# Solutions to Fuzzy Inventory Model with Fuzzy Demand Rate Using Heptagonal Fuzzy Numbers 

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#### Abstract

The objective of this article focuses a solution to fuzzy inventory model with fuzzy demand rate. Mathematical model has been described for finding the cycle length and total inventory cost in both crisp and fuzzy environments. Robust ranking is used to defuzzify the total cost function and this model is solved by using GP technique and Kuhn Tucker method. Next, comparative analysis between GP technique and Kuhn Tucker method. Additionally, a numerical examples and sensitivity analysis are given to the proposed model.


Keywords:Total cost, GP technique, Kuhn-tucker condition, Robust's Ranking defuzzification.

## 1. Introduction

In general, the analytic complexity of inventory models depends on whether the demand for an item is deterministic or probabilistic. In real life, demand is usually probabilistic, but in some cases the simpler deterministic approximation may be acceptable. Within either category, the demand may or may not vary with time .[3 ] Burwell, T. H., Dave, D. S., Fitzpatrick, K. E., \& Roy, M. R. (1997) investigated an Economic lot size model for price-dependent demand under quantity and freight discounts. The Kuhn tucker conditions provide the most unifying theory for all non linear programming problems. The main contribution of this model is the development of the general Kuhn tucker necessary conditions for determining the stationary points. [5] MangasarianO.L, [6] McCormick G.P solved nonlinear Programming. [9] Yang J H, Cao B Y (2005) presented Geometric programming with fuzzy relation equation constraints. [4] Cao B Y (1993) describes an extended fuzzy geometric programming. The basic experiment towards concepts of fuzziness was made by zadeh [10]. Several researchers have carried out examine on several fuzzy numbers. [13] Zimmermann H.J, proposed results obtained by Fuzzy set theory and its applications.

This paper is standardized as follows. Section 2 contains the notations and arithmetic operations used for this model. The G.P technique development is formulated both crisp and fuzzy set in Section 3. A Section 4 describes the solution procedure for Kuhn tucker condition both crisp and fuzzy set. Numerical examples, sensitivity analysis and graphs are provided in Section 5 related to our models. Finally, conclusion is included in section 6 .

## 2. Model's Criterion and Enlargement

### 2.1 Notation:

d- Demand
O- Ordering Cost
H- Holding Cost
C- Purchase Cost
S- Shortage Cost
L- Cycle Length
$\theta$ - Time $0 \leq \theta \leq \mathrm{L}$
$\lambda$ - Shortage Period
TIC-Total Inventory Cost.

### 2.2 Robust's Ranking

Robust's ranking which satisfy compensation, linearity, and additively properties and provides results which are consistent with human intuition. If $\tilde{A}_{H}$ is a fuzzy number then the ranking is defined by

$$
R\left(\tilde{A}_{H}\right)=\int_{0}^{1} 0.5\left(a_{h \alpha}^{L}, a_{h \alpha}^{U}\right) d \alpha
$$

Where $\left(a_{h \alpha}^{L}, a_{h \alpha}^{U}\right)$ is the $\alpha$ level cut of fuzzy number $\tilde{A}_{H}$.

## Proposed Ranking

If $\left(h_{1}, h_{2}, h_{3}, h_{4}, h_{5}, h_{6}, h_{7}\right)$ are heptagonal fuzzy numbers, then
$\left(a_{h \alpha}^{L}, a_{h \alpha}^{U}\right)=\left\{\left(h_{2}-h_{1}\right) \alpha+h_{1}, h_{4}-\left(h_{4}-h_{3}\right) \alpha,\left(h_{6}-h_{5}\right) \alpha+h_{5}, h_{7}-\left(h_{7}-h_{5}\right) \alpha\right\}$
Hence, the fuzzy version of heptagonal ranking is
$R\left(\tilde{A}_{H}\right)=\int_{0}^{1} 0.5\left\{\left(h_{2}-h_{1}\right) \alpha+h_{1}, h_{4}-\left(h_{4}-h_{3}\right) \alpha,\left(h_{6}-h_{5}\right) \alpha+h_{5}, h_{7}-\left(h_{7}-h_{5}\right) \alpha\right\} d \alpha$.

