

MULTI-ENTITY STOCK DEPENDENT MODEL WITH CAPACITY AND MANUFACTURE COST RESTRAINT'S

Atma Nand^{1a}, N. S. Chauhan^{2b}

Abstract: Inventory has an impact on the manufacturing process as well as supply chain operations. The fundamental goal of this research paper is to optimize the cost associated with inventories and to provide flow less continuous production process in time. Normally, demand rate of any entity in inventory control model are treated as predictable and at the same time constant too, and that the cost associated to unit inventory must be independent and non-variable in nature. Nevertheless, in practical circumstances, the unit price and demand rate of an entity may be interconnected. When the asking for an article is enormous, an entity is manufactured in huge quantities and the static charges of manufacturing being diffused over a multiple component. Henceforth, per unit article cost decreases significantly. i.e., per unit article cost and the demand of an article are related under inverse variation. So, better to consider the demand rate of an article as a variable constraint than to fixed one. In this research article, a mathematical model for multiple articles through permitted and restricted shortage and per article cost based on demand accompanied by upper and lower limits viz restricted storage space and manufacturing expenses has been constructed. Overall, investigating the simultaneous effect of storage space and manufacturing expenses in an inventory model provides valuable insights that enable cost optimization, resource allocation, capacity planning, and risk mitigation. It helps companies make informed decisions and improve their overall operational efficiency and profitability. The Multi-Entity Stock Dependent Model with Capacity and Manufacture Cost Restraints can be customized and used in a variety of sectors that include managing inventory across numerous entities and complicated supply chain networks. Here are a few examples of industries that can benefit from such a model: manufacturing industry, the retail and distributor sector, e-commerce companies, pharmaceutical and healthcare industry, automotive industry and food and beverage industry. The article cost is explored at this juncture in a fuzzy atmosphere and solutions of the model being obtained through KKT condition. Finally, a conclusion is offered in the final portion.

Keywords: *Integrated inventory model, restricted shortages, KKT, fuzzification, variable constraint.*

1. Introduction

The economic order quantity model has an essential and key functioning in the field of inventory. When applying the EOQ model to particular practical scenarios encountered in real life, it is difficult to precisely estimate the cost associated with the various terminology of inventory viz, setup, carrying, shortage, demand etc. Only approximate values may be found. Mostly, the situation which are unpredictable and not certain are studied under the influence of stochastic inventory theory.

Inventory control is a crucial area for both real-world applications and research reasons. The most often used inventory model is the Economic Order Quantity model, in which the sequential operations are classed as supply and demand. The first quantitative treatment of inventory was the basic EOQ model. This model was created by Harris (1915) later, Hadley & Whitin (1963) analysed several inventory methods. Abu Hashan Md Mashudand et al. (2021), Mishra U et al. (21), ¹Rahman et al. (2022), ²Rahman et al. (2022) studies the two warehouse inventory problems to get rid of stockout situation but renting or

owning two warehouse increases the investment and overall profitability of the entire model decreases. Miah et al. (2021) developed limited time price discount inventory model but restricted this model for the electronic products only. This model could be more generalize to cover more industries.⁴Roy D et al. (2022), ²Roy D et al. (2022) studies the inventory models with preservation technology and cap-and-trade policy. Sultana et al. (2022) described the role of the discount policy and its impact on the inventory control model. They extended the earlier work of Shaikh A. A. et al. (2017, 2020) in which author considered fully backlogged inventory model and application of preservation facility with ramp type demand. Md. Alamin Khan et al. (2017) proposed the solution of nonlinear system of equation that could be raised during the development of mathematical formulation of inventory model.

The fundamental goal of this mathematical model is to optimize the cost associated with inventories and to provide flow less continuous production process in time. Normally, demand rate of any entity in inventory control model are treated as predictable and at the same time constant too, and that the cost associated to unit inventory must be independent and non-variable in nature. Nevertheless, in practical circumstances, the unit price and demand rate of an entity may be interconnected. When the asking for an article is enormous, an entity is manufactured in

Authors information:

^aDepartment of Mathematics, SALS, Uttaranchal University, Dehradun-248007, INDIA. E-mail: atmanand.prasad@gmail.com¹.

^bDepartment of Mathematics, Faculty of Engineering (FOE), Teerthanker Mahaveer University, Moradabad-244001, INDIA.

*Corresponding Author: atmanand.prasad@gmail.com

Received: January 17, 2023

Accepted: July 5, 2023

Published: September 30, 2024

huge quantities and the static charges of manufacturing being diffused over a multiple component. Henceforth, per unit article cost decreases significantly. i.e., per unit article cot and the demand of an article are related under inverse variation. So, better to consider the demand rate of an article as a variable constraint than to fixed one.

In this chapter, a mathematical model for multiple articles through permitted and restricted shortage and per article cost based on demand accompanied by upper and lower limits viz restricted storage space and manufacturing expenses has been constructed. The article cost is explored at this juncture in a fuzzy atmosphere and solutions of the model being obtained through KKT condition. Finally, a conclusion is offered in the final portion.

2. K-K-T Conditions

Taha (2007) presented how to achieve the optimum solution of a nonlinear programming issue associated to inequality restrictions by applying the Kuhn-Tucker criteria. The construction of the Kuhn-Tucker conditions is based on the Lagrangean approach. Assume that the issue is stated by

$$\text{Minimize } y = f(x) \quad \dots \dots 1$$

$$\text{Associated to } h_{\sigma}(x) \geq 0, \sigma = \{x: x \in N\}. \quad \dots \dots 2$$

The non-negative restrictions $x \geq 0$, if any, are included into the limitations of a natural numbers. The inequality restrictions can be imposed to the equations by implementing non-negative slack variables. Suppose s_{σ}^2 the amount of slack eliminated from the σ^{th} constraint $h_{\sigma}(x) \geq 0$.

$$\text{Let } \varepsilon = (\varepsilon_1, \varepsilon_2, \dots, \varepsilon_n), \quad \dots \dots 4$$

$$h(x) = (g_1(x), g_2(x), \dots, g_n(x)) \quad \dots \dots 5$$

$$\text{and } s^2 = (s_1^2, s_2^2, \dots, s_n^2) \quad \dots \dots 6$$

Then the Lagrangean functions are given by

$$H(x, s, \varepsilon) = f(x) - \varepsilon[h(x) - s^2] \quad \dots \dots 7$$

Considering the fractional differentiation of H associated to x, s, and ε criteria that are also sufficient if the objective function and solution domain meet the following limitations:

Sense of optimization	Prerequisite conditions	
	Objective function	Solution space
Maximization	Concave	Convex Set
Minimization	Convex	Convex Set

The following is a summary of the requirements for determining if the Kuhn-Tucker conditions are satisfied:

Problem	Kuhn-Tucker conditions
1. Max z = f(x) associated to $k^{\sigma}(x) \leq 0, x \geq 0, \sigma = 1, 2, \dots, n$	$\frac{\partial}{\partial x_j} f(x) - \sum_{\sigma=1}^n \varepsilon_{\sigma} \frac{\partial}{\partial x_j} k^{\sigma}(x) = 0 \quad \dots \dots 8$ $\varepsilon_{\sigma} k^{\sigma}(x) = 0, k^{\sigma}(x) \leq 0, \sigma = 1, 2, \dots, n \quad \dots \dots 9$ $\varepsilon_{\sigma} \geq 0, \sigma = 1, 2, \dots, n \quad \dots \dots 10$
2. Max z = f(x) associated to $k^{\sigma}(x) \geq 0, x \geq 0, \sigma = 1, 2, \dots, n$	$\frac{\partial}{\partial x_j} f(x) - \sum_{\sigma=1}^n \varepsilon_{\sigma} \frac{\partial}{\partial x_j} k^{\sigma}(x) = 0 \quad \dots \dots 11$ $\varepsilon_{\sigma} k^{\sigma}(x) = 0, k^{\sigma}(x) \geq 0, \sigma = 1, 2, \dots, n \quad \dots \dots 12$ $\varepsilon_{\sigma} \geq 0, \sigma = 1, 2, \dots, n \quad \dots \dots 13$

Karush was one who firstly introduced and developed the K-K-T conditions in 1939.

