A Particle Swarm Optimization based a twostorage model for deteriorating items with Transportation Cost and Advertising Cost: The Auto Industry

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Abstract

A deterministic inventory model of Auto industry article deterioration using Particle Swarm Optimization (PSO) has been developed. He advocates a type of ramp with inflation effects in dual storage facilities of Auto industry. The own warehouse of Auto industry has a fixed capacity of W. units. The rented warehouse (RW) of Auto industry has unlimited capacity. We have assumed here that storage costs of Auto industry in RW were higher than in OW. Inventory of Auto industry bottlenecks are allowed and partially deferred using Particle Swarm Optimization (PSO) assumes that inventory of Auto industry will deteriorate over time with a variable rate of deterioration of Auto industry. The inflation effect has also been taken into account for various costs associated with the inventory system of Auto industry and the using Particle Swarm Optimization (PSO). The numerical example is also used to examine the behaviour of the model. The cost minimization technique is used to obtain the terms of total cost and other parameters.

Keywords:- Two-warehouses, deterministic inventory, Transportation Cost, Advertising Cost and Particle Swarm Optimization

1. Introduction

Basically, the inventory management and control system responds to the problems of demand and the supply chain of Auto industry. The business is entirely based on the demand and supply of goods, whether they are finished products or raw materials of Auto industry. In order to meet the demand of the consumer or supplier, it is necessary to interrogate the articles at all times and, for the same purpose, sufficient space is required to store the goods in order to meet the requirements of Auto industry. The room in which the goods are stored is called a warehouse of Auto industry. Traditional models assume that demand and ownership costs are constant and goods are delivered on demand, but over time many researchers believe that demand may vary with price and time of Auto industry. Other factors and holding costs may also vary over time and depending on other factors. Many models have been developed taking into account different time-dependent demands, with deficit and without defect of Auto industry. For all of these models, which take into account the variability of demand in response to inventories, it is assumed that the costs of ownership are constant throughout the inventory cycle of Auto industry. Review of bearing models often assumes unlimited storage capacity. However, in very busy markets such as supermarkets, supermarket markets, etc., storage space for items may be limited of Auto industry. Another case of insufficient storage space can occur when it is decided to purchase a large number of items. This may be because an attractive discount is available for bulk purchases or when the cost of purchasing goods is greater than other inventory costs, or when the demand for items is very high or if Article considered is An article deals with a seasonal product such as the yield of a crop or when problems arise with frequent supply of Auto industry. In this case, these items can not be housed in your existing warehouse of Auto industry. To store surplus items, an additional warehouse has been installed, possibly located near its own warehouse of Auto industry. First, it was 1 who discussed a rolling model with two storage systems of Auto industry. In order to

reduce storage costs, it is necessary to consume RW's assets at the earliest due to the higher holding costs of Auto industry.

Particle swarm optimization is initialized by a population of random solution and each potential solution is assigned a randomized velocity. The potential solutions called particles are then flown through the problem space. Each particle keeps track of its coordinates in the problem space which are associated with the best solution or fitness achieved so far the fitness value is also stored this value is called pbest. Another best value that is tracked by the global version of the PSO is the overall best value and its location obtained so far by any particle in the population. This value is termed gbest. Thus at each time step the particle change its velocity and moves towards its pbest and gbest this is the global version of PSO when in addition to pbest each particle keeps track of the best solution called nbest or lbest attained within a local topological neighbourhood of the particles the process is known as the local version of PSO.

