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OWA OPERATORS FOR DISCRETE GRADUAL INTERVALS: IMPLICATIONS TO FUZZY INTERVALS AND MULTI-EXPERT DECISION MAKING

ZDENKO TAKÁČ

A new concept in fuzzy sets theory, namely that of gradual element, was introduced recently. It is known that the set of gradual real numbers is not ordered linearly. We restrict our attention to a discrete case and propose a class of linear orders for discrete gradual real numbers. Then, using idea of the so-called admissible order of intervals, we present a class of linear orders for discrete gradual intervals. Once we have the linear orders it is possible to define OWA operator for discrete gradual real numbers and OWA operator for discrete gradual intervals. Recall that gradual intervals also encompass fuzzy intervals, hence our results are applicable to the setting of fuzzy intervals. Our approach is illustrated on a multi-expert decision making problem.

Keywords: OWA operator, ordered weighted averaging operator, gradual number, gradual interval, fuzzy interval, linear order, total order, multi-expert decision making, type-2 fuzzy set

Classification: 03E72, 68T37

1. INTRODUCTION

In [30] Yager proposed an ordered weighted averaging operator (OWA for short) as follows: let $\mathbf{w} = (w_1, \dots, w_n) \in [0, 1]^n$ with $w_1 + \dots + w_n = 1$ be a weighting vector, an OWA operator associated with \mathbf{w} is a mapping $OWA_{\mathbf{w}} : [0, 1]^n \rightarrow [0, 1]$ defined by

$$OWA_{\mathbf{w}}(x_1, \dots, x_n) = \sum_{i=1}^n w_i x_{(i)}$$

where $x_{(i)}$ denotes the i th largest number among x_1, \dots, x_n . This is one of the most widely used aggregation methods. It is of special significance in solving decision making problems.

Most of the existing OWA operators focus on aggregation of real numbers, however, there is a growing interest of scholars to study OWA operators on some more difficult structures. An OWA operator for fuzzy sets (type-1 OWA operator) is proposed in [31]

and more efficient α -level approach to implementing the type-1 OWA operator is given in [32]. Another point of view to OWA operators for fuzzy sets is proposed in [15, 23]. In [4] an OWA operator for intervals (IVOWA) is proposed, and there is proved that it is a special case of the type-1 OWA operator. Another approach to OWA operator for intervals is given in [29].

The distinction between fuzziness and imprecision is discussed in [9] and [10] which leads to proposing a new concept in fuzzy sets theory, namely a gradual element. A gradual real number possess fuzziness but not uncertainty, instead of what is debatable called 'fuzzy number' (we will refer to as fuzzy interval, see [20, 28]) which possess both. A gradual interval is simply a crisp interval of gradual real numbers. This is also a new way of looking at fuzzy intervals [16]: instead of considering them as fuzzy sets, one can see them as crisp intervals of gradual real numbers. Note that the similar point of view on gradualness is presented in [14, 25] and [17].

The aim of this paper is to propose an OWA operator for gradual numbers and gradual intervals. The crucial point of this is the existence of a linear order. As is stated in [10] the set of gradual numbers is not linearly ordered and the same holds for the set of gradual intervals. Hence our first objection is to introduce some linear order on these two sets. This is of course only possible in discrete case, thus we first introduce a linear order on the set of all discrete gradual real numbers. Then, based on the idea of admissible orders [3], we propose a linear order on the set of all discrete gradual intervals. Having these linear orders we can proceed to the definition of OWA operator for discrete gradual real numbers (DGOWA) and for discrete gradual intervals (DGIOWA).

There are two essential differences between our DGIOWA and OWA operators defined in [31] and [15]: 1. the weights of later are fuzzy sets whereas the weights of former are gradual numbers; 2. the former operates on the set of all discrete gradual intervals in $[0, 1]$ whereas the later operates on the set of all normal convex fuzzy sets in $[0, 1]$ which is in a discrete case the proper subset of the set of all discrete gradual intervals, Hence, from this point of view the DGIOWA is more general than the existing OWA operators. In some situations it is more natural to represent information by a gradual interval (which is not a fuzzy set) than by a fuzzy set. For instance, in multi-expert decision making problem, a decision maker has more freedom in assessment of alternatives in case of using gradual intervals and subsequent application of DGIOWA. We discuss this issue in Section 4.

It is worth pointing out that we also show the relation between partial order of gradual intervals and partial order of fuzzy intervals (fuzzy truth values) used in the type-2 fuzzy sets setting (see e. g. [5, 18]).

We apply proposed linear order and DGIOWA operator in a multi-expert decision making (MEDM) problem. An MEDM problem [24] can be described as follows: we have a set of alternatives, a set of experts, and each expert provides a preference on the set of alternatives. A decision maker (DM) is looking for the best alternative. A general model for solving an MEDM problem is introduced in [6, 12]:

1. *Making the information uniform.* It is not our purpose to study this point. We assume that the experts' preferences are uniform, represented by utility functions (in our case functions that associate each alternative with a discrete gradual interval).

2. *Application of a selection process*, which is applied in two steps [2]:

- *Aggregation phase*: the aim is to combine the individual preferences to obtain a collective preference of each alternative. We apply our DGIOWA operator in this phase.
- *Exploitation phase*: the aim is to obtain an ordering of the collective preferences, and to choose the best alternative(s). We apply our linear order for discrete gradual intervals in this phase.

The paper is organized as follows. In Section 2 we introduce a linear order on the set of all discrete gradual real numbers and define an OWA operator on this set. In Section 3 we discuss the relation between gradual intervals and fuzzy intervals, then introduce a linear order and OWA operator on the set of all discrete gradual intervals. Section 4 presents an application of proposed OWA operators and linear orders to real world MEDM problem. The conclusions are discussed in Section 5.

2. OWA OPERATORS FOR DISCRETE GRADUAL REAL NUMBERS

The aim of this section is to propose OWA operator for discrete gradual real numbers.

2.1. Discrete gradual real numbers

Definition 2.1. (Dubois and Prade [9], Fortin et al. [10]) A gradual real number \check{r} is defined by an assignment function $B_{\check{r}} : (0, 1] \rightarrow \mathbb{R}$. The set of all gradual real numbers is denoted by $G(\mathbb{R})$.

In this article we consider *discrete gradual real numbers* $B_{\check{r}} : \{\alpha_1, \dots, \alpha_k\} \rightarrow \mathbb{R}$, where $0 < \alpha_1 < \alpha_2 < \dots < \alpha_k = 1$ (mostly, $\alpha_i = \frac{i}{k}$). For given k and $0 < \alpha_1 < \alpha_2 < \dots < \alpha_k = 1$, the set of all discrete gradual real numbers is denoted by $DG_k(\mathbb{R})$. For simplicity we do not distinguish between gradual number and its assignment function, thus we write $\check{r}(\alpha)$ instead of $B_{\check{r}}(\alpha)$. We say that $\check{r}, \check{s} \in DG_k(\mathbb{R})$ are equal, write $\check{r} = \check{s}$, if $\check{r}(\alpha_i) = \check{s}(\alpha_i)$ for all $i = 1, \dots, k$.

Gradual numbers encompass real numbers: for every $r \in \mathbb{R}$ there exists $\check{r} \in DG_k(\mathbb{R})$ such that $\check{r}(\alpha_i) = r$ for all $i = 1, \dots, k$. We will write $\check{r} \stackrel{\circ}{=} r$ to emphasize that r and \check{r} are objects of different kinds. For instance, instead of $\check{r}(\alpha_i) = 1$ for all $i = 1, \dots, k$ we will write $\check{r} \stackrel{\circ}{=} 1$, or simply $\check{1}$.

2.2. Extended aggregation functions

Definition 2.2. A function $A^{(n)} : [0, 1]^n \rightarrow [0, 1]$ is called an n -ary aggregation function on $[0, 1]$ if it satisfies the conditions:

- (1) $A^{(n)}(0, \dots, 0) = 0$ and $A^{(n)}(1, \dots, 1) = 1$;
- (2) $A^{(n)}$ is nondecreasing in each variable;
- (3) $A^{(1)}(x) = x$ for all $x \in [0, 1]$.

Definition 2.3. (Grabisch et al. [11]) A mapping $\mathbb{A} : \bigcup_{n \in \mathbb{N}} [0, 1]^n \rightarrow [0, 1]$ is called an extended aggregation function on $\bigcup_{n \in \mathbb{N}} [0, 1]^n$ if its restriction $A^{(n)} := \mathbb{A}|_{[0,1]^n}$ to $[0, 1]^n$ is an aggregation function on $[0, 1]^n$ for all $n \in \mathbb{N}$.

We will sometimes use symbol $\mathbb{A}_{\mathbf{v}}$ instead of \mathbb{A} to emphasize that extended aggregation function may (not necessarily) depend on a weighting vector \mathbf{v} , e.g. weighted averages.

Definition 2.4. We say that an extended aggregation operator $\mathbb{A}_{\mathbf{v}}$ satisfies property (P1) for a fixed $k \in \mathbb{N}$ if:

(P1) For all $(x_1, \dots, x_k), (y_1, \dots, y_k) \in [0, 1]^k$ it holds

$$\left. \begin{aligned} \mathbb{A}_{\mathbf{v}_1}(x_1) &= \mathbb{A}_{\mathbf{v}_1}(y_1) \\ \mathbb{A}_{\mathbf{v}_2}(x_1, x_2) &= \mathbb{A}_{\mathbf{v}_2}(y_1, y_2) \\ &\vdots \\ \mathbb{A}_{\mathbf{v}_k}(x_1, \dots, x_k) &= \mathbb{A}_{\mathbf{v}_k}(y_1, \dots, y_k) \end{aligned} \right\} \Rightarrow \begin{cases} x_1 = y_1 \\ \vdots \\ x_k = y_k \end{cases}$$

where $\mathbf{v}_1 = (v_{11}), \dots, \mathbf{v}_k = (v_{k1}, \dots, v_{kk})$ are any weighting vectors with $v_{ij} \neq 0$ for all $i = 1, \dots, k, j = 1, \dots, i$.

Remark 2.5. Examples of extended aggregation functions which do not satisfy property (P1) are e.g. maximum and minimum. On the other hand, typical function which satisfies the property is arithmetic mean, or more generally, any weighted average with non-zero weights.

If $I = \{i_1, \dots, i_m\}$ is a finite index set, we will write shortly $\mathbb{A}(x_i)_{i \in I}$ instead of $\mathbb{A}(x_{i_1}, \dots, x_{i_m})$.

2.3. Linear order of discrete gradual real numbers

We propose a linear order of discrete gradual real numbers in this section.

