Some aggregation operators of neutrosophic Z-numbers and their multicriteria decision making method

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Abstract
As the generalization of the classical fuzzy number, the concept of Z-number introduced by Zadeh indicates more ability to depict the human knowledge and judgments of both restraint and reliability as an order pair of fuzzy numbers. In indeterminacy and inconsistent environment, a neutrosophic set is described by the truth, falsity, and indeterminacy degrees, but they lack measures related to reliability. To describe the hybrid information of combining the truth, falsity and indeterminacy degrees with their corresponding reliability degrees, this paper first proposes the concept of a neutrosophic Z-number (NZN) set, which is a new framework of neutrosophic values combined with the neutrosophic measures of reliability, as the generalization of the Z-number and the neutrosophic set. Then, we define the operations of neutrosophic Z-numbers (NZNs) and a score function for ranking NZNs. Next, we present NZN weighted arithmetic averaging (NZNWAA) and NZN weighted geometric averaging (NZNWGA) operators to aggregate NZN information and investigate their properties. Regarding the NZNWAA and NZNWGA operators and the score function, a multicriteria decision making (MDM) approach is developed in the NZN environment. Finally, an illustrative example about the selection problem of business partners is given to demonstrate the applicability and effectiveness of the developed MDM approach in NZN setting.

Keywords Neutrosophic Z-number set · Neutrosophic Z-number · Neutrosophic Z-number weighted arithmetic averaging operator · Neutrosophic Z-number weighted geometric averaging operator · Multicriteria decision making

Introduction
It is known that fuzzy sets proposed by Zadeh [1] play an essential role in the current scientific and technical applications [2–7]. In 2011, Zadeh [8] further introduced the concept of Z-numbers to describe the restraint and reliability of the evaluation by an order pair of fuzzy numbers in uncertain situations. Compared with the classical fuzzy number, it is a more generalized notion closely related to reliability. Hence, the Z-number implies more ability to describe the human knowledge and judgments by an order pair of fuzzy numbers corresponding to the restriction and reliability. Since then, it has obtained a lot of attentions. Some researchers presented theoretical studies of Z-numbers, like Z*-numbers [9], arithmetic operations of discrete and continuous Z-numbers [10, 11], modeling of Z-number [12], approximate reasoning of Z-numbers [13], functions based on a Z-number set [14], total utility of Z-numbers [15] and so on; while other researchers developed some applications of Z-numbers, such as Z-evaluations [16], sensor data fusion using Z-numbers [17], decision making approaches with Z-numbers [18–24], Z-numbers-based stable strategies analysis in evolutionary game [25], Z-numbers-based medicine selection of the patients with mild symptoms of the COVID-19 [26], Z-numbers-based environmental assessment under uncertainty [27] and so on.

In indeterminate and inconsistent environment, neutrosophic sets [28, 29] are described independently by the truth, falsity, and indeterminacy membership degrees, but...
the aforementioned Z-numbers cannot depict them. Then, neutrosophic sets have been applied in various areas, such as image processing [30], decision making [31–34], medical diagnosis [35–37], and mechanical fault diagnosis [38]. However, the truth, falsity, and indeterminacy membership degrees in the neutrosophic set lack the reliability measures related to them. If the Z-number notion is extended to the neutrosophic set, we can describe the hybrid information of combining the truth, falsity and indeterminacy degrees with their corresponding reliability degrees by three order pairs of fuzzy numbers. In multicriteria decision making (MDM) problems, the information expressions and decision making methods are vital research topics [39–42]. Motivated based on the ideas of combining the Z-number with the neutrosophic set and enhancing MDM reliability, the objects of this study are to present a more generalized neutrosophic notion closely related to reliability and to use it for MDM problems. To do so, this paper proposes the concept of a neutrosophic Z-number (NZN) set, which is a new framework of neutrosophic values combined with the neutrosophic measures of reliability, as the generalization of the Z-number and the neutrosophic set. Then, we define the operations of neutrosophic Z-numbers (NZNs) and a score function for ranking NZNs and propose NZN weighted arithmetic averaging (NZNWAA) and NZN weighted geometric averaging (NZNWGAA) operators to aggregate NZN information. Regarding the NZNWAA and NZNWGAA operators and the score function, a MDM approach is developed in the NZN environment. An illustrative example is used to demonstrate the applicability and effectiveness of the developed MDM approach in NZN setting. Lastly, conclusions and further study are presented in section “Conclusion”.

Neutrosophic Z-number set

In 2011, Zadeh [8] firstly introduced the concept of Z-number by an order pair of fuzzy numbers $Z = (V, R)$ associated with a real-valued uncertain variable $X$, where the first component $V$ is a fuzzy restriction on the values that $X$ can take and the second component $R$ is a measure of reliability for $V$.

Based on an extension of the Z-number concept [8] and the neutrosophic set, we can give the definition of a NZN set.

