# A Comparative Study of Fuzzy Optimization through Fuzzy Number 

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(Received on August 1, 2021; Accepted on August 14, 2021)


#### Abstract

This paper presents a model for analyzing fuzzy reliability of the system using fuzzy number mathematical operation, where a triangular fuzzy number represents the stability of each computer system. Because the method is using simple fuzzy mathematical operation of fuzzy numbers instead of difficult arithmetic and logical of intervals. We are introducing a rule-based approach for estimating precisely all the broadest collections of blurry contextual disparities, considered an expansion to fuzzy relational coefficients, fuzzy relational inequalities can be extended to other fuzzy logic areas.


Keywords- Fuzzy optimization; triangular fuzzy number; fuzzy logic.

## 1. Introduction

Operational research is already properly approved as applicable to the strengthening optimization. LPP help to make efficient use successful mechanisms (Chandrawat et al., 2017). This offers more resources to meet the current environment (Singh and Dhiman, 2017). It's dealing with the big issue of highest profit. Optimization is an analysis technique used for optimization. Optimization is a mathematical method in use for objective function (Singh and Dhiman, 2018). It is a scientific theory to sales management resolving that surfaced during the Second World War, when maximize the operation of a number of variables based on a set of difficulties (Kandel and Byatt, 1978). Optimization is a theoretical method for solving practical and functional problem of a linear equation, in which the function is limited inequalities. The real image of the LPP can be described by;

$$
\operatorname{Maximum} \Omega=\sum_{j=1, i=1}^{m} A^{*} y_{j}
$$

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Subject to

$$
\sum_{j=1, i=1}^{n, m} C_{j i}{ }^{*} y_{i} \leq d^{*} \forall y_{i} \geq 0
$$

Fuzzy relationships, which are often found in various fields, such as mathematics, decisionmaking, and cluster analysis (Zadeh, 1965), are specific circumstances of L-relations where L is the standard sequence $[0,1]$ (Rasheed and Sarhan, 2019).

Fuzzy logic is already adjusted for different areas; from linear programming to Al., it has already been required to assist the computer to decide the variations between data that are not correct nor incorrect (Negoiţă and Ralescu, 1975). If there is anything measurements of the human reasoning process such as darkness, some brightness etc. Fuzzy sets extend modernist sets, as the measurement processes (aka unique systems) of classical sets are specific cases of the objective function of fuzzy sets, if maybe it only takes values 0 or 1 . Classical ions sets are commonly referred to as triangular fuzzy numbers in fuzzy logic (Mendel, 2003). The fuzzy number principle are used in a broad range of domains, like bioinformatics, that the knowledge is incorrect or imperfect (Liu et al., 2021).

Fuzzy logic is a type of multi-valued theory in which variable realized can be any real number between 0 and 1 including of it. It is used to resolve the theory of false statement, in which the result of truth can vary from truly true to completely false. In comparison, in Boolean logic, only integer values 0 or 1 can be the truth values of variables (Kumar et al., 2021).

## 2. Literature Review

Within the general Linear programming problem, the variables stay unchanged but we may assume that such quantity of limitation stays unchanged, however if the price of the rise or drop below a certain period in time, so this specific linear programming problems is known the uncertain Linear programming problems its member rank for the steadily increased value is implemented by fuzzy linear programming problems methods. This participation role consists can achieve best outcomes within the bottom and top limit scope of the linear programming problems. So, our initial issue would be turned in to equal, crisp query. After that, a crisp deal only with word that is defined so there is no fuzziness. Fuzzy suggests a loss of clearness and any idea that is unclear in every way. For example, the level of satisfaction for the fantasy character might be different in different study social characters. Usually, in the mathematical logic, we research the characteristics of the sets or the items that do not relate to the set but to the fuzzy set. For e.g., whenever the characteristic relates to the group, people classify membership degree 1 to it, if not 0 because in fuzzy logic, researchers can attach standard approach between 0 and 1. Zadeh (Zadeh, 1965) first decided to introduce the fuzzy logic idea. If this fuzzy logic is usable to linear decision making, then FLP came into being. Through using FLP, we can calculate improvements in other decision variables. There are many everyday situations requests from fuzzy linear programming, like business and technology online methods. The nature of optimization techniques and FLP some researchers have developed methods for solving this problem in some modern system using intuitive fuzzy number, and LPP operations have been implemented. Throughout this paper, we use the hexagonal fuzzy number linear programming problems to come to terms with either the predictive increase (pi) throughout the essential quality (bi) of classical optimization and discuss the results with the specified membership degree. Triangle
fuzzy sets can be used to interpret feasible uncertainties and missing data in decision-making, probability of the risk and optimization techniques (Türk, 2021).

