Applications of complex picture fuzzy soft power aggregation operators in multi-attribute decision making

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The major theme of this analysis is to suggest a new theory in the form of complex picture fuzzy soft (CPFS) information and to initiate their major algebraic laws, score value, and accuracy values. The mathematical form of the CPFS set includes three main functions, called supporting, abstinence, and supporting against terms with a prominent characteristic that the sum of the triplet will lie in the unit interval. Further, in the consideration of the power aggregation operator using generalized t-norm and t-conorm and CPFS information, we diagnosed the mathematical concept of CPFS power averaging (CPFSPA), CPFS weighted power averaging (CPFSWPA), CPFS ordered weighted power averaging (CPFSOWPA), CPFS power geometric (CPFSPG), CPFS weighted power geometric (CPFSWPG), CPFS ordered weighted power geometric (CPFSOWPG). Moreover, the major results and their particular investigation of the invented approaches are also deliberated. Additionally, in the consideration of diagnosed operators using CPFS information, we illustrated a MADM ("multi-attribute decision-making") tool to find the best option from the family of decisions. Finally, we have shown the supremacy and feasibility of the diagnosed operators with the help of sensitive analysis and geometrical representations.

Pattern recognition is a vital part of the decision-making strategy used as an application in the environment of engineering science, networking systems, and medical diagnoses. Pattern recognition is the computerized recognition of patterns and consistencies in information. It has been used in statistical information analysis, image analysis, signal procedure, bioinformatics, and machine learning. Pattern recognition has a lot of implementations in different areas, but various deficiencies were involved in the environment genuine life troubles in the consideration of crisp sets. To enhance the major theme of the research work, Atanassov1 invented the major and well-known theory of intuitionistic fuzzy set (IFS), by generalizing the theory of fuzzy set (FS)2. The main IFS is very feasible and flexible because of its shape, the prominent condition of IFS is described: \[ 0 \leq \mu_{\text{IFS}}(\overline{A}) + \nu_{\text{IFS}}(\overline{A}) \leq 1. \]

The fundamental concept of IFS has massively improved than FS for handling awkward and unreliable information. Various applications are diagnosed in the circumstances of many fields. For sample, bipolar soft sets3, geometric operators4, generalized operators5, simple operators6, simple measures7, measures using Hausdorff distance8, similarity measures9, Bonferroni operators10, generalized Heronian operators11, Heronian operators12, interval-valued IFSs13, cubic IFSs14 and generalized cubic IFSs15. Cuong16,17 diagnosed the main theory of picture FS (PFS) and its applications. The main PFS is very feasible and flexible because of its shape, the prominent condition of PFS is described: \[ 0 \leq \mu_{\text{PFS}}(\overline{A}) + \nu_{\text{PFS}}(\overline{A}) \leq 1. \]

The fundamental concept of PFS has massively improved over IFS and FS for handling awkward and unreliable information. Various applications are diagnosed in the circumstances of many fields of decision-making strategy18–20 and operators21–24.

FS has a huge number of implementations in the circumstances of medical diagnoses, pattern recognition, clustering analysis, and networking systems. But in various places, the theory of FS has also faced a lot of troubles. For example, if the owner of some well-known company wants to lunch a novel sort of various software in a market, they have given two sorts of information related to each software, called name and production date of the software. For handling the above-cited information, the novel theory of complex FS (CFS)25 was diagnosed as a new strategy for managing genuine life ambiguities. Further, Alkouri and Salleh26 introduced the strategy.
of complex IFS (CIFS) as a helper for scholars, those who faced complications in selecting the best option during the decision-making process. CIFS has its level for handling awkward and ambiguous information, certain people have considered the theory of CIFS for utilizing it in the circumstances of different fields, for instance, Ali et al.27 diagnosed the complex intuitionistic fuzzy soft sets. Further, the concept of the relationships among any two CIFS was invented by Jan et al.28, the mathematical shape of complex interval-valued IFSs was diagnosed by Garg and Rani29 and the theory of group utilized in the region of CIFSs was explored by Gulzar et al.30. Akram et al.31 diagnosed the main theory of complex PFS (CPFS) and its applications. The main CPFS is very feasible and flexible because of its shape, the prominent condition of CPFS is described: 0 ≤ L_α(ϕ) + A_β(ϕ) + R_γ(ϕ) ≤ 1 and 0 ≤ L_α(ϕ) + A_β(ϕ) + R_γ(ϕ) ≤ 1. The fundamental concept of CPFS has massively improved over CIFS and CFS for handling awkward and unreliable information. Various applications are diagnosed in the circumstances of many fields of decision-making strategy. To explain the information in the above paragraph, we noticed that the prevailing information has the following major dilemmas:

1. How do we develop new and more effective ideas?
2. How do we develop a superior shape of aggregation operators, used for evaluating the collection of information?
3. How do we evaluate the beneficial preference from the collection of information?

Therefore, the main influence of this theory is to find the solution to the above dilemmas with the help of diagnosed power aggregation operators based on CPFS information.

Additionally, Molodtsov32 introduced the theory of soft set (SS) by extending the theory of FS and because of their shape all scholars have employed it in different fields, for instance, the theory of fuzzy SS was invented by Maji et al.34. Further, the intuitionistic fuzzy SS was discovered by Maji et al.35 and the interval-valued intuitionistic fuzzy SS was explored by Jiang et al.36. The generalized intuitionistic fuzzy SS was invented by Agarwal et al.37. Similarly, the theory of intuitionistic fuzzy SS and its application in decision-making was discovered by Jiang et al.38. The theory of power aggregation operators for IFS was discovered by Xu39. Further, Rani and Garg40 diagnosed the power aggregation operators for CIFSSs and CPFSs. In various situations, the theory of CIFSS information has failed because of many complications, for instance, if someone proved information in the shape of yes, abstinence, and no against the value of parameters, where the value of yes, abstinence, and no in the shape of a complex number, then the theory of complex intuitionistic fuzzy soft sets have been invalid. For evaluating the above-complicated situations, the theory of CPFS and its operational laws are the parts of this manuscript. From the above-cited theory, we also clear that every theory has its limitations because of its structures. Similarly, all prevailing theories have a lot of benefits, keeping the benefits of the SSs and CPFSs, the main contribution of this analysis is to explore the well-known concept of CPFS setting and their laws. The major theme of this analysis is described with the help of various points:

1. To pioneer the theory of CPFS information and evaluated their major algebraic laws, score value, and accuracy values.
2. To present the theory of CPFS operator, CPFSWPA, CPFSOWPA, CPFSWG, CPFSWPG, and CPFSOWPG operators, and diagnosed their particular cases of the invented approaches.
3. To diagnose some real-life situations, we evaluate a MADM tool under the consideration of diagnosed operators to find the best option from the family of decisions.
4. To show the supremacy and feasibility of the diagnosed operators with the help of sensitive analysis and geometrical representations.

The major consequence of this organization is the shape: In section Preliminaries, we highlighted various principles of CIFSSs and their feasible and dominant algebraic laws. Additionally, we also recall the basic idea of PAOs and the generalized t-norm (TN) and t-conorm (TCN). In section “Complex picture fuzzy soft settings”, we diagnosed the mathematical concept of CPFS operator, CPFSWPA, CPFSOWPA, CPFSWG, CPFSWPG, and CPFSOWPG. Moreover, the major results and their particular investigation of the invented approaches are also deliberated. In section “Power aggregation operators under CPSF informationPower aggregation operators under CPSF information”, the major results and their particular investigation of the invented approaches are also deliberated. Additionally, in the consideration of diagnosed operators using CPFS information, we illustrated a MADM tool to find the best option from the family of decisions. Finally, we have shown the supremacy and feasibility of the diagnosed operators with the help of sensitive analysis and geometrical representations. Section “Conclusion” contains various concluding remarks.

Preliminaries

The fundamental theme of this section is to highlight various principles of CIFSSs and their feasible and dominant algebraic laws. Additionally, we also recall the basic idea of PAOs and the generalized TN and TCN. Where the mathematical term \( N \), proves universal sets and the truth, abstinence, and facility grades are demonstrated in the shape: \( N_{\alpha}(\phi) = N_{\alpha}(\phi)e^{2\pi(i\phi)} \), \( A_{\beta}(\phi) = A_{\beta}(\phi)e^{2\pi(i\phi)} \), and \( R_{\gamma}(\phi) = R_{\gamma}(\phi)e^{2\pi(i\phi)} \).

