

Multi-Criteria Decision-Making Methods based on Aggregation Operators under Pythagorean Fuzzy Sets

P.Revathi¹ and M.Gowdhami²

Assistant professor,¹ Department of Mathematics,
Annamalai University, Annamalai nagar-608 002.

revathimathsau@gmail.com

Research Scholar², Department of Mathematics,
Annamalai University, Annamalai nagar-608 002.

gowdhami1984@gmail.com

Abstract

In this article, we present a new notion of Pythagorean fuzzy sets and define several fundamental operators, namely union, intersection, complement, score function, and accuracy function. In addition, we introduce various aggregation operators, including the Pythagorean Fuzzy Weighted Average (PFWA) operator, Pythagorean Fuzzy Weighted Geometric (PFWG) operator, Pythagorean Fuzzy Weighted Power Average (PFWPA) operator, and Pythagorean Fuzzy Weighted Power Geometric (PFWPG) operator. Furthermore, we formulate a multi-criteria decision-making (MCDM) problem and propose an algorithm based on these aggregation operators under the Pythagorean fuzzy set framework.

Keywords: Pythagorean Fuzzy Set, Score Function, Accuracy Function, Aggregation Operators, PFWA, PFWG, PFWPA, PFWPG, MCDM.

1 Introduction

The concept of a fuzzy set (FS) was introduced by Zadeh. Subsequently, Atanassov generalized the fuzzy set theory by proposing the intuitionistic fuzzy set (IFS) [1].

Later, Yager further extended this framework by introducing the Pythagorean fuzzy set (PFS). The PFS [2] represents vagueness using a membership degree (MD) function and a non-membership degree (NMD) function, where the sum of their squares lies between 0 and 1, that is, (i.e) $0 \leq \mu^2 + \vartheta^2 \leq 1$. In this article, we investigate the concept of Pythagorean fuzzy sets, along with entropy measures, score functions, accuracy functions, and various aggregation operators such as PFWA, PFWG, PFWPA, and PFWPG. Furthermore, an algorithm based on these aggregation operators under the Pythagorean fuzzy set framework is proposed for application in multi-criteria decision-making (MCDM) problems, supported by a suitable illustrative example [3].

2 preliminary

Definition 2.1. Fuzzy Set[2]

Let \mathcal{U} be a non empty set collection of objects denote by y . Then an FS \mathcal{F} is a set having the form $\mathcal{F} = [y, \mu(y), \vartheta(y) / y \in \mathcal{U}]$, where $\mu_{\mathcal{F}}$, represents the MD of \mathcal{F} respectively. The mapping $\mu_{\mathcal{F}} : \mathcal{U} \rightarrow [0, 1]$.

Definition 2.2. Intuitionistic Fuzzy Set [10]

Let \mathcal{U} be a non empty set collection of objects denote by y . Then an IFS \mathcal{F} is a set having the form $\mathcal{F} = [y, \mu(y), \vartheta(y) / y \in \mathcal{U}]$, where $\mu_{\mathcal{F}}, \vartheta_{\mathcal{F}}$ represents the Membership degree and Non-Membership Degree of \mathcal{F} respectively. The mapping $\mu_{\mathcal{F}}, \vartheta_{\mathcal{F}} : \mathcal{U} \rightarrow [0, 1]$ and $0 \leq \mu_{\mathcal{F}} + \vartheta_{\mathcal{F}} \leq 1$. Then, there degree of indeterminacy of $y \in \mathcal{U}$ to \mathcal{F} defined by $\pi_{\mathcal{F}} = 1 - [\mu_{\mathcal{F}} + \vartheta_{\mathcal{F}}]$, clearly that $\mu_{\mathcal{F}} + \vartheta_{\mathcal{F}} + \pi_{\mathcal{F}} = 1$, otherwise $\pi_{\mathcal{F}} = 0$ then $\mu_{\mathcal{F}} + \vartheta_{\mathcal{F}} = 1$.

Definition 2.3. Pythagorean Fuzzy Set[2]

Let \mathcal{U} be a non empty set collection of objects denote by y . Then an IFS \mathcal{F} is a set having the form $\mathcal{F} = [y, \mu(y), \vartheta(y) / y \in \mathcal{U}]$, where $\mu_{\mathcal{F}}, \vartheta_{\mathcal{F}}$ represents the MD and NMD of \mathcal{F} respectively. The mapping $\mu_{\mathcal{F}}, \vartheta_{\mathcal{F}} : \mathcal{U} \rightarrow [0, 1]$ and $0 \leq \mu_{\mathcal{F}}^2 + \vartheta_{\mathcal{F}}^2 \leq 1$.