**Definition 1** Set $X$ as a universe set. Then a NZN set in a universe set $X$ is defined as the following form:

$$S_Z = \{ (x, T(V, R)(x), I(V, R)(x), F(V, R)(x)) | x \in X \},$$

where $T(V, R)(x) = \langle TV(x), FR(x) \rangle$, $I(V, R)(x) = \langle IV(x), IR(x) \rangle$, $F(V, R)(x) = \langle FV(x), FR(x) \rangle$: $X \rightarrow [0, 1]^2$ are the order pairs of truth, indeterminacy and falsity fuzzy values, then the first component $V$ is neutrosophic values in a universe set $X$ and the second component $R$ is neutrosophic measures of reliability for $V$, along with the conditions $0 \leq TV(x) + IV(x) + FV(x) \leq 3$ and $0 \leq TR(x) + IR(x) + FR(x) \leq 3$.

For the convenient representation, the element $(x, T(V, R)(x), I(V, R)(x), F(V, R)(x))$ in $S_Z$ is simply denoted as $s_Z = \langle T(V, R), I(V, R), F(V, R) \rangle = \langle TV, TR \rangle, \langle IV, IR \rangle, \langle FV, FR \rangle$, which is named NZN.

**Definition 2** Let $s_{Z1} = \langle T_1(V, R), I_1(V, R), F_1(V, R) \rangle = \langle TV_1, TR_1 \rangle, \langle IV_1, IR_1 \rangle, \langle FV_1, FR_1 \rangle$ and $s_{Z2} = \langle T_2(V, R), I_2(V, R), F_2(V, R) \rangle = \langle TV_2, TR_2 \rangle, \langle IV_2, IR_2 \rangle, \langle FV_2, FR_2 \rangle$ be two NZNs and $\lambda > 0$. Then, we give the following relations:

1. $s_{Z1} \supseteq s_{Z2} \iff TV_1 \geq TV_2, TR_1 \geq TR_2, IV_1 \leq IV_2, IR_1 \leq IR_2, FV_1 \leq FV_2, and FR_1 \leq FR_2$.
2. $s_{Z1} = s_{Z2} \iff s_{Z1} \supseteq s_{Z2} and s_{Z2} \supseteq s_{Z1}$.
To compare NZNs $s_{Zi} = \langle T_i(V, R), I_i(V, R), F_i(V, R) \rangle$ ($i = 1, 2$), we introduce a score function:

$$Y(s_{Zi}) = \frac{2 + T_iV_i - I_iV_i - F_iV_i}{3}$$

for $Y(s_{Zi}) \in [0, 1]$ (1)

Thus, if $Y(s_{Zi}) \geq Y(s_{Zj})$, there is the ranking $s_{Zi} \geq s_{Zj}$.

**Example 1** Set two NZNs as $s_{Z1} = <(0.7, 0.8), (0.1, 0.7), (0.3, 0.8)>$ and $s_{Z2} = <(0.6, 0.9), (0.3, 0.8), (0.2, 0.7)>$. Then, their ranking is given as follows:

By Eq. (1), we have $Y(s_{Z1}) = (2 + 0.7 \times 0.8 - 0.1 \times 0.7 - 0.3 \times 0.8)/3 = 0.75$ and $Y(s_{Z2}) = (2 + 0.6 \times 0.9 - 0.3 \times 0.8 - 0.2 \times 0.7)/3 = 0.72$. Since $Y(s_{Z1}) > Y(s_{Z2})$, their ranking is $s_{Z1} > s_{Z2}$.

**Two weighted aggregation operators of neutrosophic Z-numbers**

Based on the operations (6)–(9) in Definition 2, we can propose the two weighted aggregation operators of NZNs in this section.

**NZNWAA operator**

Based on the operations (6) and (8) in Definition 2, we can present the NZNWAA operator of NZNs.

**Definition 3** Let $s_{Zi} = \langle T_i(V, R), I_i(V, R), F_i(V, R) \rangle$ ($i = 1, 2, \ldots, n$) be a group of NZNs and NZNWAA: $\Omega^n \rightarrow \Omega$. Then, the NZNWAA operator is defined as

$$\text{NZNWAA}(s_{Z1}, s_{Z2}, \ldots, s_{Zn}) = \sum_{i=1}^{n} \lambda_i s_{Zi},$$

where $\lambda_i (i = 1, 2, \ldots, n)$ is the weight of $s_{Zi}$ with $0 \leq \lambda_i \leq 1$ and $\sum_{i=1}^{n} \lambda_i = 1$.

**Theorem 1** Let $s_{Zi} = \langle T_i(V, R), I_i(V, R), F_i(V, R) \rangle$ ($i = 1, 2, \ldots, n$) be a group of NZNs. Then, the collected value of the NZNWAA operator is a NZN, which is obtained by the following formula:

$$\text{NZNWAA}(s_{Z1}, s_{Z2}, \ldots, s_{Zn}) = \sum_{i=1}^{n} \lambda_i s_{Zi}$$

$$= \left( \prod_{i=1}^{n} I_i^\lambda, \prod_{i=1}^{n} F_i^\lambda \right),$$

where $\lambda_i$ is the weight of $s_{Zi} (i = 1, 2, \ldots, n)$ with $0 \leq \lambda_i \leq 1$ and $\sum_{i=1}^{n} \lambda_i = 1$.

