

Cost Analysis on A Multi-Item Production Inventory Model With Remanufacturing of Defective And Return Items Under Two Constraints

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Abstract - This paper explores a Multi-Item production Inventory Model with Remanufacturing of defective and return items under two constraints using Hexagonal fuzzy number. Imperfect quality items are unavoidable in an inventory system due to production process, natural disasters, damages and many other reasons. Our proposed model is considered as the regular production, during the production time some items are defected, wholesaler found that some items are damaged before the sales and after sales the customers identify the defective items. The cost parameters represented by Hexagonal fuzzy number. Expected total cost is derived and the model is defuzzified by Mean Deviation Method. Finally, a numerical example is given to illustrate the model and sensitivity analysis also made.

Keywords - Inventory, Defected items, Remanufacturing, reworked items, Dagum Distribution, Hexagonal Fuzzy number, Mean Deviation Method.

AMS Subject Classification(2010): 90B05

I. INTRODUCTION

Inventory planning and control is concerned with the acquisition and storage of the materials required for supporting various business operations. In classical model such as EOQ and EPQ developed by Harris.F.W (1913) [6] and Taft .E.W (1918) [16], it is assumed that the rate of replenishment /production and unit price of an item are constant. But later theorists worked out more comprehensive, realistic market – sensitive models. Multi - item classical inventory models are presented in well known books by Hadley.G and Whitin.T.M (1958)[5], Silver.E.A and Peterson. R(1985) [15] and Taha.H.A (2005)[17]., etc. Cheng.T.C.H (1989) [4] was related to the EOQ model with demand dependent unit cost using geometric programming technique. Pureto.J et.al (1997)[10] considered inventory models in both constrained and unconstrained situation.

While modeling an inventory problem, it is assumed that demand and various relevant costs are defined with certainty. But, in real life, these parameters are not exactly known. In this situation, uncertainties are treated as randomness and are handled through probability theory. In certain situations, uncertainties are due to fuzziness and in such cases the fuzzy set theory, originally introduced by Zadeh.L.A (1965) [18] may be applied. After Bellman.R.E

and Zadeh.L.A (1970) [3] initiated fuzzy optimization through aggregation operations that combine fuzzy goals and fuzzy - decision space. Arnold Kaufmann and Madan M Gupta (1991) [2] provided an introduction of fuzzy arithmetic operations. Later, the fuzzy linear programming model was formulated and an approach for solving problem in linear programming model with fuzzy numbers has been presented by Zimmerman.H.J (1996) [19].

Generally the inventory model are formulated by considering that only the perfect items are produced. However, in reality production items may not always be perfect. So, a proportion of the produced items can be found to be defective. Samanta.G.P (2004) [13] worked a continuous production control inventory model for deteriorating items with shortages. Hejazi.S.R et.al(2008) [7] examined the EPQ model was investigated by considering production of various types of non-perfect products. OlhaYegorova (2014)[9] developed the economic order quantity model for deteriorating items with two level of trade credit in one replacement cycle. Ritha.W and Nivetha Martin (2013) [11] explained the inventory model with waste disposal method. The inventory model with imperfect quality items with shortages has been presented by Ritha.W and Rexlin Jayakumari.S (2013) [12]. Arindum Mukhopadhyay and Adrijit Goswami (2014) [1] investigates an economic production quantity(EPQ) model

with imperfect quality items with varying set-up cost. Medhi.J (1994) [8] provided the concept of probability distribution.

In our proposed inventory model, the demand rate is taken as random variable which follows degum distribution. Miner defects are reworked and such returned products are rechecked and send to the market. The defective items are occurred at the time of manufacturing and products that are beyond repair items are remanufacturing. Remanufacturing

Assumptions and Notations

Our proposed model is constructed under the following assumptions and notations.