3. Formulation and Evaluation of The Model

Let the stock volume of σ^{th} entity ($\sigma \in N$) be R_{σ} for instance $t = 0$. In specified range $(0, T_{\sigma} (= t_{1\sigma} + t_{2\sigma}))$, demand being fulfilled by systematic declination of stock level. This technique results in the inventory level being 0 at the time $t_{1\sigma}$ then the range is allowed to experience shortages in the domain $(t_{1\sigma}, T_{\sigma})$.

Figure 3.1 interpret a mathematical model with stock backorder. R_{σ} is the maximum inventory quantity and Q is the order quantity for one period? Also, $t_{1\sigma}$ indicates the time needed for the R_{σ} entities to be required. The length of time during one period over which backorder will be incurred will be given by Figure-1.

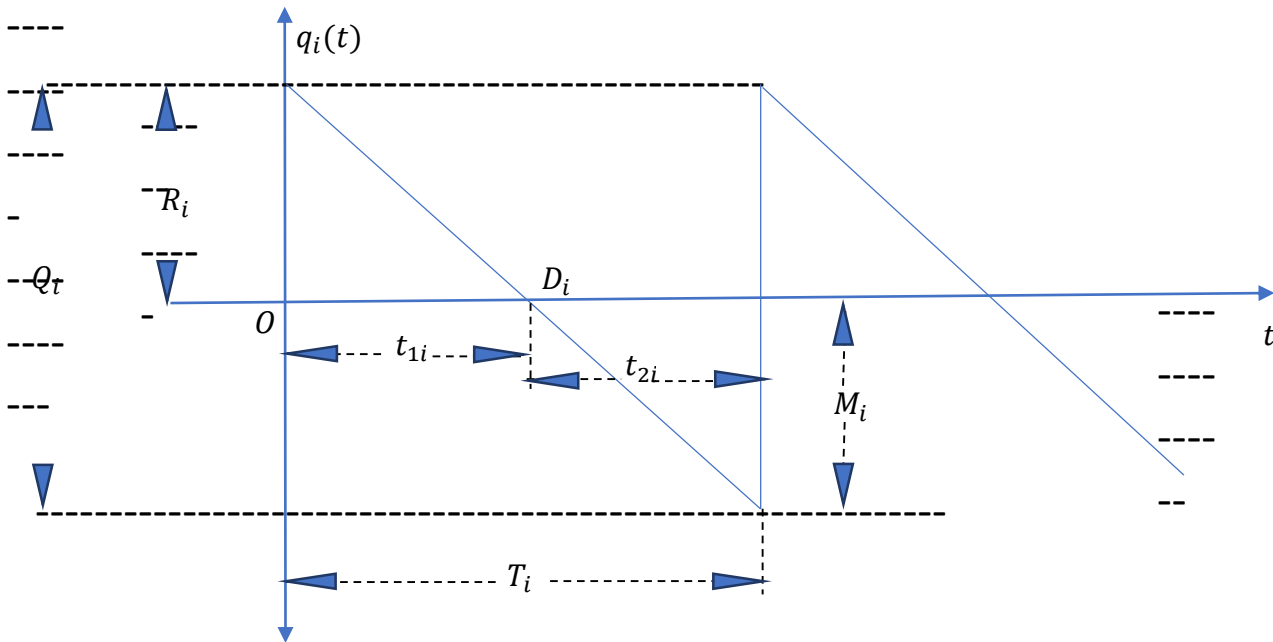


Figure -1 Stock level of the σ^{th} entity

Let $q_\sigma(t)$ be the on-hand inventory at time t ($0 \leq t \leq T$). In this model, uniform replenishment rate starts with inventory level q_σ . The inventory level decreases with demand. Ultimately the inventory reaches 0 at the end of the cycle time $t_{1\sigma}$.

The derivative function describing inventory level $q_\sigma(t)$ of σ^{th} entity in the range $0 \leq t \leq T_\sigma$ is given by

$$\frac{dq_\sigma(t)}{dt} = \begin{cases} -D_\sigma, & \text{for } 0 \leq t \leq t_{1\sigma} \\ -D_\sigma, & \text{for } t_{1\sigma} \leq t \leq T_\sigma \end{cases} \dots\dots 14$$

With the conditions $q_\sigma(0) = R_\sigma (= Q_\sigma - M_\sigma)$, $q_\sigma(T_\sigma) = -M_\sigma$ and $q_\sigma(t_{1\sigma}) = 0$ 15

On every interval a certain quantity of shortfall is permitted and there is a penalty cost m_1 per entity of unmet demand per unit time.

For $0 \leq t \leq t_{1\sigma}$ 16

$$\int_0^t dq_\sigma(t) = - \int_0^t D_\sigma dt \dots\dots 17$$

$$q_\sigma(t) - q_\sigma(0) = -D_\sigma t \dots\dots 18$$

Hence, $q_\sigma(t) = R_\sigma - D_\sigma(t)$ 19

For $t_{1\sigma} \leq t \leq T_\sigma$ 20

$$\int_{t_{1\sigma}}^t dq_\sigma(t) = - \int_{t_{1\sigma}}^t D_\sigma dt \dots\dots 21$$

$$q_\sigma(t) - q_\sigma(t_{1\sigma}) = -D_\sigma(t - t_{1\sigma}) \dots\dots 22$$

Hence, $q_\sigma(t) = D_\sigma(t_{1\sigma} - t)$ 23

Thus $q_\sigma(t) = R_\sigma - D_\sigma(t)$, for $0 \leq t \leq t_{1\sigma}$ 24

$D_\sigma(t_{1\sigma} - t)$ for $t_{1\sigma} \leq t \leq T_\sigma$ 25

Also $D_\sigma t_{1\sigma} = R_\sigma$ 26

$M_\sigma = D_\sigma t_{2\sigma}$ 27

$Q_\sigma = D_\sigma T_\sigma$ 28

The holding cost is related to the cost of carrying (or holding) inventory. This cost frequently encompasses the expenditures such as rent for space, usage for storage, interest on the money locked-up, insurance of stored equipment, manufacturing, taxes, depreciation of equipment and furnishings utilised, etc.

This is derived by evaluating the integral in the range $(0, t_{1\sigma})$ 29

Holding cost = $H_\sigma \int_0^{t_{1\sigma}} q_\sigma(t) dt$ 30

$$= \frac{H_\sigma(Q_\sigma - M_\sigma)^2}{2D_\sigma} \dots\dots 31$$

$$= \frac{H_\sigma T_\sigma(Q_\sigma - M_\sigma)^2}{2Q_\sigma} \dots\dots 32$$

Since, $Q_\sigma = D_\sigma T_\sigma$ 33

The penalty cost for running out of stock (i.e., when an entity cannot be provided on the customer's demand) is known as the shortfall cost. This cost encompasses the loss of projected profit via sales of items and loss of goodwill in terms of permanent loss of customers, and it is tied to lost earnings in future sales. Thus, the shortfall cost is calculated by evaluating the integral in the range $(t_{1\sigma}, T_\sigma)$. This is because the shortage develops only after all the present stockpiles are consumed. Hence,

Shortage cost = $n_\sigma \int_{t_{1\sigma}}^{T_\sigma} -q_\sigma(t) dt$ 34

$$= -n_\sigma \left[\{T_\sigma q_\sigma(T_\sigma) - t_{1\sigma} q_\sigma(t_{1\sigma})\} - \int_{t_{1\sigma}}^{T_\sigma} t(-D_\sigma) dt \right] \dots\dots 35$$

Since, $dq_\sigma(t) = -D_\sigma dt$ for $t_{1\sigma} \leq t \leq T_\sigma$ 36

The shortage cost becomes,

Shortage cost = $n_\sigma \{T_\sigma(-M_\sigma) - 0\} + D_\sigma \int_{t_{1\sigma}}^{T_\sigma} t dt$ 37

$$= n_\sigma T_\sigma M_\sigma - \frac{n_\sigma D_\sigma}{2} [T_\sigma^2 - t_{1\sigma}^2] \dots\dots 38$$