## 3. Preliminaries

## Definition 1

A $\alpha$-cut of a fuzzy group of X is the set where the membership values in X , are higher than or equal to $\alpha$, is indicated by $\tilde{X}^{\alpha}$

$$
\begin{equation*}
\tilde{X}^{\alpha}=\left\{U \mid \lambda_{X}(U) \geq \alpha, u \in U\right\} \tag{2}
\end{equation*}
$$

## Definition 2

Let U be the universal space and a Fuzzy set $\tilde{X}$, a set in which each element of the set U is linked to a membership grade specified as:

$$
\begin{equation*}
\tilde{\mathrm{X}}=\left\{\left(\mathrm{U}, \lambda_{\mathrm{x}}(U)\right): u \in U, \lambda_{X}(U) \rightarrow[01]\right\} \tag{3}
\end{equation*}
$$

## Definition 3

A powerful $\alpha$-cut of a fluffy set of U is the set in which the membership values in X , are higher than $\alpha$, is indicated by $\tilde{X}^{\alpha+}$

$$
\begin{equation*}
\tilde{X}^{\alpha+}=\left\{U \mid \lambda_{X}(U) \succ \alpha, u \in U\right\} \tag{4}
\end{equation*}
$$

## Definition 4

The height of a fuzzy set denoted by $\mathrm{h}(\mathrm{X})$ is defined as the elements contained in that set's largest membership values.

## Definition 5

If a fuzzy set $h(X)=1$ then fuzzy set $A$ is called normal.

## Definition 6

If the given condition is satisfied then the set is called convex fuzzy set.

$$
\begin{equation*}
\lambda_{X}\left\{\mu u_{1}+(1-\mu) u_{2}\right\} \geq \min \left\{\lambda_{X}\left(u_{1}\right), \lambda_{X}\left(u_{2}\right)\right\}, \text { where } 0 \leq \mu \leq 1, u_{1}, u_{2} \in U \tag{5}
\end{equation*}
$$

## Definition 7

The fuzzy number $\mathrm{P}=\langle p, q, r\rangle$ are called triangular fuzzy number, if the membership function $\lambda_{B}$ are defined

$$
\begin{gather*}
\lambda_{\mathrm{B}(\mathrm{x})}=\left\{\begin{array}{c}
\frac{\mathrm{y}-\mathrm{b}}{\mathrm{a}-\mathrm{b}} ; \mathrm{b} \leq \mathrm{y}<a \\
1 ; \mathrm{y}=\mathrm{a} \\
\frac{\mathrm{~d}-\mathrm{y}}{\mathrm{~d}-\mathrm{a}} ; \mathrm{a} \leq \mathrm{y}<d \\
0 ; \text { otherwise }
\end{array}\right\}  \tag{6}\\
B_{\alpha=}[b, d]=\left[(a-b)_{\alpha}+b,-(d-a)_{\alpha}+d\right] \tag{7}
\end{gather*}
$$

## 4. Mathematics Operation

It is a fundamental method of mathematics operation; that is addition, subtraction, multiplication and division.