Our order (comparison of \check{r} and \check{s}) is based on the comparison of extended aggregation functions (EAFs, for short) $\mathbb{A}_{\mathbf{v}}$ of $\check{r}(\alpha_1), \dots, \check{r}(\alpha_k)$ and $\check{s}(\alpha_1), \dots, \check{s}(\alpha_k)$ (step 2 of the following algorithm) w.r.t. given weighting vector \mathbf{v} (even if $\mathbb{A}_{\mathbf{v}}$ does not depend on \mathbf{v}). If the EAFs are equal, we omit the set of all elements with minimal weight (step 3), redistribute the sum of their weights to the rest of the elements (step 5, we discuss the technique of redistribution below the algorithm) and compare EAFs again (step 2). We repeat these steps till we get a strict inequality or we omit all the elements. If the later is true, we refine our procedure as follows: we always omit exactly one element - the element corresponding to minimal α_i among the elements with minimal weight (step 4), redistribute its weight to the rest of the elements (step 5) and compare EAFs (step 2). We repeat these steps till we get a strict inequality or we omit all the elements.

Algorithm 1.

Let $\mathbb{A}_{\mathbf{v}} : \bigcup_{n \in \mathbb{N}} [0, 1]^n \rightarrow [0, 1]$ be an extended aggregation function, $\check{r}, \check{s} \in DG_k(\mathbb{R})$ be discrete gradual real numbers, $\mathbf{v} = (v_1, \dots, v_k) \in [0, 1]^k$ be a weighting vector with $v_1 + \dots + v_k = 1$ and $g : \{1, 2, \dots, k\} \rightarrow [0, 1]$ be a function with appropriate properties specified in the algorithm.

1. *Initialization:*

- $I := \{1, \dots, k\}.$
- $M := \emptyset.$
- $s := 1.$
- $v'_i := v_i$ for all $i \in I.$

2. *Comparison:*

- If $\mathbb{A}_{\mathbf{v}'}(\check{r}(\alpha_i))_{i \in I} < \mathbb{A}_{\mathbf{v}'}(\check{s}(\alpha_i))_{i \in I}$, then $\check{r} \prec_{\mathbb{A}, \mathbf{v}, g} \check{s}$ (END).
- If $\mathbb{A}_{\mathbf{v}'}(\check{r}(\alpha_i))_{i \in I} > \mathbb{A}_{\mathbf{v}'}(\check{s}(\alpha_i))_{i \in I}$, then $\check{s} \prec_{\mathbb{A}, \mathbf{v}, g} \check{r}$ (END).
- If $s = 1$, then go to step 3.
- If $s = 2$, then go to step 4.

3. *Omitting 1:*

- $M := M \cup \{j \mid v_j = \min\{v_i \mid i \in I\}\}.$
- $I := I - M.$
- If $I \neq \emptyset$, then go to step 5,
- else $M := \emptyset, I := \{1, \dots, k\}, s := 2.$

4. *Omitting 2:*

- $M := M \cup \min\{j \mid v_j = \min\{v_i \mid i \in I\}\}.$
- $I := I - M.$
- If $I = \emptyset$, then $\check{r} \approx \check{s}$ (END).

5. *Redistribution of weights:*

- $\mathbf{v}' := (v'_1, \dots, v'_k)$, where $v'_i := v_i + g(i)$ for all $i \in I$, and $g : \{1, 2, \dots, k\} \rightarrow [0, 1]$ is a function with $\sum_{i \in I} v'_i = 1.$
- Go to step 2.

Remark 2.6. Algorithm 1 is a generalization of algorithm given in [27] where only weighted averages were used in step 2 (Comparison). For more detailed explanation of algorithm see above mentioned paper.

Remark 2.7. Redistribution of weights of the omitted elements is done by a *redistribution function* $g(i)$ in the following way: $v'_i = v_i + g(i)$, where $g(i) \geq 0$ for all i . The redistribution function g should satisfy $\sum_{i \in I} g(i) = \sum_{j \in M} v_j$ which clearly forces $\sum_{i \in I} v'_i = 1$ (the sum of all new weights is equal to 1). For instance, $g(i) = \frac{v_s}{k - |M|}$ where $|M|$ denotes the cardinality of M , $v_s = \sum_{j \in M} v_j$, i. e., $v'_i = v_i + \frac{v_s}{k - |M|}$ (the weight of each element is increased by the same value); or $g(i) = \frac{v_i}{1 - v_s} \cdot v_s$, i. e., $v'_i = v_i + \frac{v_i}{1 - v_s} \cdot v_s$ (the weights of individual elements are increased in proportion of the original weights).

Theorem 2.8. Let $\prec_{\mathbb{A}, \mathbf{v}, g}$ be given by Algorithm 1 for some fixed redistribution function g , weighting vector $\mathbf{v} = (v_1, \dots, v_k) \in (0, 1]^k$ with $v_1 + \dots + v_k = 1$ and extended aggregation function $\mathbb{A}_{\mathbf{v}}$ satisfying property (P1). A binary relation $\preceq_{\mathbb{A}, \mathbf{v}, g}$ defined by

$$\check{r} \preceq_{\mathbb{A}, \mathbf{v}, g} \check{s} \quad \text{iff} \quad (\check{r} \prec_{\mathbb{A}, \mathbf{v}, g} \check{s} \quad \text{or} \quad \check{r} = \check{s})$$

is a linear order on the set of all discrete gradual real numbers.

Proof. It is obvious that $\preceq_{\mathbb{A}, \mathbf{v}, g}$ is a partial order on the set of all gradual real numbers. Hence it is sufficient to show that each pair \check{r}, \check{s} of gradual real numbers is comparable by $\preceq_{\mathbb{A}, \mathbf{v}, g}$. From Algorithm 1 we obtain $\check{r} \prec_{\mathbb{A}, \mathbf{v}, g} \check{s}, \check{s} \prec_{\mathbb{A}, \mathbf{v}, g} \check{r}$, or

$$\begin{aligned} \mathbb{A}_{\mathbf{v}_1}(\check{r}(\alpha_i))_{i \in I_1} &= \mathbb{A}_{\mathbf{v}_1}(\check{s}(\alpha_i))_{i \in I_1}, & I_1 &= \{1, 2, \dots, k\}, \\ \mathbb{A}_{\mathbf{v}_2}(\check{r}(\alpha_i))_{i \in I_2} &= \mathbb{A}_{\mathbf{v}_2}(\check{s}(\alpha_i))_{i \in I_2}, & I_2 &\subseteq I_1, |I_2| = k - 1, \\ &\vdots & & \\ \mathbb{A}_{\mathbf{v}_k}(\check{r}(\alpha_i))_{i \in I_k} &= \mathbb{A}_{\mathbf{v}_k}(\check{s}(\alpha_i))_{i \in I_k}, & I_k &\subseteq I_{k-1}, |I_k| = 1, \end{aligned}$$

for some weighting vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ with non-zero coordinates. Hence, by property (P1) it follows that $\check{r}(\alpha_i) = \check{s}(\alpha_i)$ for all $i = 1, \dots, k$, which gives $\check{r} = \check{s}$. \square

It is clear that $\preceq_{\mathbb{A}, \mathbf{v}, g}$ refines partial order of gradual real numbers [10]: $\check{r} \leq \check{s}$ iff $\check{r}(\alpha_i) \leq \check{s}(\alpha_i)$ for all $i = 1, \dots, k$. This means that, for all extended aggregation functions \mathbb{A} , weighting vectors \mathbf{v} and redistribution functions g it holds:

$$\check{r} \preceq_{\mathbb{A}, \mathbf{v}, g} \check{s} \quad \text{whenever} \quad \check{r} \leq \check{s}. \tag{1}$$

Moreover, our linear order $\preceq_{\mathbb{A}, \mathbf{v}, g}$ generalizes the 'standard' order of real numbers \leq , i. e.,

$$\check{r} \preceq_{\mathbb{A}, \mathbf{v}, g} \check{s} \quad \text{iff} \quad r \leq s \tag{2}$$

for all $r, s \in R$ where $\check{r} \overset{\circ}{=} r$ and $\check{s} \overset{\circ}{=} s$.

Example 1. Let extended aggregation function $\mathbb{A}_{\mathbf{v}}$ be a weighted average, weighting vector \mathbf{v} be given by the fifth column of Table 1, discrete gradual real numbers \check{r}, \check{s} (see Figure 1 on the left) by the third and fourth columns, and let redistribution function be $g(i) = \frac{v_s}{k-|M|}$. Then $\check{r} \prec_{\mathbb{A}, \mathbf{v}, g} \check{s}$ (the detailed application of Algorithm 1 is showed in Table 1).

Remark 2.9. Although gradual numbers are functions from $(0,1]$ to real numbers, we depict them in a reversed form, i. e. independent variable $\alpha_i \in (0, 1]$ is on the vertical axis and dependent variables $\check{r}(\alpha_i), \check{s}(\alpha_i)$ are on the horizontal axis. Note that we work with discrete gradual numbers, hence only the points are important and the lines which connect the points are added just for better lucidity.

i	α_i	$\check{r}(\alpha_i)$	$\check{s}(\alpha_i)$	v_i	v'_i	v'_i	v'_i	v'_i	v'_i
10	1	3	2	0.14	0.186	0.3	0.147	0.1575	0.17
9	0.9	3	3	0.14	0.186	0.3	0.147	0.1575	0.17
8	0.8	3	4	0.14	0.186	0.3	0.147	0.1575	0.17
7	0.7	4	5	0.1	0.146		0.107	0.1175	0.13
6	0.6	4	4	0.1	0.146		0.107	0.1175	0.13
5	0.5	4	3	0.1	0.146		0.107	0.1175	0.13
4	0.4	3	4	0.07			0.07	0.0875	0.1
3	0.3	3	2	0.07			0.07	0.0875	
2	0.2	2	2	0.07			0.07		
1	0.1	3	3	0.07					
$WA(\check{r})$				3.23	3.44	3	3.245	3.3525	3.39
$WA(\check{s})$				3.23	3.44	3	3.245	3.3525	3.49
order				?	?	?	?	?	$\check{r} \prec_{\mathbb{A}, \mathbf{v}, g} \check{s}$

Tab. 1. Application of Algorithm 1 (Example 1). By $WA(\check{r})$ we denote weighted average of $\check{r}(\alpha_i)$ s.