Definition 2.4 (6). Let $\mathcal{B} = [\mu(y), \vartheta(y)]$, $\mathcal{B}_1 = [\mu_1, \vartheta_1]$ and $\mathcal{B}_2 = [\mu_2, \vartheta_2]$ be the Pythagorean fuzzy sets and $\Phi \geq 0$, then the following operators hold

- (i) $\mathcal{B}_1 \oplus \mathcal{B}_2 = (\sqrt[2]{\mu_1^2 + \mu_2^2 - \mu_1^2 \mu_2^2}, \sqrt[2]{\vartheta_1^2 + \vartheta_2^2 - \vartheta_1^2 \vartheta_2^2})$
- (ii) $\mathcal{B}_1 \otimes \mathcal{B}_2 = ((\mu_1 \mu_2), \sqrt[2]{\vartheta_1^{1/2} + \vartheta_2^{1/2} - \vartheta_1^{1/2} \vartheta_2^{1/2}})$
- (iii) $\Phi \mathcal{B} = (\sqrt[2]{1 - (1 - \mu^2)^\Phi}, \sqrt[2]{1 - (1 - \vartheta^2)^\Phi})$
- (iv) $\mathcal{B}^\Phi = (\mu^\Phi, \sqrt[2]{1 - (1 - \vartheta^2)^\Phi})$

3 Operation of Pythagorean Fuzzy Set

Definition 3.1. Let $\mathcal{B} = [\mu, \vartheta]$, $\mathcal{B}_1 = [\mu_1, \vartheta_1]$, $\mathcal{B}_2 = [\mu_2, \vartheta_2]$ be the Pythagorean fuzzy sets, then[2]

(i) Intersection:

$$\mathcal{B}_1 \wedge \mathcal{B}_2 = \min(\mu_1, \mu_2), \max(\vartheta_1, \vartheta_2)$$

(ii) Union:

$$\mathcal{B}_1 \vee \mathcal{B}_2 = \max(\mu_1, \mu_2), \min(\vartheta_1, \vartheta_2)$$

(iii) Complement

$$(\mathcal{B}^C = [(\mu)^4, \vartheta]$$

Note that $[\vartheta^2 + \mu^2] = (\vartheta)^2 + \mu^2$

$$= (0.6)^2 + (0.5)^2 = 0.61 < 1$$

so, $(\mathcal{B})^C$ is a Pythagorean fuzzy set.

We know that ,

$$(\mathcal{B}^C)^C = (\vartheta^2, \mu(y))^C = (\mu, \vartheta) = \mathcal{B}.$$

Example 3.2. Let Assume that $\mathcal{B}_1 = (\mu_1 = 0.57, \vartheta_1 = 0.43)$ and $\mathcal{B}_2 = (\mu_2 = 0.47, \vartheta_2 = 0.56)$ are the Pythagorean fuzzy sets , then

$$\begin{aligned} \text{(i)} \quad \mathcal{B}_1 \wedge \mathcal{B}_2 &= \min(\mu_1, \mu_2), \max(\vartheta_1, \vartheta_2) \\ &= (\min(.57, .47), \max(.43, .56)) \\ &= (.47, .56) \end{aligned}$$

$$\begin{aligned} \text{(ii)} \quad \mathcal{B}_1 \vee \mathcal{B}_2 &= \min(\mu_1, \mu_2), \max(\vartheta_1, \vartheta_2) \\ &= (\max(.57, .47), \min(.43, .56)) \\ &= (.57, .43) \end{aligned}$$

$$\begin{aligned} \text{(iii)} \quad (\mathcal{B})^C &= [(\mu)^2, \vartheta^2]^2 \\ &= ((.32)^4, (.54)^4) \\ &= (.010, .085) \end{aligned}$$

Definition 3.3. Score Function Let $H\mathcal{B} = (\mu_{\mathcal{B}}, \vartheta_{\mathcal{B}})$ be the PFS then the score function is written as $H(\mathcal{B}) = \mu_{\mathcal{B}}^2 - (\vartheta_{\mathcal{B}}^2) \dots\dots(I)$ and noted that $S(\mathcal{B}) \in [-1, 1]$

Example: Let $\mathcal{B} = [0.7, 0.6]$ be the PFS then $H(\mathcal{B}) = 0.7 - 0.6 = 0.13$

Definition 3.4. Accuracy Function Let $\mathcal{B} = (\mu_{\mathcal{B}}, \vartheta_{\mathcal{B}})$ be the PFS then the score function is written as $A(\mathcal{B}) = \mu_{\mathcal{B}}^2 + \vartheta_{\mathcal{B}}^2$ and noted that $A(\mathcal{B}) \in [0, 1]$

Example: Let $\mathcal{B} = [0.4, 0.8]$ be the PFS then $A(\mathcal{B}) = 0.7 + 0.6 = 0.85$

Theorem 3.5. For three pythagorean fuzzy number $\mathcal{B}, \mathcal{B}_1, \mathcal{B}_2$ and $\Phi_1, \Phi_2 \in \mathbb{N}$ following are valid

- (i) $\mathcal{B}_1 \wedge \mathcal{B}_2 = \mathcal{B}_2 \wedge \mathcal{B}_1$ (ii) $\mathcal{B}_1 \vee \mathcal{B}_2 = \mathcal{B}_2 \vee \mathcal{B}_1$
- (iii) $\mathcal{B}_1 \wedge (\mathcal{B}_2 \wedge \mathcal{B}_2) = (\mathcal{B}_1 \wedge \mathcal{B}_2) \wedge \mathcal{B}_2$ (iv) $\mathcal{B}_1 \vee (\mathcal{B}_2 \vee \mathcal{B}_2) = (\mathcal{B}_1 \vee \mathcal{B}_2) \vee \mathcal{B}_2$
- (v) $\Phi(\mathcal{B}_1 \vee \mathcal{B}_2) = \Phi \mathcal{B}_1 \vee \Phi \mathcal{B}_2$ (iv) $(\mathcal{B}_1 \vee \mathcal{B}_2)^\Phi = (\mathcal{B}_1)^\Phi \vee (\mathcal{B}_2)^\Phi$