But $T_\sigma = \frac{Q_\sigma}{D_\sigma}$

$t_{1\sigma} = \frac{R_\sigma}{D_\sigma} = \frac{(Q_\sigma - M_\sigma)}{D_\sigma}$ and $Q_\sigma = R_\sigma + M_\sigma$ 39

$$= n_\sigma M_\sigma T_\sigma = \frac{n_\sigma D_\sigma}{2} \left[\frac{Q_\sigma^2}{D_\sigma^2} - \frac{R_\sigma^2}{D_\sigma^2} \right] \dots\dots 40$$

$$= n_{\sigma} M_{\sigma} T_{\sigma} - \frac{n_{\sigma}}{2D_{\sigma}} [(R_{\sigma} + M_{\sigma})^2 - R_{\sigma}^2] \quad \dots \dots \dots 41$$

$$= \frac{n_{\sigma} M_{\sigma}^2}{2D_{\sigma}} \quad \dots \dots \dots 42$$

$$= \frac{n_{\sigma} M_{\sigma}^2 T_{\sigma}}{2Q_{\sigma}} \quad \dots \dots \dots 43$$

Since, $D_{\sigma} = \frac{Q_{\sigma}}{T_{\sigma}}$

The cost of manufacturing each component of the system is provided by

Production cost = $P_{\sigma} Q_{\sigma}$

Additionally, each entity's setup price is provided by

Setup cost = S_{σ}

The total cost is defined as the sum of production cost, setup cost, holding cost and shortage cost.

$$TC = p_{\sigma} Q_{\sigma} + S_{\sigma} + \frac{H_{\sigma} T_{\sigma} (Q_{\sigma} - M_{\sigma})^2}{2Q_{\sigma}} + \frac{n_{\sigma} M_{\sigma}^2 T_{\sigma}}{2Q_{\sigma}}, \text{ for } \sigma = \dots \dots \dots 44$$

1, 2, ..., n.

The overall average cost of the σ^{th} entity of the system is then given by

$$TC(p_{\sigma}, Q_{\sigma}, M_{\sigma}) = p_{\sigma} D_{\sigma} + \frac{S_{\sigma} D_{\sigma}}{Q_{\sigma}} + \frac{H_{\sigma} (Q_{\sigma} - M_{\sigma})^2}{2Q_{\sigma}} + \frac{n_{\sigma} M_{\sigma}^2}{2Q_{\sigma}} \quad \dots \dots \dots 45$$

The unit price of an entity is often thought of as being constant and independent in nature, and the classical inventory concerns are formed by taking these assumptions into account. By making the assumption that the demand rate and unit price are constants and independent of one another, Silver & Peterson (1985) developed an inventory model. However, in real-world situations, a company's unit price and demand rate may be connected. When there is a high demand for something, it is produced in vast quantities and the fixed production costs are spread among a lot of different goods. As a result, the entity's unit cost declines, making its unit price inversely linked to its demand. This strategy was used by Jung & Klein (2001) to propose and resolve the Economic Order Quantity problem. The inventory model with demand-dependent unit pricing is solved in the current study by utilising the Karush Kuhn-Tucker method.

4. Assumptions of the Inventory Model

Under the following presumptions, a multi-entity stock non-restricted shortage model has been developed.

Instant replenishment is available.

No lead time exists.

Demand is correlated with unit price as $p_{\sigma} = A_{\sigma}^{\beta} D_{\sigma}^{-\beta}$

where $A_{\sigma} (> 0)$ and $\beta (\beta > 1)$ being non-variable, non-imaginary numbers chosen to provide the best fit of the estimated price function $A_{\sigma} > 0$ is a necessary constraint as individually D_{σ} and P_{σ} need to be positive.

5. Objective Function of the Model

The objectives of the issue are as follows:

According to the objective of modelling, the total cost of materials must minimise in the system which comprises three parts with shortage cost.

The annual total cost rendering to the fundamental valuation of the mathematical model for economic order quantity is:

Overall cost = manufacture cost + setup cost + holding cost + shortage cost

$$TC(p_{\sigma}, Q_{\sigma}, M_{\sigma}) = p_{\sigma} D_{\sigma} + \frac{S_{\sigma} D_{\sigma}}{Q_{\sigma}} + \frac{H_{\sigma} (Q_{\sigma} - M_{\sigma})^2}{2Q_{\sigma}} + \frac{n_{\sigma} M_{\sigma}^2}{2Q_{\sigma}} \quad \dots \dots \dots 46$$

Substituting for p_{σ} gives

$$TC(D_{\sigma}, Q_{\sigma}, M_{\sigma}) = A_{\sigma}^{\beta} D_{\sigma}^{1-\beta} + \frac{S_{\sigma} D_{\sigma}}{Q_{\sigma}} + \frac{H_{\sigma} (Q_{\sigma} - M_{\sigma})^2}{2Q_{\sigma}} + \frac{n_{\sigma} M_{\sigma}^2}{2Q_{\sigma}} \quad \dots \dots \dots 47$$

for $\sigma = 1, 2, 3, \dots, n$

The inventory model's primary goal is to reduce the overall cost specified by

$$\text{Min } TC(D_{\sigma}, Q_{\sigma}, M_{\sigma}) = \sum_{\sigma=1}^n \left[A_{\sigma}^{\beta} D_{\sigma}^{1-\beta} + \frac{S_{\sigma} D_{\sigma}}{Q_{\sigma}} + \frac{H_{\sigma} (Q_{\sigma} - M_{\sigma})^2}{2Q_{\sigma}} + \frac{n_{\sigma} M_{\sigma}^2}{2Q_{\sigma}} \right] \quad \dots \dots \dots 48$$

6. Constraints of the Model

The following limitations have been put on the proposed model: In order to achieve the ideal overall cost, it is necessary to take into account a variety of resource limitations. The amount of warehouse floor space that can be used to store the items is limited.

i. e; $\sum_{\sigma=1}^n w_{\sigma} Q_{\sigma} \leq W \quad \dots \dots \dots 49$

The amount that can be invested in overall production costs is finite and may have a maximum investment level.

i. e; $\sum_{\sigma=1}^n p_{\sigma} Q_{\sigma} \leq B \quad \dots \dots \dots 50$

$$\Rightarrow \text{i. e; } \sum_{\sigma=1}^n A_{\sigma}^{\beta} D_{\sigma}^{-\beta} Q_{\sigma} \leq B \quad \dots \dots \dots 51$$

7. Uncertain Stock Model

When p_{σ} 's are uncertain decision constraints, the above crisp formulation underneath uncertain situation reduces to

$$\text{Min } TC(p_{\sigma}, D_{\sigma}, Q_{\sigma}, M_{\sigma}) \quad \dots \dots \dots 52$$

$$= \sum_{\sigma=1}^n \left[A_{\sigma}^{\beta} D_{\sigma}^{1-\beta} + \frac{S_{\sigma} D_{\sigma}}{Q_{\sigma}} + \frac{H_{\sigma} (Q_{\sigma} - M_{\sigma})^2}{2Q_{\sigma}} + \frac{n_{\sigma} M_{\sigma}^2}{2Q_{\sigma}} \right]$$

associated to the limitations

$$\sum_{\sigma=1}^n w_{\sigma} Q_{\sigma} \leq W \quad \dots \dots \dots 53$$

$$\sum_{\sigma=1}^n A_{\sigma}^{\beta} D_{\sigma}^{-\beta} Q_{\sigma} \leq B \quad \dots \dots \dots 54$$

Where $\tilde{p}_{\sigma} = A_{\sigma}^{\beta} D_{\sigma}^{-\beta}$ and \tilde{p}_{σ} represents fuzzification of the parameters.

