Disjunctive Representation of Triangular Bipolar Neutrosophic Numbers, De-Bipolarization Technique and Application in Multi-Criteria Decision-Making Problems

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Abstract: This research paper adds to the theory of the generalized neutrosophic number from a distinctive frame of reference. It is universally known that the concept of a neutrosophic number is generally associated with and strongly related to the concept of positive, indeterminacy and non-belongingness membership functions. Currently, all membership functions always lie within the range of 0 to 1. However, we have generated bipolar concept in this paper where the membership contains both positive and negative parts within the range $-1$ to 0 and 0 to 1. We describe different structures of generalized triangular bipolar neutrosophic numbers, such as category-1, category-2, and category-3, in relation to the membership functions containing dependency or independency with each other. Researchers from different fields always want to observe the co-relationship and interdependence between fuzzy numbers and crisp numbers. In this platform, we also created the perception of de-bipolarization for a triangular bipolar neutrosophic number with the help of well-known techniques so that any bipolar neutrosophic fuzzy number of any type can be smoothly converted into a real number instantly. Creating a problem using bipolar neutrosophic perception is a more reliable, accurate, and trustworthy method than others. In this paper, we have also taken into account a multi-criteria decision-making problem (MCDM) for different users in the bipolar neutrosophic domain.

Keywords: bipolar neutrosophic number; de-bipolarization; multi-criteria decision-making problem; MCDM
1. Introduction

Fuzzy set theory, which deals with the concept of vagueness and uncertainty theory, was first presented by Zadeh in his paper [1] (1965). Vagueness theory plays a key role in solving problems related to engineering and statistical computation. It is widely used in social science, networking, and decision-making problems or any kind of real-life problem. Based on Zadeh’s paper, Atanassov [2] presented in 1986 the legerdemain idea of an intuitionistic fuzzy set in the field of mathematics where he considered the concept of membership function as well as non-membership function in the case of an intuitionistic fuzzy set. Subsequently, in 2007, Liu and Yuan [3] invented the concept of triangular intuitionistic fuzzy set, which in reality is the mixture of triangular fuzzy set and intuitionistic fuzzy set. Later, Ye [4] introduced the elementary idea of trapezoidal intuitionistic fuzzy set where both truth function and falsity function are a trapezoidal number in nature instead of triangular. Uncertainty theory plays an influential role in creating interesting models in various fields having scientific and technological problems. However, an elementary question arises: how can we develop or utilize the uncertainty concepts in our mathematical modeling with respect to daily life? Researchers everywhere have invented many approaches and methods to define those concepts and have offered different recommendations for using uncertainty philosophy. The literature presents different types of suggestions to classify some of the basic ambiguous parameters. It should be noted that there is no exclusive representation of the vagueness parameter. For solving a problem, the decision maker’s choice can be variously conferred in different applications.

In 1995, Smarandache (published in 1998) [5] put forward the idea of a neutrosophic set having three different components, namely, (i) truthiness, (ii) indeterminacies, and (iii) falseness. All aspects of the neutrosophic set are very relevant to our real-life systems. The neutrosophic concept is a very effective and exciting idea in real life. Later, Wang et al. [6] advanced the perception of a single typed neutrosophic set, which is very useful in solving any complex problem. Chakraborty et al. [7] introduced the concept of triangular neutrosophic as well as its classification. Chakraborty et al. [8] presented the perception of defuzzification using the removal area method. Maity et al. [9] also developed the idea of ranking and defuzzification in a new way.

To tackle human decision making based on positive and negative parts, Boscand Pivert [10] put forth the idea of bipolarity. They introduced the idea of the positive part of a membership function as well as the negative part of a membership function. Subsequently, Lee [11,12] defined the concept of bipolar fuzzy set in their research articles. Later, Kang and Kang [13] extended this idea into semi-groups and group structures. As research continued, Deli et al. [14] generated the idea of a bipolar neutrosophic set and tried to apply it to a decision-making problem. Broumi et al. [15] developed the concept of bipolar neutrosophic graph theory and, afterward, Ali and Smarandache [16] introduced the concept of the uncertain complex neutrosophic set. Molodtsov [17] introduced the concept of the soft bipolar set and, afterward, Aslam et al. [18] used it in an application-based problem. Vakkas et al. [19] invented the idea of similar measure on a bipolar set. Later, Wang et al. [20] also introduced the idea of operators in a bipolar neutrosophic set and used it in a decision-making problem. Recently, Raja et al. [21] developed a hope function in a bipolar neutrosophic set. We can find many applications of neutrosophic theory and development of multi criteria decision making problem in the literature surveys presented in [22–40]. The development of fuzzy set theory continues [41–49].

In this research article, we developed the concept of a different bipolar neutrosophic number in the triangular aspect. We invented both linear and nonlinear forms of a single typed triangular bipolar neutrosophic number for different categories. There are three categories of numbers when the three membership functions are maybe dependent or independent among each other. We introduced the concept of a category-1, -2, or -3 triangular bipolar neutrosophic numbers and also a generalized linear and nonlinear bipolar neutrosophic number. Researchers from everywhere are very interested in defuzzification techniques. As research goes on, they continually develop many techniques to solve the defuzzification problem. We introduced the removal area methods and built up the de-bipolarization
technique concept in the case of a linear triangular bipolar fuzzy number. Using that particular result, we converted a triangular bipolar neutrosophic set into a crisp one.

Currently, researchers from around the globe are very interested in solving multi-criteria decision-making problems. For that type of problem, we considered a finite number of alternatives as well as a finite number of attributes with different types of weight function for different numbers of decision makers. The goal of this method was to find a comparison between the alternatives and the attributes while maintaining the weight of the decision makers so that we could easily discover the best alternatives and the worst one. Many researchers have already proposed ideas about multi-criteria decision-making (MCDM) problems, but in this new triangular bipolar neutrosophic arena, we considered a MCDM problem and we focused and analyzed this problem using our developed de-polarization technique.

1.1. Motivation

The concept of vagueness plays a key role in mathematical modeling, engineering problem solving and medical diagnosis problem solving, among others. An important issue then arises if one considers a triangular bipolar neutrosophic number: what will be the linear and nonlinear forms and what will be the geometrical figure? How should we specify a category-1, -2, or -3 bipolar neutrosophic number when the membership functions are related to each other? Based on this perspective, we developed the subject of this research article. We succeeded in producing certain interesting results on de-bipolarization techniques and other applications.

1.2. Novelties

Numerous works have already been published in this bipolar fuzzy set context. Researchers have already developed several formulations and applications in various fields. However, many interesting results are still unknown. Our work aimed to develop ideas for those unknown aspects:

(i) Introduction of a distinctive form of triangular bipolar neutrosophic fuzzy number and its definition for different cases.
(ii) Graphical representation of a triangular bipolar neutrosophic fuzzy number.
(iii) Development of a de-bipolarization technique.
(iv) Application in an MCDM problem.

1.3. Verbal Phrases in the Neutrosophic Arena

With respect to daily life, an interesting question often arises about how we can connect the concept of vagueness and neutrosophic theory to real life and, in that case, what are the verbal phrases that can be used.

Example 1. Let us consider a problem of vote casting. Suppose, in an election, that one must select a candidate among a finite number of candidates. People have different emotions, feelings, demands, ethics, dreams, etc. Therefore, according to their viewpoint, the result can be any kind of fuzzy number, such as an interval number, triangular fuzzy number, intuitionistic number, or neutrosophic fuzzy number. Let us check the verbal phrases in each different case for the given problem in Table 1.