Let P and Q are two fuzzy numbers.

$$
\begin{equation*}
P=\left[p_{1}, p_{3}\right] Q=\left[q_{1}, q_{3}\right] \in R \tag{8}
\end{equation*}
$$

(a) Addition of Two Fuzzy Set

$$
\begin{gather*}
P+Q=\left[p_{1}, q_{3}\right]+\left[q_{1}, q_{3}\right]  \tag{9}\\
{\left[p_{1}+q_{1}, q_{3}+p_{3}\right]}
\end{gather*}
$$

(b) Subtraction of Two Fuzzy Set

$$
\begin{gather*}
P-Q=\left[p_{1}, p_{3}\right]-\left[q_{1}, q_{3}\right]  \tag{10}\\
{\left[p_{1}-q_{1}, p_{3}-q_{3}\right]}
\end{gather*}
$$

(c) Multiplication of Two Fuzzy Set

$$
\begin{gather*}
P(.) Q=\left[p_{1} q_{1} \& p_{1} q_{3} \& p_{3} q_{1} \& p_{3} q_{3}\right]  \tag{11}\\
{\left[p_{1} q_{1} \operatorname{orp}_{1} q_{3} \operatorname{orp}_{3} q_{1} \operatorname{orp}_{3} q_{3}\right]}
\end{gather*}
$$

(d) Division of Two Fuzzy Set

$$
\begin{align*}
P / Q= & {\left[p_{1} / q_{1} \& p_{1} / q_{3} \& p_{3} / q_{1} \& p_{3} / q_{3}\right] }  \tag{12}\\
& {\left[p_{1} / q_{1} \text { or } p_{1} / q_{3} \text { or } p_{3} / q_{1} \text { or } p_{3} / q_{3}\right] }
\end{align*}
$$

## 5. Mathematical Modeling

The graphical LPP methodology uses the maximum or minimum critical points of the objective function line (Torra, 2010) and the possible regions to solve problem the graphical methodology is used for solving LPP involving two variable decision $\mathrm{y}, \mathrm{x}$ and instead of $y_{1}, y_{2}$

### 5.1 Classical Linear Programming

The problem with classical Linear Programming is the finding under limitations described by Linear equations or inequities.

$$
\begin{gather*}
\text { Maximum/Minimum } d_{1} y_{1}+d_{2} y_{2}+\ldots+d_{m} y_{m}  \tag{13}\\
b_{11} y_{1}+b_{12} y_{2}+\ldots+b_{1 m} y_{m} \leq a_{1} \\
b_{21} y_{1}+b_{22} y_{2}+\ldots+b_{1 m} y_{m} \leq a_{2} \\
\ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \\
b_{n 1} y_{1}+b_{n 2} y_{2}+\cdots+d_{n 2 m} y_{m} \leq a_{n} \\
y_{1} y_{2}, \geq 0
\end{gather*}
$$

The maximized or minimized variable is known as objective function. Let us assume it is $\beta$. The expression of the question can be completed to use this symbol.

$$
\begin{gather*}
\text { Minimum } \beta=d_{y}  \tag{14}\\
\beta_{y} \leq a \\
y \geq 0
\end{gather*}
$$

### 5.2 Fuzzy linear programming

The following forms are the most particular form of linear fuzzy programming.

$$
\begin{equation*}
\text { Maximum } \sum_{i=1}^{m} D_{i} Y_{i} \tag{15}
\end{equation*}
$$

Subject to

$$
\begin{gathered}
\sum_{i=1}^{m} B_{j i} Y_{i} \leq A_{i}\left(j \in M_{m}\right) \\
y_{i} \geq 0\left(\mathrm{i} \in M_{m}\right)
\end{gathered}
$$

Case 1 Fuzzy LPP where on the right side only.

$$
\begin{equation*}
\sum_{i=1}^{m} D_{i} Y_{i} \tag{16}
\end{equation*}
$$

Subject to

$$
\begin{gathered}
\sum_{i=1}^{m} b_{j i} Y_{i} \leq A_{j}\left(j \in M_{m}\right) \\
y_{i} \geq 0\left(\mathrm{i} \in M_{n}\right)
\end{gathered}
$$

Case 2 Fuzzy LPP where on the right side just the function is the limits matrix the fuzzy numbers.