### 2.3 Arithmetic operations on heptagonal fuzzy numbers

Let $\tilde{A}=\left(a_{1}, a_{2}, a_{3}, a_{4}, a_{5}, a_{6}, a_{7}\right), \tilde{B}=\left(b_{1}, b_{2}, b_{3}, b_{4}, b_{5}, b_{6}, b_{7}\right)$ be two heptagonal fuzzy numbers. Then

1) The addition of $\tilde{A}$ and $\tilde{B}$ is $\tilde{A} \oplus \tilde{B}=\left(a 1+b_{1}, a_{2}+b_{2}, a_{3}+b_{3}, a_{4}+b_{4}, a_{5}+b_{5}, a_{6}+b_{6}, a_{7}+b_{7}\right)$.
2) The multiplication of $\tilde{A}$ and $\tilde{B}$ is $\tilde{A} \otimes \tilde{B}=\left(\mathrm{a}_{1} \mathrm{~b}_{1}, \mathrm{a}_{2} \mathrm{~b}_{2}, \mathrm{a}_{3} \mathrm{~b}_{3}, \mathrm{a}_{4} \mathrm{~b}_{4}, \mathrm{a}_{5} \mathrm{~b}_{5}, \mathrm{a}_{6} \mathrm{~b}_{6}, \mathrm{a}_{7} \mathrm{~b}_{7}\right)$.
3) $\frac{1}{\tilde{B}}=\left(\frac{1}{b 7}, \frac{1}{b 6}, \frac{1}{b 5}, \frac{1}{b 4}, \frac{1}{b 3}, \frac{1}{b 2}, \frac{1}{b 1}\right)$, then the division of $\tilde{A}$ and $\tilde{B}$ is $\stackrel{\widetilde{A}}{\widetilde{B}}=\left(\frac{a 1}{b 7}, \frac{a 2}{b 6}, \frac{a 3}{b 5}, \frac{a 4}{b 4}, \frac{a 5}{b 3}, \frac{a 6}{b 2}, \frac{a 7}{b 1}\right)$.

## 3. Geometric Programming Technique

### 3.1 Crisp Model Using GP Technique

$$
\begin{equation*}
\operatorname{TIC}(L)=C d+\frac{O}{L}+\frac{h d \theta^{2}}{2 L}+S d \frac{(L-\theta)^{2}}{2 L} \tag{1}
\end{equation*}
$$

Here (C- $\theta \mathrm{S}$ ) d is constant.
Hence the total cost becomes,

$$
\begin{equation*}
\operatorname{TIC}(L)=\frac{1}{L}\left(O+\frac{(h+S) \theta^{2} d}{2}\right)+\frac{S d L}{2} \tag{2}
\end{equation*}
$$

Using the GP techniques to eqn (1), we get
$t(\ell)=\left(\frac{O+\frac{(h+S) \theta^{2} d}{2}}{\ell_{1}}\right)^{\ell_{1}}\left(\frac{S d}{2 \ell_{2}}\right)^{\ell_{2}}$
S.t
$\ell_{1}+\ell_{2}=1$
$-\ell_{1}+\ell_{2}=0$
Solving we get,

$$
\ell_{1}=\ell_{2}=\frac{1}{2}
$$

i.e., $t(\ell)=\left(\frac{O+\frac{(h+S) \theta^{2} d}{2}}{1 / 2}\right)^{1 / 2}\left(\frac{S d}{21 / 2}\right)^{1 / 2}$

From primal-dual relation
$\frac{1}{L}\left(O+\frac{(h+S) \theta^{2} d}{2}\right)=\ell_{1} t(\ell)$
$\frac{S d L}{2}=\ell_{2} t(\ell)$
Solving the above eqns, we get the cycle length and total cost is given as follows
$L=\sqrt{\frac{2}{S d}\left(O+\frac{\theta^{2} d}{2}(h+S)\right)}$
$T I C(L)=\sqrt{2 S d\left(O+\frac{\theta^{2} d}{2}(h+S)\right)}+(C-S \theta) d$

