INTUITIONISTIC FUZZY LINEAR FRACTIONAL PROGRAMMING PROBLEM USING DENOMINATOR OBJECTIVE RESTRICTION METHOD

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Abstract: The present paper describes for solving intuitionistic fuzzy linear fractional programming problem where subject to the constraints of right hand side is symmetric trapezoidal intuitionistic fuzzy number. To rank the symmetric trapezoidal intuitionistic fuzzy number by using ranking function. The denominator objective restriction method for finding an optimal solution to the intuitionistic fuzzy linear fractional programming problem. The proposed procedure is illustrated with a numerical example. To find the membership function and nonmembership function of \((\alpha, \beta)-\text{cut method by using MATLAB.}\)

AMS Subject Classification: 90C05, 90C30, 90C70, 90C90

Key Words: linear fractional programming problem, intuitionistic fuzzy set, intuitionistic fuzzy number, ranking function, denominator objective restriction method, intuitionistic fuzzy linear fractional programming problem

1. Introduction

In mathematical optimization, linear-fractional programming (LFP) is a generalization of linear programming (LP). Whereas the objective function in a

In this paper introduces a linear fractional programming problem with intuitionistic fuzzy number. The intuitionistic fuzzy linear fractional programming problem is that in which the objective function is the ratio of numerator and denominator. To find the intuitionistic fuzzy linear fractional programming
problem by using denominator objective restriction method. The denominator objective restriction method is based on the intuitionistic fuzzy simplex method.

This paper is organized as follows: In Section 2 provides the basic definitions and notations of IFS, IFN, TIFN, STIFN. In Section 3 some formulation of the problem are recalled. In Section 4 describes the algorithm for Denominator Objective Restriction Method. In Section 5, the proposed method is illustrated with a numerical example. In Section 6, finally follows the conclusion.

2. Preliminaries

**Definition 2.1.** (Intuitionistic Fuzzy Set) An Intuitionistic Fuzzy Set (IFS) A in X is defined as an object of the form $A = \{< x, \mu_A(x), \nu_A(x) >: x \in X\}$ where the functions $\mu_A : X \to [0, 1]$ and $\nu_A : X \to [0, 1]$ define the degree of membership and the degree of nonmembership of the element $x \in X$ respectively, and for every $x \in X$ in A, $0 \leq \mu_A(x) + \nu_A(x) \leq 1$ holds.

**Definition 2.2.** (Intuitionistic Fuzzy Number) An Intuitionistic Fuzzy Number (IFN) $\tilde{A}$ is:

(i) an intuitionistic fuzzy subset of the real line,

(ii) normal, that is, there is some $x_0 \in \mathbb{R}$ such that $\mu_{\tilde{A}}(x_0) = 1$, $\nu_{\tilde{A}}(x_0) = 0$,

(iii) convex for the membership function $\mu_{\tilde{A}}(x)$, that is

$$\mu_{\tilde{A}}(\lambda x_1 + (1 - \lambda)x_2) \geq \min(\mu_{\tilde{A}}(x_1), \mu_{\tilde{A}}(x_2)),$$

for every $x_1, x_2 \in \mathbb{R}$, $\lambda \in [0, 1]$,

(iv) concave for the nonmembership function $\nu_A(x)$, that is

$$\nu_{\tilde{A}}(\lambda x_1 + (1 - \lambda)x_2) \leq \max(\nu_{\tilde{A}}(x_1), \nu_{\tilde{A}}(x_2)),$$

for every $x_1, x_2 \in \mathbb{R}$, $\lambda \in [0, 1]$.

