Research Article

The Unbalanced Linguistic Aggregation Operator in Group Decision Making

Li Zou,1 Zheng Pei,2 Hamid Reza Karimi,3 and Peng Shi4,5

1 School of Computer and Information Technology, Liaoning Normal University, Dalian 116029, China
2 School of Mathematic and Computer Engineering, Xihua University, Sichuan, Chengdu 610039, China
3 Department of Engineering Faculty of Engineering and Science, University of Agder, 4898 Grimstad, Norway
4 Department of Computing and Mathematical Sciences, University of Glamorgan, Pontypridd CF37 1DL, UK
5 School of Engineering and Science, Victoria University, Melbourne, VIC 8001, Australia

Correspondence should be addressed to Hamid Reza Karimi, hamid.r.karimi@uia.no

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1. Introduction

The need of uncertain linguistic information processing has become an important topic in many areas dealing with vague information such as in universal design used for developing products and environments that are usable by all people to the greatest extent possible; although a set of acknowledged principles has been developed and commonly used by industry and academia, it is difficult to quantitatively evaluate whether a product is indeed a good example of universal design, and linguistic evaluation approach for universal design can improve the shortcomings of traditional analytical hierarchy process methods, quantitative judgments, priority analysis, and aggregation performance [1]. Formally, the fuzzy logic framework and especially fuzzy sets themselves are not always easy to obtain from
the linguistic value sets. That is why we choose to keep the words themselves without going through a fuzzy modeling [2]. In fact, linguistic value is the granularity of information that allows for a better approximation of the concepts when it is needed [3].

In linguistic group decision-making analysis, the problems are associated with the following [4]: (1) the choice of the linguistic value set with its semantic; (2) the choice of the aggregation operator of linguistic information; (3) the choice of the best alternatives. In the above-mentioned three steps, the aim of (1) consists of establishing the linguistic variable [5] or linguistic expression domain with a view to provide the linguistic performance values. In practice, fuzzy numbers or an ordered structure of linguistic values can be used to explain their semantic [5, 6]. The aim of (2) is to carry out the aggregation of linguistic information; there are many numeric aggregation operators [7–9] and linguistic aggregation operators [4, 10–13] to aggregate them. The aim of (3) consists of obtaining a collective performance value over each alternative and finding a solution set of alternatives. The solution set of alternatives is the best alternative that is the most satisfied alternative for the experts.

In most cases, information of group decision-making problems can be assessed in a qualitative form rather than a quantitative one, and experts use natural language instead of numerical values to express their evaluations of decision-making problems [4]; hence, we need the linguistic approach to solve group decision-making problems with linguistic assessment [3, 6, 10, 14]. Up to now, many linguistic approaches have been proposed and applied to solve problems with linguistic assessment, for example, personnel management [4, 12], web information processing [15, 16], and sensor evaluation and fuzzy risk analysis [17–24]. In fact, linguistic approaches are the core of computing with words proposed by Zadeh [25]. In [26], Herrera et al. make a review of the developments of Computing with Words in decision-making and explore different linguistic computational models that have been applied to the decision making field. In this paper, unbalanced linguistic values proposed in [27] are used to deal with group decision making, in practice, experts need to assess a number of values in a side of reference domain higher than in the other one; that is, experts use unbalanced linguistic values to express their evaluation for problems, to deal with unbalanced linguistic values in group decision making, and we adopt 2-tuple representation model of linguistic values [10] and linguistic hierarchies [4] to express unbalanced linguistic values; moreover, inspired by the ordered weighted geometric operator [28–30], we propose the unbalanced linguistic ordered weighted geometric operator to aggregate unbalanced linguistic evaluation values, and some interesting properties of the unbalanced linguistic ordered weighted geometric operator are also discussed.

The organization of this paper is as follows. In Section 2, we briefly make a review of the 2-tuple fuzzy linguistic representation model and unbalanced linguistic terms. In Section 3, we present the unbalanced linguistic ordered weighted geometric operator and some properties of the operator. An example is given in Section 4 to deal with a group decision-making problem with unbalanced linguistic evaluation values, and it is shown that our method is an alternative aggregation operator for linguistic evaluation approach. We conclude in Section 5.