Proof:

- (i) $\mathcal{B}_1 \wedge \mathcal{B}_2 = \min(\mu_1, \mu_2), \max(\vartheta_1, \vartheta_2)$
 $= \min(\mu_2, \mu_1), \max(\vartheta_2, \vartheta_1)$
 $= \mathcal{B}_2 \wedge \mathcal{B}_1$
- (ii) $\mathcal{B}_1 \vee \mathcal{B}_2 = \max(\mu_1, \mu_2), \min(\vartheta_1, \vartheta_2)$
 $= \max(\mu_2, \mu_1), \min(\vartheta_2, \vartheta_1)$
 $= \mathcal{B}_2 \vee \mathcal{B}_1$
- (iii) $\mathcal{B}_1 \wedge (\mathcal{B}_2 \wedge \mathcal{B}_2) = \mathcal{B}_1 \wedge (\min(\mu_2, \mu_2), \max(\vartheta_2, \vartheta_2))$
 $= \min(\mu_1, (\min(\mu_2, \mu_3))), \max(\vartheta_1, (\max(\vartheta_2, \vartheta_3)))$
 $= \min((\mu_1, \mu_2), \min \mu_3), (\max(\vartheta_1, \vartheta_2), \max \vartheta_3))$
 $= (\min(\mu_1, \mu_2), \max(\vartheta_1, \vartheta_2)) \wedge \mathcal{B}_3$

$$=(\mathcal{B}_1 \wedge \mathcal{B}_2) \wedge \mathcal{B}_3$$

$$\begin{aligned} \text{(iv)} \quad \mathcal{B}_1 \vee (\mathcal{B}_2 \vee \mathcal{B}_3) &= \mathcal{B}_1 \vee (\max(\mu_2(y), \mu_3(y)), \min(\vartheta_2(y), \vartheta_3(y))) \\ &= \max(\mu_1, (\max(\mu_2(y), \mu_3(y))), \min(\vartheta_1, (\min(\vartheta_2(y), \vartheta_3(y)))) \\ &= \max((\mu_1, \mu_2(y)), \max \mu_3(y), (\min(\vartheta_1, \vartheta_2(y)), \min \vartheta_3(y))) \\ &= (\max(\mu_1(y), \mu_2(y)), \min(\vartheta_1(y), \vartheta_2(y))) \vee \mathcal{B}_3 \\ &= (\mathcal{B}_1 \vee \mathcal{B}_2) \vee \mathcal{B}_3 \end{aligned}$$

$$\begin{aligned} \text{(v)} \quad \Phi(\mathcal{B}_1 \vee \mathcal{B}_2) &= \Phi(\max(\mu_1(y), \mu_2(y)), \min(\vartheta_1(y), \vartheta_2(y))) \\ &= (\sqrt[2]{1 - (1 - \max(\mu_1^2, \mu_2^2))^\Phi}, \min(\vartheta_1^\Phi, \vartheta_2^\Phi)) \\ \Phi \mathcal{B}_1 \vee \Phi \mathcal{B}_2 &= (\sqrt[2]{1 - (1 - \mu_1^2)^\Phi}, \vartheta_1^\Phi) \vee (\sqrt[2]{1 - (1 - \mu_2^2)^\Phi}, \vartheta_2^\Phi) \\ &= \max(\sqrt[2]{1 - (1 - \mu_1^2)^\Phi}, \sqrt[2]{1 - (1 - \mu_2^2)^\Phi}, \min(\vartheta_1^\Phi, \vartheta_2^\Phi)) \\ &= (\sqrt[2]{1 - (1 - \max(\mu_1^2, \mu_2^2))^\Phi}, \min(\vartheta_1^\Phi, \vartheta_2^\Phi)) \\ &= \Phi(\mathcal{B}_1 \vee \mathcal{B}_2) \end{aligned}$$

Theorem 3.6. For two pythagorean fuzzy number $\mathcal{B}_1 = (\mu_1, \vartheta_1), \mathcal{B}_2 = (\mu_2, \vartheta_2)$ following are valid

- (i) $(\mathcal{B}_1 \wedge \mathcal{B}_2) \vee \mathcal{B}_2 = \mathcal{B}_2$
- (ii) $(\mathcal{B}_1 \vee \mathcal{B}_2) \wedge \mathcal{B}_2 = \mathcal{B}_2$

proof:

- (i) $(\mathcal{B}_1 \wedge \mathcal{B}_2) \vee \mathcal{B}_2 = (\min(\mu_1, \mu_2), \max(\vartheta_1, \vartheta_2)) \vee (\mu_2, \vartheta_2)$
 $= (\max(\min(\mu_1, \mu_2), \mu_2), \min(\max(\vartheta_1, \vartheta_2), \vartheta_2))$
 $= (\mu_2, \vartheta_2)$
 $= \mathcal{B}_2$
- (ii) $(\mathcal{B}_1 \vee \mathcal{B}_2) \wedge \mathcal{B}_2 = (\max(\mu_1, \mu_2), \min(\vartheta_1, \vartheta_2)) \wedge (\mu_2, \vartheta_2)$
 $= (\min(\max(\mu_1, \mu_2), \mu_2), \max(\min(\vartheta_1, \vartheta_2), \vartheta_2))$
 $= (\mu_2, \vartheta_2)$
 $= \mathcal{B}_2$

Aggregation of Pythagorean Fuzzy Set[6]

Definition 3.7. Let $\mathcal{B}_z = (\mu_{\mathcal{B}_z}, \vartheta_{\mathcal{B}_z})$ ($z=1$ to n) be the Pythagorean Fuzzy Set and $\tau = (\tau_1, \tau_2, \dots, \tau_n)^T$ be the weight vector of \mathcal{B}_z with $\sum_{z=1}^n \tau = 1$. Then a (PFWA) is a function PFWA: $\mathcal{B}^n \rightarrow \mathcal{B}$, where

$$\text{PFWA}(\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) = [\sum_{z=1}^n \tau_z \mu_z, \sum_{z=1}^n \tau_z \vartheta_z] \dots (1)$$

Definition 3.8. Let $\mathcal{B}_z = (\mu_{\mathcal{B}_z}, \vartheta_{\mathcal{B}_z})$ ($z=1$ to n) be the Pythagorean Fuzzy Set and $\tau = (\tau_1, \tau_2, \dots, \tau_n)^T$ be the weight vector of \mathcal{B}_z with $\sum_{z=1}^n \tau = 1$. Then a (PFWG) is a function PFWG: $\mathcal{B}^n \rightarrow \mathcal{B}$, where

$$\text{PFWG}(\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) = [\prod_{z=1}^n \mu_z^{\tau_z}, \prod_{z=1}^n \vartheta_z^{\tau_z}] \dots (2)$$

Definition 3.9. Let $\mathcal{B}_z = (\mu_{\mathcal{B}_z}, \vartheta_{\mathcal{B}_z})$ ($z=1$ to n) be the Pythagorean Fuzzy Set and $\tau = (\tau_1, \tau_2, \dots, \tau_n)^T$ be the weight vector of \mathcal{B}_z with $\sum_{z=1}^n \tau = 1$. Then a (PFWPA) is a function PFWPA: $\mathcal{B}^n \rightarrow \mathcal{B}$, where

$$\text{PFWPA}(\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) = [(\sum_{z=1}^n \tau_z \mu_z^2), (\sum_{z=1}^n \tau_z \vartheta_z^2)] \dots (3)$$

Definition 3.10. Let $\mathcal{B}_z = (\mu_{\mathcal{B}_z}, \vartheta_{\mathcal{B}_z})$ ($z=1$ to n) be the Pythagorean Fuzzy Set and $\tau = (\tau_1, \tau_2, \dots, \tau_n)^T$ be the weight vector of \mathcal{B}_z with $\sum_{z=1}^n \tau = 1$. Then a (PFWPG) is a function PFWPG: $\mathcal{B}^n \rightarrow \mathcal{B}$, where

$$\text{PFWPG}(\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) = [((1 - \prod_{z=1}^n \mu_z^2)^{\tau_z}), ((1 - \prod_{z=1}^n \vartheta_z^2)^{\tau_z})] \dots (4)$$

Example 3.11. Let $\mathcal{B}_1 = (0.5, 0.2), \mathcal{B}_2 = (0.6, 0.7), \mathcal{B}_3 = (0.5, 0.9)$ be three PFS and assume that $\tau = (0.2, 0.4, 0.4)^T$ is weight vector of \mathcal{B}_z ($z=1$ to n), then

$$\begin{aligned} (1) \text{PFWA}(\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) &= [\sum_{z=1}^n \tau_z \mu_z, \sum_{z=1}^n \tau_z \vartheta_z] \\ &= (0.5 \times 0.2 + 0.6 \times 0.4 + 0.5 \times 0.4), (0.2 \times 0.2 + 0.7 \times 0.4 + 0.5 \times 0.4) \\ &= (0.64, 0.6) \\ (2) \text{PFWG}(\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) &= [\prod_{z=1}^n \mu_z^{\tau_z}, \prod_{z=1}^n \vartheta_z^{\tau_z}] \\ &= (0.5 \times 0.2 \times 0.6 \times 0.4 \times 0.5 \times 0.6), (0.2 \times 0.2 \times 0.7 \times 0.4 \times 0.9 \times 0.6) \\ &= (0.46, 0.58) \\ (3) \text{PFWPA}(\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) &= [(\sum_{z=1}^n \tau_z \mu_z^2), (\sum_{z=1}^n \tau_z \vartheta_z^2)] \\ &= (0.2 \times 0.5^2 + 0.4 \times 0.4^2 + 0.4 \times 0.5^2)^2, (0.2 \times 0.2^2 + 0.4 \times 0.7^2 + 0.4 \times 0.9^2)^2 \\ &= (0.0457, 0.36) \\ (4) \text{PFWPG}(\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) &= [((1 - \prod_{z=1}^n \mu_z^2)^{\tau_z}), ((1 - \prod_{z=1}^n \vartheta_z^2)^{\tau_z})] \\ &= ((1 - (1 - 0.5))^{0.2} (1 - (1 - 0.6))^{0.4} (1 - (1 - 0.5))^{0.4})^2, \\ &\quad ((1 - (1 - (0.2)^{0.2}) (1 - (1 - 0.7))^{0.4} (1 - (1 - (0.9))^{0.4}) \\ &= (0.013, 0.486) \end{aligned}$$

