



# **An EOQ Model for Weibull Deteriorating Items with Linear Demand and Partial Backlogging in Fuzzy Environment**

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**Abstract ---** *In this paper, we developed an EOQ model for Weibull deteriorating items with linear demand rate in fuzzy environment. Shortages are allowed and partially backlogged. Holding cost, deterioration cost, ordering cost, shortage cost and opportunity cost are assumed as a triangular fuzzy numbers. The purpose of this paper is to minimize the total cost function in fuzzy environment. Graded mean representation, signed distance and centroid methods are used to defuzzify the total cost function and the results obtained by these methods are compared with the help of a numerical example. Sensitivity analysis is also carried out to detect the most sensitive parameters of the system.*

**Keywords---** *EOQ model, Two parameter Weibull deteriorating items, Shortages, Triangular fuzzy number, Graded mean representation method, Signed distance method, Centroid method.*

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## **I. INTRODUCTION**

In the traditional inventory models, one of the assumptions was that the items preserved their physical characteristics while they were kept stored in the inventory. This assumption is evidently true for most items, but not for all. However, the deteriorating items are subject to a continuous loss in their masses or utility throughout their lifetime due to decay, damage, spoilage, and penalty of other reasons. Owing to this fact, controlling and maintaining the inventory of deteriorating items becomes a challenging problem for decision makers. Harris (1915) [1] developed the first inventory model, Economic Order Quantity, which was generalized by Wilson (1934) [2] who gave a formula to obtain economic order quantity. Whitin (1957) [3] considered the deterioration of the fashion goods at the end of the prescribed shortage period. Ghare and Schrader (1963) [4] developed a model for an exponentially decaying inventory. Dave and Patel (1981) [5] were the first to study a deteriorating inventory with linear increasing demand when shortages are not allowed.

Some of the recent work in this field has been done by Chung and Ting (1993) [6]; Wee (1995) [7] studied an inventory model with deteriorating items. Chang and Dye (1999) [8] developed an inventory model with time-varying demand and partial backlogging. Skouri et al. (2009) [9] developed an inventory model with ramp-type demand rate, partial backlogging, and Weibull's deterioration rate. Mishra, V.K. and Singh, L.S. (2010) [10] developed an deteriorating inventory model with time dependent demand and partial backlogging. Covert and Philip [11] extended Ghare and Schrader's constant deterioration rate to a two-parameter Weibull distribution. In 2012, C. K. Jaggi, S. Pareek, A. Sharma and Nidhi [12] presented a fuzzy inventory model for deteriorating items with time-varying demand and shortages. C.K. Tripathy and U. Mishra [2010] [13] they developed an inventory model for Weibull deteriorating items with price dependent demand and time varying holding cost. Bhunia and Maiti (1994) [14] developed an a two warehouse inventory model for a linear trend in demand,

Bhunia and Maiti (1998) [15] developed an a two warehouse inventory model for deteriorating items with a linear demand and shortages. Goswami and Chandhuri (1991) [16] developed an EOQ model for deteriorating items with a linear trend in demand.

In conventional inventory models, uncertainties are treated as randomness and are being handled by applying the probability theory. However, in certain situations uncertainties are due to fuzziness, and such cases are dilated in the fuzzy set theory which was demonstrated by Zadeh in [17]. Kauffmann and Gupta [18] provided an introduction to fuzzy arithmetic operation and Zimmermann [19] discussed the concept of the fuzzy set theory and its applications. Considering the fuzzy set theory in inventory modelling renders an authenticity to the model formulated since fuzziness is the closest possible approach to reality. As reality is imprecise and can only be approximated to a certain extent, same way, fuzzy theory helps one to incorporate uncertainties in the formulation of the model, thus bringing it closer to reality. Park [20] applied the fuzzy set concepts to EOQ formula by representing the inventory carrying cost with a fuzzy number and solved the economic order quantity model using fuzzy number operations based on the extension principle. Vujosevic et al. [21] used trapezoidal fuzzy number to fuzzify the order cost in the total cost of the inventory model without backorder, and got fuzzy total cost. Yao and Lee [22] introduced a backorder inventory model with fuzzy order quantity as triangular and trapezoidal fuzzy numbers and shortage cost as a crisp parameter.

Chang [23] discussed the fuzzy production inventory model for fuzzify the product quantity as triangular fuzzy number. Yao and Chiang [24] considered the total cost of inventory without backorder. They fuzzified the total demand and cost of storing one unit per day into triangular fuzzy numbers and defuzzify by the centroid and the signed distance methods. Gani and Maheswari[25] developed an EOQ model with imperfect quality items with shortages where defective rate, demand, holding cost, ordering cost and shortage cost are taken as triangular fuzzy numbers. Graded mean integration method is used for defuzzification of the total profit. Uthayakumar and Valliathal [26] developed an economic production model for Weibull deteriorating items over an infinite horizon under fuzzy environment and considered some cost component as triangular fuzzy numbers and using the signed distance method to defuzzify the cost function.

In this paper, an inventory model for Weibull deteriorating items and linear demand with shortages is considered where , ordering cost, holding cost, deterioration rate, , shortage cost and opportunity cost are assumed as a triangular fuzzy numbers. For defuzzification of the total cost function, Graded Mean Representation, Signed distance and Centroid methods are used. By comparing the results obtained by these methods, we get the better one as an estimate of the total cost in the fuzzy sense.