### 3.2 Fuzzy Model Using GP Technique

In this model we take the criterions ordering cost and demand are heptagonal fuzzy numbers.
Then from eqn (1) we have
$\operatorname{TIC}(L)=\frac{1}{L}\left(\tilde{O}+\frac{(h+S) \theta^{2} \tilde{d}}{2}\right)+\frac{S \tilde{d} L}{2}$
Where $d=\left(d_{1}, d_{2}, d_{3}, d_{4}, d_{5}, d_{6}, d_{7}\right), O=\left(O_{1}, O_{2}, O_{3}, O_{4}, O_{5}, O_{6}, O_{7}\right)$.
Using robust ranking defuzzification for heptagonal fuzzy numbers
Corresponding dual form of eqn (8) is given by
$t(\ell)=\left(\frac{2\left(O_{1}+O_{2}+O_{3}+O_{4}+2 O_{5}+O_{6}+O_{7}\right)+(h+S) \theta^{2}\left(d_{1}+d_{2}+d_{3}+d_{4}+2 d_{5}+d_{6}+d_{7}\right)}{8 \ell_{1}}\right)^{\ell_{1}}\left(\frac{S\left(d_{1}+d_{2}+d_{3}+d_{4}+2 d_{5}+d_{6}+d_{7}\right)}{8 \ell_{2}}\right)^{\ell_{2}}$
Subject to,
$\ell_{1}+\ell_{2}=1$
$-\ell_{1}+\ell_{2}=0$
Solving we get,

$$
\ell_{1}=\ell_{2}=\frac{1}{2}
$$

Putting the values in the above dual form then we get the optimal solutions are obtained from GP technique as follows.

$$
\begin{aligned}
& L=\sqrt{\frac{\left(2\left(O_{1}+O_{2}+O_{3}+O_{4}+2 O_{5}+O_{6}+O_{7}\right)+\theta^{2}\left(d_{1}+d_{2}+d_{3}+d_{4}+2 d_{5}+d_{6}+d_{7}\right)(h+S)\right)}{S\left(d_{1}+d_{2}+d_{3}+d_{4}+2 d_{5}+d_{6}+d_{7}\right)}} \\
& T I C(L)=\sqrt{\frac{S\left(d_{1}+d_{2}+d_{3}+d_{4}+2 d_{5}+d_{6}+d_{7}\right)}{16}\left(2\left(O_{1}+O_{2}+O_{3}+O_{4}+2 O_{5}+O_{6}+O_{7}\right)+\theta^{2}\left(d_{1}+d_{2}+d_{3}+d_{4}+2 d_{5}+d_{6}+d_{7}\right)(h+S)\right)} \\
& +\frac{(C-S \theta)\left(d_{1}+d_{2}+d_{3}+d_{4}+2 d_{5}+d_{6}+d_{7}\right)}{4}
\end{aligned}
$$

## 4. Kuhn-Tucker Condition Approach

### 4.1 Crisp Model Using Kuhn-Tucker Condition

The crisp total inventory cost is given by,
$\operatorname{TIC}(L)=C d+\frac{O}{L}+\frac{h d \theta^{2}}{2 L}+S d \frac{(L-\theta)^{2}}{2 L}$
Differentiating partially w.r.t L and equating it to zero then implies the cycle length is,

$$
L=\sqrt{\frac{2}{S d}\left(O+\frac{\theta^{2} d}{2}(h+S)\right)}
$$

### 4.2 Fuzzy Model Using Kuhn-Tucker Condition

The fuzzy total inventory cost is given by,

$$
\operatorname{TIC}(L)=C \tilde{d}+\frac{\tilde{O}}{L}+\frac{h \tilde{d} \theta^{2}}{2 L}+S \tilde{d} \frac{(L-\theta)^{2}}{2 L}
$$

Suppose $d=\left(d_{1}, d_{2}, d_{3}, d_{4}, d_{5}, d_{6}, d_{7}\right)$, are heptagonal fuzzy numbers. Using robust ranking defuzzification for heptagonal fuzzy numbers and then we solve the cycle length is given as follows.