Definition 1 27 In the presence of the universal set \( N \), the CIFSS \( N_{CIFS} \) is organized in the structure:
$$\Sigma_{\text{CIFS-}c_{y}} = \left\{ (\mathcal{M}_{\Sigma_{y}}(\varphi), \mathcal{M}_{\Sigma_{z}}(\varphi)) : \varphi \in \mathbb{F} \right\}$$

(1)

With $0 \leq \mathcal{M}_{R_{y}}(\varphi) + \mathcal{M}_{R_{z}}(\varphi) \leq 1$ and $0 \leq \mathcal{M}_{I_{y}}(\varphi) + \mathcal{M}_{I_{z}}(\varphi) \leq 1$. Further, $\Sigma_{\Sigma_{y}}(\varphi) = \mathcal{R}_{R_{y}}(\varphi)e^{2\pi(\mathcal{R}_{I_{y}}(\varphi))}$, represented the neutral grade. The mathematical organization $\Sigma_{\text{CIFS-}i_{t}e_{i}} = (\mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})}, \mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})})$, $i = 1, 2, \ldots, n$, represented the CIFSNs.

**Definition 2** Considered any two $\Sigma_{\text{IF-}i_{t}e_{i}} = (\mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})}, \mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})})$, $i = 1, 2$, then

$$\Sigma_{\text{IF-}i_{t}e_{i}} = \left( \begin{array}{c}
\mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})}, \\
\mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})}
\end{array} \right)$$

(2)

$$\Sigma_{\text{IF-}i_{t}e_{i}} = \left( \begin{array}{c}
\mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})}, \\
\mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})}
\end{array} \right)$$

(3)

$$\sigma \Sigma_{\text{IF-}i_{t}e_{i}} = \left( \begin{array}{c}
\mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})}, \\
\mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})}
\end{array} \right)$$

(4)

$$\sigma \Sigma_{\text{IF-}i_{t}e_{i}} = \left( \begin{array}{c}
\mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})}, \\
\mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})}
\end{array} \right)$$

(5)

**Definition 3** Considered any two $\Sigma_{\text{IF-}i_{t}e_{i}} = (\mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})}, \mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})})$, $i = 1, 2$, the score value (SV) and accuracy value (AV) are simplified by:

$$\mathfrak{S} \left( \Sigma_{\text{IF-}i_{t}e_{i}} \right) = |\mathcal{M}_{R_{i_{t}}} - \mathcal{M}_{I_{i_{t}}} + \mathcal{M}_{I_{i_{t}}} - \mathcal{M}_{I_{i_{t}}}|, \mathfrak{S} \left( \Sigma_{\text{IF-}i_{t}e_{i}} \right) \in [0, 1]$$

(6)

$$\mathfrak{H} \left( \Sigma_{\text{IF-}i_{t}e_{i}} \right) = |\mathcal{M}_{R_{i_{t}}} + \mathcal{M}_{I_{i_{t}}} + \mathcal{M}_{I_{i_{t}}} + \mathcal{M}_{I_{i_{t}}}|, \mathfrak{H} \left( \Sigma_{\text{IF-}i_{t}e_{i}} \right) \in [0, 1]$$

(7)

Considered any two $\Sigma_{\text{CIFS-}i_{t}e_{i}} = (\mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})}, \mathcal{M}_{R_{i_{t}}}e^{2\pi(\mathcal{M}_{I_{i_{t}}})})$ and $\Sigma_{\text{CIFS-}i_{t}e_{i}}$, then

1. When $\mathfrak{S} \left( \Sigma_{\text{CIFS-}i_{t}e_{i}} \right) > \mathfrak{S} \left( \Sigma_{\text{CIFS-}i_{t}e_{i}} \right)$, then $\Sigma_{\text{CIFS-}i_{t}e_{i}} > \Sigma_{\text{CIFS-}i_{t}e_{i}}$;
2. When $\mathfrak{S} \left( \Sigma_{\text{CIFS-}i_{t}e_{i}} \right) < \mathfrak{S} \left( \Sigma_{\text{CIFS-}i_{t}e_{i}} \right)$, then $\Sigma_{\text{CIFS-}i_{t}e_{i}} < \Sigma_{\text{CIFS-}i_{t}e_{i}}$;
3. When $\mathfrak{S} \left( \Sigma_{\text{CIFS-}i_{t}e_{i}} \right) = \mathfrak{S} \left( \Sigma_{\text{CIFS-}i_{t}e_{i}} \right)$, then.

(i) When $\mathfrak{H} \left( \Sigma_{\text{CIFS-}i_{t}e_{i}} \right) > \mathfrak{H} \left( \Sigma_{\text{CIFS-}i_{t}e_{i}} \right)$, then $\Sigma_{\text{CIFS-}i_{t}e_{i}} > \Sigma_{\text{CIFS-}i_{t}e_{i}}$;
(ii) When $\mathfrak{H} \left( \Sigma_{\text{CIFS-}i_{t}e_{i}} \right) < \mathfrak{H} \left( \Sigma_{\text{CIFS-}i_{t}e_{i}} \right)$, then $\Sigma_{\text{CIFS-}i_{t}e_{i}} < \Sigma_{\text{CIFS-}i_{t}e_{i}}$;
(iii) When $\mathfrak{H} \left( \Sigma_{\text{CIFS-}i_{t}e_{i}} \right) = \mathfrak{H} \left( \Sigma_{\text{CIFS-}i_{t}e_{i}} \right)$, then $\Sigma_{\text{CIFS-}i_{t}e_{i}} = \Sigma_{\text{CIFS-}i_{t}e_{i}}$.

**Definition 4** Considered any collection of attributes $\Sigma_{\text{PI-}i}$, $i = 1, 2, \ldots, n$, the PAO is simplified by:

$$PA \left( \Sigma_{\text{PI-}1}, \Sigma_{\text{PI-}2}, \ldots, \Sigma_{\text{PI-}n} \right) = \sum_{i=1}^{n} \left( 1 + \mathfrak{T} \left( \Sigma_{\text{PI-}i} \right) \right) / \sum_{i=1}^{n} \left( 1 + \mathfrak{T} \left( \Sigma_{\text{PI-}i} \right) \right)$$

(8)

The mathematical term $\mathfrak{T} \left( \Sigma_{\text{PI-}i} \right) = \sum_{k=1}^{n} \sum_{k \neq i} \mathfrak{T} \left( \Sigma_{\text{PI-}i} \right)$, stated the support for $\Sigma_{\text{PI-}i}$ and $\Sigma_{\text{PI-}k}$.

Particularized by:

$$\text{Sup} \left( \Sigma_{\text{PI-}i}, \Sigma_{\text{PI-}k} \right) = 1 - d \left( \Sigma_{\text{PI-}i}, \Sigma_{\text{PI-}k} \right)$$

(9)
where \( d(\overline{p}_{l-1}, \overline{p}_{l-k}) \) acknowledged the measure for \( \overline{p}_{l-1} \) and \( \overline{p}_{l-k} \), with a technique that:

1. \( \text{Sup}(\overline{p}_{l-1}, \overline{p}_{l-k}) \in [0, 1] \);
2. \( \text{Sup}(\overline{p}_{l-1}, \overline{p}_{l-k}) = \text{Sup}(\overline{p}_{l-k}, \overline{p}_{l-1}) \);
3. If \( d(\overline{p}_{l-1}, \overline{p}_{l-k}) \leq d(\overline{p}_{l-1}, \overline{p}_{l-q}) \) then \( \text{Sup}(\overline{p}_{l-1}, \overline{p}_{l-k}) \geq \text{Sup}(\overline{p}_{l-1}, \overline{p}_{l-q}) \).

**Definition 5** A mapping \( \overline{F} : [0, 1] \times [0, 1] \rightarrow [0, 1] \) is particularized TN when \( \overline{F} \), justified the boundary, monotonocity, commutative, and associativity techniques. Where \( \overline{\varphi}(\varphi, \overline{\pi}) = 1 - \overline{F}(1 - \overline{\varphi}, 1 - \overline{\pi}) \), diagnosed the TN and the general shape of Archimedeian TN and TCN is diagnosed by:

\[
\overline{F}(\varphi, \overline{\pi}) = \begin{cases} 
\overline{\varphi}^{-1} & (\overline{\varphi}(\varphi) + \overline{\pi}(\overline{\pi})) \\
\overline{\pi}^{-1} & (\overline{\varphi}(\varphi) + \overline{\pi}(\overline{\pi})) 
\end{cases}
\]

and \( \overline{\varphi}(\varphi, \overline{\pi}) = \overline{\varphi}^{-1} (\overline{\varphi}(\varphi) + \overline{\pi}(\overline{\pi})) \), based on the continuous increasing (or decreasing) function with \( \overline{\varphi}(1) = 0, \overline{\varphi}(0) = 0 \) and \( \overline{\varphi}(1 - \overline{\varphi}) = (1 - \overline{\varphi}) \).