Table 1. Verbal phrases.

| Distinct Parameter       | Verbal Phrase | Information                                           |
|--------------------------|---------------|-------------------------------------------------------|
| Interval Number          | [Low, High]   | Voter will select according to their first priority within a certain range, like [2nd,3rd] candidate. |
| Triangular Fuzzy Number  | [Low, Median, High] | Voter will select according to their first priority containing an intermediate candidate, like [1st,2nd,3rd] |
Table 1. Cont.

| Distinct Parameter                  | Verbal Phrase                    | Information                                                                 |
|-------------------------------------|----------------------------------|-----------------------------------------------------------------------------|
| Intuitionistic (Triangular)         | [Standard, Median, High; Very Low, Poor, Low] | Voters will select some candidates directly and reject others immediately according to their viewpoint. |
| Neutrosophic (Triangular Bipolar)   | [High, Standard, Very High; Intermediate, Average, Median; Very Low, Poor, Low] | Some voters will select some candidates directly, some will hesitate when casting their vote, and some will directly reject voting according to their viewpoint. |

1.4. Logical Relationship between the Objective and the Subjective Part of this Paper

The objective part of this paper is to invent the disjunctive form of a triangular bipolar fuzzy number and the geometrical representation of it for different cases. The subjective part is to apply the linear form of the bipolar neutrosophic number to a real-life multi-criteria decision-making problem. To do so, we invented the logical removal area technique to compute the de-bipolarized value of the defined number. Using this de-bipolarization technique, we were able to relate a crisp number and a triangular bipolar neutrosophic number very easily. For the multi-criteria decision-making problem, we considered the weighted mean approach and the normalized approach as well as the de-bipolarization method to compute the best alternatives.

1.5. Structure of this Paper

In this research article, Section 1 contains the introduction presenting the basic concepts and the literature survey. It also includes the novelties, the motivation for this work, and verbal phrase perceptions on the neutrosophic domain. Section 2 presents the preliminary portion, some established definitions, and other elements. Section 3 contains the concept and development of disjunctive forms of a linear triangular bipolar fuzzy number. Section 4 presents its nonlinear and generalized form. In Section 5, we present the developed de-bipolarization technique of a linear bipolar fuzzy number (with disjunctive figures), using the concept of a removal area technique presented in Section 6. We consider a multi-criteria decision-making (MCDM) problem in a bipolar neutrosophic environment and solve it using the results from the previous sections with a real-life example accompanied by a sensitivity analysis. Lastly, Section 7 contains the conclusions reached about the research undertaken.

2. Preliminaries

Definition 1. Fuzzy Set: [1] A set $\tilde{X}$, defined as $\tilde{X} = \{(y, \mu_{\tilde{X}}(y)) : y \in X, \mu_{\tilde{X}}(y) \in [0, 1]\}$ and generally denoted by the pair $(y, \mu_{\tilde{X}}(y))$, $y$ belongs to the crisp set $X$ and $\mu_{\tilde{X}}(y)$ belongs to the interval $[0, 1]$, then set $\tilde{X}$ is called a fuzzy set.

Definition 2. Neutrosophic Set: [5] A set $\tilde{\text{neutroS}}$ in the universal discourse $X$ generally specified by $x$ is called a neutrosophic set if $\tilde{\text{neutroS}} = \{(x; \varepsilon_{\text{neutroS}}(x), \ell_{\text{neutroS}}(x), \mu_{\text{neutroS}}(x)) : x \in X\}$, where $\varepsilon_{\text{neutroS}}(x) : X \rightarrow [0, 1]$ represents the degree of confidence, $\ell_{\text{neutroS}}(x) : X \rightarrow [0, 1]$ represents the degree of hesitation and $\mu_{\text{neutroS}}(x) : X \rightarrow [0, 1]$ represents the degree of falseness of the decision. Where, $\varepsilon_{\text{neutroS}}(x), \ell_{\text{neutroS}}(x)$ & $\mu_{\text{neutroS}}(x)$ satisfies the relation

$$0 \leq \varepsilon_{\text{neutroS}}(x) + \ell_{\text{neutroS}}(x) + \mu_{\text{neutroS}}(x) \leq 3.$$
Definition 3. Single Typed Neutrosophic Number: [10] Single Typed Neutrosophic Number \( \tilde{z} \) is specified as
\[
\tilde{z} = \left\{ \left[ \left( p^1, q^1, r^1, s^1 \right); \alpha \right], \left( p^2, q^2, r^2, s^2 \right); \beta \right\} \quad \text{where} \quad \alpha, \beta, \gamma \in [0, 1], \text{where} \quad (\varepsilon_{\tilde{z}}) : \mathbb{R} \rightarrow [0, \alpha], (\xi_{\tilde{z}}) : \mathbb{R} \rightarrow [\beta, 1] \text{ and } (\mu_{\tilde{z}}) : \mathbb{R} \rightarrow [\gamma, 1] \text{ is given as:}
\]
\[
\varepsilon_{\tilde{z}}(x) = \left\{ \begin{array}{ll}
\varepsilon_{\tilde{z}}^1(x) & p^1 \leq x \leq q^1 \\
\varepsilon_{\tilde{z}}^2(x) & q^1 \leq x \leq r^1 \\
\varepsilon_{\tilde{z}}^3(x) & r^1 \leq x \leq s^1 \\
0 & \text{otherwise}
\end{array} \right.
\]
\[
\xi_{\tilde{z}}(x) = \left\{ \begin{array}{ll}
\xi_{\tilde{z}}^1(x) & p^2 \leq x \leq q^2 \\
\xi_{\tilde{z}}^2(x) & q^2 \leq x \leq r^2 \\
\xi_{\tilde{z}}^3(x) & r^2 \leq x \leq s^2 \\
1 & \text{otherwise}
\end{array} \right.
\]
\[
\mu_{\tilde{z}}(x) = \left\{ \begin{array}{ll}
\mu_{\tilde{z}}^1(x) & p^3 \leq x \leq q^3 \\
\mu_{\tilde{z}}^2(x) & q^3 \leq x \leq r^3 \\
\mu_{\tilde{z}}^3(x) & r^3 \leq x \leq s^3 \\
1 & \text{otherwise}
\end{array} \right.
\]

Definition 4. Bipolar Neutrosophic Set: [11] A bipolar neutrosophic set is specified as,
\[
\text{Bipolar Neutrosophic Set: } \left\{ \left( \varepsilon_{\text{Bineutros}}^+(x), \varepsilon_{\text{Bineutros}}^-(x), \mu_{\text{Bineutros}}^+(x), \mu_{\text{Bineutros}}^-(x) \right) \right\} \quad \text{where} \quad \varepsilon_{\text{Bineutros}}^+(x) : \mathbb{R} \rightarrow [0, 1], \varepsilon_{\text{Bineutros}}^-(x) : \mathbb{R} \rightarrow [-1, 0], \mu_{\text{Bineutros}}^+(x) : \mathbb{R} \rightarrow [0, 1], \mu_{\text{Bineutros}}^-(x) : \mathbb{R} \rightarrow [-1, 0]
\]

represents the degree of confidence, represents the degree of hesitation and represents the degree of falseness of the decision.

3. Single Typed Linear Triangular Bipolar Neutrosophic Number

In this section, we define different types of a single typed linear bipolar neutrosophic number. To help researchers, we present the following block diagram as in Figure 1:

![Figure 1. Block diagram for different types of a single typed linear triangular bipolar neutrosophic number.](image-url)
3.1. Triangular Single Typed Bipolar Neutrosophic Number of Category-1: The Portion of the Authenticity, Hesitation, and Untrue Are Independent

This case may arise with the following problem: one must select one member or one party in the election system of a country. Suppose there is a finite number of candidates, and one of them is X. A certain percentage of the people will surely cast their vote in favor of X, which is the authenticity function. A certain percentage of them will surely cast their vote against X, which is the untrue function. Apart from these two groups of people, a few people will hesitate to give their vote. Here, all components are independent.