$$
\begin{equation*}
\sum_{i=1}^{m} d_{i} y_{i} \tag{17}
\end{equation*}
$$

Subject to

$$
\begin{gathered}
\sum_{i=1}^{m} B_{j i} Y_{i} \leq A_{i}\left(j \in M_{n}\right) \\
y_{i} \geq 0\left(\mathrm{i} \in M_{m}\right)
\end{gathered}
$$

Usually, fuzzy LPP will first be translated into analogous simple or anti liner problems which will then be solved by usual methods the final results of fuzzy LPP is therefore real values representing a solution in the form of the fuzzy numbers used. Here is the usual $A_{i(y)}$ from was used for $A_{j}\left(j \in M_{m}\right)$ fuzzy number.

$$
A_{i}(y)=\left\{\begin{array}{c}
1 \text { when } y \leq a_{j}  \tag{18}\\
\frac{a_{j}+q_{j}-y}{q_{j}} \text { when } a_{j}<y<a_{j}+q_{j} \\
0 \text { when } a_{j}+q_{j} \leq y_{1}
\end{array}\right\}
$$

In which $y \in S$ we next measure the degree of $\beta_{j(y)}$ by which y approaches the $i^{t h}$ limit with (i $\in$ $\left.M_{n}\right)$ the method for any vector. $y=y_{1} y_{2} \ldots, y_{n}$.

$$
\begin{equation*}
C_{j}(Y)=A_{j} \sum_{i=1}^{m} b_{j i} Y_{i} \tag{19}
\end{equation*}
$$

This is a fuzzy set on $S^{m}$ and their junction $\bigcap_{j=1}^{n} D$ is a fuzzy set to function the fuzzy set ofoptimal value is now calculated. Last it is achieved by first measuring the high and low limits of the optimal values $\beta_{k}$. By solving the classical linear programming problems and can be achieved by trying to reduce the optical values.

$$
\begin{equation*}
\text { Maximum } \beta=d_{y} \tag{20}
\end{equation*}
$$

Subject to

$$
\begin{gathered}
\sum_{i=1}^{m} B_{j i} Y_{i} \leq A_{i}\left(j \in M_{n}\right) \\
y_{i} \geq 0\left(\mathrm{i} \in M_{m}\right.
\end{gathered}
$$

A similar linear programming problem occurs in the upper limit for the best possible value $\beta_{k}$ in which each is substituted by $\left(a_{j}+q_{j}\right)$.

$$
\begin{equation*}
\operatorname{Max} \beta=d_{y} \tag{21}
\end{equation*}
$$

Subject to

$$
\begin{gathered}
\sum_{i=1}^{m} B_{j i} Y_{i} \leq A_{i}\left(j \in M_{n}\right) \\
y_{i} \geq 0\left(\mathrm{i} \in M_{m}\right)
\end{gathered}
$$

The fuzzy set of feasible variables H is then defined by the fuzzy $S^{m}$ subset.

$$
\mathrm{H}(\mathrm{Y})=\left\{\begin{array}{c}
1 \text { when } y \leq a_{j}  \tag{22}\\
\frac{d_{j}+\beta_{k}}{\beta_{v}-\beta_{k}} \text { when } \beta_{k} \leq D_{y} \leq \beta_{v} \\
0 \text { when } D_{Y} \leq \beta_{k}
\end{array}\right\}
$$

The problems have the standard problems of optimizing.

$$
\begin{equation*}
\Upsilon\left(\beta_{k}-z_{k}\right)-d_{y} \leq-\beta_{k} \tag{23}
\end{equation*}
$$

Subject to

$$
\begin{gathered}
\gamma_{\mathrm{qk}}+\sum_{i=1}^{m} B_{j i} Y_{i} \leq a_{j}+q_{j}\left(j \in M_{n}\right) \\
\Upsilon, y_{i} \geq 0\left(\mathrm{i} \in M_{m}\right)
\end{gathered}
$$