2. Literature Review

Yaday and Swami (2018) analyzed a integrated supply chain model for deteriorating items with linear stock dependent demand under imprecise and inflationary environment. Yaday and Swami (2018) discuss a partial backlogging production-inventory lot-size model with time-varying holding cost and weibull deterioration. Yadav, et., al. (2018) presented a supply chain inventory model for decaying items with two ware-house and partial ordering under inflation. Yadav, et., al. (2018) proposed an inventory model for deteriorating items with two warehouses and variable holding cost. Yadav, et., al. (2018) analyzed a inventory of electronic components model for deteriorating items with warehousing using genetic algorithm. Yadav, et., al. (2018) discuss a analysis of green supply chain inventory management for warehouse with environmental collaboration and sustainability performance using genetic algorithm. Yadav and kumar (2017) presented a electronic components supply chain management for warehouse with environmental collaboration & neural networks. Yaday, et., al. (2017) analyzed a effect of inflation on a two-warehouse inventory model for deteriorating items with time varying demand and shortages. Yadav, et., al. (2017) discuss an inflationary inventory model for deteriorating items under two storage systems. Yadav, et., al. (2017) proposed a fuzzy based two-warehouse inventory model for non instantaneous deteriorating items with conditionally permissible delay in payment. Yadav (2017) analyzed a analysis of supply chain management in inventory optimization for warehouse with logistics using genetic algorithm. Yadav, et., al. (2017) discuss a supply chain inventory model for two warehouses with soft computing optimization. Yadav, et., al. (2016) presented a multi objective optimization for electronic component inventory model & deteriorating items with two-warehouse using genetic algorithm. Yadav (2017) analyzed a modeling and analysis of supply chain inventory model with two-warehouses and economic load dispatch problem using genetic algorithm. Yadav, et., al. 2018 discuss a particle swarm optimization for inventory of auto industry model for two warehouses with deteriorating items. Yaday, et., al. (2018) analyzed a hybrid techniques of genetic algorithm for inventory of auto industry model for deteriorating items with two warehouses. Yaday, et., al. (2018) discuss a supply chain management of pharmaceutical for deteriorating items using genetic algorithm. Yadav, et., al. (2018) analyzed a particle swarm optimization of inventory model with two-warehouses. Yadav, et., al. (2018) presented a supply chain management of chemical industry for deteriorating items with warehouse using genetic algorithm. Yadav (2017) discuss a analysis of seven stages supply chain management in electronic component inventory optimization for warehouse with economic load dispatch using ga and PSO. Yadav, et., al. (2017) gives a multi-objective genetic algorithm optimization in inventory model for deteriorating items with shortages using supply chain management. Yadav, et., al. (2017) analyzed a supply chain management in inventory optimization for deteriorating items with genetic algorithm. Yaday, et., al. (2017) discuss a modeling & analysis of supply chain management in inventory optimization for deteriorating items with genetic algorithm and particle swarm optimization. Yadav, et., al. (2017) presented a multi-objective particle swarm optimization and genetic algorithm in inventory model for deteriorating items with shortages using supply chain management. Yadav, et., al. (2017) proposed soft computing optimization of two warehouse inventory model with genetic algorithm. Yadav, et., al. (2017) analyzed a multi-objective genetic algorithm involving green supply chain management. Yadav, et., al. (2017) presented a multi-objective particle swarm optimization algorithm involving green supply chain inventory management. Yadav, et., al. (2017) gives a green supply chain management for warehouse with particle swarm optimization algorithm. Yadav, et., al. (2017) analyzed a analysis of seven stages supply chain management in electronic component inventory optimization for warehouse with economic load dispatch using genetic algorithm. Yadav, et., al. (2017) discuss a analysis of six stages supply chain management in inventory optimization for warehouse with artificial bee colony algorithm using genetic algorithm. Yaday, et., al. (2016) presented a analysis of electronic component inventory optimization in six stages supply chain management for warehouse with abc using genetic algorithm and PSO. Yadav, et., al. (2016) analyzed a two-warehouse inventory model for deteriorating items with variable holding cost, time-dependent demand and shortages. Yadav, et., al. (2016) discuss a two warehouse inventory