A triangular single typed neutrosophic number of category-1 is specified as $\overline{A_{BiNeu}} = (i_1, i_2, i_3; j_1, j_2, j_3; k_1, k_2, k_3)$ and whose authenticity membership, hesitation, and untrue membership are specified as follows:

$$T^{+}_{\overline{A_{BiNeu}}} (x) = \begin{cases} \frac{x-i_1}{i_2-i_1} & \text{when } i_1 \leq x < i_2 \\ \frac{1}{1} & \text{when } x = i_2 \\ \frac{i_3-x}{i_3-i_2} & \text{when } i_2 < x \leq i_3 \end{cases}, T^{-}_{\overline{A_{BiNeu}}} (x) = \begin{cases} \frac{j_1-x}{i_2-i_1} & \text{when } i_1 \leq x < i_2 \\ \frac{-1}{1} & \text{when } x = i_2 \\ \frac{j_3-x}{i_3-i_2} & \text{when } i_2 < x \leq i_3 \end{cases}$$

and

$$I^{+}_{\overline{A_{BiNeu}}} (x) = \begin{cases} \frac{j_1-x}{j_2-j_1} & \text{when } j_1 \leq x < j_2 \\ \frac{0}{1} & \text{when } x = j_2 \\ \frac{j_3-x}{j_3-j_2} & \text{when } j_2 < x \leq j_3 \end{cases}, I^{-}_{\overline{A_{BiNeu}}} (x) = \begin{cases} \frac{x-j_1}{j_2-j_1} & \text{when } j_1 \leq x < j_2 \\ \frac{0}{1} & \text{when } x = j_2 \\ \frac{x-j_3}{j_3-j_2} & \text{when } j_2 < x \leq j_3 \end{cases}$$

and

$$F^{+}_{\overline{A_{BiNeu}}} (x) = \begin{cases} \frac{k_1-x}{k_2-k_1} & \text{when } k_1 \leq x < k_2 \\ \frac{0}{1} & \text{when } x = k_2 \\ \frac{k_3-x}{k_3-k_2} & \text{when } k_2 < x \leq k_3 \end{cases}, F^{-}_{\overline{A_{BiNeu}}} (x) = \begin{cases} \frac{x-k_1}{k_2-k_1} & \text{when } k_1 \leq x < k_2 \\ \frac{0}{1} & \text{when } x = k_2 \\ \frac{x-k_3}{k_3-k_2} & \text{when } k_2 < x \leq k_3 \end{cases}$$

where $-3 \leq T^{-}_{\overline{A_{BiNeu}}} (x) + I^{-}_{\overline{A_{BiNeu}}} (x) + F^{-}_{\overline{A_{BiNeu}}} (x) \leq 3$, $x \in \overline{A_{BiNeu}}$.

The parametric representation of the above category-1 number is defined as follows:

$$(\overline{A_{BiNeu}})_{\alpha, \beta, \gamma} = [T^{+}_{BiNeu1}(\alpha), T^{+}_{BiNeu2}(\alpha); I^{+}_{BiNeu1}(\beta), I^{+}_{BiNeu2}(\beta); F^{+}_{BiNeu1}(\gamma), F^{+}_{BiNeu2}(\gamma)]$$

where

$$T^{+}_{BiNeu1}(\alpha) = i_1 + \alpha(i_2 - i_1), T^{+}_{BiNeu2}(\alpha) = i_3 - \alpha(i_3 - i_2)$$
$$T^{-}_{BiNeu1}(\alpha) = i_2 - \alpha(i_2 - i_1), T^{-}_{BiNeu2}(\alpha) = i_3 + \alpha(i_3 - i_2)$$
$$I^{+}_{BiNeu1}(\beta) = j_2 - \beta(j_2 - j_1), I^{+}_{BiNeu2}(\beta) = j_2 + \beta(j_3 - j_2)$$
$$I^{-}_{BiNeu1}(\beta) = j_2 - \beta(j_2 - j_1), I^{-}_{BiNeu2}(\beta) = j_2 + \beta(j_3 - j_2)$$
$$F^{+}_{BiNeu1}(\gamma) = k_2 - \gamma(k_2 - k_1), F^{+}_{BiNeu2}(\gamma) = k_2 + \gamma(k_3 - k_2)$$
$$F^{-}_{BiNeu1}(\gamma) = k_2 + \gamma(k_2 - k_1), F^{-}_{BiNeu2}(\gamma) = k_2 - \gamma(k_3 - k_2)$$

Here, $-1 \leq \alpha \leq 1, -1 \leq \beta \leq 1, -1 \leq \gamma \leq 1$ and $-3 \leq \alpha + \beta + \gamma \leq 3$.

3.2. Triangular Single Typed Bipolar Neutrosophic Number of Category-2: The Portion of Hesitation and Untrue Are Dependent

This case may arise with the following problem: one must select one member or one party in the election system of a country. Suppose there is a finite number of candidates, and one of them is X.
A certain percentage of the people will surely cast their vote in favor of X, which is the authenticity function. However, a certain percentage of the people will surely cast their vote against X, while they hesitate to cast their vote for the other candidates in this election. In this case, the hesitation membership function and the untrue portion are dependent on each other.

A triangular single typed bipolar neutrosophic number of Category-2 is specified as \( \vec{A}_{BiNeu} = (i_1, i_2, j_1, j_2, j_3; u_{BiN}, y_{BiN}) \) and whose authenticity membership, hesitation, and untrue membership are specified as follows:

\[
T^+_{A_{BiNeu}}(x) = \begin{cases} 
\frac{x - i_1}{i_2 - i_1} & \text{when } i_1 \leq x < i_2 \\
1 & \text{when } x = i_2 \\
\frac{i_3 - x}{i_3 - i_2} & \text{when } i_2 < x \leq i_3 \\
0 & \text{otherwise}
\end{cases}
\]

\[
T^-_{A_{BiNeu}}(x) = \begin{cases} 
\frac{x - i_1}{i_2 - i_1} & \text{when } i_1 \leq x < i_2 \\
-1 & \text{when } x = i_2 \\
\frac{x - i_3}{i_3 - i_2} & \text{when } i_2 < x \leq i_3 \\
0 & \text{otherwise}
\end{cases}
\]

and

\[
I^+_{A_{BiNeu}}(x) = \begin{cases} 
\frac{x - i_1 + y_{Ne}(i_1)}{i_3 - i_1} & \text{when } j_1 \leq x < j_2 \\
u_{BN} & \text{when } x = j_2 \\
\frac{x - j_3 + y_{Ne}(j_3 - x)}{j_3 - j_2} & \text{when } j_2 \leq x \leq j_3 \\
0 & \text{otherwise}
\end{cases}
\]

and

\[
I^-_{A_{BiNeu}}(x) = \begin{cases} 
\frac{x - i_1 - y_{Ne}(i_1)}{i_3 - i_1} & \text{when } j_1 \leq x < j_2 \\
u_{BN} & \text{when } x = j_2 \\
\frac{x - j_3 - y_{Ne}(j_3 - x)}{j_3 - j_2} & \text{when } j_2 \leq x \leq j_3 \\
0 & \text{otherwise}
\end{cases}
\]

\[
F^+_{A_{BiNeu}}(x) = \begin{cases} 
\frac{i_2 - x + y_{Ne}(x - i_2)}{i_3 - i_2} & \text{when } j_1 \leq x < j_2 \\
y_{Ne} & \text{when } x = j_2 \\
\frac{i_2 - x + y_{Ne}(x - i_2)}{i_3 - i_2} & \text{when } j_2 \leq x \leq j_3 \\
0 & \text{otherwise}
\end{cases}
\]

\[
F^-_{A_{BiNeu}}(x) = \begin{cases} 
\frac{i_2 - x - y_{Ne}(x - i_2)}{i_3 - i_2} & \text{when } j_1 \leq x < j_2 \\
y_{Ne} & \text{when } x = j_2 \\
\frac{i_2 - x - y_{Ne}(x - i_2)}{i_3 - i_2} & \text{when } j_2 \leq x \leq j_3 \\
0 & \text{otherwise}
\end{cases}
\]

where \(-2 \leq T^+_{A_{BiNeu}}(x) + I^+_{A_{BiNeu}}(x) + F^+_{A_{BiNeu}}(x) \leq 2, x \in \vec{A}_{BiNeu}\).