**Proof** Regarding mathematical induction, Eq. (3) is verified below.

1. If $n = 2$, according to the operations (6) and (8) in Definition 2 we yield the following result:
2. If $n = m$, Eq. (3) has the following form:

\[ \text{NZNWAAs}_Z(s_{Z1}, s_{Z2}, \ldots, s_{Zm}) = \sum_{i=1}^{m} \lambda_i s_{Zi} \]

\[ = \left( 1 - \prod_{i=1}^{m} (1 - T_{Vi})^{\lambda_i}, 1 - \prod_{i=1}^{m} (1 - T_{Ri})^{\lambda_i} \right), \]

\[ \left( \prod_{i=1}^{m} I_{Vi}, \prod_{i=1}^{m} I_{Ri} \right) \left( \prod_{i=1}^{m} F_{Vi}, \prod_{i=1}^{m} F_{Ri} \right) \right). \]  

(5)

3. If $n = m + 1$, according to the operations (6) and (8) in Definition 2 and Eqs. (4) and (5), there is the following result:

\[ \text{NZNWAAs}_Z(s_{Z1}, s_{Z2}, \ldots, s_{Zm}, s_{Zm+1}) \]

\[ = \sum_{i=1}^{m} \lambda_i s_{Zi} \oplus \lambda_{m+1} s_{Zm+1} \]

\[ = \left( 1 - \prod_{i=1}^{m+1} (1 - T_{Vi})^{\lambda_i}, 1 - \prod_{i=1}^{m+1} (1 - T_{Ri})^{\lambda_i} \right), \]

\[ \left( \prod_{i=1}^{m+1} I_{Vi}, \prod_{i=1}^{m+1} I_{Ri} \right) \left( \prod_{i=1}^{m+1} F_{Vi}, \prod_{i=1}^{m+1} F_{Ri} \right) \right). \]  

Based on the above results, Eq. (3) can keep for any $n$. Thus, the verification is finished. \(\square\)

Theorem 2 The NZNWAAs operator of Eq. (3) implies the following properties:

1. Idempotency: Set $s_{Zi} = (T_i(V, R), I_i(V, R), F_i(V, R)) = (T_{Vi}, T_{Ri}), (I_{Vi}, I_{Ri}), (F_{Vi}, F_{Ri}) \right) (i = 1, 2, \ldots, n)$ as a group of NZNs. If $s_{Z} = s_{Z}(i = 1, 2, \ldots, n)$, there is NZNWAAs$_{Z}(s_{Z1}, s_{Z2}, \ldots, s_{Zn}) = s_{Z}$.  

2. Boundedness: Set $s_{Zi} = (T_i(V, R), I_i(V, R), F_i(V, R)) = (T_{Vi}, T_{Ri}), (I_{Vi}, I_{Ri}), (F_{Vi}, F_{Ri}) \right) (i = 1, 2, \ldots, n)$ as a group of NZNs and let

\[ s_{Z_{min}} = \left\{ \left( \text{min}_i(T_i(V, R)), \text{max}_i(I_i(V, R)), \text{max}_i(F_i(V, R)) \right) \right\} \]

\[ = \left\{ \left( \text{min}_i(T_{Vi}), \text{min}_i(T_{Ri}) \right), \left( \text{max}_i(I_{Vi}), \text{max}_i(I_{Ri}) \right), \left( \text{max}_i(F_{Vi}), \text{max}_i(F_{Ri}) \right) \right\}. \]

\[ s_{Z_{max}} = \left\{ \left( \text{max}_i(T_i(V, R)), \text{min}_i(I_i(V, R)), \text{min}_i(F_i(V, R)) \right) \right\} \]

\[ = \left\{ \left( \text{max}_i(T_{Vi}), \text{max}_i(T_{Ri}) \right), \left( \text{min}_i(I_{Vi}), \text{min}_i(I_{Ri}) \right), \left( \text{min}_i(F_{Vi}), \text{min}_i(F_{Ri}) \right) \right\}. \]

Then, $s_{Z_{min}} \leq \text{NZNWAAs}_Z(s_{Z1}, s_{Z2}, \ldots, s_{Zn}) \leq s_{Z_{max}}$ can keep.