model with ramp type demand and partial backordering for weibull distribution deterioration. Yadav, et., al. (2016) proposed a two-storage model for deteriorating items with holding cost under inflation and genetic algorithms. Singh, et., al. (2016) analyzed a two-warehouse model for deteriorating items with holding cost under particle swarm optimization. Singh, et., al. (2016) presented a two-warehouse model for deteriorating items with holding cost under inflation and soft computing techniques. Sharma, et., al. (2016) gives an optimal ordering policy for non-instantaneous deteriorating items with conditionally permissible delay in payment under two storage management. Yadav, et., al. (2016) discuss a analysis of genetic algorithm and particle swarm optimization for warehouse with supply chain management in inventory control. Swami, et., al. (2015) analyzed an inventory policies for deteriorating item with stock dependent demand and variable holding costs under permissible delay in payment. Swami, et., al. (2015) presented an inventory model for decaying items with multivariate demand and variable holding cost under the facility of trade-credit. Swami, et., al. (2015) discuss an inventory model with price sensitive demand, variable holding cost and trade-credit under inflation. Gupta, et., al. (2015) proposed a binary multi-objective genetic algorithm &PSO involving supply chain inventory optimization with shortages, inflation. Yadav, et., al. (2015) analyzed a soft computing optimization based two ware-house inventory model for deteriorating items with shortages using genetic algorithm. Gupta, et., al. (2015) discuss a fuzzy-genetic algorithm based inventory model for shortages and inflation under hybrid & PSO. Yadav, et., al. (2015) presented a two warehouse inventory model for deteriorating items with shortages under genetic algorithm and PSO, taygi, et., al. (2015) analyzed an inventory model with partial backordering, weibull distribution deterioration under two level of storage. Yadav and Swami (2014) presented a twowarehouse inventory model for deteriorating items with ramp-type demand rate and inflation. Yadav and Swami (2013) discuss a effect of permissible delay on two-warehouse inventory model for deteriorating items with shortages. Yadav and Swami (2013) analyzed a two-warehouse inventory model for decaying items with exponential demand and variable holding cost. Yadav and Swami (2013) presented a partial backlogging twowarehouse inventory models for decaying items with inflation.

3. Assumption and notations

The mathematical model of two warehouse inventory model for deteriorating items is based on the following notation and assumptions

Notations:

O_c: Cost of ordering per Order of Auto industry φ: Capacity of OW of Auto industry.

T: The length of replenishment cycle of Auto industry.

M: Maximum inventory level per cycle to be ordered of Auto industry.

t₁. the time up to which product has no deterioration of Auto industry.

t₂: The time at which inventory level reaches to zero in RW of Auto industry.

t₃: The time at which inventory level reaches to zero in OW of Auto industry.

 H^{OW} : The holding cost per unit time in OW of Auto industry i.e. $H^{OW} = (ab+1)_1t$; where $(ab+1)_1$ is positive constant.

H^{RW}: The holding cost per unit time in RW of Auto industry i.e. H^{RW}=(ab+1)₂twhere (ab+1)₂>0 and H^{RW}>H^{OW}.

S_c: The shortages cost per unit per unit time of Auto industry.

 $I^{1RW}(t)$: The level of inventory in RW at time $[0 \ t_1]$ in which the product has no deterioration of Auto industry.

 $I^{2RW}(t)$: The level of inventory in RW at time $[t_1 \ t_2]$ in which the product has deterioration of Auto industry.

 $I^{10W}(t)$: The level of inventory in OW at time $[0 \ t_1]$ in which the product has no Deterioration of Auto industry.

 $I^{2OW}(t)$: The level of inventory in OW at time $[t_1 \ t_2]$ in which only Deterioration takes place of Auto industry.

 $I^{3OW}(t)$: The level of inventory in OW at time [t_2 t_3] in which Deterioration takes place of Auto industry.

Is(t): Determine the inventory level at time t in which the product has shortages of Auto industry.

 $(\gamma + \sigma + \phi)$: Deterioration rate in RW of Auto industry Such that $0 < (\gamma + \sigma + \phi) < 1$;

 $(\Phi + \sigma)$: Deterioration rate in OW of Auto industry such that $0 < (\Phi + \sigma) < 1$;

R_d: Deterioration cost per unit in RW of Auto industry.

O_d: Deterioration cost per unit in OW of Auto industry.

T_C: Cost of transportation per unit per cycle of Auto industry

r: (Discount rate – inflation i.e., r=d-i)

T^{IC}(t₂,t₃, T): The total relevant inventory cost per unit time of inventory system of Auto industry.

Assumption

- 1 Replenishment rate is infinite and lead time is negligible i.e. zero.
- 2 Holding cost is variable and is linear function of time of Auto industry.
- The time horizon of the inventory system is infinite of Auto industry.
- Goods of OW are consumed only after the consumption of goods kept in RW due to the more holding cost in RW than in OW of Auto industry.
- The OW has the limited capacity of storage and RW has unlimited capacity of Auto industry.
- Demand vary with time and is linear function of time and given by $D(t)=(\Psi\eta+\theta)t$; where $(\Psi\eta+\theta)>0$;
- For deteriorating items a fraction of on hand inventory deteriorates per unit time in both the warehouse with different rate of Deterioration of Auto industry.
- 8 Shortages are allowed and demand is fully backlogged at the beginning of next replenishment of Auto industry.
- The unit inventory cost (Holding cost of Auto industry + Deterioration cost of Auto industry) in RW>OW.