## II. FUZZY PRELIMINARIES

In order to treat fuzzy inventory model by using graded mean representation, signed distance and centroid to defuzzify , we need the following definitions.

**Definition 2.1** A fuzzy set  $\tilde{a}$  on  $R = (-\infty, \infty)$  is called a fuzzy point if its membership function is

$$\mu_{\tilde{a}}(x) = \begin{cases} 1, & x = a \\ 0, & x \neq a \end{cases} \quad \text{----- (1)}$$

where the point  $a$  is called the support of fuzzy set  $\tilde{a}$

**Definition 2.2** A fuzzy set  $[a_\alpha, b_\alpha]$  where  $0 \leq \alpha \leq 1$  and  $a < b$  defined on  $R$ , is called a level of a fuzzy interval if its membership function is

$$\mu_{[a_\alpha, b_\alpha]}(x) = \begin{cases} \alpha, & a \leq x \leq b \\ 0, & \text{otherwise} \end{cases} \quad \text{----- (2)}$$

**Definition 2.3** A fuzzy number  $\tilde{A} = (a, b, c)$  where  $a < b < c$  and defined on  $R$ , is called a triangular fuzzy number if its membership function is

$$\mu_A = \begin{cases} \frac{x-a}{b-a}, & a \leq x \leq b \\ \frac{c-x}{c-b}, & b \leq x \leq c \\ 0, & \text{otherwise} \end{cases} \quad \text{----- (3)}$$

When  $a = b = c$ , we have fuzzy point  $(c, c, c) = \tilde{c}$ . The family of all triangular fuzzy numbers on  $R$  is denoted as  $F_N = \{(a, b, c) / a < b < c \forall a, b, c \in R\}$ .

The  $\alpha$ -cut of  $\tilde{A} = (a, b, c) \in F_N, 0 \leq \alpha \leq 1$ , is  $A(\alpha) = [A_L(\alpha), A_R(\alpha)]$  Where  $A_L(\alpha) = a + (b-a)\alpha$  and  $A_R(\alpha) = c - (c-b)\alpha$  are the left and right endpoints of  $A(\alpha)$ .

**Definition 2.4** If  $\tilde{A} = (a, b, c)$  is a triangular fuzzy number then the graded mean integration representation of  $\tilde{A}$  is defined as

$$P(\tilde{A}) = \frac{\int_0^{w_A} h \left( \frac{L^{-1}(h) + R^{-1}(h)}{2} \right) dh}{\int_0^{w_A} h dh} \quad \text{With } 0 < h \leq w_A \text{ and } 0 < w_A \leq 1$$

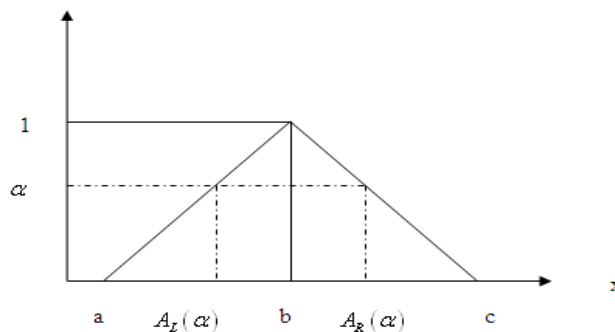
$$P(\tilde{A}) = \frac{1}{2} \frac{\int_0^1 h [a + h(b-a) + c - h(c-a)] dh}{\int_0^1 h dh} = \frac{a + 4b + c}{6} \quad \text{----- (4)}$$

**Definition 2.5** If  $\tilde{A} = (a, b, c)$  is a triangular fuzzy then the signed distance of  $\tilde{A}$  is defined as

$$d(\tilde{A}, \tilde{0}) = \int_0^1 d([A_L(\alpha)_\alpha, A_R(\alpha)_\alpha], \tilde{0}) = \frac{1}{4} (a + 2b + c) \quad \text{----- (5)}$$

**Definition 2.6** The centroid method on the triangular fuzzy number  $\tilde{A} = (a, b, c)$  is defined as

$$C(\tilde{A}) = \frac{a + b + c}{3} \quad \text{----- (6)}$$



**Figure 1.**  $\alpha$ -cut of a triangular fuzzy number

### III. NOTATIONS AND ASSUMPTIONS

The proposed inventory model having following notations and assumptions:

#### 3.1 Notations

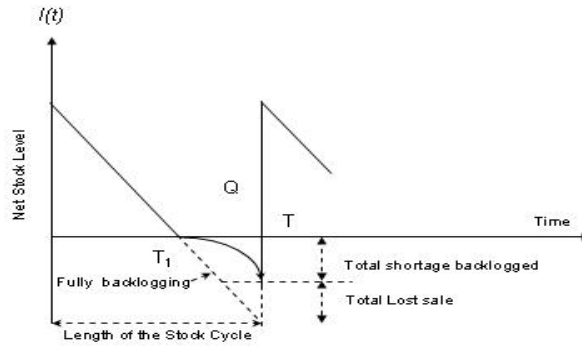
- $I(t)$  : the inventory level at time  $t$ .
- $W$  : the maximum inventory level for each ordering cycle.
- $IB$  : the maximum amount of demand backlogged for each ordering cycle.
- $Q$  : the economic order quantity for each ordering cycle.
- $T$  : length of each ordering cycle.
- $C_1$  : deterioration cost, \$/per unit.
- $C_2$  : shortage Cost, \$/per unit /per unit time.
- $C_3$  : opportunity cost,\$/ per unit /per unit time.
- $h(t)$  : holding cost, \$/per unit/ per unit time.
- $A$  : ordering cost of inventory, \$/ per order
- $\tilde{C}_1$  : fuzzy deterioration cost, \$/per unit.
- $\tilde{C}_2$  : fuzzy shortage cost, \$/per unit/ per unit time.
- $\tilde{C}_3$  : fuzzy opportunity cost,\$/ per unit /per unit time.
- $\tilde{h}$  : fuzzy holding cost, \$/per unit/ per unit time.
- $\tilde{A}$  : fuzzy ordering cost of inventory, \$/ per order.
- $TC(T_1, T)$  : total inventory cost per unit time.
- $T\tilde{C}(T_1, T)$  : total fuzzy inventory cost per unit time.
- $TC_{dG}(T_1, T)$  : defuzzify value of  $T\tilde{C}(T_1, T)$  by applying Graded mean integration method
- $TC_{dS}(T_1, T)$  : defuzzify value of  $T\tilde{C}(T_1, T)$  by applying Signed distance method
- $TC_{dC}(T_1, T)$  : defuzzify value of  $T\tilde{C}(T_1, T)$  by applying Centroid method

#### 3.2 Assumptions:

- (i) The inventory system involves only one item and the planning horizon is infinite.
- (ii) Replenishment occurs instantaneously at an infinite rate.
- (iii) The demand rate,  $D(t) = \begin{cases} a + bt, & \text{when } I(t) > 0 \\ D_0, & \text{when } I(t) \leq 0 \end{cases}$   
 where  $a > 0$ ,  $b > 0$  and  $a$  is initial demand.
- (iv) The deterioration of time as follows by Weibull parameters (two) distribution  $\theta(t) = \alpha\beta t^{\beta-1}$ ,  
 where  $0 < \alpha < 1$  is the scale parameter and  $\beta > 0$  is the shape parameter.
- (v) During the shortage period, the backlogging rate is variable and is dependent on the length of the waiting time for the next replenishment. The longer the waiting time is, the smaller the backlogging rate would be. Hence, the proportion of customers who would like to accept backlogging at time  $t$  is decreasing with the waiting time  $T - t$  waiting for the next replenishment. To take care of this situation we have defined the backlogging rate to be  $\frac{1}{1 + \delta(T - t)}$  when inventory is negative. The backlogging parameter  $\delta$  is a positive constant,  $T_1 \leq t \leq T$ .

**IV. MATHEMATICAL MODEL**

**4.1 Crisp Model:**



**Figure 2: Graphical representation of the inventory system**

We consider the deteriorating inventory model with linear demand. Replenishment occurs at time  $t=0$  when the inventory level attains its maximum  $W$ . From  $t=0$  to  $T_1$ , the inventory level reduces due to demand and deterioration. At time  $T_1$  the inventory level achieves zero, then shortage is allowed to occur during the time interval  $[T_1, T]$  and all of the demand during shortage period  $[T_1, T]$  is partially backlogged.

As the inventory level reduces due to demand rate as well as deterioration during the inventory interval  $[T_1, T]$ , the differential equation representing the inventory status is governed by

$$\frac{dI(t)}{dt} + \theta(t)I(t) = -D(t) \quad ; 0 \leq t \leq T_1$$

Where  $\theta(t) = \alpha\beta t^{\beta-1}$  and  $D(t) = a + bt$

$$\frac{dI(t)}{dt} + \alpha\beta t^{\beta-1} I(t) = -(a + bt) \quad ; 0 \leq t \leq T_1 \quad \text{----- (7)}$$

During the shortage interval  $[T_1, T]$ , the demand at time  $t$  is partially backlogged at the fraction  $\frac{1}{1 + \delta(T-t)}$ . Therefore, the differential equation governing the amount of demand backlogged is

$$\frac{dI(t)}{dt} = -\frac{D_0}{1 + \delta(T-t)} \quad ; T_1 \leq t \leq T \quad \text{----- (8)}$$

where  $D(t) = D_0$  with the boundary condition  $I(T_1) = 0$  and  $I(0) = W$ . The solution of equation (7) and (8) is given by

$$I(t) = \left\{ \begin{array}{l} a(T_1 - t) + \frac{b}{2}(T_1^2 - t^2) + \frac{a\alpha(T_1^{\beta+1} - t^{\beta+1})}{\beta + 1} + \frac{b\alpha(T_1^{\beta+2} - t^{\beta+2})}{\beta + 2} \\ -a\alpha(T_1 t^\beta - t^{\beta+1}) - \frac{b\alpha}{2}(T_1^2 t^\beta - t^{\beta+2}) - \frac{a\alpha^2}{\beta + 1}(T_1^{\beta+1} t^\beta - t^{2\beta+1}) \\ -\frac{b\alpha^2}{\beta + 2}(T_1^{\beta+2} t^\beta - t^{2\beta+2}) \end{array} \right\} ; 0 \leq t \leq T_1 \quad \text{----- (9)}$$

$$I(t) = \frac{D_0}{\delta} \log[1 + \delta(T - t)] - \frac{D_0}{\delta} \log[1 + \delta(T - T_1)] \quad ; T_1 \leq t \leq T \quad \text{----- (10)}$$