**Complex picture fuzzy soft settings**

The major theme of this analysis is to suggest a new theory in the form of CPFS information and invented their major algebraic laws, score value, and accuracy value. The mathematical form of the CPFS set includes three main functions, called supporting, abstinence, and supporting against terms with a prominent characteristic that is the sum of the triplet will lie in the unit interval.

**Definition 6** In the presence of the universal set \( \overline{X} \), the CPFS \( \overline{\varphi}_{\overline{CIFS}} \) is organized in the structure:

\[
\overline{\varphi}_{\overline{CIFS}} \in \{ (\overline{M}_{I_{1}}, \overline{A}_{I_{1}}, \overline{N}_{I_{1}}) : \varphi \in \overline{X} \}.
\]

With \( 0 \leq \overline{M}_{I_{1}}(\varphi) + \overline{A}_{I_{1}}(\varphi) + \overline{N}_{I_{1}}(\varphi) \leq 1 \) and \( 0 \leq \overline{M}_{I_{2}}(\varphi) + \overline{A}_{I_{2}}(\varphi) + \overline{N}_{I_{2}}(\varphi) \leq 1 \). Further, \( \overline{M}_{I_{1}}(\varphi) = \overline{M}_{I_{2}}(\varphi) = \overline{M}_{I_{1}}(\varphi)\overline{e}^{2\pi(\overline{M}_{I_{1}}(\varphi) - \overline{A}_{I_{1}}(\varphi) + \overline{N}_{I_{1}}(\varphi)))} \), represented the neutral grade. The mathematical organization \( \overline{\varphi}_{\overline{CIFS}} = \{ (\overline{M}_{I_{1}}e^{2\pi(\overline{M}_{I_{1}})}, \overline{A}_{I_{1}}e^{2\pi(\overline{A}_{I_{1}})}, \overline{N}_{I_{1}}e^{2\pi(\overline{N}_{I_{1}})}) \} \), \( i = 1, 2, \ldots, n \), represented the CPFSNs.

**Definition 7** Considered any two \( \overline{\varphi}_{\overline{CIFS}} \in \{ (\overline{M}_{I_{1}}, \overline{A}_{I_{1}}, \overline{N}_{I_{1}}) : \varphi \in \overline{X} \} \), \( \overline{M}_{I_{1}} \leq \overline{M}_{I_{2}}, \overline{A}_{I_{1}} \leq \overline{A}_{I_{2}}, \overline{N}_{I_{1}} \leq \overline{N}_{I_{2}} \), then

\[
\overline{\varphi}_{\overline{CIFS}} \otimes \overline{\varphi}_{\overline{CIFS}} = \left\{ \left( \overline{M}_{I_{1} + I_{2}}, \overline{A}_{I_{1} + I_{2}}, \overline{N}_{I_{1} + I_{2}} \right) : \overline{M}_{I_{1}}, \overline{A}_{I_{1}}, \overline{N}_{I_{1}} \leq \overline{M}_{I_{2}}, \overline{A}_{I_{2}}, \overline{N}_{I_{2}} \right\}
\]

\[
\overline{\varphi}_{\overline{CIFS}} \oplus \overline{\varphi}_{\overline{CIFS}} = \left\{ \left( \overline{M}_{I_{1} + I_{2}}, \overline{A}_{I_{1} + I_{2}}, \overline{N}_{I_{1} + I_{2}} \right) : \overline{M}_{I_{1}}, \overline{A}_{I_{1}}, \overline{N}_{I_{1}} \leq \overline{M}_{I_{2}}, \overline{A}_{I_{2}}, \overline{N}_{I_{2}} \right\}
\]

\[
\overline{\varphi}_{\overline{CIFS}} \ominus \overline{\varphi}_{\overline{CIFS}} = \left\{ \left( \overline{M}_{I_{1}}, \overline{A}_{I_{1}} - \overline{A}_{I_{2}}, \overline{N}_{I_{1}} - \overline{N}_{I_{2}} \right) : \overline{M}_{I_{1}}, \overline{A}_{I_{1}}, \overline{N}_{I_{1}} \leq \overline{M}_{I_{2}}, \overline{A}_{I_{2}}, \overline{N}_{I_{2}} \right\}
\]

\[
\overline{\varphi}_{\overline{CIFS}} \odot \overline{\varphi}_{\overline{CIFS}} = \left\{ \left( \overline{M}_{I_{1}}, \overline{A}_{I_{1}} - \overline{A}_{I_{2}}, \overline{N}_{I_{1}} - \overline{N}_{I_{2}} \right) : \overline{M}_{I_{1}}, \overline{A}_{I_{1}}, \overline{N}_{I_{1}} \leq \overline{M}_{I_{2}}, \overline{A}_{I_{2}}, \overline{N}_{I_{2}} \right\}
\]

**Definition 8** Considered any two \( \overline{\varphi}_{\overline{CIFS}} = \{ (\overline{M}_{I_{1}}e^{2\pi(\overline{M}_{I_{1}})}, \overline{A}_{I_{1}}e^{2\pi(\overline{A}_{I_{1}})}, \overline{N}_{I_{1}}e^{2\pi(\overline{N}_{I_{1}})}) \} \), \( i = 1, 2 \), the score value (SV) and accuracy value (AV) are simplified by:

\[
\overline{F}(\overline{\varphi}_{\overline{CIFS}}) = \{ \overline{M}_{I_{1}} - \overline{A}_{I_{1}} - \overline{A}_{I_{2}} - \overline{A}_{I_{2}} \} \overline{\varphi}_{\overline{CIFS}} \in [-1, 1]
\]
\[
\mathcal{H}(\underline{\Sigma}_{\text{CIF}}-it) = \left| \mathfrak{M}_{R_{1\Sigma}} + A_{R_{1\Sigma}} + \mathfrak{M}_{R_{1\Sigma}} + A_{R_{1\Sigma}} + \mathfrak{N}_{1\Sigma} \right|, \quad \mathcal{H}(\underline{\Sigma}_{\text{CIF}}-it) \in [0, 1]
\] (16)

Considered any two \( \underline{\Sigma}_{\text{CIF}}-it = (\mathfrak{M}_{R_{1\Sigma}}e^{i2\pi(\mathfrak{M}_{1\Sigma})}, A_{R_{1\Sigma}}e^{i2\pi(A_{1\Sigma})}, \mathfrak{N}_{R_{1\Sigma}}e^{i2\pi(\mathfrak{N}_{1\Sigma})}) \) and \( \underline{\Sigma}_{\text{CIF}}-it = (\mathfrak{M}_{R_{2\Sigma}}e^{i2\pi(\mathfrak{M}_{2\Sigma})}, A_{R_{2\Sigma}}e^{i2\pi(A_{2\Sigma})}, \mathfrak{N}_{R_{2\Sigma}}e^{i2\pi(\mathfrak{N}_{2\Sigma})}) \), then

1. When \( \mathcal{S}(\underline{\Sigma}_{\text{CIF}}-it) > \mathcal{S}(\underline{\Sigma}_{\text{CIF}}-it) \), then \( \underline{\Sigma}_{\text{CIF}}-it > \underline{\Sigma}_{\text{CIF}}-it \);  
2. When \( \mathcal{S}(\underline{\Sigma}_{\text{CIF}}-it) < \mathcal{S}(\underline{\Sigma}_{\text{CIF}}-it) \), then \( \underline{\Sigma}_{\text{CIF}}-it < \underline{\Sigma}_{\text{CIF}}-it \);  
3. When \( \mathcal{S}(\underline{\Sigma}_{\text{CIF}}-it) = \mathcal{S}(\underline{\Sigma}_{\text{CIF}}-it) \), then.

Further, we try to construct the general shape of certain algebraic laws based on CPFSSs.

