Research Article

Interval-Valued Complex Fuzzy Geometric Aggregation Operators and Their Application to Decision Making

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This paper investigates the geometric aggregation operators for aggregating the interval-valued complex fuzzy sets (IVCFSs) whose membership grades are a special set of complex numbers. We develop some geometric aggregation operators under the interval-valued complex fuzzy environment, namely, interval-valued complex fuzzy geometric (IVCFG), interval-valued complex fuzzy weighted geometric (IVCFWG), and interval-valued complex fuzzy ordered weighted geometric (IVCFOWG) operators. Then, we investigate the rotational and reflectional invariances of these operators. Further, a decision-making approach based on these operators is presented under the interval-valued complex fuzzy environment and an example is illustrated to demonstrate the efficiency of the proposed approach.

1. Introduction

The aggregation operator is a powerful method for decision making, pattern recognition, and cluster analysis. In the past decades, in both theoretical and applied studies, aggregation operators have attained great advances. Many types of aggregation operators have been proposed under different environments, such as fuzzy environment [1–4], intuitionistic fuzzy environment [5–9], interval-valued intuitionistic fuzzy environment [10–14], Pythagorean fuzzy environment [15–17], neutrosophic fuzzy environment [18–20], and hesitant fuzzy environment [21–25].

In the above fuzzy environments, membership degrees are the subsets of real numbers. As a generalization of traditional fuzzy set [26], Ramot et al. [27] introduced the concept of complex fuzzy set (CFS), which is characterized by a complex-valued membership function. In many practical situations, complex fuzzy sets are useful [28–40]. Moreover, many researchers extended the concept of CFS to interval-valued complex fuzzy set (IVCFS) [41, 42] and complex intuitionistic fuzzy set (CIFS) [43]. Therefore, many researchers discussed how to aggregate CFSs. Ramot et al. [32] introduced the concept of complex fuzzy aggregation. Ma et al. [38] proposed a product-sum aggregation operator under complex fuzzy environment. Bi et al. [39, 40] proposed several aggregation operators under complex fuzzy environment. Garg and Rani [44, 45] investigated the aggregation operators under complex intuitionistic fuzzy environment.

However, we still have the key question: why complex fuzzy aggregation? Moreover, why complex fuzzy sets? As mentioned in [46], from mathematical and practical viewpoints, complex fuzzy sets are natural and useful. But complex fuzzy sets (CFSs) remain a puzzle from the intuitive viewpoint. Fuzzy sets and other extensions give intuitively clear way to describe how humans deal with different types of uncertainty. So before we start to examine the information aggregation issue under interval-valued complex fuzzy environment, we first discuss some phenomena which maybe ignored in real life. When we ask the way, two persons may
give the answers “it is about 1 km away,” and then we think that 1 km is a reasonable result. However, if their answers are not exclusively same about direction, as shown in Figure 1, 0.95 km also is a reasonable result since \( C = (A + B)/2 \).

How does this phenomenon affect human decision making? For example, there are two hospitals \( H_1 \) and \( H_2 \); which one is the nearest hospital? Then, we get data about distance and direction from strangers. It is a very interesting case; two strangers both agree that hospital \( H_1 \) is nearer than hospital \( H_2 \), but after data aggregation, the result that \( H_2 \) is nearer than hospital \( H_1 \) is also reasonable, as shown in Figure 2. The order only relying on the distance is reasonable since we want to go to the nearest hospital. The method based on complex fuzzy aggregation is reasonable since it is a center-based method.

Complex fuzzy aggregation operator can perfectly describe above phenomenon in human decision making since it does not satisfy the property of amplitude monotonicity [40]. Monotonicity is a basic property, which holds in many types of aggregation operators under different fuzzy environments [1–25]. Complex fuzzy aggregation operator as a nonmonotone average is very natural and gives an intuitive way to describe how humans deal with such type of uncertainty.

In this paper, we focus on the aggregation operator under interval-valued complex fuzzy environment. IVCFs also are very appropriate for some applications in real life. For example, when we get lost, we often ask strangers for directions in our daily life. Then, we get some answers, such as “0.5–0.6 km, east” and “0.5–0.7 km, northeast.” These answers can be represented in terms of IVCFs. We present the theory of the weighted geometric aggregation operators among the IVCFs. First we review necessary concepts and some basic properties related to this paper in Section 2. In Section 3, we present an interval-valued complex fuzzy weighted geometric (IVCFWG) operator on CFs. In Section 4, we present an interval-valued complex fuzzy ordered geometric (IVCFOWG) operator. In Section 5, we present a decision-making approach based on the proposed operator under IVCFs environment. Finally, conclusions are given in Section 6.

2. Preliminaries

In this paper, our discussion is based on interval-valued complex fuzzy set theory. Some basic concepts are recalled below, whereas for other concepts, refer to reports from Ref. [27, 28, 41, 42, 47].

Let \( D \) be the set of complex numbers on complex unit disk, i.e., \( D = \{ a \in \mathbb{C} | |a| \leq 1 \} \). Let \( U \) be a fixed universe, and a mapping \( A: U \longrightarrow D \) is called a complex fuzzy set on \( U \).