2. Preliminaries

In this section, we briefly review the 2-tuple linguistic representation model and unbalanced linguistic values we refer to [10, 27] for more details.
Mathematical Problems in Engineering

The 2-tuple linguistic representation model was introduced by Herrera and Martínez [10]. Let $S = \{s_0, \ldots, s_g\}$ be the initial finite linguistic value set. Formally, the 2-tuple linguistic representation model is formed by $(s_i, \alpha)$, in which $s_i \in S(i \in \{0, 1, \ldots, g\})$ and $\alpha \in [-0.5, 0.5]$; that is, linguistic information is encoded in the space $S \times [-0.5, 0.5]$. Based on the representation $(s_i, \alpha)$, we can easily obtain the following symbolic translation of linguistic values from $\beta \in [0, g]$ to $S \times [-0.5, 0.5]$, that is, $\Delta : [0, g] \rightarrow S \times [-0.5, 0.5]$, $\beta \mapsto (s_i, \alpha)$, in which $i = \text{round}(\beta)$ (round(·) is the usual round operation), $\alpha = \beta - i \in [-0.5, 0.5]$. Intuitively, $\Delta(\beta) = (s_i, \alpha)$ expresses that $s_i$ is the closest linguistic value to $\beta$, and $\alpha$ is the value of the symbolic translation. Additionally, there is a $\Delta^{-1}$ function such that from a 2-tuple it returns its equivalent numerical value $\beta \in [0, g]$, that is, $\Delta^{-1} : S \times [-0.5, 0.5] \rightarrow [0, g]$, $\Delta^{-1}(s_i, \alpha) = i + \alpha = \beta$. In fact, the model defines a set of transformation functions between linguistic values and 2-tuple linguistic representations as well as numeric values and 2-tuple linguistic representations. This allows us to easily process linguistic information by numeric value; for example, we have the following linguistic aggregation operators: Let a set of the 2-tuple linguistic representations be $x = \{(s_1, \alpha_1), \ldots, (s_n, \alpha_n)\}$, and let $W = \{w_1, \ldots, w_n\}$ be an associated weight such that $w_i \in [0, 1]$ and $\sum_{i=1}^{n} w_i = 1$:

1. the 2-tuple arithmetic mean operator $\overline{x}^2: \overline{x}^2 = \Delta(\sum_{i=1}^{n} 1/n \times \Delta^{-1}(s_i, \alpha_i))$;

2. the 2-tuple weighted aggregation operator $F^w: F^w = \Delta(\sum_{i=1}^{n} w_i \times \Delta^{-1}(s_i, \alpha_i))$;

3. the 2-tuple ordered weighted aggregation operator $F^c: F^c((s_1, \alpha_1), \ldots, (s_n, \alpha_n)) = \Delta(\sum_{j=1}^{n} w_j \times \beta_j^r)$, in which $\beta_j^r$ is the $j$th largest linguistic value of $\{\beta_i = \Delta^{-1}(s_i, \alpha_i) | i = 1, \ldots, n\}$.

Unbalanced linguistic values proposed in [27] are used to deal with scales for assessing preferences where the experts need to assess a number of values in a side of reference domain higher than in the other one. Generally, an unbalanced linguistic term set $S$ has a minimum label, a maximum label, and a central label, and the remaining labels are nonuniformly and nonsymmetrically distributed around the central one on both left and right lateral sets; that is, we can represent $S$ in the following form: $S = S_l \cup S_c \cup S_r$, in which $S_l$ contains all left lateral labels but the central label, $S_c$ just contains the central label, and $S_r$ contains all right lateral labels higher than the central label.

Example 2.1. $S = \{\text{none (N)}, \text{low (L)}, \text{medium (M)}, \text{almost high (AH)}, \text{high (H)}, \text{quite high (QH)}, \text{very high (VH)}, \text{almost total (AT)}, \text{total (T)}\}$ is an unbalanced linguistic term set, in which $S_l = \{N, L\}$, $S_c = \{M\}$, and $S_r = \{AH, H, QH, VH, AT, T\}$.

To obtain 2-tuple fuzzy linguistic representations of unbalanced linguistic values without loss of information, we need the concept of linguistic hierarchies [4], that is, $LH = \bigcup_t l(t, n(t))$ it takes into account a set of levels where each level is a linguistic term set with different granularity from the remaining levels of the hierarchy, where $l(t, n(t))$ is a linguistic hierarchy with $t$ being a number that indicates the level of the hierarchy and $n(t)$ the granularity of the linguistic term set of $t$. Generally, the linguistic term set $S^{n(t+1)}$ of the level $t + 1$ is obtained from its predecessor $S^{n(t)}$ as $l(t, n(t)) \rightarrow l(t + 1, 2 \times n(t) + 1)$. In linguistic hierarchies $LH$, transformation function between labels from different levels to represent
2-tuple fuzzy linguistic representations is as follows [4]: for any linguistic levels \( t \) and \( t' \), \( TF^t_{t'}: l(t, n(t)) \rightarrow l(t', n(t')) \) such that

\[
TF^t_{t'}(s_i^{n(t)}, a^{n(t)}) = \Delta_t \left( \frac{\Delta_1^{-1}(s_i^{n(t)}, a^{n(t)}) \times (n(t') - 1)}{n(t) - 1} \right).
\]  

(2.1)

By using linguistic hierarchies \( LH = \bigcup_i l(t, n(t)) \), we can obtain the 2-tuple fuzzy linguistic representation of each term of unbalanced linguistic term set in \( LH \); here, we use the following example to explain the converting process, and we refer to [27] for more details.