8. Karush Kuhn-Tucker Conditions for Solving the Stock Model

The impartial function of a stock model is

$$\text{Min TC } (D_\sigma, Q_\sigma, M_\sigma) = \sum_{\sigma=1}^n \left[A_\sigma^\beta D_\sigma^{1-\beta} + \frac{S_\sigma D_\sigma}{Q_\sigma} + \frac{H_\sigma(Q_\sigma - M_\sigma)^2}{2Q_\sigma} + \frac{n_\sigma M_\sigma^2}{2Q_\sigma} \right] \dots\dots\dots 55$$

associated to the constraints

$$\sum_{\sigma=1}^n w_\sigma Q_\sigma \leq W \dots\dots\dots 56$$

$$\sum_{\sigma=1}^n A_\sigma^\beta D_\sigma^{-\beta} Q_\sigma \leq B \dots\dots\dots 57$$

Here the decision constraints are the demand D_σ , lot size Q and the shortage level M_1 . The problem is, to solve the above inventory model with these decision variables associated to the inequality constraints (3.3) and (3.4) in order to minimize the overall cost function. The problem is solved for a unit entity. For a single entity the objective function and the constraints can be written as follows.

$$\text{Min TC } (D, Q, M) = \sum_{\sigma=1}^n \left[A^\beta D^{1-\beta} + \frac{SD}{Q} + \frac{H(Q - M)^2}{2Q} + \frac{nM^2}{2Q} \right] \dots\dots\dots 58$$

associated to the inequality constraints

$$wQ \leq W \dots\dots\dots 59$$

$$A^\beta D^{-\beta} Q \leq B \dots\dots\dots 60$$

To minimize the objective function, the Lagrangean function has been constructed by introducing the variables s_1 and s_2 as follows:

9. Relationship Function

The relationship function for the vague variable P_σ is defined as follows

$$\mu_{p_\sigma}(X) = \begin{cases} 1, & p_\sigma \leq L_{L_\sigma} \\ \frac{U_{L_\sigma} - p_\sigma}{U_{L_\sigma} - L_{L_\sigma}}, & L_{L_\sigma} \leq p_\sigma \leq U_{L_\sigma} \\ 0, & p_\sigma \geq U_{L_\sigma} \end{cases}$$

Here U_L and L_L are superior bound and inferior bound of P_σ correspondingly.

10. Numerical Example

To illustrate the suggested mathematical model for stock with and without shortage instances, the following input data are examined in correct units for a single entity. A numerical example has been constructed for a single entity with the set of input parametric values given in Table-1.

Table-1 The input values of parameters in the mathematical model

Parameter	Notation	Value (in rupees)
Number of entities	n	1
Constant	A_1	20
Setup cost of the entity 1	S_1	80
Holding cost of the entity 1	H_1	0.7
Storage space for the entity 1	w_1	3sq.ft.
Storage space available	W	280sq.ft.
Total investment cost	B	40
Shortage cost per unit entity	n_1	10
Lower limit of the component cost of the entity 1	L_{L_1}	1
Upper limit of the component cost of the entity 1	U_{L_1}	2

For the above data, the objective function becomes

$$G = 20^\beta D^{1-\beta} + 80DQ^{-1} + 0.35(Q - M)^2 Q^{-1} + 5M^2 Q^{-1} - \varepsilon_1(280 - 3Q - s_1^2) - \varepsilon_2(40 - 20^\beta D^{-\beta} Q - s_2^2) \dots\dots\dots 62$$

$$G = 20^\beta D^{1-\beta} + 80DQ^{-1} + 0.35Q - 0.7M + 5.35M^2 Q^{-1} - \varepsilon_1(280 - 3Q - s_1^2) - \varepsilon_2(40 - 20^\beta D^{-\beta} Q - s_2^2) \dots\dots\dots 63$$

Differentiating (63) partially with respect to D, Q and M respectively we get

By the Kuhn-Tucker conditions

$$\frac{\partial G}{\partial D} = 0 \Rightarrow (1 - \beta)20^\beta D^{-\beta} + 80Q^{-1} - \beta \epsilon_2 20^\beta D^{-\beta-1} Q = 0 \dots\dots\dots 67$$

$$\frac{\partial G}{\partial Q} = 0 \Rightarrow 80DQ^{-2} + 0.35 - 5.35 Q^{-2}M^2 + 3\epsilon_1 + \epsilon_2 20^\beta D^{-\beta} = 0 \dots\dots\dots 68$$

$$\frac{\partial G}{\partial M} = -0.7 + 10.7Q^{-1}M = 0 \dots\dots\dots 69$$

Hence, an optimal solution has been obtained by solving the Equations (6), (7) & (8) by implementing K-K-T conditions with demand *D*, portion size *Q* and the deficiency level *M* as the decision constraints by varying the parametric value β . Also, an optimum solution has been obtained by fuzzifying the unit cost and the results are discussed in Table 2.

The value of the parameter β has been chosen between 2 and 3. The most suitable values are obtained for the parametric values β such as 2.4, 2.5, 2.6 and 2.8 by trial-and-error method that minimizes the objective function.

A sensitivity analysis for optimum solution with shortages corresponding to the parameter β is given in Table 2.

Table-2 Optimal results for the model with shortages

β	P_1	μ_{p_0} value	D_1	Q_1	M_1	Expected Total cost
2.4	1.2853	0.7147	18.014	62.092	4.061	55.151
2.5	1.3053	0.6947	17.978	62.899	4.114	54.872
2.6	1.3278	0.6722	17.934	63.627	4.161	54.633
2.8	1.3806	0.6194	17.824	64.834	4.240	54.197

In Table 2, a study of expected total cost with demand and lot size including shortages is given for different values of β . We can

conclude that when demand decreases, lot size increases but the annual total cost decreases.

$$\frac{\partial G}{\partial D} = (1 - \beta)20^\beta D^{-\beta} + 80Q^{-1} - \beta \epsilon_2 20^\beta D^{-\beta-1} Q \dots\dots\dots 64$$

$$\frac{\partial G}{\partial Q} = 80DQ^{-2} + 0.35 - 5.35 Q^{-2}M^2 + 3\epsilon_1 + \epsilon_2 20^\beta D^{-\beta} \dots\dots\dots 65$$

$$\frac{\partial G}{\partial M} = -0.7 + 10.7Q^{-1}M \dots\dots\dots 66$$

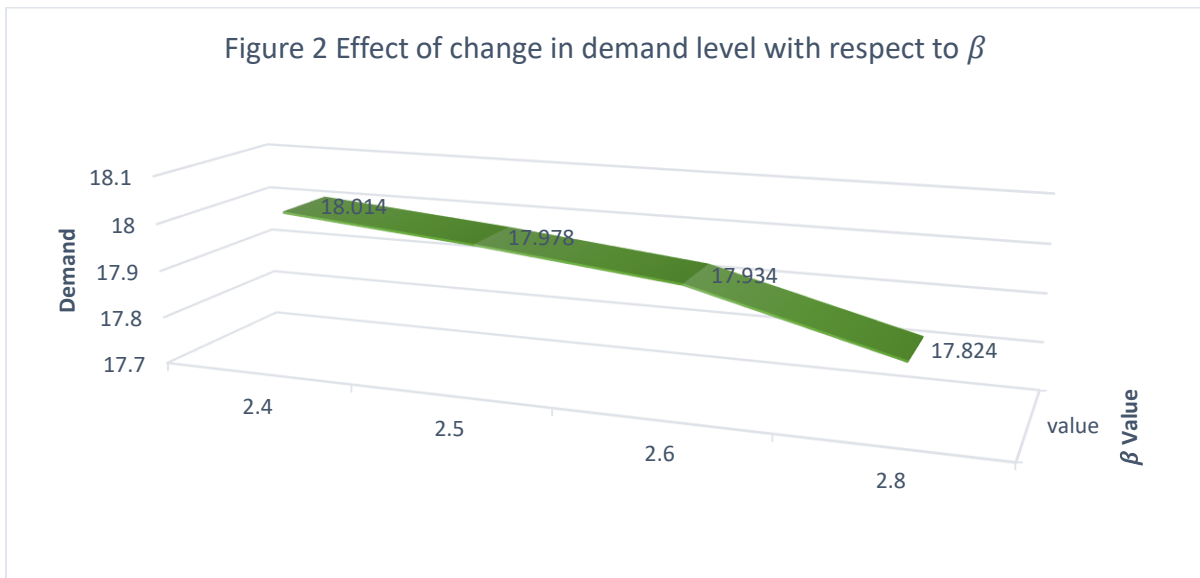
From the above table it follows that 1.2853 has the extreme relationship value 0.7147. Henceforth the enforced optimal resolution is $Q_1 = 62.092$, $D_1 = 18.014$, $M_1 = 4.061$ and Minimum expected total cost 55.151.