### 5.3 Fuzzy Optimal Solution to Fully Fuzzy Linear Programming Problems

$$
\begin{equation*}
\operatorname{Max} / \operatorname{Min} \tilde{Z}=\sum_{i=1}^{\dot{m}} \tilde{\Omega}^{T} \otimes \tilde{Y}_{i} \tag{24}
\end{equation*}
$$

Subject to

$$
\begin{gathered}
\tilde{B} \otimes \tilde{Y}=\tilde{a}, \\
\tilde{Y} \text { is a non }- \text { negative fuzzy number. }
\end{gathered}
$$

The proposed method takes the following steps:

## STEP-1

Putting

$$
\begin{aligned}
& \tilde{\Omega}^{T}=\left[\tilde{\Omega}_{i}\right]_{1 \times m}, \tilde{Y}=\left[\tilde{y}_{i}\right]_{m \times 1}, \tilde{B}=\left[\tilde{b}_{j i}\right]_{n \times m}, \tilde{a}=\left[\tilde{a}_{j}\right]_{n \times 1} \\
& \text { weget } \\
& \quad \operatorname{Max} \backslash \operatorname{Min}\left(\sum_{i=1}^{m} \tilde{\Omega}_{i} \otimes \tilde{y}_{i}\right) \\
& \text { subject to } \sum_{i=1}^{m} \tilde{b}_{j i} \otimes \tilde{y}_{i}=\tilde{a}_{j} \forall i \in(N) \\
& \tilde{y}_{i} \text { is the fuzzy number. }
\end{aligned}
$$

## STEP 2

If all the limits $\tilde{c}_{n}, \tilde{b}_{n}, \tilde{a}_{n}, \&, \tilde{x}_{m n}$ is showed by triangular fuzzy number $\left(\mathrm{f}_{\mathrm{n}}, \mathrm{g}_{\mathrm{n}}, \mathrm{h}_{\mathrm{n}}\right)$, $\left(\mathrm{a}_{\mathrm{n}}, \mathrm{b}_{\mathrm{n}}, \mathrm{c}_{\mathrm{n}}\right),\left(\mathrm{x}_{\mathrm{mn}}, \mathrm{y}_{\mathrm{mn}}, \mathrm{z}_{\mathrm{mn}}\right)$, and $\left(p_{n}, q_{n}, r_{n}\right)$ showed, then the FFLP problem.

$$
\begin{equation*}
\operatorname{Max}(\text { or } \operatorname{Min}) \mathrm{R}\left(\sum_{n 1}^{j}\left(f_{n}, g_{n}, h_{n}\right) \otimes\left(a_{n}, b_{n}, c_{n}\right)\right) \tag{26}
\end{equation*}
$$

Subject to $\sum_{n 1}^{j}\left(a_{n}, b_{n}, c_{n}\right) \otimes\left(x_{m n}, y_{m n}, z_{m n}\right)=\left(p_{j}, q_{j}, r_{j}\right) \forall i=$ natural number $(N)$ $\left(x_{n}, y_{n}, z_{n}\right)$ is a postive triangular $f u z z y$ number.

## STEP 3

$\operatorname{Consider}\left(x_{m n}, y_{m n}, z_{m n}\right) \otimes\left(a_{n}, b_{n}, c_{n}\right)=\left(m_{m n}, n_{m n}, o_{m n}\right)$

$$
\begin{equation*}
\text { Maximize }(\text { or Minimize }) \notin \mathrm{R}\left(\sum_{n=1}^{j}\left(f_{n}, g_{n}, \mathrm{~h}_{n}\right) \otimes\left(a_{n}, b_{n}, \mathrm{c}_{n}\right)\right) \tag{27}
\end{equation*}
$$

Subject to

$$
\begin{gathered}
\sum_{\substack{n=1 \\
\left(a_{n}, b_{n}, \mathrm{c}_{n}\right)}}^{j}\left(m_{\mathrm{mn}}, n_{\mathrm{mn}}, \mathrm{o}_{\mathrm{mn}}\right)=\left(p_{n}, q_{n}, \mathrm{r}_{n}\right) \forall i=N \\
\text { a positive triangular fuzzy number. }
\end{gathered}
$$

