IJCRT.ORG

ISSN: 2320-2882



INTERNATIONAL JOURNAL OF CREATIVE RESEARCH THOUGHTS (IJCRT)

An International Open Access, Peer-reviewed, Refereed Journal

Two-Godown Inventory Model for Deteriorating Items using Multi-Objective Genetic Algorithm

¹Veenita Sharma, ²A.B. Chandramouli, ³Ajay Singh Yadav

¹Research Scholar, Chaudhary Charan Singh University, Meerut, Uttar Pradesh, India.

²Department of Mathematics, Meerut College Meerut, Uttar Pradesh, India.

³Department of Mathematics, SRM Institute of Science and Technology, Delhi-NCR Campus, Ghaziabad, Uttar Pradesh, India.

Abstract

In this chapter a deterministic inventory model has been developed for deteriorating items having a ramp type demand with the effects of inflation with two-storage facilities using Genetic Algorithm. The owned godown (OG) has a fixed capacity of W units; the rented godown (RG) has unlimited capacity using Genetic Algorithm. Here, we assumed that the inventory holding cost in RG is higher than those in OWG. Shortages in inventory are allowed and partially backlogged and it is assumed that the inventory deteriorates over time at a variable deterioration rate using Genetic Algorithm. The effect of inflation has also been considered for various costs associated with the inventory system. Numerical example is also used to study the behaviour of the model using Genetic Algorithm. Cost minimization technique is used to get the expressions for total cost and other parameters using Genetic Algorithm.

Keywords:- Inventory model, owned godown, rented godown, deteriorating items and Genetic Algorithm

1. Introduction

Many researchers extended the EOQ model to time-varying demand patterns. Some researchers discussed of inventory models with linear trend in demand. The main limitations in linear-time varying demand rate is that it implies a uniform change in the demand rate per unit time. This rarely happens in the case of any commodity in the market. In recent years, some models have been developed with a demand rate that changes exponentially with time. For seasonal products like clothes, Air conditions etc. at the end of their seasons the demand of these items is observed to be exponentially decreasing for some initial period. Afterwards, the demand for the products becomes steady rather than decreasing exponentially. It is believed that such type of demand is quite realistic. Such type situation can be represented by ramp type demand rate.

An important issue in the inventory theory is related to how to deal with the unfulfilled demands which occur during shortages or stock outs. In most of the developed models researchers assumed that the shortages are either completely backlogged or completely lost. The first case, known as backordered or backlogging case, represent a situation where the unfulfilled demand is completely back ordered. In the second case, also known as lost sale case, we assume that the unfulfilled demand is completely lost.

Furthermore, when the shortages occur, some customers are willing to wait for backorder and others would turn to buy from other sellers. In many cases customers are conditioned to a shipping delay and may be willing to wait for a short time in order to get their first choice. For instance, for fashionable commodities and high-tech products with short product life cycle, the willingness of a customer to wait for backlogging is diminishing with the length of the waiting time. Thus the length of the waiting time for the next replenishment would determine whether the backlogging would be accepted or not. In many real life situations, during a shortage period, the longer the waiting time is, the smaller is the backlogging rate would be. Therefore, for realistic business situations the backlogging rate should be variable and dependent on the waiting time for the next replenishment. Many researchers have modified inventory policies by considering the "time proportional partial backlogging rate".

Supply chain management can be defined as: "Supply chain management is the coordination of production, stock, location and transportation between actors in supply chain to achieve the best combination of responsiveness and efficiency to a given market. Many researchers in the inventory system have focused on products that do not exceed deterioration. However, there are a number of things whose significance does not remain the same over time. The deterioration of these substances plays an important role and cannot be stored for long {Yaday, et. al. (1 to 10)}. Deterioration of an object can be described as deterioration, evaporation, obsolescence and loss of use or limit of an object, resulting in lower stock consumption compared to natural conditions. When commodities are placed in stock as inventory to meet future needs, there may be deterioration of items in the system of arithmetic that may occur for one or more reasons, etc. Storage conditions, weather or humidity. {Yaday, et. al. (11 to 20)}. Inach it is generally claimed that management owns a warehouse to store purchased inventory. However, management can, for a variety of reasons, buy or give more than it can store in its warehouse and name it OW, with an additional number in a rented warehouse called RW located near OW or slightly away from it {Yadav, et. al. (21 to 53)}. Inventory costs (including holding costs and depreciation costs) in RW are usually higher than OW costs due to additional costs of handling, equipment maintenance, etc. To reduce the cost of inventory will economically use RW products as soon as possible. Actual customer service is provided only by OW, and in order to reduce costs, RW stocks are first cleaned. Such arithmetic examples are called two arithmetic examples in the warehouse {Yadav and swami. (54 to 61)}. Management of supply of electronic storage devices and integration of environment and nervous networks {Yadav and Kumar (62)}. Analysis of seven supply chain management measures in improving the inventory of electronic devices for storage by sending an economic load using GA and PSO and analysis of supply chain management in improving the inventory of storage and equipment using genetic calculation and model design and analysis of chain inventory from bi warehouse and economic difficulty of freight transport using genetic calculation {Yadav, AS (63, 64, 65)}. Inventory policies of inventory and inventory requirements and different storage costs under allowable payments and inventory delays An

example of depreciation of goods and services of various types and costs of holding down a Business-Loan and an inventory model with sensitive needs of prices, holding costs in contrast to loans of business expenses under inflation {Swami, et. al. (66, 67, 68)}. The objectives of the Multiple Objective Genetic Algorithm and PSO, which include the improvement of supply and deficit inventory, inflation, and a calculation model based on a genetic calculation of scarcity and low inflation by PSO (Gupta, et. al. (69, 70,)). An example with two warehouses depreciation of items and storage costs under particle upgrade and an example with two warehouses of material damage and storage costs in inflation and soft computer techniques (Singh, et. al. (71, 72). Delayed alcohol supply management and refinement of particles and green cement supply system and inflation using particle enhancement and electronic inventory calculation system and distribution center using genetic calculations {Kumar, et. al. (73, 74.75)}. Example of depreciation inventory with two warehouses and stock-based stocks using a genetic inventory and vehicle inventory system for demand and inflation of stocks with two distribution centers using genetic inventory (Chauhan and Yadav (76, 77)). Marble Analysis Improvement of industrial reserves based on genetic engineering and multi-particle improvement {Pandey, et. al. (78). White wine industry in supply chain management using nervous networks {Ahlawat, et. al. (79)}. Best policy for importing damaged items immediately and payment of conditional delays under the supervision of two warehouses {Singh, et. al. (80)}.