**Example 2.2.** Let linguistic hierarchies be

\[
LH = l(1, 3) \cup l(2, 5) \cup l(3, 9) \cup l(4, 17)
\]

\[
= \{s_0^3, s_1^3, s_2^3\} \cup \{s_5^9, s_5^9, \ldots, s_4^9\} \cup \{s_0^9, s_1^9, \ldots, s_8^9\}
\]

\[
\cup \{s_0^{17}, s_1^{17}, \ldots, s_{16}^{17}\}.
\]

\((s_5^9, 0.3)\) is a 2-tuple fuzzy linguistic representation of level 3, and it is 2-tuple fuzzy linguistic representation in level 2 is

\[
TF^3_{2}(s_5^9, 0.3) = \Delta_2 \left( \frac{\Delta_3^{-1}(s_5^9, 0.3) \times (5 - 1)}{9 - 1} \right)
\]

\[
= \Delta_2 \left( \frac{5.3 \times 4}{8} \right) = \Delta_2(2.65)
\]

\[
= (s_{5}^{3}, -0.35).
\]

For unbalanced linguistic term set \( S = \{N, L, M, AH, H, QH, VH, AT, T\} \), (1) due to \( n(2) = 5 \) and \( (n(2) - 1)/2 = |S_l| = |\{N, L\}| = 2 \), the representation of \( S_l \) is obtained from level 2 of \( LH \) as follows: \( \{L \leftarrow s_5^5, N \leftarrow s_5^5\} \); (2) due to \( n(3) = 9, n(4) = 17, \) and \( (n(3) - 1)/2 = 4 < |S_r| = |\{AH, H, QH, VH, AT, T\}| = 6 < (n(4) - 1)/2 = 8 \), we use level 3 and level 4 to represent \( S_r = \{AH, H, QH, VH, AT, T\} \), according to lab3 and lab4 defined in [27], and \( \{AH, H\} \) and \( \{QH, VH, AT, T\} \) are represented in level 3 and level 4, respectively, \( \{AH \leftarrow s_5^9, H \leftarrow s_6^9\}, \{QH \leftarrow s_1^{13}, VH \leftarrow s_1^{17}, AT \leftarrow s_1^{15}, T \leftarrow s_1^{16}\} \); (3) according to density and bridging representation gaps defined in [27], the upside and the downside of the central label \( M \) are represented in level 2 and 3 of \( LH \) by means of \( s_2^5 \) and \( s_4^9 \) respectively; the upside and the downside of the label \( H \) are represented in level 3 and 4 of \( LH \) by means of \( s_6^9 \) and \( s_1^{12} \) respectively; (4) the final 2-tuple fuzzy linguistic representations of \( S \) in \( LH \) are

\[
S_l : \{N \leftarrow s_0^3, L \leftarrow s_1^3\}, S_c : \{M \leftarrow s_2^5 \cup s_1^3\}, S_r : \{AH \leftarrow s_5^9, H \leftarrow s_6^9 \cup s_1^{12}, QH \leftarrow s_1^{17}, VH \leftarrow s_1^{14}, AT \leftarrow s_1^{15}, T \leftarrow s_1^{16}\}.
\]

Formally, for any 2-tuple fuzzy linguistic representation \( (s_i, a_i) \) \((s_i \in S, a_i \in [-0.5, 0.5])\), it can be converted by the following unbalanced linguistic transformation functions in \( LH \)
and vice versa, that is, \( L \mathcal{E} : S \times [-0.5, 0.5) \rightarrow LH \times [-0.5, 0.5), (s_i, a_i) \mapsto (s_{L(i)}^{G(i)}, a_i) \) such that \( s_{L(i)}^{G(i)} \in LH \). \( L \mathcal{E}^{-1} : LH \times [-0.5, 0.5) \rightarrow S \times [-0.5, 0.5), (s_k^{n(i)}, a_k) \mapsto (s_i, \lambda) \), for example, \( L \mathcal{E}(H, 0.3) = (s_k^0, 0.3), L \mathcal{E}^{-1}(s_{13}^{2}, -0.2) = (QH, -0.2) \).

3. The Unbalanced Linguistic Ordered Weighted Geometric Operator

To aggregate unbalanced linguistic values, we present the unbalanced linguistic ordered weighted geometric operator \[20\], which is inspired by 2-tuple fuzzy linguistic representation model \[27\] and the ordered weighted geometric operator \[29\].

**Definition 3.1.** An ordered weighted geometric operator of dimension \( n \) is a mapping \( g : (R^+)^n \rightarrow R^+ \) that has a weighting vector \( W = (w_1, w_2, \ldots, w_n) \) associated with it, with \( w_i \in [0, 1] \) and \( \sum_{i=1}^{n} w_i = 1 \), such that \( g(a_1, a_2, \ldots, a_n) = \prod_{j=1}^{n} b_j^{w_j} \), where \( b_j \) is the \( j \)th largest linguistic value of the \( a_i \) (\( i = 1, \ldots, n \)).

