 | 0.213669 | 0.011647 | 5.355552 | 0.223109 | 0.011815 |
|  |  | 13 | 6.262317 | 0.247337 | 0.013765 | 5.991815 | 0.264043 | 0.013808 |


[^0]:    ${ }^{1}$ Tables (1-11) are included at the end of the article.