Example 2. It is easy to check that the linear order $\prec_{\mathbb{A}, \mathbf{v}, g}$ depends on weighting vector \mathbf{v} . We will now show that it also depends on redistribution function g . Let extended aggregation function $\mathbb{A}_{\mathbf{v}}$ be a weighted average, weighting vector \mathbf{v} be given by the fifth column of Table 2, discrete gradual real numbers \check{r}, \check{s} by the third and fourth columns, and let redistribution functions be $g_1(i) = \frac{v_i}{1-v_s} \cdot v_s$ (the sixth and seventh columns) and $g_2(i) = \frac{v_s}{k-|M|}$ (the eighth column). Then $\check{s} \prec_{\mathbb{A}, \mathbf{v}, g_1} \check{r}$ and $\check{r} \prec_{\mathbb{A}, \mathbf{v}, g_2} \check{s}$ (see Table 2).

i	α_i	$\check{r}(\alpha_i)$	$\check{s}(\alpha_i)$	v_i	$v'_i(g_1)$	$v'_i(g_1)$	$v'_i(g_2)$
10	1	5	2	0.14	0.1591	0.1944	0.155
9	0.9	5	2	0.14	0.1591	0.1944	0.155
8	0.8	4	3	0.12	0.1364	0.16	0.135
7	0.7	4	3	0.12	0.1364	0.16	0.135
6	0.6	3	5	0.1	0.1136	0.1389	0.115
5	0.5	2	6	0.1	0.1136	0.1389	0.115
4	0.4	2	5	0.08	0.0909		0.095
3	0.3	2	5	0.08	0.0909		0.095
2	0.2	3	4	0.06			
1	0.1	3	2	0.06			
$WA(\check{r})$				3.54	3.6136	3.9722	3.585
$WA(\check{s})$				3.54	3.6136	3.3056	3.645
order				?	?	$\check{s} \prec_{\mathbb{A}, \mathbf{v}, g_1} \check{r}$	$\check{r} \prec_{\mathbb{A}, \mathbf{v}, g_2} \check{s}$

Tab. 2. Application of Algorithm 1 (Example 2). By $v'_i(g_1)$ and $v'_i(g_2)$ are denoted weights calculated via redistribution functions g_1 and g_2 , respectively.

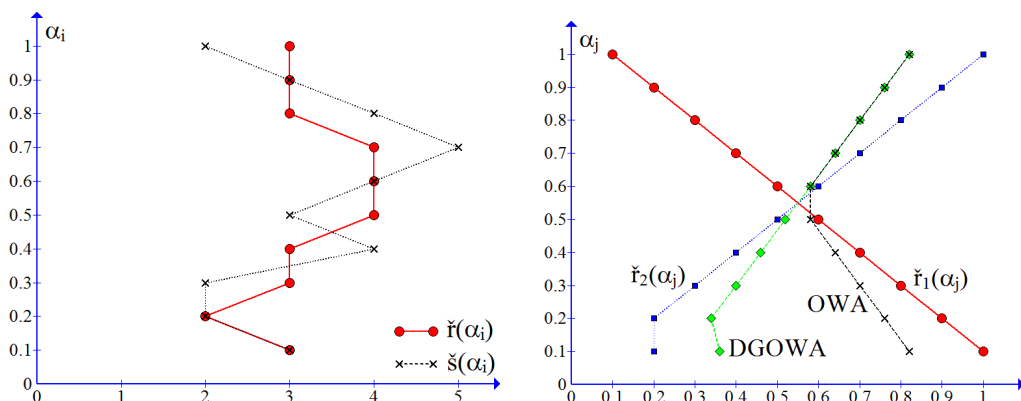


Fig. 1. See Example 1 (left) or Example 3 (right).

2.4. OWA operators for discrete gradual real numbers

Since we have a linear order, we can define an OWA operator for discrete gradual numbers. Instead of using real numbers we employ gradual numbers as weights.

Note that the sum of discrete gradual real numbers $\check{w}_1, \check{w}_2 \in DG_k(\mathbb{R})$ is given by $(\check{w}_1 + \check{w}_2)(\alpha_i) = \check{w}_1(\alpha_i) + \check{w}_2(\alpha_i)$ for all $i = 1, \dots, k$.

Definition 2.10. Let $\check{\mathbf{w}} = (\check{w}_1, \dots, \check{w}_n) \in DG_k([0, 1])^n$ with $\check{w}_1 + \dots + \check{w}_n = \check{1}$ be a weighting vector of discrete gradual numbers. A discrete gradual numbers OWA operator (DGOWA for short) associated with $\check{\mathbf{w}}$ is a mapping $DGOWA_{\check{\mathbf{w}}}^{\preceq_{\mathbb{A}, \mathbf{v}, g}} : DG_k(\mathbb{R})^n \rightarrow DG_k(\mathbb{R})$ defined by

$$DGOWA_{\check{\mathbf{w}}}^{\preceq_{\mathbb{A}, \mathbf{v}, g}}(\check{r}_1, \dots, \check{r}_n) = \sum_{i=1}^n \check{w}_i \check{r}_{(i)}, \tag{3}$$

where $\check{r}_{(i)}$, $i = 1, \dots, n$, denote the i th greatest component of the input $(\check{r}_1, \dots, \check{r}_n)$ with respect to a linear order $\preceq_{\mathbb{A}, \mathbf{v}, g}$.

Application of the gradual number arithmetic [10] to equation (3) leads to computation of k independent results:

$$DGOWA_{\check{\mathbf{w}}}^{\preceq_{\mathbb{A}, \mathbf{v}, g}}(\check{r}_1, \dots, \check{r}_n)(\alpha_j) = \sum_{i=1}^n \check{w}_i(\alpha_j) \check{r}_{(i)}(\alpha_j) \tag{4}$$

for all $j = 1, \dots, k$. This means that DGOWA operator is very easily computable.

Obviously, the proposed DGOWA operator can also be used with real weights w_1, \dots, w_n . Moreover, the following theorem shows that our operator encompasses 'standard' OWA operator for real numbers.

Theorem 2.11. Let $\mathbf{w} = (w_1, \dots, w_n) \in [0, 1]^n$ satisfy $w_1 + \dots + w_n = 1$, and $r_1, \dots, r_n \in \mathbb{R}$. Let $\check{w}_i \doteq w_i$, $\check{r}_i \doteq r_i$, for $i = 1, \dots, n$, be discrete gradual real numbers, and $\check{\mathbf{w}} = (\check{w}_1, \dots, \check{w}_n)$. Then

$$DGOWA_{\check{\mathbf{w}}}^{\check{\succeq}_{\mathbb{A}, \mathbf{v}, g}}(\check{r}_1, \dots, \check{r}_n) \doteq OWA_{\mathbf{w}}(r_1, \dots, r_n)$$

for any linear order $\check{\succeq}_{\mathbb{A}, \mathbf{v}, g}$.

Proof. The proof starts with the observation that $r_{(n)} \leq \dots \leq r_{(1)}$ implies $\check{r}_{(n)} \check{\succeq}_{\mathbb{A}, \mathbf{v}, g} \dots \check{\succeq}_{\mathbb{A}, \mathbf{v}, g} \check{r}_{(1)}$ for any linear order $\check{\succeq}_{\mathbb{A}, \mathbf{v}, g}$. Then

$$DGOWA_{\check{\mathbf{w}}}^{\check{\succeq}_{\mathbb{A}, \mathbf{v}, g}}(\check{r}_1, \dots, \check{r}_n) = \sum_{i=1}^n \check{w}_i \check{r}_{(i)} \doteq \sum_{i=1}^n w_i r_{(i)} = OWA_{\mathbf{w}}(r_1, \dots, r_n).$$

□

It is worth pointing out that, in general, DGOWA operator cannot be represented by k OWA operators in the following form:

$$DGOWA_{(\check{w}_1, \dots, \check{w}_n)}^{\check{\succeq}_{\mathbb{A}, \mathbf{v}, g}}(\check{r}_1, \dots, \check{r}_n)(\alpha_j) = OWA_{(\check{w}_1(\alpha_j), \dots, \check{w}_n(\alpha_j))}(\check{r}_1(\alpha_j), \dots, \check{r}_n(\alpha_j)) \quad (5)$$

for all $j = 1, \dots, k$. The reason is that the i th greatest element of $\check{r}_1, \dots, \check{r}_n$ with respect to $\check{\succeq}_{\mathbb{A}, \mathbf{v}, g}$ need not correspond to the i th greatest element of $\check{r}_1(\alpha_j), \dots, \check{r}_n(\alpha_j)$ with respect to the 'standard' order of real numbers. We next give a counterexample.

Example 3. Let discrete gradual real numbers \check{r}_1, \check{r}_2 be given by Table 3, see Figure 1 (on the right), and let \mathbb{A} be weighted average. Clearly, $\check{r}_1 \check{\succeq}_{\mathbb{A}, \mathbf{v}, g} \check{r}_2$ for weighting vector $\mathbf{v} = (0.1, 0.1, \dots, 0.1)$. We can see that $DGOWA_{(0.8, 0.2)}^{\check{\succeq}_{\mathbb{A}, \mathbf{v}, g}}(\check{r}_1, \check{r}_2)(\alpha_j)$ is not equal to $OWA_{(0.8, 0.2)}(\check{r}_1(\alpha_j), \check{r}_2(\alpha_j))$ for $j \in \{1, \dots, 5\}$. The difference lays in the fact that in DGOWA the greater weight 0.8 is assigned to the greater element \check{r}_2 for all α_j , however, in OWA operator the greater weight is assigned to the greater number of the pair $\check{r}_1(\alpha_j), \check{r}_2(\alpha_j)$ which is $\check{r}_1(\alpha_j)$ for $j = 1, \dots, 5$, and $\check{r}_2(\alpha_j)$ for $j = 6, \dots, 10$:

$$DGOWA_{(0.8, 0.2)}^{\check{\succeq}_{\mathbb{A}, \mathbf{v}, g}}(\check{r}_1, \check{r}_2)(\alpha_j) = 0.8 \cdot \check{r}_2(\alpha_j) + 0.2 \cdot \check{r}_1(\alpha_j)$$

$$OWA_{(0.8, 0.2)}(\check{r}_1(\alpha_j), \check{r}_2(\alpha_j)) = 0.8 \cdot \max\{\check{r}_1(\alpha_j), \check{r}_2(\alpha_j)\} + 0.2 \cdot \min\{\check{r}_1(\alpha_j), \check{r}_2(\alpha_j)\}$$

For complete results see Table 3 and Figure 1. Recall that the difference disappears if $\check{r}_1 \leq \check{r}_2$.

3. OWA OPERATORS FOR DISCRETE GRADUAL INTERVALS

In this section we extend the results of previous section to the setting of discrete gradual intervals (which encompasses also fuzzy intervals). First, gradual intervals and aggregation functions for gradual numbers are described, then linear order of discrete gradual intervals is proposed, and finally, OWA operator for discrete gradual intervals is defined and discussed.

j	α_j	$\check{r}_1(\alpha_j)$	$\check{r}_2(\alpha_j)$	$DGOWA$	OWA
10	1	0.1	1	0.82	0.82
9	0.9	0.2	0.9	0.76	0.76
8	0.8	0.3	0.8	0.7	0.7
7	0.7	0.4	0.7	0.64	0.64
6	0.6	0.5	0.6	0.58	0.58
5	0.5	0.6	0.5	0.52	0.58
4	0.4	0.7	0.4	0.46	0.64
3	0.3	0.8	0.3	0.4	0.7
2	0.2	0.9	0.2	0.34	0.76
1	0.1	1	0.2	0.36	0.84

Tab. 3. A counterexample to equation (5). See Example 3.