Let, \( \mathcal{J}^{[0,1]} \) be the set of all closed subintervals of \([0, 1]\), i.e., \( \mathcal{J}^{[0,1]} = \{ [a, b] | 0 \leq a \leq b \leq 1 \} \). Let \( D \) be the boundary set of \( D \) i.e., \( \hat{D} = \{ a \in \mathbb{C} | |a| = 1 \} \). A mapping \( A: U \longrightarrow \mathcal{J}^{[0,1]} \), \( \hat{D} \) is called an IVCF on \( U \). For any \( x \in U \), its membership degree \( \mu_A(x) \) is

\[
\begin{aligned}
\left[ r_A(x), \overline{r_A(x)} \right] \cdot e^{j\rho_x(x)},
\end{aligned}
\]

where \( j = \sqrt{-1}, \mathcal{J}^{[0,1]} \cdot \hat{D} \) is the dot product set of \( \mathcal{J}^{[0,1]} \) and \( \hat{D} \), \( [r_A(x), \overline{r_A(x)}] \in \mathcal{J}^{[0,1]} \) is the interval-valued amplitude part, and \( v_A(x) \in \mathbb{R} \) is the phase part.

For convenience, we only consider the values on \( \mathcal{J}^{[0,1]} \cdot \hat{D} \), which are called interval-valued complex fuzzy values (IVCFVs). Let \( a = [r_a, \overline{r_a}] \cdot e^{j\rho_a} \) be an IVCF, where the interval-valued amplitude part is \( [r_a, \overline{r_a}] \in \mathcal{J}^{[0,1]} \) and the phase part is \( v_a \in \mathbb{R} \). The modulus of \( a \) is the interval-valued amplitude part \( [r_a, \overline{r_a}] \), denoted by \( |a| \). An order of IVCFVs is defined by the interval-valued amplitude part, i.e.,

\[
|a| \leq |b| \text{ if } r_a \leq r_b \text{ and } \overline{r_a} \leq \overline{r_b}. \tag{2}
\]

Let \( a = [r_a, \overline{r_a}] \cdot e^{j\rho_a} \) and \( b = [r_b, \overline{r_b}] \cdot e^{j\rho_b} \) be two IVCFVs and let the parameters be \( \lambda > 0 \) and \( \theta \in \mathbb{R} \); four operators of IVCFVs include multiplication, power, rotation, and reflection which are defined as follows.

(i) Multiplication of two IVCFVs \( a, b \in \mathcal{J}^{[0,1]} \cdot \hat{D} \):

\[
\begin{aligned}
a \otimes b &= [r_a \cdot r_b, \overline{r_a} \cdot \overline{r_b}] \cdot e^{j(\rho_a + \rho_b)}. \tag{3}
\end{aligned}
\]

(ii) Power of an IVCFV \( a \in \mathcal{J}^{[0,1]} \cdot \hat{D} \):

\[
\begin{aligned}
a^\lambda &= \left[ \frac{r_a}{r_a + \overline{r_a}} \right] \cdot e^{j\lambda\rho_a}. \tag{4}
\end{aligned}
\]

(iii) Rotation of an IVCFV \( a \in \mathcal{J}^{[0,1]} \cdot \hat{D} \):

\[
\begin{aligned}
\text{Rot}_\theta(a) &= [r_a, \overline{r_a}] \cdot e^{j(\rho_a + \theta)}. \tag{5}
\end{aligned}
\]

(iv) Reflection of an IVCFV \( a \in \mathcal{J}^{[0,1]} \cdot \hat{D} \):

\[
\begin{aligned}
\text{Ref}(a) &= [r_a, \overline{r_a}] \cdot e^{-j\rho_a}. \tag{6}
\end{aligned}
\]

When \( a, b \in \hat{D} \), \( \odot \) is a complex intersection defined by Ramot et al. [27]. When \( a, b \in [0, 1] \), \( \odot \) is a t-norm [48].

**Theorem 1.** Suppose that \( a, b, c \) are three IVCFVs, and the parameters are \( \lambda_1 > 0, \lambda_2 > 0, \text{ and } \theta_1, \theta_2 \in \mathbb{R} \). Then, we have

\[
\begin{aligned}
(1) & a \odot b = b \odot a \\
(2) & (a \odot b) \odot c = a \odot (b \odot c) \\
(3) & (a \odot b)^\lambda = a^{\lambda_1} \odot b^{\lambda_2}.
\end{aligned}
\]
Proof. Let \( a = [r_a, \overline{r_a}] \cdot e^{j\gamma_a}, b = [r_b, \overline{r_b}] \cdot e^{j\gamma_b}, \) and \( c = [r_c, \overline{r_c}] \cdot e^{j\gamma_c}. \)

1. By the definition of multiplication of two IVCFVs, we have

\[
(a \otimes b) = \left[ r_a \cdot r_b, \overline{r_a} \cdot \overline{r_b} \right] \cdot e^{j(\gamma_a + \gamma_b)}
\]

(7) Ref. \(( \text{Ref} (a)) = \text{Ref} \left( [r_a, \overline{r_a}] \cdot e^{j\gamma_a} \right) = [r_a, \overline{r_a}] \cdot e^{j\gamma_a} \]

(8) Ref. \(( a \otimes b) = \text{Ref} (a) \otimes \text{Ref} (b) \)