**Definition 2.3.** (Trapezoidal Intuitionistic Fuzzy Number) A Trapezoidal Intuitionistic Fuzzy Number (TIFN) $\tilde{A}$ is an IFS in $\mathbb{R}$ with membership function and nonmembership function as follows:

$$\mu_{\tilde{A}}(x) = \begin{cases} 
\frac{x-(a_1-\alpha)}{\alpha} & \text{for } x \in [a_1 - \alpha, a_1] \\
1 & \text{for } x \in [a_1, a_2] \\
\frac{a_2+\beta-x}{\beta} & \text{for } x \in [a_2, a_2 + \beta] \\
0 & \text{otherwise;}
\end{cases}$$
\[ v_{\tilde{A}}(x) = \begin{cases} \frac{a_1 - x}{\alpha} & \text{for } x \in [a_1 - \alpha', a_1] \\ 0 & \text{for } x \in [a_1, a_2] \\ \frac{x - a_2}{\beta'} & \text{for } x \in [a_2, a_2 + \beta'] \\ 1 & \text{otherwise} \end{cases} \]

where \( \tilde{A} = [a_1, a_2, \alpha, \beta; a_1, a_2, \alpha', \beta'] \).

**Definition 2.4.** (Symmetric Trapezoidal Intuitionistic Fuzzy Number) A TIFN is said to be STIFN if \( \alpha = \beta \) (say \( h \)) and \( \alpha' = \beta' \) (say \( h' \)). Hence the definition of STIFN is as follows:

An IFS \( \tilde{A} \) in \( \mathbb{R} \) is said to be a Symmetric Trapezoidal Intuitionistic Fuzzy Number (STIFN) if there exist real numbers \( a_1, a_2, h, h' \) where \( a_1 \leq a_2, h \leq h' \) and \( h, h' > 0 \) such that the membership and nonmembership functions are as follows:

\[ \mu_{\tilde{A}}(x) = \begin{cases} \frac{x -(a_1-h)}{h} & \text{for } x \in [a_1 - h, a_1] \\ 1 & \text{for } x \in [a_1, a_2] \\ \frac{a_2 + h - x}{h} & \text{for } x \in [a_2, a_2 + h] \\ 0 & \text{otherwise} \end{cases} \]

\[ v_{\tilde{A}}(x) = \begin{cases} \frac{a_1 - x}{h} & \text{for } x \in [a_1 - h', a_1] \\ 0 & \text{for } x \in [a_1, a_2] \\ \frac{x - a_2}{h'} & \text{for } x \in [a_2, a_2 + h'] \\ 1 & \text{otherwise} \end{cases} \]

where \( \tilde{A} = [a_1, a_2, h, h; a_1, a_2, h', h'] \).

### 3. Formulation of the Problem

#### 3.1. Linear Fractional Programming Problem

The Linear Fractional Programming Problem can be formulated as follows:

Given objective function

\[ Q(x) = \frac{P(x)}{D(x)} = \frac{\sum_{j=1}^{n} p_j x_j + p_0}{\sum_{j=1}^{n} d_j x_j + d_0}, \]  

(1)
which must be maximized (or minimized) subject to

\[
\begin{align*}
\sum_{j=1}^{n} a_{ij}x_j & \leq b_i, & i = 1, 2, \ldots, m_1, \\
\sum_{j=1}^{n} a_{ij}x_j & \geq b_i, & i = m_1 + 1, m_1 + 2, \ldots, m_2, \\
\sum_{j=1}^{n} a_{ij}x_j & = b_i, & i = m_2 + 1, m_2 + 2, \ldots, m.
\end{align*}
\]

(2)

\[x_j \geq 0, \quad j = 1, 2, \ldots, n_1,\]  

(3)

where \(m_1 \leq m_2 \leq m, n_1 \leq n\). Here and in what follows we suppose that \(D(x) \neq 0, \forall x = (x_1, x_2, \ldots, x_n) \in S\), where S denotes a feasible set or set of feasible solutions defined by constraints (2)-(3).

### 3.2. Fuzzy Linear Fractional Programming Problem

Fuzzy linear fractional programming problem is defined as:

\[
\text{Maximize } (\tilde{C}\tilde{X} + \alpha)/(\tilde{D}\tilde{X} + \beta) = \tilde{R}(\tilde{X}),
\]

subject to the constraints: \(\tilde{A}\tilde{X} \leq \tilde{b}\),

\(\tilde{X} \geq \tilde{0}\).

With additional assumption: the denominator is positive for all possible solutions.