**Definition 9** Considered any two \( \underline{\Sigma}_{\text{CIF}}-it = (\mathfrak{M}_{R_{1\Sigma}}e^{i2\pi(\mathfrak{M}_{1\Sigma})}, A_{R_{1\Sigma}}e^{i2\pi(A_{1\Sigma})}, \mathfrak{N}_{R_{1\Sigma}}e^{i2\pi(\mathfrak{N}_{1\Sigma})}) \), \( \mathfrak{A}_{1, 2} \) with \( \hat{\sigma} > 0 \), then

\[
\underline{\Sigma}_{\text{CIF}}-it \oplus \underline{\Sigma}_{\text{CIF}}-it = \left( \mathfrak{M}_{R_{1\Sigma}}e^{i2\pi(\mathfrak{M}_{1\Sigma})}, A_{R_{1\Sigma}}e^{i2\pi(A_{1\Sigma})}, \mathfrak{N}_{R_{1\Sigma}}e^{i2\pi(\mathfrak{N}_{1\Sigma})} \right), \quad \mathfrak{A}_{1, 2}
\] (17)

\[
\underline{\Sigma}_{\text{CIF}}-it \ominus \underline{\Sigma}_{\text{CIF}}-it = \left( \mathfrak{M}_{R_{1\Sigma}}e^{i2\pi(\mathfrak{M}_{1\Sigma})}, A_{R_{1\Sigma}}e^{i2\pi(A_{1\Sigma})}, \mathfrak{N}_{R_{1\Sigma}}e^{i2\pi(\mathfrak{N}_{1\Sigma})} \right), \quad \mathfrak{A}_{1, 2}
\] (18)

**Theorem 1** Considered any two \( \underline{\Sigma}_{\text{CIF}}-it = (\mathfrak{M}_{R_{1\Sigma}}e^{i2\pi(\mathfrak{M}_{1\Sigma})}, A_{R_{1\Sigma}}e^{i2\pi(A_{1\Sigma})}, \mathfrak{N}_{R_{1\Sigma}}e^{i2\pi(\mathfrak{N}_{1\Sigma})}) \), \( \mathfrak{A}_{1, 2} \) with \( \hat{\sigma}_{1} > 0, i = 1, 2 \), then

\[
\underline{\Sigma}_{\text{CIF}}-it = \left( \mathfrak{M}_{R_{1\Sigma}}e^{i2\pi(\mathfrak{M}_{1\Sigma})}, A_{R_{1\Sigma}}e^{i2\pi(A_{1\Sigma})}, \mathfrak{N}_{R_{1\Sigma}}e^{i2\pi(\mathfrak{N}_{1\Sigma})} \right), \quad \mathfrak{A}_{1, 2}
\] (19)
1. $\Sigma_{CIF-11} \oplus \Sigma_{CIF-12} = \Sigma_{CIF-12} \oplus \Sigma_{CIF-11};$
2. $\Sigma_{CIF-11} \otimes \Sigma_{CIF-12} = \Sigma_{CIF-12} \otimes \Sigma_{CIF-11};$
3. $\sigma \left( \Sigma_{CIF-11} \oplus \Sigma_{CIF-12} \right) = \sigma \Sigma_{CIF-11} \oplus \sigma \Sigma_{CIF-12};$
4. $\left( \Sigma_{CIF-11} \otimes \Sigma_{CIF-12} \right)^\sigma = \Sigma_{CIF-11}^\sigma \otimes \Sigma_{CIF-12}^\sigma;$
5. $\sigma_1 \Sigma_{CIF-11} \oplus \sigma_2 \Sigma_{CIF-11} = \left( \sigma_1 + \sigma_2 \right) \Sigma_{CIF-11};$
6. $\Sigma_{CIF-11}^\sigma \otimes \Sigma_{CIF-11}^\sigma = \Sigma_{CIF-11} \left( \sigma_1 + \sigma_2 \right).$

**Proof** We prove Eqs. (3) and (5), because Parts (1), (2), (4), and (6) are straightforward. Assume that $\sigma \left( \Sigma_{CIF-11} \oplus \Sigma_{CIF-12} \right),$ then

$$\begin{align*}
\sigma \left( \Sigma_{CIF-11} \oplus \Sigma_{CIF-12} \right) &= \sigma \left( \Sigma_{CIF-11} + \Sigma_{CIF-12} \right) = \\
&= \left( \Sigma_{CIF-11} + \Sigma_{CIF-12} \right)^\sigma = \\
&= \left( \Sigma_{CIF-11} \oplus \Sigma_{CIF-12} \right)^\sigma.
\end{align*}$$

Hence, $\sigma \left( \Sigma_{CIF-11} \oplus \Sigma_{CIF-12} \right) = \sigma \Sigma_{CIF-11} \oplus \sigma \Sigma_{CIF-12}.$ Assume that $\sigma \left( \Sigma_{CIF-11} \oplus \Sigma_{CIF-12} \right), then
\[ \sigma_1^{CIF-11} \oplus \sigma_2^{CIF-11} = \left( f^{-1} \left( \sigma_1 \oplus R_{1i} \right) e^{i2\pi \left( f^{-1} \left( \sigma_1 \oplus R_{1i} \right) \right)} \right) \left( f^{-1} \left( \sigma_1 \oplus A_{1i} \right) e^{i2\pi \left( f^{-1} \left( \sigma_1 \oplus A_{1i} \right) \right)} \right) \left( f^{-1} \left( \sigma_1 \oplus M_{1i} \right) e^{i2\pi \left( f^{-1} \left( \sigma_1 \oplus M_{1i} \right) \right)} \right) \]

\[ \oplus \left( f^{-1} \left( \sigma_2 \oplus R_{1i} \right) e^{i2\pi \left( f^{-1} \left( \sigma_2 \oplus R_{1i} \right) \right)} \right) \left( f^{-1} \left( \sigma_2 \oplus A_{1i} \right) e^{i2\pi \left( f^{-1} \left( \sigma_2 \oplus A_{1i} \right) \right)} \right) \left( f^{-1} \left( \sigma_2 \oplus M_{1i} \right) e^{i2\pi \left( f^{-1} \left( \sigma_2 \oplus M_{1i} \right) \right)} \right) \]

\[ = \left( f^{-1} \left( \sigma_1 \oplus M_{1i} \right) e^{i2\pi \left( f^{-1} \left( \sigma_1 \oplus M_{1i} \right) \right)} \right) \left( f^{-1} \left( \sigma_1 \oplus A_{1i} \right) e^{i2\pi \left( f^{-1} \left( \sigma_1 \oplus A_{1i} \right) \right)} \right) \left( f^{-1} \left( \sigma_1 \oplus R_{1i} \right) e^{i2\pi \left( f^{-1} \left( \sigma_1 \oplus R_{1i} \right) \right)} \right) \]

Hence, \( \sigma_1^{CIF-11} \oplus \sigma_2^{CIF-11} = \sigma_1^{CIF-11} \).

**Power aggregation operators under CPSF information**

The major theme of this analysis is to investigate the consideration of power aggregation operator using generalized t-norm and t-conorm and CPSF information, we diagnosed the mathematical concept of CPFSPA, CPF-SWPA, CPFSOWPA, CPFSPG, CPFSWPG, CPFSOWPG. Moreover, the major results and their particular investigation of the invented approaches are also deliberated. Various important techniques called averaging, Einstein, and Hamacher operators are investigated using \( \widetilde{g}(\varphi) = -\log(\varphi) \), \( \widetilde{g}(\varphi) = \log \left( \frac{2}{\varphi} \right) \), \( \varphi \neq 0 \), and \( \widetilde{g}(\varphi) = \log \left( \sigma + \frac{1 - \sigma}{\varphi} \right) \), \( \sigma \in (0, \infty) \), \( \varphi \neq 0 \). Throughout this manuscript, the term \( \overline{\sigma_i} \) represents the CFSPNs with \( \overline{\sigma_i} > 0 \).

**Definition 10** When \( \overline{\sigma_i} \in \overline{\Sigma} \), then the CPFPA operator is simplified by:

\[ CFSPA : \overline{\Sigma} \rightarrow \overline{\Sigma} \]

by

\[ CFSPA \left( \overline{\sigma_{CIF-11}}, \overline{\sigma_{CIF-12}}, \ldots, \overline{\sigma_{CIF-nm}} \right) = \oplus_1^n \left( M_{1i} \oplus \left( \overline{\sigma_i}, \overline{\Sigma} \right) \right) \]

where \( \overline{M_i} = \frac{(1+\overline{\sigma_i})}{\sum_{i=1}^n (1+\overline{\sigma_i})} \), \( \overline{\Sigma_i} = \frac{(1+\overline{\sigma_i})}{\sum_{i=1}^n (1+\overline{\sigma_i})} \), and \( \overline{\Sigma_i} = \sum_{k=1}^m \text{Sup} \left( \overline{\sigma_{PF-it}}, \overline{\sigma_{PF-kt}} \right) \), \( \overline{\sigma_{PF-it}} \), \( \overline{\sigma_{PF-kt}} \), simplified the support for \( \overline{\sigma_{PF-it}} \) and \( \overline{\sigma_{PF-kt}} \).