2. And

\[
(a \otimes b) \otimes c = \left[ r_a \cdot r_b \cdot r_c, \overline{r_a} \cdot \overline{r_b} \cdot \overline{r_c} \right] \cdot e^{j(\gamma_a + \gamma_b + \gamma_c)}
\]

(8) For any two IVCFVs \( a, b \) and real value \( \lambda > 0 \), we have

\[
\text{Ref} (a)^{\lambda} = \left[ [r_a, \overline{r_a}] \cdot e^{-j\gamma_a} \right]^{\lambda} = [r_a, \overline{r_a}] \cdot e^{-j\lambda\gamma_a} \]

(13) For any two IVCFVs \( a, b \), we have

\[
\text{Ref} (a) \otimes \text{Ref} (b) = [r_a, \overline{r_a}] \cdot e^{-j\gamma_a} \otimes [r_b, \overline{r_b}] \cdot e^{-j\gamma_b} = [r_a \cdot r_b, \overline{r_a} \cdot \overline{r_b}] \cdot e^{-j(\gamma_a + \gamma_b)}
\]

Moreover, the multiplication and power operators have the following properties.

\[\square\]
(2) (Amplitude boundedness) \(|r_a \land r_b, \overline{r_a \land r_b}| \leq |a \otimes b| \leq 1\).

(3) If \(\lambda_1 < \lambda_2\), then \(|a^{\lambda_1}| \leq |a^{\lambda_2}|\).

(4) If \(|a| \leq |b|\), then \(|a^{\lambda_1}| \leq |b^{\lambda_1}|\).

\(|\text{Ref}(a)| \leq |\text{Ref}(b)|\), \(|\text{Rot}_\theta(a)| \leq |\text{Rot}_\theta(b)|\).

**Proof.** Since the amplitude terms of IVCFVs are interval values, we can obtain the above results from the properties of multiplication operator of interval values.

Rotational invariance and reflectional invariance are two important geometric properties of complex fuzzy operators [28, 29, 39]. They show that a complex fuzzy operator is invariant under a rotation and a reflection, respectively. We define the following two similar properties for interval-valued complex fuzzy operators as follows.

(i) An interval-valued complex fuzzy operator \(v: \mathcal{J}^{[0,1]} \times \mathcal{J}^{[0,1]} \rightarrow \mathcal{J}^{[0,1]} \). \(D\) is rotationally invariant if and only if, for any \(\theta\),

\[v(\text{Rot}_\theta(a_1), \ldots, \text{Rot}_\theta(a_n)) = \text{Rot}_\theta(v(a_1, \ldots, a_n)). \quad (15)\]

(ii) An interval-valued complex fuzzy operator \(v: \mathcal{J}^{[0,1]} \times \mathcal{J}^{[0,1]} \rightarrow \mathcal{J}^{[0,1]} \). \(D\) is reflectionally invariant if and only if

\[v(\text{Ref}(a_1), \ldots, \text{Ref}(a_n)) = \text{Ref}(v(a_1, \ldots, a_n)). \quad (16)\]

Then, for the multiplication operator of IVCFVs, we have the following results.

**Theorem 3.** The multiplication operator of IVCFVs is reflectionally invariant.

**Proof.** Trivial from Theorem 1 (16).

**Theorem 4.** The multiplication operator of IVCFVs is not rotationally invariant.

**Proof.** For any two IVCFVs \(a = [r_a, r_a, \epsilon^{\gamma_a}], b = [r_b, r_b, \epsilon^{\gamma_b}]\) and real value \(\theta\), we have

\[
\text{Rot}_\theta(a) \otimes \text{Rot}_\theta(b) = [r_a, r_a, \epsilon^{\gamma_a + \theta}] \otimes [r_b, r_b, \epsilon^{\gamma_b + \theta}]

= [r_a \cdot r_b, r_a \cdot r_b, \epsilon^{\gamma_a + \gamma_b + \theta}],

\text{Rot}_\theta(a \otimes b) = [r_a \cdot r_b, r_a \cdot r_b, \epsilon^{\gamma_a + \gamma_b + \theta}].

Since \(\gamma_a + \gamma_b + 2\theta \neq \gamma_a + \gamma_b + \theta\), the multiplication operator is not rotationally invariant.

**3. Interval-Valued Complex Fuzzy Weighted Geometric Operators**

In this section, we introduce the weighted geometric operators in an interval-valued complex fuzzy environment and discuss their fundamental characteristics.

**Definition 1.** Let \(a_i (i = 1, 2, \ldots, n)\) be a collection of IVCFVs; an interval-valued complex fuzzy weighted geometric (IVCFWG) operator is defined as

\[\text{IVCFWG}(a_1, a_2, \ldots, a_n) = \otimes_{i=1}^{n} w_i a_i, \quad (18)\]

where \(w_i \in [0, 1]\) for all \(i\) and \(\sum_{i=1}^{n} w_i = 1\).

When \(w_i = 1/n \quad (i = 1, 2, \ldots, n)\), then the IVCFWG operator is denoted by interval-valued complex fuzzy geometric (IVCFG) operator, i.e.,

\[\text{IVCFWG}(a_1, a_2, \ldots, a_n) = \otimes_{i=1}^{n} a_i^{1/n}. \quad (19)\]

When \(a_i \in \mathcal{J}^{[0,1]} \quad (i = 1, 2, \ldots, n)\), the IVCFWG operator can reduce to a traditional interval-valued fuzzy weighted geometric operator of values on unit interval \(\mathcal{J}^{[0,1]}\).

