

## New aggregation operators of intuitionistic trapezoidal fuzzy multi-numbers based on Archimedean norms

İRFAN DELI, HÜMEYRA KARADÖL

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**ABSTRACT.** In trapezoidal intuitionistic fuzzy multi numbers, an aggregation operator performs the task of aggregate the described information to generate a ranking of alternatives which are two critical tasks: a flexible and superior tool for the first task and an effective tool for the second task is aggregation operator. Therefore, in the study, we give new aggregation operators, based on Archimedean T-norm and T-conorm operations which is called Archimedean norms operator for aggregating trapezoidal intuitionistic fuzzy multi-information. Then, we develop a multi criteria decision making method based on the given operators. Finally, the feasibility and effectiveness of the method is demonstrated via a numerical example.

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**Keywords:** Fuzzy sets, Fuzzy number, Intuitionistic trapezoidal fuzzy multi numbers, Archimedean T-norm and T-conorm, Multiple criteria decision making.

**Corresponding Author:** İrfan Deli ([irfandeli20@gmail.com](mailto:irfandeli20@gmail.com))

### 1. INTRODUCTION

**T**hroughout our daily life including development of modern science and artificial intelligence technology, more uncertain data processing problems urgently need to be solved and uncertain values need to be modelled. Therefore, we always encounter the uncertain values such as; good, very good, very very good, bad, very bad, hot, very hot, very very hot, cold, very cold, very very cold, and so on. These linguistic terms differ from individual to individual, from time to time and from environment to environment, and it is very difficult to model and solve an event or problem under uncertainty. For this, interval mathematics, probability theory, fuzzy set theory [1], intuitionistic fuzzy set theory [2] with strong applications, different theories are presented. The most comprehensive of set theory is fuzzy set theory proposed by

Zadeh [1] in 1965. Fuzzy sets and especially fuzzy numbers which are a fuzzy set on  $\mathbb{R}$  real numbers have been studied increasingly by many authors to express more abundant and flexible information than classical sets and fuzzy sets. For example: Ban et al. [3] found nearest trapezoidal approximation and the nearest symmetric trapezoidal approximation to a given fuzzy number, with respect to the average Euclidean distance, preserving the value and ambiguity. Cheng [4] developed a centroid-based distance method was suggested for ranking fuzzy numbers and Wang et al. [5] showed incorrect and have led to some misapplications of the method and updated the concept of centroid for fuzzy numbers. Wei and Chen [6] proposed a new similarity measure between generalized fuzzy numbers by combining the concepts of geometric distance, the perimeter and the height of generalized fuzzy numbers for calculating the degree of similarity between generalized fuzzy numbers. The proposed theory which contain uncertain information have gained much attention from past and latter researchers for applications in various fields in [7, 8].

Since intuitionistic fuzzy sets proposed by Atanassov [2] are successful to handle the uncertain situations of data in the decision making problems they have great practical potential in a variety of areas. For example; Wang and Xin [9] gave the axiom definition of distance measure between intuitionistic fuzzy sets. Luo and Zhao [10] defined a new distance measure between intuitionistic fuzzy sets, which is based on a matrix norm and a strictly increasing (or decreasing) binary function and applied to pattern recognitions. Szmidt and Kacprzyk [11] showed how to calculate distances for intuitionistic fuzzy sets not only from a mathematical point of view but also of an intuitive appeal making use of all the relevant information. Park et al. [12] proposed some measures applied to pattern recognitions. Liang and Shi [13] proposed new similarity measures and proved the relationships between some similarity measures. Garg [14] developed improved cosine similarity measure for an intuitionistic fuzzy sets by considering the interaction between the pairs of the membership degrees. based on the Hausdor metric, are suggested. Recently, Lei et al. [15] defined two subtraction and division operations of intuitionistic fuzzy numbers (IFNs), and develop a sequence of general integrals dealing with continuous intuitionistic fuzzy data based on Archimedean t-conorm and t-norm. Xia et al. [16] proposed some operations on intuitionistic fuzzy sets under Archimedean t-conorm and t-norms, Also, intuitionistic fuzzy sets, including Archimedean t-conorm and t-norm, gained attention of many researchers in [17, 18, 19, 10, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30].

Recently, Uluçay et al. [31] defined trapezoidal fuzzy multi-number which are more than one with the possibility of the same or the different membership functions allowing the repeated occurrences of any element. Deli and Keleş [32] and Deli and Keleş [33] developed new methods to solve multi-criteria decision-making problems using trapezoidal fuzzy multi-numbers under distance measures and similarity based on the concept of value and ambiguity of trapezoidal fuzzy multi-numbers. Then, Sahin et al. [34] developed a novel approach based on multi-criteria decision making(MCDM) trapezoidal fuzzy multi-number by defining Dice vector similarity and weighted Dice vector similarity measure. Also, Sahin et al. [35] presented some similarity measures for trapezoidal fuzzy multi numbers such as; Jaccard similarity measure, weighted Jacard similarity measure, Cosine similarity measure, weighted

cosine similarity measure, Hybrid vector similarity measure and weighted Hybrid vector similarity measure. Also, Kesen and Deli [36, 37] initiated Archimedean T-norm and T-conorm operations of trapezoidal fuzzy multi-numbers to aggregating trapezoidal fuzzy multi-numbers. Moreover, Uluçay et al. [38] defined intuitionistic trapezoidal fuzzy multi numbers which are more than one with the possibility of the same or the different membership functions allowing the repeated occurrences of any element under intuitionistic fuzzy numbers.

An aggregation operator performs the task of generate a ranking of alternatives in uncertainly information and considering the increasing complexity of decision-making situations, it is imperative to extend aggregation operators for fusing uncertain information with the different forms of attribute values. As we know, no studies about archimedean norms on intuitionistic trapezoidal fuzzy multi numbers have been conducted until now. This study focuses on the development of intuitionistic fuzzy multi numbers and aims to design a managerial decision-making solving method by generalized Kesen and Deli [36, 37]. Some operational principles of intuitionistic fuzzy multi numbers on account of the Archimedean t-norm and t-conorm are initiated, on which two intuitionistic fuzzy multi numbers operators are established by taking various weight forms. Moreover, we explore the aggregation operators' idempotency, boundedness, and monotonicity, as well as analyze some particular forms of these operators. Furthermore, these aggregation operators are employed to design a method of deriving an overall performance from evaluation of experts with intuitionistic fuzzy multi numbers. This paper is derived from the second author's master's thesis [39] supervised by the first author.

## 2. PRELIMINARIES

**Definition 2.1** ([1]). Let  $X$  be a non-empty set. A *fuzzy set*  $F$  on  $X$  is defined as:

$$F = \{ \langle x, \mu_F(x) \rangle : x \in X \},$$

where  $\mu_F : X \rightarrow [0, 1]$  is a mapping.

**Definition 2.2** ([40]). Let  $X$  be a non-empty set. A *multi-fuzzy set*  $G$  on  $X$  is defined as:

$$G = \{ \langle x, \mu_G^1(x), \mu_G^2(x), \dots, \mu_G^i(x), \dots \rangle : x \in X \},$$

where  $\mu_G^i : X \rightarrow [0, 1]$  is a mapping for all  $i \in \{1, 2, \dots, p\}$ .

**Definition 2.3** ([31]). Let  $\eta_A^i \in [0, 1]$   $i \in \{1, 2, \dots, p\}$  and  $a, b, c, d \in R$  such that  $a \leq b \leq c \leq d$ . Then a *trapezoidal fuzzy multi-number* (TFM-number)  $A = \langle (a, b, c, d); \eta_A^1, \eta_A^2, \dots, \eta_A^p \rangle$  is a special fuzzy multi-set on the real number set  $R$ , whose membership functions are defined as:

$$\mu_A^i(x) = \begin{cases} (x - a)\eta_A^i / (b - a) & a \leq x < b \\ \eta_A^i & b \leq x < c \\ (d - x)\eta_A^i / (d - c) & c \leq x < d \\ 0 & otherwise \end{cases}$$

Note that the set of all TFM-number on  $\mathbb{R}^+$  will be denoted by  $\mathcal{U}(\mathbb{R}^+)$ .

**Definition 2.4** ([41]). A function  $\tau : [0, 1] \times [0, 1] \rightarrow [0, 1]$  is called a *t-norm*, if it satisfies the following four conditions:

- (i)  $\tau(1, x) = x$  for all  $x \in [0, 1]$ ,
- (ii)  $\tau(x, y) = \tau(y, x)$  for all  $x, y \in [0, 1]$ ,
- (iii)  $\tau(x, \tau(y, z)) = \tau(\tau(x, y), z)$  for all  $x, y, z \in [0, 1]$ ,
- (iv) if  $x \leq x'$  and  $y \leq y'$ , then  $\tau(x, y) \leq \tau(x', y')$ , where  $x, y, x', y' \in [0, 1]$ .

**Definition 2.5** ([41]). A function  $s : [0, 1] \times [0, 1] \rightarrow [0, 1]$  is called a *t-conorm*, if it satisfies the following four conditions:

- (i)  $s(0, x) = x$  for all  $x \in [0, 1]$ ,
- (ii)  $s(x, y) = s(y, x)$  for all  $x, y \in [0, 1]$ ,
- (iii)  $s(x, s(y, z)) = s(s(x, y), z)$  for all  $x, y, z \in [0, 1]$ ,
- (iv) if  $x \leq x'$  and  $y \leq y'$  then  $s(x, y) \leq s(x', y')$ , where  $x, y, x', y' \in [0, 1]$ .

**Definition 2.6** ([41]). A t-norm function  $\tau(x, y)$  is called an *Archimedean t-norm*, if it is continuous and  $\tau(x, x) < x$  for all  $x \in (0, 1)$ . An Archimedean t-norm is called a *strict Archimedean t-norm*, if it is strictly increasing in each variable for  $x, y \in (0, 1)$ . A t-conorm function  $s(x, y)$  is called Archimedean t-conorm if it is continuous and  $s(x, x) > x$  for all  $x \in (0, 1)$ . An Archimedean t-conorm is called a *strict Archimedean t-conorm*, if it is strictly increasing in each variable for  $x, y \in (0, 1)$ .

In [42], a strict Archimedean t-norm  $s$  is expressed and Archimedean t-conorm  $T$  is expressed via function  $h$  and  $g$ , respectively as: for all  $x, Y, T \in [0, 1]$ ,

$$s(x, y) = h^{-1}(h(x) + h(y)),$$

and

$$\tau(x, y) = g^{-1}(g(x) + g(y)) \quad \text{with} \quad h(t) = g(1 - t),$$

where  $g$  is continuous Archimedean t-norm and strictly decreasing function such that  $g : [0, 1] \rightarrow [0, 1]$  and  $g(1) = 0$ .

In [8, 43], according to specific forms of function  $g$ , some well-known t-conorms and t-norms are given as follows.

- (1) Let  $g_1(t) = -Int$ . Then  $h_1(t) = -\ln(1 - t)$ ,  $g_1^{-1}(t) = e^{-t}$ ,  $h_1^{-1}(t) = 1 - e^{-t}$  and Algebraic t-conorm and t-norm are obtained as follows:

$$s_A(x, y) = x + y - xy, \tau(x, y) = xy.$$

- (2) Let  $g_2(t) = \ln(\frac{2-t}{t})$ . Then  $h_2(t) = \ln(\frac{2-(1-t)}{1-t})$ ,  $g_2^{-1}(t) = \frac{2}{e^t+1}$ ,  $h_2^{-1}(t) = 1 - \frac{2}{e^t+1}$  and we can get Einstein t-conorm and t-norm:

$$s_E(x, y) = \frac{x + y}{1 + xy}, \tau_E(x, y) = \frac{xy}{1 + (1 - x)(1 - y)}.$$

- (3) Let  $g_3(t) = \ln(\frac{\gamma+(1-\gamma)t}{t})$ ,  $\gamma \in (0, +\infty)$ . Then we have  $h_3(t) = \ln(\frac{\gamma+(1-\gamma)(1-t)}{1-t})$ ,  $g_3^{-1}(t) = \frac{\gamma}{e^t+\gamma-1}$ ,  $h_3^{-1}(t) = 1 - \frac{\gamma}{e^t+\gamma-1}$  and Hamacher t-conorm and t-norm are obtained as follows:

$$s_H(x, y) = \frac{x+y-xy-(1-\gamma)xy}{1-(1-\gamma)xy}, \quad \gamma \in (0, +\infty)$$

$$\tau_H(x, y) = \frac{xy}{\gamma+(1-\gamma)(x+y-xy)}, \quad \gamma \in (0, +\infty).$$

(4) Let  $g_4(t) = \ln(\frac{\gamma-1}{\gamma^t-1})$ ,  $\gamma \in (0, +\infty)$ . Then  $h_4(t) = \ln(\frac{\gamma}{(1-\gamma)(t)-1})$ ,  $g_4^{-1}(t) = \log_\gamma \frac{\gamma-1+e^t}{e^{(t)}}$ ,  $h_4^{-1}(t) = \log_\gamma \frac{\gamma+e^t}{e^{t(1-\gamma)}}$  and get Frank t-conorm and t-norm:

$$s_F(x, y) = 1 - \ln_\gamma(1 + \frac{(\gamma^{1-x}-1)(\gamma^{1-y}-1)}{\gamma-1}), \quad \gamma \in (1, +\infty)$$

$$\tau_F(x, y) = \ln_\gamma(1 + \frac{(\gamma^x-1)(\gamma^y-1)}{\gamma-1}), \quad \gamma \in (1, +\infty).$$

**Definition 2.7** ([16, 41, 43]). Let  $\dot{A} = (\theta_{\dot{A}}, \sigma_{\dot{A}})$ ,  $\dot{A}_1 = (\rho_{\dot{A}_1}, \sigma_{\dot{A}_1})$  and  $\dot{A}_2 = (\rho_{\dot{A}_2}, \sigma_{\dot{A}_2})$  be three intuitionistic fuzzy elements. Then the operations of intuitionistic fuzzy elements based on Archimedean t-norm and Archimedean t-conorm is defined as:

- (i)  $\dot{A}_1 \oplus \dot{A}_2 = (s(\rho_{\dot{A}_1}, \rho_{\dot{A}_2}), \tau(\sigma_{\dot{A}_1}, \sigma_{\dot{A}_2}))$   
 $= (h^{-1}(h(\rho_{\dot{A}_1}) + h(\rho_{\dot{A}_2})), g^{-1}(g(\sigma_{\dot{A}_1}) + g(\sigma_{\dot{A}_2})))$ ,
- (ii)  $\dot{A}_1 \otimes \dot{A}_2 = (\tau(\rho_{\dot{A}_1}, \rho_{\dot{A}_2}), s(\sigma_{\dot{A}_1}, \sigma_{\dot{A}_2}))$   
 $= (g^{-1}(g(\rho_{\dot{A}_1}) + g(\rho_{\dot{A}_2})), h^{-1}(h(\sigma_{\dot{A}_1}) + h(\sigma_{\dot{A}_2})))$ ,
- (iii)  $\lambda \dot{A} = (h^{-1}(\lambda h(\rho_{\dot{A}})), g^{-1}(\lambda g(\sigma_{\dot{A}}))), \lambda > 0$ ,
- (iv)  $\dot{A}^\lambda = (g^{-1}(\lambda g(\rho_{\dot{A}})), h^{-1}(\lambda h(\sigma_{\dot{A}}))), \lambda > 0$ .