**Theorem 2** Using the information in Eq. (21), we get
Hence Eq. (22) is held for $m = k \leq n$. Then for $k = 1$, then using the information in Eq. (22), we get

$$\text{CPFSPA} \left( \overline{\Sigma_{\text{CIF} - 11}}, \overline{\Sigma_{\text{CIF} - 12}}, \ldots, \overline{\Sigma_{\text{CIF} - 1m}} \right) = \Theta_{t=1}^{n} \left( \overline{M_{t} \Sigma_{P1-t}} \right)$$

Similarly, assume that $m = 1$, then we get

$$\text{CPFSPA} \left( \overline{\Sigma_{\text{CIF} - 11}}, \overline{\Sigma_{\text{CIF} - 12}}, \ldots, \overline{\Sigma_{\text{CIF} - n1}} \right) = \Theta_{t=1}^{n} \left( \overline{N_{t} \Sigma_{P1-t}} \right)$$

Hence Eq. (22) is held for $m = n = 1$. Additionally, we suppose that Eq. (22) is also held for $m = k_{1}, n = k_{2} + 1$ and $m = k_{1} + 1, n = k_{2}$, then for $m = k_{1} + 1, n = k_{2} + 1$, we have

$$\Theta_{t=1}^{k_{1}+1} \left( \overline{M_{t} \Theta_{i=1}^{k_{2}+1} \left( \overline{N_{t} \Sigma_{P1-t}} \right)} \right) = \Theta_{t=1}^{k_{1}} \left( \overline{M_{t} \Theta_{i=1}^{k_{2}+1} \left( \overline{N_{t} \Sigma_{P1-t}} \right)} \right) \oplus \Theta_{t=1}^{k_{1}+1} \left( \overline{N_{t} \Sigma_{P1-t}} \right)$$
Hence Eq. (22) is held for $m = k_1 + 1, n = k_2 + 1$, therefore it is also true for all positive integer $m, n$.

Property 1 (Idempotency) When $\Sigma_{\text{CIF}} = \Sigma_{\text{CIF}},$ then

$$\Sigma_{\text{CIF}} \Sigma_{\text{CIF}} - \text{it} = \Sigma_{\text{CIF}},$$

$$\Sigma_{\text{CPFSPA}} \Sigma_{\text{CPFSPA}} - 11, \Sigma_{\text{CPFSPA}} - 12, \ldots, \Sigma_{\text{CPFSPA}} - \text{nn} = \Sigma_{\text{CPFSPA}} \Sigma_{\text{CPFSPA}} - 11, \Sigma_{\text{CPFSPA}} - 12, \ldots, \Sigma_{\text{CPFSPA}} - \text{nn} = \Sigma_{\text{CPFSPA}}$$

Proof When $\Sigma_{\text{CIF}} = \Sigma_{\text{CIF}},$ then based on Eq. (22), we get

$$\Sigma_{\text{CPFSPA}} \Sigma_{\text{CPFSPA}} - 11, \Sigma_{\text{CPFSPA}} - 12, \ldots, \Sigma_{\text{CPFSPA}} - \text{nn}$$
Property 2 When $\Sigma_{CIF-11}$, $\Sigma_{CIF-12}$, and $\Sigma_{CIF}$ be any two CPFSNs, then

$$\text{CPFSPA}(\Sigma_{CIF-11}, \Sigma_{CIF-12}, \Sigma_{CIF}) = \text{CPFSPA}(\Sigma_{CIF-11}, \Sigma_{CIF-12}, \Sigma_{CIF})$$

Proof When $\Sigma_{CIF-11}$ and $\Sigma_{CIF}$ be any two CPFSNs, then

$$\Sigma_{CIF-11} \oplus \Sigma_{CIF} = \left( \sum_{t=1}^{m} M_t \sum_{i=1}^{n} N_i \begin{pmatrix} f^{-1} \left( f (M_{R,t}) + f (M_{R}) \right) e^{i2\pi \left( f^{-1} \left( f (M_{R,t}) + f (M_{R}) \right) \right)} \\
                        g^{-1} \left( g (A_{R,t}) + g (A_R) \right) e^{i2\pi \left( g^{-1} (g (A_{R,t}) + g (A_R)) \right)} \\
                        g^{-1} \left( g (N_{R,t}) + g (N_R) \right) e^{i2\pi \left( g^{-1} (g (N_{R,t}) + g (N_R)) \right)} \end{pmatrix} \right)$$
Prove that

\[ \text{CPFSPA} \left( \sigma \mathbf{d} \cdot \mathbf{CIF} \cdot \mathbf{d} \cdot A_{R, e} \mathbf{d}^{-1} \cdot (\mathbf{m}_{R, e} + \mathbf{m}_{R}) \right) \]

\[ \oplus \mathbf{m}_{R} e^{2\pi(\mathbf{N}_{CIF} \cdot A_{i} \cdot \mathbf{d}^{2} + \mathbf{N}_{CIF} \cdot A_{i} \cdot \mathbf{d})} \]

\[ = \text{CPFSPA} \left( \mathbf{d} \cdot \mathbf{CIF} \cdot \mathbf{d} \cdot A_{R, e} \mathbf{d}^{-1} \cdot (\mathbf{m}_{R, e} + \mathbf{m}_{R}) \right) \]

\[ = \sigma \text{CPFSPA} \left( \mathbf{d} \cdot \mathbf{CIF} \cdot \mathbf{d} \cdot A_{R, e} \mathbf{d}^{-1} \cdot (\mathbf{m}_{R, e} + \mathbf{m}_{R}) \right) \]

\[ (25) \]
Property 4 We know that $\tilde{\sigma} > 0$, then

$$
\sigma \overline{\Sigma}_{\text{CIF}-it} = \left( f^{-1}(\tilde{\sigma} \int \mathbb{M}_{R,t}) e^{i2\pi f^{-1}(\tilde{\sigma} \int \mathbb{M}_{R,t})} \right) \left( g^{-1}(\sigma g \mathbb{A}_{R,t}) e^{i2\pi g^{-1}(\sigma g \mathbb{A}_{R,t})} \right)
$$

then,

$$
\text{CPFSPA} \left( \sigma \overline{\Sigma}_{\text{CIF}-11}, \sigma \overline{\Sigma}_{\text{CIF}-12}, \ldots, \sigma \overline{\Sigma}_{\text{CIF}-nm} \right)
$$

$$
= \left( f^{-1} \left( \sum_{t=1}^{m} \mathbb{M}_{t} \left( \sum_{i=1}^{n} \tilde{N}_{i} \mathbb{M}_{R,t} \right) \right) e^{i2\pi f^{-1} \left( \sum_{t=1}^{m} \mathbb{M}_{t} \left( \sum_{i=1}^{n} \tilde{N}_{i} \mathbb{M}_{R,t} \right) \right)} \right) \left( g^{-1} \left( \sum_{t=1}^{m} \mathbb{M}_{t} \left( \sum_{i=1}^{n} \tilde{N}_{i} \mathbb{A}_{R,t} \right) \right) e^{i2\pi g^{-1} \left( \sum_{t=1}^{m} \mathbb{M}_{t} \left( \sum_{i=1}^{n} \tilde{N}_{i} \mathbb{A}_{R,t} \right) \right)} \right)
$$

$$
= \left( f^{-1} \left( \sum_{t=1}^{m} \mathbb{M}_{t} \left( \sum_{i=1}^{n} \tilde{N}_{i} \mathbb{M}_{R,t} \right) \right) e^{i2\pi f^{-1} \left( \sum_{t=1}^{m} \mathbb{M}_{t} \left( \sum_{i=1}^{n} \tilde{N}_{i} \mathbb{M}_{R,t} \right) \right)} \right) \left( g^{-1} \left( \sum_{t=1}^{m} \mathbb{M}_{t} \left( \sum_{i=1}^{n} \tilde{N}_{i} \mathbb{A}_{R,t} \right) \right) e^{i2\pi g^{-1} \left( \sum_{t=1}^{m} \mathbb{M}_{t} \left( \sum_{i=1}^{n} \tilde{N}_{i} \mathbb{A}_{R,t} \right) \right)} \right)
$$

$$
= \left( f^{-1} \left( \sigma \int \mathbb{M}_{R,t} \right) e^{i2\pi f^{-1} \left( \sigma \int \mathbb{M}_{R,t} \right)} \right) \left( g^{-1} \left( \sigma g \mathbb{A}_{R,t} \right) e^{i2\pi g^{-1} \left( \sigma g \mathbb{A}_{R,t} \right)} \right)
$$

$$
= \sigma \text{CPFSPA} \left( \overline{\Sigma}_{\text{CIF}-11}, \overline{\Sigma}_{\text{CIF}-12}, \ldots, \overline{\Sigma}_{\text{CIF}-nm} \right).
$$
Property 4 Assume that

\[ \overline{CIF}_{k} = \left( m_{R_{t}}, e^{i2\pi(m_{R_{t}})}, A_{R_{t}}e^{i2\pi(A_{R_{t}})}, R_{t}e^{i2\pi(R_{t})} \right), k = 1, 2, \] and

\[ \overline{CIF}_{k} = \left( m_{R_{t}}, e^{i2\pi(m_{R_{t}})}, A_{R_{t}}e^{i2\pi(A_{R_{t}})}, R_{t}e^{i2\pi(R_{t})} \right), \] then

\[ \overline{CPFSPA \left( \Sigma_{CIF-11} \oplus \Sigma_{CIF-12} \oplus \Sigma_{CIF-nm} \oplus \Sigma_{CIF-nm} \right) = CPFSPA \left( \Sigma_{CIF-11} \oplus \Sigma_{CIF-12} \oplus \Sigma_{CIF-nm} \oplus \Sigma_{CIF-nm} \right) } \]