11. Sensitivity Analysis

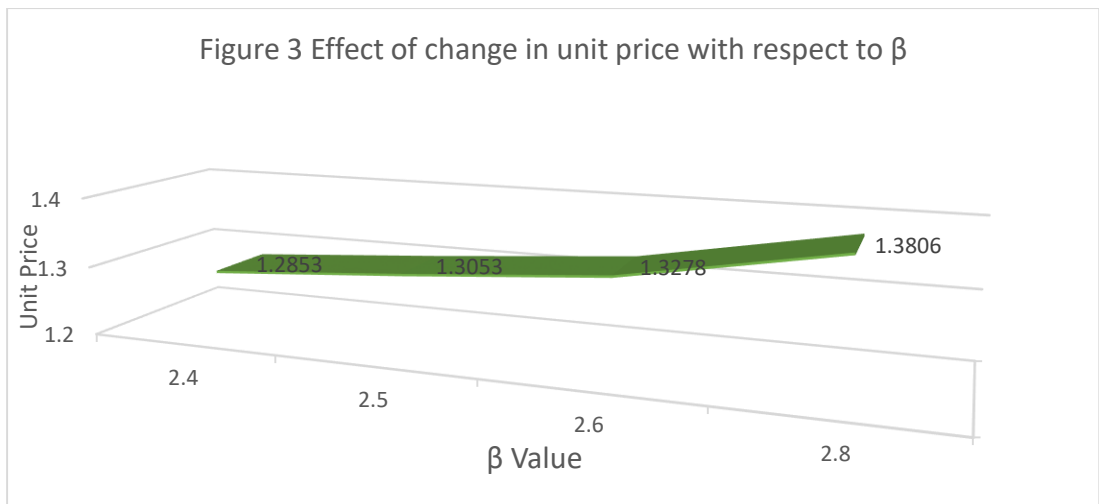
Sensitivity examination is investigated to check in what way the productivity of the mathematical model is influenced by modifications or errors in its input parameters based on the numerical example. The outcomes are demonstrated through the assistance of numerical example. A model with and without shortage is discussed in this chapter.

One of the most fundamental inventory models is the stochastic model. The model is significant because it continues to be one of the most widely used inventory models in the sector and acts as a foundation for more sophisticated inventory models.

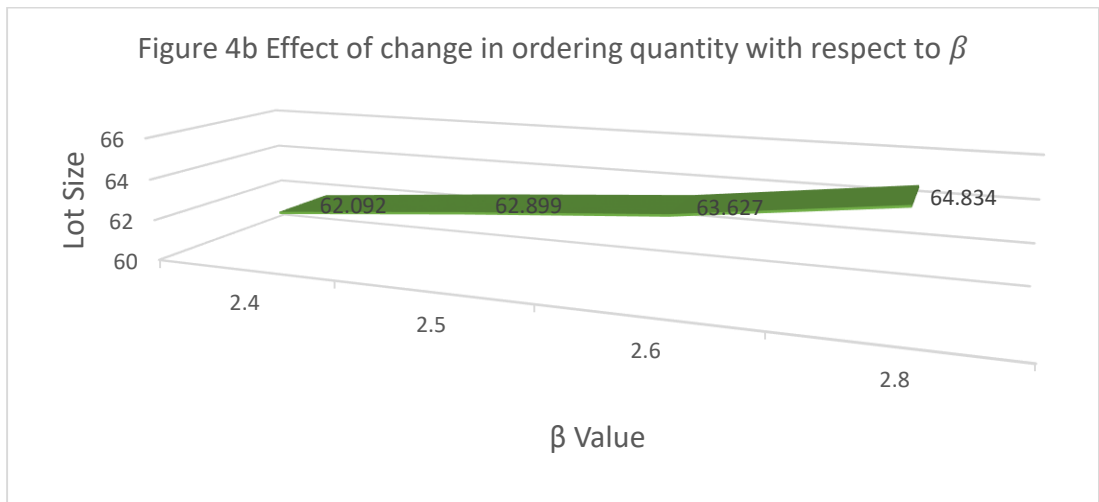
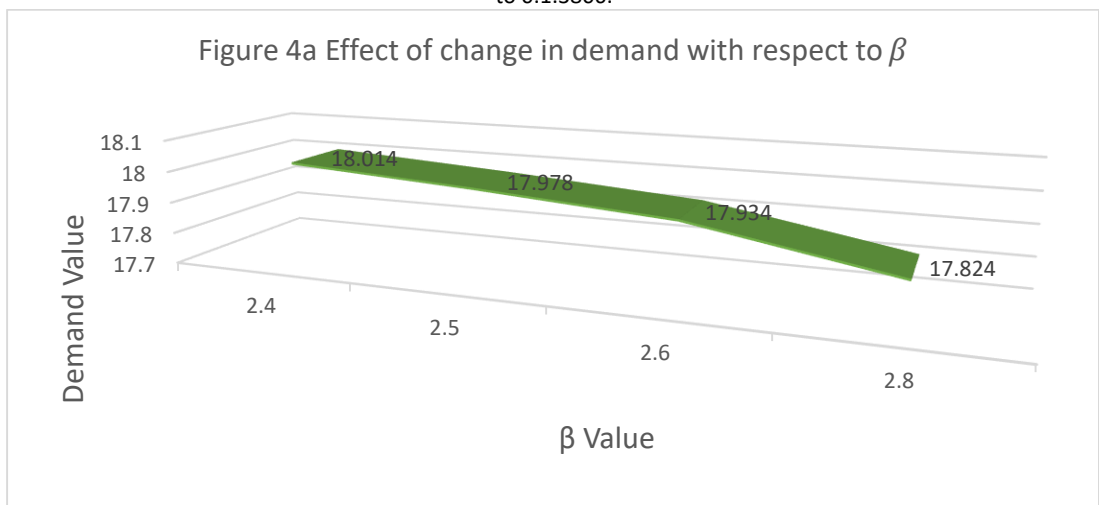
The values of ambiguous variables needed for decision-making are obtained by solving the inventory model using the Kuhn-Tucker condition technique. The maximum membership function value of 0.7147 corresponds to the best values of the choice variables and the overall cost. Hence the optimal solution is $Q_1 = 62.092$, $D_1 = 18.014$, $M_1 = 4.061$ and $TC = 55.151$. Results due to different values of β for the model has been calculated and depicted in the following Figures 2, 3, 4, 5 and 6.



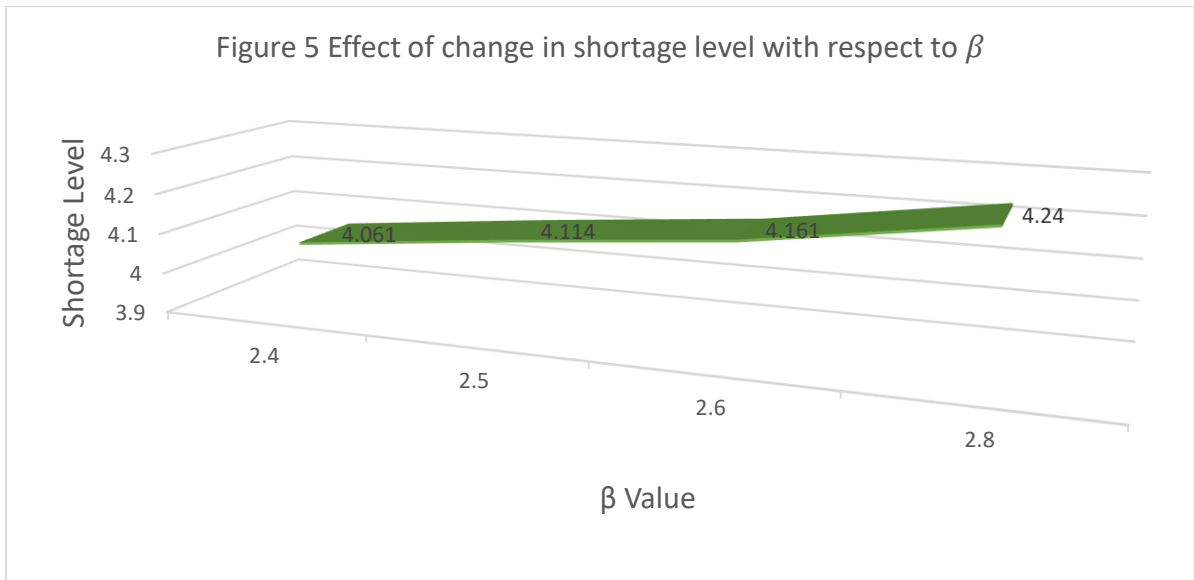
The above Figure 2 shows that as the value of the parameter β increases from 2.4 to 2.8, the value of the demand decreases from 18.014 to 17.824.