4. Mathematical formulation of model and analysis

In the beginning of the cycle at t=0 a lot size of M units of inventory of Auto industry enters into the system in which backlogged (M-R) units are cleared and the remaining units R is kept into two storage of Auto industry as W units in OW and RW units in RW.

$$\frac{dI^{1RW}(t)}{dt} = -(\Psi \eta + \theta) t \quad ; \qquad 0 \le t \le t_1$$
 (1)

$$\frac{\mathrm{d} I^{2RW}(t)}{\mathrm{d} t} = - \left(\gamma + \sigma + \phi \right) I^{2RW}(t) - \left(\Psi \eta + \theta \right) t \quad ; \qquad t_1 \leq t \leq t_2 \tag{2}$$

$$\frac{\mathrm{d}I^{1\mathbf{w}}(t)}{\mathrm{d}t} = 0 ; \qquad 0 \le t \le t_1$$
 (3)

$$\frac{\mathrm{d}I^{2\mathrm{w}}(t)}{\mathrm{d}t} = -\left(\Phi + \sigma\right) I^{2\mathrm{w}}(t) ; \qquad \qquad t_1 \leq t \leq t_2 \tag{4}$$

$$\frac{\mathrm{d}I^{3W}(t)}{\mathrm{d}t} = -(\Phi + \sigma) I^{3W}(t) - (\Psi \eta + \theta)t; \qquad t_2 \leq t \leq t_3$$
 (5)

$$\frac{\mathrm{d}I^{4S}(t)}{\mathrm{d}t} = -(\Psi \eta + \theta)t; \qquad t_3 \le t \le T$$
 (6)

Now inventory level at different time intervals is given by solving the above differential equations (1) to (6) with boundary conditions as follows:

At t=0, $I^{1RW}(t)=R-W$; therefore Differential eq. (1) gives

$$I^{1RW}(t) = R - W - \frac{(\Psi \eta + u)t^2}{2}$$
 ; $0 \le t \le t_1$ (7)

Differential eq. (2) is solved at $t=t_2$ and boundary condition $I^{2RW}(t_2)=0$, which yields

$$I^{2RW}(t) = \frac{(\Psi \eta + u)}{\alpha^2} \{ ((\gamma + \sigma + \phi)t_2 - 1)e^{(\gamma + \sigma + \phi)(t_2 - t)} - ((\gamma + \sigma + \phi)t - 1) \} ; t_1 \le t \le t_2$$
 (8)

Solution of differential eq. (3) with boundary condition at t=0 and I^{10W}(0)=W

$$I^{1OW}(t) = \varphi; \qquad 0 \le t \le t_1$$

Differential eq. (4) yields at $t=t_1$ and $I^{2OW}(t_1)=\varphi$

$$I^{2OW}(t) = \varphi e^{(\Phi + \sigma)(t_1 - t)} \qquad \qquad t_1 \le t \le t_2 \tag{10}$$

Solution of eq. (5) at $t = t_3$ and $I^{3OW}(t_3) = 0$ gives

$$I^{3OW}(t) = \frac{(\Psi \eta + \theta)}{(\Phi + \sigma)^2} \{ (\Phi t_3 - 1) e^{(\Phi + \sigma)(t_3 - t)} - ((\Phi + \sigma)t - 1) \} ; t_2 \leq t \leq t_3$$
(11)

Lastly the solution of eq. (6) at $t=t_3$ and $I^{4S}(t_3)=0$, is given as

$$I^{4S}(t) = \frac{(\Psi \eta + \theta)}{2} \{ t_3^2 - t^2 \}; \qquad t_3 \le t \le T$$
 (12)

Now considering the continuity of $I^{1R}(t_1) = I^{2R}(t_1)$, at $t=t_1$ from eq. (7) & (8) we have

$$R = \varphi + \frac{(\Psi \eta + \theta)t_1^2}{2} + \frac{(\Psi \eta + \theta)}{\alpha^2} \{ ((\gamma + \sigma + \phi)t_2 - 1)e^{(\gamma + \sigma + \phi)(t_2 - t_1)} - ((\gamma + \sigma + \phi)t_1 - 1) \};$$
 (13)