3. Monotonicity: Set $s_{Zi} = (T_i(V, R), I_i(V, R), F_i(V, R)) = (T_{Vi}, T_{Ri}), (I_{Vi}, I_{Ri}), (F_{Vi}, F_{Ri}) \right) (i = 1, 2, \ldots, n)$ as a group of NZNs. When $s_{Z} \leq s_{Z}^*$, there is NZNWAAs$_{Z}(s_{Z1}, s_{Z2}, \ldots, s_{Zn}) \leq \text{NZNWAAs}_Z(s_{Z1}^*, s_{Z2}^*, \ldots, s_{Zn}^*)$.  

\[ \text{NZNWAAs}_Z(s_{Z1}, s_{Z2}, \ldots, s_{Zm}) = \lambda_1 s_{Z1} \oplus \lambda_2 s_{Z2} \]

\[ = \left( 1 - (1 - T_{Vi})^{\lambda_1}, 1 - (1 - T_{Ri})^{\lambda_2} \right), \]

\[ \left( F_{Vi}^{\lambda_1}, F_{Ri}^{\lambda_2} \right). \]  

(4)
Proof. 1. If \( s_{Zi} = s_Z (i = 1, 2, \ldots, n) \), the result of Eq. (3) is given by

\[
\text{NZNWAA}(s_{Z1}, s_{Z2}, \ldots, s_{Zn}) = \sum_{i=1}^{n} \lambda_i s_{Zi} = \left(1 - \prod_{i=1}^{n} (1 - T_{Vi})^{\lambda_i}, 1 - \prod_{i=1}^{n} (1 - T_{Ri})^{\lambda_i}\right).
\]

Thus, the verification of all properties is completed. \( \square \)

2. Since \( s_{Z\min} \) and \( s_{Z\max} \) are given by the minimum NZN and the maximum NZN, the inequality \( s_{Z\min} \leq s_Z \leq s_{Z\max} \) exists. Thus, there is \( \sum_{i=1}^{n} \lambda_i s_{Z\min} \leq \sum_{i=1}^{n} \lambda_i s_{Zi} \leq \sum_{i=1}^{n} \lambda_i s_{Z\max} \). Based on the above property (1), \( s_{Z\min} \leq \sum_{i=1}^{n} \lambda_i s_{Zi} \leq s_{Z\max} \) can exist, i.e., there is \( s_{Z\min} \leq \text{NZNWAA}(s_{Z1}, s_{Z2}, \ldots, s_{Zn}) \leq s_{Z\max} \).

3. Since \( s_{Zi} \leq s_{Z\max} \), there is \( \sum_{i=1}^{n} \lambda_i s_{Zi} \leq \sum_{i=1}^{n} \lambda_i s_{Z\max} \), i.e., \( \text{NZNWAA}(s_{Z1}, s_{Z2}, \ldots, s_{Zn}) \leq \text{NZNWAA}(s_{Z\min}, s_{Z\min}, \ldots, s_{Z\min}) \).

Thus, the verification of all properties is completed. \( \square \)

NZNWGA operator

Based on the operations (7) and (9) in Definition 2, we can present the NZNWGA operator of NZNs.

Definition 4 Let \( s_{Zi} = (T_i(V, R), I_i(V, R), F_i(V, R)) = (\langle V_i, R_i \rangle, (I_i(V, R), (F_i(V, R))) (i = 1, 2, \ldots, n) \) be a group of NZNs and NZNWGA: \( \Omega^n \rightarrow \Omega \). Then, the NZNWGA operator is defined as

\[
\text{NZNWGA}(s_{Z1}, s_{Z2}, \ldots, s_{Zn}) = \prod_{i=1}^{n} \lambda_i s_{Zi} = \left(\prod_{i=1}^{n} T_{Vi}^{\lambda_i}, \prod_{i=1}^{n} F_{Vi}^{\lambda_i}\right) (\lambda_i = 1, 2, \ldots, n)
\]

where \( \lambda_i (i = 1, 2, \ldots, n) \) is the weight of \( s_{Zi} \) with \( 0 \leq \lambda_i \leq 1 \) in \( \sum_{i=1}^{n} \lambda_i = 1 \).

Theorem 3 Let \( s_{Zi} = (T_i(V, R), I_i(V, R), F_i(V, R)) = (\langle V_i, R_i \rangle, (I_i(V, R), (F_i(V, R))) (i = 1, 2, \ldots, n) \) be a group of NZNs. Then, the collected value of the NZNWGA operator is a NZN, which is obtained by the following formula:

\[
\text{NZNWGA}(s_{Z1}, s_{Z2}, \ldots, s_{Zn}) = \prod_{i=1}^{n} \lambda_i s_{Zi} = \left(\prod_{i=1}^{n} T_{Vi}^{\lambda_i}, \prod_{i=1}^{n} F_{Vi}^{\lambda_i}\right)
\]

where \( \lambda_i \) is the weight of \( s_{Zi} (i = 1, 2, \ldots, n) \) with \( 0 \leq \lambda_i \leq 1 \) and \( \sum_{i=1}^{n} \lambda_i = 1 \).

By the similar verification process of Theorem 1, we can also verify that the NZNWGA operator of Eq. (7) is true, which is not repeated here.