**Definition 3.2.** Assume that the set of unbalanced linguistic values \( V = \{s_i \mid i = 1, 2, \ldots, n\} \) will be aggregated, in which \( s_i \in S = \{s_0, s_1, \ldots, s_m\} \) is an unbalanced linguistic value, a weighting vector is \( W = (w_1, w_2, \ldots, w_n) \) with \( w_i \in [0, 1] \), and \( \sum_{i=1}^{n} w_i = 1 \). Then the unbalanced linguistic ordered weighted geometric operator (the ULOWG operator) is defined as follows:

\[
\begin{align*}
\text{f}_{ULOWG}(V) &= \text{f}_{ULOWG}(\{s_i \mid i = 1, 2, \ldots, n\}) \\
&= \text{f}_{ULOWG}\left(\{TF_{L(i)}^{h_i}(L \mathcal{E}(s_i)) \mid i = 1, \ldots, n\}\right) \\
&= L \mathcal{E}^{-1}(s_k^{n(h)}, a_k),
\end{align*}
\]

where \( t_i \) is the level of \( L \mathcal{E}(s_i) \) in LH, \( t_0 \) is a level of LH, fixed by users, \( s_k^{n(h)} \in S^{n(h)} \subset LH \) and \( a_k \in [-0.5, 0.5) \) such that

\[
k + a_k = \prod_{i=1}^{n} \left( \Delta_{t_0}^{-1}\left(\Delta_{t_0}^{h_i}(L \mathcal{E}(s_i))\right) \right)^{w_i},
\]

where for any \( i \in \{1, \ldots, n\}, \Delta_{t_0}^{-1}(\Delta_{t_0}^{h_i}(L \mathcal{E}(s_i))) \geq \Delta_{t_0}^{-1}(\Delta_{t_0}^{h_i}(L \mathcal{E}(s_i))) \); that is, \( TF_{L(i)}^{h_i}(L \mathcal{E}(s_i)) \) is the \( i \)th largest linguistic value in \( \{TF_{L(i)}^{h_i}(L \mathcal{E}(s_i)) \mid i = 1, 2, \ldots, n\} \).

**Remark 3.3.** In Definition 3.2, \( TF_{L(i)}^{h_i}(L \mathcal{E}(s_i)) \) means that unbalanced linguistic values are represented at level \( t_0 \) of LH, and \( t_0 \) is decided by users. In fact, by using \( TF_{L(i)}^{h_i}(\cdot) \) and \( L \mathcal{E}(\cdot) \), \( S \) is converted in \( S^{n(h)} \) of LH. By using \( L \mathcal{E}^{-1}(s_k^{n(h)}, a_k) \), aggregation result \( f_{ULOWG}(V) \) is converted to the 2-tuple fuzzy linguistic representation of unbalanced linguistic value.
In Definition 3.2, a natural question is how to obtain the associated weighting vector $W$. Following Yager’s ideas [31] on quantifier-guided aggregation, we could compute the weighting vector of an ULOWG operator using a linguistic quantifier $Q$ [31] as follows

$$w_i = Q\left(\frac{i}{r}, a, b\right) - Q\left(\frac{i-1}{r}, a, b\right), \quad i = 1, \ldots, r. \tag{3.3}$$

$$Q(x, a, b) = \begin{cases} 0, & \text{if } x < a, \\ \frac{x-a}{b-a}, & \text{if } a \leq x < b, \\ 1, & \text{if } x \geq b, \end{cases} \tag{3.4}$$

where $x, a, b \in [0, 1]$. Some examples of $Q(x, a, b)$ (see Figure 1) are most, at least half, and as many as possible; their parameters $(a, b)$ are $(0.3, 0.8), (0, 0.5),$ and $(0.5, 1)$, respectively.

**Example 3.4.** We select Yager’s linguistic quantifier *most*, that is, $Q(x, a, b) = Q(x, 0.3, 0.8)$, according to (3.3), $r = 3$, and

$$w_1 = Q\left(\frac{1}{3}, 0.3, 0.8\right) - Q\left(\frac{0}{3}, 0.3, 0.8\right) = \frac{1}{15},$$

$$w_2 = Q\left(\frac{2}{3}, 0.3, 0.8\right) - Q\left(\frac{1}{3}, 0.3, 0.8\right) = \frac{2}{3},$$

$$w_3 = Q\left(\frac{3}{3}, 0.3, 0.8\right) - Q\left(\frac{2}{3}, 0.3, 0.8\right) = \frac{4}{15}. \tag{3.5}$$

so weighting vectors $W = (1/15, 2/3, 4/15)$.