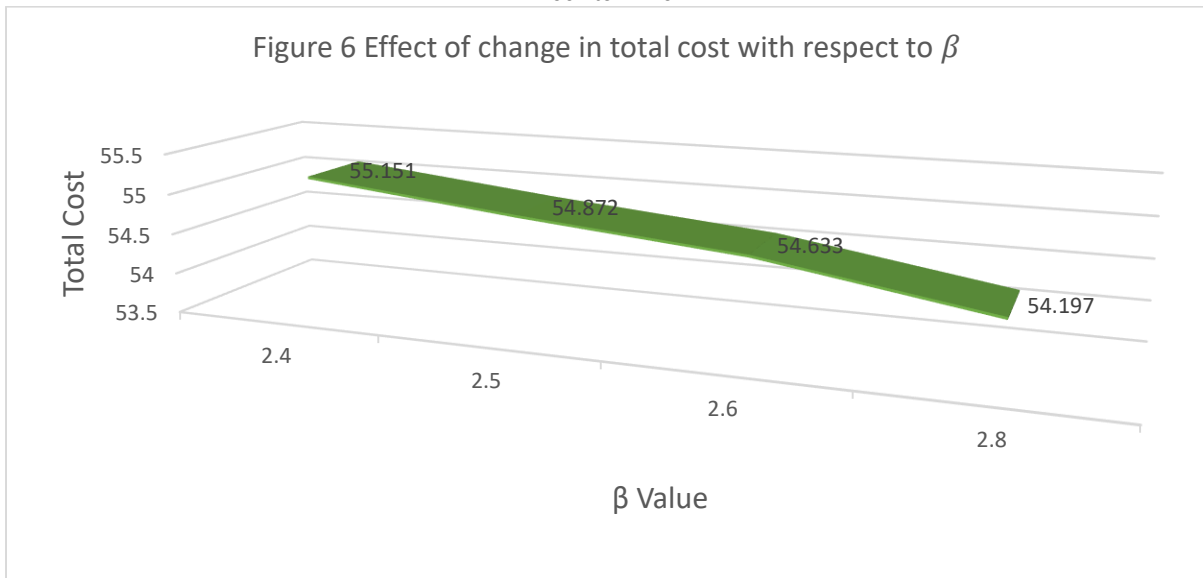
The above Figure 3 shows that as the value of the parameter β increases from 2.4 to 2.8, the value of the unit price increases from 1.2853 to 1.3806.



The above Figure 4 shows that as the value of the parameter β increases from 2.4 to 2.8, the value of the lot size increases from 62.092 to 64.834.



The above Figure 5 shows that as the value of the parameter β increases from 2.4 to 2.8, the value of the shortage level increases from 4.061 to 4.240.



The above Figure 6 shows that as the value of the parameter β increases from 2.4 to 2.8, the value of the annual total cost decreases from 55.151 to 54.197.

12. Inventory Model Without Shortages as a Singular Instance

It is possible to get the scenario without a shortage by inserting in the preceding model. The only variables that affect the total cost function in this scenario are demand and lot size. An inventory model without shortages can be reduced to

$$M\sigma n TC(D_\sigma, Q_\sigma) = \sum_{\sigma=1}^n \left[A_\sigma^\beta D_\sigma^{1-\beta} + \frac{S_\sigma D_\sigma}{Q_\sigma} + \frac{H_\sigma Q_\sigma}{2} \right] \dots \dots \dots 70$$

associated to the constraints

$$\sum_{\sigma=1}^n w_\sigma Q_\sigma \leq W \dots \dots \dots 71$$

$$\sum_{\sigma=1}^n A_\sigma^\beta D_\sigma^{-\beta} Q_\sigma \leq B \dots \dots \dots 72$$

For a single entity the inventory model can be stated as

$$M\sigma_n TC(D, Q) = A^\beta D^{1-\beta} + \frac{SD}{Q} + \frac{HQ}{2} \dots \dots \dots 73$$

associated to the constraints

$$wQ \leq W \text{ and } \dots \dots \dots 74$$

$$A^\beta D^{-\beta} Q \leq B \dots \dots \dots 75$$

The Lagrangean function corresponding to this objective function can be written as

$$G = A^\beta D^{1-\beta} + SDQ^{-1} + 0.5HQ - \epsilon_1(W - wQ - s_1^2) - \epsilon_2(B - A^\beta D^{-\beta} Q - s_2^2) \dots \dots \dots 76$$

Differentiating the above function partially with respect to D and Q gives the following derivatives.

$$\frac{\partial G}{\partial D} = (1 - \beta)A^\beta D^{-\beta} + SQ^{-1} - \epsilon_2\beta A^\beta D^{-\beta-1}Q \dots \dots \dots 77$$

By Karush Kuhn-Tucker conditions

$$\frac{\partial G}{\partial D} = 0 \Rightarrow (1 - \beta)A^\beta D^{-\beta} + SQ^{-1} - \varepsilon_2 \beta A^\beta D^{-\beta-1} Q = 0 \dots\dots\dots 78$$

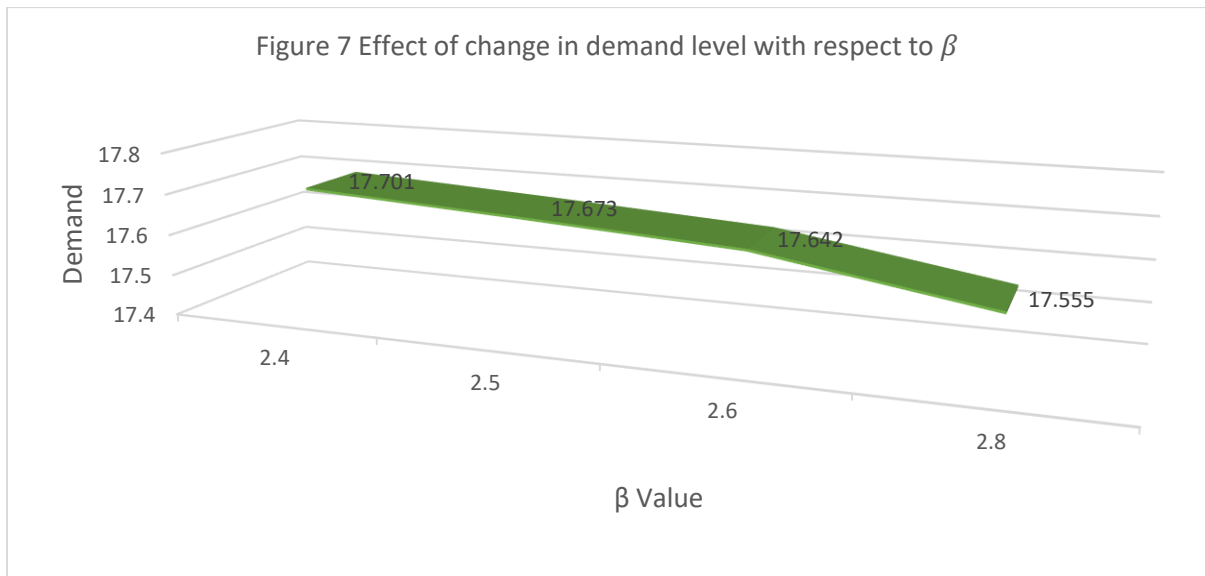
$$\frac{\partial G}{\partial D} = 0 \Rightarrow -SDQ^{-2} + 0.5H + \varepsilon_1 W + \varepsilon_2 A^\beta D^{-\beta} = 0 \dots\dots\dots 79$$

Solving the Equations (14) and (15) gives the required optimum solution. Ideal resolution in deprived of shortages is assumed in the subsequent Table 3 for the same set of input values given in Table I.