3.1. Gradual intervals and fuzzy intervals

Gradual interval [10], i.e. interval of gradual real numbers, is given by an ordered pair of gradual numbers $\tilde{X} = [\check{x}^-, \check{x}^+]$, where \check{x}^- is called the *left profile* (or fuzzy lower bound) of \tilde{X} and \check{x}^+ the *right profile* (or fuzzy upper bound). The set of all gradual intervals will be denoted by $\mathbb{I}(G(\mathbb{R}))$:

$$\mathbb{I}(G(\mathbb{R})) = \{[\check{x}^-, \check{x}^+] \mid \check{x}^-, \check{x}^+ \in G(\mathbb{R}), \check{x}^- \leq \check{x}^+\}.$$

If \check{x}^-, \check{x}^+ are discrete gradual numbers on the same set $\{\alpha_1, \dots, \alpha_k\}$, then $[\check{x}^-, \check{x}^+]$ is called a *discrete gradual interval*. The set of all discrete gradual intervals is denoted by $\mathbb{I}(DG_k(\mathbb{R}))$.

A discrete gradual interval $[\check{x}^-, \check{x}^+]$ with strictly increasing left profile \check{x}^- and strictly decreasing right profile \check{x}^+ on $\{\alpha_1, \dots, \alpha_k\}$ characterizes *fuzzy interval* (fuzzy subset of real line whose α -cuts are closed intervals [7, 22]) $f : \mathbb{R} \rightarrow \{\alpha_1, \dots, \alpha_k\}$ in the following way:

$$f(x) = \begin{cases} \alpha_i, & \text{if } x \in [\check{x}^-(\alpha_i), \check{x}^-(\alpha_{i+1})] \cup (\check{x}^+(\alpha_{i+1}), \check{x}^+(\alpha_i)], \quad i = 1, \dots, k - 1, \\ 1, & \text{if } x \in [\check{x}^-(1), \check{x}^+(1)], \\ 0, & \text{otherwise.} \end{cases} \tag{6}$$

Fuzzy intervals f given by (6) will be called *fuzzy intervals with finite range* to emphasize the difference between them and discrete fuzzy intervals, the latter having discrete domain and the former discrete range. Obviously, α -cuts of f in $\alpha_i, i = 1, \dots, k$, are closed intervals (see Figure 2):

$$f_{\alpha_i} = [\check{x}^-(\alpha_i), \check{x}^+(\alpha_i)].$$

It is worth pointing out that all our results for discrete gradual intervals hold also for fuzzy intervals with finite range.

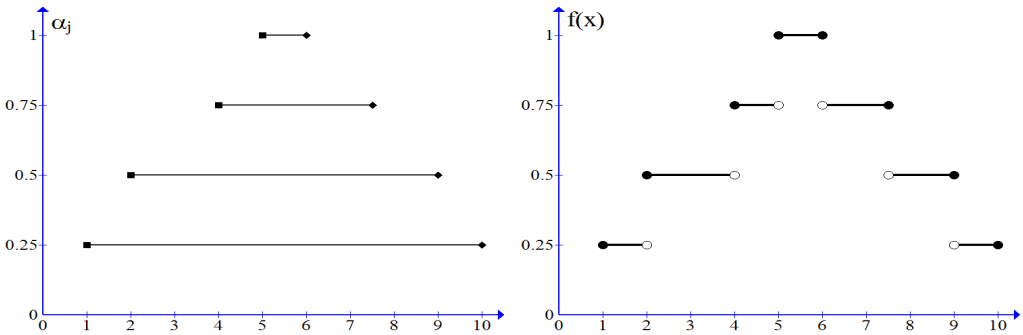


Fig. 2. Discrete gradual interval $[\check{x}^-, \check{x}^+]$ with strictly increasing left profile \check{x}^- and strictly decreasing right profile \check{x}^+ (on the left) and corresponding fuzzy interval f given by (6) (on the right).

We have two natural partial orders on $\mathbb{I}(DG_k(\mathbb{R}))$ based on the partial order \leq and linear order $\preceq_{\mathbb{A}, \mathbf{v}, g}$ on $DG_k(\mathbb{R})$:

$$[\check{x}^-, \check{x}^+] \preceq_2 [\check{y}^-, \check{y}^+] \quad \text{iff} \quad (\check{x}^- \leq \check{y}^- \quad \text{and} \quad \check{x}^+ \leq \check{y}^+) \tag{7}$$

$$[\check{x}^-, \check{x}^+] \preceq_2^{\mathbb{A}, \mathbf{v}, g} [\check{y}^-, \check{y}^+] \quad \text{iff} \quad (\check{x}^- \preceq_{\mathbb{A}, \mathbf{v}, g} \check{y}^- \quad \text{and} \quad \check{x}^+ \preceq_{\mathbb{A}, \mathbf{v}, g} \check{y}^+) \tag{8}$$

which correspond to commonly used partial order of 'standard' intervals: $[a, b] \leq_2 [c, d]$ iff $a \leq c, b \leq d$. Clearly, neither \preceq_2 nor $\preceq_2^{\mathbb{A}, \mathbf{v}, g}$ is linear. We emphasize that the order \preceq_2 coincides (in the sense of the following theorem) with the order \sqsubseteq_{T_2} of membership grades f, g of type-2 fuzzy sets. Recall that \sqsubseteq_{T_2} is given by:

$$f \sqsubseteq_{T_2} g \quad \text{iff} \quad f \sqcap g = f, \quad (f \sqcap g)(z) = \sup_{x \wedge y = z} (f(x) \wedge g(y)),$$

where f, g are functions from $[0, 1]$ to $[0, 1]$ (called also fuzzy truth values). See [13, 20, 26, 28].

Theorem 3.1. The orders \preceq_2 and \sqsubseteq_{T_2} coincide on the set $GV \subseteq \mathbb{I}(DG_k([0, 1]))$ of all discrete gradual intervals with strictly increasing left profiles and strictly decreasing right profiles, i. e.,

$$[\check{x}^-, \check{x}^+] \preceq_2 [\check{y}^-, \check{y}^+] \quad \text{iff} \quad f \sqsubseteq_{T_2} g, \tag{9}$$

where $[\check{x}^-, \check{x}^+], [\check{y}^-, \check{y}^+] \in GV$, and f, g arise by equation (6) from $[\check{x}^-, \check{x}^+], [\check{y}^-, \check{y}^+]$, respectively. Moreover,

$$[\check{x}^-, \check{x}^+] \preceq_2^{\mathbb{A}, \mathbf{v}, g} [\check{y}^-, \check{y}^+] \quad \text{whenever} \quad f \sqsubseteq_{T_2} g. \tag{10}$$

Proof. By [28, Proposition 16] (or [13, Theorem 1]), $f \sqsubseteq_{T_2} g$ holds if and only if $\check{x}^- \leq \check{y}^-$ and $\check{x}^+ \leq \check{y}^+$, hence if and only if $[\check{x}^-, \check{x}^+] \preceq_2 [\check{y}^-, \check{y}^+]$ by (7). Thus (9) is proved. Now let $f \sqsubseteq_{T_2} g$. Then $[\check{x}^-, \check{x}^+] \preceq_2 [\check{y}^-, \check{y}^+]$ by (9), hence $\check{x}^- \leq \check{y}^-$ and $\check{x}^+ \leq \check{y}^+$ by (7), consequently, $\check{x}^- \preceq_{\mathbb{A}, \mathbf{v}, g} \check{y}^-$ and $\check{x}^+ \preceq_{\mathbb{A}, \mathbf{v}, g} \check{y}^+$ by (1), and finally, $[\check{x}^-, \check{x}^+] \preceq_2^{\mathbb{A}, \mathbf{v}, g} [\check{y}^-, \check{y}^+]$ by (8). Thus (10) is proved. \square

3.2. Aggregation functions for discrete gradual real numbers

We generalize definition of aggregation functions on $[0, 1]$ (see e. g. [1, 8, 11, 19]) to the gradual numbers setting in a straightforward way.

Definition 3.2. A function $\check{A} : DG_k([0, 1])^n \rightarrow DG_k([0, 1])$ is called an n -ary aggregation function on $DG_k([0, 1])$ if it satisfies the conditions:

- (A1) $\check{A}(\check{0}, \dots, \check{0}) = \check{0}$;
- (A2) $\check{A}(\check{1}, \dots, \check{1}) = \check{1}$;
- (A3) $\check{x}_1 \leq \check{y}_1, \dots, \check{x}_n \leq \check{y}_n$ imply $\check{A}(\check{x}_1, \dots, \check{x}_n) \leq \check{A}(\check{y}_1, \dots, \check{y}_n)$ for all $\check{x}_1, \check{y}_1, \dots, \check{x}_n, \check{y}_n \in DG_k([0, 1])$.

An n -ary aggregation function \check{A} on $DG_k([0, 1])$ is said to be *induced by* n -ary aggregation functions A_1, \dots, A_k on $[0, 1]$ if

$$\check{A}(\check{x}_1, \dots, \check{x}_n)(\alpha_j) = A_j(\check{x}_1(\alpha_j), \dots, \check{x}_n(\alpha_j))$$

for all $j = 1, \dots, k$. If $A_1 = A_2 = \dots = A_k = A$, \check{A} is said to be *induced by* A . However, there are aggregation functions on $DG_k([0, 1])$ which cannot be induced by any aggregation functions A_1, \dots, A_k , for instance:

$$\check{A}(\check{x}_1, \check{x}_2)(\alpha_j) = \frac{\sum_{i=1}^k \check{x}_1(\alpha_i) + \sum_{i=1}^k \check{x}_2(\alpha_i)}{2k} \cdot \check{x}_1(\alpha_j)\check{x}_2(\alpha_j)$$

for all $j = 1, \dots, k$.

It is easy to check that the DGOWA operator given by Definition 2.10 satisfies (A1) – (A3) on $DG_k([0, 1])$.

Theorem 3.3. DGOWA operator on $DG_k([0, 1])$ is an aggregation function according to Definition 3.2.

Proof. The proof is straightforward and therefore omitted. □

3.3. Linear order of discrete gradual intervals

Construction of our linear order of gradual intervals generalizes the ideas of [3, 4] where the authors proposed so-called admissible order of intervals.