Let $S = \{N, L, M, AH, H, QH, VH, AT, T\}$ be a set of unbalanced linguistic values. Suppose that $V = \{AH, QH, H\}$ will be aggregated, and a weighting vector is $W = (1/15, 2/3, 4/15)$. In this example, we select $t_0 = 3$ and density $s_6$ is extreme [27]; hence,

$$TF_3^I(\mathcal{L}(AH)) = TF_3^I(s^9_5) = s^9_5,$$

$$TF_3^I(\mathcal{L}(QH)) = TF_3^I(s^{17}_{13}) = (s^9_6, 0.5), \tag{3.6}$$

$$TF_3^I(\mathcal{L}(H)) = TF_3^I(s^9_6) = s^9_6.$$
Proposition 3.5. Let the set of unbalanced linguistic values be $V = \{s_i \mid i = 1, 2, \ldots, n\}$, and a weighting vector is $W = (w_1, \ldots, w_n)$ with $w_i \in [0, 1]$ and $\sum_{i=1}^{n} w_i = 1$.

(1) If $w_1 = 1$ and $w_2 = \cdots = w_n = 0$, then $f_{UL\text{OWG}}(s_1, \ldots, s_n) = \max \{s_1, \ldots, s_n\}$;

(2) if $w_n = 1$ and $w_1 = \cdots = w_{n-1} = 0$, then $f_{UL\text{OWG}}(s_1, \ldots, s_n) = \min \{s_1, \ldots, s_n\}$;

(3) if $w_1 = w_n = 0$, then $f_{UL\text{OWG}}$ reduces to the linguistic Olympic operator; that is, the smallest and largest linguistic values are deleted from linguistic evaluation values;

(4) if for any $i \in \{1, \ldots, n\}$, $w_i = 1$, then

$$f_{UL\text{OWG}}(\{s_1, \ldots, s_n\}) = s_i.$$  \hspace{1cm} (3.8)

Proof. (1) According to (3.2), we have $k + \alpha_k = \prod_{i=1}^{n} (\Delta_{b_i}^{-1}(TF_{b_i}^{H}(LH(s_i))))^{w_i} = (\Delta_{b_1}^{-1}(TF_{b_1}^{H}(LH(s_1))))^{w_1} \times (\Delta_{b_2}^{-1}(TF_{b_2}^{H}(LH(s_2))))^{w_2} \times \cdots \times (\Delta_{b_n}^{-1}(TF_{b_n}^{H}(LH(s_n))))^{w_n} = (\Delta_{b_1}^{-1}(TF_{b_1}^{H}(LH(s_1))))^{1} \times (\Delta_{b_2}^{-1}(TF_{b_2}^{H}(LH(s_2))))^{0} \times \cdots \times (\Delta_{b_n}^{-1}(TF_{b_n}^{H}(LH(s_n))))^{0} = (\Delta_{b_1}^{-1}(TF_{b_1}^{H}(LH(s_1))))$, where $(\Delta_{b_i}^{-1}(TF_{b_i}^{H}(LH(s_i))))$ is the largest linguistic value of the $\{TF_{b_i}^{H}(LH(s_i)) \mid i = 1, 2, \ldots, n\}$; then, we can get

$$f_{UL\text{OWG}}(s_1, \ldots, s_n) = LH^{-1}(\Delta_{b_1}(k + \alpha_k)) = LH^{-1} \left( \left( \Delta_{b_1}^{-1}(TF_{b_1}^{H}(LH(s_1))) \right) \right)$$

$$= \max \{s_1, \ldots, s_n\}.$$  \hspace{1cm} (3.9)

(2) According to (3.2), we have $k + \alpha_k = \prod_{i=1}^{n} (\Delta_{b_i}^{-1}(TF_{b_i}^{H}(LH(s_i))))^{w_i} = (\Delta_{b_1}^{-1}(TF_{b_1}^{H}(LH(s_1))))^{w_1} \times (\Delta_{b_2}^{-1}(TF_{b_2}^{H}(LH(s_2))))^{w_2} \times \cdots \times (\Delta_{b_n}^{-1}(TF_{b_n}^{H}(LH(s_n))))^{w_n} = (\Delta_{b_1}^{-1}(TF_{b_1}^{H}(LH(s_1))))^{0} \times (\Delta_{b_2}^{-1}(TF_{b_2}^{H}(LH(s_2))))^{0} \times \cdots \times (\Delta_{b_n}^{-1}(TF_{b_n}^{H}(LH(s_n))))^{1} = (\Delta_{b_1}^{-1}(TF_{b_1}^{H}(LH(s_n))))$, where $(\Delta_{b_i}^{-1}(TF_{b_i}^{H}(LH(s_n))))$ is the smallest linguistic value of the $\{TF_{b_i}^{H}(LH(s_i)) \mid i = 1, 2, \ldots, n\}$; then, we can get

$$f_{UL\text{OWG}}(s_1, \ldots, s_n) = LH^{-1}(\Delta_{b_1}(k + \alpha_k)) = \min \{s_1, \ldots, s_n\}.$$  \hspace{1cm} (3.10)

(3) It is obvious.