Maximum amount of demand backlogged per cycle is obtained by putting in  $t = T$  in Equation (10). Therefore,

$$IB = -I(T) = \frac{D_0}{\delta} \log[1 + \delta(T - T_1)] \quad \text{----- (11)}$$

Maximum inventory level for each cycle is obtained by putting the boundary condition  $I(0) = W$  in equation (9). Therefore,

$$W = aT_1 + \frac{b}{2}T_1^2 + \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} \quad \text{----- (12)}$$

Hence, the economic order quantity per cycle is

$$Q = W + IB = aT_1 + \frac{b}{2}T_1^2 + \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} + \frac{D_0}{\delta} \log[1 + \delta(T - T_1)] \quad \text{----- (13)}$$

The total cost per cycle consists of following cost components

- (i) Inventory holding cost over the cycle is given by

$$\begin{aligned}
 HC &= \int_0^{T_1} h(t)I(t)dt \\
 &= h \left\{ \begin{aligned} &\frac{aT_1^2}{2} + \frac{bT_1^3}{3} + \frac{a\alpha\beta T_1^{\beta+2}}{(\beta+1)(\beta+2)} + \frac{b\alpha\beta T_1^{\beta+3}}{(\beta+1)(\beta+3)} \\ &-\frac{a\alpha^2 T_1^{2\beta+2}}{(\beta+1)(2\beta+2)} - \frac{b\alpha^2 T_1^{2\beta+3}}{(\beta+1)(2\beta+3)} \end{aligned} \right\} \quad \text{----- (14)}
 \end{aligned}$$

- (ii) Deterioration cost over the cycle is given by

$$\begin{aligned}
 DC &= C_1 \left[ W - \int_0^{T_1} D(t)dt \right] \\
 &= C_1 \left\{ \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} \right\} \quad \text{----- (15)}
 \end{aligned}$$

- (iii) Shortage cost over the cycle is given by

$$\begin{aligned}
 SC &= -C_2 \int_{T_1}^T I(t)dt \\
 &= \frac{-C_2 D_0}{\delta} \int_{T_1}^T \{ \log[1 + \delta(T - t)] - \log[1 + \delta(T - T_1)] \} dt \\
 &= C_2 D_0 \left\{ \frac{T - T_1}{\delta} - \frac{1}{\delta^2} \log[1 + \delta(T - T_1)] \right\} \quad \text{----- (16)}
 \end{aligned}$$

(iv) The opportunity cost due to lost sales per cycle is

$$\begin{aligned}
 OC &= C_3 \int_{T_1}^T \left\{ D_0 \left[ 1 - \frac{1}{1 + \delta(T-t)} \right] \right\} dt \\
 &= C_3 D_0 \left\{ T - T_1 - \frac{1}{\delta} \log[1 + \delta(T-t)] \right\} \quad \text{----- (17)}
 \end{aligned}$$

(v) Ordering cost per cycle

$$OC = A \quad \text{----- (18)}$$

Total cost of the system per unit time is given by

$$\begin{aligned}
 TC(T_1, T) &= \frac{1}{T} \{ HC + DC + SC + OC + A \} \\
 &= \left\{ \begin{aligned} &\frac{h}{T} \left\{ \frac{aT_1^2}{2} + \frac{bT_1^3}{3} + \frac{a\alpha\beta T_1^{\beta+2}}{(\beta+1)(\beta+2)} + \frac{b\alpha\beta T_1^{\beta+3}}{(\beta+1)(\beta+3)} - \frac{a\alpha^2 T_1^{2\beta+2}}{(\beta+1)(2\beta+2)} - \frac{b\alpha^2 T_1^{2\beta+3}}{(\beta+1)(2\beta+3)} \right\} \\ &+ \frac{C_1}{T} \left\{ \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} \right\} + \frac{C_2 D_0}{T} \left\{ \frac{T-T_1}{\delta} - \frac{1}{\delta^2} \log[1 + \delta(T-T_1)] \right\} \\ &+ \frac{C_3 D_0}{T} \left\{ T - T_1 - \frac{1}{\delta} \log[1 + \delta(T-t)] \right\} + \frac{A}{T} \end{aligned} \right\} \quad \text{-----(19)}
 \end{aligned}$$

**4.2 Fuzzy model:**

Due to uncertainty in the environment it is not easy to define all the parameters precisely, accordingly we assume some of these parameters namely  $\tilde{h}, \tilde{C}_1, \tilde{C}_2, \tilde{C}_3, \tilde{A}$  may change within some limits. Let  $\tilde{h} = (h_1, h_2, h_3); \tilde{C}_1 = (C_{11}, C_{12}, C_{13}); \tilde{C}_2 = (C_{21}, C_{22}, C_{23}); \tilde{C}_3 = (C_{31}, C_{32}, C_{33}); \tilde{A} = (A_1, A_2, A_3)$  are as triangular fuzzy numbers.