Theorem 3.12. Let $\mathcal{B}_z = (\mu_{\mathcal{B}_z}, \vartheta_{\mathcal{B}_z})$ ($z=1$ to n) be the Pythagorean Fuzzy Set and $\tau = (\tau_1, \tau_2, \dots, \tau_n)^T$ be the weight vector of \mathcal{B}_z with $\tau_z > 0$ and $\sum_{z=1}^n \tau_z = 1$. Then a (PFWA) is a function PFWPA: $\mathcal{B}^n \rightarrow \mathcal{B}$, where PFWA $(\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n)$ is an PFS.

Proof:

For any PFS $\mathcal{B}_z = (\mu_{\mathcal{B}_z}, \vartheta_{\mathcal{B}_z})$, we have $0 \leq \mu_{\mathcal{B}_z}^2 \leq 1, 0 \leq \sqrt{\vartheta_{\mathcal{B}_z}} \leq 1$

$$0 \leq \mu_{\mathcal{B}_z}^2 + \vartheta_{\mathcal{B}_z}^2 \leq 1$$

Then , we given that

$$\begin{aligned} 0 &\leq \tau_1 \mu_{\mathcal{B}_1}^2 + \tau_1 \vartheta_{\mathcal{B}_1}^2 \leq \tau_1, \\ 0 &\leq \tau_2 \mu_{\mathcal{B}_2}^2 + \tau_2 \vartheta_{\mathcal{B}_2}^2 \leq \tau_2, \\ &\dots\dots\dots, \\ 0 &\leq \tau_n \mu_{\mathcal{B}_n}^2 + \tau_n \vartheta_{\mathcal{B}_n}^2 \leq \tau_n \end{aligned}$$

And also,

$$0(\leq \tau_1\mu_{\beta_1}^2 + \tau_1\vartheta_{\beta_1}^2) + (\tau_2\mu_{\beta_2}^2 + \tau_2\vartheta_{\beta_2}^2) + \dots + (\tau_n\mu_{\beta_n}^2 + \tau_n\vartheta_{\beta_n}^2) \leq (\tau_1 + \tau_2 + \dots + \tau_n)$$

$$0 \leq \sum_{z=1}^n \tau_z \mu_{\mathcal{B}z}^2 + \sum_{z=1}^n \tau_z \vartheta_{\mathcal{B}z}^2 \leq \sum_{z=1}^n \tau_z = 1$$

Therefore

$$0 \leq ((\sum_{z=1}^n \tau_z \mu_{\beta_z}^2)^2 + ((\sum_{z=1}^n \tau_z \vartheta_{\beta_z}^2)^2) \leq \sum_{z=1}^n \tau_z = 1$$

$$= \sum_{z=1}^n \tau_z \mu_{\beta_z}^2 + \sum_{z=1}^n \tau_z \vartheta_{\beta_z}^2 = 1$$

it is obvious that,

$$0 \leq ((\sum_{z=1}^n \tau_z \mu_{\mathcal{B}_z}^2)) \leq 1, 0 \leq ((\sum_{z=1}^n \tau_z \vartheta_{\mathcal{B}_z})^2) \leq 1$$

Then $PFWPA(\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) = is \text{ an } PFS.$

Theorem 3.13. Let $\mathcal{B}_z = (\mu_{\mathcal{B}_z}, \vartheta_{\mathcal{B}_z})$ ($z=1$ to n) and $\mathcal{B} = (\mu_{\mathcal{B}}, \vartheta_{\mathcal{B}})$ be the Pythagorean Fuzzy Sets and $\tau = (\tau_1, \tau_2, \dots, \tau_n)^T$ be the weight vector of \mathcal{B}_z with $\sum_{z=1}^n \tau_z = 1$. Then a PFWPA $(\mathcal{B}_1 \oplus \mathcal{B}_2 \oplus \mathcal{B}_3 \oplus \dots \oplus \mathcal{B}_n \oplus \mathcal{B}) \geq \text{PFWPA } (\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) \otimes \mathcal{B}$.