2. Assumptions and Notations

In developing the mathematical model of the inventory system the following assumptions are being made: In developing the mathematical model of the inventory system the following assumptions are being made:

1. The demand rate
$$D(t) = (A_0 + 1)e^{-(\lambda_0 + 1)\left\{t - \left[t - (\mu_0 + 1)\right]H\left\{t - \left[\mu_0 + 1\right]\right\}\right\}}, (A_0 + 1) > 0, (\lambda_0 + 1) > 0$$

2. The backlogging rate is $\exp -(\delta_0 + 1)(t)$, when inventory is in shortage. The backlogging parameter δ_0 is a positive constant.

The variable rate of deterioration in both storehouses is taken as $\theta_0(t) = (\theta_0 + 1)t$ where $0 < (\theta_0 + 1) << 1$ and only applied to on hand inventory.

In addition, the following notations are used throughout this paper:

 $\Upsilon_{og}(t)$ The inventory level in OG at any time t. $\Upsilon_{rg}(t)$ The inventory level in RG at any time t. W_0 The capacity of the own godown. Q_o The ordering quantity per cycle. The Planning horizon.

 r_o Inflation rate.

 G_1 The holding cost per unit per unit time in OG.

www.ijcrt.org	© 2021 l
G_2	The holding cost per unit per unit time in R
G_d	The deterioration cost per unit.
G_3	The shortage cost per unit per unit time.
G_4	The opportunity cost due to lost sales.

The replenishment cost per order.

3. Formulation and Solution of the Model

The inventory levels at OW are governed by the following differential equations:

$$\frac{d\Upsilon_{og}(t)}{dt} = -(\theta_0 + 1)(t)\Upsilon(t) \qquad 0 \le t < \mu_0 \qquad \dots (1)$$

$$\frac{d\Upsilon_{og}(t)}{dt} + (\theta_0 + 1)(t)\Upsilon(t) = -(A_0 + 1)e^{-(\lambda_0 + 1)(\mu_0 + 1)}, \qquad \mu_0 \le t \le t_1 \qquad \dots (2)$$

And

 G_O

$$\frac{d\Upsilon_{og}(t)}{dt} = -(A_0 + 1)e^{-(\lambda_0 + 1)(\mu_0 + 1)}e^{-(\delta_0 + 1)t}, \qquad t_1 \le t \le T \qquad (3)$$

with the boundary conditions,

$$\Upsilon_{og}(0) = (W_0 + 1) \text{ and } \Upsilon_{og}(t_1) = 0$$
 ...(4

The solutions of equations (1), (2) and (3) are given by

$$\Upsilon_{og}(t) = (W_0 + 1)e^{-(\theta_0 + 1)t^2/2}, \qquad 0 \le t < \mu_0 \qquad \dots (5)$$

$$\Upsilon_{og}(t) = (A_0 + 1)e^{-(\lambda_0 + 1)(\mu_0 + 1)} \left\{ \frac{(t_1 + t) + (\theta_0 + 1)(t_1^3 - t^3)}{6} \right\} e^{-(\theta_0 + 1)t^2/2}, \qquad \mu_0 \le t \le t_1 \qquad \dots (6)$$

And
$$\Upsilon_{og}(t) = \frac{(A_0 + 1)}{(\delta_0 + 1)} e^{-(\lambda_0 + 1)(\mu_0 + 1)} \begin{cases} e^{-(\delta_0 + 1)t} - \\ e^{-(\delta_0 + 1)t_1} \end{cases}$$
, $t_1 \le t \le T$... (7)

respectively.

The inventory level at RW is governed by the following differential equations:

$$\frac{d\Upsilon_{rg}(t)}{dt} + (\theta_0 + 1)(t)I(t) = -(A_0 + 1)e^{-(\lambda_0 + 1)t}, \qquad 0 \le t < \mu_0 \qquad \dots (8)$$

With the boundary condition $I_r(0) = 0$, the solution of the equation (3.8) is

$$\Upsilon_{rg}(t) = (A_0 + 1) \left\{ \frac{((\mu_0 + 1) - t) - ((\lambda_0 + 1)^2 - t^2)}{2} ((\mu_0 + 1)^2 - t^2) + \right\} e^{-(\theta_0 + 1)t^2/2}, \quad \mu \le t \le t_1 \quad (9)$$

Due to continuity of $\Upsilon_{os}(t)$ at point $t = \mu_0$ it follows from equations (5) and (6), one has

$$(W_{0}+1)e^{-(\theta_{0}+1)(\mu_{0}+1)^{2}/2} = (A_{0}+1)e^{-(\lambda_{0}+1)(\mu_{0}+1)} \begin{cases} [t_{1}-(\mu_{0}+1)] + \\ (\theta_{0}+1)[t_{1}^{3}-(\mu_{0}+1)^{3}] \end{cases} e^{-(\theta_{0}+1)(\mu_{0}+1)^{2}/2}$$

$$(W_{0}+1) = (A_{0}+1)e^{-(\lambda_{0}+1)(\mu_{0}+1)} \begin{cases} [t_{1}-(\mu_{0}+1)] + \\ (\theta_{0}+1)[t_{1}^{3}-(\mu_{0}+1)^{3}] \end{cases} (10)$$

The total average cost consists of following elements:

- (i) Ordering cost per cycle = $(C_0 + 1)$ (11)
- (ii) Holding cost per cycle in OG

$$G_{HOG} = G_{I} \begin{bmatrix} (\mu_{0}+1) \\ \int \\ 0 \end{bmatrix} \Upsilon_{o}(t)e^{-(\eta_{0}+1)t} dt + \int \\ (\mu_{0}+1) \\ (\mu_{0}+1) - \frac{(\eta_{0}+1)(\mu_{0}+1)^{2} - (\theta_{0}+1)(\mu_{0}+1)^{3}}{2} - \frac{(\theta_{0}+1)(\mu_{0}+1)^{3}}{6} \end{bmatrix} + G_{HOG} = G_{I} \begin{bmatrix} (\mu_{0}+1) \\ (\mu_{0}+1) - \frac{(\eta_{0}+1)(\mu_{0}+1)^{2} - (\theta_{0}+1)(\mu_{0}+1)^{3}}{2} - \frac{(\eta_{0}+1)t_{1}^{4}}{12} \\ - \frac{(\eta_{0}+1)(\theta_{0}+1)t_{1}^{5} - (\mu_{0}+1)[2t_{1}-(\mu_{0}+1)]}{2} - \frac{(\theta_{0}+1)(\mu_{0}+1)^{2}(3t_{1}-2(\mu_{0}+1))}{6} \\ + \frac{(\eta_{0}+1)(\mu_{0}+1)^{2}(3t_{1}-2(\mu_{0}+1))}{6} \\ + \frac{(\eta_{0}+1)(\mu_{0}+1)^{2}(5t_{1}^{3}-3(\mu_{0}+1)^{3})}{30} \\ + \frac{(\theta_{0}+1)(\mu_{0}+1)^{3}(4t_{1}-3(\mu_{0}+1))}{24} \end{bmatrix} \end{bmatrix}$$
(12)

(iii) Holding cost per cycle in RG

$$G_{HRG} = G_2 \begin{bmatrix} \left(\mu_0 + 1\right) \\ \int \\ 0 \end{bmatrix} \Upsilon_o(t) e^{-\left(\eta_0 + 1\right)t} dt$$

$$G_{HGC} = G_2 \left[(A_0 + 1) \begin{cases} \frac{(\mu_0 + 1)^2}{2} - \frac{\left[3(\lambda_0 + 1) + (r_0 + 1)\right](\mu_0 + 1)^3}{6} + \left(\frac{(\theta_0 + 1)}{12} + \frac{(\lambda_0 + 1)(r_0 + 1)}{8}\right)(\mu_0 + 1)^4 \\ -\left(\frac{(r_0 + 1)(\theta_0 + 1)}{20} - \frac{(\lambda_0 + 1)(\theta_0 + 1)}{30}\right)(\mu_0 + 1)^5 \end{cases} \right]$$

(13)

(iv) Cost of deteriorated units per cycle



$$G_{DC} = G_{d} \begin{bmatrix} (\mu_{0}+1) & (\partial_{0}+1)t Y_{o}(t) e^{-(\eta_{0}+1)t} dt + \int_{0}^{(\mu_{0}+1)} (\partial_{0}+1)t Y_{o}(t) e^{-(\eta_{0}+1)t} dt + \int_{0}^{t} (\partial_{0}+1)t Y_{o}(t) e^{-(\eta_{0}+1)t} dt + \int_{0}^{t} (\partial_{0}+1)t Y_{o}(t) e^{-(\eta_{0}+1)(t+(\mu_{0}+1))} dt \end{bmatrix}$$

$$= G_{d} (\partial_{0}+1) \left\{ (M_{0}+1) \left\{ \frac{(\partial_{0}+1)}{4} + \frac{(\partial_{0}+1)}{4} + \frac{(\partial_{0}+1)}{12} (\mu_{0}+1)^{5} - \frac{(\eta_{0}+1)(\partial_{0}+1)}{36} - \frac{(\eta_{0}+1)(\partial_{0}+1)}{24} \right\} + \left\{ \frac{(\partial_{0}+1)}{2} - \frac{(\eta_{0}+1)(\mu_{0}+1)^{3}}{3} + \frac{(\partial_{0}+1)(\mu_{0}+1)^{4}}{8} \right\} + \left\{ \frac{(\partial_{0}+1)}{6} - \frac{(\eta_{0}+1)(\partial_{0}+1)(\mu_{0}+1)^{4}}{40} - \frac{(\eta_{0}+1)(\partial$$

(v) Shortage cost per cycle

$$G_{SC} = G_3 \left[\int_{t_1}^{T} -\Upsilon_o(t) e^{-(r_0 + 1)(t_1 + t)} dt \right]$$

$$= \left[\frac{-(A_0+1)G_3 e^{-\left[\binom{r_0+1}{t_0}t_1 + \binom{r_0+1}{t_0}\right]}}{(\delta_0+1)} \right] \begin{cases} \int_{t_1}^{T} e^{-(r_0+1)+(\delta_0+1)t} dt - \frac{r_0}{t_0} \\ e^{-(\delta_0+1)t_1} \int_{t_1}^{T} e^{-(r_0+1)t} dt \end{cases}$$

$$= \left[\frac{-(A_0+1)G_3 e^{-\left[\binom{r_0+1}{t_0}t_1 + \binom{r_0+1}{t_0}\right]}}{(\delta_0+1)(r_0+1)\left[(\delta_0+1)+(r_0+1)\right]} \right\} \begin{cases} (\delta_0+1)e^{-\left[(\delta_0+1)+(r_0+1)\right]t_1} + \frac{r_0}{t_0} \\ e^{-r_0T} \left\{ (r_0+1)e^{-(\delta_0+1)T} - \binom{r_0}{t_0} \right\} \end{cases}$$

$$= \left[\frac{-(A_0+1)G_3 e^{-\left[\binom{r_0+1}{t_0}t_1 + \binom{r_0}{t_0}\right]}}{(\delta_0+1)(r_0+1)\left[(\delta_0+1)+(r_0+1)\right]} \right\}$$

$$= \left[\frac{-(A_0+1)G_3 e^{-\left[\binom{r_0+1}{t_0}t_1 + \binom{r_0}{t_0}\right]}}{(\delta_0+1)(r_0+1)\left[\binom{r_0+1}{t_0} + \binom{r_0+1}{t_0}\right]} \right]$$

$$= \left[\frac{-(A_0+1)G_3 e^{-\left[\binom{r_0+1}{t_0}t_1 + \binom{r_0+1}{t_0}\right]}}{(\delta_0+1)(r_0+1)\left[\binom{r_0+1}{t_0} + \binom{r_0+1}{t_0}\right]}} \right]$$

$$= \left[\frac{-(A_0+1)G_3 e^{-\left[\binom{r_0+1}{t_0}t_1 + \binom{r_0+1}{t_0}\right]}}{(\delta_0+1)(r_0+1)\left[\binom{r_0+1}{t_0} + \binom{r_0+1}{t_0}\right]}} \right]$$