$$
\operatorname{TIC}(L)=\frac{1}{4}\left[\begin{array}{l}
\frac{1}{L}\left(O_{1}+\frac{(h+S) \theta^{2} d_{1}}{2}\right)+\frac{S d_{1} L}{2}+(C-\theta S) d_{1}+ \\
\frac{1}{L}\left(O_{2}+\frac{(h+S) \theta^{2} d_{2}}{2}\right)+\frac{S d_{2} L}{2}+(C-\theta S) d_{2}+ \\
\frac{1}{L}\left(O_{3}+\frac{(h+S) \theta^{2} d_{3}}{2}\right)+\frac{S d_{3} L}{2}+(C-\theta S) d_{3}+ \\
\frac{1}{L}\left(O_{4}+\frac{(h+S) \theta^{2} d_{4}}{2}\right)+\frac{S d_{4} L}{2}+(C-\theta S) d_{4}+ \\
\frac{1}{L}\left(2 O_{5}+\frac{2(h+S) \theta^{2} d_{5}}{2}\right)+\frac{2 S d_{5} L}{2}+2(C-\theta S) d_{5}+ \\
\frac{1}{L}\left(O_{6}+\frac{(h+S) \theta^{2} d_{6}}{2}\right)+\frac{S d_{6} L}{2}+(C-\theta S) d_{6}+ \\
\left.O_{7}+\frac{(h+S) \theta^{2} d_{7}}{2}\right)+\frac{S d_{7} L}{2}+(C-\theta S) d_{7}
\end{array}\right]
$$

Differentiating partially w.r.t L and equating it to zero.

$$
\begin{aligned}
& \frac{\partial T I C(L)}{\partial L}=0 \\
& \frac{\partial T I C(L)}{\partial L}=\frac{-1}{L^{2}}\left(\left(O_{1}+O_{2}+O_{3}+O_{4}+2 O_{5}+O_{6}+O_{7}\right)+\frac{(h+S) \theta^{2}\left(d_{1}+d_{2}+d_{3}+d_{4}+2 d_{5}+d_{6}+d_{7}\right)}{2}\right) \\
& +\frac{S\left(d_{1}+d_{2}+d_{3}+d_{4}+2 d_{5}+d_{6}+d_{7}\right)}{2}=0
\end{aligned}
$$

Therefore,

$$
\tilde{L}=\sqrt{\frac{2\left(O_{1}+O_{2}+O_{3}+O_{4}+2 O_{5}+O_{6}+O_{7}\right)+(h+S) \theta^{2}\left(d_{1}+d_{2}+d_{3}+d_{4}+2 d_{5}+d_{6}+d_{7}\right)}{S\left(d_{1}+d_{2}+d_{3}+d_{4}+2 d_{5}+d_{6}+d_{7}\right)}}
$$

Fuzzy optimal cycle length using Kuhn -tucker condition:
Suppose $L$ be a heptagonal fuzzy number $\tilde{L}=\left(L_{1}, L_{2}, L_{3}, L_{4}, L_{5}, L_{6}, L_{7}\right)$ with $0<L_{1} \leq L_{2} \leq L_{3} \leq L_{4} \leq L_{5} \leq L_{6} \leq L_{7}$.