(4) If for any $i \in \{1, \ldots, n\}$, $w_i = 1$; thus, for any $j \neq i$, $w_j = 0$, then we have $k + \alpha_k = (\Delta_{b_i}(TF_{b_i}^{H}(LH(s_i))))$

$$f_{UL\text{OWG}}(s_1, \ldots, s_n) = LH^{-1}(\Delta_{b_1}(k + \alpha_k)) = LH^{-1}(\Delta_{b_1}(k + \alpha_k)) = s_i.$$  \hspace{1cm} (3.11)
Proposition 3.6. Assume that a set of unbalanced linguistic values \( V = \{s_i \mid i = 1, \ldots, n\} \) will be aggregated and a weighting vector is \( W = (w_1, w_2, \ldots, w_n) \) with \( w_i \in [0, 1] \) and \( \sum_{i=1}^{n} w_i = 1 \). The ULOWG operator satisfies the following:

1. \( \min\{s_i \mid i = 1, \ldots, n\} \leq f_{\text{ULOWG}}(\{s_i \mid i = 1, \ldots, n\}) \leq \max\{s_i \mid i = 1, \ldots, n\} \);
2. \( f_{\text{ULOWG}} \) is idempotent; that is, if \( s_1 = s_2 = \cdots = s_n \), then \( f_{\text{ULOWG}}(\{s_1, \ldots, s_n\}) = s_i \);
3. \( f_{\text{ULOWG}} \) is monotone in relation to the input values \( s_i \); \( f_{\text{ULOWG}} \) is commutative;
4. \( f_{\text{ULOWG}} \) reduces to the linguistic geometric mean if \( w_i = 1/n \) for all \( i = 1, \ldots, n \), that is,

\[
f_{\text{ULOWG}}(\{s_1, \ldots, s_n\}) = (s_k, a_k),
\]

\[
k + a_k = \sqrt[n]{\prod_{i=1}^{n} \Delta_{t_0}^{-1}\left( T_{b_i}^{l_i}(\mathcal{L}(s_i)) \right)^{w_i}},
\]

Proof. (1) For any \( V = \{s_i \mid i = 1, \ldots, n\} \), \( W = (w_1, w_2, \ldots, w_n) \), \( k + a_k = \prod_{i=1}^{n} \Delta_{t_0}^{-1}\left( T_{b_i}^{l_i}(\mathcal{L}(s_i)) \right)^{w_i} \); according to Proposition 3.5 (1) and (2), we know that \( \prod_{i=1}^{n} \Delta_{t_0}^{-1}\left( T_{b_i}^{l_i}(\mathcal{L}(s_i)) \right)^{w_i} \leq \prod_{i=1}^{n} \Delta_{t_0}^{-1}\left( T_{b_i}^{l_i}(\mathcal{L}(s_i)) \right)^{w_i} \leq \prod_{i=1}^{n} \Delta_{t_0}^{-1}\left( T_{b_i}^{l_i}(\mathcal{L}(s_i)) \right)^{w_i} \), that is, \( \min\{s_i \mid i = 1, \ldots, n\} \leq f_{\text{ULOWG}}(\{s_i \mid i = 1, \ldots, n\}) \leq \max\{s_i \mid i = 1, \ldots, n\} \).

(2) Due to \( s_1 = s_2 = \cdots = s_n \), and according to (3.2), we have

\[
k + a_k = \prod_{i=1}^{n} \Delta_{t_0}^{-1}\left( T_{b_i}^{l_i}(\mathcal{L}(s_i)) \right)^{w_i} = \left( \prod_{i=1}^{n} \Delta_{t_0}^{-1}\left( T_{b_i}^{l_i}(\mathcal{L}(s_i)) \right) \right)^{w_1 + w_2 + \cdots + w_n},
\]

then, \( f_{\text{ULOWG}}(s_1, \ldots, s_n) = \mathcal{L}^{-1}(\Delta_{t_0}(k + a_k) = \mathcal{L}^{-1}(\Delta_{t_0}(k + a_k)) = s_i \).

(3) Assume that \( W = (w_1, w_2, \ldots, w_n) \) denotes the weight of \( \{s_1, s_2, \ldots, s_n\} \) and \( W = (w'_1, w'_2, \ldots, w'_n) \) denotes the weight of \( \{s'_1, s'_2, \ldots, s'_n\} \). If for any \( i \in N, s_i \leq s'_i \), then according to (3.2),

\[
k + a_k = \prod_{i=1}^{n} \Delta_{t_0}^{-1}\left( T_{b_i}^{l_i}(\mathcal{L}(s_i)) \right)^{w_i} \leq k' + a'_k = \prod_{i=1}^{n} \Delta_{t_0}^{-1}\left( T_{b_i}^{l_i}(\mathcal{L}(s'_i)) \right)^{w_i},
\]

that is,

\[
f_{\text{ULOWG}}(s_1, \ldots, s_n) = \mathcal{L}^{-1}(\Delta_{t_0}(k + a_k)) \leq f_{\text{ULOWG}}(s'_1, \ldots, s'_n) = \mathcal{L}^{-1}(\Delta_{t_0}(k' + a'_k)).
\]