Especially, in [16, 41, 43, 44] the operational laws based on the function  $g$  are given as:

- (1) for  $g_1(t) = -\log(t)$ ,
  - (i)  $\dot{A}_1 \oplus \dot{A}_2 = (\rho_{\dot{A}_1} + \rho_{\dot{A}_2} - \rho_{\dot{A}_1} \rho_{\dot{A}_2}, \delta_{\dot{A}_1} \delta_{\dot{A}_2})$ ,
  - (ii)  $\dot{A}_1 \otimes \dot{A}_2 = (\rho_{\dot{A}_1} \rho_{\dot{A}_2}, \delta_{\dot{A}_1} + \delta_{\dot{A}_2} - \delta_{\dot{A}_1} \delta_{\dot{A}_2})$ ,
  - (iii)  $\lambda \dot{A} = (1 - (1 - \rho_{\dot{A}})^\lambda, \delta_{\dot{A}}^\lambda), \lambda > 0$ ,
  - (iv)  $\dot{A}^\lambda = (\rho_{\dot{A}}^\lambda, 1 - (1 - \delta_{\dot{A}})^\lambda), \lambda > 0$ ,
- (2) for  $g_2(t) = -\log(\frac{2-t}{t})$ ,
  - (i)  $\dot{A}_1 \oplus \dot{A}_2 = (\frac{\rho_{\dot{A}_1} + \rho_{\dot{A}_2}}{1 + \rho_{\dot{A}_1} \rho_{\dot{A}_2}}, \frac{\delta_{\dot{A}_1} \delta_{\dot{A}_2}}{1 + (1 - \delta_{\dot{A}_1})(1 - \delta_{\dot{A}_2})})$ ,
  - (ii)  $\dot{A}_1 \otimes \dot{A}_2 = (\frac{\rho_{\dot{A}_1} \rho_{\dot{A}_2}}{1 + (1 - \rho_{\dot{A}_1})(1 - \rho_{\dot{A}_2})}, \frac{\delta_{\dot{A}_1} + \delta_{\dot{A}_2}}{1 + \delta_{\dot{A}_1} \delta_{\dot{A}_2}})$ ,
  - (iii)  $\lambda \dot{A} = (\frac{(1 + \rho_{\dot{A}}^\lambda - (1 - \rho_{\dot{A}})^\lambda)}{(1 + \rho_{\dot{A}})^\lambda + (1 - \rho_{\dot{A}})^\lambda}, \frac{2\delta_{\dot{A}}^\lambda}{(2 - \delta_{\dot{A}})^\lambda + \delta_{\dot{A}}^\lambda}), \lambda > 0$ ,
  - (iv)  $\dot{A}^\lambda = (\frac{2\rho_{\dot{A}}^\lambda}{(2 - \rho_{\dot{A}})^\lambda + \rho_{\dot{A}}^\lambda}, \frac{(1 + \delta_{\dot{A}})^\lambda - (1 - \delta_{\dot{A}})^\lambda}{(1 + \delta_{\dot{A}})^\lambda + (1 - \delta_{\dot{A}})^\lambda}), \lambda > 0$ ,
- (3) for  $g_3(t) = \log(\frac{\gamma + (1-\gamma)t}{t})$ ,  $\gamma \in (0, +\infty)$ ,
  - (i)  $\dot{A}_1 \oplus \dot{A}_2 = (\frac{\rho_{\dot{A}_1} + \rho_{\dot{A}_2} - \rho_{\dot{A}_1} \rho_{\dot{A}_2} - (1-\gamma)\rho_{\dot{A}_1} \rho_{\dot{A}_2}}{1 - (1-\gamma)\rho_{\dot{A}_1} \rho_{\dot{A}_2}}, \frac{\delta_{\dot{A}_1} \delta_{\dot{A}_2}}{\gamma + (1-\gamma)(\delta_{\dot{A}_1} + \delta_{\dot{A}_2} - \delta_{\dot{A}_1} \delta_{\dot{A}_2})})$ ,
  - (ii)  $\dot{A}_1 \otimes \dot{A}_2 = (\frac{\rho_{\dot{A}_1} \rho_{\dot{A}_2}}{\gamma + (1-\gamma)(\rho_{\dot{A}_1} + \rho_{\dot{A}_2} - \rho_{\dot{A}_1} \rho_{\dot{A}_2})}, \frac{\delta_{\dot{A}_1} + \delta_{\dot{A}_2} - \delta_{\dot{A}_1} \delta_{\dot{A}_2} - (1-\gamma)\delta_{\dot{A}_1} \delta_{\dot{A}_2}}{1 - (1-\gamma)\delta_{\dot{A}_1} \delta_{\dot{A}_2}})$ ,

- (iii)  $\lambda \dot{A} = \left( \frac{(1+(\gamma-1)\rho_A)^\lambda - (1-\rho_A)^\lambda}{(1+(\gamma-1)\rho_A)^\lambda + (\gamma-1)(1-\rho_A)^\lambda}, \frac{\gamma \delta_{\dot{A}}^\lambda}{(1+(\gamma-1)(1-\delta_{\dot{A}})^\lambda)^\lambda + (\gamma-1)\delta_{\dot{A}}^\lambda} \right), \lambda > 0,$
- (iv)  $\dot{A}^\lambda = \left( \frac{\gamma \rho_{\dot{A}}^\lambda}{(1+(\gamma-1)(1-\rho_{\dot{A}})^\lambda)^\lambda + (\gamma-1)\rho_{\dot{A}}^\lambda}, \frac{(1+(\gamma-1)\delta_{\dot{A}})^\lambda - (1-\delta_{\dot{A}})^\lambda}{(1+(\gamma-1)\delta_{\dot{A}})^\lambda + (\gamma-1)(1-\delta_{\dot{A}})^\lambda} \right), \lambda > 0,$
- (4) for  $g_4(t) = \ln\left(\frac{\gamma-1}{\gamma^t-1}\right), \gamma \in (1, \infty),$ 
  - (i)  $\dot{A}_1 \oplus \dot{A}_2 = \left( 1 - \ln_\gamma\left(1 + \frac{(\gamma^{1-\rho_{\dot{A}_1}}-1)(\gamma^{1-\rho_{\dot{A}_2}}-1)}{\gamma-1}\right), \ln_\gamma\left(1 + \frac{(\gamma^{\delta_{\dot{A}_1}}-1)(\gamma^{\delta_{\dot{A}_2}}-1)}{\gamma-1}\right) \right), \lambda > 1,$
  - (ii)  $\dot{A}_1 \otimes \dot{A}_2 = \left( \ln_\gamma\left(1 + \frac{(\gamma^{\rho_{\dot{A}_1}}-1)(\gamma^{\rho_{\dot{A}_2}}-1)}{\gamma-1}\right), 1 - \ln_\gamma\left(1 + \frac{(\gamma^{1-\delta_{\dot{A}_1}}-1)(\gamma^{1-\delta_{\dot{A}_2}}-1)}{\gamma-1}\right) \right), \lambda > 1,$
  - (iii)  $\lambda \dot{A} = \left( 1 - \ln_\gamma\left(1 + \frac{(\gamma^{1-\rho_A}-1)^\lambda}{(\gamma-1)^{\lambda-1}}\right), \ln_\gamma\left(1 + \frac{(\gamma^{\delta_A}-1)^\lambda}{(\gamma-1)^{\lambda-1}}\right) \right), \lambda > 0, \lambda > 1,$
  - (iv)  $\dot{A}^\lambda = \left( \ln_\gamma\left(1 + \frac{(\gamma^{\rho_A}-1)^\lambda}{(\gamma-1)^{\lambda-1}}\right), 1 - \ln_\gamma\left(1 + \frac{(\gamma^{1-\delta_A}-1)^\lambda}{(\gamma-1)^{\lambda-1}}\right) \right), \lambda > 0, \lambda > 1.$

**Definition 2.8** ([2]). Let  $X$  be a non-empty set. An intuitionistic fuzzy set  $\mathbb{IF}$  on  $X$  is defined as:

$$\mathbb{IF} = \{ \langle x, \mu_{\mathbb{IF}}(x), \nu_{\mathbb{IF}}(x) \rangle : x \in X \},$$

where  $\mu_{\mathbb{IF}}(x), \nu_{\mathbb{IF}}(x) : X \rightarrow [0, 1]$  such that  $0 \leq \mu_{\mathbb{IF}}(x) + \nu_{\mathbb{IF}}(x) \leq 1$  for  $x \in X$ .

**Definition 2.9** ([38]). Let  $\rho_A^i, \delta_A^i \in [0, 1]$  ( $i \in \{1, 2, \dots, P\}$ ) and  $a, b, c, d \in \mathbb{R}$  such that  $a \leq b \leq c \leq d$ . Then a trapezoidal intuitionistic fuzzy multi-numbers (TIFM-numbers)  $\dot{A} = \langle [a, b, c, d]; (\rho_A^1, \rho_A^2, \dots, \rho_A^P), (\delta_A^1, \delta_A^2, \dots, \delta_A^P) \rangle$  is a special intuitionistic fuzzy multi-set on the real number set  $\mathbb{R}$ , whose membership functions and non-membership functions are defined, respectively as:

$$\mu_{\dot{A}}^i(x) = \begin{cases} \frac{(x-a)}{(b-a)} \rho_A^i, & a \leq x < b \\ \rho_A^i, & b \leq x \leq c \\ \frac{(d-x)}{(d-c)} \rho_A^i, & c < x \leq d \\ 0, & \text{otherwise,} \end{cases} \quad \text{and} \quad \nu_{\dot{A}}^i(x) = \begin{cases} \frac{(b-x) + \delta_A^i(x-a)}{(b-a)}, & a \leq x < b \\ \delta_A^i, & b \leq x \leq c \\ \frac{(x-c) + \delta_A^i(d-x)}{(d-c)}, & c < x \leq d \\ 1, & \text{otherwise,} \end{cases}$$

Note that the set of all TIFM-numbers on  $[0, 1]$  will be denoted by  $\Lambda$ .

Let  $\dot{A}, \dot{B} \in \Lambda$  and  $\gamma \neq 0$ , where  $\dot{A} = \langle [a_1, b_1, c_1, d_1]; (\rho_A^1, \rho_A^2, \dots, \rho_A^P), (\delta_A^1, \delta_A^2, \dots, \delta_A^P) \rangle,$   
 $\dot{B} = \langle [a_2, b_2, c_2, d_2]; (\rho_B^1, \rho_B^2, \dots, \rho_B^P), (\delta_B^1, \delta_B^2, \dots, \delta_B^P) \rangle.$  Then  $\dot{A} + \dot{B}, \dot{A} \cdot \dot{B}, \gamma \dot{A}$  and  $\dot{A}^\gamma$  are respectively defined as follows:

- (i)  $\dot{A} + \dot{B} = \langle [a_1 + a_2, b_1 + b_2, c_1 + c_2, d_1 + d_2]; (s(\rho_A^1, \rho_B^1), s(\rho_A^2, \rho_B^2), \dots, s(\rho_A^P, \rho_B^P)), (t(\delta_A^1, \delta_B^1), t(\delta_A^2, \delta_B^2), \dots, t(\delta_A^P, \delta_B^P)) \rangle,$
- (ii)  $\dot{A} \cdot \dot{B} = \langle [a_1 a_2, b_1 b_2, c_1 c_2, d_1 d_2]; (t(\rho_A^1, \rho_B^1), \dots, t(\rho_A^P, \rho_B^P)), (s(\delta_A^1, \delta_B^1), \dots, s(\delta_A^P, \delta_B^P)) \rangle,$
- (iii)  $\gamma \dot{A} = \langle [\gamma a_1, \gamma b_1, \gamma c_1, \gamma d_1]; (1 - (1 - \rho_A^1)^\gamma, 1 - (1 - \rho_A^2)^\gamma, \dots, 1 - (1 - \rho_A^P)^\gamma), ((\delta_A^1)^\gamma, (\delta_A^2)^\gamma, \dots, (\delta_A^P)^\gamma) \rangle (\gamma \geq 0),$
- (iv)  $\dot{A}^\gamma = \langle [a_1^\gamma, b_1^\gamma, c_1^\gamma, d_1^\gamma]; ((\rho_A^1)^\gamma, (\rho_A^2)^\gamma, \dots, (\rho_A^P)^\gamma), (1 - (1 - \delta_A^1)^\gamma, 1 - (1 - \delta_A^2)^\gamma, \dots, 1 - (1 - \delta_A^P)^\gamma) \rangle (\gamma \geq 0),$

where  $s$  is s-norm and  $t$  is a t-norm.

**Definition 2.10** ([38]). Let  $\dot{A}_j = \langle [a_j, b_j, c_j, d_j]; (\rho_{\dot{A}_j}^1, \rho_{\dot{A}_j}^2, \dots, \rho_{\dot{A}_j}^P), (\delta_{\dot{A}_j}^1, \delta_{\dot{A}_j}^2, \dots, \delta_{\dot{A}_j}^P) \rangle, j \in I_n$  be a collection of TIFM-numbers and let  $S\dot{A}_w : \Lambda^n \rightarrow \Lambda$ , if

$$SA_w(\dot{A}_1, \dot{A}_2, \dot{A}_3, \dots, \dot{A}_n) = \sum_{j=1}^n w_j \dot{A}_j$$

then  $S\dot{A}_w$  is called a *TIFM-number weighted arithmetic operator of dimension n*, where  $w = (w_1, w_2, w_3, \dots, w_n)^T$  is the weight vector of  $\dot{A}_j, j \in I_n$ , with  $w_j \in [0, 1]$  and  $\sum_{j=1}^n w_j = 1$ .

**Definition 2.11.** Let  $\dot{A} = \langle (a, b, c, d); (\rho_{\dot{A}}^1, \rho_{\dot{A}}^2, \dots, \rho_{\dot{A}}^P), (\delta_{\dot{A}}^1, \delta_{\dot{A}}^2, \dots, \delta_{\dot{A}}^P) \rangle$  be a TIFM-numbers. Then

(i) *1 score value* of TIFM-number  $\dot{A}$  based on [45, 46], denoted by  $\bar{S}_1(\dot{A})$ , is defined as:

$$(2.1) \quad \bar{S}_1(\dot{A}) = \left( \frac{d-a+c-b}{2} \right) \times \sum_{i=1}^P \frac{1 - (\delta_{\dot{A}}^i)^2}{2 - (\rho_{\dot{A}}^i)^2 - (\delta_{\dot{A}}^i)^2},$$

(ii) *2 score value* of TIFM-number  $\dot{A}$  based on [45, 47], denoted by  $\bar{S}_2(\dot{A})$ , is defined as:

$$(2.2) \quad \bar{S}_2(\dot{A}) = \left( \frac{d-a+c-b}{2} \right) \times \sum_{i=1}^P \frac{e^{(\rho_{\dot{A}}^i)^2 - (\delta_{\dot{A}}^i)^2}}{2 - (\rho_{\dot{A}}^i)^2 - (\delta_{\dot{A}}^i)^2},$$

(iii) *3 score value* of TIFM-number  $\dot{A}$  based on Zhang [45, 48, 49, 50], denoted by  $\bar{S}_3(\dot{A})$ , is defined as:

$$(2.3) \quad \bar{S}_3(\dot{A}) = \left( \frac{d-a+c-b}{2} \right) \times \sum_{i=1}^P \left( \frac{1}{2} + \sqrt{(\rho_{\dot{A}}^i)^2 + (\delta_{\dot{A}}^i)^2} \left( \frac{1}{2} - \arcsin \left( \frac{\delta_{\dot{A}}^i}{\sqrt{(\rho_{\dot{A}}^i)^2 + (\delta_{\dot{A}}^i)^2}} \right) \right) \right),$$

(iv) *4 score value* of TIFM-number  $\dot{A}$  based on [45, 51], denoted by  $\bar{S}_4(\dot{A})$ , is defined as:

$$(2.4) \quad \bar{S}_4(\dot{A}) = \left( \frac{d-a+c-b}{2} \right) \times \sum_{i=1}^P \left( (\rho_{\dot{A}}^i)^2 - (\delta_{\dot{A}}^i)^2 \right) + \left( \frac{e^{(\rho_{\dot{A}}^i)^2 - (\delta_{\dot{A}}^i)^2}}{e^{(\rho_{\dot{A}}^i)^2 - (\delta_{\dot{A}}^i)^2} + 1} - \frac{1}{2} \right).$$

Let  $\dot{A} = \langle (a, b, c, d); (\rho_{\dot{A}}^1, \rho_{\dot{A}}^2, \dots, \rho_{\dot{A}}^P), (\delta_{\dot{A}}^1, \delta_{\dot{A}}^2, \dots, \delta_{\dot{A}}^P) \rangle$  and

$\dot{B} = \langle (a, b, c, d); (\rho_{\dot{B}}^1, \rho_{\dot{B}}^2, \dots, \rho_{\dot{B}}^P), (\delta_{\dot{B}}^1, \delta_{\dot{B}}^2, \dots, \delta_{\dot{B}}^P) \rangle$  be two TIFM-numbers. Then comparison of  $\dot{A}$  and  $\dot{B}$  is given as:

- (1)  $\bar{S}_j(\dot{A}) < \bar{S}_j(\dot{B})$ , then  $\dot{A} < \dot{B}$  (j=1,2,3,4),
- (2)  $\bar{S}_j(\dot{A}) = \bar{S}_j(\dot{B})$ , then  $\dot{A} = \dot{B}$  (j=1,2,3,4).