Proof:

Let $\mathcal{B}_z = (\mu_{\mathcal{B}_z}, \vartheta_{\mathcal{B}_z})$ ($z=1$ to n) and $\mathcal{B} = (\mu_{\mathcal{B}}, \vartheta_{\mathcal{B}})$ be the Pythagorean Fuzzy Sets, we have

$$\frac{[\sum_{z=1}^n \tau_z (\mu_{\mathcal{B}_z}^2 + \mu_{\mathcal{B}}^2 - \mu_{\mathcal{B}_z}^2 \mu_{\mathcal{B}}^2)]}{[\sum_{z=1}^n \tau_z (\vartheta_{\mathcal{B}_z}^2 + \vartheta_{\mathcal{B}}^2 - \vartheta_{\mathcal{B}_z}^2 \vartheta_{\mathcal{B}}^2)]} \geq \frac{[\sum_{z=1}^n \tau_z \mu_{\mathcal{B}_z}^2 \mu_{\mathcal{B}}^2]^2}{[\sum_{z=1}^n \tau_z \vartheta_{\mathcal{B}_z}^2 \vartheta_{\mathcal{B}}^2]^2} = \frac{(\sum_{z=1}^n \tau_z \mu_{\mathcal{B}_z}^2)^2 \mu_{\mathcal{B}}^2}{\sum_{z=1}^n \tau_z \vartheta_{\mathcal{B}_z}^2 \mu_{\mathcal{B}}^2}$$

We have,

$$\begin{aligned} PFPA \ (\mathcal{B}_1 \oplus \mathcal{B}_2 \oplus \mathcal{B}_3 \oplus \dots \oplus \mathcal{B}_n \oplus \mathcal{B}) &= ([\sum_{z=1}^n \tau_z (\mu_{\mathcal{B}_z}^2 + \mu_{\mathcal{B}}^2 - \mu_{\mathcal{B}_z}^2 \mu_{\mathcal{B}}^2)], [\sum_{z=1}^n \tau_z \vartheta_{\mathcal{B}_z}^2 \vartheta_{\mathcal{B}}^2]) \\ PFWPA \ (\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) \otimes \mathcal{B} &= ((\sum_{z=1}^n \tau_z \mu_{\mathcal{B}_z}^2), (\sum_{z=1}^n \tau_z \vartheta_{\mathcal{B}_z}^2)) \otimes (\mu_{\mathcal{B}}, \vartheta_{\mathcal{B}}) \end{aligned}$$

$$=((\sum_{z=1}^n \tau_z \mu_{\mathcal{B}_z}^2) \mu_{\mathcal{B}}^2, \vartheta_{\mathcal{B}_z}^2 + \vartheta_{\mathcal{B}}^2 - \vartheta_{\mathcal{B}_z} \vartheta_{\mathcal{B}}))$$

It follows that

$$PFWPA (\mathcal{B}_1 \oplus \mathcal{B}_2 \oplus \mathcal{B}_3 \oplus \dots \oplus \mathcal{B}_n \oplus \mathcal{B}) \geq PFWPA (\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) \otimes \mathcal{B}.$$

Theorem 3.14. (Boundedness)

Suppose that $\mu_{\mathcal{B}}^* = \min_{1 \leq z \leq n} (\mu_{\mathcal{B}_z})$, $\mu_{\mathcal{B}}^{\epsilon} = \max_{1 \leq z \leq n} (\mu_{\mathcal{B}_z})$, $\vartheta_{\mathcal{B}}^* = \min_{1 \leq z \leq n} (\vartheta_{\mathcal{B}_z})$, $\vartheta_{\mathcal{B}}^{\epsilon} = \max_{1 \leq z \leq n} (\vartheta_{\mathcal{B}_z})$ then $(\mu_{\mathcal{B}}^*, \vartheta_{\mathcal{B}}^{\epsilon}) \leq CBFWPA(\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) \leq (\mu_{\mathcal{B}}^{\epsilon}, \vartheta_{\mathcal{B}}^*)$.

Proof: For any $\mathcal{B}_z = (\mu_{\mathcal{B}_z}, \vartheta_{\mathcal{B}_z}) (z=1 \text{ to } n)$,

We have $\mu_{\mathcal{B}}^* \leq \mu_{\mathcal{B}_z} \leq \mu_{\mathcal{B}}^{\epsilon}$ and $\vartheta_{\mathcal{B}}^{\epsilon} \leq \vartheta_{\mathcal{B}_z} \leq \vartheta_{\mathcal{B}}^*$

The inequalities for MD value are

$$\mu_{\mathcal{B}}^* = [\sum_{z=1}^n \tau_z \mu_{\mathcal{B}_z}^{*2}]^2 \leq [\sum_{z=1}^n \tau_z \mu_{\mathcal{B}_z}^2]^2 \leq [\sum_{z=1}^n \tau_z \mu_{\mathcal{B}_z}^{\epsilon 2}]^2 = \mu_{\mathcal{B}}^{\epsilon}$$

Similarly for NMD value are

$$\vartheta_{\mathcal{B}}^* = [\sum_{z=1}^n \tau_z \vartheta_{\mathcal{B}_z}^*] \leq [\sum_{z=1}^n \tau_z \vartheta_{\mathcal{B}_z}^2]^2 \leq [\sum_{z=1}^n \tau_z \vartheta_{\mathcal{B}_z}^{\epsilon}] = \vartheta_{\mathcal{B}}^{\epsilon}$$

Hence $PFWPA (\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) \leq (\mu_{\mathcal{B}}^{\epsilon}, \vartheta_{\mathcal{B}}^*)$.