The fuzzy total inventory cost is,

$$
T I C(L)=\left[\begin{array}{l}
\frac{1}{L_{7}}\left(O_{1}+\frac{(h+S) \theta^{2} d_{1}}{2}\right)+\frac{S d_{1} L_{1}}{2}+(C-\theta S) d_{1}, \\
\frac{1}{L_{6}}\left(O_{2}+\frac{(h+S) \theta^{2} d_{2}}{2}\right)+\frac{S d_{2} L_{2}}{2}+(C-\theta S) d_{2}, \\
\frac{1}{L_{5}}\left(O_{3}+\frac{(h+S) \theta^{2} d_{3}}{2}\right)+\frac{S d_{3} L_{3}}{2}+(C-\theta S) d_{3}, \\
\frac{1}{L_{4}}\left(O_{4}+\frac{(h+S) \theta^{2} d_{4}}{2}\right)+\frac{S d_{4} L_{4}}{2}+(C-\theta S) d_{4}, \\
\frac{1}{L_{3}}\left(O_{5}+\frac{(h+S) \theta^{2} d_{5}}{2}\right)+\frac{S d_{5} L_{5}}{2}+(C-\theta S) d_{5}, \\
\frac{1}{L_{2}}\left(O_{6}+\frac{(h+S) \theta^{2} d_{6}}{2}\right)+\frac{S d_{6} L_{6}}{2}+(C-\theta S) d_{6}, \\
\frac{1}{L_{1}}\left(O_{7}+\frac{(h+S) \theta^{2} d_{7}}{2}\right)+\frac{S d_{7} L_{7}}{2}+(C-\theta S) d_{7}
\end{array}\right]
$$

We defuzzify the total inventory cost using robust's ranking method.

$$
T I C(L)=\frac{1}{4}\left[\begin{array}{l}
\frac{1}{L_{7}}\left(O_{1}+\frac{(h+S) \theta^{2} d_{1}}{2}\right)+\frac{S d_{1} L_{1}}{2}+(C-\theta S) d_{1}+ \\
\frac{1}{L_{6}}\left(O_{2}+\frac{(h+S) \theta^{2} d_{2}}{2}\right)+\frac{S d_{2} L_{2}}{2}+(C-\theta S) d_{2}+ \\
\frac{1}{L_{5}}\left(O_{3}+\frac{(h+S) \theta^{2} d_{3}}{2}\right)+\frac{S d_{3} L_{3}}{2}+(C-\theta S) d_{3}+ \\
\frac{\frac{2}{L_{4}}\left(O_{4}+\frac{(h+S) \theta^{2} d_{4}}{2}\right)+\frac{S d_{4} L_{4}}{2}+(C-\theta S) d_{4}+}{}\left(O_{5}+\frac{(h+S) \theta^{2} d_{5}}{2}\right)+\frac{2 S d_{5} L_{5}}{2}+2(C-\theta S) d_{5}+ \\
\frac{1}{L_{2}}\left(O_{6}+\frac{(h+S) \theta^{2} d_{6}}{2}\right)+\frac{S d_{6} L_{6}}{2}+(C-\theta S) d_{6}+ \\
\frac{1}{L_{1}}\left(O_{7}+\frac{(h+S) \theta^{2} d_{7}}{2}\right)+\frac{S d_{7} L_{7}}{2}+(C-\theta S) d_{7}
\end{array}\right]
$$

Now differentiating partially w.r.t $L_{1}, L_{2}, L_{3}, L_{4}, L_{5}, L_{6}, L_{7}$ respectively and equate it to zero.

$$
\begin{aligned}
& \frac{\partial T I C(L)}{\partial L_{1}}=0 \\
& \frac{-1}{L_{1}^{2}}\left(O_{7}+\frac{(h+S) \theta^{2} d_{7}}{2}\right)+\frac{S d_{1}}{2}=0 \\
& L_{1}=\sqrt{\frac{2 O_{7}+(h+S) \theta^{2} d_{7}}{S d_{1}}}
\end{aligned}
$$

Similarly,
$\frac{\partial T I C(L)}{\partial L_{2}}=0 \Rightarrow L_{2}=\sqrt{\frac{2 O_{6}+(h+S) \theta^{2} d_{6}}{S d_{2}}}$
$\frac{\partial T I C(L)}{\partial L_{3}}=0 \Rightarrow L_{3}=\sqrt{\frac{2 O_{5}+(h+S) \theta^{2} d_{5}}{S d_{3}}}$
$\frac{\partial T I C(L)}{\partial L_{4}}=0 \Rightarrow L_{4}=\sqrt{\frac{2 O_{4}+(h+S) \theta^{2} d_{4}}{S d_{4}}}$
$\frac{\partial T I C(L)}{\partial L_{5}}=0 \Rightarrow L_{5}=\sqrt{\frac{2 O_{3}+(h+S) \theta^{2} d_{3}}{S d_{5}}}$
$\frac{\partial T I C(L)}{\partial L_{6}}=0 \Rightarrow L_{6}=\sqrt{\frac{2 O_{2}+(h+S) \theta^{2} d_{2}}{S d_{6}}}$
$\frac{\partial T I C(L)}{\partial L_{7}}=0 \Rightarrow L_{7}=\sqrt{\frac{2 O_{1}+(h+S) \theta^{2} d_{1}}{S d_{7}}}$
Next, the Kuhn Tucker condition is used to minimize the total cost and to find the solution of
$L_{1}, L_{2}, L_{3}, L_{4}, L_{5}, L_{6}, L_{7}$.
The Kuhn Tucker conditions are,