## STEP 4

$$
\begin{equation*}
\operatorname{Max}(\text { or min }) \mathrm{R}\left(\sum_{j=1}^{n}\left(f_{j}, g_{j}, h_{j}\right) \otimes\left(a_{j}, b_{j}, c_{j}\right)\right) \tag{28}
\end{equation*}
$$

Subject to

$$
\begin{gathered}
\sum_{n=1}^{j} m_{\mathrm{mn}}=f_{i}, \forall i=N \\
\sum_{n=1}^{j} m_{\mathrm{mn}}=f_{i}, \forall i=N \\
\sum_{n=1}^{j} m_{\mathrm{mn}}=f_{i}
\end{gathered}
$$

### 5.4 S, L, R METHOD

Suppose all the fuzzy number is triangular. We can write $\mathrm{B}=\langle s, l, r\rangle$

$$
\begin{equation*}
\text { Maximum } \sum_{i=1}^{m} d_{i} y_{i} \tag{29}
\end{equation*}
$$

Subject to

$$
\begin{gathered}
\sum_{i=1}^{m}\left\langle s_{j i} l_{j i} r_{j i}\right\rangle y_{j i} \leq\left\langle u_{j}, l_{j i} r_{j i}\right\rangle\left\langle j \in M_{n}\right\rangle \\
y_{i} \geq o\left\langle i \in M_{n}\right\rangle
\end{gathered}
$$

Where $B_{j i}=\left\langle s_{j i}, l_{j i}, r_{j i}\right\rangle \& A_{j i}=\left\langle u_{j}, v_{j}, w_{j}\right\rangle$ is fuzzy number. $\mathrm{B} \leq A$ iff and only if
Maximum $(\mathrm{B}, \mathrm{A})=\mathrm{A}$. it proved for any 2 triangular fuzzy numbers.
$\mathrm{B}=\left\langle s_{2}, l_{2}, r_{2}\right\rangle$ and $A=\left\langle s_{1}, l_{1}, r_{1}\right\rangle, \mathrm{B} \leq A$
if and only if $s_{2} \leq s_{1}, s_{2}-l_{2} \leq s_{1}-l_{1}$ and $s_{2}+r_{2} \leq s_{1}+r_{1}$

$$
\left\langle s_{2}, l_{2}, r_{2}\right\rangle+\left\langle s_{1}, l_{1}, r_{1}\right\rangle=\left\langle s_{2}+s_{1}, l_{2}+l_{1}, r_{2}+r_{1}\right\rangle \&\left\langle s_{2}, l_{2}, r_{2}\right\rangle y=\left\langle s_{2 y}, l_{2 y}, r_{2 y}\right\rangle
$$

Where y is non- negative number.

$$
\begin{equation*}
\text { Maximum } \sum_{i=1}^{m} d_{i} y_{i} \tag{30}
\end{equation*}
$$

Subject to

$$
\begin{gathered}
\sum_{i=1}^{m} s_{j i} y_{i} \leq p_{j} \\
\left.\sum_{i=1}^{m} s_{j i}-l_{j i}\right) y_{i} \leq p_{j}-q_{j} \\
\left.\sum_{i=1}^{m} s_{j i}+l_{j i}\right) y_{i} \leq p_{j}+q_{j}\left\langle j \in M_{n}\right\rangle \\
y_{i} \geq o\left\langle i \in M_{n}\right\rangle
\end{gathered}
$$

## 6. Numerical experiment

Example 1 Let suppose that a registered firm is making only two products and assume it has name $A_{1}$ and $A_{2}$ having cost $\$ .30$ and $\$ .40$ per unit gain as shown in Figure 1. The each unit of product $A_{1}$ needs twice as many labor hours as each product $A_{2}$ as shown in Figure 2. The total availability of labour are 500 per day that too minimum and suppose it need to be extended up to 600 hours per day because of arranging some important overtime work as shown in Figure 3. The supplying of material is ok for 400 units for the both $A_{1}$ and $A_{2}$ product in a day but it can be possible to extend it 500 units in a day as per previous experiences as shown in Figure 4. The question is now how much units of product $A_{1}$ and $A_{2}$ could be made in a day to have maximum gain?