Assumptions:

1. Production rate is finite.
2. All demands must be satisfied.
3. Some returned items are allowed and reworked.
4. Defected items are remanufactured.
5. Alternative power supply(solar plants, their operating and maintenance) costs are allowed.
6. The production rate for i^{th} items is strictly greater than the sum of remanufacturing and reworked rate for i^{th} items.
7. Demand rate is random variable, which follows Dagum distribution.

$$E(d_i) = \begin{cases} \frac{B_i}{A_i} \times \frac{\Gamma\left(-\frac{1}{A_i}\right)\Gamma\left(\frac{1}{A_i} + P_i\right)}{\Gamma P_i}, & P_i > 0, A_i > 1, B_i > 0 \\ \infty & \text{otherwise} \end{cases}$$

Notations: The following notations are for the i^{th} item ($i=1,2,3,\dots,n$)

- Q_{si} - Inventory level at time t_1 .
- Q_{mi} - Maximum inventory level at time t_2 .
- d_i - demand rate(random variable).
- R_{ri} - rate of remanufacturing items. (*i.e.*, $R_{ri} = a(\theta_i + (1-b)(R_{1i} + R_{2i}))$)
- R_{wi} - rate of reworked items. (*i.e.*, $R_{wi} = b(R_{1i} + R_{2i})$)
- θ_i - rate of defective items.
- γ_i - rate of wastage items. (*i.e.*, $\gamma_i = (1-a)(\theta_i + (1-b)(R_{1i} + R_{2i}))$)
- R_{1i} - rate of return items during $[0, t_1]$.
- R_{2i} - rate of return items during $[t_1, t_2]$.
- R_{3i} - rate of return items during $[t_2, t_3]$.
- N - number of cycle.
- F - maximum maintenance cost of machines.
- \tilde{k}_i - fuzzy production rate.
- \tilde{H}_i - fuzzy holding cost product per unit per unit time.
- \tilde{s}_i - fuzzy setup cost per cycle.
- \tilde{R}_{ci} - fuzzy reworking cost.
- \tilde{W}_c - fuzzy wastage cost.
- \tilde{s}_{ei} - fuzzy operating and maintenance cost of solar plants per cycle.
- \tilde{m}_m - fuzzy maintenance cost of machines per cycle

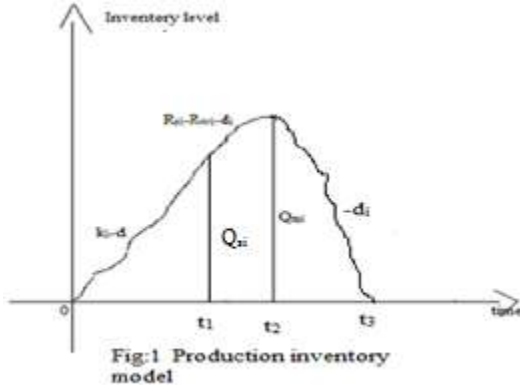
goods are also send to the market and remaining unused products should be contaminate.

In recent time, power scarcity has affected the large scale industries in manufacturing goods, to solve this problem, solar plants are being installed in many forms. It incurs a cost. The cost as alternative power supply cost, its operating and maintenance cost. This cost is also added to construct the mathematical model, in our proposed inventory model.

- \tilde{t}_c - fuzzy transportation cost
- W_t - working time of solar plants per cycle.
- S_E - fixed solar plants cost for plan period.
- a,b - arbitrary constant($0 < a, b < 1$)

II. FORMULATION OF THE CRISP MODEL

To derive the inventory level function, divide the time interval $[0, t_3]$ into three parts: $[0, t_1]$, $[t_1, t_2]$ and $[t_2, t_3]$. The regular production started at time $t=0$ and stop at time $t=t_1$. During $[0, t_1]$, the inventory level gradually increases and some items defective and returned. During the period $[t_1, t_2]$, remanufacturing process and set right the returned items are done. So, stock builds up during the period $[0, t_2]$ and declines during the period $[t_2, t_3]$. The stock is reduced to zero at t_3 .