Total cost of the system per unit time in fuzzy sense is given by

$$\begin{aligned}
 T\tilde{C}(T_1, T) &= \left\{ \begin{aligned} &\frac{\tilde{h}}{T} \left\{ \frac{aT_1^2}{2} + \frac{bT_1^3}{3} + \frac{a\alpha\beta T_1^{\beta+2}}{(\beta+1)(\beta+2)} + \frac{b\alpha\beta T_1^{\beta+3}}{(\beta+1)(\beta+3)} - \frac{a\alpha^2 T_1^{2\beta+2}}{(\beta+1)(2\beta+2)} - \frac{b\alpha^2 T_1^{2\beta+3}}{(\beta+1)(2\beta+3)} \right\} \\ &+ \tilde{C}_2 \left\{ \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} \right\} + \tilde{C}_1 D_0 \left\{ \frac{T-T_1}{\delta} - \frac{1}{\delta^2} \log[1 + \delta(T-T_1)] \right\} \\ &+ \tilde{C}_3 D_0 \left\{ T - T_1 - \frac{1}{\delta} \log[1 + \delta(T-t)] \right\} + \frac{\tilde{A}}{T} \end{aligned} \right\} \quad \text{----- (20)}
 \end{aligned}$$

We defuzzify the fuzzy total cost  $T\tilde{C}(T_1, T)$  by Graded mean representation, Signed distance and Centroid methods.

(i) By Graded Mean Representation Method, Total cost is given by

$$TC_{dG}(T_1, T) = \frac{1}{6} [TC_{dG_1}(T_1, T) + 4TC_{dG_2}(T_1, T) + TC_{dG_3}(T_1, T)]$$

$$TC_{dG_1}(T_1, T) = \left\{ \begin{aligned} & \frac{h_1}{T} \left\{ \frac{aT_1^2}{2} + \frac{bT_1^3}{3} + \frac{a\alpha\beta T_1^{\beta+2}}{(\beta+1)(\beta+2)} + \frac{b\alpha\beta T_1^{\beta+3}}{(\beta+1)(\beta+3)} - \frac{a\alpha^2 T_1^{2\beta+2}}{(\beta+1)(2\beta+2)} - \frac{b\alpha^2 T_1^{2\beta+3}}{(\beta+1)(2\beta+3)} \right\} \\ & + C_{21} \left\{ \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} \right\} + C_{11} D_0 \left\{ \frac{T-T_1}{\delta} - \frac{1}{\delta^2} \log[1 + \delta(T-T_1)] \right\} \\ & + C_{31} D_0 \left\{ T-T_1 - \frac{1}{\delta} \log[1 + \delta(T-t)] \right\} + \frac{A_1}{T} \end{aligned} \right\} \quad \text{-----(21)}$$

$$TC_{dG_2}(T_1, T) = \left\{ \begin{aligned} & \frac{h_2}{T} \left\{ \frac{aT_1^2}{2} + \frac{bT_1^3}{3} + \frac{a\alpha\beta T_1^{\beta+2}}{(\beta+1)(\beta+2)} + \frac{b\alpha\beta T_1^{\beta+3}}{(\beta+1)(\beta+3)} - \frac{a\alpha^2 T_1^{2\beta+2}}{(\beta+1)(2\beta+2)} - \frac{b\alpha^2 T_1^{2\beta+3}}{(\beta+1)(2\beta+3)} \right\} \\ & + C_{22} \left\{ \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} \right\} + C_{12} D_0 \left\{ \frac{T-T_1}{\delta} - \frac{1}{\delta^2} \log[1 + \delta(T-T_1)] \right\} \\ & + C_{32} D_0 \left\{ T-T_1 - \frac{1}{\delta} \log[1 + \delta(T-t)] \right\} + \frac{A_2}{T} \end{aligned} \right\} \quad \text{-----(22)}$$

$$TC_{dG_3}(T_1, T) = \left\{ \begin{aligned} & \frac{h_3}{T} \left\{ \frac{aT_1^2}{2} + \frac{bT_1^3}{3} + \frac{a\alpha\beta T_1^{\beta+2}}{(\beta+1)(\beta+2)} + \frac{b\alpha\beta T_1^{\beta+3}}{(\beta+1)(\beta+3)} - \frac{a\alpha^2 T_1^{2\beta+2}}{(\beta+1)(2\beta+2)} - \frac{b\alpha^2 T_1^{2\beta+3}}{(\beta+1)(2\beta+3)} \right\} \\ & + C_{23} \left\{ \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} \right\} + C_{13} D_0 \left\{ \frac{T-T_1}{\delta} - \frac{1}{\delta^2} \log[1 + \delta(T-T_1)] \right\} \\ & + C_{33} D_0 \left\{ T-T_1 - \frac{1}{\delta} \log[1 + \delta(T-t)] \right\} + \frac{A_3}{T} \end{aligned} \right\} \quad \text{----- (23)}$$

$$TC_{dG}(T_1, T) = \frac{1}{6} [TC_{dG_1}(T_1, T) + 4TC_{dG_2}(T_1, T) + TC_{dG_3}(T_1, T)] \quad \text{----- (24)}$$

The necessary condition for  $TC_{dG}(T_1, T)$  to be minimize is that  $\frac{\partial TC_{dG}(T_1, T)}{\partial T_1} = 0$  and  $\frac{\partial TC_{dG}(T_1, T)}{\partial T} = 0$  .