(4) Since \( \{s'_1, s'_2, \ldots, s'_n\} \) is the permutation of the \( \{s_1, s_2, \ldots, s_n\} \), so we can get the same sort result \( \{b_1, b_2, \ldots, b_n\} \), that is to say, \( b_i \) is the \( j \)th largest element in the set \( \{s'_1, s'_2, \ldots, s'_n\} \) and \( \{s_1, s_2, \ldots, s_n\} \). According to Definition 3.2,

\[
k + a_k = \prod_{i=1}^{n} \Delta_{t_0}^{-1}\left( T_{b_i}^{l_i}(\mathcal{L}(s_i)) \right)^{w_i} = k' + a'_k = \prod_{i=1}^{n} \Delta_{t_0}^{-1}\left( T_{b_i}^{l_i}(\mathcal{L}(s'_i)) \right)^{w_i},
\]
Table 1: The four alternatives \((x_i)\) and consultancy departments \((d_i)\).

<table>
<thead>
<tr>
<th>Alternatives</th>
<th>(x_1)</th>
<th>(x_2)</th>
<th>(x_3)</th>
<th>(x_4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Departments</td>
<td>(d_1)</td>
<td>(d_2)</td>
<td>(d_3)</td>
<td>(d_4)</td>
</tr>
<tr>
<td>Cost analysis</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>System analysis</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Risk analysis</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Technology analysis</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Table 2: Evaluations provided by four experts.

<table>
<thead>
<tr>
<th>Alternatives</th>
<th>(e_1)</th>
<th>(e_2)</th>
<th>(e_3)</th>
<th>(e_4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>(x_1)</td>
<td>(L)</td>
<td>(AH)</td>
<td>(H)</td>
<td>(H)</td>
</tr>
<tr>
<td>(x_2)</td>
<td>(AH)</td>
<td>(H)</td>
<td>(QH)</td>
<td>(L)</td>
</tr>
<tr>
<td>(x_3)</td>
<td>(AH)</td>
<td>(L)</td>
<td>(AH)</td>
<td>(M)</td>
</tr>
<tr>
<td>(x_4)</td>
<td>(M)</td>
<td>(H)</td>
<td>(AH)</td>
<td>(M)</td>
</tr>
</tbody>
</table>

that is,

\[
f_{ULOWG}(s_1, \ldots, s_n) = f_{ULOWG}(s'_1, \ldots, s'_n). \tag{3.17}
\]

(5) It is obviously set up.

4. Comparison between ULOWA Operator and \(F^\pi\) Operator, and \(F^e\) Operator

In this section, we introduced an application in group decision-making. In the following example, we used the ULOWA operator, ULOWA operator and \(F^e\) operator to choose the better alternative, respectively. Then, we analyze the difference of the ULOWA operator, \(F^\pi\) operator, and \(F^e\) operator based on evaluation results.

Example 4.1. A distribution company needs to upgrade its computing system, so it hires a consulting company to survey the different possibilities existing on the market, to decide which is the best option for its needs. The options (alternatives) and consultancy departments are shown in Table 1. In each of the departments, there is one expert providing evaluation for each alternative (shown in Table 2); these evaluations are assessed in the initial finite unbalanced linguistic value set \(S = \{\text{none (N), low (L), medium (M), almost high (AH), high (H), quite high (QH), very high (VH), almost total (AT), total (T)}\}.

In this example, a linguistic hierarchy is \(LH = I(1,3) \cup I(2,5) \cup I(3,9) \cup I(4,17) = \{s_3^3, s_3^5, s_3^7\} \cup \{s_5^3, s_5^5, \ldots, s_5^9\} \cup \{s_9^3, s_9^5, \ldots, s_9^9\} \cup \{s_{17}^3, s_{17}^5, \ldots, s_{17}^9\}\) of Example 2.2. We select \(t_0 = 3\), and density \(s_8\) is extreme, according to Table 2.

\[
TF_{3}^i(LH(L)) = TF_{3}^3(s_1^5) = (s_2^9, 0),
\]

\[
TF_{3}^i(LH(M)) = TF_{3}^3(s_4^9) = (s_4^9, 0),
\]

\[
TF_{3}^i(LH(AH)) = TF_{3}^3(s_5^9) = (s_5^9, 0),
\]
Mathematical Problems in Engineering

\[
TF^4_3(\mathcal{H}(H)) = TF^4_3(s^9_6) = (s^9_6, 0),
\]

\[
TF^4_3(\mathcal{H}(QH)) = TF^4_3(s_{13}^7) = (s^9_6, 0.5).
\]

(4.1)

According to (3.3), we can get the weighting vector \( W = (0, 0.4, 0.5, 0.1) \) (1) Evaluations based on \( F^W \).