**Definition 2.12.** Let  $\dot{A}_i = \langle (a_i, b_i, c_i, d_i); (\rho_{\dot{A}_i}^1, \rho_{\dot{A}_i}^2, \dots, \rho_{\dot{A}_i}^P), (\delta_{\dot{A}_i}^1, \delta_{\dot{A}_i}^2, \dots, \delta_{\dot{A}_i}^P) \rangle$  for  $i=1,2$  be two TIFM-numbers. Then the operations of TIFM-numbers under Archimedean t-norm and Archimedean t-conorm are defined as:

(i) the *sum* of  $\dot{A}_1$  and  $\dot{A}_2$ , denoted by  $\dot{A}_1 \oplus \dot{A}_2$ , is defined as:

$$\dot{A}_1 \oplus \dot{A}_2$$

$$\begin{aligned}
 &= \langle (s(a_1, a_2), s(b_1, b_2), s(c_1, c_2), s(d_1, d_2)); s(\rho_{\dot{A}_1}^1, \rho_{\dot{A}_2}^1), s(\rho_{\dot{A}_1}^2, \rho_{\dot{A}_2}^2), \\
 &\quad \dots, s(\rho_{\dot{A}_1}^P, \rho_{\dot{A}_2}^P), (t(\delta_{\dot{A}_1}^1, \delta_{\dot{A}_2}^1), t(\delta_{\dot{A}_1}^2, \delta_{\dot{A}_2}^2), \dots, t(\delta_{\dot{A}_1}^P, \delta_{\dot{A}_2}^P))) \rangle \\
 &= \langle (h_1^{-1}(h_1(a_1) + h_1(a_2)), h_1^{-1}(h_1(b_1) + h_1(b_2)), \\
 &\quad h_1^{-1}(h_1(c_1) + h_1(c_2)), h_1^{-1}(h_1(d_1) + h_1(d_2))); \\
 &\quad h_1^{-1}(h_1(\rho_{\dot{A}_1}^1) + h_1(\rho_{\dot{A}_2}^1)), h_1^{-1}(h_1(\rho_{\dot{A}_1}^2) + h_1(\rho_{\dot{A}_2}^2)), \dots, h_1^{-1}(h_1(\rho_{\dot{A}_1}^P) + h_1(\rho_{\dot{A}_2}^P)), \\
 &\quad g_1^{-1}(g_1(\delta_{\dot{A}_1}^1) + g_1(\delta_{\dot{A}_2}^1)), g_1^{-1}(g_1(\delta_{\dot{A}_1}^2) + g_1(\delta_{\dot{A}_2}^2)), \dots, g_1^{-1}(g_1(\delta_{\dot{A}_1}^P) + g_1(\delta_{\dot{A}_2}^P))) \rangle,
 \end{aligned}$$

(ii) the *product* of  $\dot{A}_1$  and  $\dot{A}_2$ , denoted by  $\dot{A}_1 \otimes \dot{A}_2$ , is defined as:

$$\begin{aligned}
 &\dot{A}_1 \otimes \dot{A}_2 \\
 &= \langle (\tau(a_1, a_2), \tau(b_1, b_2), \tau(c_1, c_2), \tau(d_1, d_2)); (\tau(\rho_{\dot{A}_1}^1, \rho_{\dot{A}_2}^1), \tau(\rho_{\dot{A}_1}^2, \rho_{\dot{A}_2}^2), \\
 &\quad \dots, \tau(\rho_{\dot{A}_1}^P, \rho_{\dot{A}_2}^P)), s(\delta_{\dot{A}_1}^1, \delta_{\dot{A}_2}^1), s(\delta_{\dot{A}_1}^2, \delta_{\dot{A}_2}^2), \dots, s(\delta_{\dot{A}_1}^P, \delta_{\dot{A}_2}^P)) \rangle \\
 &= \langle (g_1^{-1}(g_1(\rho_{\dot{A}_1}^1) + g_1(\rho_{\dot{A}_2}^1)), g_1^{-1}(g_1(\rho_{\dot{A}_1}^2) + g_1(\rho_{\dot{A}_2}^2)), \dots, g_1^{-1}(g_1(\rho_{\dot{A}_1}^P) + g_1(\rho_{\dot{A}_2}^P)), \\
 &\quad (h_1^{-1}(h_1(\delta_{\dot{A}_1}^1) + h_1(\delta_{\dot{A}_2}^1)), h_1^{-1}(h_1(\delta_{\dot{A}_1}^2) + h_1(\delta_{\dot{A}_2}^2)), \dots, h_1^{-1}(h_1(\delta_{\dot{A}_1}^P) + h_1(\delta_{\dot{A}_2}^P)))) \rangle,
 \end{aligned}$$

(iii) the *scalar multiple* of  $\dot{A}_i$ , denoted by  $\lambda \dot{A}_i$ , is defined as:

$$\begin{aligned}
 &\lambda \dot{A}_i \\
 &= \langle (h_1^{-1}(\lambda h_1(a_i)), h_1^{-1}(\lambda h_1(b_i)), h_1^{-1}(\lambda h_1(c_i)), h_1^{-1}(\lambda h_1(d_i))), \\
 &\quad (g_1^{-1}(\lambda g_1(a_i)), g_1^{-1}(\lambda g_1(b_i)), g_1^{-1}(\lambda g_1(c_i)), g_1^{-1}(\lambda g_1(d_i))); \\
 &\quad h_1^{-1}(\lambda h_1(\rho_{\dot{A}_i}^1)), h_1^{-1}(\lambda h_1(\rho_{\dot{A}_i}^2)), \dots, h_1^{-1}(\lambda h_1(\rho_{\dot{A}_i}^P)), \\
 &\quad (g_1^{-1}(\lambda g_1(\delta_{\dot{A}_i}^1)), g_1^{-1}(\lambda g_1(\delta_{\dot{A}_i}^2)), \dots, g_1^{-1}(\lambda g_1(\delta_{\dot{A}_i}^P)))) \rangle,
 \end{aligned}$$

(iv) the *scalar power* of  $\dot{A}_i$ , denoted by  $\dot{A}_i^\lambda$ , is defined as:

$$\begin{aligned}
 &\dot{A}_i^\lambda \\
 &= \langle (g_1^{-1}(\lambda g_1(a_i)), g_1^{-1}(\lambda g_1(b_i)), g_1^{-1}(\lambda g_1(c_i)), g_1^{-1}(\lambda g_1(d_i))), \\
 &\quad ((h_1^{-1}(\lambda h_1(a_i)), h_1^{-1}(\lambda h_1(b_i)), h_1^{-1}(\lambda h_1(c_i)), h_1^{-1}(\lambda h_1(d_i))))); \\
 &\quad g_1^{-1}(\lambda g_1(\rho_{\dot{A}_i}^1)), g_1^{-1}(\lambda g_1(\rho_{\dot{A}_i}^2)), \dots, g_1^{-1}(\lambda g_1(\rho_{\dot{A}_i}^P)), \\
 &\quad (h_1^{-1}(\lambda h_1(\delta_{\dot{A}_i}^1)), h_1^{-1}(\lambda h_1(\delta_{\dot{A}_i}^2)), \dots, h_1^{-1}(\lambda h_1(\delta_{\dot{A}_i}^P)))) \rangle.
 \end{aligned}$$

Especially, under Archimedean t-norm and Archimedean t-conorm, the operational laws based on the function  $g$  are given as:

(1) for  $g_1(t) = -\log(t)$ ,

(i)  $\dot{A}_1 \oplus \dot{A}_2 = \langle (a_1 + a_2 - a_1 a_2, b_1 + b_2 - b_1 b_2, c_1 + c_2 - c_1 c_2, d_1 + d_2 - d_1 d_2);$   
 $(\rho_{\dot{A}_1}^1 + \rho_{\dot{A}_2}^1 - \rho_{\dot{A}_1}^1 \rho_{\dot{A}_2}^1, \rho_{\dot{A}_1}^2 + \rho_{\dot{A}_2}^2 - \rho_{\dot{A}_1}^2 \rho_{\dot{A}_2}^2, \dots,$   
 $\rho_{\dot{A}_1}^P + \rho_{\dot{A}_2}^P - \rho_{\dot{A}_1}^P \rho_{\dot{A}_2}^P), (\delta_{\dot{A}_1}^1 \delta_{\dot{A}_2}^1, \delta_{\dot{A}_1}^2 \delta_{\dot{A}_2}^2, \dots, \delta_{\dot{A}_1}^P \delta_{\dot{A}_2}^P) \rangle,$

(ii)  $\dot{A}_1 \otimes \dot{A}_2 = \langle (a_1 a_2, b_1 b_2, c_1 c_2, d_1 d_2); (\rho_{\dot{A}_1}^1 \rho_{\dot{A}_2}^1, \rho_{\dot{A}_1}^2 \rho_{\dot{A}_2}^2, \dots, \rho_{\dot{A}_1}^P \rho_{\dot{A}_2}^P),$   
 $(\delta_{\dot{A}_1}^1 + \delta_{\dot{A}_2}^1 - \delta_{\dot{A}_1}^1 \delta_{\dot{A}_2}^1, \delta_{\dot{A}_1}^2 + \delta_{\dot{A}_2}^2 - \delta_{\dot{A}_1}^2 \delta_{\dot{A}_2}^2, \dots,$   
 $\delta_{\dot{A}_1}^P + \delta_{\dot{A}_2}^P - \delta_{\dot{A}_1}^P \delta_{\dot{A}_2}^P) \rangle,$

(iii)  $\lambda \dot{A}_i = \langle (1 - (1 - a_i)^\lambda, 1 - (1 - b_i)^\lambda, 1 - (1 - c_i)^\lambda, 1 - (1 - d_i)^\lambda);$   
 $1 - (1 - \rho_{\dot{A}_i}^1)^\lambda, 1 - (1 - \rho_{\dot{A}_i}^2)^\lambda, \dots,$   
 $1 - (1 - \rho_{\dot{A}_i}^P)^\lambda), ((\delta_{\dot{A}_i}^1)^\lambda, (\delta_{\dot{A}_i}^2)^\lambda, \dots, (\delta_{\dot{A}_i}^P)^\lambda) \rangle, \quad \lambda > 0,$

(iv)  $\dot{A}_i^\lambda = \langle ((a_i)^\lambda, (b_i)^\lambda, (c_i)^\lambda, (d_i)^\lambda); (\rho_{\dot{A}_i}^1)^\lambda, (\rho_{\dot{A}_i}^2)^\lambda, \dots,$   
 $(\rho_{\dot{A}_i}^P)^\lambda, (1 - (1 - \delta_{\dot{A}_i}^1)^\lambda, 1 - (1 - \delta_{\dot{A}_i}^2)^\lambda, \dots, 1 - (1 - \delta_{\dot{A}_i}^P)^\lambda) \rangle, \quad \lambda > 0,$