Solving these equations we find the optimum values of  $T_1$  and T say  $T_1^*$  and  $T^*$  for which cost is minimum and the sufficient condition is

$$\left( \frac{\partial^2 TC_{dG}(T_1, T)}{\partial T^2} \right) \left( \frac{\partial^2 TC_{dG}(T_1, T)}{\partial T_1^2} \right) - \left( \frac{\partial^2 TC_{dG}(T_1, T)}{\partial T \partial T_1} \right)^2 > 0,$$



$$\frac{\partial^2 TC_{dG}(T_1, T)}{\partial T_1^2} > 0, \quad \frac{\partial^2 TC_{dG}(T_1, T)}{\partial T^2} > 0$$

The optimal solution of the equations (24) can be obtained by using appropriate software. This has been illustrated by the following numerical example.

(ii) By signed distance Method, Total cost is given by

$$TC_{dS}(T_1, T) = \frac{1}{4} [TC_{dS_1}(T_1, T) + 2TC_{dS_2}(T_1, T) + TC_{dS_3}(T_1, T)]$$

$$TC_{dS_1}(T_1, T) = \left\{ \begin{aligned} & \left[ \frac{h_1}{T} \left\{ \frac{aT_1^2}{2} + \frac{bT_1^3}{3} + \frac{a\alpha\beta T_1^{\beta+2}}{(\beta+1)(\beta+2)} + \frac{b\alpha\beta T_1^{\beta+3}}{(\beta+1)(\beta+3)} - \frac{a\alpha^2 T_1^{2\beta+2}}{(\beta+1)(2\beta+2)} - \frac{b\alpha^2 T_1^{2\beta+3}}{(\beta+1)(2\beta+3)} \right\} \right] \\ & + C_{21} \left\{ \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} \right\} + C_{11} D_0 \left\{ \frac{T-T_1}{\delta} - \frac{1}{\delta^2} \log[1 + \delta(T-T_1)] \right\} \\ & + C_{31} D_0 \left\{ T-T_1 - \frac{1}{\delta} \log[1 + \delta(T-t)] \right\} + \frac{A_1}{T} \end{aligned} \right\}$$

----- (25)

$$TC_{dS_2}(T_1, T) = \left\{ \begin{aligned} & \left[ \frac{h_2}{T} \left\{ \frac{aT_1^2}{2} + \frac{bT_1^3}{3} + \frac{a\alpha\beta T_1^{\beta+2}}{(\beta+1)(\beta+2)} + \frac{b\alpha\beta T_1^{\beta+3}}{(\beta+1)(\beta+3)} - \frac{a\alpha^2 T_1^{2\beta+2}}{(\beta+1)(2\beta+2)} - \frac{b\alpha^2 T_1^{2\beta+3}}{(\beta+1)(2\beta+3)} \right\} \right] \\ & + C_{22} \left\{ \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} \right\} + C_{12} D_0 \left\{ \frac{T-T_1}{\delta} - \frac{1}{\delta^2} \log[1 + \delta(T-T_1)] \right\} \\ & + C_{32} D_0 \left\{ T-T_1 - \frac{1}{\delta} \log[1 + \delta(T-t)] \right\} + \frac{A_2}{T} \end{aligned} \right\}$$

----- (26)

$$TC_{dS_3}(T_1, T) = \left\{ \begin{aligned} & \left[ \frac{h_3}{T} \left\{ \frac{aT_1^2}{2} + \frac{bT_1^3}{3} + \frac{a\alpha\beta T_1^{\beta+2}}{(\beta+1)(\beta+2)} + \frac{b\alpha\beta T_1^{\beta+3}}{(\beta+1)(\beta+3)} - \frac{a\alpha^2 T_1^{2\beta+2}}{(\beta+1)(2\beta+2)} - \frac{b\alpha^2 T_1^{2\beta+3}}{(\beta+1)(2\beta+3)} \right\} \right] \\ & + C_{23} \left\{ \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} \right\} + C_{13} D_0 \left\{ \frac{T-T_1}{\delta} - \frac{1}{\delta^2} \log[1 + \delta(T-T_1)] \right\} \\ & + C_{33} D_0 \left\{ T-T_1 - \frac{1}{\delta} \log[1 + \delta(T-t)] \right\} + \frac{A_3}{T} \end{aligned} \right\}$$

----- (27)

$$TC_{dS}(T_1, T) = \frac{1}{4} [TC_{dS_1}(T_1, T) + 2TC_{dS_2}(T_1, T) + TC_{dS_3}(T_1, T)] \quad \text{----- (28)}$$

The necessary condition for  $TC_{dS}(T_1, T)$  to be minimize is that  $\frac{\partial TC_{dS}(T_1, T)}{\partial T_1} = 0$  and  $\frac{\partial TC_{dS}(T_1, T)}{\partial T} = 0$  .

Solving these equations we find the optimum values of  $T_1$  and T say  $T_1^*$  and  $T^*$  for which cost is minimum and the sufficient condition is

$$\left( \frac{\partial^2 TC_{dS}(T_1, T)}{\partial T^2} \right) \left( \frac{\partial^2 TC_{dS}(T_1, T)}{\partial T_1^2} \right) - \left( \frac{\partial^2 TC_{dS}(T_1, T)}{\partial T \partial T_1} \right)^2 > 0,$$

$$\frac{\partial^2 TC_{ds}(T_1, T)}{\partial T_1^2} > 0, \frac{\partial^2 TC_{ds}(T_1, T)}{\partial T^2} > 0$$

The optimal solution of the equations (28) can be obtained by using appropriate software. This has been illustrated by the following numerical example.