Theorem 4 The NZNWGA operator of Eq. (7) also implies the following properties:

1. Idempotency: Set \( s_{Zi} = (T_i(V, R), I_i(V, R), F_i(V, R)) = (\langle V_i, R_i \rangle, (I_i(V, R), (F_i(V, R))) (i = 1, 2, \ldots, n) \) as a group of NZNs. If \( s_{Zi} = s_Z (i = 1, 2, \ldots, n) \), there is \( \text{NZNWGA}(s_{Z1}, s_{Z2}, \ldots, s_{Zn}) = s_Z \).

2. Boundedness: Set \( s_{Zi} = (T_i(V, R), I_i(V, R), F_i(V, R)) = (\langle V_i, R_i \rangle, (I_i(V, R), (F_i(V, R))) (i = 1, 2, \ldots, n) \) as a group of NZNs and let

\[
s_{Z\min} = \left(\min(T_{Vi}), \min(I_{Vi}), \max(F_{Vi})\right), \quad s_{Z\max} = \left(\max(T_{Vi}), \min(I_{Vi}), \max(F_{Vi})\right)
\]

where \( \lambda_i (i = 1, 2, \ldots, n) \) is the weight of \( s_{Zi} \) with \( 0 \leq \lambda_i \leq 1 \) in \( \sum_{i=1}^{n} \lambda_i = 1 \).
\[ s_{Z\text{max}} = \left( \max_{i}(T_i(V, R)), \min_{i}(I_i(V, R)), \min_{i}(F_i(V, R)) \right) \]
\[ = \left( \left( \max_{i}(T_{V_i}), \max_{i}(T_{R_i}) \right), \left( \min_{i}(I_{V_i}), \min_{i}(I_{R_i}) \right), \left( \min_{i}(F_{V_i}), \min_{i}(F_{R_i}) \right) \right). \]

Then, \(s_{Z\text{min}} \leq \text{NZNWGA}(s_{Z1}, s_{Z2}, \ldots, s_{Zn}) \leq s_{Z\text{max}}\) can keep.

3. Monotonicity: Set \(s_{Zi} = (T_i(V, R), I_i(V, R), F_i(V, R))\) and \(s^*_Zi = (T^*_i(V, R), I^*_i(V, R), F^*_i(V, R))\) as two groups of NZNs. When \(s_{Zi} \leq s^*_Zi\), there is \(\text{NZNWGA}(s_{Z1}, s_{Z2}, \ldots, s_{Zn}) \leq \text{NZNWGA}(s^*_{Z1}, s^*_{Z2}, \ldots, s^*_{Zn})\).

Obviously, the above properties corresponding to the NZNWGA operator can also be verified by the similar verification process of Theorem 2, which is not repeated here.

### MDM approach using the NZNWAA and NZNWGA operators and the score function

Regarding the proposed NZNWAA and NZNWGA operators and the score function, this section develops a MDM approach to solve MDM problems with the evaluation information of both neutrosophic values and neutrosophic measures of reliability related to the neutrosophic values in NZN setting.

Suppose that in a MDM problem a set of \(m\) alternatives \(Q = \{Q_1, Q_2, \ldots, Q_m\}\) is presented and assessed by a set of \(n\) criteria \(X = \{x_1, x_2, \ldots, x_n\}\). Then, the importance of each criterion \(x_i (i = 1, 2, \ldots, n)\) is considered by the weight \(\lambda_i\), which is constructed as the weight vector \(\lambda = (\lambda_1, \lambda_2, \ldots, \lambda_n)\). Decision makers are requested to give the suitability assessment of each criterion \(x_i (i = 1, 2, \ldots, n)\) for each alternative \(Q_j (j = 1, 2, \ldots, m)\) by both the truth, falsity, indeterminacy fuzzy values and the measures of corresponding reliabilities, which are constructed as a NZN \(s_{Zji} = (T_{ji}(V, R), I_{ji}(V, R), F_{ji}(V, R)) = (T_{Vji}, T_{Rji}), (I_{Vji}, I_{Rji}), (F_{Vji}, F_{Rji})\), where \(T_{Vji}, I_{Vji}, F_{Vji} \in [0, 1]\) and \(T_{Rji}, I_{Rji}, F_{Rji} \in [0, 1]\). Thus, the decision matrix of NZNs can be represented as \(S_Z = \{(s_{Zji})_{m \times n}\}\). In this MDM problem, the decision process is described below:

**Step 1** Based on Eq. (3) or Eq. (7), the overall NZN \(s_{Zj}\) is obtained by

\[ s_{Zj} = \text{NZNWAA}(s_{Zj1}, s_{Zj2}, \ldots, s_{Zjn}) = \sum_{i=1}^{n} \lambda_i s_{Zji} \]
\[ = \left( \left(1 - \prod_{i=1}^{n} (1 - T_{Vji})^{\lambda_i}, 1 - \prod_{i=1}^{n} (1 - T_{Rji})^{\lambda_i}\right), \left(\prod_{i=1}^{n} I_{Vji}^{\lambda_i}, \prod_{i=1}^{n} I_{Rji}^{\lambda_i}\right), \left(\prod_{i=1}^{n} F_{Vji}^{\lambda_i}, \prod_{i=1}^{n} F_{Rji}^{\lambda_i}\right) \right) \]