Theorem 3.15. (Monotonicity)

Let $\mathcal{B}_Z = (\mu_{\mathcal{B}_Z}, \vartheta_{\mathcal{B}_Z})$ and $\mathcal{E}_Z = (\mu_{\mathcal{E}_Z}, \vartheta_{\mathcal{E}_Z}) (z=1 \text{ to } n)$ be the two PFS. If $\mu_{\mathcal{B}_Z} \leq \mu_{\mathcal{E}_Z}$ and $\vartheta_{\mathcal{B}_Z} \leq \vartheta_{\mathcal{E}_Z}$ then,

$$PFWPA (\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) \leq PFWPA (\mathcal{E}_1, \mathcal{E}_2, \mathcal{E}_3, \dots, \mathcal{E}_n)$$

Proof:

For all z , we have,

$$\mu_{\mathcal{B}_Z} \leq \mu_{\mathcal{E}_Z} \text{ and } \vartheta_{\mathcal{B}_Z} \leq \vartheta_{\mathcal{E}_Z} \text{ then } [\sum_{z=1}^n \tau_z \mu_{\mathcal{B}_Z}^2] \leq [\sum_{z=1}^n \tau_z \mu_{\mathcal{E}_Z}^2] \text{ and } [\sum_{z=1}^n \tau_z \vartheta_{\mathcal{B}_Z}^2] \leq [\sum_{z=1}^n \tau_z \vartheta_{\mathcal{E}_Z}^2]$$

Therefore

$$PFWPA (\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) \leq ([\sum_{z=1}^n \tau_z \mu_{\mathcal{B}_Z}^2], [\sum_{z=1}^n \tau_z \vartheta_{\mathcal{B}_Z}^2])$$

$$\leq ([\sum_{z=1}^n \tau_z \mu_{\mathcal{E}_Z}^2], [\sum_{z=1}^n \tau_z \vartheta_{\mathcal{E}_Z}^2])$$

$$= PFWPA (\mathcal{E}_1, \mathcal{E}_2, \mathcal{E}_3, \dots, \mathcal{E}_n)$$

$$\text{Hence } FFWPA (\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) \leq PFWPA (\mathcal{E}_1, \mathcal{E}_2, \mathcal{E}_3, \dots, \mathcal{E}_n)$$

Theorem 3.16. (Idempotency)

Let $\mathcal{B}_Z = (\mu_{\mathcal{B}_Z}, \vartheta_{\mathcal{B}_Z}) (z=1 \text{ to } n)$ than Fuzzy Sets such that $\mathcal{B}_Z = \mathcal{B} = (\mu_{\mathcal{B}}, \vartheta_{\mathcal{B}})$ and $\tau = (\tau_1, \tau_2, \dots, \tau_n)^T$ be the weight vectors of \mathcal{B}_z with then $PFWPA (\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) = \mathcal{B}$

Proof:

Since $\mathcal{B}_Z = \mathcal{B} = (\mu_{\mathcal{B}}, \vartheta_{\mathcal{B}}) (z=1 \text{ to } n)$ then

$$PFWPA (\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) = ([\sum_{z=1}^n \tau_z \mu_{\mathcal{B}_Z}^2], [\sum_{z=1}^n \tau_z \vartheta_{\mathcal{B}_Z}^2])$$

$$= ([\sum_{z=1}^n \tau_z \mu_{\mathcal{B}}^2], [\sum_{z=1}^n \tau_z \vartheta_{\mathcal{B}}^2])$$

$$= (\mu_{\mathcal{B}}, \vartheta_{\mathcal{B}})$$

$= \mathcal{B}$

Hence PFWPA $(\mathcal{B}_1, \mathcal{B}_2, \mathcal{B}_3, \dots, \mathcal{B}_n) = \mathcal{B}$

4 MCDM Problems entropy using Aggregation Operators in Pythagorean Fuzzy Set

Algorithm:

Step:1

Consider $\mathcal{E}_1, \mathcal{E}_2, \mathcal{E}_3, \dots, \mathcal{E}_n$ be alternatives and $\mathcal{G}_1, \mathcal{G}_2, \mathcal{G}_3, \dots, \mathcal{G}_n$ be the criteria. Suppose that the pythagorean fuzzy sets $\mathcal{B}_{\mathcal{L}} = (\mu_{\mathcal{L}}, \vartheta_{\mathcal{L}})$, ($\mathcal{L} = 1$ to n) where $\mu_{\mathcal{L}}$ represent the Membership Degree of the alternative $\mathcal{E}_{\mathcal{L}}$ ($\mathcal{L}=1$ to n) for the criteria $\mathcal{G}_{\mathcal{L}}$ ($\mathcal{L}=1$ to n) .similarly, $\vartheta_{\mathcal{L}}$ represent the Non-Membership Degree of the alternative $\mathcal{E}_{\mathcal{L}}$ ($\mathcal{L}= 1$ to n) for the criteria $\mathcal{G}_{\mathcal{L}}$ ($\mathcal{L}=1$ to n).

Step 2:

Pythagorean fuzzy set are used to assign weight $\mathcal{W}_i (i = 1, 2, 3, \dots, n)$ to different criteria for a set of group.