(iii) By Centroid Method, Total cost is given by

$$TC_{dc_1}(T_1, T) = \left\{ \begin{aligned} & \frac{h_1}{T} \left\{ \frac{aT_1^2}{2} + \frac{bT_1^3}{3} + \frac{a\alpha\beta T_1^{\beta+2}}{(\beta+1)(\beta+2)} + \frac{b\alpha\beta T_1^{\beta+3}}{(\beta+1)(\beta+3)} - \frac{a\alpha^2 T_1^{2\beta+2}}{(\beta+1)(2\beta+2)} - \frac{b\alpha^2 T_1^{2\beta+3}}{(\beta+1)(2\beta+3)} \right\} \\ & + C_{21} \left\{ \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} \right\} + C_{11} D_0 \left\{ \frac{T-T_1}{\delta} - \frac{1}{\delta^2} \log[1 + \delta(T-T_1)] \right\} \\ & + C_{31} D_0 \left\{ T-T_1 - \frac{1}{\delta} \log[1 + \delta(T-t)] \right\} + \frac{A_1}{T} \end{aligned} \right\} \tag{29}$$

$$TC_{dc_2}(T_1, T) = \left\{ \begin{aligned} & \frac{h_2}{T} \left\{ \frac{aT_1^2}{2} + \frac{bT_1^3}{3} + \frac{a\alpha\beta T_1^{\beta+2}}{(\beta+1)(\beta+2)} + \frac{b\alpha\beta T_1^{\beta+3}}{(\beta+1)(\beta+3)} - \frac{a\alpha^2 T_1^{2\beta+2}}{(\beta+1)(2\beta+2)} - \frac{b\alpha^2 T_1^{2\beta+3}}{(\beta+1)(2\beta+3)} \right\} \\ & + C_{22} \left\{ \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} \right\} + C_{12} D_0 \left\{ \frac{T-T_1}{\delta} - \frac{1}{\delta^2} \log[1 + \delta(T-T_1)] \right\} \\ & + C_{32} D_0 \left\{ T-T_1 - \frac{1}{\delta} \log[1 + \delta(T-t)] \right\} + \frac{A_2}{T} \end{aligned} \right\} \tag{30}$$

$$TC_{dc_3}(T_1, T) = \left\{ \begin{aligned} & \frac{h_3}{T} \left\{ \frac{aT_1^2}{2} + \frac{bT_1^3}{3} + \frac{a\alpha\beta T_1^{\beta+2}}{(\beta+1)(\beta+2)} + \frac{b\alpha\beta T_1^{\beta+3}}{(\beta+1)(\beta+3)} - \frac{a\alpha^2 T_1^{2\beta+2}}{(\beta+1)(2\beta+2)} - \frac{b\alpha^2 T_1^{2\beta+3}}{(\beta+1)(2\beta+3)} \right\} \\ & + C_{23} \left\{ \frac{a\alpha T_1^{\beta+1}}{\beta+1} + \frac{b\alpha T_1^{\beta+2}}{\beta+2} \right\} + C_{13} D_0 \left\{ \frac{T-T_1}{\delta} - \frac{1}{\delta^2} \log[1 + \delta(T-T_1)] \right\} \\ & + C_{33} D_0 \left\{ T-T_1 - \frac{1}{\delta} \log[1 + \delta(T-t)] \right\} + \frac{A_3}{T} \end{aligned} \right\} \tag{31}$$

$$TC_{dc}(T_1, T) = \frac{1}{3} [TC_{dc_1}(T_1, T) + TC_{dc_2}(T_1, T) + TC_{dc_3}(T_1, T)] \tag{32}$$

The necessary condition for  $TC_{dc}(T_1, T)$  to be minimize is that  $\frac{\partial TC_{dc}(T_1, T)}{\partial T_1} = 0$  and  $\frac{\partial TC_{dc}(T_1, T)}{\partial T} = 0$ .

Solving these equations we find the optimum values of  $T_1$  and T say  $T_1^*$  and  $T^*$  for which cost is minimum and the sufficient condition is

$$\left( \frac{\partial^2 TC_{dc}(T_1, T)}{\partial T^2} \right) \left( \frac{\partial^2 TC_{dc}(T_1, T)}{\partial T_1^2} \right) - \left( \frac{\partial^2 TC_{dc}(T_1, T)}{\partial T \partial T_1} \right)^2 > 0,$$

$$\frac{\partial^2 TC_{dc}(T_1, T)}{\partial T_1^2} > 0, \frac{\partial^2 TC_{dc}(T_1, T)}{\partial T^2} > 0$$

The optimal solution of the equations (32) can be obtained by using appropriate software. This has been illustrated by the following numerical example.