Let $Q_i(t)$ be the inventory level during the period $[0, t_3]$. Then the differential equations governing the instantaneous state of any time t are given by,

$$\frac{dQ_i(t)}{dt} + (\theta_i + R_{1i})Q_i(t) = k_i - E(d_i) \quad 0 \leq t \leq t_1 \quad \text{-----(1)}$$

$$\frac{dQ_i(t)}{dt} + (\gamma_i + R_{2i})Q_i(t) = R_{ri} + R_{wi} - E(d_i) \quad t_1 \leq t \leq t_2 \quad \text{-----(2)}$$

$$\frac{dQ_i(t)}{dt} + R_{3i}Q_i(t) = -E(d_i) \quad t_2 \leq t \leq t_3 \quad \text{-----(3)}$$

with the boundary conditions are

$$Q_i(0) = 0, Q_i(t_1) = Q_{si}, Q_i(t_2) = Q_{mi}, Q_i(t_3) = 0.$$

The solution of the equations (1)-(3) are given by

$$Q_i(t) = \begin{cases} \frac{k_i - E(d_i)}{\theta_i + R_{1i}} + [1 - e^{-(\theta_i + R_{1i})t}] & 0 \leq t \leq t_1 \\ \frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} + \left[\left(\frac{k_i - E(d_i)}{\theta_i + R_{1i}} \right) [1 - e^{-(\theta_i + R_{1i})t_1}] - \left(\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right) \right] e^{(\gamma_i + R_{2i})(t_1 - t)} & t_1 \leq t \leq t_2 \\ \frac{E(d_i)}{R_{3i}} + [e^{R_{3i}(t_3 - t)} - 1] & t_2 \leq t \leq t_3 \end{cases}$$

$$Q_{mi} = \left[\frac{E(d_i)}{R_{3i}} \right] [e^{R_{3i}(t_3 - t_2)} - 1]$$

$$\begin{aligned} \text{Holding cost} &= H_i \int_0^{t_3} Q_i(t) dt \\ &= H_i \left[\int_0^{t_1} Q_i(t) dt + \int_{t_1}^{t_2} Q_i(t) dt + \int_{t_2}^{t_3} Q_i(t) dt \right] \\ &= H_i \left\{ \left[\left(\frac{k_i - E(d_i)}{\theta_i + R_{1i}} \right) \left[t_1 + \frac{e^{-(\theta_i + R_{1i})t_1} - 1}{(\theta_i + R_{1i})} \right] \right] \right. \\ &\quad + \left[\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right] (t_2 - t_1) - \left[Q_{si} - \left(\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right) \left[\frac{e^{(\gamma_i + R_{2i})(t_1 - t_2)} - 1}{(\gamma_i + R_{2i})} \right] \right] \\ &\quad \left. + \frac{E(d_i)}{R_{3i}} \left[\frac{e^{R_{3i}(t_3 - t_2)} - 1}{R_{3i}} + (t_2 - t_3) \right] \right\} \end{aligned}$$

Setup cost = s_i
Wastage cost = $w_c \gamma_i (t_2 - t_1)$
Cost of reworked items = $R_{ci} b (R_{1i} + R_{2i}) (t_2 - t_1)$
Transportation cost = $t_c E(d_i)$
Solar plants operating and maintenance cost = $s_{ei} W_t$
Solar Plant cost = S_E

Expected total cost = [Holding cost + Setup cost + wastage cost + cost of reworked items
Transportation cost + Solar plants operating and maintenance cost
+ Machines maintenance cost + Solar Plant cost]

Expected Total cost is given by

$$E(TC) = \sum_{i=1}^n H_i \left\{ \left[\left(\frac{k_i - E(d_i)}{\theta_i + R_{1i}} \right) \left[t_1 + \frac{e^{-(\theta_i + R_{1i})t_1} - 1}{(\theta_i + R_{1i})} \right] \right] \right. \\ + \left[\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right] (t_2 - t_1) \\ - \left[\left(\frac{k_i - E(d_i)}{\theta_i + R_{1i}} \right) \left[1 - e^{-(\theta_i + R_{1i})t_1} \right] \right] \left[\frac{e^{(\gamma_i + R_{2i})(t_1 - t_2)} - 1}{(\gamma_i + R_{2i})} \right] \\ - \left(\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right) \\ \left. + \frac{E(d_i)}{R_{3i}} \left[\frac{e^{R_{3i}(t_3 - t_2)} - 1}{R_{3i}} + (t_2 - t_3) \right] \right\} \quad \text{---(12)}$$

$$+ s_i + w_c \gamma_i (t_2 - t_1) + R_{ci} b (R_{1i} + R_{2i}) (t_2 - t_1) + t_c E(d_i) + m_m + s_{ei} W_t + S_E$$