(2) for  $g_2(t) = -\log(\frac{2-t}{t})$ ,



$$\begin{aligned}
 \text{(i)} \quad \dot{A}_1 \oplus \dot{A}_2 &= \left\langle \left( \frac{a_1+a_2}{1+a_1a_2}, \frac{b_1+b_2}{1+b_1b_2}, \frac{c_1+c_2}{1+c_1c_2}, \frac{d_1+d_2}{1+d_1d_2}; \frac{\rho_{A_1}^1+\rho_{A_2}^1}{1+\rho_{A_1}^1\rho_{A_2}^1}, \frac{\rho_{A_1}^2+\rho_{A_2}^2}{1+\rho_{A_1}^2\rho_{A_2}^2}, \dots, \frac{\rho_{A_1}^P+\rho_{A_2}^P}{1+\rho_{A_1}^P\rho_{A_2}^P}, \right. \right. \\
 &\quad \left. \left. \left( \frac{\delta_{A_1}^1\delta_{A_2}^1}{1+(1-\delta_{A_1}^1)(1-\delta_{A_2}^1)}, \frac{\delta_{A_1}^2\delta_{A_2}^2}{1+(1-\delta_{A_1}^2)(1-\delta_{A_2}^2)}, \dots, \frac{\delta_{A_1}^P\delta_{A_2}^P}{1+(1-\delta_{A_1}^P)(1-\delta_{A_2}^P)} \right) \right\rangle, \\
 \text{(ii)} \quad \dot{A}_1 \otimes \dot{A}_2 &= \left\langle \left( \frac{a_1a_2}{1+(1-a_1)(1-a_2)}, \frac{b_1b_2}{1+(1-b_1)(1-b_2)}, \frac{c_1c_2}{1+(1-c_1)(1-c_2)}, \frac{d_1d_2}{1+(1-d_1)(1-d_2)}; \right. \right. \\
 &\quad \left. \left. \frac{\rho_{A_1}^1\rho_{A_2}^1}{1+(1-\rho_{A_1}^1)(1-\rho_{A_2}^1)}, \frac{\rho_{A_1}^2\rho_{A_2}^2}{1+(1-\rho_{A_1}^2)(1-\rho_{A_2}^2)}, \dots, \frac{\rho_{A_1}^P\rho_{A_2}^P}{1+(1-\rho_{A_1}^P)(1-\rho_{A_2}^P)}, \right. \right. \\
 &\quad \left. \left. \left( \frac{\delta_{A_1}^1+\delta_{A_2}^1}{1+\delta_{A_1}^1\delta_{A_2}^1}, \frac{\delta_{A_1}^2+\delta_{A_2}^2}{1+\delta_{A_1}^2\delta_{A_2}^2}, \dots, \frac{\delta_{A_1}^P+\delta_{A_2}^P}{1+\delta_{A_1}^P\delta_{A_2}^P} \right) \right\rangle, \\
 \text{(iii)} \quad \lambda \dot{A}_i &= \left\langle \left( \frac{(1+a_i)^\lambda - (1-a_i)^\lambda}{(1+a_i)^\lambda + (1-a_i)^\lambda}, \frac{(1+b_i)^\lambda - (1-b_i)^\lambda}{(1+b_i)^\lambda + (1-b_i)^\lambda}, \frac{(1+c_i)^\lambda - (1-c_i)^\lambda}{(1+c_i)^\lambda + (1-c_i)^\lambda}, \frac{(1+d_i)^\lambda - (1-d_i)^\lambda}{(1+d_i)^\lambda + (1-d_i)^\lambda}; \right. \right. \\
 &\quad \left. \left. \frac{(1+\rho_{A_i}^1)^\lambda - (1-\rho_{A_i}^1)^\lambda}{(1+\rho_{A_i}^1)^\lambda + (1-\rho_{A_i}^1)^\lambda}, \frac{(1+\rho_{A_i}^2)^\lambda - (1-\rho_{A_i}^2)^\lambda}{(1+\rho_{A_i}^2)^\lambda + (1-\rho_{A_i}^2)^\lambda}, \dots, \frac{(1+\rho_{A_i}^P)^\lambda - (1-\rho_{A_i}^P)^\lambda}{(1+\rho_{A_i}^P)^\lambda + (1-\rho_{A_i}^P)^\lambda}, \right. \right. \\
 &\quad \left. \left. \left( \frac{2(\delta_{A_i}^1)^\lambda}{(2-\delta_{A_i}^1)^\lambda + (\delta_{A_i}^1)^\lambda}, \frac{2(\delta_{A_i}^2)^\lambda}{(2-\delta_{A_i}^2)^\lambda + (\delta_{A_i}^2)^\lambda}, \dots, \frac{2(\delta_{A_i}^P)^\lambda}{(2-\delta_{A_i}^P)^\lambda + (\delta_{A_i}^P)^\lambda} \right) \right\rangle, \\
 \text{(iv)} \quad \dot{A}_i^\lambda &= \left\langle \left( \frac{2(a_i)^\lambda}{(2-a_i)^\lambda + (a_i)^\lambda}, \frac{2(b_i)^\lambda}{(2-b_i)^\lambda + (b_i)^\lambda}, \frac{2(c_i)^\lambda}{(2-c_i)^\lambda + (c_i)^\lambda}, \frac{2(d_i)^\lambda}{(2-d_i)^\lambda + (d_i)^\lambda}; \right. \right. \\
 &\quad \left. \left. \frac{2(\rho_{A_i}^1)^\lambda}{(2-\rho_{A_i}^1)^\lambda + (\rho_{A_i}^1)^\lambda}, \frac{2(\rho_{A_i}^2)^\lambda}{(2-\rho_{A_i}^2)^\lambda + (\rho_{A_i}^2)^\lambda}, \dots, \frac{2(\rho_{A_i}^P)^\lambda}{(2-\rho_{A_i}^P)^\lambda + (\rho_{A_i}^P)^\lambda}, \right. \right. \\
 &\quad \left. \left. \left( \frac{(1+\delta_{A_i}^1)^\lambda - (1-\delta_{A_i}^1)^\lambda}{(1+\delta_{A_i}^1)^\lambda + (1-\delta_{A_i}^1)^\lambda}, \frac{(1+\delta_{A_i}^2)^\lambda - (1-\delta_{A_i}^2)^\lambda}{(1+\delta_{A_i}^2)^\lambda + (1-\delta_{A_i}^2)^\lambda}, \dots, \frac{(1+\delta_{A_i}^P)^\lambda - (1-\delta_{A_i}^P)^\lambda}{(1+\delta_{A_i}^P)^\lambda + (1-\delta_{A_i}^P)^\lambda} \right) \right\rangle, \\
 \text{(3) for } g_3(t) &= \log\left(\frac{\gamma+(1-\gamma)t}{t}\right), \quad \gamma \in (0, +\infty), \\
 \text{(i)} \quad \dot{A}_1 \oplus \dot{A}_2 &= \left\langle \left( \frac{a_1+a_2-a_1a_2-(1-\gamma)a_1a_2}{1-(1-\gamma)a_1a_2}, \frac{b_1+b_2-b_1b_2-(1-\gamma)b_1b_2}{1-(1-\gamma)b_1b_2}, \right. \right. \\
 &\quad \left. \left. \frac{c_1+c_2-c_1c_2-(1-\gamma)c_1c_2}{1-(1-\gamma)c_1c_2}, \frac{d_1+d_2-d_1d_2-(1-\gamma)d_1d_2}{1-(1-\gamma)d_1d_2}; \right. \right. \\
 &\quad \left. \left. \frac{\rho_{A_1}^1+\rho_{A_2}^1-\rho_{A_1}^1\rho_{A_2}^1-(1-\gamma)\rho_{A_1}^1\rho_{A_2}^1}{1-(1-\gamma)\rho_{A_1}^1\rho_{A_2}^1}, \frac{\rho_{A_1}^2+\rho_{A_2}^2-\rho_{A_1}^2\rho_{A_2}^2-(1-\gamma)\rho_{A_1}^2\rho_{A_2}^2}{1-(1-\gamma)\rho_{A_1}^2\rho_{A_2}^2}, \dots, \right. \right. \\
 &\quad \left. \left. \frac{\rho_{A_1}^P+\rho_{A_2}^P-\rho_{A_1}^P\rho_{A_2}^P-(1-\gamma)\rho_{A_1}^P\rho_{A_2}^P}{1-(1-\gamma)\rho_{A_1}^P\rho_{A_2}^P}, \left( \frac{\delta_{A_1}^1\delta_{A_2}^1}{\gamma+(1-\gamma)(\delta_{A_1}^1+\delta_{A_2}^1-\delta_{A_1}^1\delta_{A_2}^1)}, \right. \right. \\
 &\quad \left. \left. \frac{\delta_{A_1}^2\delta_{A_2}^2}{\gamma+(1-\gamma)(\delta_{A_1}^2+\delta_{A_2}^2-\delta_{A_1}^2\delta_{A_2}^2)}, \dots, \frac{\delta_{A_1}^P\delta_{A_2}^P}{\gamma+(1-\gamma)(\delta_{A_1}^P+\delta_{A_2}^P-\delta_{A_1}^P\delta_{A_2}^P)} \right) \right\rangle, \\
 \text{(ii)} \quad \dot{A}_1 \otimes \dot{A}_2 &= \left\langle \left( \frac{a_1a_2}{\gamma+(1-\gamma)(a_1+a_2-a_1a_2)}, \frac{b_1b_2}{\gamma+(1-\gamma)(b_1+b_2-b_1b_2)}, \right. \right. \\
 &\quad \left. \left. \frac{c_1c_2}{\gamma+(1-\gamma)(c_1+c_2-c_1c_2)}, \frac{d_1d_2}{\gamma+(1-\gamma)(d_1+d_2-d_1d_2)}; \right. \right. \\
 &\quad \left. \left. \frac{\rho_{A_1}^1\rho_{A_2}^1}{\gamma+(1-\gamma)(\rho_{A_1}^1+\rho_{A_2}^1-\rho_{A_1}^1\rho_{A_2}^1)}, \frac{\rho_{A_1}^2\rho_{A_2}^2}{\gamma+(1-\gamma)(\rho_{A_1}^2+\rho_{A_2}^2-\rho_{A_1}^2\rho_{A_2}^2)}, \dots, \right. \right. \\
 &\quad \left. \left. \frac{\rho_{A_1}^P\rho_{A_2}^P}{\gamma+(1-\gamma)(\rho_{A_1}^P+\rho_{A_2}^P-\rho_{A_1}^P\rho_{A_2}^P)}, \left( \frac{\delta_{A_1}^1+\delta_{A_2}^1-\delta_{A_1}^1\delta_{A_2}^1-(1-\gamma)\delta_{A_1}^1\delta_{A_2}^1}{1-(1-\gamma)\delta_{A_1}^1\delta_{A_2}^1}, \right. \right. \\
 &\quad \left. \left. \frac{\delta_{A_1}^2+\delta_{A_2}^2-\delta_{A_1}^2\delta_{A_2}^2-(1-\gamma)\delta_{A_1}^2\delta_{A_2}^2}{1-(1-\gamma)\delta_{A_1}^2\delta_{A_2}^2}, \dots, \frac{\delta_{A_1}^P+\delta_{A_2}^P-\delta_{A_1}^P\delta_{A_2}^P-(1-\gamma)\delta_{A_1}^P\delta_{A_2}^P}{1-(1-\gamma)\delta_{A_1}^P\delta_{A_2}^P} \right) \right\rangle, \\
 \text{(iii)} \quad \lambda \dot{A}_i &= \left\langle \left( \frac{(1+(\gamma-1)a_i)^\lambda - (1-a_i)^\lambda}{(1+(\gamma-1)a_i)^\lambda + (\gamma-1)(1-a_i)^\lambda}, \frac{(1+(\gamma-1)b_i)^\lambda - (1-b_i)^\lambda}{(1+(\gamma-1)b_i)^\lambda + (\gamma-1)(1-b_i)^\lambda}, \right. \right. \\
 &\quad \left. \left. \frac{(1+(\gamma-1)c_i)^\lambda - (1-c_i)^\lambda}{(1+(\gamma-1)c_i)^\lambda + (\gamma-1)(1-c_i)^\lambda}, \frac{(1+(\gamma-1)d_i)^\lambda - (1-d_i)^\lambda}{(1+(\gamma-1)d_i)^\lambda + (\gamma-1)(1-d_i)^\lambda}; \right. \right. \\
 &\quad \left. \left. \frac{(1+(\gamma-1)\rho_{A_i}^1)^\lambda - (1-\rho_{A_i}^1)^\lambda}{(1+(\gamma-1)\rho_{A_i}^1)^\lambda + (\gamma-1)(1-\rho_{A_i}^1)^\lambda}, \frac{(1+(\gamma-1)\rho_{A_i}^2)^\lambda - (1-\rho_{A_i}^2)^\lambda}{(1+(\gamma-1)\rho_{A_i}^2)^\lambda + (\gamma-1)(1-\rho_{A_i}^2)^\lambda}, \dots, \right. \right. \\
 &\quad \left. \left. \frac{(1+(\gamma-1)\rho_{A_i}^P)^\lambda - (1-\rho_{A_i}^P)^\lambda}{(1+(\gamma-1)\rho_{A_i}^P)^\lambda + (\gamma-1)(1-\rho_{A_i}^P)^\lambda} \right\rangle,
 \end{aligned}$$

$$\begin{aligned}
 & \left( \frac{\gamma(\delta_{A_i}^1)^\lambda}{(1+(\gamma-1)(1-\delta_{A_i}^1))^\lambda+(\gamma-1)(\delta_{A_i}^1)^\lambda}, \frac{\gamma(\delta_{A_i}^2)^\lambda}{(1+(\gamma-1)(1-\delta_{A_i}^2))^\lambda+(\gamma-1)(\delta_{A_i}^2)^\lambda}, \dots, \right. \\
 & \left. \frac{\gamma(\delta_{A_i}^P)^\lambda}{(1+(\gamma-1)(1-\delta_{A_i}^P))^\lambda+(\gamma-1)(\delta_{A_i}^P)^\lambda} \right), \lambda > 0, \\
 \text{(iv) } \dot{A}_i^\lambda = & \left\langle \left( \frac{\gamma(a_i)^\lambda}{(1+(\gamma-1)(1-a_i))^\lambda+(\gamma-1)(a_i)^\lambda}, \frac{\gamma(b_i)^\lambda}{(1+(\gamma-1)(1-b_i))^\lambda+(\gamma-1)(b_i)^\lambda}, \right. \right. \\
 & \left. \frac{\gamma(c_i)^\lambda}{(1+(\gamma-1)(1-c_i))^\lambda+(\gamma-1)(c_i)^\lambda}, \frac{\gamma(d_i)^\lambda}{(1+(\gamma-1)(1-d_i))^\lambda+(\gamma-1)(d_i)^\lambda}; \right. \\
 & \left. \frac{\gamma(\rho_{A_i}^1)^\lambda}{(1+(\gamma-1)(1-\rho_{A_i}^1))^\lambda+(\gamma-1)(\rho_{A_i}^1)^\lambda}, \frac{\gamma(\rho_{A_i}^2)^\lambda}{(1+(\gamma-1)(1-\rho_{A_i}^2))^\lambda+(\gamma-1)(\rho_{A_i}^2)^\lambda}, \dots, \right. \\
 & \left. \frac{\gamma(\rho_{A_i}^P)^\lambda}{(1+(\gamma-1)(1-\rho_{A_i}^P))^\lambda+(\gamma-1)(\rho_{A_i}^P)^\lambda}, \left( \frac{(1+(\gamma-1)\rho_{A_i}^1)^\lambda-(1-\delta_{A_i}^1)^\lambda}{(1+(\gamma-1)\delta_{A_i}^1)^\lambda+(\gamma-1)(1-\delta_{A_i}^1)^\lambda}, \right. \right. \\
 & \left. \left. \frac{(1+(\gamma-1)\delta_{A_i}^2)^\lambda-(1-\delta_{A_i}^2)^\lambda}{(1+(\gamma-1)\delta_{A_i}^2)^\lambda+(\gamma-1)(1-\delta_{A_i}^2)^\lambda}, \dots, \frac{(1+(\gamma-1)\delta_{A_i}^P)^\lambda-(1-\delta_{A_i}^P)^\lambda}{(1+(\gamma-1)\delta_{A_i}^P)^\lambda+(\gamma-1)(1-\delta_{A_i}^P)^\lambda} \right) \right), \lambda > 0, \\
 \text{(4) for } g_4(t) = & \ln\left(\frac{\gamma-1}{\gamma^t-1}\right), \gamma \in (1, \infty), \\
 \text{(i) } \dot{A}_1 \oplus \dot{A}_2 = & \left\langle \left( 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-a_1}-1)(\gamma^{1-a_2}-1)}{\gamma-1} \right), 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-b_1}-1)(\gamma^{1-b_2}-1)}{\gamma-1} \right), \right. \right. \\
 & \left. \left. 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-c_1}-1)(\gamma^{1-c_2}-1)}{\gamma-1} \right), 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-d_1}-1)(\gamma^{1-d_2}-1)}{\gamma-1} \right); \right. \right. \\
 & \left. \left. 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-\rho_{A_1}^1}-1)(\gamma^{1-\rho_{A_2}^1}-1)}{\gamma-1} \right), \right. \right. \\
 & \left. \left. 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-\rho_{A_1}^2}-1)(\gamma^{1-\rho_{A_2}^2}-1)}{\gamma-1} \right), \right. \right. \\
 & \left. \left. \dots, 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-\rho_{A_1}^P}-1)(\gamma^{1-\rho_{A_2}^P}-1)}{\gamma-1} \right), \right. \right. \\
 & \left. \left. \log_\gamma \left( 1 + \frac{(\gamma^{\delta_{A_1}^2}-1)(\gamma^{\delta_{A_2}^2}-1)}{\gamma-1} \right), \dots, \log_\gamma \left( 1 + \frac{(\gamma^{\delta_{A_1}^P}-1)(\gamma^{\delta_{A_2}^P}-1)}{\gamma-1} \right) \right) \right), \gamma > 1, \\
 \text{(ii) } \dot{A}_1 \otimes \dot{A}_2 = & \left\langle \left( \log_\gamma \left( 1 + \frac{(\gamma^{a_1}-1)(\gamma^{a_2}-1)}{\gamma-1} \right), \log_\gamma \left( 1 + \frac{(\gamma^{b_1}-1)(\gamma^{b_2}-1)}{\gamma-1} \right), \right. \right. \\
 & \left. \left. \log_\gamma \left( 1 + \frac{(\gamma^{c_1}-1)(\gamma^{c_2}-1)}{\gamma-1} \right), \log_\gamma \left( 1 + \frac{(\gamma^{d_1}-1)(\gamma^{d_2}-1)}{\gamma-1} \right); \right. \right. \\
 & \left. \left. \log_\gamma \left( 1 + \frac{(\gamma^{\rho_{A_1}^1}-1)(\gamma^{\rho_{A_2}^1}-1)}{\gamma-1} \right), \log_\gamma \left( 1 + \frac{(\gamma^{\rho_{A_1}^2}-1)(\gamma^{\rho_{A_2}^2}-1)}{\gamma-1} \right), \right. \right. \\
 & \left. \left. \dots, \log_\gamma \left( 1 + \frac{(\gamma^{\rho_{A_1}^P}-1)(\gamma^{\rho_{A_2}^P}-1)}{\gamma-1} \right), \right. \right. \\
 & \left. \left. (1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-\delta_{A_1}^1}-1)(\gamma^{1-\delta_{A_2}^1}-1)}{\gamma-1} \right), 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-\delta_{A_1}^2}-1)(\gamma^{1-\delta_{A_2}^2}-1)}{\gamma-1} \right), \right. \right. \\
 & \left. \left. \dots, 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-\delta_{A_1}^P}-1)(\gamma^{1-\delta_{A_2}^P}-1)}{\gamma-1} \right) \right) \right), \gamma > 1, \\
 \text{(iii) } \lambda \dot{A}_i = & \left\langle \left( 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-a_i}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right), 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-b_i}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right), \right. \right. \\
 & \left. \left. 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-c_i}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right), 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-d_i}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right); \right. \right. \\
 & \left. \left. 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-\rho_{A_i}^1}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right), 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-\rho_{A_i}^2}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right), \right. \right. \\
 & \left. \left. \dots, 1 - \log_\gamma \left( 1 + \frac{(\gamma^{1-\rho_{A_i}^P}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right), \left( \log_\gamma \left( 1 + \frac{(\gamma^{\delta_{A_i}^1}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right), \right. \right. \\
 & \left. \left. \log_\gamma \left( 1 + \frac{(\gamma^{\delta_{A_i}^2}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right), \dots, \log_\gamma \left( 1 + \frac{(\gamma^{\delta_{A_i}^P}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right) \right) \right), \lambda > 0, \gamma > 1, \\
 \text{(iv) } \dot{A}_i^\lambda = & \left\langle \left( \log_\gamma \left( 1 + \frac{(\gamma^{a_i}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right), \log_\gamma \left( 1 + \frac{(\gamma^{b_i}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right), \log_\gamma \left( 1 + \frac{(\gamma^{c_i}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right), \right. \right. \\
 & \left. \left. \log_\gamma \left( 1 + \frac{(\gamma^{d_i}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right); \log_\gamma \left( 1 + \frac{(\gamma^{\rho_{A_i}^1}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right), \log_\gamma \left( 1 + \frac{(\gamma^{\rho_{A_i}^2}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right), \right. \right. \\
 & \left. \left. \log_\gamma \left( 1 + \frac{(\gamma^{\rho_{A_i}^P}-1)^\lambda}{(\gamma-1)^{\lambda-1}} \right) \right) \right),
 \end{aligned}$$