When \(a_i \in [0, 1] \quad (i = 1, 2, \ldots, n)\), the IVCFWG operator can reduce to a traditional weighted geometric operator [49] of real numbers on unit interval \([0, 1]\).

When \(a_i \in [0, 1]\) and \(w_i = 1/n \quad (i = 1, 2, \ldots, n)\), the CFWP operator can reduce to a traditional geometric mean operator [50] of real numbers on unit interval \([0, 1]\).

**Theorem 5.** Let \(a_i (i = 1, 2, \ldots, n)\) be a collection of IVCFVs; then, the aggregated value \(\text{IVCFWG}(a_1, a_2, \ldots, a_n)\) is also an IVCFV and

\[\text{IVCFWG}(a_1, a_2, \ldots, a_n) = \prod_{i=1}^{n} \frac{r_{a_i}}{r_{a_1}} \cdot e^{i \theta} \cdot \left(\sum_{i=1}^{n} w_i \gamma_i\right), \quad (20)\]

where \(w_i \in [0, 1]\) for all \(i\) and \(\sum_{i=1}^{n} w_i = 1\).

**Proof.** It is easy to check that the aggregated value is also an IVCFV. Now, we prove equation (20) by mathematical induction method.

For \(n = 2\), we have

\[a_1^{w_1} = [r_{a_1}^{w_1}, r_{a_1}^{w_1}, \epsilon^{\gamma_{a_1} + \theta}], \quad a_2^{w_2} = [r_{a_2}^{w_2}, r_{a_2}^{w_2}, \epsilon^{\gamma_{a_2} + \theta}];\]

then,

\[\text{IVCFWG}(a_1, a_2) = a_1^{w_1} \otimes a_2^{w_2} = [r_{a_1}^{w_1}, r_{a_1}^{w_1}, \epsilon^{\gamma_{a_1} + w_1 \gamma_{a_2}}] \cdot e^{i \theta} \cdot \left(\sum_{i=1}^{2} w_i \gamma_i\right), \quad (21)\]

If equation (20) holds for \(n = k\), i.e.,

\[\text{IVCFWG}(a_1, a_2, \ldots, a_k) = \prod_{i=1}^{k} \frac{r_{a_i}}{r_{a_1}} \cdot e^{i \theta} \cdot \left(\sum_{i=1}^{k} w_i \gamma_i\right), \quad (22)\]

then for \(n = k + 1\),
IVCFWG\((a_1, a_2, \ldots, a_{k+1})\)

\[
= \text{IVCFWG}(a_1, a_2, \ldots, a_k) \otimes a_{k+1}^{w_{k+1}}
\]

\[
= \left[ \prod_{i=1}^{k} \frac{r_{a_i}^{w_i}}{r_{a_i}^{w_i}} \right] \cdot e^{(\sum_{i=1}^{k} w_{a_i})} \otimes a_{k+1}^{w_{k+1}}
\]

\[
= \left[ \left( \prod_{i=1}^{k} \frac{r_{a_i}^{w_i}}{r_{a_i}^{w_i}} \right) \cdot r_{a_{k+1}}^{w_{k+1}} \cdot \left( \prod_{i=1}^{k} \frac{r_{a_i}^{w_i}}{r_{a_i}^{w_i}} \right) \right] \cdot e^{(\sum_{i=1}^{k} w_{a_i})} \cdot e^{(\sum_{i=1}^{k} w_{a_i})}
\]

\[
= \left[ \prod_{i=1}^{k+1} \frac{r_{a_i}^{w_i}}{r_{a_i}^{w_i}} \right] \cdot e^{(\sum_{i=1}^{k+1} w_{a_i})}
\]

Therefore, equation (20) holds for all \(n\).

The working of the proposed IVCFWG operator is explained with a numerical example as follows.

**Example 1.** Let \(a_1 = [0.6, 0.7] \cdot e^{0.1\pi}, a_2 = [0.3, 0.5] \cdot e^{0.2\pi}, a_3 = [0.6, 0.8] \cdot e^{0.7\pi},\) and \(a_4 = [0.4, 0.7] \cdot e^{0.4\pi}\) be four IVCFVs and \(w = (0.1, 0.2, 0.3, 0.4)^T\) be the associated weight vector. Then, by using equation (20),

\[
\prod_{i=1}^{k} \frac{r_{a_i}^{w_i}}{r_{a_i}^{w_i}} = 0.6^{0.1} \cdot 0.3^{0.2} \cdot 0.6^{0.3} \cdot 0.4^{0.4} = 0.4441,
\]

\[
\prod_{i=1}^{k} \frac{r_{a_i}^{w_i}}{r_{a_i}^{w_i}} = 0.7^{0.1} \cdot 0.5^{0.2} \cdot 0.8^{0.3} \cdot 0.7^{0.4} = 0.6812,
\]

\[
\sum_{i=1}^{k} w_{a_i} = 0.1 \cdot 0.1\pi + 0.2 \cdot 0.2\pi + 0.3 \cdot 0.3\pi + 0.4 \cdot 0.4\pi = 0.42\pi,
\]

we obtain IVCFWG\((a_1, a_2, a_3, a_4) = [0.4441, 0.6812] \cdot e^{0.42\pi}\).