Solution: Let $y_{1}, y_{2}$ in a notation of the numbers of units of the products $A_{1}$ and $A_{2}$ produced in a day a solution can be by using FLPP given below.

$$
\begin{equation*}
\text { Maximum } \Omega=3 y_{1}+4 y_{2} \tag{31}
\end{equation*}
$$

Subject to

$$
\begin{gathered}
y_{1}+y_{2} \leq C_{1} \\
2 y_{1}+y_{2} \leq C_{2} \\
y_{1,} y_{2} \geq 0
\end{gathered}
$$

Where $C_{1}$ defined as

$$
C_{1}(y)=\left\{\begin{array}{c}
1 \text { when } y \leq 400  \tag{32}\\
\frac{500-y}{100} \text { when } 400<y \leq 500 \\
0 \text { when } 500<y
\end{array}\right\}
$$

Where $C_{2}$ is defined as

$$
C_{2}(y)=\left\{\begin{array}{c}
1 \text { when } y \leq 500  \tag{33}\\
\frac{600-y}{100} \text { when } 500<y \leq 600 \\
0 \text { when } 600<y
\end{array}\right\}
$$

Further solving we need upper limits and lower.

$$
\begin{gathered}
\Omega_{l}=130 \text { and } \Omega_{u}=160 \\
A_{1} \text { Maximum } \Omega=.3 y_{1}+.4 y_{2}
\end{gathered}
$$

Subject to

$$
\begin{aligned}
y_{1}+y_{2} & \leq 400 \\
2 y_{1}+y_{2} & \leq 500 \\
y_{1,} y_{2} & \geq 0
\end{aligned}
$$



Figure 1: Solution of equation number (34).

$$
\begin{equation*}
A_{2} \text { Maximum } \Omega=.3 y_{1}+.4 y_{2} \tag{35}
\end{equation*}
$$

Subject to


Figure 2: Solution of equation number (35).
Example 2 Suppose the following fully fuzzy programming problems and solved as given below.

$$
\begin{equation*}
\max (2,7,8) \otimes \widetilde{P_{1}} \oplus(1,4,9) \otimes \tilde{P}_{2} \tag{36}
\end{equation*}
$$

Subject to

$$
\begin{aligned}
& (1,4,5) \otimes \tilde{P}_{1} \oplus(2,3,4) \otimes \tilde{P}_{2}=(7,17,33) \\
& (-2,2,3) \otimes \tilde{P}_{1} \oplus(2,4,5) \otimes \tilde{P}_{2}=(2,18,34) \\
& \tilde{P}_{1}, \tilde{P}_{2} \text { arenon }- \text { negativefuzzynumber } .
\end{aligned}
$$

Solution: Consider $\tilde{P}_{1}=\left(P_{1}, Q_{1}, R_{1}\right)$ and $\tilde{P}_{2}\left(P_{2}, Q_{2}, R_{2}\right)$ then fully fuzzy linear programming problem be written below as.

$$
\operatorname{Max}(2,7,8) \otimes\left(P_{1}, Q_{1}, R_{1}\right) \oplus(1,4,9) \otimes\left(P_{2}, Q_{2}, R_{2}\right)
$$

Subject to

$$
\begin{aligned}
& (1,4,5) \otimes\left(P_{1}, Q_{1}, R_{1}\right) \oplus(2,3,4) \otimes\left(P_{2}, Q_{2}, R_{2}\right)=(7,17,33) \\
& (-2,2,3) \otimes\left(P_{1}, Q_{1}, R_{1}\right) \oplus(2,4,5) \otimes\left(P_{2}, Q_{2}, R_{2}\right)=(2,18,34) \\
& \left(P_{1}, Q_{1}, R_{1}\right) \text { and }\left(P_{2}, Q_{2}, R_{2}\right) \text { are non -negative fuzzy number. }
\end{aligned}
$$

Using step 3, the above FLLP problems.