$$\dots, \log_\gamma(1 + \frac{\gamma^{\rho_{\dot{A}_i}^P} - 1}{(\gamma - 1)^{\lambda - 1}}), (1 - \log_\gamma(1 + \frac{\gamma^{1 - \delta_{\dot{A}_i}^1} - 1}{(\gamma - 1)^{\lambda - 1}}), \\ 1 - \log_\gamma(1 + \frac{\gamma^{1 - \delta_{\dot{A}_i}^2} - 1}{(\gamma - 1)^{\lambda - 1}}), \dots, 1 - \log_\gamma(1 + \frac{\gamma^{1 - \delta_{\dot{A}_i}^P} - 1}{(\gamma - 1)^{\lambda - 1}})), \lambda > 0, \gamma > 1.$$

3. TIFM AGGREGATION OPERATORS BASED ON ARCHIMEDEAN T-CONORM AND T-NORM

**Definition 3.1.** Let  $\dot{A}_i = \langle (a_i, b_i, c_i, d_i); (\rho_{\dot{A}_i}^1, \rho_{\dot{A}_i}^2, \dots, \rho_{\dot{A}_i}^P), (\delta_{\dot{A}_i}^1, \delta_{\dot{A}_i}^2, \dots, \delta_{\dot{A}_i}^P) \rangle$  ( $i = 1, 2, \dots, n$ ) be a collection of TIFM-numbers and  $w = (w_1, w_2, \dots, w_n)^T$  be the weight vector of the TIFM-numbers  $\dot{A}_i$  ( $i = 1, 2, \dots, n$ ). Then *intuitionistic trapezoidal fuzzy multi weighted averaging operator* of  $\dot{A}_i$  ( $i = 1, 2, \dots, n$ ) based on Archimedean t-norm and Archimedean t-conorm, denoted by  $\text{TIFMA}(\dot{A}_1, \dot{A}_2, \dots, \dot{A}_n)$ , is defined as:

$$\text{TIFMA}(\dot{A}_1, \dot{A}_2, \dots, \dot{A}_n) = \bigoplus_{i=1}^n (w_i \dot{A}_i),$$

where  $w_i$  indicates the importance degree of  $\dot{A}_i$  ( $i = 1, 2, \dots, n$ ) such that  $w_i \in [0, 1]$  and  $\sum_{i=1}^n w_i = 1$ .

**Theorem 3.2.** Let  $\dot{A}_i = \langle (a_i, b_i, c_i, d_i); (\rho_{\dot{A}_i}^1, \rho_{\dot{A}_i}^2, \dots, \rho_{\dot{A}_i}^P), (\delta_{\dot{A}_i}^1, \delta_{\dot{A}_i}^2, \dots, \delta_{\dot{A}_i}^P) \rangle$  ( $i = 1, 2, \dots, n$ ) be a collection of TIFM-numbers,  $\dot{A} = \langle (a, b, c, d); (\rho_{\dot{A}}^1, \rho_{\dot{A}}^2, \dots, \rho_{\dot{A}}^P), (\delta_{\dot{A}}^1, \delta_{\dot{A}}^2, \dots, \delta_{\dot{A}}^P) \rangle$  be a TIFM-number and  $\lambda, \lambda_1, \lambda_2 > 0$ . The aggregated value by using the *ATS-IFWA* operator is also an IFV, and

(3.1)

$$\text{TIFMA}(\dot{A}_1, \dot{A}_2, \dots, \dot{A}_n) = \langle (h^{-1}(\sum_{i=1}^n w_i h(a_i)), h^{-1}(\sum_{i=1}^n w_i h(b_i)), h^{-1}(\sum_{i=1}^n w_i h(c_i)), h^{-1}(\sum_{i=1}^n w_i h(d_i))), \\ h^{-1}(\sum_{i=1}^n w_i h(\rho_{\dot{A}_i}^1)), h^{-1}(\sum_{i=1}^n w_i h(\rho_{\dot{A}_i}^2)), \dots, h^{-1}(\sum_{i=1}^n w_i h(\rho_{\dot{A}_i}^P)), \\ (g^{-1}(\sum_{i=1}^n w_i g(\delta_{\dot{A}_i}^1)), g^{-1}(\sum_{i=1}^n w_i g(\delta_{\dot{A}_i}^2)), \dots, g^{-1}(\sum_{i=1}^n w_i g(\delta_{\dot{A}_i}^P))) \rangle$$

which has been investigated by Beliaikov et al. [52], Xu and Yager [53], Wu and Cai [54] and next we give a further study.

**Definition 3.3.** Let  $\dot{A}_i = \langle (a_i, b_i, c_i, d_i); (\rho_{\dot{A}_i}^1, \rho_{\dot{A}_i}^2, \dots, \rho_{\dot{A}_i}^P), (\delta_{\dot{A}_i}^1, \delta_{\dot{A}_i}^2, \dots, \delta_{\dot{A}_i}^P) \rangle$  ( $i = 1, 2, \dots, n$ ) be a collection of TIFM-numbers and  $w = (w_1, w_2, \dots, w_n)^T$  weight vector of  $\dot{A}_i$  ( $i = 1, 2, \dots, n$ ) where  $w_i \geq 0$  ( $i = 1, 2, \dots, n$ ) and  $\sum_{i=1}^n w_i = 1$ . If

$$\text{TIFMA}(\dot{A}_1, \dot{A}_2, \dots, \dot{A}_n) = \bigoplus_{i=1}^n (w_i \dot{A}_i)$$

then TIFMA is called an *Archimedean t-conorm and t-norm based intuitionistic trapezoidal fuzzy multi weighted averaging (TIFMA) operator*.

**Theorem 3.4** ([55]). Let  $\dot{A}_i = \langle (a_i, b_i, c_i, d_i); (\rho_{\dot{A}_i}^1, \rho_{\dot{A}_i}^2, \dots, \rho_{\dot{A}_i}^P), (\delta_{\dot{A}_i}^1, \delta_{\dot{A}_i}^2, \dots, \delta_{\dot{A}_i}^P) \rangle$ ,

$$\dot{B}_i = \langle (\dot{a}_i, \dot{b}_i, \dot{c}_i, \dot{d}_i); (\rho_{\dot{B}_i}^1, \rho_{\dot{B}_i}^2, \dots, \rho_{\dot{B}_i}^P), (\delta_{\dot{B}_i}^1, \delta_{\dot{B}_i}^2, \dots, \delta_{\dot{B}_i}^P) \rangle \quad (i=1,2,\dots,n)$$

be two collection of TIFM-numbers. If

$$a_i \leq \dot{a}_i, b_i \leq \dot{b}_i, c_i \leq \dot{c}_i, d_i \leq \dot{d}_i,$$

$$\rho_{\dot{A}_i}^1 \leq \rho_{\dot{B}_i}^1, \rho_{\dot{A}_i}^2 \leq \rho_{\dot{B}_i}^2, \dots, \rho_{\dot{A}_i}^P \leq \rho_{\dot{B}_i}^P$$

and

$$\delta_{\dot{B}_i}^1 \leq \delta_{\dot{A}_i}^1, \delta_{\dot{B}_i}^2 \leq \delta_{\dot{A}_i}^2, \dots, \delta_{\dot{B}_i}^P \leq \delta_{\dot{A}_i}^P,$$

then

$$\text{TIFMA}(\dot{A}_1, \dot{A}_2, \dots, \dot{A}_n) \leq \text{TIFMA}(\dot{B}_1, \dot{B}_2, \dots, \dot{B}_n)$$

For any  $a, b \in \mathbb{R}$ , we will denote  $\max\{a, b\}$  and  $\min\{a, b\}$  by  $a \vee b$  and  $a \wedge b$  respectively.

**Theorem 3.5.** Let  $\dot{A}_i = \langle (a_i, b_i, c_i, d_i); (\rho_{\dot{A}_i}^1, \rho_{\dot{A}_i}^2, \dots, \rho_{\dot{A}_i}^P), (\delta_{\dot{A}_i}^1, \delta_{\dot{A}_i}^2, \dots, \delta_{\dot{A}_i}^P) \rangle$  ( $i = 1, 2, \dots, n$ ) be a collection of TFM-numbers. Let  $\dot{A}^-$  and  $\dot{A}^+$  be given by ( $i = 1, 2, \dots, n$ )

$$\dot{A}^- \leq \text{TIFMA}(\dot{A}_1, \dot{A}_2, \dots, \dot{A}_n) \leq \dot{A}^+,$$

where

$$\begin{aligned} & \langle (\bigvee_{i=1}^n a_i, \bigvee_{i=1}^n b_i, \bigvee_{i=1}^n c_i, \bigvee_{i=1}^n d_i); \\ & \quad (s(\rho_{\dot{A}_i}^1), s(\rho_{\dot{A}_i}^2), \dots, s(\rho_{\dot{A}_i}^P), (\tau(\delta_{\dot{B}_i}^1), \tau(\delta_{\dot{B}_i}^2), \dots, \tau(\delta_{\dot{B}_i}^P)), \\ & \langle (\bigvee_{i=1}^n a_i, \bigvee_{i=1}^n b_i, \bigvee_{i=1}^n c_i, \bigvee_{i=1}^n d_i); (\bigvee_{i=1}^n \rho_{\dot{A}_i}^1, \bigvee_{i=1}^n \rho_{\dot{A}_i}^2, \dots, \bigvee_{i=1}^n \rho_{\dot{A}_i}^P), \\ & \quad (\bigwedge_{i=1}^n \delta_{\dot{B}_i}^1, \bigwedge_{i=1}^n \delta_{\dot{B}_i}^2, \dots, \bigwedge_{i=1}^n \delta_{\dot{B}_i}^P) \rangle \end{aligned}$$

and

$$\begin{aligned} \dot{A}^- &= \langle (\bigwedge_{i=1}^n a_i, \bigwedge_{i=1}^n b_i, \bigwedge_{i=1}^n c_i, \bigwedge_{i=1}^n d_i); \\ & \quad \tau(\rho_{\dot{A}_i}^1), \tau(\rho_{\dot{A}_i}^2), \dots, \tau(\rho_{\dot{A}_i}^P), (s(\delta_{\dot{B}_i}^1), s(\delta_{\dot{B}_i}^2), \dots, s(\delta_{\dot{B}_i}^P)), \\ \dot{A}^+ &= \langle (\bigwedge_{i=1}^n a_i, \bigwedge_{i=1}^n b_i, \bigwedge_{i=1}^n c_i, \bigwedge_{i=1}^n d_i); (\bigwedge_{i=1}^n \rho_{\dot{A}_i}^1, \bigwedge_{i=1}^n \rho_{\dot{A}_i}^2, \dots, \bigwedge_{i=1}^n \rho_{\dot{A}_i}^P), \\ & \quad (\bigvee_{i=1}^n \delta_{\dot{B}_i}^1, \bigvee_{i=1}^n \delta_{\dot{B}_i}^2, \dots, \bigvee_{i=1}^n \delta_{\dot{B}_i}^P) \rangle. \end{aligned}$$

If we choose the additive generator  $g$  in different forms, then some specific intuitionistic trapezoidal fuzzy aggregation operators can be obtained (See [55]).

**Theorem 3.6.** Let  $\dot{A}_i = \langle (a_i, b_i, c_i, d_i); (\rho_{\dot{A}_i}^1, \rho_{\dot{A}_i}^2, \dots, \rho_{\dot{A}_i}^P), (\delta_{\dot{A}_i}^1, \delta_{\dot{A}_i}^2, \dots, \delta_{\dot{A}_i}^P) \rangle$  ( $i = 1, 2, \dots, n$ ) be a collection of TIFM-numbers and  $w = (w_1, w_2, \dots, w_n)^T$  weight vector of  $\dot{A}_i$  ( $i = 1, 2, \dots, n$ ) where  $w_i \geq 0$  ( $i = 1, 2, \dots, n$ ) and  $\sum_{i=1}^n w_i = 1$ . If  $r > 0$ , then

$$\text{TIFMA}(r\dot{A}_1, r\dot{A}_2, \dots, r\dot{A}_n) = r\text{TIFMA}(\dot{A}_1, \dot{A}_2, \dots, \dot{A}_n).$$

If the additive generator  $g$  is assigned different forms, then some specific intuitionistic fuzzy aggregation operators can be obtained (See [55]).