Based on Theorem 5, the proposed IVCFWG operator satisfies the following properties.

**Theorem 6.** Let \(a_i (i = 1, 2, \ldots, n)\) and \(b_i (i = 1, 2, \ldots, n)\) be two collections of IVCFVs, the weights be \(w_i \in [0, 1] (i = 1, 2, \ldots, n),\) and \(\sum_{i=1}^{n} w_i = 1.\) Then, we have the following properties:

1. (Idempotency) If \(a_i = a (i = 1, 2, \ldots, n),\) then
   \[
   \text{IVCFWG}(a_1, a_2, \ldots, a_n) = a.
   \]

2. (Amplitude monotonicity) If \(|a_i| \leq |b_i| (i = 1, 2, \ldots, n),\) then
   \[
   \text{IVCFWG}(a_1, a_2, \ldots, a_n) \leq \text{IVCFWG}(b_1, b_2, \ldots, b_n).
   \]

3. (Amplitude boundedness)

   \[
   r_1 \leq \text{IVCFWG}(a_1, a_2, \ldots, a_n) \leq r_2,
   \]
   \[
   \text{where } r_1, r_2 \text{ are two interval values:}
   \]
   \[
   r_1 = \left[ \min_{i} r_{a_i}, \min_{i} r_{a_i} \right],
   \]
   \[
   r_2 = \left[ \max_{i} r_{a_i}, \max_{i} r_{a_i} \right].
   \]

**Proof.**

1. Let \(a_i = a = [r_{a_1}, r_{a_1}] (i = 1, 2, \ldots, n);\) then,
   \[
   \text{IVCFWG}(a_1, a_2, \ldots, a_n) = \left[ \prod_{i=1}^{n} r_{a_i}^{w_i}, \prod_{i=1}^{n} r_{a_i}^{w_i} \right] \cdot e^{(\sum_{i=1}^{n} w_{a_i})} \cdot e^{(\sum_{i=1}^{n} w_{a_i})} = \left[ r_{a_1}, r_{a_1} \right] \cdot e^{(\sum_{i=1}^{n} w_{a_i})} = a.
   \]

2. The property of the amplitude monotonicity is trivial from Theorem 2 (4).

3. The property of the amplitude boundedness is easily obtained from idempotency and amplitude monotonicity.

Note that idempotency is concerned with both the phase part and amplitude part of IVCFVs. Amplitude boundedness and amplitude monotonicity are only concerned with the amplitude part of IVCFVs.

**Theorem 7.** The IVCFWG operator is reflectionally invariant.

**Proof.** For any collection of IVCFVs \(a_i (i = 1, 2, \ldots, n),\) from equation (20), we have

IVCFWG\((\text{Ref} \,(a_1), \text{Ref} \,(a_2), \ldots, \text{Ref} \,(a_n))\)

\[
= \left[ \prod_{i=1}^{n} \left( \frac{r_{a_i}^{w_i}}{r_{a_i}^{w_i}} \right), \prod_{i=1}^{n} \left( \frac{r_{a_i}^{w_i}}{r_{a_i}^{w_i}} \right) \right] \cdot e^{(\sum_{i=1}^{n} w_{a_i})} \cdot e^{(\sum_{i=1}^{n} w_{a_i})},
\]

\[
\text{Ref} \,(\text{IVCFWG}(a_1, a_2, \ldots, a_n))
\]

\[
= \text{Ref} \,(\otimes_{i=1}^{n} a_{i}^{w_i})
\]

\[
= \left[ \prod_{i=1}^{n} \left( \frac{r_{a_i}^{w_i}}{r_{a_i}^{w_i}} \right), \prod_{i=1}^{n} \left( \frac{r_{a_i}^{w_i}}{r_{a_i}^{w_i}} \right) \right] \cdot e^{(\sum_{i=1}^{n} w_{a_i})} \cdot e^{(\sum_{i=1}^{n} w_{a_i})}.\]
Then, \( \text{IVCFWG} \left( \text{Ref} \left( a_1 \right), \text{Ref} \left( a_2 \right), \ldots, \text{Ref} \left( a_n \right) \right) = \text{Ref} \left( \bigotimes_{i=1}^{n} w_i a_i \right) \) is reflectionally invariant. \( \square \)

**Theorem 8.** The IVCFWG operator is rotationally invariant.

**Proof.** For any collection of IVCFVs \( a_i (i = 1, 2, \ldots, n) \) and any \( \theta \in \mathbb{R} \), we have

\[
\begin{align*}
\text{IVCFWG} \left( \text{Rot}_\theta \left( a_1 \right), \text{Rot}_\theta \left( a_2 \right), \ldots, \text{Rot}_\theta \left( a_n \right) \right) \\
= \left[ \prod_{i=1}^{n} \left( r_{a_i} \right)^{w_i} \right] \cdot \prod_{i=1}^{n} \left( r_{a_i}^{w_i} \right)^{w_i} \cdot e^{\left( \sum_{i=1}^{n} w_i \left( a_i, \sigma(i) \right) \right)} \\
= \left[ \prod_{i=1}^{n} \left( r_{a_i} \right)^{w_i} \right] \cdot \prod_{i=1}^{n} \left( r_{a_i}^{w_i} \right)^{w_i} \cdot e^{\left( \sum_{i=1}^{n} w_i \sigma(i) \right) \cdot \left( \sum_{i=1}^{n} \theta \right)} \\
= \text{Rot}_\theta \left( \text{IVCFWG} \left( a_1, a_2, \ldots, a_n \right) \right) \\
\end{align*}
\]

Then, the IVCFWG operator is rotationally invariant. \( \square \)

### 4. Interval-Valued Complex Fuzzy Ordered Weighted Geometric Operators

Based on the order of IVCFVs defined by equation (2) and the ordered weighted averaging (OWA) operator introduced by Yager [1], we define an interval-valued complex fuzzy ordered weighted geometric (IVCFOWG) operator as follows.