or

\[ s_{Zj} = \text{NZNWGA}(s_{Zj1}, s_{Zj2}, \ldots, s_{Zjn}) = \prod_{i=1}^{n} s_{Zji}^\lambda \]
\[ = \left( \left(1 - \prod_{i=1}^{n} (1 - T_{Vji})^{\lambda_i}, 1 - \prod_{i=1}^{n} (1 - T_{Rji})^{\lambda_i}\right), \left(1 - \prod_{i=1}^{n} (1 - I_{Vji})^{\lambda_i}, 1 - \prod_{i=1}^{n} (1 - I_{Rji})^{\lambda_i}\right), \left(1 - \prod_{i=1}^{n} (1 - F_{Vji})^{\lambda_i}, 1 - \prod_{i=1}^{n} (1 - F_{Rji})^{\lambda_i}\right) \right). \]

**Step 2** The score values of \(Y(s_{Zj}) (j = 1, 2, \ldots, m)\) are calculated using Eq. (1)

**Step 3** The alternatives are ranked based on the score values and the best one is chosen among ones

**Step 4** End

### An illustrative example and relative comparative analysis

#### An illustrative example

This part provides an illustrative example about the selection problem of business partners to demonstrate the applicability and effectiveness of the developed MDM approach with NZN information.

Suppose a manufacturing company needs to choose a suitable supplier in potential business partners. The expert panel provides a set of four suppliers/alternatives \(Q = \{Q_1, Q_2, Q_3, Q_4\}\) from potential business partners, which must satisfy the assessment requirements of the three criteria: (1) \(x_1\) is the cost of product; (2) \(x_2\) is the quality of product; (3) \(x_3\) is the quality of service. The weight vector of the three criteria is specified as \(\lambda = (0.33, 0.35, 0.32)\) to indicate the importance of the three criteria. Then, the experts/decision makers are invited to assess the four suppliers/alternatives over the three criteria by the NZNs that are composed of their truth, falsity,
indeterminacy fuzzy values and the measures of corresponding reliabilities. Thus, all NZNs can be constructed as the following NZN decision matrix:

\[
S_Z = (s_{Zj})_{4 \times 3} = \\
\begin{bmatrix}
(0.6, 0.8), (0.1, 0.7), (0.2, 0.8) \\
(0.7, 0.6), (0.1, 0.8), (0.2, 0.7) \\
(0.8, 0.7), (0.1, 0.8), (0.2, 0.6) \\
(0.6, 0.6), (0.2, 0.6), (0.1, 0.7) \\
\end{bmatrix}
\]

On the one hand, the developed MDM approach using the NZNWAA operator can be used for this MDM problem and depicted by the following decision process:

Step 1 Using Eq. (8), the overall collected NZNs \(s_{Zj}(j = 1, 2, 3, 4)\) are given as follows:

\[
s_{Z1} = < (0.6701, 0.7098), (0.1248, 0.7655), (0.2000, 0.7928) > ,
\]

\[
s_{Z2} = < (0.7451, 0.7365), (0.1558, 0.7315), (0.2000, 0.6943) > ,
\]

\[
s_{Z3} = < (0.6701, 0.7138), (0.2277, 0.6943), (0.2606, 0.7335) > ,
\]

\[
s_{Z4} = < (0.6682, 0.7123), (0.1248, 0.7000), (0.1421, 0.7977) > .
\]

Step 2 By Eq. (1), the score values of \(Y(s_{Zj})\) for the alternative \(Q_j (j = 1, 2, 3, 4)\) are yielded below:

\[
Y(s_{Z1}) = 0.7405 , \quad Y(s_{Z2}) = 0.7653 , \quad Y(s_{Z3}) = 0.7097 , \quad \text{and} \quad Y(s_{Z4}) = 0.7584 .
\]

Step 3 According to the score values \(Y(s_{Z2}) > Y(s_{Z4}) > Y(s_{Z1}) > Y(s_{Z3})\), the four alternatives are ranked as \(Q_2 > Q_4 > Q_1 > Q_3\). Hence, the best supplier is \(Q_2\).

On the other hand, the developed MDM approach using the NZNWGA operator can be also used for this MDM problem and depicted by the following decision process:

Step 1’ By Eq. (9), the overall collected NZNs \(s_{Zj}(j = 1, 2, 3, 4)\) are obtained as follows:

\[
s_{Z1} = < (0.6653, 0.6931), (0.1333, 0.7714), (0.2000, 0.8154) > ,
\]

\[
s_{Z2} = < (0.7234, 0.7306), (0.2095, 0.7376), (0.2000, 0.7103) > ,
\]

\[
s_{Z3} = < (0.665, 0.6971), (0.2335, 0.7103), (0.3642, 0.7397) > ,
\]

\[
s_{Z4} = < (0.6632, 0.6963), (0.1333, 0.7000), (0.1695, 0.8206) > .
\]