Subject to

$$\sum_{i=1}^n (R_{ri} + R_{wi}) < \sum_{i=1}^n k_i$$

$$m_m N < F$$

Using Lagrange multiplier method, the Lagrange function is

$$L(t_1, t_2, t_3, \lambda) = \sum_{i=1}^n H_i \left\{ \begin{aligned} & \left[\left(\frac{k_i - E(d_i)}{\theta_i + R_{1i}} \right) \left[t_1 + \frac{e^{-(\theta_i + R_{1i})t_1} - 1}{(\theta_i + R_{1i})} \right] \right. \\ & + \left. \left[\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right] (t_2 - t_1) \right. \\ & - \left. \left[\left(\frac{k_i - E(d_i)}{\theta_i + R_{1i}} \right) \left[1 - e^{-(\theta_i + R_{1i})t_1} \right] \right] \left[\frac{e^{(\gamma_i + R_{2i})(t_1 - t_2)} - 1}{(\gamma_i + R_{2i})} \right] \right. \\ & - \left. \left[\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right] \right. \\ & + \left. \frac{E(d_i)}{R_{3i}} \left[\frac{e^{R_{3i}(t_3 - t_2)} - 1}{R_{3i}} + (t_2 - t_3) \right] \right\} \\ & + s_i + w_c \gamma_i (t_2 - t_1) + R_{ci} b (R_{1i} + R_{2i}) (t_2 - t_1) + t_c E(d_i) + m_m + s_{ei} W_t + S_E \\ & - \lambda_1 \sum_{i=1}^n \frac{(R_{ri} + R_{wi})}{k_i} - 1 - \lambda_2 (m_m N - F) \end{aligned} \right.$$

By using Kuhn-Tucker necessary condition

$$\frac{\partial L}{\partial t_1} = \sum_{i=1}^n H_i \left\{ \begin{aligned} & \left[\left(\frac{k_i - E(d_i)}{\theta_i + R_{1i}} \right) \left[1 - \frac{e^{-(\theta_i + R_{1i})t_1}}{(\theta_i + R_{1i})^2} \right] - \left[\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right] - \left[\left(\frac{k_i - E(d_i)}{\theta_i + R_{1i}} \right) \left[1 - e^{-(\theta_i + R_{1i})t_1} \right] \right] \left[\frac{e^{(\gamma_i + R_{2i})(t_1 - t_2)}}{(\gamma_i + R_{2i})^2} \right] \right. \\ & + \left. \left[\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right] \left[\frac{e^{(\gamma_i + R_{2i})(t_1 - t_2)}}{(\gamma_i + R_{2i})^2} \right] - \left[\frac{e^{(\gamma_i + R_{2i})(t_1 - t_2)} - 1}{(\gamma_i + R_{2i})} \right] \left[\left(\frac{k_i - E(d_i)}{\theta_i + R_{1i}} \right) \left[\frac{e^{-(\theta_i + R_{1i})t_1}}{(\theta_i + R_{1i})} \right] \right. \\ & \left. \left. - w_c \gamma_i - R_{ci} b (R_{1i} + R_{2i}) \right] \right\} \\ \\ \frac{\partial L}{\partial t_2} = \sum_{i=1}^n H_i \left\{ \begin{aligned} & \left[\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right] + \left[\left(\frac{k_i - E(d_i)}{\theta_i + R_{1i}} \right) \left[1 - e^{-(\theta_i + R_{1i})t_1} \right] - \left(\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right) \right] \left[\frac{e^{(\gamma_i + R_{2i})(t_1 - t_2)}}{(\gamma_i + R_{2i})^2} \right] \\ & - \frac{E(d_i)}{R_{3i}} \left(\frac{e^{R_{3i}(t_3 - t_2)}}{R_{3i}^2} - 1 \right) \right. \\ & \left. + w_c \gamma_i + R_{ci} b (R_{1i} + R_{2i}) \right\} \\ \\ \frac{\partial L}{\partial t_3} = \sum_{i=1}^n H_i \left\{ - \frac{E(d_i)}{R_{3i}} \left(\frac{e^{R_{3i}(t_3 - t_2)}}{R_{3i}^2} - 1 \right) \right\} \\ \\ \frac{\partial L}{\partial \lambda_1} = \sum_{i=1}^n (k_i - (R_{ri} + R_{wi})) \\ \\ \frac{\partial L}{\partial \lambda_2} = m_m N - F \end{aligned} \right.$$

Solve the above equations, the values t_1, t_2 and t_3 are obtained.