$$
\begin{aligned}
& \nabla f(T I C(L))-\lambda_{i} \nabla g\left(L_{i}\right) \\
& \lambda_{i} g_{i}\left(L_{i}\right)=0 \\
& g_{i}\left(L_{i}\right) \geq 0
\end{aligned}
$$

These above simplify to the following, $\lambda_{1}, \lambda_{2}, \lambda_{3}, \lambda_{4}, \lambda_{5}, \lambda_{6}, \lambda_{7} \leq 0$.

$$
\begin{aligned}
& \nabla f(T I C(L))-\lambda_{i} \nabla g\left(L_{i}\right) \Rightarrow \\
& \operatorname{TIC}(L)=\frac{1}{4}\left\{\left[\frac{1}{L_{h}}\left(O_{1}+\frac{(h+S) \theta^{2} d_{1}}{2}\right)+\frac{S d_{1} L_{1}}{2}+(C-\theta S) d_{1}\right]+\left[\frac{1}{L_{6}}\left(O_{2}+\frac{(h+S) \theta^{2} d_{2}}{2}\right)+\frac{S d_{2} L_{2}}{2}+(C-\theta S) d_{2}\right]\right. \\
& +\left[\frac{1}{L_{5}}\left(O_{3}+\frac{(h+S) \theta^{2} d_{3}}{2}\right)+\frac{S d_{3} L_{3}}{2}+(C-\theta S) d_{3}\right]+\left[\frac{1}{L_{4}}\left(O_{4}+\frac{(h+S) \theta^{2} d_{4}}{2}\right)+\frac{S d_{4} L_{4}}{2}+(C-\theta S) d_{4}\right]+ \\
& {\left[\frac{2}{L_{3}}\left(O_{5}+\frac{(h+S) \theta^{2} d_{5}}{2}\right)+\frac{2 S d_{5} L_{5}}{2}+2(C-\theta S) d_{5}\right]+\left[\frac{1}{L_{2}}\left(O_{6}+\frac{(h+S) \theta^{2} d_{6}}{2}\right)+\frac{S d_{6} L_{6}}{2}+(C-\theta S) d_{6}\right]+} \\
& \left.\left[\frac{1}{L_{1}}\left(O_{7}+\frac{(h+S) \theta^{2} d_{7}}{2}\right)+\frac{S d_{7} L_{7}}{2}+(C-\theta S) d_{7}\right]\right\}-\lambda_{1}\left(L_{2}-L_{1}\right)-\lambda_{2}\left(L_{3}-L_{2}\right)-\lambda_{3}\left(L_{4}-L_{3}\right) \\
& -\lambda_{4}\left(L_{5}-L_{4}\right)-\lambda_{5}\left(L_{6}-L_{5}\right)-\lambda_{6}\left(L_{7}-L_{6}\right)-\lambda_{7}\left(L_{1}\right)=0 \\
& \frac{1}{4}\left[\frac{1}{L_{7}}\left(O_{1}+\frac{(h+S) \theta^{2} d_{1}}{2}\right)+\frac{S d_{1} L_{1}}{2}+(C-\theta S) d_{1}\right]+\lambda_{1}-\lambda_{7}=0 \\
& \frac{1}{4}\left[\frac{1}{L_{6}}\left(O_{2}+\frac{(h+S) \theta^{2} d_{2}}{2}\right)+\frac{S d_{2} L_{2}}{2}+(C-\theta S) d_{2}\right]-\lambda_{1}+\lambda_{2}=0 \\
& \frac{1}{4}\left[\frac{1}{L_{5}}\left(O_{3}+\frac{(h+S) \theta^{2} d_{3}}{2}\right)+\frac{S d_{3} L_{3}}{2}+(C-\theta S) d_{3}\right]-\lambda_{2}+\lambda_{3}=0 \\
& \frac{1}{4}\left[\frac{1}{L_{4}}\left(O_{4}+\frac{(h+S) \theta^{2} d_{4}}{2}\right)+\frac{S d_{4} L_{4}}{2}+(C-\theta S) d_{4}\right]-\lambda_{3}+\lambda_{4}=0 \\
& \frac{1}{4}\left[\frac{2}{L_{3}}\left(O_{5}+\frac{(h+S) \theta^{2} d_{5}}{2}\right)+\frac{2 S d_{5} L_{5}}{2}+2(C-\theta S) d_{5}\right]-\lambda_{4}+\lambda_{5}=0 \\
& \frac{1}{4}\left[\frac{1}{L_{2}}\left(O_{6}+\frac{(h+S) \theta^{2} d_{6}}{2}\right)+\frac{S d_{6} L_{6}}{2}+(C-\theta S) d_{6}\right]-\lambda_{5}+\lambda_{6}=0 \\
& \frac{1}{4}\left[\frac{1}{L_{1}}\left(O_{7}+\frac{(h+S) \theta^{2} d_{7}}{2}\right)+\frac{S d_{7} L_{7}}{2}+(C-\theta S) d_{7}\right]-\lambda_{6}=0
\end{aligned}
$$