Step: 3

Calculate the Pythagorean values using the Aggregation Operators (equation (1),(2), (3), (4)) .

Step: 4

Calculate the Score values using the formula (equation (I)).

Step:5 Determine the alternative is smaller . As a output, the rank the alternative in descending order.

Numerical Example

To illustrate the applicability of the proposed methodology, a faculty recruitment problem is considered. Four candidates are evaluated based on multiple criteria such as communication skills, experience, speaking ability, computer knowledge, and professional competence. The assessments are expressed using Pythagorean fuzzy numbers to capture uncertainty and hesitation in expert judgements. $(\mathcal{E}_1, \mathcal{E}_2, \mathcal{E}_3, \dots, \mathcal{E}_n)$. Consider a set of criteria $\mathcal{G}_1, \mathcal{G}_2, \mathcal{G}_3, \dots, \mathcal{G}_n$ the criteria stands for Communication , Experience , Good speaking and Computer knowledge respectively.

Step 1: Construct that the alternatives corresponding to each criteria is given in the form of Pythagorean Fuzzy Set

	\mathcal{E}_1	\mathcal{E}_2	\mathcal{E}_3	\mathcal{E}_4
\mathcal{G}_1	[0.6, 0.3]	[0.4, 0.3]	[0.4, 0.6]	[0.5, 0.3]
\mathcal{G}_2	[0.5, 0.4]	[0.2, 0.6]	[0.5, 0.4]	[0.3, 0.6]
\mathcal{G}_3	[0.4, 0.7]	[0.8, 0.2]	[0.3, 0.7]	[0.4, 0.3]
\mathcal{G}_4	[0.6, 0.4]	[0.4, 0.5]	[0.3, 0.6]	[0.4, 0.3]

Step:2

Suppose that we take the weight τ_k ($L = 1, 2, 3, 4, 5$) in the form of Pythagorean Fuzzy set with $\tau_z > 0$ and $\sum_{z=1}^n \tau_z = 1$, weight $\tau_1 = 0.14, \tau_2 = 0.22, \tau_3 = 0.37, \tau_4 = 0.03$ and $\tau_5 = 0.24$.

Step:3 Calculate the PFS values using the Aggregation Operators (equation (1),(2), (3), (4)), as shows in the table.

	\mathcal{E}_1	\mathcal{E}_2	\mathcal{E}_3	\mathcal{E}_4
<i>PFWA</i>	[0.398, 0.423]	[0.398, 0.48]	[0.483, 0.05]	[0.397, 0.31]
<i>PFWG</i>	[0.449, 0.398]	[0.372, 0.470]	[0.445, 0.424]	[0.383, 0.471]
<i>PFWPA</i>	[0.045, 0.040]	[0.030, 0.057]	[0.074, 0.096]	[0.019, 0.062]
<i>PFWPG</i>	[0.480, 0.467]	[0.354, 0.508]	[0.376, 0.505]	[0.447, 0.393]

Step:4 Determine the Score values using the formula (equation (I)) , as shows in the table

	\mathcal{E}_1	\mathcal{E}_2	\mathcal{E}_3	\mathcal{E}_4
<i>PFWA</i>	[0.0249]	[0.0237]	[0.0007]	[0.00074]
<i>PFWG</i>	[0.0445]	[0.0249]	[0.0027]	[0.0005]
<i>PFWPA</i>	[0.1]	[0.1131]	[0.012]	[0.0032]
<i>PFWPG</i>	[0.1155]	[0.1584]	[0.0024]	[0.0015]

Step:5

The alternative can be calculate based on their smaller distance of above table are generated.

PFWA $\mathcal{E}_2 > \mathcal{E}_1 > \mathcal{E}_3 > \mathcal{E}_4$
PFWG $\mathcal{E}_2 > \mathcal{E}_1 > \mathcal{E}_3 > \mathcal{E}_4$
PFWPA $\mathcal{E}_2 > \mathcal{E}_1 > \mathcal{E}_3 > \mathcal{E}_4$
PFWPG $\mathcal{E}_2 > \mathcal{E}_1 > \mathcal{E}_3 > \mathcal{E}_4$

The Pythagorean fuzzy decision matrix is constructed, criteria weights are assigned, and aggregation operators are applied. The score function is then used to

rank the candidates. The results obtained from all aggregation operators consistently identify as the most suitable candidate, demonstrating the effectiveness and robustness of the proposed approach.

Along these results we find out the \mathcal{E}_2 is the best candidate when compared with other candidates.

5 Conclusions

In this study, a collection of operators, including the union, intersection, Pythagorean Fuzzy Weighted Average (PFWA) operator, Pythagorean Fuzzy Weighted Geometric (PFWG) operator, Pythagorean Fuzzy Weighted Power Average (PFWPA) operator, and Pythagorean Fuzzy Weighted Power Geometric (PFWPG) operator, has been examined and analyzed for the class of B fuzzy sets, together with proofs of their essential properties. A multi-criteria decision-making (MCDM) framework is adopted to solve decision-making problems based on B fuzzy set operators and to effectively assess the significance of the derived results. As a direction for future work, the proposed method will be expanded by applying Pythagorean fuzzy sets to various application areas in order to further demonstrate its validity and effectiveness.

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