III. HEXAGONAL FUZZY NUMBER AND ITS PROPERTIES

A Hexagonal fuzzy number \tilde{A} is described as a fuzzy subset on the real line R whose membership function $\mu_{\tilde{A}}(x)$ is defined as follows

$$m_{\tilde{A}}(x) = \begin{cases} 0 & \text{for } x \notin a \\ W_A \frac{ax - a\bar{0}}{c - a\bar{0}} & \text{for } a \leq x \leq b \\ W_A + (1 - W_A) \frac{ax - b\bar{0}}{c - b\bar{0}} & \text{for } b \leq x \leq c \\ 1 & \text{for } c \leq x \leq d \\ W_A + (1 - W_A) \frac{ex - x\bar{0}}{e - d\bar{0}} & \text{for } d \leq x \leq e \\ W_A \frac{ef - x\bar{0}}{f - e\bar{0}} & \text{for } e \leq x \leq f \\ 0 & \text{for } x \notin f \end{cases}$$

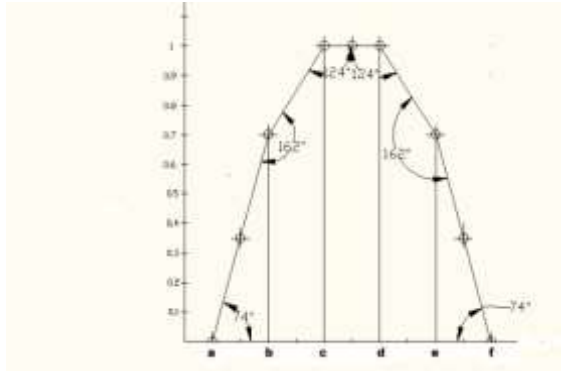


Figure 2: Graphical Representation of Hexagonal Fuzzy Number

Where $0.6 \leq W_A < 1$, a, b, c, d, e and f are real numbers.

This type of fuzzy number is denoted by $\tilde{A} = (a, b, c, d, e, f; W_A)_{HFN}$

$\mu_{\tilde{A}}$ satisfies the following conditions:

1. $\mu_{\tilde{A}}$ is a continuous mapping from \mathbb{R} to the closed interval $[0, 1]$.
2. $\mu_{\tilde{A}}$ is a convex function.
3. $\mu_{\tilde{A}}(x) = 0, -\infty < x \leq a$.
4. $\mu_{\tilde{A}}(x) = \delta_l(x)$ is strictly increasing on (a, c) .
5. $\mu_{\tilde{A}}(x) = 1, x \in [c, d]$.
6. $\mu_{\tilde{A}}(x) = \delta_r(x)$ is strictly decreasing on (d, f) .
7. $\mu_{\tilde{A}}(x) = 0, f < x \leq \infty$.

Remark:

If $0 < W_A < 0.6$, then \tilde{A} becomes a trapezoidal fuzzy number.

IV. MEAN DEVIATION OF THE FUZZY NUMBER

Let us consider \tilde{A} is a Hexagonal fuzzy number such that $\tilde{A} = (a, b, c, d, e, f)_{HFN}$, then the parameters of Mean deviation of Hexagonal fuzzy number is explained as follows.

The membership function is divided into two parts. One at the left of $x_{m1}, x_l(\alpha)$ and the other at right of

$x_{m2}, x_r(\alpha)$. The quantity $\delta_l(\tilde{A}) = \int_{\alpha=0}^1 [x_m - x_l(\alpha)] d\alpha$ is called the left mean deviation, the quantity

$\delta_r(\tilde{A}) = \int_{\alpha=0}^1 [x_r(\alpha) - x_m] d\alpha$ is called the right mean deviation and $\delta(\tilde{A}) = \delta_l(\tilde{A}) + \delta_r(\tilde{A})$ is called the mean deviation

of the fuzzy number.