$$= \left[\frac{-(A_0+1)G_3 e^{-\left[\binom{r_0+1}{t_0}t_1 + \binom{r_0+1}{t_0}\right]}}{(\delta_0+1)(r_0+1)\left[\binom{r_0+1}{t_0} + \binom{r_0+1}{t_0}\right]}} \right]$$

$$= \left[\frac{-(A_0+1)G_3 e^{-\left[\binom{r_0+1}{t_0}t_1 + \binom{r_0+1}{t_0}\right]}}{(\delta_0+1)(r_0+1)\left[\binom{r_0+1}{t_0} + \binom{r_0+1}{t_0}\right]}} \right]$$

$$= \left[\frac{-(A_0+1)G_3 e^{-\left[\binom{r_0+1}{t_0}t_1 + \binom{r_0+1}{t_0}\right]}}{(\delta_0+1)(r_0+1)\left[\binom{r_0+1}{t_0} + \binom{r_0+1}{t_0}\right]}} \right]$$

$$= \left[\frac{-(A_0+1)G_3 e^{-\left[\binom{r_0+1}{t_0}t_1 + \binom{r_0+1}{t_0}\right]}}{(\delta_0+1)(r_0+1)\left[\binom{r_0+1}{t_0} + \binom{r_0+1}{t_0}\right]}} \right]$$

(vi) Opportunity cost due to lost sales per cycle

$$G_{OPC} = G_4 \int_{t_1}^{T} (A_0 + 1) \left[1 - e^{-(\delta_0 + 1)t} \right] e^{-(\lambda_0 + 1)(\mu_0 + 1)} e^{-(\eta_0 + 1)(t_1 + t)} dt$$

$$= \frac{G_4 (A + 1)e^{-\left[(\lambda_0 + 1)(\mu_0 + 1) + \right]}}{(\eta_0 + 1) \left[(\eta_0 + 1) + \right]} \left[e^{-(\eta_0 + 1)t_1} \left\{ (\delta_0 + 1) + (\eta_0 + 1) - (\eta_0 + 1)t_1 \right\} - (\eta_0 + 1) e^{-(\delta_0 + 1)T} \right\}$$

$$= \frac{G_4 (A + 1)e^{-\left[(\lambda_0 + 1)(\mu_0 + 1) + \right]}}{(\eta_0 + 1) \left[(\eta_0 + 1) + (\eta_0 + 1) - (\eta_0 + 1) + ($$

Therefore, the total average cost per unit time of our model is obtained as follows

$$K(t_1, T) = \frac{1}{T} \left[G_{OC} + G_{HOS} + G_{HRS} + G_{DC} + G_{SC} + G_{OPC} \right]$$
(17)

4. Multi-Objective Genetic Algorithm

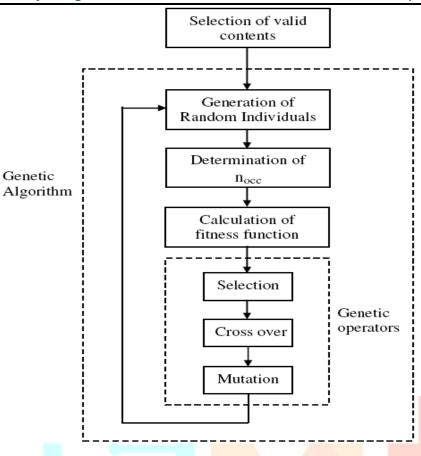
Discussions so far were limited to GA that handled the optimization of a single parameter. The optimization criteria are represented by fitness functions and are used to lead towards an acceptable solution. A typical single objective optimization problem is the TSP. There the sole optimization criterion is the cost of the tour undertaken by the salesperson and this cost is to be minimized. However, In real life we often face problem which require simultaneous optimization of several criteria. For example, in VLSI circuit design the critical parameters are chip area power consumption delay fault tolerance etc. While designing a VLSI circuit the designer may like to minimize area power consumption and delay while at the same time would like to maximize fault tolerance. The problem gets more complicated when the optimizing criteria are conflicting. For instance an attempt to design low-power VLSI circuit may affect its fault tolerance capacity adversely. Such problems are known as multi-objective optimization (MOO). Multi-objective optimization is the process of systematically and simultaneously optimizing a number of objective functions. Multiple objective problems usually have conflicting objectives which prevents simultaneous optimization of each objective. As GAs are population based optimization processes they are inherently suited to solve MOO problem. However traditional GAs are to be customized to accommodate such problem. This is achieved by using specialized

fitness functions as well as incorporating methods promoting solution diversity. Rest of this section presents the features of multi-objective GAs.

Multi-objective GA is designed by incorporating pareto-ranked niche count based fitness sharing into the traditional GA process.

This is presented as **Procedure Multi-Objective-GA**

- Step 1:-Generate the initial population randomly.
- Step 2:determine the pareto-optimal fronts U_{po1} , U_{po2} ,..... U_{pok} .
- Step 3:-If stopping criteria is satisfied then Return the pareto-optimal front U_{pol} Stop
- Step 4:for each solution x—p, evaluate the fitness.
- Step 5:-Generate the matting pool MP from population P applying appropriate selection operator.
- Apply crossover and mutation operations on the chromosomes of the mating pool to produce Step 6:the next generation p' of population from MP.
- Step 7:-Replace the old generation of population p by the new generation of 1JCR1
 - Population p'
- Step 8:-Go to step 2.



5. Numerical Illustration

To illustrate the model numerically the following parameter values are considered.

 A_0 =50 units, G_0 = Rs. 100 per order, r_0 = 0.05 unit, G_1 = Rs. 3.0 per unit per year, G_3 = Rs. 12.0 per unit per year, G_4 = Rs. 4.0 per unit, G_4 = Rs. 4.0 per unit, G_4 = Rs. 10.0 per

S = 38.597235 units and K = Rs.158.115354 per year.