**Result 1** If  $g(t) = -\log t$ , then the TIFMA operator reduces to the following:

$$\begin{aligned} \text{TIFMA}(\dot{A}_1, \dot{A}_2, \dots, \dot{A}_n) &= \langle (\bigoplus_{i=1}^n (w_i a_i), \bigoplus_{i=1}^n (w_i b_i), \bigoplus_{i=1}^n (w_i c_i), \bigoplus_{i=1}^n (w_i d_i)); \\ &1 - \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^1)^{w_i}, 1 - \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^2)^{w_i}, \dots, 1 - \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^P)^{w_i}, \\ &1 - \prod_{i=1}^n (1 - \delta_{\dot{A}_i}^1)^{w_i}, 1 - \prod_{i=1}^n (1 - \delta_{\dot{A}_i}^2)^{w_i}, \dots, 1 - \prod_{i=1}^n (1 - \delta_{\dot{A}_i}^P)^{w_i} \rangle \\ &= \langle (1 - \prod_{i=1}^n (1 - a_i)^{w_i}, 1 - \prod_{i=1}^n (1 - b_i)^{w_i}, 1 - \prod_{i=1}^n (1 - c_i)^{w_i}, 1 - \prod_{i=1}^n (1 - d_i)^{w_i}); \\ &1 - \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^1)^{w_i}, 1 - \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^2)^{w_i}, \dots, 1 - \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^P)^{w_i}, \\ &\prod_{i=1}^n (\delta_{\dot{A}_i}^1)^{w_i}, \prod_{i=1}^n (\delta_{\dot{A}_i}^2)^{w_i}, \dots, \prod_{i=1}^n (\delta_{\dot{A}_i}^P)^{w_i} \rangle \end{aligned}$$

**Result 2** If  $g(t) = \log(\frac{2-t}{t})$ , then the TIFMA operator reduces to the following:

$$\begin{aligned} \text{TIFMA}(\dot{A}_1, \dot{A}_2, \dots, \dot{A}_n) &= \langle \left( \frac{\prod_{i=1}^n (1 + a_i)^{w_i} - \prod_{i=1}^n (1 - a_i)^{w_i}}{\prod_{i=1}^n (1 + a_i)^{w_i} + \prod_{i=1}^n (1 - a_i)^{w_i}}, \frac{\prod_{i=1}^n (1 + b_i)^{w_i} - \prod_{i=1}^n (1 - b_i)^{w_i}}{\prod_{i=1}^n (1 + b_i)^{w_i} + \prod_{i=1}^n (1 - b_i)^{w_i}}, \right. \\ &\left. \frac{\prod_{i=1}^n (1 + c_i)^{w_i} - \prod_{i=1}^n (1 - c_i)^{w_i}}{\prod_{i=1}^n (1 + c_i)^{w_i} + \prod_{i=1}^n (1 - c_i)^{w_i}}, \frac{\prod_{i=1}^n (1 + d_i)^{w_i} - \prod_{i=1}^n (1 - d_i)^{w_i}}{\prod_{i=1}^n (1 + d_i)^{w_i} + \prod_{i=1}^n (1 - d_i)^{w_i}} \right); \\ &\frac{\prod_{i=1}^n (1 + \rho_{\dot{A}_i}^1)^{w_i} - \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^1)^{w_i}}{\prod_{i=1}^n (1 + \rho_{\dot{A}_i}^1)^{w_i} + \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^1)^{w_i}}, \frac{\prod_{i=1}^n (1 + \rho_{\dot{A}_i}^2)^{w_i} - \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^2)^{w_i}}{\prod_{i=1}^n (1 + \rho_{\dot{A}_i}^2)^{w_i} + \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^2)^{w_i}}, \dots, \\ &\frac{\prod_{i=1}^n (1 + \rho_{\dot{A}_i}^P)^{w_i} - \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^P)^{w_i}}{\prod_{i=1}^n (1 + \rho_{\dot{A}_i}^P)^{w_i} + \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^P)^{w_i}}, \left( \frac{2 \prod_{i=1}^n (\delta_{\dot{A}_i}^1)^{w_i}}{\prod_{i=1}^n (2 - \delta_{\dot{A}_i}^1)^{w_i} + \prod_{i=1}^n (\delta_{\dot{A}_i}^1)^{w_i}}, \right. \\ &\left. \frac{2 \prod_{i=1}^n (\delta_{\dot{A}_i}^2)^{w_i}}{\prod_{i=1}^n (2 - \delta_{\dot{A}_i}^2)^{w_i} + \prod_{i=1}^n (\delta_{\dot{A}_i}^2)^{w_i}}, \dots, \frac{2 \prod_{i=1}^n (\delta_{\dot{A}_i}^P)^{w_i}}{\prod_{i=1}^n (2 - \delta_{\dot{A}_i}^P)^{w_i} + \prod_{i=1}^n (\delta_{\dot{A}_i}^P)^{w_i}} \right) \rangle \end{aligned}$$

which is called an *Einstein TIFMA operator*.

**Result 3** If  $g(t) = \log\left(\frac{\gamma+(1-\gamma)t}{t}\right)$ ,  $\gamma \in (0, +\infty)$ , then the TIFMA operator reduces to the following:

$$\begin{aligned} & \text{TIFMA}(\dot{A}_1, \dot{A}_2, \dots, \dot{A}_n) \\ &= \left\langle \left( \frac{\prod_{i=1}^n (1 + (\gamma - 1)a_i)^{w_i} - \prod_{i=1}^n (1 - a_i)^{w_i}}{\prod_{i=1}^n (1 + (\gamma - 1)a_i)^{w_i} + (\gamma - 1) \prod_{i=1}^n (1 - a_i)^{w_i}}, \frac{\prod_{i=1}^n (1 + (\gamma - 1)b_i)^{w_i} - \prod_{i=1}^n (1 - b_i)^{w_i}}{\prod_{i=1}^n (1 + (\gamma - 1)b_i)^{w_i} + (\gamma - 1) \prod_{i=1}^n (1 - b_i)^{w_i}}, \right. \right. \\ & \quad \left. \frac{\prod_{i=1}^n (1 + (\gamma - 1)c_i)^{w_i} - \prod_{i=1}^n (1 - c_i)^{w_i}}{\prod_{i=1}^n (1 + (\gamma - 1)c_i)^{w_i} + (\gamma - 1) \prod_{i=1}^n (1 - c_i)^{w_i}}, \frac{\prod_{i=1}^n (1 + (\gamma - 1)d_i)^{w_i} - \prod_{i=1}^n (1 - d_i)^{w_i}}{\prod_{i=1}^n (1 + (\gamma - 1)d_i)^{w_i} + (\gamma - 1) \prod_{i=1}^n (1 - d_i)^{w_i}}, \dots, \right. \\ & \quad \left. \frac{\prod_{i=1}^n (1 + (\gamma - 1)\rho_{\dot{A}_i}^1)^{w_i} - \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^1)^{w_i}}{\prod_{i=1}^n (1 + (\gamma - 1)\rho_{\dot{A}_i}^1)^{w_i} + (\gamma - 1) \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^1)^{w_i}}, \frac{\prod_{i=1}^n (1 + (\gamma - 1)\rho_{\dot{A}_i}^2)^{w_i} - \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^2)^{w_i}}{\prod_{i=1}^n (1 + (\gamma - 1)\rho_{\dot{A}_i}^2)^{w_i} + (\gamma - 1) \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^2)^{w_i}}, \dots, \right. \\ & \quad \left. \frac{\prod_{i=1}^n (1 + (\gamma - 1)\rho_{\dot{A}_i}^P)^{w_i} - \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^P)^{w_i}}{\prod_{i=1}^n (1 + (\gamma - 1)\rho_{\dot{A}_i}^P)^{w_i} + (\gamma - 1) \prod_{i=1}^n (1 - \rho_{\dot{A}_i}^P)^{w_i}}, \frac{\gamma \prod_{i=1}^n (\delta_{\dot{A}_i}^1)^{w_i}}{\prod_{i=1}^n (1 + (\gamma - 1)(1 - \delta_{\dot{A}_i}^1))^{w_i} + (\gamma - 1) \prod_{i=1}^n (\delta_{\dot{A}_i}^1)^{w_i}}, \right. \\ & \quad \left. \frac{\gamma \prod_{i=1}^n (\delta_{\dot{A}_i}^2)^{w_i}}{\prod_{i=1}^n (1 + (\gamma - 1)(1 - \delta_{\dot{A}_i}^2))^{w_i} + (\gamma - 1) \prod_{i=1}^n (\delta_{\dot{A}_i}^2)^{w_i}}, \dots, \frac{\gamma \prod_{i=1}^n (\delta_{\dot{A}_i}^P)^{w_i}}{\prod_{i=1}^n (1 + (\gamma - 1)(1 - \delta_{\dot{A}_i}^P))^{w_i} + (\gamma - 1) \prod_{i=1}^n (\delta_{\dot{A}_i}^P)^{w_i}} \right) \end{aligned}$$

which is called a *Hammer TIFMA operator*.

**Result 4** If  $g(t) = \log\left(\frac{\gamma-1}{\gamma t-1}\right)$ ,  $\gamma \in (1, +\infty)$ , then the TIFMA operator reduces to the following:

$$\begin{aligned} & \text{ITMA}(A_1, A_2, \dots, A_n) = \left\langle \left( 1 - \log_\gamma \left( 1 + \frac{\prod_{i=1}^n (\gamma^{1-a_i} - 1)^{w_i}}{\gamma - 1} \right), 1 - \log_\gamma \left( 1 + \frac{\prod_{i=1}^n (\gamma^{1-b_i} - 1)^{w_i}}{\gamma - 1} \right), \right. \right. \\ & \quad \left. \left( 1 - \log_\gamma \left( 1 + \frac{\prod_{i=1}^n (\gamma^{1-c_i} - 1)^{w_i}}{\gamma - 1} \right), 1 - \log_\gamma \left( 1 + \frac{\prod_{i=1}^n (\gamma^{1-d_i} - 1)^{w_i}}{\gamma - 1} \right) \right); \right. \\ & \quad \left. \left( 1 - \log_\gamma \left( 1 + \frac{\prod_{i=1}^n (\gamma^{1-\rho_{A_i}^1} - 1)^{w_i}}{\gamma - 1} \right), 1 - \log_\gamma \left( 1 + \frac{\prod_{i=1}^n (\gamma^{1-\rho_{A_i}^2} - 1)^{w_i}}{\gamma - 1} \right), \dots, \right. \right. \\ & \quad \left. \left( 1 - \log_\gamma \left( 1 + \frac{\prod_{i=1}^n (\gamma^{1-\rho_{A_i}^P} - 1)^{w_i}}{\gamma - 1} \right), \ln \gamma \left( 1 + \frac{\prod_{i=1}^n (\gamma^{\delta_{A_i}^1} - 1)^{w_i}}{\gamma - 1} \right), \right. \right. \\ & \quad \left. \left. \ln \gamma \left( 1 + \frac{\prod_{i=1}^n (\gamma^{\delta_{A_i}^2} - 1)^{w_i}}{\gamma - 1} \right), \dots, \ln \gamma \left( 1 + \frac{\prod_{i=1}^n (\gamma^{\delta_{A_i}^P} - 1)^{w_i}}{\gamma - 1} \right) \right) \right) \end{aligned}$$

which is called a *Frank TIFMA operator*.

We give following definition inspired by geometric mean.

**Definition 3.7.** Let  $\dot{A}_i = \langle (a_i, b_i, c_i, d_i); (\rho_{\dot{A}_i}^1, \rho_{\dot{A}_i}^2, \dots, \rho_{\dot{A}_i}^P), (\delta_{\dot{A}_i}^1, \delta_{\dot{A}_i}^2, \dots, \delta_{\dot{A}_i}^P) \rangle$  ( $i = 1, 2, \dots, n$ ) be a collection of TIFM-numbers and  $w = (w_1, w_2, \dots, w_n)^T$  weight vector of  $\dot{A}_i$  ( $i = 1, 2, \dots, n$ ) such that  $\sum_{i=1}^n w_i = 1$ . If

$$\text{TIFMG}(\dot{A}_1, \dot{A}_2, \dots, \dot{A}_n) = \bigotimes_{i=1}^n (\dot{A}_i)^{w_i},$$

then TIFMG is called an *Archimedean t-cornorm* and *t-norm* based intuitionistic trapezoidal fuzzy multi geometric (TIFMG) operator.

**Theorem 3.8.** Let  $\dot{A}_i = \langle (a_i, b_i, c_i, d_i); (\rho_{\dot{A}_i}^1, \rho_{\dot{A}_i}^2, \dots, \rho_{\dot{A}_i}^P), (\delta_{\dot{A}_i}^1, \delta_{\dot{A}_i}^2, \dots, \delta_{\dot{A}_i}^P) \rangle$  ( $i = 1, 2$ ) be a collection of TIFM-numbers,  $\dot{A} = \langle (a, b, c, d); (\rho_{\dot{A}}^1, \rho_{\dot{A}}^2, \dots, \rho_{\dot{A}}^P), (\delta_{\dot{A}}^1, \delta_{\dot{A}}^2, \dots, \delta_{\dot{A}}^P) \rangle$  be a TIFM-number and  $\lambda, \lambda_1, \lambda_2 > 0$ . The aggregated value by using the TIFMA operator is also an IFV, and The aggregated value by using the TIFMG operator is also a TIFM-number and

$$\begin{aligned} \text{TIFMG}(\dot{A}_1, \dot{A}_2, \dots, \dot{A}_n) &= \bigotimes_{i=1}^n (\dot{A}_i)^{w_i} \\ &= \langle (g^{-1}(\sum_{i=1}^n w_i g(a_i)), g^{-1}(\sum_{i=1}^n w_i g(b_i)), g^{-1}(\sum_{i=1}^n w_i g(c_i)), g^{-1}(\sum_{i=1}^n w_i g(d_i))); \\ &\quad g^{-1}(\sum_{i=1}^n w_i g(\rho_{\dot{A}_i}^1)), g^{-1}(\sum_{i=1}^n w_i g(\rho_{\dot{A}_i}^2)), \dots, g^{-1}(\sum_{i=1}^n w_i g(\rho_{\dot{A}_i}^P)), \\ &\quad h^{-1}(\sum_{i=1}^n w_i h(\delta_{\dot{A}_i}^1)), h^{-1}(\sum_{i=1}^n w_i h(\delta_{\dot{A}_i}^2)), \dots, h^{-1}(\sum_{i=1}^n w_i h(\delta_{\dot{A}_i}^P))) \rangle. \end{aligned}$$

*Proof.* The operator can be easily proven similar to the TIFMA operator. This is why no need to prove TIFMG operator here again.  $\square$

If the additive generator  $g$  is assigned different forms then following intuitionistic trapezoidal fuzzy aggregation operators can be obtained (See [55]).

**Result 1** If  $g(t) = -\log t$ , then the TIFMG operator reduces to following:

$$\begin{aligned} \text{TIFMG}(\dot{A}_1, \dot{A}_2, \dots, \dot{A}_n) &= \langle (\prod_{i=1}^n (a_i)^{w_i}, \prod_{i=1}^n (b_i)^{w_i}, \prod_{i=1}^n (c_i)^{w_i}, \prod_{i=1}^n (d_i)^{w_i}); \\ &\quad \prod_{i=1}^n (\rho_{\dot{A}_i}^1)^{w_i}, \prod_{i=1}^n (\rho_{\dot{A}_i}^2)^{w_i}, \dots, \prod_{i=1}^n (\rho_{\dot{A}_i}^P)^{w_i}, \\ &\quad 1 - \prod_{i=1}^n (1 - \delta_{\dot{A}_i}^1)^{w_i}, 1 - \prod_{i=1}^n (1 - \delta_{\dot{A}_i}^2)^{w_i}, \dots, 1 - \prod_{i=1}^n (1 - \delta_{\dot{A}_i}^P)^{w_i} \rangle. \end{aligned}$$

Similarly, some new operators based on  $\log(\frac{2-t}{t})$ ,  $g(t) = \log(\frac{\gamma+(1-\gamma)t}{t})$ ,  $\gamma \in (0, +\infty)$ ,  $g(t) = \log(\frac{\gamma-1}{\gamma t-1})$ ,  $\gamma \in (1, +\infty)$  can be given.