(26)

Proof Assume \( \overline{CIF}_{k} = \left( m_{R_{t}}, e^{i2\pi(m_{R_{t}})}, A_{R_{t}}e^{i2\pi(A_{R_{t}})}, R_{t}e^{i2\pi(R_{t})} \right) \), then

\[ \overline{CIF}_{k} = \left( m_{R_{t}}, e^{i2\pi(m_{R_{t}})}, A_{R_{t}}e^{i2\pi(A_{R_{t}})}, R_{t}e^{i2\pi(R_{t})} \right) \]

then,

\[ \overline{CPFSPA \left( \Sigma_{CIF-11} \oplus \Sigma_{CIF-12} \oplus \Sigma_{CIF-nm} \oplus \Sigma_{CIF-nm} \right) = CPFSPA \left( \Sigma_{CIF-11} \oplus \Sigma_{CIF-12} \oplus \Sigma_{CIF-nm} \oplus \Sigma_{CIF-nm} \right) } \]

(26)
Important cases of the invented work using the information in Eq. (22) are described here. 

1. Assume \( \hat{g}(\varphi) = -\log(\varphi) \) in Eq. (22), then

\[
\text{CPFSPA}\left(\overline{\Sigma_{\text{CIF}}-11}, \overline{\Sigma_{\text{CIF}}-12}, \ldots, \overline{\Sigma_{\text{CIF}}-nm}\right) \oplus \text{CPFSPA}\left(\overline{\Sigma_{\text{CIF}}-11}, \overline{\Sigma_{\text{CIF}}-12}, \ldots, \overline{\Sigma_{\text{CIF}}-nm}\right)
\]

is equal to

\[
\left\{ \begin{array}{l}
1 - \prod_{t=1}^{m} \left( \prod_{i=1}^{n} \left( 1 - \mathcal{M}_{R_{it}} \right) \right) \\
\prod_{t=1}^{m} \left( \prod_{i=1}^{n} (\mathcal{A}_{R_{it}}) \right) e^{2\pi i} \\
\prod_{t=1}^{m} \left( \prod_{i=1}^{n} (\mathcal{M}_{R_{it}}) \right) e^{2\pi i}
\end{array} \right\}
\]
2. Assume the CPFS Archimedean weighted averaging (CPFS A) operator.

$$\text{CPFSA} (\Sigma_{C_{IF-11}}, \Sigma_{C_{IF-12}}, \ldots, \Sigma_{C_{IF-nm}}) =$$

$$\frac{\prod_{I=1}^{n} \left( \prod_{i=1}^{m} \left( 1 + \mathbf{m}_{R_{x}} \mathbf{N}_{i} \right) \right) - \prod_{I=1}^{n} \left( \prod_{i=1}^{m} \left( 1 - \mathbf{m}_{R_{x}} \mathbf{N}_{i} \right) \right)}{\prod_{I=1}^{n} \left( \prod_{i=1}^{m} \left( 1 + \mathbf{m}_{R_{x}} \mathbf{N}_{i} \right) \right) + \prod_{I=1}^{n} \left( \prod_{i=1}^{m} \left( 1 - \mathbf{m}_{R_{x}} \mathbf{N}_{i} \right) \right)} \mathbf{M}_{I}$$

$$\text{CPFSA} (\Sigma_{C_{IF-11}}, \Sigma_{C_{IF-12}}, \ldots, \Sigma_{C_{IF-nm}}) =$$

$$\frac{2 \prod_{I=1}^{n} \left( \prod_{i=1}^{m} \left( \mathbf{A}_{R_{x}} \mathbf{N}_{i} \right) \right)}{\prod_{I=1}^{n} \left( \prod_{i=1}^{m} \left( 2 - \mathbf{A}_{R_{x}} \mathbf{N}_{i} \right) \right) + \prod_{I=1}^{n} \left( \prod_{i=1}^{m} \left( \mathbf{A}_{R_{x}} \mathbf{N}_{i} \right) \right)} \mathbf{M}_{I}$$

$$\text{CPFSA} (\Sigma_{C_{IF-11}}, \Sigma_{C_{IF-12}}, \ldots, \Sigma_{C_{IF-nm}}) =$$

$$\frac{2 \prod_{I=1}^{n} \left( \prod_{i=1}^{m} \left( \mathbf{M}_{R_{x}} \mathbf{N}_{i} \right) \right)}{\prod_{I=1}^{n} \left( \prod_{i=1}^{m} \left( 2 - \mathbf{M}_{R_{x}} \mathbf{N}_{i} \right) \right) + \prod_{I=1}^{n} \left( \prod_{i=1}^{m} \left( \mathbf{M}_{R_{x}} \mathbf{N}_{i} \right) \right)} \mathbf{M}_{I}$$

Stated the CPFS Einstein weighted averaging (CPFSEWA) operator.

3. When $$\hat{g} (\varphi) = \log \left( \frac{1 - \hat{\sigma}}{\varphi} \right)$$, $$\hat{\sigma} \in (0, \infty), \varphi \neq 0$$ in Eq. (22), then
Stated the CPFS Hamacher weighted averaging (CPFSHWA) operator.

**Definition 11** When \( \mathcal{L}_{\text{CIF}} \in \mathbb{E} \), the CPFSWPA operator is simplified by:

\[
\text{CPFSWPA} : \mathbb{E} \rightarrow \mathbb{E}
\]

by

\[
\text{CPFSWPA} \left( \mathcal{L}_{\text{CIF}−11}, \mathcal{L}_{\text{CIF}−12}, \ldots, \mathcal{L}_{\text{CIF}−nm} \right) = \bigoplus_{i=1}^{m} \left( \mathcal{N}_{i}^{\prime}, \mathcal{L}_{\text{PI}−it} \right)
\]

where

\[
\bar{\mathcal{N}}_{i}^{\prime} = \frac{\hat{\mu}_{i} \left( 1 + \frac{1}{\mathcal{L}_{i}} \right)}{\sum_{k=1}^{m} \hat{\mu}_{k} \left( 1 + \frac{1}{\mathcal{L}_{k}} \right)}, \quad \text{and} \quad \bar{\mathcal{N}} = \sum_{k=1}^{m} \text{Sup} \left( \mathcal{L}_{\text{PI}−it} \mathcal{L}_{\text{PI}−kt} \right),
\]

\[
\bar{\mathcal{L}}_{i} = \sum_{k=1}^{m} \text{Sup} \left( \mathcal{L}_{\text{PI}−it} \mathcal{L}_{\text{PI}−kt} \right), \quad \text{and} \quad \text{Sup} \left( \mathcal{L}_{\text{PI}−it} \mathcal{L}_{\text{PI}−kt} \right), \quad \text{simplified the support for } \mathcal{L}_{\text{PI}−it} \text{ and } \mathcal{L}_{\text{PI}−kt}, \text{ where}
\]

\( \hat{\mu}_{i} \) and \( \eta_{i} \), expressed the weight vector with \( \sum_{i=1}^{n} \hat{\mu}_{i} = 1 \) and \( \sum_{i=1}^{n} \eta_{i} = 1 \).