Definition 3.4. An order \sqsubseteq on $\mathbb{I}(DG_k(\mathbb{R}))$ is called admissible on the set of all discrete gradual intervals, if it is linear and refines the order \leq_2 given by (7), i. e., for all $[\check{x}^-, \check{x}^+], [\check{y}^-, \check{y}^+] \in \mathbb{I}(DG_k(\mathbb{R}))$,

$$[\check{x}^-, \check{x}^+] \sqsubseteq [\check{y}^-, \check{y}^+] \quad \text{whenever} \quad [\check{x}^-, \check{x}^+] \leq_2 [\check{x}^-, \check{x}^+].$$

Some admissible orders can be *generated* by a pair of aggregation functions A, B (see [3, Proposition 3.2]). From now on, we only turn our attention to gradual numbers and intervals on $[0, 1]$. However, our results remain valid for gradual numbers and intervals on any bounded partially ordered set (L, \leq) .

Theorem 3.5. Let $\check{A}, \check{B} : DG_k([0, 1])^2 \rightarrow DG_k([0, 1])$ be binary aggregation functions for gradual numbers on $[0, 1]$ such that for all $\check{x}, \check{y}, \check{u}, \check{v} \in DG_k([0, 1])$ the equalities $\check{A}(\check{x}, \check{y}) = \check{A}(\check{u}, \check{v})$ and $\check{B}(\check{x}, \check{y}) = \check{B}(\check{u}, \check{v})$ hold if and only if $\check{x} = \check{u}$ and $\check{y} = \check{v}$. Let $\sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}$ be a relation on $\mathbb{I}(DG_k([0, 1]))$ given by:

$$[\check{x}^-, \check{x}^+] \sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} [\check{y}^-, \check{y}^+] \quad \text{iff} \quad \check{A}(\check{x}^-, \check{x}^+) \prec_{\mathbb{A}, \mathbf{v}, g} \check{A}(\check{y}^-, \check{y}^+) \quad \text{or} \\ (\check{A}(\check{x}^-, \check{x}^+) = \check{A}(\check{y}^-, \check{y}^+) \quad \text{and} \quad \check{B}(\check{x}^-, \check{x}^+) \preceq_{\mathbb{A}, \mathbf{v}, g} \check{B}(\check{y}^-, \check{y}^+)). \quad (11)$$

Then $\sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}$ is an admissible order on $\mathbb{I}(DG_k([0, 1]))$.

Proof. Let us first show that $\sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}$ is a partial order. Reflexivity is obvious. Antisymmetry: Let $[\check{x}^-, \check{x}^+] \sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} [\check{y}^-, \check{y}^+]$ and $[\check{y}^-, \check{y}^+] \sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} [\check{x}^-, \check{x}^+]$. Since $\preceq_{\mathbb{A}, \mathbf{v}, g}$ is linear, it follows that

$$\check{A}(\check{x}^-, \check{x}^+) = \check{A}(\check{y}^-, \check{y}^+), \check{B}(\check{x}^-, \check{x}^+) \preceq_{\mathbb{A}, \mathbf{v}, g} \check{B}(\check{y}^-, \check{y}^+), \check{B}(\check{y}^-, \check{y}^+) \preceq_{\mathbb{A}, \mathbf{v}, g} \check{B}(\check{x}^-, \check{x}^+),$$

hence

$$\check{A}(\check{x}^-, \check{x}^+) = \check{A}(\check{y}^-, \check{y}^+), \check{B}(\check{x}^-, \check{x}^+) = \check{B}(\check{y}^-, \check{y}^+),$$

and consequently, by the assumption of theorem, $\check{x}^- = \check{y}^-$ and $\check{x}^+ = \check{y}^+$.

Transitivity: assumption $[\check{x}^-, \check{x}^+] \sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} [\check{y}^-, \check{y}^+]$ and $[\check{y}^-, \check{y}^+] \sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} [\check{z}^-, \check{z}^+]$ leads to one of the following four cases:

1. $\check{A}(\check{x}^-, \check{x}^+) \prec_{\mathbb{A}, \mathbf{v}, g} \check{A}(\check{y}^-, \check{y}^+)$ and $\check{A}(\check{y}^-, \check{y}^+) \prec_{\mathbb{A}, \mathbf{v}, g} \check{A}(\check{z}^-, \check{z}^+)$, which implies $\check{A}(\check{x}^-, \check{x}^+) \prec_{\mathbb{A}, \mathbf{v}, g} \check{A}(\check{z}^-, \check{z}^+)$.
2. $\check{A}(\check{x}^-, \check{x}^+) = \check{A}(\check{y}^-, \check{y}^+)$, $\check{B}(\check{x}^-, \check{x}^+) \preceq_{\mathbb{A}, \mathbf{v}, g} \check{B}(\check{y}^-, \check{y}^+)$, $\check{A}(\check{y}^-, \check{y}^+) = \check{A}(\check{z}^-, \check{z}^+)$ and $\check{B}(\check{y}^-, \check{y}^+) \preceq_{\mathbb{A}, \mathbf{v}, g} \check{B}(\check{z}^-, \check{z}^+)$, which gives $\check{A}(\check{x}^-, \check{x}^+) = \check{A}(\check{z}^-, \check{z}^+)$ and $\check{B}(\check{x}^-, \check{x}^+) \preceq_{\mathbb{A}, \mathbf{v}, g} \check{B}(\check{z}^-, \check{z}^+)$.
3. $\check{A}(\check{x}^-, \check{x}^+) \prec_{\mathbb{A}, \mathbf{v}, g} \check{A}(\check{y}^-, \check{y}^+)$, $\check{A}(\check{y}^-, \check{y}^+) = \check{A}(\check{z}^-, \check{z}^+)$ and $\check{B}(\check{y}^-, \check{y}^+) \preceq_{\mathbb{A}, \mathbf{v}, g} \check{B}(\check{z}^-, \check{z}^+)$, which leads to $\check{A}(\check{x}^-, \check{x}^+) \prec_{\mathbb{A}, \mathbf{v}, g} \check{A}(\check{z}^-, \check{z}^+)$.
4. $\check{A}(\check{x}^-, \check{x}^+) = \check{A}(\check{y}^-, \check{y}^+)$, $\check{B}(\check{x}^-, \check{x}^+) \preceq_{\mathbb{A}, \mathbf{v}, g} \check{B}(\check{y}^-, \check{y}^+)$ and $\check{A}(\check{y}^-, \check{y}^+) \prec_{\mathbb{A}, \mathbf{v}, g} \check{A}(\check{z}^-, \check{z}^+)$, which implies $\check{A}(\check{x}^-, \check{x}^+) \prec_{\mathbb{A}, \mathbf{v}, g} \check{A}(\check{z}^-, \check{z}^+)$.

From each of the four cases it follows that $[\check{x}^-, \check{x}^+] \sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} [\check{z}^-, \check{z}^+]$, thus $\sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}$ is a partial order. Linearity immediately follows from (11) and linearity of $\preceq_{\mathbb{A}, \mathbf{v}, g}$. \square

Aggregation functions \check{A}, \check{B} for discrete gradual numbers which satisfy the assumption of Theorem 3.5 can be obtained by aggregation functions for real numbers with similar property.

Lemma 3.6. Let $\check{A}, \check{B} : DG_k([0, 1])^2 \rightarrow DG_k([0, 1])$ be binary aggregation functions for gradual numbers on $[0, 1]$ induced by binary aggregation functions $A, B : [0, 1]^2 \rightarrow [0, 1]$, respectively. Let for all $x, y, u, v \in [0, 1]$ the equalities $A(x, y) = A(u, v)$ and $B(x, y) = B(u, v)$ hold if and only if $x = u$ and $y = v$. Then for all $\check{x}, \check{y}, \check{u}, \check{v} \in DG_k([0, 1])$ the equalities $\check{A}(\check{x}, \check{y}) = \check{A}(\check{u}, \check{v})$ and $\check{B}(\check{x}, \check{y}) = \check{B}(\check{u}, \check{v})$ hold if and only if $\check{x} = \check{u}$ and $\check{y} = \check{v}$.

Proof. The proof is straightforward and therefore omitted. □

From an admissible order of discrete gradual intervals one can induce a linear order of fuzzy intervals (of a special kind).

Corollary 3.7. Let V be the set of all fuzzy intervals on $[0, 1]$ with finite range $f : [0, 1] \rightarrow \{\alpha_1, \dots, \alpha_k\}$, where $0 < \alpha_1 < \dots < \alpha_k = 1$, whose α -cuts are nested closed intervals $f_{\alpha_i} = [\underline{f}_{\alpha_i}, \overline{f}_{\alpha_i}]$ for all $i = 1, \dots, k$. Then the following relation \sqsubseteq_V on V is a linear order:

$$f \sqsubseteq_V h \quad \text{iff} \quad [\check{x}^-, \check{x}^+] \sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} [\check{y}^-, \check{y}^+]$$

where $\check{x}^-(\alpha_i) = \underline{f}_{\alpha_i}$, $\check{x}^+(\alpha_i) = \overline{f}_{\alpha_i}$, $\check{y}^-(\alpha_i) = \underline{h}_{\alpha_i}$, $\check{y}^+(\alpha_i) = \overline{h}_{\alpha_i}$, and $\sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}$ is an admissible order given in Theorem 3.5.

Proof. The proof is based on the observation that equation (6) defines a bijection from the set of all discrete gradual intervals with strictly increasing left profiles and strictly decreasing right profiles to the set V . □

Example 4. Let extended aggregation function $\mathbb{A}_{\mathbf{v}}$ be weighted average. Consider the discrete gradual intervals $\check{X} = [\check{x}^-, \check{x}^+]$, $\check{Y} = [\check{y}^-, \check{y}^+]$ and $\check{Z} = [\check{z}^-, \check{z}^+]$ given by Table 4. Let an aggregation function \check{A} be induced by arithmetic mean $A(a, b) = \frac{a+b}{2}$ and \check{B} be induced by geometric mean $B(a, b) = \sqrt{ab}$. Then, for weighting vector $\mathbf{v} = (0.1, \dots, 0.1)$, WAs are:

$$\begin{aligned} \sum_{j=1}^{10} v_j \check{A}(\check{x}^-, \check{x}^+)(\alpha_j) &= \sum_{j=1}^{10} v_j \check{A}(\check{y}^-, \check{y}^+)(\alpha_j) = 0.385, \\ \sum_{j=1}^{10} v_j \check{A}(\check{z}^-, \check{z}^+)(\alpha_j) &= 0.975, \end{aligned}$$

which means that $\check{X}, \check{Y} \sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} \check{Z}$. Since $\check{A}(\check{x}^-, \check{x}^+) = \check{A}(\check{y}^-, \check{y}^+)$, we apply \check{B} to compare \check{X} and \check{Y} . From

$$\sum_{j=1}^{10} v_j \check{B}(\check{x}^-, \check{x}^+)(\alpha_j) = 0.351, \quad \sum_{j=1}^{10} v_j \check{B}(\check{y}^-, \check{y}^+)(\alpha_j) = 0.331,$$

it follows that $\check{Y} \sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} \check{X} \sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} \check{Z}$.