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$$
\operatorname{Max} \mathrm{R}\left(2 P_{1}+P_{2}, 7 Q_{1}+4 Q_{2}, 8 R_{1}, 9 R_{2}\right)
$$

Subject to

$$
\begin{gather*}
\left(P_{1}+2 P_{2}, 4 Q_{1}+3 Q_{2}, 5 R_{1}, 4 R_{2}\right)=(7,17,33)  \tag{37}\\
\left(-2 P_{1}+2 P_{2}, 2 Q_{1}+4 Q_{2}, 3 R_{1}, 5 R_{2}=(2,18,34)\right. \tag{38}
\end{gather*}
$$

Using step 4

$$
\begin{gather*}
\operatorname{Max}\left(2\left(2 P_{1}+P_{2}\right)+\frac{1}{4}\left(7 Q_{1}+4 Q_{2}\right)+\left(8 R_{1}, 9 R_{2}\right)\right. \\
P_{1}+2 P_{2}=7  \tag{39}\\
7 Q_{1}+4 Q_{2}=17  \tag{40}\\
8 R_{1}+9 R_{2}=33  \tag{41}\\
-2 P_{1}+2 P_{2}=2  \tag{42}\\
2 Q_{1}+4 Q_{2}=18  \tag{43}\\
3 Q_{2}, 5 R_{1}=34 \tag{44}
\end{gather*}
$$



Figure 3: Solution of step number 4.
Example 3(Problem on Fully fuzzy linear programming problem)

$$
\begin{equation*}
\text { Maximum } \Omega=6 y_{1}+3 y_{2} \tag{45}
\end{equation*}
$$

Subject to

$$
\begin{gathered}
\langle 6,4,2\rangle y_{1}+\langle 7,5,3\rangle y_{2} \leq\langle 12,6,3\rangle \\
\langle 6,4,2\rangle y_{1}+\langle 3,2.5,3\rangle y_{2} \leq\langle 24,5,8\rangle \\
y_{1,} y_{2} \geq 0
\end{gathered}
$$

## Solution: Using SLR method

$$
\begin{equation*}
\text { Maximum } \Omega=6 y_{1}+3 y_{2} \tag{46}
\end{equation*}
$$

Subject to

$$
\begin{align*}
& 6 y_{1}+7 y_{2} \leq 12  \tag{47}\\
& 6 y_{1}+3 y_{2} \leq 24  \tag{48}\\
& 2 y_{1}+2 y_{2} \leq 6  \tag{49}\\
& 4 y_{1}+5 y_{2} \leq 19  \tag{50}\\
& 8 y_{1}+10 y_{2} \leq 15  \tag{51}\\
& 10 y_{1}+6 y_{2} \leq 32  \tag{52}\\
& y_{1,} y_{2} \geq 0 \tag{53}
\end{align*}
$$



Figure 4: Graphical solution of equation (45).

## 7. Conclusion

The benefits of Linear Programming System Linear programming always apply in cases at which parameters and optimal solution are static where they'll be represented as straight line variables. This approach will not be used in daily - life situations where parameters or parameter values are not linear. Things like complexity and resources are not taken into account. The benefits of Linear Programming System Linear programming always apply in cases at which parameters and optimal solution are static where they will be represented as straight line variables. This approach will not be used in real life situations where parameters or parameter values are not linear. Things like complexity and resources are not taken into account. The user's decision-making technique is

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becoming more objective but less subjective. Such devices will be in great supply in a factory, for instance, while others may make false claims idle for certain period.

## Conflict of Interest

The authors confirm that there is no conflict of interest to declare for this publication.

## Acknowledgments

This research did not receive any specific grant from funding agencies in the public, commercial, or not-for-profit sectors. The authors would like to thank the editor and anonymous reviewers for their comments that help improve the quality of this work.

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