The parametric representation of the above category-2 number is defined as follows:

\[
\left( \vec{A}_{BiNeu} \right)_{\alpha, \beta, \gamma} = [T_{BiNeu1}(\alpha), T_{BiNeu2}(\alpha); I_{BiNeu1}(\beta), I_{BiNeu2}(\beta), F_{BiNeu1}(\gamma), F_{BiNeu2}(\gamma)]
\]

where

\[
T^+_{BiNeu1}(\alpha) = i_1 + \alpha(i_2 - i_1), T^+_{BiNeu2}(\alpha) = i_3 - \alpha(i_3 - i_2)
\]

\[
T^-_{BiNeu1}(\alpha) = i_2 - \alpha(i_2 - i_1), T^-_{BiNeu2}(\alpha) = i_3 + \alpha(i_3 - i_2)
\]

\[
I^+_{BiNeu1}(\beta) = \frac{i_2 - u_{BN}i_1 - \beta(i_2 - i_1)}{1 - u_{BN}}, I^+_{BiNeu2}(\beta) = \frac{i_2 - u_{BN}i_1 + \beta(i_2 - i_1)}{1 - u_{BN}}
\]

\[
I^-_{BiNeu1}(\beta) = \frac{i_1 - u_{BN}i_2 + \beta(i_2 - i_1)}{1 - u_{BN}}, I^-_{BiNeu2}(\beta) = \frac{i_1 - u_{BN}i_2 - \beta(i_2 - i_1)}{1 - u_{BN}}
\]

\[
F^+_{BiNeu1}(\gamma) = \frac{i_2 - y_{BN}i_1 - \gamma(i_2 - i_1)}{1 - y_{BN}}, F^+_{BiNeu2}(\gamma) = \frac{i_2 - y_{BN}i_1 + \gamma(i_2 - i_1)}{1 - y_{BN}}
\]

\[
F^-_{BiNeu1}(\gamma) = \frac{i_1 - y_{BN}i_2 + \gamma(i_2 - i_1)}{1 - y_{BN}}, F^-_{BiNeu2}(\gamma) = \frac{i_1 - y_{BN}i_2 - \gamma(i_2 - i_1)}{1 - y_{BN}}
\]

Here, \(-1 \leq \alpha \leq 1, u_{BN} \leq \beta \leq 1, y_{BN} \leq \gamma \leq 1\) and \(-1 \leq \beta + \gamma \leq 1\) and \(-1 \leq \alpha + \beta + \gamma \leq 2\).

3.3. Triangular Single Typed Bipolar Neutrosophic Number of Category-3: The Portion of the Authenticity, Hesitation, and Untrue Are Dependent

This case may arise with the following problem: suppose a company manufactures some useful products and they have launched them into the market. They do not know whether they will be accepted in the market (hesitant function). After product launching, people can either accept then (authenticity function) or reject them (untrue function). Here, all three components are dependent on each other.
A triangular single typed bipolar neutrosophic number of Category-3 is specified as \( \tilde{A}_{\text{BiNeu}} = (i_1, i_2, i_3; w_{\text{BN}}, u_{\text{BN}}, y_{\text{BN}}) \) and whose authenticity membership, hesitation, and untrue membership are specified as follows:

\[
T^+ \tilde{A}_{\text{BiNeu}} (x) = \begin{cases} 
  w_{\text{BN}} \frac{x \cdot i_1}{i_2 - i_1} & \text{when } i_1 \leq x < i_2 \\
  w_{\text{BN}} & \text{when } x = i_2 \\
  w_{\text{BN}} \frac{x \cdot i_3}{i_3 - i_2} & \text{when } i_2 < x \leq i_3 \\
  0 & \text{otherwise}
\end{cases}
\]

\[
T^- \tilde{A}_{\text{BiNeu}} (x) = \begin{cases} 
  w_{\text{BN}} \frac{\overline{x} \cdot i_1}{i_2 - i_1} & \text{when } i_1 \leq x < i_2 \\
  w_{\text{BN}} & \text{when } x = i_2 \\
  w_{\text{BN}} \frac{\overline{x} \cdot i_3}{i_3 - i_2} & \text{when } i_2 < x \leq i_3 \\
  0 & \text{otherwise}
\end{cases}
\]

and

\[
I^+ \tilde{A}_{\text{BiNeu}} (x) = \begin{cases} 
  u_{\text{BN}} \frac{i_2 - x + u_{\text{BN}}(x - i_1)}{i_2 - i_1} & \text{when } i_1 \leq x < i_2 \\
  u_{\text{BN}} & \text{when } x = i_2 \\
  u_{\text{BN}} \frac{i_2 - x + u_{\text{BN}}(i_3 - x)}{i_3 - i_2} & \text{when } i_2 < x \leq i_3 \\
  1 & \text{otherwise}
\end{cases}
\]

\[
I^- \tilde{A}_{\text{BiNeu}} (x) = \begin{cases} 
  y_{\text{BN}} \frac{i_2 - x + y_{\text{BN}}(x - i_1)}{i_2 - i_1} & \text{when } i_1 \leq x < i_2 \\
  y_{\text{BN}} & \text{when } x = i_2 \\
  y_{\text{BN}} \frac{i_2 - x + y_{\text{BN}}(i_3 - x)}{i_3 - i_2} & \text{when } i_2 < x \leq i_3 \\
  1 & \text{otherwise}
\end{cases}
\]

and

\[
F^+ \tilde{A}_{\text{BiNeu}} (x) = \begin{cases} 
  u_{\text{BN}} \frac{i_2 - x + u_{\text{BN}}(x - i_1)}{i_2 - i_1} & \text{when } i_1 \leq x < i_2 \\
  u_{\text{BN}} (i_2 - x) & \text{when } x = i_2 \\
  u_{\text{BN}} \frac{i_2 - x + u_{\text{BN}}(i_3 - x)}{i_3 - i_2} & \text{when } i_2 < x \leq i_3 \\
  1 & \text{otherwise}
\end{cases}
\]

\[
F^- \tilde{A}_{\text{BiNeu}} (x) = \begin{cases} 
  y_{\text{BN}} \frac{i_2 - x + y_{\text{BN}}(x - i_1)}{i_2 - i_1} & \text{when } i_1 \leq x < i_2 \\
  y_{\text{BN}} (i_2 - x) & \text{when } x = i_2 \\
  y_{\text{BN}} \frac{i_2 - x + y_{\text{BN}}(i_3 - x)}{i_3 - i_2} & \text{when } i_2 < x \leq i_3 \\
  1 & \text{otherwise}
\end{cases}
\]

where \(-1 \leq T_{\tilde{A}_{\text{BiNeu}}} (x) + I_{\tilde{A}_{\text{BiNeu}}} (x) + F_{\tilde{A}_{\text{BiNeu}}} (x) \leq 1, \ x \in \tilde{A}_{\text{BiNeu}}\).