**Definition 2.** Let \( a_i (i = 1, 2, \ldots, n) \) be a collection of CFVs; then, an IVCFOWG operator is defined as

\[
\text{IVCFOWG} \left( a_1, a_2, \ldots, a_n \right) = \bigotimes_{i=1}^{n} a_i \sigma(i),
\]

where \( w_i \in [0, 1] \) \((i = 1, 2, \ldots, n)\) and \( \sum_{i=1}^{n} w_i = 1 \), \( \sigma(1), \sigma(2), \ldots, \sigma(n) \) is a permutation of \((1, 2, \ldots, n)\) such that \( a_{\sigma(i-1)} \geq a_{\sigma(i)} \) for all \( i \).

**Theorem 9.** Let \( a_i (i = 1, 2, \ldots, n) \) be a collection of IVCFVs; then, the aggregated value \( \text{IVCFOWG} \left( a_1, a_2, \ldots, a_n \right) \) is also an IVCFV and

\[
\begin{align*}
\text{IVCFOWG} \left( a_1, a_2, \ldots, a_n \right) \\
= \left[ \prod_{i=1}^{n} \left( r_{a_{\sigma(i)}} \right)^{w_i} \right] \cdot \prod_{i=1}^{n} \left( r_{a_{\sigma(i)}} \right)^{w_i} \cdot e^{\left( \sum_{i=1}^{n} w_i a_{\sigma(i)} \right)},
\end{align*}
\]

where \( w_i \in [0, 1] \) \((i = 1, 2, \ldots, n)\) and \( \sum_{i=1}^{n} w_i = 1 \), \( \sigma(1), \sigma(2), \ldots, \sigma(n) \) is a permutation of \((1, 2, \ldots, n)\) such that \( a_{\sigma(i-1)} \geq a_{\sigma(i)} \) for all \( i \).

**Proof.** The proof is similar to Theorem 6. \( \square \)

**Theorem 10.** Let \( a_i (i = 1, 2, \ldots, n) \) be a collection of CFVs, IVCFOWG weights be \( w_i \in [0, 1] \) \((i = 1, 2, \ldots, n)\), and \( \sum_{i=1}^{n} w_i = 1 \). Then, we have the following properties.

(1) (Idempotency) If \( a_i = a (i = 1, 2, \ldots, n) \), then

\[
\text{IVCFOWG} \left( a_1, a_2, \ldots, a_n \right) = a.
\]

(2) (Amplitude monotonicity) If \( |a_i| \leq |b_i| \) \((i = 1, 2, \ldots, n)\), then

\[
|\text{IVCFOWG} \left( a_1, a_2, \ldots, a_n \right)| \leq |\text{IVCFOWG} \left( b_1, b_2, \ldots, b_n \right)|,
\]

(3) (Amplitude boundedness)

\[
|r_1| \leq |\text{IVCFOWG} \left( a_1, a_2, \ldots, a_n \right)| \leq r_2,
\]

where \( r_1, r_2 \) are two interval values:

\[
\begin{align*}
r_1 &= \left[ \min_{i} r_{a_{\sigma(i)}}, \min_{i} r_{a_{\sigma(i)}} \right], \\
r_2 &= \left[ \max_{i} r_{a_{\sigma(i)}}, \max_{i} r_{a_{\sigma(i)}} \right].
\end{align*}
\]

**Proof.** The proof is similar to Theorem 6. \( \square \)

**Theorem 11.** The IVCFWG operator is reflectionally invariant and rotationally invariant.

**Proof.** The proof is similar to Theorems 7 and 8. \( \square \)
Note that the operator $\otimes$ does not have the property of rotational invariance, but the IVCFWG and IVCFOWG operators defined based on $\oplus$ operator have the property of rotational invariance.

Here, we investigate the aggregation operators based on special class of the IVCFVs, which belong to subsets of the upper-right quadrant of the complex unit disk. Let $D_1 = \{ e^{i\phi} | \phi \in [0, \pi/2] \}$; we consider the aggregation operator on $\mathcal{F}^{[0,1]} \cdot D_1$.

Let us consider the closeness of IVCFVs on $\mathcal{F}^{[0,1]} \cdot D_1$ under the IVCFWG and IVCFOWG operations. For the IVCFWG operator, we have the following result.