### 3.3. Intuitionistic Fuzzy Linear Fractional Programming Problem

The intuitionistic fuzzy set theory is an extension of the fuzzy set theory by Atanassov [1]. The intuitionistic fuzzy linear fractional programming problem (IFLFPP) is a special type of fuzzy linear fractional programming problem (FLFPP). In real life, information available for certain situations is vague and such uncertainty is unavoidable. The possible solution is to consider the parameters involved as intuitionistic fuzzy data.

Consider the Intuitionistic Fuzzy Linear Fractional Programming Problem

\[
\begin{align*}
\text{Maximize: } & \tilde{Z}^I = (\tilde{C}^I\tilde{X}^I + \alpha)/(\tilde{D}^I\tilde{X}^I + \beta), \\
& \text{subject to the constraints: } \tilde{A}^I\tilde{X}^I \leq \tilde{b}^I, \\
& \tilde{X}^I \geq \tilde{0}^I.
\end{align*}
\]
3.4. Ranking Function

Let $\mathcal{R}$ be any linear ranking function. Then:

(i) $\tilde{A}^I \geq \tilde{B}^I$ if and only if $\tilde{A}^I - \tilde{B}^I \geq 0$ iff $-\tilde{B}^I \geq -\tilde{A}^I$.

(ii) If $\tilde{A}^I \geq \tilde{B}^I$ and $\tilde{C}^I \geq \tilde{D}^I$, then $\tilde{A}^I + \tilde{C}^I \geq \tilde{B}^I + \tilde{D}^I$.

Since there are many ranking functions for comparing IF numbers, a linear ranking function is introduced in this paper which is actually a ranking function $\mathcal{R}$ such that

$$\mathcal{R}(K\tilde{A}^I + \tilde{B}^I) = K\mathcal{R}(\tilde{A}^I) + \mathcal{R}(\tilde{B}^I).$$

Below a method of ranking STIFN is suggested which is actually a direct method for ranking classical intuitionistic fuzzy numbers.

$$\mathcal{R}(\tilde{A}^I) = a_1 + a_2 + \frac{1}{2}(h' - h),$$

where $\tilde{A}^I = [a_1, a_2, h, h'; a_1, a_2, h', h'] \in F(R)$.

4. Algorithm for Denominator Objective Restriction Method

**Step 1:** Consider the Intuitionistic Fuzzy Linear Fractional Programming Problem

Maximize: $\tilde{Z}^I = (\tilde{C}^I \tilde{X}^I + \alpha)/(\tilde{D}^I \tilde{X}^I + \beta)$,

subject to the constraints: $\tilde{A}^I \tilde{X}^I = \tilde{b}^I$,

$\tilde{X}^I \geq \tilde{0}^I$.

**Step 2:** Construct two single objective intuitionistic fuzzy linear programming problems namely, the problem (N) as well as the problem (D) from the given problem.

**Step 3:** Compute to the optimal solution to the problem (N) by means of the intuitionistic fuzzy simplex method. Introduce slack variables to convert the inequalities into equations and let the optimal solution to the problem (N) be $\tilde{X}^I_0$ and Max $\tilde{Z}^I(\tilde{X}^I_0) = \tilde{Z}^I_0$.

**Step 4:** Using the optimal table of the problem (N) as an initial intuitionistic fuzzy simplex table to the problem (D), continue to find a sequence of improved
basic feasible solutions \( \tilde{X}^I_n \) to the problem (D) and the value of \( \tilde{Z}^I \) at each of the improved basic feasible solution by the intuitionistic fuzzy simplex method.

**Step 5:** (a) If \( \tilde{Z}^I(\tilde{X}^I_k) \leq \tilde{Z}^I(\tilde{X}^I_{k+1}) \) for all \( k = 0, 1, 2, \ldots, n - 1 \) and \( \tilde{Z}^I(\tilde{X}^I_n) \geq \tilde{Z}^I(\tilde{X}^I_{n+1}) \) for some \( n \), stop the computation process and then go to Step 7.