By construction, a fuzzy number have $x_l(\alpha) \leq x_{m1} \leq x_m \leq x_{m2} \leq x_r(\alpha)$, $\delta_l(\tilde{A}) \geq 0, \delta_r(\tilde{A}) \geq 0, \delta(\tilde{A}) \geq 0$

.Also $\delta_l(\tilde{A})$ represents the area to the left of x_m and $\delta_r(\tilde{A})$ represents the area of the right. Where $\delta(\tilde{A})$ is the sum of the two areas.

Therefore the mean deviation of the Hexagonal fuzzy number is

$$\delta(\tilde{A}) = \left[\begin{aligned} &\left(\frac{W_A}{2(a-b)} \right) \{ (x_m - b)^2 - (x_m - a)^2 \} + \left(\frac{1-W_A}{2(b-c)} \right) \{ (x_m - c)^2 - (x_m - b)^2 \} \\ &+ \left(\frac{W_A}{2(e-f)} \right) \{ (e - x_m)^2 - (f - x_m)^2 \} + \left(\frac{1-W_A}{2(d-e)} \right) \{ (d - x_m)^2 - (e - x_m)^2 \} \end{aligned} \right]$$

V. THE PROPOSED INVENTORY MODEL IN FUZZY ENVIRONMENT

If the cost parameters are fuzzy number, then the problem (13) is transformed to

$$E(\tilde{TC}) = \sum_{i=1}^n \tilde{H}_i \left\{ \begin{aligned} &\left[\left(\frac{\tilde{k}_i - E(d_i)}{\theta_i + R_{1i}} \right) \left[t_1 + \frac{e^{-(\theta_i + R_{1i})t_1} - 1}{(\theta_i + R_{1i})} \right] \right. \\ &+ \left. \left[\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right] (t_2 - t_1) \right. \\ &- \left. \left[\left(\frac{\tilde{k}_i - E(d_i)}{\theta_i + R_{1i}} \right) \left[1 - e^{-(\theta_i + R_{1i})t_1} \right] \right] \left[\frac{e^{(\gamma_i + R_{2i})(t_1 - t_2)} - 1}{(\gamma_i + R_{2i})} \right] \right. \\ &- \left. \left(\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right) \right. \\ &+ \left. \frac{E(d_i)}{R_{3i}} \left[\frac{e^{R_{3i}(t_3 - t_2)} - 1}{R_{3i}} + (t_2 - t_3) \right] \right. \\ &\left. + \tilde{s}_i + \tilde{w}_c \gamma_i (t_2 - t_1) + \tilde{R}_{ci} b (R_{1i} + R_{2i}) (t_2 - t_1) + \tilde{t}_c E(d_i) + \tilde{m}_m + \tilde{s}_{ei} W_t + S_E \right\} \quad \text{-----(13)} \end{aligned} \right.$$

Subject to

$$\underset{i=1}{\overset{n}{\tilde{a}}} (R_{ri} + R_{wi}) < \underset{i=1}{\overset{n}{\tilde{a}}} k_i$$

$$\tilde{m}_m N < F$$

Where ~ represents the fuzzification of the parameters.

In the proposed model, the parameters $k_i, H_i, s_i, w_c, t_c, s_{ei}$ are considered as Hexagonal fuzzy number.

$$\tilde{k}_i = [k_1, k_2, k_3, k_4, k_5, k_6] \quad \tilde{H}_i = [H_{m1}, H_{m2}, H_{m3}, H_{m4}, H_{m5}, H_{m6}]$$

$$\tilde{s}_i = [s_1, s_2, s_3, s_4, s_5, s_6] \quad \tilde{w}_c = [w_{c1}, w_{c2}, w_{c3}, w_{c4}, w_{c5}, w_{c6}]$$

$$\tilde{t}_c = [t_{c1}, t_{c2}, t_{c3}, t_{c4}, t_{c5}, t_{c6}] \quad \tilde{R}_{ci} = [R_{c1}, R_{c2}, R_{c3}, R_{c4}, R_{c5}, R_{c6}] \quad \tilde{s}_{ei} = [s_{e1}, s_{e2}, s_{e3}, s_{e4}, s_{e5}, s_{e6}]$$