Now we will consider the same situation but use another weighting vector $\mathbf{u} = (0.04, 0.06, 0.08, 0.1, 0.1, 0.1, 0.1, 0.12, 0.14, 0.16)$. Then

$$\sum_{j=1}^{10} u_j \check{A}(\check{x}^-, \check{x}^+)(\alpha_j) = 0.335, \quad \sum_{j=1}^{10} u_j \check{A}(\check{y}^-, \check{y}^+)(\alpha_j) = 0.34,$$

$$\sum_{j=1}^{10} u_j \check{A}(\check{z}^-, \check{z}^+)(\alpha_j) = 0.973,$$

which implies $\check{X} \sqsubseteq_{\check{A}, \check{B}}^{\check{A}, \mathbf{u}, g} \check{Y} \sqsubseteq_{\check{A}, \check{B}}^{\check{A}, \mathbf{u}, g} \check{Z}$.

We can see that we obtained different order for different weighting vectors. Recall that, in this example, all the results hold for any redistribution function g because we did not need to redistribute the weights.

α_j	\check{X}	\check{Y}	\check{Z}	$\check{A}(\check{x}^-, \check{x}^+)$	$\check{A}(\check{y}^-, \check{y}^+)$	$\check{A}(\check{z}^-, \check{z}^+)$	$\check{B}(\check{x}^-, \check{x}^+)$	$\check{B}(\check{y}^-, \check{y}^+)$
1	[0.1,0.6]	[0.3,0.4]	[1,1]	0.35	0.35	1	0.245	0.346
0.9	[0.1,0.6]	[0.3,0.4]	[1,1]	0.35	0.35	1	0.245	0.346
0.8	[0.2,0.6]	[0.3,0.5]	[0.9,1]	0.4	0.4	0.95	0.346	0.387
0.7	[0.2,0.5]	[0.2,0.5]	[0.9,1]	0.35	0.35	0.95	0.316	0.316
0.6	[0.3,0.5]	[0.2,0.6]	[0.9,1]	0.4	0.4	0.95	0.387	0.346
0.5	[0.3,0.5]	[0.2,0.6]	[0.9,1]	0.4	0.4	0.95	0.387	0.346
0.4	[0.3,0.5]	[0.2,0.6]	[0.9,1]	0.4	0.4	0.95	0.387	0.346
0.3	[0.4,0.4]	[0.2,0.6]	[1,1]	0.4	0.4	1	0.4	0.346
0.2	[0.4,0.4]	[0.1,0.7]	[1,1]	0.4	0.4	1	0.4	0.265
0.1	[0.4,0.4]	[0.1,0.7]	[1,1]	0.4	0.4	1	0.4	0.265
$WA_{\mathbf{v}}$				0.385	0.385	0.975	0.351	0.331
$WA_{\mathbf{u}}$				0.335	0.34	0.973		

Tab. 4. The discrete gradual intervals $\check{X}, \check{Y}, \check{Z}$ from Example 4. Aggregation function \check{A} is induced by arithmetic mean, \check{B} is induced by geometric mean, $WA_{\mathbf{v}}$ denotes the weighted average associated with weighting vector \mathbf{v} and similarly for $WA_{\mathbf{u}}$ and \mathbf{u} .

3.4. OWA operators for discrete gradual intervals

Since we proposed a linear order of discrete gradual intervals, we can define OWA operator for these objects.

Definition 3.8. Let $\check{\mathbf{w}} = (\check{w}_1, \dots, \check{w}_n) \in DG_k([0, 1])^n$ with $\check{w}_1 + \dots + \check{w}_n = \check{1}$ be a weighting vector of discrete gradual numbers. A discrete gradual intervals OWA operator (DGIOWA for short) associated with $\check{\mathbf{w}}$ is a mapping $DGIOWA_{\check{\mathbf{w}}}^{\check{A}} : \mathbb{I}(DG_k([0, 1]))^n \rightarrow \mathbb{I}(DG_k([0, 1]))$ defined by

$$DGIOWA_{\check{\mathbf{w}}}^{\check{A}}([\check{x}_1^-, \check{x}_1^+], \dots, [\check{x}_n^-, \check{x}_n^+]) = \sum_{i=1}^n \check{w}_i \cdot [\check{x}_{(i)}^-, \check{x}_{(i)}^+], \tag{12}$$

where $[\check{x}_{(i)}^-, \check{x}_{(i)}^+]$, $i = 1, \dots, n$, denote the i th greatest component of the input vector $([\check{x}_1^-, \check{x}_1^+], \dots, [\check{x}_n^-, \check{x}_n^+])$ with respect to an admissible order \sqsubseteq .

Application of the interval [21] and gradual number [10] arithmetic to equation (12) leads to computation of k independent results:

$$DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq}([\check{x}_1^-, \check{x}_1^+], \dots, [\check{x}_n^-, \check{x}_n^+]) (\alpha_j) = \sum_{i=1}^n [\check{w}_i(\alpha_j) \cdot \check{x}_{(i)}^-(\alpha_j), \check{w}_i(\alpha_j) \check{x}_{(i)}^+(\alpha_j)] \quad (13)$$

for all $j = 1, \dots, k$. We emphasize that $DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq}$ is very easily computable by (13).

The DGIOWA operator satisfies basic properties of aggregation functions, i. e. the boundary conditions and monotonicity (with respect to partial order \preceq_2).

Theorem 3.9. For any operator $DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq}$ on $\mathbb{I}(DG_k([0, 1]))$ it holds:

- $DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq}([\check{0}, \check{0}], \dots, [\check{0}, \check{0}]) = [\check{0}, \check{0}]$;
- $DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq}([\check{1}, \check{1}], \dots, [\check{1}, \check{1}]) = [\check{1}, \check{1}]$;
- $[\check{x}_1^-, \check{x}_1^+] \preceq_2 [\check{y}_1^-, \check{y}_1^+], \dots, [\check{x}_n^-, \check{x}_n^+] \preceq_2 [\check{y}_n^-, \check{y}_n^+]$ imply $DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq}([\check{x}_1^-, \check{x}_1^+], \dots, [\check{x}_n^-, \check{x}_n^+]) \preceq_2 DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq}([\check{y}_1^-, \check{y}_1^+], \dots, [\check{y}_n^-, \check{y}_n^+])$ for all $[\check{x}_1^-, \check{x}_1^+], [\check{y}_1^-, \check{y}_1^+], \dots, [\check{x}_n^-, \check{x}_n^+], [\check{y}_n^-, \check{y}_n^+] \in \mathbb{I}(DG_k([0, 1]))$.

Proof. The proof immediately follows from Definition 3.8 and equations (13), (7). \square

The important point to note here is that the third property of the previous theorem, which holds for the partial order \preceq_2 , does not hold for admissible orders \sqsubseteq in general. For admissible orders it only holds if the weights are real numbers, i. e., $\check{\mathbf{w}} = (\check{w}_1, \dots, \check{w}_n)$ with $\check{w}_i \doteq w_i \in \mathbb{R}$, $i = 1, \dots, n$.

Note that, for some appropriate orders of intervals and gradual intervals, $DGIOWA$ operator is an extension of $IVOWA$, which is an operator on the set $\mathbb{I}([0, 1])$ of all closed intervals of real numbers proposed in [4], and also of 'standard' OWA operator for real numbers.

Theorem 3.10. Let $\preceq_{A,B}$ be an admissible order on $\mathbb{I}([0, 1])$ generated by A, B , and let $\sqsubseteq_{\check{A}, \check{B}}^{\check{A}, \mathbf{v}, g}$ be an admissible order on $\mathbb{I}(DG_k([0, 1]))$ generated by \check{A}, \check{B} , where \check{A} is induced by A and \check{B} is induced by B . Let $\mathbf{w} = (w_1, \dots, w_n) \in [0, 1]^n$ with $w_1 + \dots + w_n = 1$, and let $\check{\mathbf{w}} = (\check{w}_1, \dots, \check{w}_n)$ with $\check{w}_i \doteq w_i$ for all $i = 1, \dots, n$. Then

$$DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq_{\check{A}, \check{B}}^{\check{A}, \mathbf{v}, g}}([\check{x}_1^-, \check{x}_1^+], \dots, [\check{x}_n^-, \check{x}_n^+]) \doteq IVOWA_{\mathbf{w}}^{\preceq_{A,B}}([x_1^-, x_1^+], \dots, [x_n^-, x_n^+])$$

where $\check{x}_i^- \doteq x_i^-$ and $\check{x}_i^+ \doteq x_i^+$ for all $i = 1, \dots, n$. Moreover, for each admissible order \sqsubseteq on $\mathbb{I}(DG_k([0, 1]))$ it holds

$$DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq}([\check{x}_1, \check{x}_1], \dots, [\check{x}_n, \check{x}_n]) \doteq OWA_{\mathbf{w}}(x_1, \dots, x_n)$$

where $\check{x}_i \doteq x_i$ for all $i = 1, \dots, n$.

Proof. 1. First observe that, for all $i, j = 1, \dots, k, j \neq k, [x_i^-, x_i^+] \preceq_{A,B} [x_j^-, x_j^+]$ implies

$$A(x_i^-, x_i^+) < A(x_j^-, x_j^+) \text{ or} \\ (A(x_i^-, x_i^+) = A(x_j^-, x_j^+) \text{ and } B(x_i^-, x_i^+) \leq B(x_j^-, x_j^+)),$$

hence

$$\check{A}(\check{x}_i^-, \check{x}_i^+) \prec_{\mathbb{A}, \mathbf{v}, g} \check{A}(\check{x}_j^-, \check{x}_j^+) \text{ or} \\ (\check{A}(\check{x}_i^-, \check{x}_i^+) = \check{A}(\check{x}_j^-, \check{x}_j^+) \text{ and } \check{B}(\check{x}_i^-, \check{x}_i^+) \preceq_{\mathbb{A}, \mathbf{v}, g} \check{B}(\check{x}_j^-, \check{x}_j^+)),$$

thus $[\check{x}_i^-, \check{x}_i^+] \sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} [\check{x}_j^-, \check{x}_j^+]$.