$$
\begin{aligned}
& \lambda_{1}\left(L_{2}-L_{1}\right)=0 \\
& \lambda_{2}\left(L_{3}-L_{2}\right)=0 \\
& \lambda_{3}\left(L_{4}-L_{3}\right)=0 \\
& \lambda_{4}\left(L_{5}-L_{4}\right)=0 \\
& \lambda_{5}\left(L_{6}-L_{5}\right)=0 \\
& \lambda_{6}\left(L_{7}-L_{6}\right)=0 \\
& \lambda_{7} L_{1}=0
\end{aligned}
$$

If $L_{1}>0$ and $\lambda_{7} L_{1}=0$ then $\lambda_{7}=0$. If $\lambda_{1}=\lambda_{2}=\lambda_{3}=\lambda_{4}=\lambda_{5}=\lambda_{6}=\lambda_{7}=0$ then
$L_{7}<L_{6}<L_{5}<L_{4}<L_{3}<L_{2}<L_{1} \cdot$ It doesn't satisfy the constrains
$0<L_{1} \leq L_{2} \leq L_{3} \leq L_{4} \leq L_{5} \leq L_{6} \leq L_{1}$.
$\therefore L_{2}=L_{1}, L_{3}=L_{2}, L_{4}=L_{3}, L_{5}=L_{4}, L_{6}=L_{5}, L_{7}=L_{6}$
i.e., $L_{1}=L_{2}=L_{3}=L_{4}=L_{5}=L_{6}=L_{7}=\tilde{L}^{*}$
$\tilde{L}^{*}=\sqrt{\frac{2\left(O_{1}+O_{2}+O_{3}+O_{4}+2 O_{5}+O_{6}+O_{7}\right)+(h+S) \theta^{2}\left(d_{1}+d_{2}+d_{3}+d_{4}+2 d_{5}+d_{6}+d_{7}\right)}{S\left(d_{1}+d_{2}+d_{3}+d_{4}+2 d_{5}+d_{6}+d_{7}\right)}}$

## 5. Numerical Example

## Crisp Model

Let us consider $\mathrm{O}=\$ 300$ per cycle, $\mathrm{h}=\$ 1.5$ per unit, $\mathrm{C}=\$ 25$ per unit, $\mathrm{S}=\$ 5$ per unit, $\mathrm{d}=500$ units, $\theta=5.46$.