4. AN APPROACH TO TRAPEZOIDAL FUZZY MULTI-ATTRIBUTE DECISION MAKING

For a decision making problem under intuitionistic trapezoidal fuzzy multi environment, let  $\dot{X}$  be set of alternatives and  $U$  be set of attributes. In order to evaluate the performance of the alternative  $\dot{X}_i$  with regards to  $u_j$ , an expert is required to provide information that the alternative  $\dot{X}_i$  satisfies the attribute  $u_j$ . This information can be expressed by  $\alpha_{ij} = \langle (a_{ij}, b_{ij}, c_{ij}, d_{ij}); (\rho_{\alpha_{ij}}^1, \rho_{\alpha_{ij}}^2, \dots, \rho_{\alpha_{ij}}^P), (\delta_{\alpha_{ij}}^1, \delta_{\alpha_{ij}}^2, \dots, \delta_{\alpha_{ij}}^P) \rangle$  which denote the degrees that the alternative  $\dot{X}_i$  satisfies the attribute  $u_j$  with the condition that  $0 < \rho_{\alpha_{ij}}^1, \rho_{\alpha_{ij}}^2, \dots, \rho_{\alpha_{ij}}^P < 1$  and  $0 < \delta_{\alpha_{ij}}^1, \delta_{\alpha_{ij}}^2, \dots, \delta_{\alpha_{ij}}^P < 1$ . When all the performances of the alternatives are provided, the intuitionistic trapezoidal fuzzy multi decision matrix  $C = (\alpha_{ij})_{m \times n}$  can be constructed. To obtain the ranking of the alternatives, the steps can be given as:

**Step 1** Construct intuitionistic trapezoidal fuzzy multi decision matrix  $C = (\alpha_{ij})_{m \times n}$  based on Table 1:

TABLE 1. TIFM-numbers for linguistic terms [36]

Linguistic terms	Linguistic values of TIFM-numbers
Absolutely low(AL)	$\langle (0.01, 0.05, 0.10, 0.15); (0.1, 0.3, 0.2, 0.4), (0.5, 0.3, 0.4, 0.7) \rangle$
Very Very Low(VVL)	$\langle (0.05, 0.10, 0.15, 0.20); (0.1, 0.3, 0.5, 0.4), (0.2, 0.3, 0.4, 0.1) \rangle$
Very Low(VL)	$\langle (0.15, 0.20, 0.25, 0.40); (0.2, 0.3, 0.5, 0.4), (0.2, 0.3, 0.4, 0.5) \rangle$
Low(L)	$\langle (0.10, 0.20, 0.25, 0.30); (0.1, 0.3, 0.3, 0.4), (0.5, 0.3, 0.4, 0.1) \rangle$
Fairly low(FL)	$\langle (0.15, 0.20, 0.25, 0.30); (0.1, 0.5, 0.2, 0.4), (0.2, 0.7, 0.4, 0.3) \rangle$
Medium(M)	$\langle (0.25, 0.30, 0.35, 0.40); (0.1, 0.3, 0.7, 0.4), (0.6, 0.3, 0.4, 0.7) \rangle$
Fairly high(FH)	$\langle (0.25, 0.35, 0.40, 0.45); (0.1, 0.3, 0.2, 0.5), (0.2, 0.5, 0.4, 0.1) \rangle$
High(H)	$\langle (0.40, 0.45, 0.50, 0.55); (0.1, 0.3, 0.2, 0.9), (0.2, 0.5, 0.3, 0.1) \rangle$
Very High(VH)	$\langle (0.45, 0.55, 0.65, 0.75); (0.1, 0.6, 0.2, 0.4), (0.6, 0.3, 0.4, 0.1) \rangle$
Very Very High(VVH)	$\langle (0.50, 0.65, 0.70, 0.80); (0.1, 0.3, 0.2, 0.4), (0.5, 0.3, 0.4, 0.1) \rangle$
Absolutely high(AH)	$\langle (0.70, 0.85, 0.90, 1.00); (0.1, 0.3, 0.3, 0.4), (0.1, 0.3, 0.4, 0.5) \rangle$

**Step 2** Transform the intuitionistic trapezoidal fuzzy multi decision matrix  $C = (c_{ij})_{m \times n}$  into the normalized intuitionistic trapezoidal fuzzy multi decision matrix  $K = (k_{ij})_{m \times n}$ , where

$$k_{ij} = \begin{cases} \langle (a_{ij}, b_{ij}, c_{ij}, d_{ij}); (\rho_{\alpha_{ij}}^1, \rho_{\alpha_{ij}}^2, \dots, \rho_{\alpha_{ij}}^P), (\delta_{\alpha_{ij}}^1, \delta_{\alpha_{ij}}^2, \dots, \delta_{\alpha_{ij}}^P) \rangle, & \text{for benefit attribute } u_j; \\ \langle (a_{ij}, b_{ij}, c_{ij}, d_{ij}); (\delta_{\alpha_{ij}}^1, \delta_{\alpha_{ij}}^2, \dots, \delta_{\alpha_{ij}}^P), (\rho_{\alpha_{ij}}^1, \rho_{\alpha_{ij}}^2, \dots, \rho_{\alpha_{ij}}^P) \rangle, & \text{for cost attribute } u_j. \end{cases}$$

**Step 3** Insert the weighted vector  $w = (w_1, w_2, \dots, w_n)$  such that  $w_j \geq 0$   $j = (1, 2, \dots, n)$  and  $\sum_{j=1}^n w_j = 1$ . In here,  $w_j$  is weighted of  $u_j$   $j = (1, 2, \dots, n)$ ;

**Step 4** Obtain values  $b_i$  ( $i = 1, 2, \dots, m$ ) for  $\dot{X}_i$  by the operators as:  
 $b_i = \text{TIFMA}(\alpha_{i1}, \alpha_{i2}, \dots, \alpha_{in})$  or  $b_i = \text{TIFMG}(\alpha_{i1}, \alpha_{i2}, \dots, \alpha_{in})$   $i = 1, 2, \dots, m$



**Step 5** Calculate the score values  $\bar{S}_p(b_i)$ (p=1,2,3,4)  $i = (1, 2, \dots, m)$  and rank the alternatives. The alternative with the highest score value is the best alternative.

**Illustrative Example**

**Example 4.1.** Assume that  $\dot{X} = \{\dot{x}_1, \dot{x}_2, \dot{x}_3, \dot{x}_4, \dot{x}_5\}$  be production of five different solar power plants depends on seasonal conditions. Then, we examine their performance according to the criterias set  $U = \{u_1 = \text{wind speed}, u_2 = \text{cloudiness}, u_3 = \text{amount of solar radiation on the earth}, u_4 = \text{surface angle of the panels}\}$ . In here, 4 different membership function was used for different season (spring, summer, autumn, winter). The solutions were given as follows:

**Step 1** We constructed intuitionistic trapezoidal fuzzy multi decision matrix  $C = (\alpha_{ij})_{m \times n}$  based on Table 1 as:

TABLE 2. The intuitionistic trapezoidal fuzzy multi decision matrix  $C = (\alpha_{ij})_{m \times n}$

	$u_1$
$\dot{X}_1$	$\langle(0.01, 0.05, 0.10, 0.15); (0.1, 0.3, 0.2, 0.4), (0.5, 0.3, 0.4, 0.7)\rangle$
$\dot{X}_2$	$\langle(0.15, 0.20, 0.25, 0.30); (0.1, 0.5, 0.2, 0.4), (0.2, 0.7, 0.4, 0.3)\rangle$
$\dot{X}_3$	$\langle(0.25, 0.30, 0.35, 0.40); (0.1, 0.3, 0.7, 0.4), (0.6, 0.3, 0.4, 0.7)\rangle$
$\dot{X}_4$	$\langle(0.45, 0.55, 0.65, 0.75); (0.1, 0.6, 0.2, 0.4), (0.6, 0.3, 0.4, 0.1)\rangle$
$\dot{X}_5$	$\langle(0.10, 0.20, 0.25, 0.30); (0.1, 0.3, 0.3, 0.4), (0.5, 0.3, 0.4, 0.1)\rangle$
	$u_2$
$\dot{X}_1$	$\langle(0.15, 0.20, 0.25, 0.40); (0.2, 0.3, 0.5, 0.4), (0.2, 0.3, 0.4, 0.5)\rangle$
$\dot{X}_2$	$\langle(0.05, 0.10, 0.15, 0.20); (0.1, 0.3, 0.5, 0.4), (0.2, 0.3, 0.4, 0.1)\rangle$
$\dot{X}_3$	$\langle(0.05, 0.10, 0.15, 0.20); (0.1, 0.3, 0.5, 0.4), (0.2, 0.3, 0.4, 0.1)\rangle$
$\dot{X}_4$	$\langle(0.15, 0.20, 0.25, 0.30); (0.1, 0.5, 0.2, 0.4), (0.2, 0.7, 0.4, 0.3)\rangle$
$\dot{X}_5$	$\langle(0.01, 0.05, 0.10, 0.15); (0.1, 0.3, 0.2, 0.4), (0.5, 0.3, 0.4, 0.7)\rangle$
	$u_3$
$\dot{X}_1$	$\langle(0.15, 0.20, 0.25, 0.30); (0.1, 0.5, 0.2, 0.4), (0.2, 0.7, 0.4, 0.3)\rangle$
$\dot{X}_2$	$\langle(0.01, 0.05, 0.10, 0.15); (0.1, 0.3, 0.2, 0.4), (0.5, 0.3, 0.4, 0.7)\rangle$
$\dot{X}_3$	$\langle(0.25, 0.35, 0.40, 0.45); (0.1, 0.3, 0.2, 0.5), (0.2, 0.5, 0.4, 0.1)\rangle$
$\dot{X}_4$	$\langle(0.70, 0.85, 0.90, 1.00); (0.1, 0.3, 0.3, 0.4), (0.1, 0.3, 0.4, 0.5)\rangle$
$\dot{X}_5$	$\langle(0.45, 0.55, 0.65, 0.75); (0.1, 0.6, 0.2, 0.4), (0.6, 0.3, 0.4, 0.1)\rangle$
	$u_4$
$\dot{X}_1$	$\langle(0.10, 0.20, 0.25, 0.30); (0.1, 0.3, 0.3, 0.4), (0.5, 0.3, 0.4, 0.1)\rangle$
$\dot{X}_2$	$\langle(0.15, 0.20, 0.25, 0.40); (0.2, 0.3, 0.5, 0.4), (0.2, 0.3, 0.4, 0.5)\rangle$
$\dot{X}_3$	$\langle(0.40, 0.45, 0.50, 0.55); (0.1, 0.3, 0.2, 0.9), (0.2, 0.5, 0.3, 0.1)\rangle$
$\dot{X}_4$	$\langle(0.50, 0.65, 0.70, 0.80); (0.1, 0.3, 0.2, 0.4), (0.5, 0.3, 0.4, 0.1)\rangle$
$\dot{X}_5$	$\langle(0.05, 0.10, 0.15, 0.20); (0.1, 0.3, 0.5, 0.4), (0.2, 0.3, 0.4, 0.1)\rangle$

**Step 2** We transformed the intuitionistic trapezoidal fuzzy multi decision matrix  $C = (c_{ij})_{m \times n}$  into the normalized intuitionistic trapezoidal fuzzy multi decision matrix  $K = (k_{ij})_{m \times n}$  as:

TABLE 3. The normalized intuitionistic trapezoidal fuzzy multi decision matrix  $K = (k_{ij})_{m \times n}$

$u_1$	
$\dot{X}_1$	$\langle(0.01, 0.05, 0.10, 0.15); (0.1, 0.3, 0.2, 0.4), (0.5, 0.3, 0.4, 0.7)\rangle$
$\dot{X}_2$	$\langle(0.15, 0.20, 0.25, 0.30); (0.1, 0.5, 0.2, 0.4), (0.2, 0.7, 0.4, 0.3)\rangle$
$\dot{X}_3$	$\langle(0.25, 0.30, 0.35, 0.40); (0.1, 0.3, 0.7, 0.4), (0.6, 0.3, 0.4, 0.7)\rangle$
$\dot{X}_4$	$\langle(0.45, 0.55, 0.65, 0.75); (0.1, 0.6, 0.2, 0.4), (0.6, 0.3, 0.4, 0.1)\rangle$
$\dot{X}_5$	$\langle(0.10, 0.20, 0.25, 0.30); (0.1, 0.3, 0.3, 0.4), (0.5, 0.3, 0.4, 0.1)\rangle$
$u_2$	
$\dot{X}_1$	$\langle(0.15, 0.20, 0.25, 0.40); (0.2, 0.3, 0.4, 0.5), (0.2, 0.3, 0.5, 0.4)\rangle$
$\dot{X}_2$	$\langle(0.05, 0.10, 0.15, 0.20); (0.2, 0.3, 0.4, 0.1), (0.1, 0.3, 0.5, 0.4)\rangle$
$\dot{X}_3$	$\langle(0.05, 0.10, 0.15, 0.20); (0.2, 0.3, 0.4, 0.1), (0.1, 0.3, 0.5, 0.4)\rangle$
$\dot{X}_4$	$\langle(0.15, 0.20, 0.25, 0.30); (0.2, 0.7, 0.4, 0.3), (0.1, 0.5, 0.2, 0.4)\rangle$
$\dot{X}_5$	$\langle(0.01, 0.05, 0.10, 0.15); (0.5, 0.3, 0.4, 0.7), (0.1, 0.3, 0.2, 0.4)\rangle$
$u_3$	
$\dot{X}_1$	$\langle(0.15, 0.20, 0.25, 0.30); (0.1, 0.5, 0.2, 0.4), (0.2, 0.7, 0.4, 0.3)\rangle$
$\dot{X}_2$	$\langle(0.01, 0.05, 0.10, 0.15); (0.1, 0.3, 0.2, 0.4), (0.5, 0.3, 0.4, 0.7)\rangle$
$\dot{X}_3$	$\langle(0.25, 0.35, 0.40, 0.45); (0.1, 0.3, 0.2, 0.5), (0.2, 0.5, 0.4, 0.1)\rangle$
$\dot{X}_4$	$\langle(0.70, 0.85, 0.90, 1.00); (0.1, 0.3, 0.3, 0.4), (0.1, 0.3, 0.4, 0.5)\rangle$
$\dot{X}_5$	$\langle(0.45, 0.55, 0.65, 0.75); (0.1, 0.6, 0.2, 0.4), (0.6, 0.3, 0.4, 0.1)\rangle$
$u_4$	
$\dot{X}_1$	$\langle(0.10, 0.20, 0.25, 0.30); (0.1, 0.3, 0.3, 0.4), (0.5, 0.3, 0.4, 0.1)\rangle$
$\dot{X}_2$	$\langle(0.15, 0.20, 0.25, 0.40); (0.2, 0.3, 0.5, 0.4), (0.2, 0.3, 0.4, 0.5)\rangle$
$\dot{X}_3$	$\langle(0.40, 0.45, 0.50, 0.55); (0.1, 0.3, 0.2, 0.9), (0.2, 0.5, 0.3, 0.1)\rangle$
$\dot{X}_4$	$\langle(0.50, 0.65, 0.70, 0.80); (0.1, 0.3, 0.2, 0.4), (0.5, 0.3, 0.4, 0.1)\rangle$
$\dot{X}_5$	$\langle(0.05, 0.10, 0.15, 0.20); (0.1, 0.3, 0.5, 0.4), (0.2, 0.3, 0.4, 0.1)\rangle$

**Step 3** We inserted the weighted vector  $w = (w_1, w_2, \dots, w_n)$  such that  $w_j \geq 0$   $j = (1, 2, \dots, n)$  and  $\sum_{j=1}^n w_j = 1$  as:  $w = (0.4, 0.3, 0.2, 0.1)$ .