**Step 6:** (b) If \( \tilde{Z}^I(\tilde{X}^I_k) \leq \tilde{Z}^I(\tilde{X}^I_{k+1}) \) for all \( k = 0, 1, 2, \ldots, n \) and \( \tilde{X}^I_{n+1} \) is an optimal solution to the problem (D) for some \( n \), stop the computation process and then go to Step 8.

**Step 7:** \( \tilde{X}^I_n \) is an optimal solution to the given problem and \( \text{Max} \tilde{Z}^I(\tilde{X}^I) = \tilde{Z}^I(\tilde{X}^I_n) \).

**Step 8:** \( \tilde{X}^I_{n+1} \) is an optimal solution to the given problem and \( \text{Max} \tilde{Z}^I(\tilde{X}^I) = \tilde{Z}^I(\tilde{X}^I_{n+1}) \).

5. Numerical Example

Consider the following IFLFP problem

Maximize: \( \tilde{Z}^I = \frac{6\tilde{x}_1^I + 3\tilde{x}_2^I + 6}{5\tilde{x}_1^I + 2\tilde{x}_2^I + 5} \),

subject to: \( 4\tilde{x}_1^I - 2\tilde{x}_2^I \leq [8, 11, 2, 2; 8, 11, 4, 4], \)
\( 3\tilde{x}_1^I + 5\tilde{x}_2^I \leq [10, 14, 4, 4; 10, 14, 6, 6], \)
\( \tilde{x}_1^I, \tilde{x}_2^I \geq 0. \)

The following two IFLFP problems can be obtained from the given problem:

(N) Maximize \( P(\tilde{X}^I) = 6\tilde{x}_1^I + 3\tilde{x}_2^I + 6 \)

subject to
\( 4\tilde{x}_1^I - 2\tilde{x}_2^I \leq [8, 11, 2, 2; 8, 11, 4, 4] \)
\( 3\tilde{x}_1^I + 5\tilde{x}_2^I \leq [10, 14, 4, 4; 10, 14, 6, 6] \)
and \( \tilde{x}_1^I, \tilde{x}_2^I \geq 0. \)

(D) Minimize \( Q(\tilde{X}^I) = 5\tilde{x}_1^I + 2\tilde{x}_2^I + 5 \)

subject to
\( 4\tilde{x}_1^I - 2\tilde{x}_2^I \leq [8, 11, 2, 2; 8, 11, 4, 4] \)
\( 3\tilde{x}_1^I + 5\tilde{x}_2^I \leq [10, 14, 4, 4; 10, 14, 6, 6] \)
and \( \tilde{x}_1^I, \tilde{x}_2^I \geq 0. \)

The standard form of the intuitionistic fuzzy linear programming problem becomes
\[ N \Rightarrow \text{Max } P(\tilde{X}^I) = 6\tilde{x}_1^I + 3\tilde{x}_2^I + 0\tilde{x}_3^I + 0\tilde{x}_4^I + 6 \]

subject to
\[ 4\tilde{x}_1^I - 2\tilde{x}_2^I + \tilde{x}_3^I = [8, 11, 2, 2; 8, 11, 4, 4] \]
\[ 3\tilde{x}_1^I + 5\tilde{x}_2^I + \tilde{x}_4^I = [10, 14, 4, 4; 10, 14, 6, 6] \]
\[ \tilde{x}_1^I, \tilde{x}_2^I, \tilde{x}_3^I, \tilde{x}_4^I \geq 0 \]

where \( \tilde{x}_3^I, \tilde{x}_4^I \) are the slack intuitionistic fuzzy variables.