The corresponding fuzzy Problem (13) is

$$E(\tilde{T}C) = \sum_{i=1}^n \delta_{\tilde{H}_{R_i}} \left\{ \left[\left(\frac{\delta_{\tilde{k}_i} - E(d_i)}{\theta_i + R_{1i}} \right) \left[t_1 + \frac{e^{-(\theta_i + R_{1i})t_1} - 1}{(\theta_i + R_{1i})} \right] \right. \right. \\ \left. \left. + \left[\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right] (t_2 - t_1) \right] \left[\left(\frac{\delta_{\tilde{k}_i} - E(d_i)}{\theta_i + R_{1i}} \right) \left[1 - e^{-(\theta_i + R_{1i})t_1} \right] \right] \left[\frac{e^{(\gamma_i + R_{2i})(t_1 - t_2)} - 1}{(\gamma_i + R_{2i})} \right] \right\} \\ \left. \left[- \left(\frac{R_{ri} + R_{wi} - E(d_i)}{\gamma_i + R_{2i}} \right) \right] \left[\frac{e^{R_{3i}(t_3 - t_2)} - 1}{R_{3i}} + (t_2 - t_3) \right] \right\} \\ + \delta_{\tilde{s}_i} + \delta_{\tilde{w}_c} \gamma_i (t_2 - t_1) + \delta_{\tilde{R}_{ci}} b(R_{1i} + R_{2i})(t_2 - t_1) + \delta_{\tilde{t}_c} E(d_i) + \delta_{\tilde{m}_m} + \delta_{\tilde{s}_{ei}} W_i + S_E$$

$$\mathring{a}_{i=1}^n (R_{ri} + R_{wi}) < \mathring{a}_{i=1}^n d_{k_i}$$

$$\delta_{\tilde{m}_m} N < F$$

Where

$$\delta_{\tilde{H}_i} = \left[\left(\frac{W_A}{2(H_1 - H_2)} \right) \left\{ (x_H - H_2)^2 - (x_H - H_1)^2 \right\} \right. \\ \left. + \left(\frac{1 - W_A}{2(H_2 - H_3)} \right) \left\{ (x_H - H_3)^2 - (x_H - H_2)^2 \right\} \right. \\ \left. + \left(\frac{W_A}{2(H_5 - H_6)} \right) \left\{ (H_5 - x_H)^2 - (H_6 - x_H)^2 \right\} \right. \\ \left. + \left(\frac{1 - W_A}{2(H_4 - H_5)} \right) \left\{ (H_4 - x_H)^2 - (H_5 - x_H)^2 \right\} \right]$$

Similarly for

$$\delta_{\tilde{k}_i}, \delta_{\tilde{s}_i}, \delta_{\tilde{w}_c}, \delta_{\tilde{t}_c}, \delta_{\tilde{R}_{ci}}, \delta_{\tilde{m}_m} \text{ and } \delta_{\tilde{s}_{ei}}.$$

VI. NUMERICAL EXAMPLE

The two wheeler manufacturing company produces two items. The relevant data for the two items are given below.

$$W_i=1200; W=0.7; S_E= 1000000 ; a=0.6; b=0.6$$

$$\tilde{s}_{ei} = [1 \ 3 \ 5 \ 7 \ 9 \ 11] \quad \tilde{t}_c = [1 \ 1.5 \ 2 \ 2.5 \ 3 \ 3.5] \quad \tilde{w}_c = [0.2 \ 0.3 \ 0.4 \ 0.5 \ 0.6 \ 0.7]$$

$$\tilde{m}_m = [1 \ 1.5 \ 2 \ 2.5 \ 3 \ 3.5]$$

Item1:

$$R_{11}=5; R_{21}=2; R_{31}=3; A=9.4; B=52.2; P=14.5; \theta_1 = 2$$

$$\tilde{k}_1 = [1000 \ 3000 \ 5000 \ 7000 \ 9000 \ 11000] \quad \tilde{H}_1 = [5 \ 7 \ 9 \ 11 \ 13 \ 15]$$