| Method | L | TIC(L) |
| :---: | :---: | :---: |
| GP | 6.24 | 14461.51 |
| Kuhn tucker | 6.24 | 14461.51 |

Graphical representation of the above table is given by,


## Fuzzy model

Let us consider $\mathrm{h}=\$ 1.5$ per unit, $\mathrm{C}=\$ 25$ per unit, $\mathrm{S}=\$ 5$ per unit, $\theta=5.46$,
$\mathrm{O}=300=(50,75,100,150,175,200,275)$ and $\mathrm{D}=500=(100,150,200,250,300,325,375)$.

| Method | $\tilde{L}$ | TIC $(L)$ |
| :---: | :---: | :---: |
| GP | 6.24 | 14461.51 |
| Kuhn tucker | 6.24 | 14461.51 |

Graphical representation of the above table is given by,


When d increases then its optimal solution TIC (L) are also increases as follows.

| D | Method | L | TIC(L) |
| :---: | :---: | :---: | :---: |
| 500 | GP | 6.24 | 14461.51 |
|  | Kuhn tucker | 6.24 | 14461.51 |
| 600 | GP | 6.24 | 17344.20 |
|  | Kuhn tucker | 6.24 | 17344.20 |
| 700 | GP | 6.24 | 20226.89 |
|  | Kuhn tucker | 6.24 | 20226.89 |
| 800 | GP | 6.24 | 23109.57 |
|  | Kuhn tucker | 6.24 | 23109.57 |
| 900 | GP | 6.24 | 25992.26 |

Graphical representation of the above table is given by,



## 6. Conclusion

In this paper GP technique and Kuhn-tucker conditions for finding the optimal solutions to minimize the total cost and maximize the total profit of fuzzy inventory model in which the ordering cost and demand of the product are taken as heptagonal fuzzy numbers and solved by robust ranking defuzzification method. The results of these proposed models, crisp sense is very close to fuzzy sense. Finally, numerical example is found to illustrate the proposed models.

## References:

[1] Beightler C.S, and Phillips D.T, Applied Geometric programming, John Wiley and Sons, New York., 1976.
[2] Benkherouf.L, Skouri .K, and Konstantaras .I, Inventory decisions for a finite horizon problem with product substitution options and time varying demand, Applied mathematical modeling. 51 (2017) 669-685.
[3] Burwell T. H., Dave, D. S., Fitzpatrick, K. E., \& Roy, M. R. (1997). Economic lot size model for price-dependent demand under quantity and freight discounts. International Journal of Production Economics, 48, 141- 155.
[4] Cao. B. Y (1993) Extended fuzzy geometric programming. The Journal of Fuzzy Mathematics 2: 285-293
[5] Chang H.-C. (2004). An application of fuzzy sets theory to the EOQ model with imperfect quality items. Computers \& Operations Research, 31, 2079-2092.
[6] Mangasarian O.L, Nonlinear Programming, McGraw-Hill, New York, 1969.
[7] McCormick G.P, Nonlinear Programming: Theory, Algorithms, and Applications, Wiley, New York, 1983.
[8] Verma, R.K., "Fuzzy Geometric Programming with several objective functions", Fuzzy Sets, and Systems, 35 (1990) 115-120.
[9] Yang J H, Cao B Y (2005) Geometric programming with fuzzy relation equation constraints. Proceedings of the IEEE International Conference on Fuzzy Systems: 557-560
[10] Yao, J. S., \& Chiang, J. (2003). Inventory without backorder with fuzzy total cost and fuzzy storing cost defuzzified by centroid and signed distance. European Journal of Operational Research, 148, 401-409.
[11] You S.P, Inventory policy for products with price and time-dependent demands, Journal of the Operation Research Society. 56(2005) 870-873.
[12] Zadeh, L.A., Fuzzy Sets, Information and Control, 8 (1965) 338-353. [11] Zener, C., Engineering Design by Geometric Programming, Wiley, 197
[13] Zimmermann H.J, Fuzzy set theory and its applications, 2nd ed. Kluwer Academic publishers, Dordrecht-Boston, 1990.