The optimal solution to the problem (N), by the intuitionistic fuzzy simplex method, is given by the following table:

<table>
<thead>
<tr>
<th>Basic Variables</th>
<th>( C_I^j )</th>
<th>6</th>
<th>3</th>
<th>0</th>
<th>0</th>
</tr>
</thead>
<tbody>
<tr>
<td>( C_B^I )</td>
<td>( Y_B^I )</td>
<td>( X_B^I )</td>
<td>( \tilde{x}_1^I )</td>
<td>( \tilde{x}_2^I )</td>
<td>( \tilde{x}_3^I )</td>
</tr>
<tr>
<td>0</td>
<td>( \tilde{x}_3^I )</td>
<td>[8, 11, 2, 2; 8, 11, 4, 4]</td>
<td>4</td>
<td>-2</td>
<td>1</td>
</tr>
<tr>
<td>0</td>
<td>( \tilde{x}_4^I )</td>
<td>[10, 14, 4, 4; 10, 14, 6, 6]</td>
<td>3</td>
<td>5</td>
<td>0</td>
</tr>
</tbody>
</table>

Since there are some \( \tilde{P}_j^I - \tilde{C}_j^I < 0 \) the current basic feasible solution is not optimal.

Here \( \tilde{x}_1^I \) is entering variable and \( \tilde{x}_3^I \) is leaving variable. Here \( \tilde{x}_2^I \) is entering variable and \( \tilde{x}_4^I \) is leaving variable.

All \( \tilde{P}_j^I - \tilde{C}_j^I \geq 0 \). The current basic feasible solution is optimal.

Therefore, the optimal solution to the problem (N) is

\[ \tilde{x}_1^I = \begin{bmatrix} 30 & 83 & 9 & 9 & 30 & 83 & 16 & 16 \\ 13 & 26 & 13 & 13 & 13 & 26 & 13 \end{bmatrix} \quad \text{and} \quad \Re[\tilde{x}_1^I] = 5.7692 \]

\[ \tilde{x}_2^I = \begin{bmatrix} 8 & 23 & 5 & 5 & 8 & 23 & 6 & 6 \\ 13 & 26 & 13 & 13 & 13 & 26 & 13 \end{bmatrix} \quad \text{and} \quad \Re[\tilde{x}_2^I] = 1.5385 \]
Table 3

<table>
<thead>
<tr>
<th>Basic Variables</th>
<th>$C_I^l$</th>
<th>6</th>
<th>3</th>
<th>0</th>
<th>0</th>
</tr>
</thead>
<tbody>
<tr>
<td>$C_I^l$</td>
<td>$Y_I^l$</td>
<td>$X_I^l$</td>
<td>$\tilde{x}_1^l$</td>
<td>$\tilde{x}_2^l$</td>
<td>$\tilde{x}_3^l$</td>
</tr>
<tr>
<td>6</td>
<td>$\tilde{x}_1^l$</td>
<td>$[30, 83, 9, 9; 30, 83, 16, 16]$</td>
<td>1</td>
<td>0</td>
<td>$\frac{21}{13}$</td>
</tr>
<tr>
<td>3</td>
<td>$\tilde{x}_2^l$</td>
<td>$[30, 83, 9, 9; 30, 83, 16, 16]$</td>
<td>0</td>
<td>1</td>
<td>$\frac{21}{13}$</td>
</tr>
</tbody>
</table>

Max $P(\tilde{X}^l) = 45.2308$ and the value of Max $\tilde{Z}^l = 1.2250$.

Now, by Step 4 of the proposed method, the initial intuitionistic fuzzy simplex table to the problem (D) is given below:

The standard form of the intuitionistic fuzzy linear programming problem of D becomes

Max $Q(\tilde{X}^l) = -5\tilde{x}_1^l - 2\tilde{x}_2^l + 0\tilde{x}_3^l + 0\tilde{x}_4^l - 5$

subject to

$4\tilde{x}_1^l - 2\tilde{x}_2^l + \tilde{x}_3^l = [8, 11, 2, 2; 8, 11, 4, 4]$

$3\tilde{x}_1^l + 5\tilde{x}_2^l + \tilde{x}_4^l = [10, 14, 4, 4; 10, 14, 6, 6]$

$\tilde{x}_1^l, \tilde{x}_2^l, \tilde{x}_3^l, \tilde{x}_4^l \geq 0$

where $\tilde{x}_3^l, \tilde{x}_4^l$ are the slack intuitionistic fuzzy variables.