Table 1:- The minimization of total average cost

P	WW	$K(t_1,T)$			
1	OPT	BEST	MAX	AVG	STD
1	72805.50	72805.50	2888.75	2808.00	77.25
2	72877.00	72877.00	2885.75	2878.75	72.69
3	72058.00	72058.00	2058.00	2058.00	78.00
4	72079.00	72079.00	2079.00	2079.00	72.00
5	72088.25	72088.25	2022.50	2022.70	78.58
6	72824.50	72824.50	2874.60	2824.70	72.88
7	72074.50	72074.50	2874.70	2068.70	77.85
8	72850.50	72850.50	2850.00	2840.00	70.48
9	72286.50	72286.50	2206.50	2286.50	70.75
10	72082.00	72082.00	2087.00	2084.00	77.50

Table 2:- GA Results

P	WW	GA			
1	OPT	BEST	MAX	AVG	STD
1	82205.50	82205 <mark>.50</mark>	82288.75	82208.00	84.25
2	82277.00	82277 <mark>.00</mark>	82285.7 <mark>5</mark>	82778.75	85.69
3	82858.00	82858 <mark>.00</mark>	82858.00	82258.00	82.50
4	82879.00	82879 <mark>.00</mark>	82879.00	82879.00	84.05
5	82888.25	82888.25	82822.50	82222.70	82.78
6	82224.50	82224.50	82274.60	82524.70	84.88
7	82874.50	82874.50	82274.70	82868 <mark>.70</mark>	84.85
8	82250.50	82250.50	82250.00	82440 <mark>.00</mark>	88.48
9	82786.50	82786.50	82706.50	82786 <mark>.50</mark>	82.75
10	82882.00	82882.00	82887.00	82884.00	86.50

6. Conclusion

This study incorporates some realistic features that are likely to be associated with the inventory of any material. Decay (deterioration) overtime for any material product and occurrence of shortages in inventory are natural phenomenon in real situations using Genetic Algorithm. Inventory shortages are allowed in the model. In many cases customers are conditioned to a shipping delay, and may be willing to wait for a short time in order to get their first choice using Genetic Algorithm. Generally speaking, the length of the waiting time for the next replenishment is the main factor for deciding whether the backlogging will be accepted or not using Genetic Algorithm. The willingness of a customer to wait for backlogging during a shortage period declines with the length of the waiting time. Thus, inventory shortages are allowed and partially backordered in the present chapter and the backlogging rate is considered as a decreasing function of the waiting time for the next replenishment using Genetic Algorithm. Demand rate is taken as exponential ramp type function of time, in which demand decreases exponentially for the some initial period and becomes steady later on using Genetic Algorithm. Since most decision makers think that the inflation does not have significant influence on

the inventory policy, the effects of inflation are not considered in some inventory models. However, from a financial point of view, an inventory represents a capital investment and must complete with other assets for a firm's limited capital funds using Genetic Algorithm. Thus, it is necessary to consider the effects of inflation on the inventory system using Genetic Algorithm. Therefore, this concept is also taken in this model using Genetic Algorithm. From the numerical illustration of the model, it is observed that the period in which inventory holds increases with the increment in backlogging and ramp parameters while inventory period decreases with the increment in deterioration and inflation parameters. Initial inventory level decreases with the increment in deterioration, inflation and ramp parameters while inventory level increases with the increment in backlogging parameter using Genetic Algorithm. The total average cost of the system goes on increasing with the increment in the backlogging and deterioration parameters while it decreases with the increment in inflation and ramp parameters. The proposed model can be further extended in several ways. For example, we could extend this deterministic model in to stochastic model using Genetic Algorithm. Also, we could extend the model to incorporate some more realistic features, such as quantity discount or the unit purchase cost, the inventory holding cost and others can also taken fluctuating with time using Genetic Algorithm.

References:

- [1] Yadav, A.S., Bansal, K.K., Shivani, Agarwal, S. And Vanaja, R. (2020) FIFO in Green Supply Chain Inventory Model of Electrical Components Industry With Distribution Centres Using Particle Swarm Optimization. Advances in Mathematics: Scientific Journal. 9 (7), 5115–5120.
- Yadav, A.S., Kumar, A., Agarwal, P., Kumar, T. And Vanaja, R. (2020) LIFO in Green Supply Chain Inventory Model of Auto-Components Industry with Warehouses Using Differential Evolution. Advances in Mathematics: Scientific Journal, 9 no.7, 5121–5126.
- [3] Yadav, A.S., Abid, M., Bansal, S., Tyagi, S.L. And Kumar, T. (2020) FIFO & LIFO in Green Supply Chain Inventory Model of Hazardous Substance Components Industry with Storage Using Simulated Annealing. Advances in Mathematics: Scientific Journal, 9 no.7, 5127–5132.
- [4] Yadav, A.S., Tandon, A. and Selva, N.S. (2020) National Blood Bank Centre Supply Chain Management For Blockchain Application Using Genetic Algorithm. International Journal of Advanced Science and Technology Vol. 29, No. 8s, 1318-1324.
- [5] Yadav, A.S., Selva, N.S. and Tandon, A. (2020) Medicine Manufacturing Industries supply chain management for Blockchain application using artificial neural networks, International Journal of Advanced Science and Technology Vol. 29, No. 8s, 1294-1301.
- [6] Yadav, A.S., Ahlawat, N., Agarwal, S., Pandey, T. and Swami, A. (2020) Red Wine Industry of Supply Chain Management for Distribution Center Using Neural Networks, Test Engraining & Management, Volume 83 Issue: March April, 11215 11222.