Step 2’ By Eq. (1), the score values of \(Y(s_{Zj})\) for the alternative \(Q_j (j = 1, 2, 3, 4)\) are given as follows:

\[
Y(s_{Z1}) = 0.7317 , \quad Y(s_{Z2}) = 0.7440 , \quad Y(s_{Z3}) = 0.6762 , \quad \text{and} \quad Y(s_{Z4}) = 0.7431 .
\]

Step 3’ According to the score values \(Y(s_{Z2}) > Y(s_{Z4}) > Y(s_{Z1}) > Y(s_{Z3})\), the four alternatives are ranked as \(Q_2 > Q_4 > Q_1 > Q_3\). Thus the best supplier is \(Q_2\).

Based on the developed MDM approach using the NZNWAA and NZNWGA operators and the score function, we can see that the above two kinds of ranking orders regarding the four alternatives and the best one are identical. Hence, the developed MDM approach is effective.

**Relative comparison**

For convenient comparison with existing relative method [33], assume we do not consider the assessment measures of corresponding reliabilities in the decision matrix \(S_Z\) as a special case of the above example. Then, the NZN decision matrix is reduced to the following single-valued neutrosophic decision matrix:

\[
\begin{bmatrix}
(0.6, 0.8), (0.1, 0.7), (0.2, 0.8) \\
(0.7, 0.6), (0.1, 0.8), (0.2, 0.7) \\
(0.8, 0.7), (0.1, 0.8), (0.2, 0.6) \\
(0.6, 0.6), (0.2, 0.6), (0.1, 0.7) \\
\end{bmatrix}
\]
Table 1 Decision results based on various weighted aggregation operators

| MDM method                                  | Aggregated value                      | Score value | Ranking                   |
|---------------------------------------------|---------------------------------------|-------------|--------------------------|
| MDM method based on the NZNWAA operator     | $s_{Z1} = \langle (0.6701, 0.7098), (0.1248, 0.7655), (0.2000, 0.7928) \rangle$ | $Y(s_{Z1}) = 0.7405$ | $Q_2 > Q_4 > Q_1 > Q_3$ |
|                                            | $s_{Z2} = \langle (0.7451, 0.7365), (0.1558, 0.7315), (0.2000, 0.6943) \rangle$ | $Y(s_{Z2}) = 0.7653$ |                         |
|                                            | $s_{Z3} = \langle (0.6701, 0.7138), (0.2277, 0.6943), (0.2606, 0.7335) \rangle$ | $Y(s_{Z3}) = 0.7097$ |                         |
|                                            | $s_{Z4} = \langle (0.6682, 0.7123), (0.1248, 0.7000), (0.1421, 0.7977) \rangle$ | $Y(s_{Z4}) = 0.7584$ |                         |
| MDM method based on the NZNWGA operator     | $s_{Z1} = \langle (0.6653, 0.6931), (0.1333, 0.7714), (0.2000, 0.8154) \rangle$ | $Y(s_{Z1}) = 0.7317$ | $Q_2 > Q_4 > Q_1 > Q_3$ |
|                                            | $s_{Z2} = \langle (0.7234, 0.7306), (0.2095, 0.7376), (0.2000, 0.7103) \rangle$ | $Y(s_{Z2}) = 0.7440$ |                         |
|                                            | $s_{Z3} = \langle (0.665, 0.6971), (0.2335, 0.7103), (0.3642, 0.7397) \rangle$ | $Y(s_{Z3}) = 0.6762$ |                         |
|                                            | $s_{Z4} = \langle (0.6632, 0.6963), (0.1333, 0.7000), (0.1695, 0.8206) \rangle$ | $Y(s_{Z4}) = 0.7431$ |                         |
| MDM method based on the SVNWAA operator [33] | $s_1 = \langle 0.6701, 0.1248, 0.2000 \rangle$ | $Y(s_1) = 0.7818$ | $Q_4 > Q_2 > Q_1 > Q_3$ |
|                                            | $s_2 = \langle 0.7451, 0.1558, 0.2000 \rangle$ | $Y(s_2) = 0.7964$ |                         |
|                                            | $s_3 = \langle 0.6701, 0.2277, 0.2606 \rangle$ | $Y(s_3) = 0.7273$ |                         |
|                                            | $s_4 = \langle 0.6682, 0.1248, 0.1421 \rangle$ | $Y(s_4) = 0.8004$ |                         |
| MDM method based on the SVNWGA operator [33] | $s_1 = \langle 0.6653, 0.1333, 0.2000 \rangle$ | $Y(s_1) = 0.7773$ | $Q_4 > Q_1 > Q_2 > Q_3$ |
|                                            | $s_2 = \langle 0.7234, 0.2095, 0.2000 \rangle$ | $Y(s_2) = 0.7713$ |                         |
|                                            | $s_3 = \langle 0.665, 0.2335, 0.3642 \rangle$ | $Y(s_3) = 0.6892$ |                         |
|                                            | $s_4 = \langle 0.6632, 0.1333, 0.1695 \rangle$ | $Y(s_4) = 0.7868$ |                         |