Here $\tilde{x}_3^l$ is entering variable and $\tilde{x}_1^l$ is leaving variable.

Table 4

<table>
<thead>
<tr>
<th>Basic Variables</th>
<th>$D^l_B$</th>
<th>$Y^l_B$</th>
<th>$X^l_B$</th>
<th>$\tilde{x}_1^l$</th>
<th>$\tilde{x}_2^l$</th>
<th>$\tilde{x}_3^l$</th>
<th>$\tilde{x}_4^l$</th>
<th>$\mathbb{R}$</th>
<th>$\theta$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$D^l_B$</td>
<td>$Y^l_B$</td>
<td>$X^l_B$</td>
<td>$\tilde{x}_1^l$</td>
<td>$\tilde{x}_2^l$</td>
<td>$\tilde{x}_3^l$</td>
<td>$\tilde{x}_4^l$</td>
<td>$\mathbb{R}$</td>
<td>$\theta$</td>
<td></td>
</tr>
<tr>
<td>-5</td>
<td>$\tilde{x}_1^l$</td>
<td>$[30, 83, 9, 9; 30, 83, 16, 16]$</td>
<td>1</td>
<td>0</td>
<td>$\frac{21}{13}$</td>
<td>$\frac{1}{13}$</td>
<td>$\frac{2}{13}$</td>
<td>$\frac{2}{13}$</td>
<td></td>
</tr>
<tr>
<td>-2</td>
<td>$\tilde{x}_2^l$</td>
<td>$[30, 83, 9, 9; 30, 83, 16, 16]$</td>
<td>0</td>
<td>1</td>
<td>$\frac{21}{13}$</td>
<td>$\frac{1}{13}$</td>
<td>$\frac{2}{13}$</td>
<td>$\frac{2}{13}$</td>
<td></td>
</tr>
</tbody>
</table>

Max $\tilde{Q}^l = -5\tilde{x}_1^l - 2\tilde{x}_2^l + 0\tilde{x}_3^l + 0\tilde{x}_4^l - 5$

subject to

$\tilde{Q}^l = [30, 83, 9, 9; 30, 83, 16, 16]$

$\tilde{Q}^l - \tilde{D}^l = [30, 83, 9, 9; 30, 83, 16, 16]$

where $\tilde{x}_1^l, \tilde{x}_2^l$ are the slack intuitionistic fuzzy variables.

Here $\tilde{x}_1^l$ is entering variable and $\tilde{x}_2^l$ is leaving variable.

\[ \tilde{x}_1^l = \begin{bmatrix} 30 \\ 13 \\ 26 \\ 13 \\ 13 \\ 13 \\ 13 \\ 13 \\ 13 \end{bmatrix} \text{ and } \mathbb{R}[\tilde{x}_1^l] = 5.7692 \]

\[ \tilde{x}_2^l = \begin{bmatrix} 8 \\ 13 \\ 26 \\ 13 \\ 13 \\ 13 \\ 13 \\ 13 \\ 13 \end{bmatrix} \text{ and } \mathbb{R}[\tilde{x}_2^l] = 1.5385 \]
Max $Q(\tilde{X}^I) = -36.923$ and the value of $\tilde{Z}^I_0 = 1.2250$. Here $\tilde{x}^I_4$ is entering variable and $\tilde{x}^I_2$ is leaving variable.