- [7] Yadav, A.S., Pandey, T., Ahlawat, N., Agarwal, S. and Swami, A. (2020) Rose Wine industry of Supply Chain Management for Storage using Genetic Algorithm. Test Engraining & Management, Volume 83 Issue: March April, 11223 11230.
- [8] Yadav, A.S., Ahlawat, N., Sharma, N., Swami, A. And Navyata (2020) Healthcare Systems of Inventory Control For Blood Bank Storage With Reliability Applications Using Genetic Algorithm. Advances in Mathematics: Scientific Journal 9 no.7, 5133–5142.
- [9] Yadav, A.S., Dubey, R., Pandey, G., Ahlawat, N. and Swami, A. (2020) Distillery Industry Inventory Control for Storage with Wastewater Treatment & Logistics Using Particle Swarm Optimization Test Engraining & Management Volume 83 Issue: May June, 15362-15370.
- [10] Yadav, A.S., Ahlawat, N., Dubey, R., Pandey, G. and Swami, A. (2020) Pulp and paper industry inventory control for Storage with wastewater treatment and Inorganic composition using genetic algorithm (ELD Problem). Test Engraining & Management, Volume 83 Issue: May June, 15508-15517.
- [11] Yadav, A.S., Pandey, G., Ahlawat, N., Dubey, R. and Swami, A. (2020) Wine Industry Inventory Control for Storage with Wastewater Treatment and Pollution Load Using Ant Colony Optimization Algorithm, Test Engraining & Management, Volume 83 Issue: May June, 15528-15535.
- [12] Yadav, A.S., Navyata, Sharma, N., Ahlawat, N. and Swami, A. (2020) Reliability Consideration costing method for LIFO Inventory model with chemical industry warehouse. International Journal of Advanced Trends in Computer Science and Engineering, Volume 9 No 1, 403-408.
- [13] Yadav, A.S., Bansal, K.K., Kumar, J. and Kumar, S. (2019) Supply Chain Inventory Model For Deteriorating Item With Warehouse & Distribution Centres Under Inflation. International Journal of Engineering and Advanced Technology, Volume-8, Issue-2S2, 7-13.
- [14] Yadav, A.S., Kumar, J., Malik, M. and Pandey, T. (2019) Supply Chain of Chemical Industry For Warehouse With Distribution Centres Using Artificial Bee Colony Algorithm. International Journal of Engineering and Advanced Technology, Volume-8, Issue-2S2, 14-19.
- [15] Yadav, A.S., Navyata, Ahlawat, N. and Pandey, T. (2019) Soft computing techniques based Hazardous Substance Storage Inventory Model for decaying Items and Inflation using Genetic Algorithm. International Journal of Advance Research and Innovative Ideas in Education, Volume 5 Issue 9, 1102-1112.
- [16] Yadav, A.S., Navyata, Ahlawat, N. and Pandey, T. (2019) Hazardous Substance Storage Inventory Model for decaying Items using Differential Evolution. International Journal of Advance Research and Innovative Ideas in Education, Volume 5 Issue 9, 1113-1122.
- [17] Yadav, A.S., Navyata, Ahlawat, N. and Pandey, T. (2019) Probabilistic inventory model based Hazardous Substance Storage for decaying Items and Inflation using Particle Swarm Optimization. International Journal of Advance Research and Innovative Ideas in Education, Volume 5 Issue 9, 1123-1133.

- [18] Yadav, A.S., Navyata, Ahlawat, N. and Pandey, T. (2019) Reliability Consideration based Hazardous Substance Storage Inventory Model for decaying Items using Simulated Annealing. International Journal of Advance Research and Innovative Ideas in Education, Volume 5 Issue 9, 1134-1143.
- [19] Yadav, A.S., Swami, A. and Kher, G. (2019) Blood bank supply chain inventory model for blood collection sites and hospital using genetic algorithm. Selforganizology, Volume 6 No.(3-4), 13-23.
- [20] Yadav, A.S., Swami, A. and Ahlawat, N. (2018) A Green supply chain management of Auto industry for inventory model with distribution centers using Particle Swarm Optimization. Selforganizology, Volume 5 No. (3-4)
- [21] Yadav, A.S., Ahlawat, N., and Sharma, S. (2018) Hybrid Techniques of Genetic Algorithm for inventory of Auto industry model for deteriorating items with two warehouses. International Journal of Trend in Scientific Research and Development, Volume 2 Issue 5, 58-65.
- [22] Yadav, A.S., Swami, A. and Gupta, C.B. (2018) A Supply Chain Management of Pharmaceutical For Deteriorating Items Using Genetic Algorithm. International Journal for Science and Advance Research In Technology, Volume 4 Issue 4, 2147-2153.
- [23] Yadav, A.S., Maheshwari, P., Swami, A., and Pandey, G. (2018) A supply chain management of chemical industry for deteriorating items with warehouse using genetic algorithm. Selforganizology, Volume 5 No.1-2, 41-51.
- [24] Yadav, A.S., Garg, A., Gupta, K. and Swami, A. (2017) Multi-objective Genetic algorithm optimization in Inventory model for deteriorating items with shortages using Supply Chain management. IPASJ International journal of computer science, Volume 5, Issue 6, 15-35.
- Yadav, A.S., Garg, A., Swami, A. and Kher, G. (2017) A Supply Chain management in Inventory Optimization for deteriorating items with Genetic algorithm. International Journal of Emerging Trends & Technology in Computer Science, Volume 6, Issue 3, 335-352.
- [26] Yadav, A.S., Maheshwari, P., Garg, A., Swami, A. and Kher, G. (2017) Modeling & Analysis of Supply Chain management in Inventory Optimization for deteriorating items with Genetic algorithm and Particle Swarm optimization. International Journal of Application or Innovation in Engineering & Management, Volume 6, Issue 6, 86-107.
- [27] Yadav, A.S., Garg, A., Gupta, K. and Swami, A. (2017) Multi-objective Particle Swarm optimization and Genetic algorithm in Inventory model for deteriorating items with shortages using Supply Chain management. International Journal of Application or Innovation in Engineering & Management, Volume 6, Issue 6, 130-144.
- [28] Yadav, A.S., Swami, A. and Kher, G. (2017) Multi-Objective Genetic Algorithm Involving Green Supply Chain Management International Journal for Science and Advance Research In Technology, Volume 3 Issue 9, 132-138.