The parametric representation of the above category-3 number is defined as follows:

\[
\left( \tilde{A}_{\text{BiNeu}} \right)_{\alpha, \beta, \gamma} = [T_{\text{BiNeu}1}(\alpha), T_{\text{BiNeu}2}(\alpha); I_{\text{BiNeu}1}(\beta), I_{\text{BiNeu}2}(\beta); F_{\text{BiNeu}1}(\gamma), F_{\text{BiNeu}2}(\gamma)]
\]

where

\[
T^+ \text{BiNeu}1(\alpha) = i_1 + \frac{\alpha}{w_{\text{BN}}} (i_2 - i_1), T^+ \text{BiNeu}2(\alpha) = i_2 + \frac{\alpha}{w_{\text{BN}}} (i_3 - i_2), T^- \text{BiNeu}1(\alpha) = i_3 - \frac{\alpha}{w_{\text{BN}}} (i_3 - i_2), T^- \text{BiNeu}2(\alpha) = i_3 - \frac{\alpha}{w_{\text{BN}}} (i_3 - i_2)
\]

\[
I^+ \text{BiNeu}1(\beta) = \frac{i_2 - u_{\text{BN}}(i_2 - i_1)}{1 - u_{\text{BN}}} , I^+ \text{BiNeu}2(\beta) = \frac{i_2 - u_{\text{BN}}(i_3 - i_2)}{1 - u_{\text{BN}}}, I^- \text{BiNeu}1(\beta) = \frac{i_2 - u_{\text{BN}}(i_2 - i_1)}{1 - u_{\text{BN}}} , I^- \text{BiNeu}2(\beta) = \frac{i_2 - u_{\text{BN}}(i_3 - i_2)}{1 - u_{\text{BN}}}
\]

\[
F^+ \text{BiNeu}1(\gamma) = \frac{i_2 - y_{\text{BN}}(i_2 - i_1)}{1 - y_{\text{BN}}} , F^+ \text{BiNeu}2(\gamma) = \frac{i_2 - y_{\text{BN}}(i_3 - i_2)}{1 - y_{\text{BN}}} , F^- \text{BiNeu}1(\gamma) = \frac{i_2 - y_{\text{BN}}(i_2 - i_1)}{1 - y_{\text{BN}}} , F^- \text{BiNeu}2(\gamma) = \frac{i_2 - y_{\text{BN}}(i_3 - i_2)}{1 - y_{\text{BN}}}
\]

Here, \(-1 \leq \alpha, \beta, \gamma \leq 1, y_{\text{BN}} \leq \gamma \leq 1 \text{ and } -1 \leq \alpha + \beta + \gamma \leq 1\)

4. Single Typed Nonlinear Triangular Bipolar Neutrosophic Number

4.1. Single Typed Nonlinear Triangular Bipolar Neutrosophic Number

A single typed nonlinear triangular bipolar neutrosophic number is specified as \( \tilde{A}_{\text{BiNeu}} = (i_1, i_2, i_3; j_1, j_2, j_3; k_1, k_2, k_3; p_1, q_1, r_1, q_2, r_2) \) and whose positive membership, hesitation, and negative membership are specified as follows and graphically in Figure 2.

\[
T^+ \tilde{A}_{\text{BiNeu}} (x) = \begin{cases} 
  \left( \frac{x \cdot j_1}{j_2 - j_1} \right)^{p_1} & \text{when } i_1 \leq x < i_2 \\
  1 & \text{when } x = i_2 \\
  \left( \frac{x \cdot j_3}{j_3 - j_2} \right)^{p_2} & \text{when } i_2 < x \leq i_3 \\
  0 & \text{otherwise}
\end{cases}
\]

\[
T^- \tilde{A}_{\text{BiNeu}} (x) = \begin{cases} 
  \left( \frac{x \cdot k_1}{k_2 - k_1} \right)^{q_1} & \text{when } i_1 \leq x < i_2 \\
  -1 & \text{when } x = i_2 \\
  \left( \frac{x \cdot k_3}{k_3 - k_2} \right)^{q_2} & \text{when } i_2 < x \leq i_3 \\
  0 & \text{otherwise}
\end{cases}
\]
and

\[
I^+_{\tilde{A}_{BiNeu}}(x) = \begin{cases} 
\frac{x-j_1}{j_2-j_1} & \text{when } j_1 \leq x < j_2 \\
0 & \text{when } x = j_2 \\
\frac{x-j_2}{j_3-j_2} & \text{when } j_2 < x \leq j_3 \\
1 & \text{otherwise}
\end{cases}
\]

\[
I^-_{\tilde{A}_{BiNeu}}(x) = \begin{cases} 
\frac{x-j_1}{j_2-j_1} & \text{when } j_1 \leq x < j_2 \\
0 & \text{when } x = j_2 \\
\frac{x-j_2}{j_3-j_2} & \text{when } j_2 < x \leq j_3 \\
-1 & \text{otherwise}
\end{cases}
\]

and

\[
F^+_{\tilde{A}_{BiNeu}}(x) = \begin{cases} 
\frac{x-k_1}{k_2-k_1} & \text{when } k_1 \leq x < k_2 \\
0 & \text{when } x = k_2 \\
\frac{x-k_2}{k_3-k_2} & \text{when } k_2 < x \leq k_3 \\
1 & \text{otherwise}
\end{cases}
\]

\[
F^-_{\tilde{A}_{BiNeu}}(x) = \begin{cases} 
\frac{x-k_1}{k_2-k_1} & \text{when } k_1 \leq x < k_2 \\
0 & \text{when } x = k_2 \\
\frac{x-k_2}{k_3-k_2} & \text{when } k_2 < x \leq k_3 \\
-1 & \text{otherwise}
\end{cases}
\]

where

\[
T^+_{\tilde{A}_{BiNeu}}(x) : X \in [0,1], T^-_{\tilde{A}_{BiNeu}}(x) : X \in [-1,0], I^+_{\tilde{A}_{BiNeu}}(x) : X \in [0,1], I^-_{\tilde{A}_{BiNeu}}(x) : X \in [-1,0],
\]

\[
F^+_{\tilde{A}_{BiNeu}}(x) : X \in [0,1], F^-_{\tilde{A}_{BiNeu}}(x) : X \in [-1,0].
\]

\[\text{Figure 2. Nonlinear triangular bipolar neutrosophic number.}\]

4.2. Single Typed Generalized Triangular Bipolar Neutrosophic Number

A single typed triangular bipolar generalized neutrosophic number is specified as

\[\tilde{A}_{BiNeu} = (i_1, i_2, i_3; j_1, j_2, j_3; k_1, k_2, k_3; \omega; \rho; \lambda)\]

and whose positive membership, hesitation and negative membership are specified as follows:

\[
T^+_{\tilde{A}_{BiNeu}}(x) = \begin{cases} 
\omega \frac{x-i_1}{i_2-i_1} & \text{when } i_1 \leq x < i_2 \\
\omega & \text{when } x = i_2 \\
\omega \frac{x-i_2}{i_3-i_2} & \text{when } i_2 < x \leq i_3 \\
0 & \text{otherwise}
\end{cases}
\]

\[
T^-_{\tilde{A}_{BiNeu}}(x) = \begin{cases} 
\omega \frac{x-i_1}{i_2-i_1} & \text{when } i_1 \leq x < i_2 \\
-\omega & \text{when } x = i_2 \\
\omega \frac{x-i_2}{i_3-i_2} & \text{when } i_2 < x \leq i_3 \\
0 & \text{otherwise}
\end{cases}
\]
and

\[
I^+_{\tilde{A}_{BiNeu}}(x) = \begin{cases} 
\rho \frac{x-j_1}{j_2-j_1} & \text{when } j_1 \leq x < j_2 \\
0 & \text{when } x = j_2 \\
\rho \frac{x-j_3}{j_3-j_2} & \text{when } j_2 < x \leq j_3
\end{cases}
\]

and

\[
F^+_{\tilde{A}_{BiNeu}}(x) = \begin{cases} 
\lambda \frac{X-\rho}{X-\rho} & \text{when } k_1 \leq X < k_2 \\
0 & \text{when } X = k_2 \\
\lambda \frac{X-\rho}{X-\rho} & \text{when } k_2 < X \leq k_3
\end{cases}
\]

where \( T^+_{\tilde{A}_{BiNeu}}(x) : X \in [0,1], T^-_{\tilde{A}_{BiNeu}}(x) : X \in [-1,0], I^+_{\tilde{A}_{BiNeu}}(x) : X \in [0,1], I^-_{\tilde{A}_{BiNeu}}(x) : X \in [-1,0], F^+_{\tilde{A}_{BiNeu}}(x) : X \in [0,1], F^-_{\tilde{A}_{BiNeu}}(x) : X \in [-1,0].