Therefore, by linearity of $\preceq_{A,B}$ and $\sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}$,

$$[\check{x}_{(n)}^-, \check{x}_{(n)}^+] \sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} \dots \sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} [\check{x}_{(1)}^-, \check{x}_{(1)}^+] \Leftrightarrow [x_{(n)}^-, x_{(n)}^+] \preceq_{A,B} \dots \preceq_{A,B} [x_{(1)}^-, x_{(1)}^+],$$

and finally,

$$DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}}([\check{x}_1^-, \check{x}_1^+], \dots, [\check{x}_n^-, \check{x}_n^+]) = \sum_{i=1}^n \check{w}_i \cdot [\check{x}_{(i)}^-, \check{x}_{(i)}^+] \doteq \\ \doteq \sum_{i=1}^n w_i \cdot [x_{(i)}^-, x_{(i)}^+] = IVOWA_{\mathbf{w}}^{\preceq_{A,B}}([x_1^-, x_1^+], \dots, [x_n^-, x_n^+]).$$

2. Observe that $x_{(n)} \leq \dots \leq x_{(1)}$ implies $\check{x}_{(n)} \leq \dots \leq \check{x}_{(1)}$, hence $[\check{x}_{(n)}^-, \check{x}_{(n)}^+] \preceq_2 \dots \preceq_2 [\check{x}_{(1)}^-, \check{x}_{(1)}^+]$, and consequently $[\check{x}_{(n)}^-, \check{x}_{(n)}^+] \sqsubseteq \dots \sqsubseteq [\check{x}_{(1)}^-, \check{x}_{(1)}^+]$.

Finally

$$DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq}([\check{x}_1^-, \check{x}_1^+], \dots, [\check{x}_n^-, \check{x}_n^+]) = \sum_{i=1}^n \check{w}_i \cdot [\check{x}_{(i)}^-, \check{x}_{(i)}^+] \doteq \\ \doteq \sum_{i=1}^n w_i \cdot [x_{(i)}^-, x_{(i)}^+] = \sum_{i=1}^n w_i \cdot x_{(i)} = OWA_{\mathbf{w}}(x_1, \dots, x_n).$$

□

We show that, for real weights, $DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq}$ is closed on the set of all discrete gradual intervals with strictly increasing left profiles and strictly decreasing right profiles, which means that, according to (6), $DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq}$ is closed on the set V of all fuzzy intervals with finite range.

Theorem 3.11. Let $GV \subseteq \mathbb{I}(DG_k([0, 1]))$ be the set of all discrete gradual intervals with strictly increasing left profiles and strictly decreasing right profiles. If

$$[\check{x}_1^-, \check{x}_1^+], \dots, [\check{x}_n^-, \check{x}_n^+] \in GV,$$

then

$$DGIOWA_{\check{\mathbf{w}}}^{\sqsubseteq}([\check{x}_1^-, \check{x}_1^+], \dots, [\check{x}_n^-, \check{x}_n^+]) \in GV$$

where $\check{\mathbf{w}} = (\check{w}_1, \dots, \check{w}_n)$ with $\check{w}_i \doteq w_i, i = 1, \dots, n$, for some weighting vector of real numbers $\mathbf{w} = (w_1, \dots, w_n)$.

Proof. For all $j = 1, \dots, k - 1, i = 1, \dots, n$ it holds: $\check{w}_i(\alpha_j) = \check{w}_i(\alpha_{j+1}), \check{x}_i^-(\alpha_j) < \check{x}_i^-(\alpha_{j+1}), \check{x}_i^+(\alpha_j) > \check{x}_i^+(\alpha_{j+1})$ and by eq. (13) $DGIOWA_{\check{\mathbf{w}}}^{\square}_{\check{A}, \check{B}}([\check{x}_1^-, \check{x}_1^+], \dots, [\check{x}_n^-, \check{x}_n^+]) \in GV$. □

Example 5. Consider $\check{Y}_1, \check{Y}_2, \check{Y}_3$ and $\check{A}, \check{A}, \check{B}, \mathbf{v}, \mathbf{u}$ from Example 4. Let $\check{\mathbf{w}} = (\check{w}_1, \check{w}_2, \check{w}_3)$ be given by Table 5. Then, by Example 4, $\check{Y}_2 \sqsubseteq_{\check{A}, \check{B}}^{\check{A}, \mathbf{v}, g} \check{Y}_1 \sqsubseteq_{\check{A}, \check{B}}^{\check{A}, \mathbf{v}, g} \check{Y}_3$. The resulting $DGIOWA_{\check{\mathbf{w}}}^{\square}_{\check{A}, \check{B}}(\check{Y}_1, \check{Y}_2, \check{Y}_3)$ is in the third column from the back of Table 5, see Figure 3. However, when we consider weighting vector \mathbf{u} instead of \mathbf{v} , we obtain different result: $DGIOWA_{\check{\mathbf{w}}}^{\square}_{\check{A}, \check{B}}(\check{Y}_1, \check{Y}_2, \check{Y}_3)$ is in the last column of the table, see Figure 3. The reason is that the order of discrete gradual intervals is different: $\check{Y}_1 \sqsubseteq_{\check{A}, \check{B}}^{\check{A}, \mathbf{u}, g} \check{Y}_2 \sqsubseteq_{\check{A}, \check{B}}^{\check{A}, \mathbf{u}, g} \check{Y}_3$ (see Example 4).

α_j	\check{Y}_1	\check{Y}_2	\check{Y}_3	\check{w}_1	\check{w}_2	\check{w}_3	\mathbf{v}	$DGIOWA_{\mathbf{v}}$	\mathbf{u}	$DGIOWA_{\mathbf{u}}$
1	[0.1,0.6]	[0.3,0.4]	[1,1]	0.1	0.3	0.6	0.1	[0.31,0.52]	0.16	[0.25,0.58]
0.9	[0.1,0.6]	[0.3,0.4]	[1,1]	0.2	0.2	0.6	0.1	[0.4,0.56]	0.14	[0.32,0.64]
0.8	[0.2,0.6]	[0.3,0.5]	[0.9,1]	0.3	0.2	0.5	0.1	[0.46,0.67]	0.12	[0.43,0.7]
0.7	[0.2,0.5]	[0.2,0.5]	[0.9,1]	0.4	0.1	0.5	0.1	[0.48,0.7]	0.1	[0.48,0.7]
0.6	[0.3,0.5]	[0.2,0.6]	[0.9,1]	0.4	0.1	0.5	0.1	[0.49,0.75]	0.1	[0.53,0.71]
0.5	[0.3,0.5]	[0.2,0.6]	[0.9,1]	0.4	0.1	0.5	0.1	[0.49,0.75]	0.1	[0.53,0.71]
0.4	[0.3,0.5]	[0.2,0.6]	[0.9,1]	0.3	0.1	0.6	0.1	[0.42,0.71]	0.1	[0.47,0.66]
0.3	[0.4,0.4]	[0.2,0.6]	[1,1]	0.3	0.1	0.6	0.1	[0.46,0.7]	0.08	[0.56,0.6]
0.2	[0.4,0.4]	[0.1,0.7]	[1,1]	0.2	0.1	0.7	0.1	[0.31,0.73]	0.06	[0.49,0.55]
0.1	[0.4,0.4]	[0.1,0.7]	[1,1]	0.1	0.1	0.8	0.1	[0.22,0.7]	0.04	[0.43,0.49]

Tab. 5. Comparison of $DGIOWA_{\check{\mathbf{w}}}^{\square}_{\check{A}, \check{B}}(\check{Y}_1, \check{Y}_2, \check{Y}_3)$ and $DGIOWA_{\check{\mathbf{w}}}^{\square}_{\check{A}, \check{B}}(\check{Y}_1, \check{Y}_2, \check{Y}_3)$ for different weighting vectors \mathbf{v} and \mathbf{u} , see Example 5.

4. APPLICATION TO MULTI-EXPERT DECISION MAKING

In this section a new procedure of solving a multi-expert decision making (MEDM) problem using discrete gradual intervals is presented. As can be seen from given example, an assessment of particular criterion in form of a gradual interval which is not a fuzzy set may arise in a natural way. Thus, the problem cannot be solved by existing methods based on aggregation of fuzzy sets.

A decision maker (DM) considers one of the n alternatives alt_1, \dots, alt_n according to the m criterions c_1, \dots, c_m . We propose the following algorithm to solve this MEDM problem:

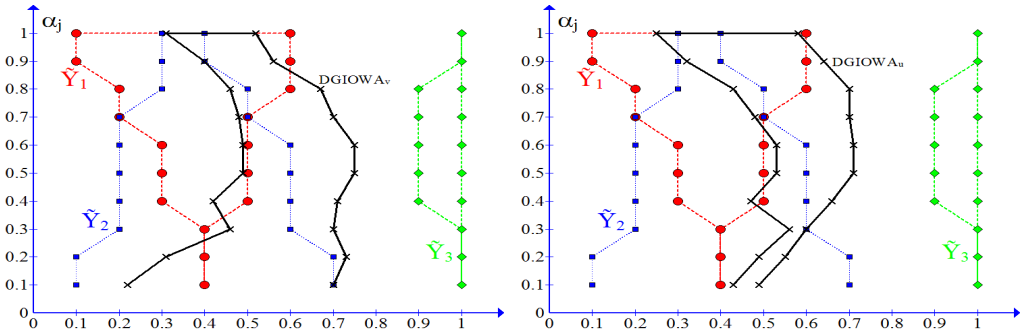


Fig. 3. Discrete gradual intervals $\tilde{Y}_1, \tilde{Y}_2, \tilde{Y}_3$ and $DGIOWA_{\tilde{w}}^{\equiv_{\tilde{A}, \tilde{B}}^{k, v, g}}(\tilde{Y}_1, \tilde{Y}_2, \tilde{Y}_3)$ for the weighting vector \mathbf{v} (on the left) and $DGIOWA_{\tilde{w}}^{\equiv_{\tilde{A}, \tilde{B}}^{k, u, g}}(\tilde{Y}_1, \tilde{Y}_2, \tilde{Y}_3)$ for the weighting vector \mathbf{u} (on the right), see Example 5.

- *Making the information uniform:*¹

DM choose:

- k supervisors s_1, \dots, s_k , possibly with different importance expressed by a weighting vector $\mathbf{v} = (v_1, \dots, v_k) \in [0, 1]^k$ with $v_1 \leq \dots \leq v_k, v_1 + \dots + v_k = 1$,
- a weighting vector $\tilde{\mathbf{w}} = (\tilde{w}_1, \dots, \tilde{w}_m) \in DG_k([0, 1])^m$ with $\tilde{w}_1 + \dots + \tilde{w}_m = \tilde{1}$ which expresses the importance of particular criterions c_1, \dots, c_m , to be more precise \tilde{w}_i expresses the importance of criterion with i th largest assessment.

Each supervisor s_j choose m experts - one expert e_{ji} for each criterion $c_i, j = 1, \dots, k, i = 1, \dots, m$. Then each expert e_{ji} assesses each alternative alt_l according to criterion c_i by interval $[x_{jli}^-, x_{jli}^+] \subseteq [0, 1], j = 1, \dots, k, i = 1, \dots, m, l = 1, \dots, n$. The following discrete gradual intervals are created from this assessment:

$$\tilde{X}_{li} = [\check{x}_{li}^-, \check{x}_{li}^+] \tag{14}$$

where $\check{x}_{li}^-(\alpha_j) = x_{jli}^-, \check{x}_{li}^+(\alpha_j) = x_{jli}^+$, i. e.,

$$\tilde{X}_{li}(\alpha_j) = [x_{jli}^-, x_{jli}^+] \tag{15}$$

for all $j = 1, \dots, k, i = 1, \dots, m, l = 1, \dots, n$, where $\alpha_j = j/k$.