If the weighting vector \( W = (0, 0.4, 0.5, 0.1) \) is for four experts and associated with their evaluations, then \( F^W \) can be used to evaluate alternatives; for example, for alternative \( x_1 \), we have

\[
F^W(L, AH, H, H) = \Delta \bigg( 0 \times \Delta^{-1} \left( s^9_6, 0 \right) + 0.4 \times \Delta^{-1} \left( s^9_6, 0 \right) + 0.5 \times \Delta^{-1} \left( s^9_6, 0 \right) \\
+ 0.1 \times \Delta^{-1} \left( s^9_6, 0 \right) \bigg) \\
= \Delta(0 \times 2 + 0.4 \times 5 + 0.5 \times 6 + 0.1 \times 6) \\
= \Delta(5.6) = \left( s^9_6, -0.4 \right) = (AH, -0.4).
\]

(2) Evaluations based on \( F^c \).

When the weighting vector \( W = (0, 0.4, 0.5, 0.1) \) is for four experts and associated with particular ordered positions rather than their evaluations, then \( F^c \) can be used to evaluate alternatives; for example, for alternative \( x_1 \), we have

\[
F^c(L, AH, H, H) = \Delta \bigg( 0 \times \Delta^{-1} \left( s^9_6, 0 \right) + 0.4 \times \Delta^{-1} \left( s^9_6, 0 \right) + 0.5 \times \Delta^{-1} \left( s^9_6, 0 \right) \\
+ 0.1 \times \Delta^{-1} \left( s^9_6, 0 \right) \bigg) \\
= \Delta(0 \times 6 + 0.4 \times 6 + 0.5 \times 5 + 0.1 \times 2) \\
= \Delta(5.1) = \left( s^9_6, 0.1 \right) = (AH, 0.1).
\]

(3) Evaluations based on \( f_{ULOWG} \).

When the weighting vector \( W = (0, 0.4, 0.5, 0.1) \) is for four experts and associated with particular ordered positions rather than their evaluations, \( f_{ULOWG} \) can also be used to evaluate alternatives; for example, for alternative \( x_1 \), we have

\[
f_{ULOWG}(L, AH, H, H) = \Delta \left( \left( \Delta^{-1} \left( s^9_6, 0 \right) \right)^0 \times \left( \Delta^{-1} \left( s^9_6, 0 \right) \right)^{0.4} \times \left( \Delta^{-1} \left( s^9_6, 0 \right) \right)^{0.5} \\
\times \left( \Delta^{-1} \left( s^9_6, 0 \right) \right)^{0.1} \right) \\
= \Delta \left( 6^0 \times 6^{0.4} \times 5^{0.5} \times 2^{0.1} \right) \\
\approx \Delta(5.3555) = \left( s^9_6, 0.3555 \right) = (AH, 0.3555).
\]
Table 3: Evaluations of alternatives based on $F^w$, $F^c$, and $f_{ULOWG}$.

<table>
<thead>
<tr>
<th>Alternatives</th>
<th>$F^w$</th>
<th>Operators</th>
<th>$F^c$</th>
<th>$f_{ULOWG}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$x_1$</td>
<td>$(H, -0.4)$</td>
<td>$(AH, 0.1)$</td>
<td>$(AH, 0.3555)$</td>
<td></td>
</tr>
<tr>
<td>$x_2$</td>
<td>$(AH, -0.4)$</td>
<td>$(H, 0)$</td>
<td>$(AH, 0.3282)$</td>
<td></td>
</tr>
<tr>
<td>$x_3$</td>
<td>$(M, -0.2)$</td>
<td>$(M, -0.2)$</td>
<td>$(H, -0.4925)$</td>
<td></td>
</tr>
<tr>
<td>$x_4$</td>
<td>$(AH, 0.1)$</td>
<td>$(H, -0.4)$</td>
<td>$(H, -0.4982)$</td>
<td></td>
</tr>
</tbody>
</table>

The best alternative $x_1$

Evaluations of all alternatives based on $F^w$, $F^c$, and $f_{ULOWG}$ are shown in Table 3, we notice that using operator $F^c$ and $f_{ULOWG}$ the same results are obtained, that is, the best alternative is $x_1$. However, evaluations of $f_{ULOWG}$ are smaller than evaluations of $F^c$ correspondingly. Similarly to $F^c$ operator, the ULOWG operator has many interesting properties. So, the ULOWA operator is a alternative aggregation operator to hand multiple attribute group decision making.

5. Conclusion

From the practical point of view, group decision making is associated with multiple information sources fusion. In this paper, we propose the ULOWG operator to solve linguistic group decision-making problems, and the ULOWG operator has many interesting properties, hence, the ULOWG operator is an alternative linguistic aggregation operator in linguistic decision-making problems.

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