4.3. Single Typed Generalized Non Linear Triangular Bipolar Neutrosophic Number

A single typed nonlinear triangular generalized bipolar neutrosophic number with nine components is specified as \( \tilde{A}_{BiNeu} = (\tilde{i}_1, \tilde{i}_2, \tilde{i}_3, \tilde{j}_1, \tilde{j}_2, \tilde{j}_3, \tilde{k}_1, \tilde{k}_2, \tilde{k}_3; p_1, p_2, q_1, q_2; r_1, r_2; \omega; \rho; ) \) and whose authenticity membership, hesitation and untruth membership are defined as follows:

\[
T^+_{\tilde{A}_{BiNeu}}(x) = \begin{cases} 
\omega \frac{X-i_1}{i_2-i_1} & \text{when } i_1 \leq X < i_2 \\
\omega & \text{when } X = i_2 \\
0 & \text{when } i_2 < X \leq i_3
\end{cases}
\]

and

\[
I^+_{\tilde{A}_{BiNeu}}(x) = \begin{cases} 
\rho \frac{j_1-X}{j_2-j_1} & \text{when } j_1 \leq X < j_2 \\
0 & \text{when } X = j_2 \\
\rho \frac{j_3-X}{j_3-j_2} & \text{when } j_2 < X \leq j_3
\end{cases}
\]

and

\[
F^+_{\tilde{A}_{BiNeu}}(x) = \begin{cases} 
\lambda \frac{X-k_1}{k_2-k_1} & \text{when } k_1 \leq X < k_2 \\
0 & \text{when } X = k_2 \\
\lambda \frac{X-k_3}{k_3-k_2} & \text{when } k_2 < X \leq k_3
\end{cases}
\]

where \( T^-_{\tilde{A}_{BiNeu}}(x) : X \in [0,1], T^-_{\tilde{A}_{BiNeu}}(x) : X \in [-1,0], I^-_{\tilde{A}_{BiNeu}}(x) : X \in [0,1], I^-_{\tilde{A}_{BiNeu}}(x) : X \in [-1,0], F^-_{\tilde{A}_{BiNeu}}(x) : X \in [0,1], F^-_{\tilde{A}_{BiNeu}}(x) : X \in [-1,0].

5. De-Bipolarization of a Linear Neutrosophic Triangular Bipolar Fuzzy Number

De-bipolarization is a process of creating a logical result in a crisp system corresponding to the bipolar neutrosophic fuzzy number and its membership function. Researchers from different countries of the world are interested in this question: if there is a bipolar neutrosophic fuzzy number having its membership function, what will be the crisp value associated with the number? As research goes on, they continually consider useful methods to convert a fuzzy number into a crisp number.

Some of the well-known methods are as follows:
1. BADD (basic defuzzification distributions)
2. BOA (bisector of area)
3. CDD (constraint decision defuzzification)
4. COA (center of area)
5. COG (center of gravity)
6. ECOA (extended center of area)
7. EQM (extended quality method)
8. FCD (fuzzy clustering defuzzification), etc.

In our triangular bipolar neutrosophic environment, researchers are very interested in finding out which conversion process will be applicable and logical to convert a triangular bipolar neutrosophic number into a crisp number. In the case of a triangular bipolar neutrosophic fuzzy number, three different kinds of membership functions are present. Lastly, we propose the “removal area method” to convert a triangular bipolar neutrosophic fuzzy number into a crisp number.

5.1. De-Bipolarization Using the Removal Area Method

Let us consider a linear bipolar neutrosophic triangular fuzzy number as follows:

\[ \tilde{A}_{\text{Bineu}} = \langle a, b, c; d, e, f; g, h, k \rangle \]

The graphical representation of a triangular bipolar neutrosophic fuzzy number is presented in Figure 3:

We assume areal number \( l \in \mathbb{R} \) and an uncertain number \( \tilde{X} \) for black line indicated triangles, the left portion area of \( \tilde{X} \) w.r.t \( l \) is \( S_l(\tilde{X}, l) \) is specified as the region enclosed by \( l \) and the left portion of the fuzzy number \( \tilde{X} \). Using the same concept, the right portion area of \( \tilde{X} \) w.r.t \( l \) is \( S^r(\tilde{X}, l) \), now consider a real number \( l \in \mathbb{R} \) together with a fuzzy number \( \tilde{Y} \) for the left most top and lower triangles (\( \Delta_{def} \)), then the left portion area of \( \tilde{Y} \) w.r.t \( l \) is \( S_l(\tilde{Y}, l) \) and is specified as the region enclosed by \( l \) and the left
portion of the fuzzy number $\bar{Y}$. Again, the right portion area of $\bar{Y}$ w.r.t $l$ is $S_l(\bar{Y}, l)$, a fuzzy number $\bar{Z}$ for the right most top and lower triangle ($\Delta ghk$), then the left portion removal of $\bar{Z}$ w.r.t $l$ is $S_l(\bar{Z}, l)$ and is specified as the region enclosed by $l$ and the left portion of the fuzzy number $\bar{Z}$. Similarly, the right portion removal of $\bar{Z}$ w.r.t $l$ is $S_l(\bar{Z}, l)$.

Mean is defined as $S(\bar{X}, l) = \frac{S_l(\bar{X}, l) + S_r(\bar{X}, l)}{2}$, $S(\bar{Y}, l) = \frac{S_l(\bar{Y}, l) + S_r(\bar{Y}, l)}{2}$, $S(\bar{Z}, l) = \frac{S_l(\bar{Z}, l) + S_r(\bar{Z}, l)}{2}$

Then, we specified the de-bipolarization of a linear bipolar neutrosophic triangular fuzzy as follows:

$$S(D_{Bipolar}, l) = \frac{S(\bar{X}, l) + S(\bar{Y}, l) + S(\bar{Z}, l)}{3}$$

For $l = 0$, $S(\bar{X}, 0) = \frac{S_l(\bar{X}, 0) + S_r(\bar{X}, 0)}{2}$, $S(\bar{Y}, 0) = \frac{S_l(\bar{Y}, 0) + S_r(\bar{Y}, 0)}{2}$, $S(\bar{Z}, 0) = \frac{S_l(\bar{Z}, 0) + S_r(\bar{Z}, 0)}{2}$

Then, $S(D_{Bipolar}, 0) = \frac{S(\bar{X}, 0) + S(\bar{Y}, 0) + S(\bar{Z}, 0)}{3}$

We take $\bar{X} = (a, b, c)$, $\bar{Y} = (d, e, f)$, $\bar{Z} = (g, h, k)$.

Then, from Figures 4–6

$$S_l(\bar{X}, 0) = \text{Area of Figure 4a} = \frac{(d+e)}{2} \times 2 = (d + e)$$
$$S_r(\bar{X}, 0) = \text{Area of Figure 4b} = \frac{(e+f)}{2} \times 2 = (e + f)$$
$$S_l(\bar{Y}, 0) = \text{Area of Figure 5a} = \frac{(a+h)}{2} \times 2 = (a + h)$$
$$S_r(\bar{Y}, 0) = \text{Area of Figure 5b} = \frac{(k+h)}{2} \times 2 = (k + h)$$
$$S_l(\bar{Z}, 0) = \text{Area of Figure 6a} = \frac{(a+b)}{2} \times 2 = (a + b)$$
$$S_r(\bar{Z}, 0) = \text{Area of Figure 6b} = \frac{(b+c)}{2} \times 2 = (b + c)$$

Hence,

$$S(\bar{X}, 0) = \frac{(a+2b+c)}{2}, S(\bar{Y}, 0) = \frac{(d+2e+f)}{2}, S(\bar{Z}, 0) = \frac{(g+2h+k)}{2}$$

So,

$$S(D_{Bipolar}, 0) = \frac{(a+2b+c+d+2e+f+g+2h+k)}{6}$$

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{figure4.png}
\caption{(a) Shaded Region of falsity portion (Step I); (b) Shaded Region of falsity portion (Step II).}
\end{figure}
6. Multi-Criteria Decision-Making in a Triangular Bipolar Neutrosophic Fuzzy Set Environment

In this process, we tried to find the best alternatives on the basis of the attribute values defined by the decision makers. It is not an easy task to evaluate the attribute value in terms of a crisp number due to the presence of impreciseness. The information of the attribute values are of triangular bipolar neutrosophic number in nature.