The result of this phase is given by Table 6.

Note that this phase gives more freedom in assessment, because the DM can make assessment without any constraints arising from the need of getting fuzzy sets (functions).

¹We added this 'preparatory' phase before aggregation phase to show one possible way of obtaining experts' preferences in the form of discrete gradual intervals. Of course, this is not the only way, but we will not develop this point here. We leave for the future works the study of this problem.

	c_1	c_2	\dots	c_m
alt_1	\tilde{X}_{11}	\tilde{X}_{12}	\dots	\tilde{X}_{1m}
alt_2	\tilde{X}_{21}	\tilde{X}_{22}	\dots	\tilde{X}_{2m}
\vdots	\vdots	\vdots		\vdots
alt_n	\tilde{X}_{n1}	\tilde{X}_{n2}	\dots	\tilde{X}_{nm}

Tab. 6. The result of the first phase. Each alternative alt_l is assessed according to each criterion c_i by discrete gradual interval \tilde{X}_{li} .

- *Aggregation phase:*
Choose $\mathbb{A}, g, \check{A}, \check{B}$ and calculate DGIOWA operators for each row of Table 6:

$$DGIOWA_{\check{w}}^{\square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}}(\tilde{X}_{11}, \dots, \tilde{X}_{1m}), \dots, DGIOWA_{\check{w}}^{\square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}}(\tilde{X}_{n1}, \dots, \tilde{X}_{nm}).$$

- *Exploitation phase:*
Take as solution the alternative corresponding to the largest DGIOWA according to linear order $\square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}$.

Example 6. A hospital (DM) plans to buy an apparatus. There are three different kinds of apparatus (alt_1, alt_2, alt_3) under consideration. The decision should be made with respect to three properties of apparatus (c_1, c_2, c_3) in such a way that the apparatus would be appropriate for five different diagnosis (d_1, \dots, d_5) with importance expressed by weighting vector $\mathbf{v} = (0.1, 0.15, 0.2, 0.25, 0.3)$. The DM chooses a supervisor (s_1, \dots, s_5) for each diagnosis, each supervisor chooses three experts and each of the experts (e_{ji}) assesses all kinds of apparatus from the point of view of particular criterion (c_i) with respect to particular diagnosis (d_j).

Let redistribution function g be $g(i) = \frac{v_i}{1-v_s} \cdot v_s$, extended aggregation function $\mathbb{A}_{\check{w}}$ be weighted average. Let aggregation function for discrete gradual real numbers \check{A} be induced by $K_{0.5}(a, b) = 0.5a + 0.5b$ and \check{B} be induced by $K_{0.6}(a, b) = 0.6a + 0.4b$ (note that \check{A}, \check{B} satisfy the assumption of Theorem 3.5, see Lemma 3.6 and [3, Theorem 3.3]). Let the experts' assessments be given by Table 7 and $\check{w} = (\check{w}_1, \check{w}_2, \check{w}_3)$ by Table 9.

Then $\tilde{X}_{11}, \dots, \tilde{X}_{33}$ arise by (14)–(15) from the assessments, see Table 8. Recall that this kind of assessment is not possible in other existing approaches [15, 31], because some \tilde{X}_{li} are gradual intervals which are not functions (neither fuzzy sets), so, it is a specific benefit of using gradual intervals and our DGIOWA operator. From (11) we have:

$$\tilde{X}_{11} \square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} \tilde{X}_{13} \square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} \tilde{X}_{12}, \quad \tilde{X}_{21} \square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} \tilde{X}_{22} \square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} \tilde{X}_{23}, \quad \tilde{X}_{33} \square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} \tilde{X}_{32} \square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} \tilde{X}_{31}.$$

The results of $DGIOWA_{\check{w}}^{\square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}}(\tilde{X}_{l1}, \tilde{X}_{l2}, \tilde{X}_{l3})$, for $l = 1, 2, 3$, are in Table 9. Finally, by (11) we have:

$$DGIOWA_{\check{w}}^{\square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}}(\tilde{X}_{11}, \tilde{X}_{12}, \tilde{X}_{13}) \square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} DGIOWA_{\check{w}}^{\square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}}(\tilde{X}_{21}, \tilde{X}_{22}, \tilde{X}_{23}) \square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} \square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g} DGIOWA_{\check{w}}^{\square_{\check{A}, \check{B}}^{\mathbb{A}, \mathbf{v}, g}}(\tilde{X}_{31}, \tilde{X}_{32}, \tilde{X}_{33})$$

hence $alt_3 \succ alt_2 \succ alt_1$, where $alt_i \succ alt_j$ means that alternative alt_i is better than alternative alt_j . Thus the best alternative is alt_3 .

From Table 9 we can see that the weighting vector $\check{w} = (\check{w}_1, \check{w}_2, \check{w}_3)$ puts stress on the largest gradual interval among $\check{X}_{l1}, \check{X}_{l2}, \check{X}_{l3}$ and only leaves slight influence to the smallest one. This has a major impact on the outcome. For instance, if $\check{w}_1 = \check{w}_2 = \check{w}_3 = \frac{1}{3}$, then the results would be reversed: $alt_1 \succ alt_2 \succ alt_3$, hence the best alternative would be alt_1 .

	s_1			s_2			
	e_{11}	e_{12}	e_{13}	e_{21}	e_{22}	e_{23}	
alt_1	[0.8,0.9]	[0.55,0.8]	[0.6,0.75]	[0.6,0.7]	[0.65,0.75]	[0.7,0.7]	...
alt_2	[0.4,0.5]	[0.45,0.85]	[0.7,0.9]	[0.5,0.6]	[0.5,0.8]	[0.9,1]	
alt_3	[0.8,1]	[0.4,0.9]	[0.4,0.5]	[0.9,1]	[0.4,0.9]	[0.2,0.3]	
	s_3			s_4			
	e_{31}	e_{32}	e_{33}	e_{41}	e_{42}	e_{43}	
alt_1	[0.7,0.8]	[0.65,0.75]	[0.6,0.8]	[0.5,0.6]	[0.65,0.7]	[0.6,0.75]	...
alt_2	[0.3,0.4]	[0.55,0.8]	[0.8,0.9]	[0.4,0.6]	[0.6,0.75]	[0.7,0.9]	
alt_3	[0.8,0.9]	[0.4,0.85]	[0.3,0.4]	[1,1]	[0.5,0.85]	[0.2,0.4]	
	s_5						
	e_{51}	e_{52}	e_{53}				
alt_1	[0.6,0.7]	[0.7,0.7]	[0.6,0.8]				
alt_2	[0.4,0.5]	[0.65,0.75]	[0.8,0.9]				
alt_3	[0.9,1]	[0.5,0.8]	[0.3,0.5]				

Tab. 7. The experts' assessments of the alternatives: expert e_{ji} , chosen by supervisor s_j , assesses criterion c_i , $j = 1, \dots, 5$, $i = 1, \dots, 3$ (Example 6).

α_j	\check{X}_{11}	\check{X}_{12}	\check{X}_{13}	\check{X}_{21}	\check{X}_{22}	\check{X}_{23}	...
1	[0.6,0.7]	[0.7,0.7]	[0.6,0.8]	[0.4,0.5]	[0.65,0.75]	[0.8,0.9]	
0.8	[0.5,0.6]	[0.65,0.7]	[0.6,0.75]	[0.4,0.6]	[0.6,0.75]	[0.7,0.9]	
0.6	[0.7,0.8]	[0.65,0.75]	[0.6,0.8]	[0.3,0.4]	[0.55,0.8]	[0.8,0.9]	
0.4	[0.6,0.7]	[0.65,0.75]	[0.7,0.7]	[0.5,0.6]	[0.5,0.8]	[0.9,1]	
0.2	[0.8,0.9]	[0.55,0.8]	[0.6,0.75]	[0.4,0.5]	[0.45,0.85]	[0.7,0.9]	
α_j	\check{X}_{31}	\check{X}_{32}	\check{X}_{33}				
1	[0.9,1]	[0.5,0.8]	[0.3,0.5]				
0.8	[1,1]	[0.5,0.85]	[0.2,0.4]				
0.6	[0.8,0.9]	[0.4,0.85]	[0.3,0.4]				
0.4	[0.9,1]	[0.4,0.9]	[0.2,0.3]				
0.2	[0.8,1]	[0.4,0.9]	[0.4,0.5]				

Tab. 8. The assessments: \check{X}_{li} assess alternative alt_l according to criterion c_i , see Example 6.

α_j	\check{w}_1	\check{w}_2	\check{w}_3	$DGIOWA_1$	$DGIOWA_2$	$DGIOWA_3$
1	0.6	0.3	0.1	[0.66,0.73]	[0.715,0.815]	[0.72,0.89]
0.8	0.6	0.3	0.1	[0.62,0.705]	[0.64,0.825]	[0.77,0.895]
0.6	0.5	0.35	0.15	[0.64,0.775]	[0.638,0.79]	[0.585,0.808]
0.4	0.5	0.35	0.15	[0.66,0.725]	[0.7,0.87]	[0.62,0.86]
0.2	0.45	0.35	0.2	[0.618,0.803]	[0.553,0.803]	[0.58,0.865]

Tab. 9. $DGIOWA_l$ denote $DGIOWA_{\check{w}}^{\square_{A,B}^{A,v,g}}(\tilde{X}_{l1}, \tilde{X}_{l2}, \tilde{X}_{l3})$ for $l = 1 \dots, 3$ (Example 6).

5. CONCLUSION

OWA operator is one of the most widely used aggregation method at present. The existence of a linear order of elements is a crucial point for applying OWA operators to the elements. This paper introduces a class of linear orders for discrete gradual real numbers and, based on the notion of admissible orders, also for discrete gradual intervals. From proposed linear order of discrete gradual intervals one can induce a linear order of fuzzy intervals with finite range. Moreover, the relation between the partial orders of gradual intervals and partial order of fuzzy intervals (membership grades of type-2 fuzzy sets) used in the type-2 fuzzy sets setting is described.

DGIOWA is an extension of IVOWA (OWA operator for intervals [3]) and 'standard' OWA operators. It is also proved that DGIOWA operator satisfies similar properties as 'standard' OWA operator (boundary conditions, monotonicity). Furthermore, the DGIOWA is also applicable to fuzzy intervals and then, for real weights, the aggregated value is fuzzy interval too.

An application of proposed linear orders and OWA operators to multi-expert decision making problem is shown.

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