Substituting eq.(13) into eq. (7) we have

$$I^{1RW}(t) = \frac{b}{2}(t_1^2 - t^2) + \frac{b}{\alpha^2} \{ ((\gamma + \sigma + \phi)t_2 - 1)e^{(\gamma + \sigma + \phi)(t_2 - t_1)} - ((\gamma + \sigma + \phi)t_1 - 1) \} ;$$
 (14)

Cost of inventory shortages during time interval [t₃ T] is given by

$$IS = \int_{t_3}^{T} [-I_s(t)] dt$$

$$= -\frac{(a\eta + \theta)}{2} \int_{t_3}^T (t_3^2 - t^2) dt$$

$$= \frac{b}{\epsilon} \{ T^3 + 2t_3^3 - 3t_3^2 T \}$$
 $t_3 \le t \le T$ (15)

The maximum Inventory to be ordered is

M = R + IS

$$= \varphi + \frac{bt_1^2}{2} + \frac{b}{\alpha^2} \{ ((\gamma + \sigma + \phi)t_2 - 1)e^{(\gamma + \sigma + \phi)(t_2 - t_1)} - ((\gamma + \sigma + \phi)t_1 - 1) \} + \frac{b}{6} \{ T^3 - 2t_3^3 - 3t_3^2 T \};$$

$$(16)$$

Next the total relevant inventory cost per cycle consists of the following elements:

(I). Ordering Cost

$$O_{c} = \rho_{0} \tag{17}$$

(II). Transportation Cost

$$T_C = \rho_1 + 1$$

(III). Advertising Cost

$$A_C = \rho_2 + 1$$

(IV). Inventory holding cost in RW denoted by I^{HR} and is given as

$$\begin{split} &I^{\text{HRW}} = \int_{0}^{t_{1}} I^{1\text{RW}}(t) (ab + 1)_{2} t dt + \int_{t_{1}}^{t_{2}} I^{2\text{RW}}(t) (ab + 1)_{2} t dt \;] \\ &= \frac{bb_{2}}{8} t_{1}^{4} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{2(\gamma + \sigma + \phi)^{2}} \{ ((\gamma + \sigma + \phi)t_{2} - 1)e^{(\gamma + \sigma + \phi)(t_{2} - t_{1})} - ((\gamma + \sigma + \phi)t_{1} - 1) \} \; \; \} t_{1}^{2} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{(\gamma + \sigma + \phi)^{4}} (\gamma t_{2} - 1) \{ (e^{\gamma(t_{2} - t_{1})} - 1) - (\gamma + \sigma + \phi)(t_{2} - t_{1}e^{\gamma(t_{2} - t_{1})}) \} - \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \; \{ \; 2(\gamma + \sigma + \phi)(t_{2}^{3} - t_{1}^{3}) - 3(t_{2}^{2} - t_{1}^{2}) \} \end{split}$$

(V). Inventory holding cost in OW denoted by I^{HOW} and is given by

$$\begin{split} \mathrm{I}^{\mathrm{HOW}} &= \int_{0}^{t_{1}} \mathrm{I}^{10\mathrm{W}}(\mathsf{t}) (\mathsf{ab} + 1)_{1} \mathsf{t} \, \mathsf{dt} + \int_{t_{1}}^{t_{2}} \mathrm{I}^{20\mathrm{W}}(\mathsf{t}) \mathsf{dt} + \int_{t_{2}}^{t_{3}} \mathrm{I}^{30\mathrm{W}}(\mathsf{t}) (\mathsf{ab} + 1)_{1} \mathsf{t} \mathsf{dt} \\ &= \frac{\varphi(\mathsf{ab} + 1)_{1} t_{1}^{2}}{2} + \frac{(\mathsf{ab} + 1)_{1} \varphi}{(\Phi + \sigma)^{2}} \{ (1 - \mathrm{e}^{-(\Phi + \sigma)(t_{2} - t_{1})}) - (\Phi + \sigma)(t_{2} \mathrm{e}^{-(\Phi + \sigma)(t_{2} - t_{1})} - t_{1}) \} \\ &+ \frac{b b_{1}}{\Phi^{4}} ((\Phi + \sigma) t_{3} - 1) \{ (\mathrm{e}^{(\Phi + \sigma)(t_{3} - t_{2})} - 1) - (\Phi + \sigma)(t_{3} - t_{2} \mathrm{e}^{(\Phi + \sigma)(t_{3} - t_{2})}) \} \\ &- \frac{(\Psi \eta + \theta)(\mathsf{ab} + 1)_{1}}{6\Phi^{2}} \{ 2(\Phi + \sigma)(t_{3}^{3} - t_{2}^{3}) - 3(t_{3}^{2} - t_{2}^{2}) \} \end{split}$$