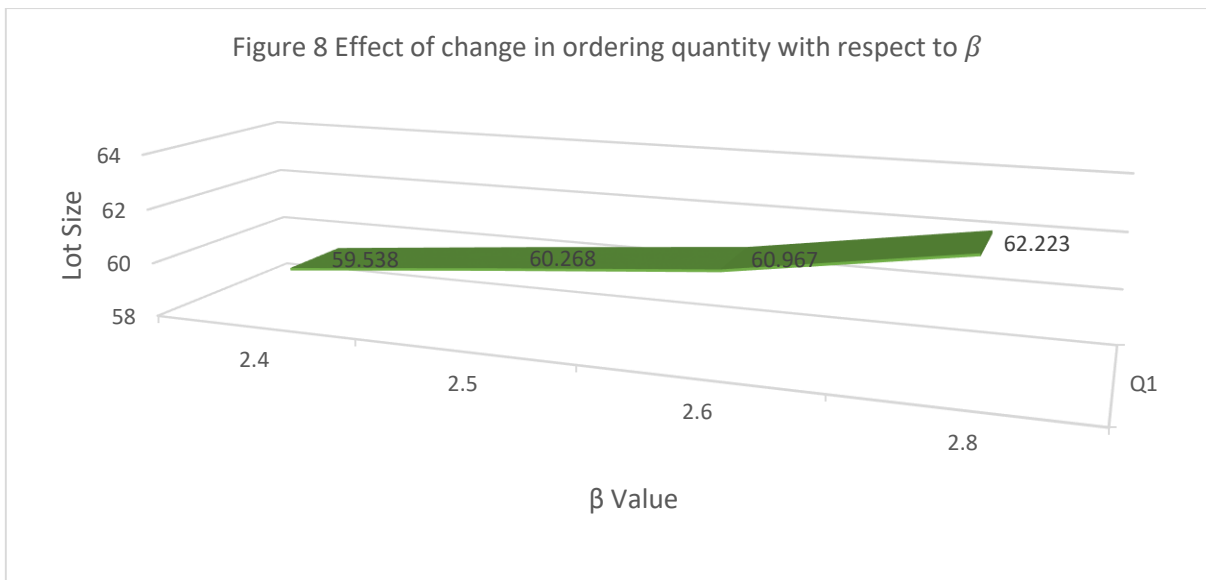
Table 3 Optimal values for various values of for no-shortage case

β	p_1	μ_{p_1} value	D_1	Q_1	Expected Total cost
2.4	1.3405	0.6595	17.701	59.538	56.519
2.5	1.3624	0.6376	17.673	60.268	56.282
2.6	1.3857	0.6143	17.642	60.967	56.059
2.8	1.4407	0.5593	17.555	62.223	55.649

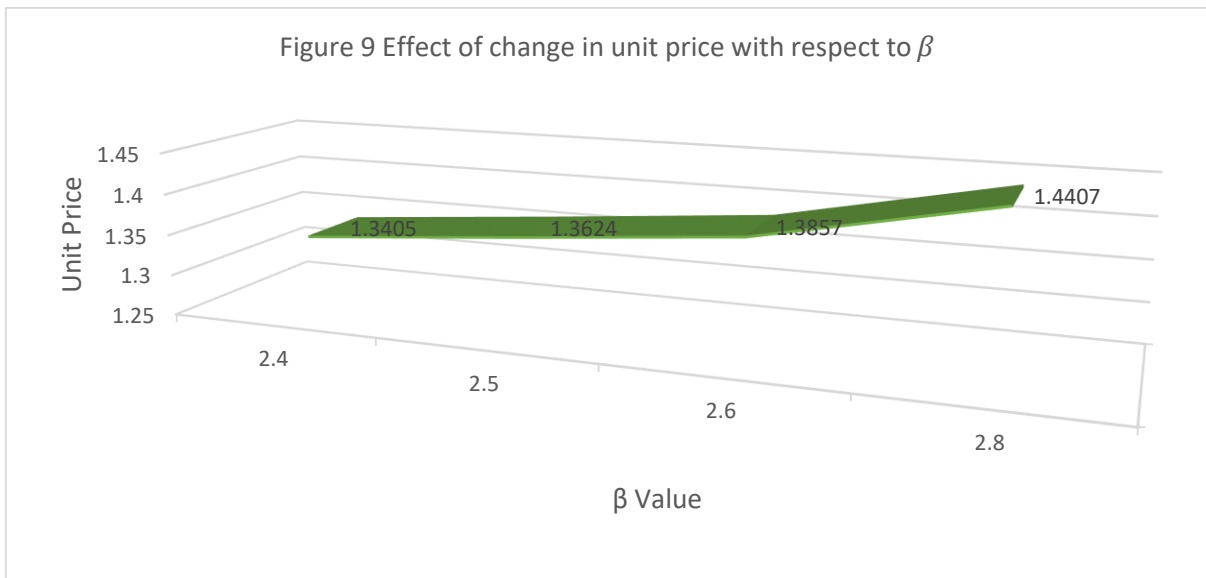
In this case, it follows that the minimum annual total cost corresponds to the determined relationship function value is 0.6595. Hence the optimal result satisfying the constraints are $D_1 = 17.701$, $Q_1 = 59.538$ and the minimum total cost is equal to 56.519. The graphical representations are shown in the Figure 7, 8,9 and 10.



The above Figure 7 shows that as the value of the parameter β increases from 2.4 to 2.8, the value of the demand decreases from 17.701 to 17.555.



The above Figure 8 shows that as the value of the parameter β increases from 2.4 to 2.8, the value of the lot size increases from 59.538 to 62.223.



The above Figure 9 shows that as the value of the parameter β increases from 2.4 to 2.8, the value of the unit price increases from 1.3405 to 1.4407.

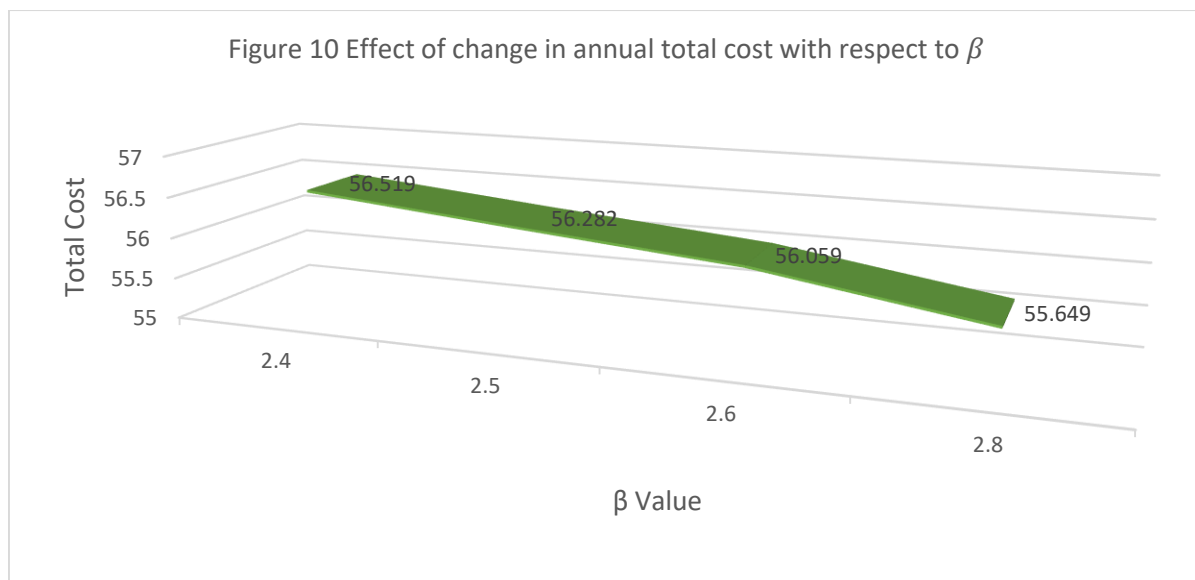


Figure 10 Effect of change in annual total cost with respect to β

The above Figure 10 shows that as the value of the parameter β increases from 2.4 to 2.8, the value of the total cost decreases from 56.519 to 55.649.