**Theorem 12.** Let $z_1, z_2, \ldots, z_n \in \mathcal{F}^{[0,1]} \cdot D_1$. Then, the aggregated value

$$\text{IVCFWG}(z_1, z_2, \ldots, z_n) \in \mathcal{F}^{[0,1]} \cdot D_1.$$ (39)

**Proof.** Denoting IVCFWG $(z_1, z_2, \ldots, z_n) = (v_1, \ldots, v_n)$, from Theorem 5, $(v_1, \ldots, v_n)$ is an interval value. Since $v = \sum_{i=1}^{n} w_i \cdot v_i$ is a weighted arithmetic aggregation operator of real numbers on $[0, \pi/2]$, then we have $v \in [0, \pi/2]$. Thus, IVCFWG $(z_1, z_2, \ldots, z_n) \in \mathcal{F}^{[0,1]} \cdot D_1$.

Similar to the above theorem, we have the following.

**Theorem 13.** Let $z_1, z_2, \ldots, z_n \in \mathcal{F}^{[0,1]} \cdot D_1$. Then, the aggregated value

$$\text{IVCFOWG}(z_1, z_2, \ldots, z_n) \in \mathcal{F}^{[0,1]} \cdot D_1.$$ (40)

The above theorems show us that the IVCFWG and the IVCFOWG operators are close under values on $\mathcal{F}^{[0,1]} \cdot D_1$. Consider other quadrants of the complex unit disk. Let

$$D_k = \left\{ z = e^{i\theta} | \theta \in \left[ \frac{(k-1)\pi}{2}, \frac{k\pi}{2} \right] \right\},$$ (41)

for $k = 1$ to $4$.

Now, we discuss the closeness of the IVCFWG and the IVCFOWG operators on values of other quadrants of the complex unit disk. Plainly, we have the following.

**Theorem 14.** For any $k \in \{1, 2, 3, 4\}$, if $z_1, z_2, \ldots, z_n \in \mathcal{F}^{[0,1]} \cdot D_k$, then we have

$$\text{IVCFWG}(z_1, z_2, \ldots, z_n) \in \mathcal{F}^{[0,1]} \cdot D_k,$$

$$\text{IVCFOWG}(z_1, z_2, \ldots, z_n) \in \mathcal{F}^{[0,1]} \cdot D_k.$$ (42)

**Proof.** Similar to Theorem 12. □

### 5. An Approach to Decision Making with the IVCFWG Operator

In this section, we present an approach using the IVCFWG operator to a decision making with interval-valued complex fuzzy information.

We consider a target selection application of CFSs. Assume that our position is fixed, and then we can measure the distance and angle of the possible alternatives by using a position sensor and an angular sensor (or other attributes from expert opinions). Assume that we get a measurement $(d \pm \epsilon, q)$. Here, we use an interval value $[d, d']$ to represent $d \pm \epsilon$ by setting $d = d - \epsilon$ and $d' = d + \epsilon$. To improve the target location accuracy, we repeatedly measure the alternatives. Then, the target is selected in the following approach according to aggregation theory.

Let $x_i (i = 1, 2, \ldots, n)$ be the possible alternatives and $e_k (k = 1, 2, \ldots, m)$ be the experts. Then, the decision maker provides a decision matrix $A = (a_{ik})_{m \times n}$, where $a_{ik}$ is an interval-valued complex fuzzy value given by the expert $e_k$ for alternative $x_i$. Further, assume that $w = (w_1, w_2, \ldots, w_n)^T$ is the weight vector of the different experts such that $w_i \geq 0$ $(i = 1, 2, \ldots, n)$ and $\sum_{i=1}^{n} w_i = 1$. The process can be summarized as follows:

**Step 1.** Transform the decision matrix $A$ into the normalized interval-valued complex fuzzy decision matrix $C = (c_{ik})_{m \times n}$ using $c_{ik} = a_{ik} / d$ where $d = \max_{i=1}^{n} |a_{ik}|$.

**Step 2.** Aggregate all the interval-valued complex fuzzy values $a_{ik}$ $(k = 1, 2, \ldots, m)$ and get the overall interval-valued complex fuzzy value $b_i$ corresponding to the alternative $x_i$ by the IVCFWG operator, $b_i = \text{IVCFWG}(a_{i1}, a_{i2}, \ldots, a_{in})$.

**Step 3.** Rank the overall IVCFVs $b_i (i = 1, 2, \ldots, n)$ using (2).

Next, we give an example to illustrate the above approach.

**Example 3.** Suppose we want to go to the nearest bank, then we often use GPS navigation system or ask strangers for directions.

Suppose that there are three alternatives $x_i (i = 1, 2, 3)$ and three experts (GPS or stranger) $e_i (i = 1, 2, 3)$ with the weight vector $(0.4, 0.3, 0.3)^T$. The three experts evaluate the three alternatives under the IVCFS environment, and their corresponding rating values are summarized in the decision matrix $C = (c_{ik})_{3 \times 3}$ represented in Table 1, where $c_{ik} = r_{ik} \cdot e^{j\epsilon_{ik}}$ is an interval-valued complex fuzzy value, $[r_{ik}, \epsilon_{ik}]$ represents the distance between $r_{ik}$ km and $\epsilon_{ik}$ km, and $\epsilon_{ik}$ represents the direction for the alternative $x_i$.