In our daily lives, we often face multi-criteria decision-making problems. Suppose someone wants to buy a good mobile phone within their financial range. Many companies and products are available in the market and, additionally, the products have different types of features like camera quality, long-lasting, type of processor, RAM, etc. Moreover, for the same features, different companies fix different prices. The buyer therefore faces a problem about which one will be the best mobile phone. Their mind is in a dilemma about buying the product. Thus, hesitation appears in the mind, so the problem belongs to the bipolar neutrosophic fuzzy environment domain. The person will accept some suggestions from friends or from other persons to form an opinion about the product. They will give their own opinion according to their choice after giving some weight to the features. This problem now becomes a multi-criteria decision-making problem and the person wants to find the best alternatives.

Figure 5. (a) Shaded Region of hesitant portion (Step I); (b) Shaded Region of hesitant portion (Step II).

Figure 6. (a) Shaded Region of truth portion (Step I); (b) Shaded Region of truth portion (Step II).
In this section, we consider a multi-criteria decision-making problem where a finite number of different decision makers is available and according to their viewpoint, we must find the best alternative. To do so, we constructed an algorithm based on the weighted mean and normalization approach so that we could solve the uncertainty problem very easily. Then, using the result of the de-bipolarization value, we were able to choose the best alternative among all of them.

6.1. Illustration of the MCDM Problem

We consider the problem as follows:

Let \( P = \{ P_1, P_2, P_3, \ldots \ldots P_m \} \) be the distinct alternative set and \( Q = \{ Q_1, Q_2, Q_3, \ldots \ldots Q_n \} \) be the distinct attribute set, respectively. Let \( W = \{ W_1, W_2, W_3, \ldots \ldots W_n \} \) be the weight set associated with the attributes \( Q \), where each \( W_i \geq 0 \) and also satisfies the relation \( \sum_{i=1}^{n} W_i = 1 \). We also consider the set of decision makers \( D = \{ D_1, D_2, D_3, \ldots \ldots D_K \} \) associated with alternatives whose weight vector is defined as \( \vartheta = \{ \vartheta_1, \vartheta_2, \vartheta_3, \ldots \ldots \vartheta_k \} \), where each \( \vartheta_i \geq 0 \) and also satisfies the relation \( \sum_{i=1}^{k} \vartheta_i = 1 \); this weight vector will be selected according to the decision maker’s quality of judgment, knowledge, thinking power, etc.

6.2. Weighted Mean and Normalisation Algorithm of the MCDM Problem

Step 1: Creation of decision matrices

First, we create the decision matrices for each decision maker’s choice associated with alternatives versus attribute functions. We consider the member of the matrices in the bipolar neutrosophic environment, so all \( a_{ij} \)’s are a member of the bipolar neutrosophic set. The associated matrix is defined as follows:

\[
X^k = \begin{bmatrix}
Q_1 & Q_2 & Q_3 & \ldots & Q_n \\
\vdots & \vdots & \vdots & & \vdots \\
P_1 & a_{11}^k & a_{12}^k & a_{13}^k & \ldots & a_{1n}^k \\
P_2 & a_{21}^k & a_{22}^k & a_{23}^k & \ldots & a_{2n}^k \\
P_3 & \vdots & \vdots & \vdots & & \vdots \\
P_m & a_{m1}^k & a_{m2}^k & a_{m3}^k & \ldots & a_{mn}^k \\
\end{bmatrix}
\]  

(1)

Step 2: Creation of weighted single-decision matrix

To obtain a single group decision matrix, we use the operation \( a_{ij}' = \left\{ \sum_{i=1}^{n} W_i X_i \right\} \) for each individual decision matrix \( X_i \), and thus we get the new matrix as follows:

\[
X = \begin{bmatrix}
Q_1 & Q_2 & Q_3 & \ldots & Q_n \\
\vdots & \vdots & \vdots & & \vdots \\
P_1 & a_{11}' & a_{12}' & a_{13}' & \ldots & a_{1n}' \\
P_2 & a_{21}' & a_{22}' & a_{23}' & \ldots & a_{2n}' \\
P_3 & \vdots & \vdots & \vdots & & \vdots \\
P_m & a_{m1}' & a_{m2}' & a_{m3}' & \ldots & a_{mn}' \\
\end{bmatrix}
\]  

(2)

Step 3: Creation of weighted priority matrix using weight vector

To obtain single-column decision matrix, we use the operation \( a_{ij}'' = \left\{ \sum_{i=1}^{n} \vartheta_i a_{pi}' \right\} \) for each individual column, and thus we get the decision matrix as follows:
environment, so all versus attribute functions. We consider the member of the matrices in the bipolar neutrosophic as follows:

\[
X = \begin{pmatrix}
\cdot & Q_1 \\
\cdot & p_1 a_{11} \\
\cdot & p_2 a_{21} \\
\cdot & \cdot \\
\cdot & p_m a_{m1}
\end{pmatrix}
\] (3)

Step 4: Ranking
Now, we consider the de-bipolarization value and convert the matrix (3) into crisp form so that we can evaluate the best alternative corresponding to the best attributes. We consider the score values according to the increasing order and choose the best fit result. The highest value gives us the best result and lowest one, the worst one.

Flowchart: The flowchart is given in Figure 7:

6.3. Illustrative Example
Let us consider a problem related to three different products and their distinctive attributes. We know that many products are available in the market and they have different components with different qualities and features. Therefore, this is multi-criteria decision-making problem with different types of decision makers. We define the problem as follows: \( P_1 = \text{Product 1}, P_2 = \text{Product 2}, P_3 = \text{Product 3} \) are the alternatives; \( Q_1 = \text{Price}, Q_2 = \text{Longibility}, Q_3 = \text{Service} \) are the attributes.

We consider there are three types of decision makers: \( D_1 = \text{Young people}, D_2 = \text{Middle-aged people}, D_3 = \text{Older people} \) having weight function \( D = \{0.35, 0.30, 0.35\} \) and we also consider the weight vector associated with the attribute function \( \delta = \{0.33, 0.30, 0.37\} \). A verbal matrix is created by the designer to assist the decision maker in the creation of their decision matrix. The verbal phrases for all the different attributes are listed in Table 2.

| Cases | Attribute               | Verbal Phrase                        |
|-------|-------------------------|--------------------------------------|
| 1     | Price of the product    | Very high (VH), High (H), Intermediate (I), Small (S), Very small (VS) |
| 2     | Legibility of the product | Very high (VHI), High (H), Mid (M), Low (L), Very low (VL) |
| 3     | Service of the product  | Very high (VHI), High (H), Mid (M), Low (L), Very low (VL) |