**Theorem 3** Considering the information in Eq. (30), we get
\[ \text{CPFSWPA} \left( \mathbf{\Sigma}_{\text{CF}-11}, \mathbf{\Sigma}_{\text{CF}-12}, \ldots, \mathbf{\Sigma}_{\text{CF}-nm} \right) \]

\[
= \left( \begin{array}{c}
\frac{1}{\pi} \left( \sum_{t=1}^{m} \mu_t \left( \sum_{i=1}^{n} \tilde{n}_i \left( \mathbf{m}_{R_{it}} \right) \right) \right) e^{-\frac{1}{2} \left( \sum_{t=1}^{m} \mu_t \left( \sum_{i=1}^{n} \tilde{n}_i \left( \mathbf{m}_{R_{it}} \right) \right)^2 \right)}
\end{array} \right) \]

Proven.\] 

**Definition 12** When \( \mathbf{\Sigma}_{\text{CF}-it} \in \mathbb{R} \), then CPFSOWPA operator is simplified by:

\[ \text{CPFSOWPA} : \mathbb{R}^n \rightarrow \mathbb{R} \]

by

\[ \text{CPFSOWPA} \left( \mathbf{\Sigma}_{\text{CF}-11}, \mathbf{\Sigma}_{\text{CF}-12}, \ldots, \mathbf{\Sigma}_{\text{CF}-nm} \right) = \otimes_{t=1}^{m} \left( \mu_t \left( \sum_{i=1}^{n} \tilde{n}_i \left( \mathbf{m}_{R_{it}} \right) \right) \right) \]

where \( \mu_t = \frac{1}{\sum_{t=1}^{m} \mu_t \left( 1+\eta_t \right) \tilde{n}_t} \), and \( \eta_t = \frac{\tilde{n}_t(1+\eta_t)}{\sum_{t=1}^{m} \tilde{n}_t(1+\eta_t)} \).

\[ \tilde{\eta}_t = \sum_{k=1}^{n} \sup_{i \neq k} \left( \frac{1}{2} \mathbf{m}_{R_{it}} \cdot \mathbf{m}_{R_{kt}} \right), \]

and \( \tilde{\eta}_t \) expressed the weight vector with \( \sum_{t=1}^{m} \tilde{\eta}_t = 1 \) and \( \sum_{i=1}^{n} \tilde{n}_i = 1 \) with \( o(i) \geq o(i-1) \) and \( io(t) \geq io(t-1) \).

**Theorem 4** Considering the information in Eq. (32), we get

\[ \text{CPFSOWPA} \left( \mathbf{\Sigma}_{\text{CF}-11}, \mathbf{\Sigma}_{\text{CF}-12}, \ldots, \mathbf{\Sigma}_{\text{CF}-nm} \right) \]

\[
= \left( \begin{array}{c}
\frac{1}{\pi} \left( \sum_{t=1}^{m} \mu_t \left( \sum_{i=1}^{n} \tilde{n}_i \left( \mathbf{m}_{R_{it}} \right) \right) \right) e^{-\frac{1}{2} \left( \sum_{t=1}^{m} \mu_t \left( \sum_{i=1}^{n} \tilde{n}_i \left( \mathbf{m}_{R_{it}} \right) \right)^2 \right)}
\end{array} \right) \]

Proven.\] 

**Definition 13** When \( \mathbf{\Sigma}_{\text{CF}-it} \in \mathbb{R} \), the CPFSPG operator is simplified by:

\[ \text{CPFSPG} : \mathbb{R}^n \rightarrow \mathbb{R} \]

by

\[ \text{CPFSPG} \left( \mathbf{\Sigma}_{\text{CF}-11}, \mathbf{\Sigma}_{\text{CF}-12}, \ldots, \mathbf{\Sigma}_{\text{CF}-nm} \right) = \otimes_{t=1}^{m} \left( \mu_t \left( \sum_{i=1}^{n} \tilde{n}_i \left( \mathbf{m}_{R_{it}} \right) \right) \right) \]

where \( \mu_t = \frac{1}{\sum_{t=1}^{m} \mu_t \left( 1+\eta_t \right) \tilde{n}_t} \), and \( \eta_t = \frac{\tilde{n}_t(1+\eta_t)}{\sum_{t=1}^{m} \tilde{n}_t(1+\eta_t)} \).

\[ \tilde{\eta}_t = \sum_{k=1}^{n} \sup_{i \neq k} \left( \frac{1}{2} \mathbf{m}_{R_{it}} \cdot \mathbf{m}_{R_{kt}} \right), \]

and \( \tilde{\eta}_t \) expressed the weight vector with \( \sum_{t=1}^{m} \tilde{\eta}_t = 1 \) and \( \sum_{i=1}^{n} \tilde{n}_i = 1 \) with \( o(i) \geq o(i-1) \) and \( io(t) \geq io(t-1) \).

**Theorem 5** Considering the information in Eq. (34), we determine
\[
\text{CPFSPG} \left( \left\{ \mathcal{CIF}^{-11}, \mathcal{CIF}^{-12}, \ldots, \mathcal{CIF}^{nm} \right\} \right) = \left( \begin{array}{c}
g^{-1} \left( \sum_{t=1}^{m} \mathcal{M}_t \sum_{j=1}^{n} \mathcal{N}_j \left( \mathcal{M}_{R_t} \right) \right) \right) i2\pi \left( g^{-1} \left( \sum_{t=1}^{m} \mathcal{M}_t \sum_{i=1}^{n} \mathcal{N}_i \left( \mathcal{M}_{R_t} \right) \right) \right), \\
\mathcal{J}^{-1} \left( \sum_{t=1}^{m} \mathcal{M}_t \sum_{j=1}^{n} \mathcal{N}_j \left( \mathcal{A}_{R_t} \right) \right) i2\pi \left( \mathcal{J}^{-1} \left( \sum_{t=1}^{m} \mathcal{M}_t \sum_{i=1}^{n} \mathcal{N}_i \left( \mathcal{A}_{R_t} \right) \right) \right), \\
\mathcal{J}^{-1} \left( \sum_{t=1}^{m} \mathcal{M}_t \sum_{j=1}^{n} \mathcal{F}_{R_t} \left( \mathcal{M}_{R_t} \right) \right) i2\pi \left( \mathcal{J}^{-1} \left( \sum_{t=1}^{m} \mathcal{M}_t \sum_{i=1}^{n} \mathcal{F}_{R_t} \left( \mathcal{M}_{R_t} \right) \right) \right) \end{array} \right) \]
\]

**Proof** Omitted.

**Property 5** When \( \mathcal{CIF}^{-11} = \mathcal{CIF}^{-12} = \mathcal{CIF} \), then

\[
\text{CPFSPG} \left( \left\{ \mathcal{CIF}^{-11}, \mathcal{CIF}^{-12}, \ldots, \mathcal{CIF}^{nm} \right\} \right) = \mathcal{CIF} \]  

(36)

**Proof** Omitted.

**Property 6** When \( \mathcal{CIF}^{-11} \) and \( \mathcal{CIF}^{-12} \) be any two CIFSNs, then

\[
\text{CPFSPG} \left( \left\{ \mathcal{CIF}^{-11} \otimes \mathcal{CIF}^{-12}, \mathcal{CIF}^{-12} \otimes \mathcal{CIF}^{-nm} \otimes \mathcal{CIF} \right\} \right) = \text{CPFSPG} \left( \left\{ \mathcal{CIF}^{-11}, \mathcal{CIF}^{-12}, \ldots, \mathcal{CIF}^{nm} \right\} \otimes \mathcal{CIF} \right) \]

(37)

**Proof** Omitted.

**Property 7** Prove that

\[
\text{CPFSPG} \left( \left\{ \mathcal{CIF}^{-11}, \mathcal{CIF}^{-12}, \mathcal{CIF}^{-nm} \right\} \right) \right) = \text{CPFSPG} \left( \left\{ \mathcal{CIF}^{-11}, \mathcal{CIF}^{-12}, \ldots, \mathcal{CIF}^{nm} \right\} \right) \]

(38)

**Proof** Omitted.

**Property 8** When \( \mathcal{CIF}^{-11} \) is

\[
\left( \mathcal{M}_{R_{k-t}}^{-1}, \mathcal{M}_{R_{k-t}}^{-1}, \mathcal{M}_{R_{k-t}}^{-1}, \mathcal{M}_{R_{k-t}}^{-1} \right) \]
then

\[
\text{CPFSPG} \left( \left\{ \mathcal{CIF}^{-11}, \mathcal{CIF}^{-12}, \ldots, \mathcal{CIF}^{nm} \right\} \right) \right) = \text{CPFSPG} \left( \left\{ \mathcal{CIF}^{-11}, \mathcal{CIF}^{-12}, \ldots, \mathcal{CIF}^{nm} \right\} \right) \]

(39)

**Proof** Omitted.

Important cases of the invented work using the information in Eq. (35) are described here.