**Step 4** We obtained values  $b_i$  ( $i = 1, 2, \dots, 5$ ) for  $\dot{X}_i$  by the TIFMA operators as:

$$(4.1) \quad b_i = \text{TIFMA}(\alpha_{i1}, \alpha_{i2}, \dots, \alpha_{in}) = \left(\bigoplus_{j=1}^n\right)(w_j \odot \alpha_{ij}) \quad i = 1, 2, \dots, 5$$

$$(4.2) \quad b_i = \text{TIFMG}(\alpha_{i1}, \alpha_{i2}, \dots, \alpha_{in}) = \left(\bigotimes_{i=1}^n\right)(\alpha_{ij})^{w_i} \quad i = 1, 2, \dots, 5.$$

or

TABLE 4. The values based on TIFMA

$b_1$	$\langle\langle(0.0914, 0.1431, 0.1933, 0.2777); (0.1312, 0.3456, 0.3144, 0.4000), (0.3162, 0.3554, 0.4000, 0.4397)\rangle\rangle$
$b_2$	$\langle\langle(0.0939, 0.1422, 0.1924, 0.2541); (0.1105, 0.3881, 0.3371, 0.4000), (0.2402, 0.4210, 0.4000, 0.2690)\rangle\rangle$
$b_3$	$\langle\langle(0.2126, 0.2740, 0.3247, 0.3754); (0.1000, 0.3000, 0.5307, 0.5164), (0.3104, 0.3497, 0.3887, 0.2178)\rangle\rangle$
$b_4$	$\langle\langle(0.4501, 0.5814, 0.6628, 0.7941); (0.1000, 0.4941, 0.2211, 0.4000), (0.3757, 0.3484, 0.2155, 0.3492)\rangle\rangle$
$b_5$	$\langle\langle(0.1562, 0.2403, 0.3113, 0.3880); (0.1000, 0.3741, 0.2764, 0.4000), (0.4732, 0.3000, 0.4000, 0.1793)\rangle\rangle$

TABLE 5. The values based on TIFMG

$b_1$	$\langle\langle(0.0488, 0.1149, 0.1733, 0.2479); (0.1231, 0.3323, 0.2564, 0.4277), (0.2367, 0.4091, 0.4319, 0.5116)\rangle\rangle$
$b_2$	$\langle\langle(0.0628, 0.1231, 0.1786, 0.2380); (0.1320, 0.3680, 0.2699, 0.2639), (0.2456, 0.5012, 0.4000, 0.4835)\rangle\rangle$
$b_3$	$\langle\langle(0.1617, 0.2317, 0.2889, 0.3434); (0.1231, 0.3000, 0.4064, 0.2993), (0.3719, 0.3672, 0.4231, 0.4865)\rangle\rangle$
$b_4$	$\langle\langle(0.3573, 0.4504, 0.5247, 0.6073); (0.1231, 0.5104, 0.2670, 0.3669), (0.3865, 0.3672, 0.3459, 0.2915)\rangle\rangle$
$b_5$	$\langle\langle(0.0632, 0.1507, 0.2185, 0.2811); (0.1621, 0.3446, 0.3174, 0.4731), (0.4022, 0.3000, 0.3459, 0.2031)\rangle\rangle$

**Step 5** We calculate the score values based on Table 4,  $\bar{S}_1(b_i) \ i = (1, 2, \dots, 5)$

as:

$$\bar{S}_1(b_1) = 0.2304$$

$$\bar{S}_1(b_2) = 0.2094$$

$$\bar{S}_1(b_3) = 0.2214$$

$$\bar{S}_1(b_4) = 0.2238$$

$$\bar{S}_1(b_5) = 0.2974$$

and all the alternatives ranked as;  $b_5 > b_1 > b_4 > b_3 > b_2$ .

Similarity, based on Table 5, we have

$$\bar{S}_1^1(b_1) = 0.2463$$

$$\bar{S}_1^1(b_2) = 0.2168$$

$$\bar{S}_1^1(b_3) = 0.2277$$

$$\bar{S}_1^1(b_4) = 0.1947$$

$$\bar{S}_1^1(b_5) = 0.2882$$

and all the alternatives ranked as;  $b_5 > b_1 > b_3 > b_2 > b_4$

## REFERENCES

- [1] L. A. Zadeh, Fuzzy sets. *Information and control* 8 (3) (1965) 338–353.
- [2] K. T. Atanassov, Intuitionistic fuzzy sets, *Fuzzy Sets and Systems* 20 (1) (1986) 87–96.
- [3] A. Ban, A. Brândaş, L. Coroianu, C. Negruțiu and O. Nica, Approximations of fuzzy numbers by trapezoidal fuzzy numbers preserving the ambiguity and value, *Computers Mathematics with Applications* 61 (5) (2011) 1379–1401.
- [4] C. H. Cheng, A new approach for ranking fuzzy numbers by distance method. *Fuzzy sets and systems* 95 (3) (1998) 307–317.
- [5] Y. M. Wang, J. B. Yang, D. L. Xu and K. S. Chin, On the centroids of fuzzy numbers, *Fuzzy sets and systems* 157 (7) (2006) 919–926.
- [6] S. H. Wei and S. M. Chen, A new similarity measure between generalized fuzzy numbers, In *SCIS ISIS SCIS ISIS 2006*, Japan Society for Fuzzy Theory and Intelligent Informatics (2006) 315–320.
- [7] D. F. Li and J. Yang, A difference-index based ranking method of trapezoidal intuitionistic fuzzy numbers and application to multiattribute decision making, *Mathematical and Computational Applications* 20 (1) (2015) 25–38.
- [8] M. Xia, Z. Xu and B. Zhu, Generalized intuitionistic fuzzy Bonferroni means, *International Journal of Intelligent Systems* 27 (2012) 23–47.
- [9] W. Wang and X. Xin, Distance measure between intuitionistic fuzzy sets, *Pattern recognition letters* 26 (13) (2005) 2063–2069.
- [10] X. Luo, Z. Xu and X. Gou, Exponential operational laws and new aggregation operators of intuitionistic fuzzy information based on Archimedean T-conorm and T-norm, *International Journal of Machine Learning and Cybernetics* 9 (2018) 1261–1269.
- [11] E. Szmidi and J. Kacprzyk, Distances between intuitionistic fuzzy sets: straightforward approaches may not work, In *2006 3rd International IEEE Conference Intelligent Systems*, IEEE (2006, September) 716–721.
- [12] J. H. Park, K. M. Lim and Y. C. Kwun, Distance measure between intuitionistic fuzzy sets and its application to pattern recognition, *Journal of The Korean Institute of Intelligent Systems* 19 (4) (2009) 556–561.
- [13] Z. Liang and P. Shi, Similarity measures on intuitionistic fuzzy sets, *Pattern recognition letters* 24 (15) (2003) 2687–2693.
- [14] G. A. R. G. Harish, An improved cosine similarity measure for intuitionistic fuzzy sets and their applications to decision-making process, *Hacetatepe Journal of Mathematics and Statistics* 47 (6) (2018) 1578–1594.
- [15] Q. Lei, Z. Xu, H. Bustince and J. Fernandez, Intuitionistic fuzzy integrals based on Archimedean t-conorms and t-norms, *Inform. Sci.* 327 (2016) 57–70.
- [16] M. Xia, Z. Xu and B. Zhu, Some issues on intuitionistic fuzzy aggregation operators based on Archimedean t-conorm and t-norm, *Knowledge-Based Systems* 31 (2012) 78–88.
- [17] L. Baccour, A. M. Alimi and R. I. John, Similarity measures for intuitionistic fuzzy sets: State of the art, *Journal of Intelligent Fuzzy Systems* 24 (1) (2013) 37–49.
- [18] H. Garg and D. Rani, Generalized geometric aggregation operators based on t-norm operations for complex intuitionistic fuzzy sets and their application to decision-making, *Cognitive Computation* 12 (3) (2020) 679–698.
- [19] P. Grzegorzewski, Distances between intuitionistic fuzzy sets and/or interval-valued fuzzy sets based on the Hausdorff metric, *Fuzzy sets and systems* 148 (2) (2004) 319–328.
- [20] J. Mo and H. L. Huang, Archimedean geometric Heronian mean aggregation operators based on dual hesitant fuzzy set and their application to multiple attribute decision making, *Soft Computing* 24 (2020) 14721–14733.
- [21] R. T. Ngan, B. C. Cuong and M. Ali, H-max distance measure of intuitionistic fuzzy sets in decision making, *Applied Soft Computing* 69 (2018) 393–425.
- [22] Z. Pei and L. Zheng, A novel approach to multi-attribute decision making based on intuitionistic fuzzy sets, *Expert Systems with Applications* 39 (3) (2012) 2560–2566.
- [23] E. Szmidi and J. Kacprzyk, Distances between intuitionistic fuzzy sets, *Fuzzy sets and systems* 114 (3) (2000) 505–518.

- [24] C. Tan, Z. Z. Jiang, X. Chen and W. H. Ip, Atanassov’s intuitionistic fuzzy Quasi-Choquet geometric operators and their applications to multicriteria decision making, *Fuzzy Optimization and Decision Making* 14 (2015) 139–172.
- [25] S. P. Wan and Z. H. Yi, Power average of trapezoidal intuitionistic fuzzy numbers using strict  $t$ -norms and  $t$ -conorms, *IEEE Transactions on Fuzzy Systems* 24 (5) (2015) 1035–1047.
- [26] M. Xia, Interval-valued intuitionistic fuzzy matrix games based on Archimedean  $t$ -conorm and  $t$ -norm, *International Journal of General Systems* 47 (3) (2018) 278–293.
- [27] Z. S. Xu and J. Chen, An overview of distance and similarity measures of intuitionistic fuzzy sets, *International Journal of Uncertainty, Fuzziness and Knowledge-Based Systems* 16 (04) (2008) 529–555.
- [28] D. Yu, Group decision making based on generalized intuitionistic fuzzy prioritized geometric operator, *International Journal of Intelligent Systems* 27 (7) (2012) 635–661.
- [29] Y. Song, X. Wang, L. Lei and A. Xue, A new similarity measure between intuitionistic fuzzy sets and its application to pattern recognition, In *Abstract and Applied Analysis* 2014 Article ID 384241 (2014) 1–11.
- [30] B. Yusoff, I. Taib and L. Abdullah, A new similarity measure on intuitionistic fuzzy sets, *International Journal of Mathematical and Computational Sciences* 5 (6) (2011) 819–823.
- [31] V. Uluçay, I. Deli and M. Şahin, Trapezoidal fuzzy multi-number and its application to multi-criteria decision-making problems, *Neural Computing and Applications* 30 (5) (2018) 1469–1478.
- [32] I. Deli and M. A. Keleş, Distance measures on trapezoidal fuzzy multi-numbers and application to multi-criteria decision-making problems, *Soft Computing* 25 (8) (2021) 5979–5992.
- [33] I. Deli and M. A. Keleş, *Neutrosophic Algebraic Structures and Their Applications*, Chapter 14: Similarity Measures on N-Valued Fuzzy Numbers and Application to Multiple Attribute Decision Making Problem, NSIA Publishing House, Neutrosophic Science International Association (NSIA), University of New Mexico, Gallup, United States (2022) 221-237.
- [34] M. Şahin, V. Uluçay and F. S. Yılmaz, Dice vector similarity measure based on multi-criteria decision making with trapezoidal fuzzy multi-Numbers, In *International Conference on Mathematics and Mathematics Education (ICMME-2017)*, Harran University, Şanlıurfa (2017, May) 11–13.
- [35] M. Şahin, V. Uluçay and F. S. Yılmaz, Improved hybrid vector similarity measures and their applications on trapezoidal fuzzy multi numbers, *Neutrosophic Triplet Structures* 1 (2019) 158–184.
- [36] D. Kesen, *Decision Making Operators of Trapezoidal Fuzzy Multi Numbers and Their Application to Decision Making Problems*, Master’s Thesis, Kilis 7 Aralık University, Graduate Education Institute (2022).
- [37] D. Kesen and I. Deli, *Data-Driven Modelling with Fuzzy Sets and their Applications to Knowledge Management*, Chapter 5: Trapezoidal Fuzzy Multi Aggregation Operator Based on Archimedean Norms and Their Application to Multi Attribute Decision-Making Problems, CRC Press/Taylor and Francis Group (2023).
- [38] V. Uluçay, I. Deli and M. Şahin, Intuitionistic trapezoidal fuzzy multi-numbers and its application to multi-criteria decision-making problems, *Complex Intelligent Systems* 5 (1) (2019) 65–78.
- [39] H. Karadöl, *Some new concepts on intuitionistic trapezoidal fuzzy Multi-numbers and application to multi-criteria decision-making Problems*, Master’s Thesis Kilis 7 Aralık University (2023).
- [40] S. Sebastian and T. V. Ramakrishnan, Multi-Fuzzy Sets, *International Mathematical Forum* 5 (50) (2010) 2471–2476.
- [41] G. Klir and B. Yuan, *Fuzzy sets and fuzzy logic: theory and applications*, Prentice Hall Upper Saddle River 1995.
- [42] E. P. Klement and R. Mesiar, *Logical, algebraic, analytic, and probabilistic aspects of triangular norms*, Elsevier, New York 2005.
- [43] G. Beliakov, A. Pradera and T. Calvo, *Aggregation functions: a guide for practitioners*. Springer, Heidelberg, Berlin, New York 2007.

- [44] W. Z. Wang and X. W. Liu, Intuitionistic fuzzy geometric aggregation operators based on Einstein operations, *Int. J. Intell. Syst.* 26 (2011) 1049–1075.
- [45] D. Zhang and G. Wang, Geometric score function of Pythagorean fuzzy numbers determined by the reliable information region and its application to group decision-making, *Engineering Applications of Artificial Intelligence* 121 (2023) 1–15.
- [46] X. Zhang, Multicriteria Pythagorean fuzzy decision analysis: A hierarchical QUALIFLEX approach with the closeness index-based ranking methods, *Inform. Sci.* 330 (2016) 104–124.
- [47] X. D. Peng and J. Dai, Approaches to Pythagorean fuzzy stochastic multi-criteria decision making based on prospect theory and regret theory with new distance measure and score function, *International Journal of intelligent system* 32 (1) (2017) 1187–1214.
- [48] D. Q. Li and W. Y. Zeng, Distance measure of Pythagorean fuzzy sets, *International Journal of intelligent systems* 33 (2) (2018) 348–361.
- [49] R. R. Yager and A. M. Abbasov, Pythagorean membership grades, complex numbers and decision making, *International Journal of Intelligent Systems* 28 (2013) 436–452.
- [50] R. R. Yager, Pythagorean membership grades in multicriteria decision making, *IEEE Transactions on Fuzzy Systems* 22 (4) (2014) 958–965.
- [51] X. D. Peng, Algorithm for Pythagorean fuzzy multi-criteria decision making based on WDBA with new score function, *Fundamenta Informaticae* 165 (2) (2019) 99–137.
- [52] G. Beliakov, H. Bustince, D. P. Goswami, U. K. Mukherjee and N. R. Pal, On averaging operators for Atanassov’s intuitionistic fuzzy sets, *Inform. Sci.* 181 (2011) 1116–1124.
- [53] Z. Xu and R. R. Yager, Some geometric aggregation operators based on intuitionistic fuzzy sets, *Int. J. General Syst.* 35 (2006) 417–433.
- [54] J. Wu and Q. W. Cao, Same families of geometric aggregation operators with intuitionistic trapezoidal fuzzy numbers. *Applied Mathematical Modelling* 37 (1-2) (2013) 318–327.
- [55] M. Xia, Z. Xu and B. Zhu, Some issues on intuitionistic fuzzy aggregation operators based on Archimedean t-conorm and t-norm, *Knowledge-Based Systems* 31 (2012) 78–88.

İRFAN DELİ ([irfandeli@kilis.edu.tr](mailto:irfandeli@kilis.edu.tr))

KILISLI MUALLİM RIFAT FACULTY OF EDUCATION, 7 ARALIK UNIVERSITY, 79000  
KILIS, TÜRKİYE, [IRFANDELI@KILIS.EDU.TR](mailto:IRFANDELI@KILIS.EDU.TR)

HÜMEYRA KARADÖL ([kara46israfil@gmail.com](mailto:kara46israfil@gmail.com))

Kilisli Muallim Rifat Faculty of Education, 7 Aralık University, 79000 Kilis, Türkiye,  
[kara46israfil@gmail.com](mailto:kara46israfil@gmail.com)