$$\tilde{R}_{c1} = [0.2 \ 0.3 \ 0.4 \ 0.5 \ 0.6 \ 0.7]$$

$$\tilde{s}_1 = [100 \ 150 \ 200 \ 250 \ 300 \ 350]$$

Item2:

$$R_{12}=4; R_{22}=2; R_{32}=2; A=2.5; B=42.120; P=14.50;$$

$$\theta_2 = 1$$

$$\tilde{k}_2 = [2000 \ 3500 \ 5000 \ 6500 \ 8000 \ 9500] \quad \tilde{H}_2 = [6 \ 8 \ 10 \ 12 \ 14 \ 16]$$

$$\tilde{R}_{c_2} = [0.2 \ 0.4 \ 0.6 \ 0.8 \ 1.0 \ 1.2] \quad \tilde{s}_2 = [80 \ 160 \ 240 \ 320 \ 400 \ 480]$$

Using MATLAB software, the expected maximum inventory level, the expected inventory level at t_1 and expected minimum total cost are obtained in both crisp and fuzzy environment and are given in the following table.

Table 1 : Comparison of crisp and fuzzy result

Model	Item	k_i	H_i	R_c	s	t_c	w_c	s_{ei}	m_m	t_1	t_2	t_3	$E(Q_{si}^*)$	$E(Q_{mi}^*)$	$E(TC)$
crisp	1	1000	5	0.2	100	1	0.2	1	1	11.39	20.2	22.41	961	2978	1356000
	2	2000	6	0.2	80								985	6997	
	1	3000	7	0.3	150	1.5	0.3	3	1.5	11.39	20.21	22.42	972	3004	1367000
	2	3500	8	0.4	160								991	7002	
	1	5000	9	0.4	200	2	0.4	5	2	11.40	20.21	22.42	981	3021	1453000
	2	5000	10	0.6	240								1011	7026	
	1	7000	11	0.5	250	2.5	0.5	7	2.5	11.40	20.22	22.44	993	3043	1579000
	2	6500	12	0.8	320								1056	7165	
	1	9000	13	0.6	300	3	0.6	9	3	11.43	20.24	22.46	1021	3132	1584000
	2	8000	14	1.0	400								1089	7212	
	1	11000	15	0.7	350	3.5	0.7	11	3.5	11.43	20.25	22.48	1037	3187	1676000
	2	9500	16	1.2	480								1134	7432	
fuzzy	1	6800	6.8	0.34	170	1.7	0.34	6.8	1.7	11.37	20.18	22.42	1123	3215	1343000
	2	5100	6.8	0.27	272								1374	8014	

Table 2 : Effect of changes in holding cost of the fuzzy inventory model

%change	Item	H_i	t_1	t_2	t_3	$E(Q_{si}^*)$	$E(Q_{mi}^*)$	$E(TC)$
-50	1	5.8	11.38	20.23	22.36	943	2634	1145000
	2	6.6				915	6325	
-25	1	6.3	11.36	20.21	22.34	957	2789	1265000
	2	7.1				974	6894	
+25	1	7.3	11.40	20.24	22.37	986	2856	1394000
	2	8.1				1097	6983	
+50	1	7.8	11.42	20.24	22.38	1026	2974	1532400
	2	8.6				1124	7025	

VII. CONCLUSION

From table 1, observed that the optimal values are given for the fuzzy models along with the crisp model. The expected inventory level in the fuzzy environment is high compared to the crisp value. The expected minimum total cost in the crisp environment is high compared to the fuzzy value. Finally, conclude that the fuzzy model can be executable in the real world.

In table 2, the sensitivity analysis are given for the fuzzy model, from the same the following are observed.

- In the regular manufacturing holding cost H_i
 - If the regular manufacturing holding cost is taken as either 25% or 50% decrease then the expected inventory level Q_{si} , the expected maximum

inventory level Q_{mi} and expected total cost will be decreases.

2. If the regular manufacturing holding cost is taken as either 25% or 50% increase then the expected inventory level Q_{si} , the expected maximum inventory level Q_{mi} and expected total cost will be increases.

- In the same manner, all cost parameters can be analyzed.

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