- [29] Yadav, A.S., Swami, A., Kher, G. (2017) Multi-Objective Particle Swarm Optimization Algorithm Involving Green Supply Chain Inventory Management. International Journal for Science and Advance Research In Technology, Volume 3 Issue, 240-246.
- [30] Yadav, A.S., Swami, A. and Pandey, G. (2017) Green Supply Chain Management for Warehouse with Particle Swarm Optimization Algorithm. International Journal for Science and Advance Research in Technology, Volume 3 Issue 10, 769-775.
- [31] Yadav, A.S., Swami, A., Kher, G. and Garg, A. (2017) Analysis of seven stages supply chain management in electronic component inventory optimization for warehouse with economic load dispatch using genetic algorithm. Selforganizology, 4 No.2, 18-29.
- [32] Yadav, A.S., Maheshwari, P., Swami, A. and Garg, A. (2017) Analysis of Six Stages Supply Chain management in Inventory Optimization for warehouse with Artificial bee colony algorithm using Genetic Algorithm. Selforganizology, Volume 4 No.3, 41-51.
- [33] Yadav, A.S., Swami, A., Gupta, C.B. and Garg, A. (2017) Analysis of Electronic component inventory Optimization in Six Stages Supply Chain management for warehouse with ABC using genetic algorithm and PSO. Selforganizology, Volume 4 No.4, 52-64.
- [34] Yadav, A.S., Maheshwari, P. and Swami, A. (2016) Analysis of Genetic Algorithm and Particle Swarm Optimization for warehouse with Supply Chain management in Inventory control. International Journal of Computer Applications, Volume 145 –No.5, 10-17.
- [35] Yadav, A.S., Swami, A. and Kumar, S. (2018) Inventory of Electronic components model for deteriorating items with warehousing using Genetic Algorithm. International Journal of Pure and Applied Mathematics, Volume 119 No. 16, 169-177.
- [36] Yadav, A.S., Johri, M., Singh, J. and Uppal, S. (2018) Analysis of Green Supply Chain Inventory Management for Warehouse With Environmental Collaboration and Sustainability Performance Using Genetic Algorithm. International Journal of Pure and Applied Mathematics, Volume 118 No. 20, 155-161.
- [37] Yadav, A.S., Ahlawat, N., Swami, A. and Kher, G. (2019) Auto Industry inventory model for deteriorating items with two warehouse and Transportation Cost using Simulated Annealing Algorithms. International Journal of Advance Research and Innovative Ideas in Education, Volume 5,Issue 1, 24-33.
- [38] Yadav, A.S., Ahlawat, N., Swami, A. and Kher, G. (2019) A Particle Swarm Optimization based a two-storage model for deteriorating items with Transportation Cost and Advertising Cost: The Auto Industry. International Journal of Advance Research and Innovative Ideas in Education, Volume 5, Issue 1, 34-44.
- [39] Yadav, A.S., Ahlawat, N., and Sharma, S. (2018) A Particle Swarm Optimization for inventory of Auto industry model for two warehouses with deteriorating items. International Journal of Trend in Scientific Research and Development, Volume 2 Issue 5, 66-74.

- [40] Yadav, A.S., Swami, A. and Kher, G. (2018) Particle Swarm optimization of inventory model with two-warehouses. Asian Journal of Mathematics and Computer Research, Volume 23 No.1, 17-26.
- [41] Yadav, A.S., Maheshwari, P.,, Swami, A. and Kher, G. (2017) Soft Computing Optimization of Two Warehouse Inventory Model With Genetic Algorithm. Asian Journal of Mathematics and Computer Research, Volume 19 No.4, 214-223.
- [42] Yadav, A.S., Swami, A., Kumar, S. and Singh, R.K. (2016) Two-Warehouse Inventory Model for Deteriorating Items with Variable Holding Cost, Time-Dependent Demand and Shortages. IOSR Journal of Mathematics, Volume 12, Issue 2 Ver. IV, 47-53.
- [43] Yadav, A.S., Sharam, S. and Swami, A. (2016) Two Warehouse Inventory Model with Ramp Type Demand and Partial Backordering for Weibull Distribution Deterioration. International Journal of Computer Applications, Volume 140 –No.4, 15-25.
- [44] Yadav, A.S., Swami, A. and Singh, R.K. (2016) A two-storage model for deteriorating items with holding cost under inflation and Genetic Algorithms. International Journal of Advanced Engineering, Management and Science, Volume -2, Issue-4, 251-258.
- [45] Yadav, A.S., Swami, A., Kher, G. and Kumar, S. (2017) Supply Chain Inventory Model for Two Warehouses with Soft Computing Optimization. International Journal of Applied Business and Economic Research, Volume 15 No 4, 41-55.
- [46] Yadav, A.S., Rajesh Mishra, Kumar, S. and Yadav, S. (2016) Multi Objective Optimization for Electronic Component Inventory Model & Deteriorating Items with Two-warehouse using Genetic Algorithm. International Journal of Control Theory and applications, Volume 9 No.2, 881-892.
- Yadav, A.S., Gupta, K., Garg, A. and Swami, A. (2015) A Soft computing Optimization based Two Ware-House Inventory Model for Deteriorating Items with shortages using Genetic Algorithm. International Journal of Computer Applications, Volume 126 No.13, 7-16.
- [48] Yadav, A.S., Gupta, K., Garg, A. and Swami, A. (2015) A Two Warehouse Inventory Model for Deteriorating Items with Shortages under Genetic Algorithm and PSO. International Journal of Emerging Trends & Technology in Computer Science, Volume 4, Issue 5(2), 40-48.
- [49] Yadav, A.S. Swami, A., and Kumar, S. (2018) A supply chain Inventory Model for decaying Items with Two Ware-House and Partial ordering under Inflation. International Journal of Pure and Applied Mathematics, Volume 120 No 6, 3053-3088.
- [50] Yadav, A.S. Swami, A. and Kumar, S. (2018) An Inventory Model for Deteriorating Items with Two warehouses and variable holding Cost. International Journal of Pure and Applied Mathematics, Volume 120 No 6, 3069-3086.
- [51] Yadav, A.S., Taygi, B., Sharma, S. and Swami, A. (2017) Effect of inflation on a two-warehouse inventory model for deteriorating items with time varying demand and shortages. International Journal Procurement Management, Volume 10, No. 6, 761-775.