13. Summary

In this research article, a mathematical model for multiple articles through permitted and restricted shortage and per article cost based on demand accompanied by upper and lower limits viz restricted storage space and manufacturing expenses has been constructed. The article cost is explored at this juncture in a fuzzy atmosphere and solutions of the model being obtained through KKT condition. A comparison study of the findings for with shortage case (Table-2) and without shortage case (Table-3) is done. In the numerical examples (Table-1), it is found that the optimum total cost in the shortage scenario is less than that of the without shortage case.

The purpose behind developing this model is to provide several managerial insights and constructive conclusion such as: the model aids managers in deciding how best to distribute resources among various supply chain organizations. To maximise overall performance, managers may take well-informed decisions on production schedules, capacity utilisation, and cost-cutting strategies. Managers may minimise stockouts, assure product availability, and optimise inventory levels while taking capacity constraints and production costs into account. It offers information on how adjustments to production procedures, capacity use, and inventory allocation might result in cost reductions. It assists managers in determining the impact of variable demand, capacity constraints, or cost variations on inventory management. Organizations may use this information to establish risk-mitigation measures such as safety stock management, contingency planning, and alternate sourcing choices. The model serves as a framework for assessing and improving supply chain performance. Managers may identify areas for improvement and apply specific initiatives to increase overall performance by analysing key performance metrics such as inventory turnover, order fulfilment rates, and manufacturing cost ratios.

Ultimately, it helps businesses to make efficient decisions about inventory management, resource allocation, and cost reduction. Managers may boost efficiency, cut costs, enhance customer happiness, and establish a competitive edge in their respective sectors by recognising the interdependencies between storage space, production expenditures, and other restrictions.

One of the possible extensions of this chapter is to establish particular conditions that guarantee the global optimality of solutions. It is also conceivable to consider the life cycle to be a stochastic function. When this model would cover the time value of money and inflation, a better reflection of real-life scenarios could be presented.

14. References

- Bardhan, S., Pal, H., & Giri, B. C. (2019). Optimal replenishment policy and preservation technology investment for a non-instantaneous deteriorating item with stock-dependent demand. *Operational Research*, 19(2), 347-368.
- Barman, A., Das, R., & De, P. K. (2021). Optimal pricing, replenishment scheduling, and preservation technology investment policy for multi-item deteriorating inventory model under shortages. *International Journal of Modeling, Simulation, and Scientific Computing*, 12(05), 2150039.
- Erlenkotter, D. (1990). Ford Whitman Harris and the economic order quantity model. *Operations Research*, 38(6), 937-946.
- Hadley, G. W., & Whitten, T. (1963). TM, 1963. Analysis of Inventory systems. *Englewood Cliffs, NJ*, 1963.
- Mashud, A. H. M., Wee, H. M., & Huang, C. V. (2021). Preservation technology investment, trade credit and partial backordering

- model for a non-instantaneous deteriorating inventory. *RAIRO-Operations Research*, 55, S51-S77.
- Md. Alamin Khan, A. H. M. M., M. A. Halim. (2017). Numerous Exact Solutions Of Nonlinear Partial Differential Equations By Tan–Cot Method. *Journal of Mechanics of Continua Mathematical sciences*, 11(2), 37-48.
- Md Mashud, A. H., Pervin, M., Mishra, U., Daryanto, Y., Tseng, M.-L., & Lim, M. K. (2021). A sustainable inventory model with controllable carbon emissions in green-warehouse farms. *Journal of Cleaner Production*, 298, 126777. doi:<https://doi.org/10.1016/j.jclepro.2021.126777>
- Miah, M. S., Islam, M. M., Hasan, M., Mashud, A. H. M., Roy, D., & Sana, S. S. (2021). A Discount Technique-Based Inventory Management on Electronics Products Supply Chain. *Journal of Risk and Financial Management*, 14(9), 398.
- Mishra, V. K. (2014). Controllable deterioration rate for time-dependent demand and time-varying holding cost. *Yugoslav Journal of Operations Research*, 24(1), 87-98.
- Mishra, U., Mashud, A. H. M., Tseng, M.-L., & Wu, J.-Z. (2021). Optimizing a Sustainable Supply Chain Inventory Model for Controllable Deterioration and Emission Rates in a Greenhouse Farm. *Mathematics*, 9(5), 495.
- Nath, B. K., & Sen, N. (2021). A partially backlogged two-warehouse EOQ model with non-instantaneous deteriorating items, price and time dependent demand and preservation technology using interval number. *International Journal of Mathematics in Operational Research*, 20(2), 149-181.
- Pal, H., Bardhan, S., & Giri, B. C. (2018). Optimal replenishment policy for non-instantaneously perishable items with preservation technology and random deterioration start time. *International Journal of Management Science and Engineering Management*, 13(3), 188-199.
- Rahman, M. M., Ahmed, R., Mashud, A. H. M., Malik, A. I., Miah, S., & Abedin, M. Z. (2022). Consumption-Based CO2 Emissions on Sustainable Development Goals of SAARC Region. *Sustainability*, 14(3), 1467.
- Rahman, M. M., Anan, N., Mashud, A. H. M., Hasan, M., & Tseng, M.-L. (2022). Consumption-based CO2 emissions accounting and scenario simulation in Asia and the Pacific region. *Environmental Science and Pollution Research*, 29(23), 34607-34623. doi:[10.1007/s11356-021-18265-w](https://doi.org/10.1007/s11356-021-18265-w)
- Roy, D., Hasan, S. M. M., Rashid, M. M., Hezam, I. M., Al-Amin, M., Chandra Roy, T., . . . Mashud, A. H. M. (2022). A Sustainable Advance Payment Scheme for Deteriorating Items with Preservation Technology. *Processes*, 10(3), 546.
- Roy, D., & Mashud, A. H. M. (2022). Optimizing profit in a controlled environment: Assessing the synergy between preservation technology and cap-and-trade policy. *Journal of King Saud University - Science*, 34(3), 101899. doi:<https://doi.org/10.1016/j.jksus.2022.101899>
- Shaikh, A. A., Mashud, A. H. M., Uddin, M. S., & Khan, M. A.-A. (2017). Non-instantaneous deterioration inventory model with price and stock dependent demand for fully backlogged shortages under inflation. *International Journal of Business Forecasting and Marketing Intelligence*, 3(2), 152-164.
- Shaikh, A. A., Panda, G. C., Khan, M. A.-A., Mashud, A. H. M., & Biswas, A. (2020). An inventory model for deteriorating items with preservation facility of ramp type demand and trade credit. *Int. J. Math. Oper. Res.*, 17(4), 514-551.
- Silver, E. A., & Peterson, R. (1985). *Decision systems for inventory management and production planning* (Vol. 18). Wiley.
- Sultana, S., Mashud, A. H. M., Daryanto, Y., Miah, S., Alrasheedi, A., & Hezam, I. M. (2022). The Role of the Discount Policy of Prepayment on Environmentally Friendly Inventory Management. *Fractal and Fractional*, 6(1), 26.
- Yang, H. L., & Chang, C. T. (2013). A two-warehouse partial backlogging inventory model for deteriorating items with permissible delay in payment under inflation. *Applied Mathematical Modelling*, 37(5), 2717-2726.
- Zia, N. P., & Taleizadeh, A. A. (2015). A lot-sizing model with backordering under hybrid linked-to-order multiple advance payments and delayed payment. *Transportation Research Part E: Logistics and Transportation Review*, 82, 19-37.