(VI). Cost of inventory deteriorated in RW is denoted and given by

$$\begin{split} &I^{DRW} = (R-W) - \int_{t_1}^{t_2} (a\eta + \theta)t \, dt \\ &= & \frac{b}{\alpha^2} \{ ((\gamma + \sigma + \phi)t_2 - 1)e^{(\gamma + \sigma + \phi)(t_2 - t_1)} - ((\gamma + \sigma + \phi)t_1 - 1) \} - \frac{(ab+1)}{2} (t_2^2 - 2t_1^2) \end{split}$$

(VII). Cost of deteriorated inventory in RW is given by

$$CI^{DRW} = D_R \left\{ \frac{b}{\alpha^2} \left\{ ((\gamma + \sigma + \phi)t_2 - 1)e^{(\gamma + \sigma + \phi)(t_2 - t_1)} - ((\gamma + \sigma + \phi)t_1 - 1) \right\} - \frac{(\Psi \eta + \theta)}{2} (t_2^2 - 2t_1^2) \right\}$$
(20)

(VIII). Cost of inventory deteriorated in OW is denoted and given by

$$I^{DOW} = \varphi - \int_{t_2}^{t_3} (\Psi \eta + \theta) t dt$$

$$= \varphi - \frac{b}{2}(t_3^2 - t_2^2)$$

(IX). Cost of deteriorated inventory in OW is given by

$$CI^{DOW} = O_d \left\{ \varphi - \frac{(\Psi \eta + \theta)}{2} (t_3^2 - t_2^2) \right\}$$
 (21)

(X). Shortages Cost

$$CIS = S_{c} \left[\frac{(\Psi \eta + \theta)}{6} \left\{ T^{3} + 2t_{3}^{3} - 3t_{3}^{2} T \right\} \right]$$
 (22)

 $T^{IC}(t_2, t_3 T) = \frac{1}{T}$ [Ordering cost + Transportation Cost + Advertising Cost + Inventory holding cost per cycle in RW + Inventory holding cost per cycle in OW + Deterioration cost per cycle in RW+ Deterioration cost per cycle in OW + Shortage cost]

$$T^{IC}(t_{2}, t_{3}, T) = \frac{1}{T}[O_{c} + T_{C} + A_{C} + I^{HRW} + I^{HOW} + CI^{DRW} + CI^{DOW} + CIS]$$
(23)

Substituting equations (17) to (22) in equation (23) we get

$$\begin{split} \mathbf{T}^{\mathrm{IC}}(\mathbf{t}_{2},\ \mathbf{t}_{3},\ \mathbf{T}) &= \frac{1}{\mathrm{T}} [(\rho_{0} + (\rho_{1} + 1) + (\rho_{2} + 1) + \frac{bb_{2}}{8} t_{1}^{4} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{2(\gamma + \sigma + \phi)^{2}} \{ ((\gamma + \sigma + \phi)t_{2} - 1)e^{(\gamma + \sigma + \phi)(t_{2} - t_{1})} - ((\gamma + \sigma + \phi)t_{1} - 1) \} t_{1}^{2} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{(\gamma + \sigma + \phi)^{4}} (\alpha t_{2} - 1) \{ (e^{\alpha(t_{2} - t_{1})} - 1) - (\gamma + \sigma + \phi)(t_{2} - t_{1}e^{\alpha(t_{2} - t_{1})}) \} - \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma + \phi)^{2}} \{ 2(\gamma + \sigma + \phi)(t_{2} - t_{1}) \} + \frac{(\Psi \eta + \theta)(ab + 1)_{2}}{6(\gamma + \sigma$$

The total relevant inventory cost is minimum if

$$\frac{\partial T^{IC}}{\partial t_2} = 0 \quad ; \quad \frac{\partial T^{IC}}{\partial t_3} = 0 \quad ; \quad \frac{\partial T^{IC}}{\partial T} = 0 \tag{25}$$