- [52] Yadav, A.S., R. P. Mahapatra, Sharma, S. and Swami, A. (2017) An Inflationary Inventory Model for Deteriorating items under Two Storage Systems. International Journal of Economic Research, Volume 14 No.9, 29-40.
- [53] Yadav, A.S., Sharma, S. and Swami, A. (2017) A Fuzzy Based Two-Warehouse Inventory Model For Non instantaneous Deteriorating Items With Conditionally Permissible Delay In Payment. International Journal of Control Theory And Applications, Volume 10 No.11, 107-123.
- [54] Yadav, A.S. and Swami, A. (2018) Integrated Supply Chain Model for Deteriorating Items With Linear Stock Dependent Demand Under Imprecise And Inflationary Environment. International Journal Procurement Management, Volume 11 No 6, 684-704.
- [55] Yadav, A.S. and Swami, A. (2018) A partial backlogging production-inventory lot-size model with time-varying holding cost and weibull deterioration. International Journal Procurement Management, Volume 11, No. 5, 639-649.
- [56] Yadav, A.S. and Swami, A. (2013) A Partial Backlogging Two-Warehouse Inventory Models For Decaying Items With Inflation. International Organization of Scientific Research Journal of Mathematics, Issue 6, 69-78.
- Yadav, A.S. and Swami, A. (2019) An inventory model for non-instantaneous deteriorating items with variable holding cost under two-storage. International Journal Procurement Management, Volume 12 No 6, 690-710.
- [58] Yadav, A.S. and Swami, A. (2019) A Volume Flexible Two-Warehouse Model with Fluctuating Demand and Holding Cost under Inflation. International Journal Procurement Management, Volume 12 No 4, 441-456.
- [59] Yadav, A.S. and Swami, A. (2014) Two-Warehouse Inventory Model for Deteriorating Items with Ramp-Type Demand Rate and Inflation. American Journal of Mathematics and Sciences Volume 3 No-1, 137-144.
- [60] Yadav, A.S. and Swami, A. (2013) Effect of Permissible Delay on Two-Warehouse Inventory Model for Deteriorating items with Shortages. International Journal of Application or Innovation in Engineering & Management, Volume 2, Issue 3, 65-71.
- [61] Yadav, A.S. and Swami, A. (2013) A Two-Warehouse Inventory Model for Decaying Items with Exponential Demand and Variable Holding Cost. International of Inventive Engineering and Sciences, Volume-1, Issue-5, 18-22.
- [62] Yadav, A.S. and Kumar, S. (2017) Electronic Components Supply Chain Management for Warehouse with Environmental Collaboration & Neural Networks. International Journal of Pure and Applied Mathematics, Volume 117 No. 17, 169-177.
- [63] Yadav, A.S. (2017) Analysis of Seven Stages Supply Chain Management in Electronic Component Inventory Optimization for Warehouse with Economic Load Dispatch Using GA and PSO. Asian Journal Of Mathematics And Computer Research, volume 16 No.4, 208-219.

- [64] Yadav, A.S. (2017) Analysis Of Supply Chain Management In Inventory Optimization For Warehouse With Logistics Using Genetic Algorithm International Journal of Control Theory And Applications, Volume 10 No.10, 1-12.
- [65] Yadav, A.S. (2017) Modeling and Analysis of Supply Chain Inventory Model with two-warehouses and Economic Load Dispatch Problem Using Genetic Algorithm. International Journal of Engineering and Technology, Volume 9 No 1, 33-44.
- [66] Swami, A., Singh, S.R., Pareek, S. and Yadav, A.S. (2015) Inventory policies for deteriorating item with stock dependent demand and variable holding costs under permissible delay in payment. International Journal of Application or Innovation in Engineering & Management, Volume 4, Issue 2, 89-99.
- [67] Swami, A., Pareek, S., Singh S.R. and Yadav, A.S. (2015) Inventory Model for Decaying Items with Multivariate Demand and Variable Holding cost under the facility of Trade-Credit. International Journal of Computer Application, 18-28.
- [68] Swami, A., Pareek, S., Singh, S.R. and Yadav, A.S. (2015) An Inventory Model With Price Sensitive Demand, Variable Holding Cost And Trade-Credit Under Inflation. International Journal of Current Research, Volume 7, Issue, 06, 17312-17321.
- [69] Gupta, K., Yadav, A.S., Garg, A. and Swami, A. (2015) A Binary Multi-Objective Genetic Algorithm &PSO involving Supply Chain Inventory Optimization with Shortages, inflation. International Journal of Application or Innovation in Engineering & Management, Volume 4, Issue 8, 37-44.
- [70] Gupta, K., Yadav, A.S., Garg, A., (2015) Fuzzy-Genetic Algorithm based inventory model for shortages and inflation under hybrid & PSO. IOSR Journal of Computer Engineering, Volume 17, Issue 5, Ver. I, 61-67.
- [71] Singh, R.K., Yadav, A.S. and Swami, A. (2016) A Two-Warehouse Model for Deteriorating Items with Holding Cost under Particle Swarm Optimization. International Journal of Advanced Engineering, Management and Science, Volume -2, Issue-6, 858-864.
- [72] Singh, R.K., Yadav, A.S. and Swami, A. (2016) A Two-Warehouse Model for Deteriorating Items with Holding Cost under Inflation and Soft Computing Techniques. International Journal of Advanced Engineering, Management and Science, Volume -2, Issue-6, 869-876.
- [73] Kumar, S., Yadav, A.S., Ahlawat, N. and Swami, A. (2019) Supply Chain Management of Alcoholic Beverage Industry Warehouse with Permissible Delay in Payments using Particle Swarm Optimization. International Journal for Research in Applied Science and Engineering Technology, Volume 7 Issue VIII, 504-509.
- [74] Kumar, S., Yadav, A.S., Ahlawat, N. and Swami, A. (2019) Green Supply Chain Inventory System of Cement Industry for Warehouse with Inflation using Particle Swarm Optimization. International Journal for Research in Applied Science and Engineering Technology, Volume 7 Issue VIII, 498-503.

IJCR

- [75] Kumar, S., Yadav, A.S., Ahlawat, N. and Swami, A. (2019) Electronic Components Inventory Model for Deterioration Items with Distribution Centre using Genetic Algorithm. International Journal for Research in Applied Science and Engineering Technology, Volume 7 Issue VIII, 433-443.
- [76] Chauhan, N. and Yadav, A.S. (2020) An Inventory Model for Deteriorating Items with Two-Warehouse & Stock Dependent Demand using Genetic algorithm. International Journal of Advanced Science and Technology, Vol. 29, No. 5s, 1152-1162.
- [77] Chauhan, N. and Yadav, A.S. (2020) Inventory System of Automobile for Stock Dependent Demand & Inflation with Two-Distribution Center Using Genetic Algorithm. Test Engraining & Management, Volume 83, Issue: March April, 6583 6591.
- [78] Pandey, T., Yadav, A.S. and Medhavi Malik (2019) An Analysis Marble Industry Inventory Optimization Based on Genetic Algorithms and Particle swarm optimization. International Journal of Recent Technology and Engineering Volume-7, Issue-6S4, 369-373.
- [79] Ahlawat, N., Agarwal, S., Pandey, T., Yadav, A.S., Swami, A. (2020) White Wine Industry of Supply Chain Management for Warehouse using Neural Networks Test Engraining & Management, Volume 83, Issue: March April, 11259 11266.
- [80] Singh, S. Yadav, A.S. and Swami, A. (2016) An Optimal Ordering Policy For Non-Instantaneous Deteriorating Items With Conditionally Permissible Delay In Payment Under Two Storage Management International Journal of Computer Applications, Volume 147 –No.1, 16-25.