Then, Eqs. (1), (8) and (9) are also reduced to existing score function, single-valued neutrosophic weighted arithmetic averaging (SVNWAA) and the single-valued neutrosophic weighted geometric averaging (SVNWGA) operators [33]:

$$Y(s_j) = \frac{2 + T_{Vj} - I_{Vj} - F_{Vj}}{3} \quad \text{for} \quad Y(s_j) \in [0, 1], \quad (10)$$

$$s_j = \text{SVNWAA}(s_{j1}, s_{j2}, \ldots, s_{jn}) = \sum_{i=1}^{n} \lambda_i s_{ji}$$

$$= \left(1 - \prod_{i=1}^{n} (1 - T_{Vji})^{\lambda_i} \right) \prod_{i=1}^{n} T_{Vji}^{\lambda_i} \prod_{i=1}^{n} F_{Vji}^{\lambda_i} \right), \quad (11)$$

$$s_j = \text{SVNWGA}(s_{j1}, s_{j2}, \ldots, s_{jn}) = \prod_{i=1}^{n} \lambda_i s_{ji}$$

$$= \left(\prod_{i=1}^{n} T_{Vji}^{\lambda_i} \right) - \left(1 - \prod_{i=1}^{n} (1 - I_{Vji})^{\lambda_i} \right) \prod_{i=1}^{n} (1 - F_{Vji})^{\lambda_i} \right), \quad (12)$$

Based on existing MDM method [33], decision results are given using Eqs. (10)–(12). Thus, all decision results obtained by existing MDM method [33] and the proposed MDM approach are shown in Table 1.

By comparing the developed MDM approach with existing neutrosophic MDM method [33] in Table 1, we see that there is the ranking difference between them. The developed MDM approach based on the NZNWAA or NZNWGA operator indicates that the ranking is $Q_2 > Q_4 > Q_1 > Q_3$ and the best alternative is $Q_2$, while the MDM method based on the NZNWAA or SVNWGA operator [33] indicates that the ranking is $Q_4 > Q_2 > Q_1 > Q_3$ or $Q_4 > Q_1 > Q_2 > Q_3$ and the best alternative is $Q_4$.

**Discussion**

From the decision results of Table 1, it is obvious that the MDM methods with different decision information can affect the ranking orders. The reason resulting in the different ranking is that the developed MDM approach uses the hybrid assessment information of both neutrosophic values and neutrosophic measures of corresponding reliabilities, while existing neutrosophic MDM method [33] only uses the assessment information of single-valued neutrosophic values without considering the related reliability measures.
Clearly, the introduced reliability measures not only can enhance the information expression and credibility of the evaluation results but also can impact on the ranking order of alternatives, which show the effectiveness and rationality of the developed MDM approach. Since NZNs indicate more ability to depict the human knowledge and judgments by neutrosophic values and the reliability measures related to the neutrosophic values, NZNs richen the measure information of reliability related to the neutrosophic values in indeterminate and inconsistent setting. Hence, the information expression of NZN is superior to that of the single neutrosophic value or the single Z-number in MDM problems. Then, the developed MDM approach in this paper indicates more generalized form to extend existing neutrosophic MDM theory and methods because the existing neutrosophic MDM method is only the special case of this study and cannot carry out the MDM problem with NZN information. Therefore, the developed MDM approach can overcome the flaw of existing neutrosophic MDM theory and methods and strengthen the MDM reliability and effectiveness, which show the highlighting advantages under the environment of NZNs.

Conclusion

In the original study, the proposed NZN set can solve the hybrid information expression problem of both neutrosophic values and corresponding reliability measures and overcome the flaw of missing reliability measures in existing neutrosophic set. Then, the proposed basic operations, score function, and NZNWAA and NZNWGA operators of NZNs provided effective and reasonable mathematical tools for the information aggregation of NZNs and MDM modeling in the setting of NZNs. Further, the developed MDM approach using the NZNWAA and NZNWGA operators and the score function can solve MDM problems with NZN information as the extension of existing neutrosophic MDM methods. The decision results of the illustrative example about the selection problem of business partners indicated the applicability and effectiveness of the developed MDM approach in NZN setting.

However, the developed MDM approach extends existing neutrosophic MDM theory and methods and provides a new way for solving MDM problems with NZNs. Since as the first time study, the ranking method based on the score function and the NZNWAA and NZNWGA operators are the most basic algorithms for MDM problems with NZNs, we should continue to propose the new aggregation operators (e.g., Bonferroni mean, Heronian mean, Dombi aggregation operators) and ranking methods of NZNs for improving neutrosophic MDM methods and to extend them to the applications of group decision making, medical diagnosis, pattern recognition, optimization programming etc.

Compliance with ethical standards

Conflict of interest The authors declare that we have no conflict of interest regarding the publication of this paper.

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