**Step 1.** The values $c_{ik}$ do not need normalization.

**Step 2.** Aggregate the interval-valued complex fuzzy values $b_i$ of the alternatives $x_i$ by using the IVCFWG operator (see equation (20)):

$$\prod_{i=1}^{n} r_{i_k} = 0.19^{0.4} \cdot 0.11^{0.3} \cdot 0.16^{0.3} = 0.1532,$$

$$\prod_{i=1}^{n} r_{i_k} = 0.25^{0.4} \cdot 0.23^{0.3} \cdot 0.23^{0.3} = 0.2378,$$

(43)

$$\sum_{i=1}^{n} w_i \epsilon_{i_k} = (0.4 \cdot 0.11 + 0.3 \cdot 0.13 + 0.3 \cdot 0.11) \cdot \pi = 0.116\pi,$$
operators. valued complex fuzzy aggregation operators are more environment considered in the present paper. In [5, 6] are unable to solve the problems under the IVCFSs the DM problem under these existing environments by the proposed aggregation operators. The proposed CFS, IVFS, and FS. IVCFScan handletwo-dimensional information in a single (CFS), interval-valued fuzzy sets (IVFS), and fuzzy sets (FS).

| Ki | Ki | Ki |
|----|----|----|
| X1 | [0.19, 0.25] | e^{0.11i} | e^{0.13i} | e^{0.11i} |
| X2 | [0.12, 0.23] | e^{0.52i} | e^{0.54i} | e^{0.52i} |
| X3 | [0.10, 0.16] | e^{1.11i} | e^{1.12i} | e^{1.14i} |

and we obtain \( b_1 = \text{IVCFWG}(a_{11}, a_{12}, a_{13}) = [0.1532, 0.2378] \cdot e^{0.116i} \).

\[
\prod_{i=1}^{3} r_{a\alpha}^{\omega_1} = 0.12^{0.4} \cdot 0.18^{0.3} \cdot 0.18^{0.3} = 0.1531,
\]

\[
\prod_{i=1}^{3} r_{a\alpha}^{\omega_2} = 0.23^{0.4} \cdot 0.24^{0.3} \cdot 0.2^{0.3} = 0.2234, \quad (44)
\]

\[
\sum_{i=1}^{3} w_i r_{a\alpha} = (0.4 \cdot 0.52 + 0.3 \cdot 0.54 + 0.3 \cdot 0.52) \cdot \pi = 0.526\pi,
\]

and we obtain \( b_2 = \text{IVCFWG}(a_{11}, a_{12}, a_{13}) = [0.1531, 0.2234] \cdot e^{0.126i}. \)

\[
\prod_{i=1}^{3} r_{a\alpha}^{\omega_1} = 0.1^{0.4} \cdot 0.19^{0.3} \cdot 0.17^{0.3} = 0.1422,
\]

\[
\prod_{i=1}^{3} r_{a\alpha}^{\omega_2} = 0.16^{0.4} \cdot 0.23^{0.3} \cdot 0.19^{0.3} = 0.1878, \quad (45)
\]

\[
\sum_{i=1}^{3} w_i r_{a\alpha} = (0.4 \cdot 1.11 + 0.3 \cdot 1.12 + 0.3 \cdot 1.14) \cdot \pi = 1.122\pi,
\]

and we obtain \( b_3 = \text{IVCFWG}(a_{11}, a_{12}, a_{13}) = [0.1422, 0.1878] \cdot e^{0.122i}. \)

Step 3. Rank the interval-valued complex fuzzy values \( b_i \) \((i = 1, 2, 3): [b_1] > [b_2] > [b_3]\), which shows that the alternative \( x_3 \) is the optimal choice, i.e., the nearest bank.

An IVCFS is a generalization of complex fuzzy sets (CFS), interval-valued fuzzy sets (IVFS), and fuzzy sets (FS). IVCFS can handle two-dimensional information in a single set. Then, IVCFS contains much more information than CFS, IVFS, and FS.

It is revealed from the present study that the aggregation operators under IVFSs and FSs [5, 6] are the special cases of the proposed aggregation operators. Thus, the proposed aggregation operators can be equivalently utilized to solve the DM problem under these existing environments by setting phase term to be zero while the existing operators [5, 6] are unable to solve the problems under the IVCFSs environment considered in the present paper. Thus, interval-valued complex fuzzy aggregation operators are more general than some (interval-valued) fuzzy aggregation operators.

6. Conclusion

As mentioned in [50], the basic feature of aggregation operators is their monotonicity property. However, aggregation operators under the complex fuzzy environment [40] are not monotone. Interestingly, such aggregation operators can explain some phenomena in our real life.

In this paper, we discussed the geometric aggregation operators under the interval-valued complex fuzzy environment. Two interval-valued complex fuzzy aggregation operators, the IVCFWG and the IVCFOWG operators, are developed and their properties are studied. It is also interesting to note that both the IVCFWG and the IVCFOWG operators are rotationally invariant and reflectionally invariant. Further, based on the proposed operators, we presented a decision-making approach under interval-valued complex fuzzy information. An illustrative example is given for illustrating the proposed approach.

It was observed that the aggregation operator based on nonadditive integrals (Choquet [51] and Sugeno integrals [52]) is one of the hot topics in this field [49]. As future work, we can consider the complex fuzzy aggregation operators based on complex integrals.

Data Availability

The data used to support the findings of this study are included in the article.

Conflicts of Interest

The authors declare that they have no conflicts of interest.

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