### Table 5

<table>
<thead>
<tr>
<th>Basic Variables</th>
<th>$D^I_B$</th>
<th>$Y^I_B$</th>
<th>$X^I_B$</th>
<th>$\tilde{x}^I_1$</th>
<th>$\tilde{x}^I_2$</th>
<th>$\tilde{x}^I_3$</th>
<th>$\tilde{x}^I_4$</th>
<th>$\Re \theta$</th>
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<tbody>
<tr>
<td>$D^I_B$</td>
<td>$Y^I_B$</td>
<td>$X^I_B$</td>
<td>$\tilde{x}^I_1$</td>
<td>$\tilde{x}^I_2$</td>
<td>$\tilde{x}^I_3$</td>
<td>$\tilde{x}^I_4$</td>
<td>$\Re \theta$</td>
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</tr>
<tr>
<td>0</td>
<td>$\tilde{x}^I_2$</td>
<td>$[12, \frac{83}{13}, \frac{9}{13}, \frac{30}{13}, 12, \frac{83}{13}, \frac{23}{13}]$</td>
<td>$\frac{27}{13}$</td>
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<td>1</td>
<td>$\frac{30}{13}$</td>
<td>$\frac{75}{13}$</td>
<td></td>
</tr>
<tr>
<td>$-2$</td>
<td>$\tilde{x}^I_2$</td>
<td>$[2, \frac{15}{13}, \frac{6}{13}, \frac{12}{13}, \frac{15}{13}, \frac{30}{13}, \frac{25}{13}]$</td>
<td>$\frac{1}{13}$</td>
<td>0</td>
<td>1</td>
<td>$\frac{5}{13}$</td>
<td>$\frac{25}{13}$</td>
<td></td>
</tr>
</tbody>
</table>

All $\tilde{Q}^I_j - \tilde{D}^I_j \geq 0$. The current basic feasible solution is optimal. Therefore, the optimal solution to the problem (D) is

$\tilde{x}^I_1 = 0$ and $\Re [\tilde{x}^I_1] = 0$

$\tilde{x}^I_2 = 0$ and $\Re [\tilde{x}^I_2] = 0$

Max $Q(\tilde{X}^I) = -5$ and the value of $\tilde{Z}^I_1 = 1.2$.

Since $\tilde{Z}^I_0 > \tilde{Z}^I_1$ and by the step 5(a) of the proposed method, the optimal solution to the given intuitionistic fuzzy linear fractional programming problem is

$\tilde{x}^I_1 = \begin{bmatrix} 30 & 83 & 9 & 9 & 30 & 83 & 16 & 16 \\ 13 & 26 & 13 & 13 & 13 & 26 & 13 & 13 \end{bmatrix}$ and $\Re [\tilde{x}^I_1] = 5.7692$

$\tilde{x}^I_2 = \begin{bmatrix} 8 & 23 & 5 & 5 & 8 & 23 & 6 & 6 \\ 13 & 26 & 13 & 13 & 13 & 26 & 13 & 13 \end{bmatrix}$ and $\Re [\tilde{x}^I_2] = 1.5385$
In this paper to define a procedure to solve intuitionistic fuzzy linear fractional programming problem using denominator objective restriction method. The denominator objective restriction method is based on the intuitionistic fuzzy

<table>
<thead>
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<th>α</th>
<th>$\mu_{\tilde{A}}(x)$ x1</th>
<th>$\beta$</th>
<th>$\nu_{\tilde{A}}(x)$ x3</th>
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<th>x3</th>
<th>x4</th>
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<td>0.2</td>
<td>2.0615</td>
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<td>3.1923</td>
<td>1</td>
<td>4.4231</td>
</tr>
</tbody>
</table>

$Max \tilde{Z}^I = 1.2250$

By using MATLAB the graph will be shown. The membership and non-membership value of $\tilde{x}_1$ by using $(\alpha, \beta)$-cut method.

The membership and nonmembership value of $\tilde{x}_2$ by using $(\alpha, \beta)$-cut method.

6. Conclusion

In this paper to define a procedure to solve intuitionistic fuzzy linear fractional programming problem using denominator objective restriction method. The denominator objective restriction method is based on the intuitionistic fuzzy...
simplex method. The intuitionistic fuzzy linear fractional programming problem can be reduced to an intuitionistic fuzzy linear programming problem in which the constraints of right hand side is symmetric trapezoidal intuitionistic fuzzy number. This approach is very useful because most theoretical results developed in intuitionistic fuzzy linear programming could be relatively easily expanded to include intuitionistic fuzzy linear fractional programming problems. It is easy to understand, compute and also, to more experiment and industrial applications concerning decision making in an uncertain environment. The methods can serve decision makers by providing an appropriate best solution to a variety of linear fractional programming models having crisp or fuzzy parameters.
References