### V. NUMERICAL EXAMPLE

**Example:**

**Crisp Model:**

Let us consider an inventory system with the following data:  $A = 30, D_0 = 100, h = 5, a = 12, b = 2,$

$$C_1 = 2, C_2 = 4, C_3 = 6, \alpha = 0.1, \beta = 2, \delta = 0.5$$

The solution of crisp model :  $TC(T_1, T) = 62.228, T_1 = 0.8426, T = 0.9048, Q = 19.930$

**Fuzzy Model:** Let us take

$$\tilde{h} = (3, 5, 7), \tilde{C}_1 = (1.7, 2, 2.3), \tilde{C}_2 = (2, 4, 6), \tilde{C}_3 = (3, 6, 9), \tilde{A} = (25, 30, 34)$$

are all triangular fuzzy numbers. The solution of fuzzy model can be determined by following three methods.

By Graded mean Integration representation method, we have

$$TC_{dG}(T_1, T) = 61.674, T_1 = 0.8365, T = 0.8982, Q = 19.743$$

By Centroid method, we have

$$TC_{dc}(T_1, T) = 61.859, T_1 = 0.8385, T = 0.9004, Q = 19.805$$

By Signed distance method, we have

$$TC_{ds}(T_1, T) = 61.951, T_1 = 0.8395, T = 0.9015, Q = 19.836$$

#### 5.1 SENSITIVITY ANALYSIS

To study the effects of changes in the system parameters, the sensitivity is analyzed. The results are shown in below tables

**Table.1**  
Sensitivity Analysis on Parameter  $C_1$

Defuzzify Value of $C_1$	Fuzzify value of parameter $\tilde{C}_1$	$T_1$	T	$TC_{dG}(T_1, T)$
3	(2,3,4)	0.8292	0.8912	61.959
4	(3,4,5)	0.8223	0.8846	62.239
5	(4,5,6)	0.8157	0.8782	62.513
6	(5,6,7)	0.8093	0.8720	62.783

**Table.2**  
**Sensitivity Analysis on Parameter  $C_2$**

Defuzzify Value of $C_2$	Fuzzify value of parameter $\tilde{C}_2$	$T_1$	T	$TC_{dG}(T_1, T)$
3	(2.5,3,3.5)	0.8339	0.9022	61.440
2.5	(2,2.5,3)	0.8324	0.9045	61.303
2	(1.5,2,2.5)	0.8307	0.9072	61.151
1.5	(1,1.5,2)	0.8288	0.9102	60.980

**Table.3**  
**Sensitivity Analysis on Parameter  $C_3$**

Defuzzify Value of $C_3$	Fuzzify value of parameter $\tilde{C}_3$	$T_1$	T	$TC_{dG}(T_1, T)$
5	(4,5,6)	0.8339	0.9022	61.439
4	(3,4,5)	0.8307	0.9072	61.151
3	(2,3,4)	0.8267	0.9135	60.787
2	(1,2,3)	0.8214	0.9220	60.311

**Table.4**  
**Sensitivity Analysis on Parameter  $h$**

Defuzzify Value of $h$	Fuzzify value of parameter $\tilde{h}$	$T_1$	T	$TC_{dG}(T_1, T)$
6	(5,6,7)	0.7644	0.8469	65.980
7	(6,7,8)	0.7109	0.7989	70.344
8	(7,8,9)	0.6665	0.7594	74.334
9	(8,9,10)	0.6288	0.7263	78.013

**Table.5**  
**Sensitivity Analysis on Parameter  $A$**

Defuzzify Value of $A$	Fuzzify value of parameter $\tilde{A}$	$T_1$	T	$TC_{dG}(T_1, T)$
31	(30,31,32)	0.8487	0.9272	62.787
32	(31,32,33)	0.8603	0.9402	63.858
33	(32,33,34)	0.8718	0.9529	64.914
34	(33,34,35)	0.8828	0.9654	65.957

**5.2 Observations:**

- 1) From Table 1, as we increase the parameter  $C_1$ , the optimum values of  $T_1$  and T decrease. By this effect, the total cost  $TC_{dG}(T_1, T)$  increases.
- 2) From Table 2, as we decrease the parameter  $C_2$ , the optimum values of  $T_1$  decrease and T increase. By this effect, the total cost  $TC_{dG}(T_1, T)$  decreases.
- 3) From Table 3, as we decrease the parameter  $C_3$ , the optimum values of  $T_1$  decrease and T increase. By this effect, the total cost  $TC_{dG}(T_1, T)$  decreases.
- 4) From Table 4, as we increase the parameter  $h$ , the optimum values of  $T_1$  and T decrease. By this effect, the total cost  $TC_{dG}(T_1, T)$  increases.
- 5) From Table 5, as we increase the parameter  $A$ , the optimum values of  $T_1$  and T decrease. By this effect, the total cost  $TC_{dG}(T_1, T)$  increases.

## VI. CONCLUSION

This paper proposed an EOQ model for Weibull deteriorating items with linear demand rate in fuzzy environment. Shortages are allowed and partially backlogged. The deterioration cost, ordering cost, holding cost, shortage cost, opportunity cost are represented by triangular fuzzy numbers. For defuzzification graded mean, signed distance and centroid method are employed to evaluate the optimal time period of positive stock  $T_1$  and total cycle length  $T$  which minimizes the total cost. By given numerical example it has been tested that graded mean representation method gives minimum cost as compared to signed distance method and centroid method. A sensitivity analysis is also conducted on the parameters  $C_1$ ,  $C_2$ ,  $C_3$ ,  $h$ ,  $A$  to explore the effects of fuzziness. The proposed model can be extended for stock dependent demand and price dependent demand in fuzzy environment.

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