7. Inventory Analysis using Particle Swarm Optimization Algorithm

2:
$$\{M_x, N_x, U_x, V_x\}_{x=1}^X := initialize()$$

4: for
$$b := 1: X$$

5: for
$$r := 1 : R$$

6:
$$n_{xc}^{(a+1)} = y n_{xc}^a + c_1 d_1 [V_{xc} - m_{xc}^a] + c_2 d_2 [U_{xc} - m_{xc}^a]$$

7:
$$M_r^{a+1} = M_r^a + mN_r^a + \epsilon^a$$

8: end

9: M_x := enforce Constraints(X)

10: $Y_x := f(M_x)$

11: if $M_x \le e \forall e \in P$

12: P:= {e \in P/ e $\not<$ M_r }

13: P:= $P \cup M_x$

14: end

15: end

16: if $M_x \leq V_x \vee (XM_x \neq V_x \wedge V_x \neq M_x)$

17: $V_x := M_x$

18: end

19: $U_x := \text{selectGuide}(X, A)$

20: end

6. Numerical Example:

In order to illustrate the above solution procedure, consider an inventory system with the following data in appropriate units: ρ_0 =50, (ρ_2+1) = 100, (ρ_2+1) = 150, ϕ =300, $(\Psi\eta+\theta)$ =26, $(ab+1)_1$ =5.7, $(ab+1)_2$ =5.4, t_1 =1.3, $(\gamma+\sigma+\phi)$ =1.25, $(\Phi+\sigma)$ =2.52, C_s =6.3, and C_t =7.3 The vales of decision variables are competed for the model and also for the models of special cases.

Population=45, Generations=55, Cognitive learning factor=34, Cooperative factor=33, Social learning factor=22.25, Inertial constant=23.15 and number of neighbors=26.

7. Sensitivity Analysis

Table:-1 Best results of purpose function by different Product's Maximum inventory Level for Maximum demand

Total Cost	Product's Maximum inventory Level							Maximum
	Ordering	Transportation	Advertising	holding	holding	Deterioration	Shortage	
	cost	Cost	Cost	cost OW	cost RW	cost	cost	
		1	7					
$\mathbf{T}^{\mathrm{IC}}\left(\mathbf{t}_{2},\mathbf{t}_{3}\;\mathbf{T}\right)$	378	587	25.4	258.4	259.6	25.4	30.5	3001.25
		460			down -	1000		

The optimization of inventory control in supply chain management based on Particle Swarm Optimization (PSO). Particle Swarm Optimization (PSO) is analyzed with the help of MATLAB. The stock levels for the three different members of the supply chain, Ordering cost, Transportation Cost, Advertising Cost, Inventory holding cost per cycle in RW, Inventory holding cost per cycle in OW, Deterioration cost per cycle in RW, Deterioration cost per cycle in OW, Shortage cost are generated using the MATLAB script and this generated data set is used for evaluating the performance of the Particle Swarm Optimization (PSO). Some sample set of data used in the implementation is given in table 2. Some 5 sets of data are given in the table 1 and these are assumed as the records of the past period.

Table:-1 Particle Swarm Optimization (PSO) model optimal solution

Р	WW	PSO					
	OPT	BEST	MAX	AVG	STD		
1	15.50	15.50	11.75	18.00	1.5		

2	17.00	17.00	15.75	18.75	1.9
3	21.00	18.00	11.00	11.00	1.0
4	39.00	19.00	19.00	19.00	2.0
5	41.25	15.25	12.50	12.70	1.8

8. Conclusion

In this article, we proposed a deterministic inventory model of Auto industry with two stocks of Auto industry for non-immediate deterioration of constant demand items of Auto industry using Particle Swarm Optimization (PSO). It is assumed that articles are shipped from RW of Auto industry to retail stores in a streaming pattern to minimize the total costing function of the stock model using Particle Swarm Optimization (PSO). We find that all relevant storage costs are affected by the selling price and the interest rate earned of Auto industry. All relevant inventory costs will be minimized if the time allowed is longer than the order cycle or if the retailer pays their total purchase cost at the end of the specified period of Auto industry using Particle Swarm Optimization (PSO). In addition, the proposed model can be used to control stocks of certain trade-damaged items and can be expanded by incorporating a time-dependent requirement, a likely demand model, variable holding costs, and so on of Auto industry using Particle Swarm Optimization (PSO).

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