Step 1:
We consider the matrices according to each decision maker’s choice related to alternatives and attribute functions (See Table 3). All the members of the matrices are of a spherical neutrosophic nature. Therefore, the decision matrices are as follows:

\[
M^1 = \begin{pmatrix}
\vdots & Q_1 & Q_2 & Q_3 \\
P_1 & <1,2,3;0,5,1,5,2.5;1,2.5,3.5> & <1,5,8;0,5,3,6;3.5,7,5> & <1,5,9;0,6,2,6;2.6,5,6,9> \\
P_2 & <0.7,2,4;0.5,1,5,3,4.5> & <2,4,6;1,5,2.5,3.5;3.5,7> & <1.5,3,5,5.5;1,3,5,2.5,4.6> \\
P_3 & <1.4,7;0.2,3,3.5,5,7,5> & <1.5,2.5,3.5,1,2.3;2,3,4> & <2.4,6;1,5,3,5,5.5;5.5,6,5> \\
\end{pmatrix}
\]

For decision maker \(D_1\)

\[
M^2 = \begin{pmatrix}
\vdots & Q_1 & Q_2 & Q_3 \\
P_1 & <1,3,5;0,5,2.5,4.5;2,4,6> & 2,5,7;1,3,5,5.5;4,6,8> & <1,2,3;0,5,1,5,2.5;1,3,5,3.5> \\
P_2 & <2,4,6;1,3,5,5,6,7> & <1.5,3,5,5.5;1,3,5,2.5,4.6> & <1.4,7;0,5,2,3,3.9,5.5,7.5> \\
P_3 & <1.5,3,5,4.5,1.3,5,2.5,4.6> & <1.4,7,0.5,2,3,3.5,5.5,7.5> & <1,5,9;0,6,2,6,2.6,5,9.5> \\
\end{pmatrix}
\]

For decision maker \(D_2\)

\[
M^3 = \begin{pmatrix}
\vdots & Q_1 & Q_2 & Q_3 \\
P_1 & <2,5,7,1.5,3,5,5,4.6,8> & <1.5,3,5,5.5;1,3,5,2.5,4.6> & <1,1.5,4;0.5,1,2.5;1.25,3.4,5> \\
P_2 & <1.2,3;0,5,1.5,2.5;1.3,2,5,3.5> & <1.5,8;1,5,3,6;4,6.8,5> & <1,5,9;0.6,2,6,2.6,5,9.5> \\
P_3 & <0.6,2.4,0.3,1.25,1.3,5.5,4.5> & <0.5,2.5,4.5,1,2,3,1,3,5,5.5> & <1,3.5;0.5,2,5,3.5,2.5,4.6> \\
\end{pmatrix}
\]

For decision maker \(D_3\)

| Alternatives/Attributes | C^1 | C^2 | C^3 |
|-------------------------|-----|-----|-----|
| A_1                     | L   | M   | H   |
| A_2                     | VL  | M   | I   |
| A_3                     | L   | I   | VH  |

Table 3. Verbal matrix.

Step 2: Creation of weighted mean single-decision matrix

\[
M = \begin{pmatrix}
\vdots & Q_1 & Q_2 & Q_3 \\
P_1 & <1.35,3.35,5,0.85,2.5,4.15,2.45,4.18,5.83> & <1.48,4.5,6,8,0.98,3.2,5.5,3.3,5.7,7.5> & <1.2,9,5.5,0.5,1.5,3.8,1.5,4.7,5> \\
P_2 & <1.2,2.6,4.25,0.6,1.78,3.25,2.4,3.72,4.9> & <1.5,4.2,6.5,1.35,2.8,4.8,3.2,5,7.3> & <1.18,4.18,7.2,0.7,4.8,2.4,8.6,5,4.7,7> \\
P_3 & <1.01,3.52,0.58,1.95,3.4,2.5,4.18,6> & <1.3,5,0.85,2.3,2.26,4.5,6> & <1.35,4.6,5,0.9,2.8,5.2,7.3,7.3> \\
\end{pmatrix}
\]

Step 3: Creation of weighted Priority matrix using weight vector

\[
M = \begin{pmatrix}
\vdots & 1.26,3.53,5.7,0.76,2.4,4.43;2.35,5.47,6.4 > \\
& 1.28,3.7,6.0,0.88,2.31,4.3;2.74,4.72,6.66 > \\
& 1.13,3.38,5.6,0.78,2.28,3.9;2.51,4.54,6.4 > \\
\end{pmatrix}
\]

Step 4: Ranking

Now, we consider the De-Bipolarization value defined in Section 5.1 and converts the triangular bipolar neutrosophic numbers into a crisp one. Thus, we obtain the final ideal decision matrix as follows:

\[
M = \begin{pmatrix}
\vdots & <6.983> \\
& <7.22> \\
& <6.79> \\
\end{pmatrix}
\]

Now, after arranging the numbers in ascending order, we obtain \(6.79 < 6.983 < 7.22\). Thus, the ranking of the priority alternatives is \(P_2 > P_1 > P_3\).

6.4. Results and Sensitivity Analysis

A sensitivity analysis was done to understand how the attribute weights of each criterion affected the relative matrix and their ranking. The main idea of a sensitivity analysis is to interchange weights
of the attribute values while keeping the rest of the terms fixed. Table 4 shows the sensitivity results. Also input data is in Figure 8 and output result shown in Figure 9.

### Table 4. Sensitivity analysis chart.

| Attribute Weight | Final Decision Matrix | Ordering     |
|------------------|------------------------|--------------|
| <(0.33,0.30,0.37)> | <(6.983,7.22,6.79)> | P₂ > P₁ > P₃ |
| <(0.25,0.30,0.45)> | <(6.87,7.37,6.95)> | P₂ > P₃ > P₁ |
| <(0.35,0.25,0.40)> | <(6.85,7.15,6.68)> | P₂ > P₃ > P₁ |
| <(0.40,0.30,0.30)> | <(7.04,7.18,6.58)> | P₂ > P₁ > P₃ |
| <(0.20,0.30,0.50)> | <(6.82,7.24,7.03)> | P₂ > P₃ > P₁ |

![Figure 8. Input table with the associated weight of the different parameters.](image1)

![Figure 9. Output table of the preferred priority alternatives.](image2)
6.5. Comparison with Other Established Work:

We compare our work with other previous work described by the authors [11,18,49] for finding the best alternatives we observed that in each cases $P_2$ becomes the best alternative (see Table 5).

| Approach      | Ranking       |
|---------------|---------------|
| Deli [11]     | $P_2 > P_1 > P_3$ |
| Aslam [18]    | $P_2 > P_3 > P_1$ |
| Garg [49]     | $P_2 > P_1 > P_3$ |
| Our approach  | See Table 4   |

Remark 1. The novelty of this paper is to understand the behavior of the weights of different criteria on the ranking of the alternatives. For this reason, a sensitivity analysis was performed by interchanging the weights of the criteria. Using a sensitivity analysis, we observe a certain level of change in the attribute values and that, ultimately, $P_2$ is the best alternative, whereas the other alternatives change their ordering.

7. Conclusions and Future Research Scope

The concept of a triangular bipolar neutrosophic number is interesting and pragmatic and has a practical use in the current research arena. The formulation of a distinctive type of triangular bipolar neutrosophic number of category-1, -2, or -3 and of a de-polarization technique is essential for researchers who deal with the ideas of uncertainty and vagueness. To solve any kind of multi-criteria decision-making problem, one can also apply the current method discussed above. In this paper, we adopted the concept of a bipolar neutrosophic number from different viewpoints and perspectives. We also used the idea of linear as well as nonlinear form with truth, false, and hesitant functions in the case of a triangular bipolar neutrosophic number when the membership functions were related to each other. The concept of de-bipolarization was very helpful when we wanted to find the best result in the case of different decision-making problems, in which the number of alternatives and attributes was finite and different decision makers were involved. Finally in the example portion we also consider a sensitivity analysis and also did comparison with the other paper’s result to tally our proposed work and we can conclude that our result is more suitable as we consider the updated De-Bipolarized value in the problem to tackle the multi criteria decision making problem. Furthermore, researchers can apply this concept of a bipolar number in various fields, such as engineering problems, diagnosis problems, mathematical modeling, among others.

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