Assume \( \hat{g}(\phi) = -\log(\phi) \) in Eq. (35), then
CPFSPG\left(\Sigma_{CIF-11}, \Sigma_{CIF-12}, \ldots, \Sigma_{CIF-nm}\right) =

\left\{\begin{array}{c}
\prod_{t=1}^{m} \left( \prod_{i=1}^{n} (\mathcal{R}_{R_{st}})_{i} \right) e^{2\pi i} \\
\prod_{t=1}^{m} \left( \prod_{i=1}^{n} (1 - A_{R_{st}})_{i} \right) e^{2\pi i} \\
\prod_{t=1}^{m} \left( \prod_{i=1}^{n} (1 - \mathcal{R}_{R_{st}})_{i} \right) e^{2\pi i}
\end{array}\right\}

(40)

Stated the CPFS Archimedean weighted geometric (CPFSAWG) operator.

Assume \(\overline{\mathcal{g}}(\phi) = \log\left(\frac{2\pi e}{\phi}\right)\), \(\phi \neq 0\) in Eq. (35), then

\[
\text{CPFSPG}\left(\Sigma_{CIF-11}, \Sigma_{CIF-12}, \ldots, \Sigma_{CIF-nm}\right) =
\left\{\begin{array}{c}
2 \prod_{t=1}^{m} \left( \prod_{i=1}^{n} (\mathcal{R}_{R_{st}})_{i} \right) e^{2\pi i} \\
\prod_{t=1}^{m} \left( \prod_{i=1}^{n} (1 - \mathcal{R}_{R_{st}})_{i} \right) e^{2\pi i} \\
\prod_{t=1}^{m} \left( \prod_{i=1}^{n} (1 - A_{R_{st}})_{i} \right) e^{2\pi i}
\end{array}\right\}
\]

(41)

Stated the CPFS Einstein weighted geometric (CPFSEWG) operator.
Assume \( g(\varphi) = \log \left( \sigma + \frac{1-\sigma}{\varphi} \right) \), \( \sigma \in (0, \infty), \varphi \neq 0 \) in Eq. \( (35) \), then

\[
\text{CPFSPG}(\Sigma_{CIF-11}, \Sigma_{CIF-12}, \ldots, \Sigma_{CIF-mm}) =
\]

\[
\left( \begin{array}{c}
\tilde{\sigma} \prod_{n=1}^{m} \left( \prod_{i=1}^{n} (\mathcal{M}_{R_{n}} \psi_{i})^{\tilde{M}}_{\tau} \right) \\
\prod_{n=1}^{m} \left( \prod_{i=1}^{n} (1 + \left( \tilde{\sigma} - 1 \right) (1 - \mathcal{M}_{R_{n}} \psi_{i})^{\tilde{M}}_{\tau} \right) + \left( \tilde{\sigma} - 1 \right) \prod_{n=1}^{m} \left( \prod_{i=1}^{n} (\mathcal{M}_{R_{n}} \psi_{i})^{\tilde{M}}_{\tau} \right)
\end{array} \right)
\]
\[
\text{CPFSWPG}\left(\Sigma_{\text{CIF}}^{-1}, \Sigma_{\text{CIF}}^{-12}, \ldots, \Sigma_{\text{CIF}}^{-nm}\right) = \otimes_{\ell=1}^{m} \left( \otimes_{i=1}^{n} \left( \mathcal{M}^{-\ell}(\bar{N}_{i}) \right) \right)
\]

where

\[
\bar{M}_{\ell} = \frac{\hat{\mu}_{\ell} \left( 1 + \bar{T}_{\ell} \right)}{\sum_{\ell=1}^{m} \hat{\mu}_{\ell} \left( 1 + \bar{T}_{\ell} \right)}, \quad \bar{N}_{i} = \frac{\bar{\eta}_{i} \left( 1 + \bar{N}_{i} \right)}{\sum_{i=1}^{n} \bar{\eta}_{i} \left( 1 + \bar{N}_{i} \right)} ,
\]

and

\[
\bar{T}_{\ell} = \sum_{k=1}^{m} \left[ \sum_{i=1}^{n} \bar{N}_{i} \right] \left( \mathcal{S}_{\text{PI}}^{-\ell} \mathcal{S}_{\text{PI}}^{-i} \right), \quad \text{and} \quad \sum_{k=1}^{n} \text{Sup}_{i} \left( \mathcal{S}_{\text{PI}}^{-\ell} \mathcal{S}_{\text{PI}}^{-kt} \right),
\]

stated the support for \( \bar{S}_{\text{PI}}^{-\ell} \) and \( \bar{S}_{\text{PI}}^{-kt} \), where \( \hat{\mu}_{\ell} \) and \( \bar{\eta}_{i} \), expressing the weight vector with \( \sum_{\ell=1}^{m} \hat{\mu}_{\ell} = 1 \) and \( \sum_{i=1}^{n} \bar{\eta}_{i} = 1 \).

**Theorem 6** Considering the information in Eq. (33), we get

\[
\text{CPFSWPG}\left(\Sigma_{\text{CIF}}^{-1}, \Sigma_{\text{CIF}}^{-12}, \ldots, \Sigma_{\text{CIF}}^{-nm}\right) =
\]

\[
\left( \mathcal{G}^{-1} \left( \sum_{\ell=1}^{m} \bar{M}_{\ell} \left( \sum_{i=1}^{n} \bar{N}_{i} \right) \mathcal{M}_{\ell} \left( \mathcal{G}_{\mathcal{R}_{\ell}} \right) \right) \right) \left( \mathcal{G}^{-1} \left( \sum_{\ell=1}^{m} \bar{M}_{\ell} \left( \sum_{i=1}^{n} \bar{N}_{i} \right) \mathcal{M}_{\ell} \left( \mathcal{G}_{\mathcal{R}_{\ell}} \right) \right) \right)
\]

**Proof** Omitted.

**Definition 14** When \( \Sigma_{\text{CIF}}^{-\ell} \in \mathcal{S} \), the CPFSOWPG operator is simplified by:

\[
\text{CPFSOWPG} : \mathcal{S}^{n} \rightarrow \mathcal{S}
\]

by

\[
\text{CPFSOWPG}\left(\Sigma_{\text{CIF}}^{-1}, \Sigma_{\text{CIF}}^{-12}, \ldots, \Sigma_{\text{CIF}}^{-nm}\right) = \otimes_{\ell=1}^{m} \left( \otimes_{i=1}^{n} \left( \mathcal{M}^{-\ell}(\bar{N}_{i}) \right) \right)
\]

where

\[
\bar{M}_{\ell} = \frac{\hat{\mu}_{\ell} \left( 1 + \bar{T}_{\ell} \right)}{\sum_{\ell=1}^{m} \hat{\mu}_{\ell} \left( 1 + \bar{T}_{\ell} \right)}, \quad \bar{N}_{i} = \frac{\bar{\eta}_{i} \left( 1 + \bar{N}_{i} \right)}{\sum_{i=1}^{n} \bar{\eta}_{i} \left( 1 + \bar{N}_{i} \right)} ,
\]

and

\[
\bar{T}_{\ell} = \sum_{k=1}^{m} \left[ \sum_{i=1}^{n} \bar{N}_{i} \right] \left( \mathcal{S}_{\text{PI}}^{-\ell} \mathcal{S}_{\text{PI}}^{-i} \right), \quad \text{and} \quad \sum_{k=1}^{n} \text{Sup}_{i} \left( \mathcal{S}_{\text{PI}}^{-\ell} \mathcal{S}_{\text{PI}}^{-kt} \right),
\]

simplified the support for \( \bar{S}_{\text{PI}}^{-\ell} \) and \( \bar{S}_{\text{PI}}^{-kt} \), where \( \hat{\mu}_{\ell} \) and \( \bar{\eta}_{i} \), expressed the weight vector with \( \sum_{\ell=1}^{m} \hat{\mu}_{\ell} = 1 \) and \( \sum_{i=1}^{n} \bar{\eta}_{i} = 1 \) with \( o(\ell) \geq o(i-1) + 1 \) and \( i \geq o(\ell - 1) \).

**Theorem 7** Considering the information in Eq. (35), we get
\[ \text{CPFSOWPG} \left( \sum_{CIF=1}^{m}, \sum_{CIF=2}^{m}, \ldots, \sum_{CIF=nm}^{m} \right) = \]

\[
\begin{align*}
&= \left( g^{-1} \left( \sum_{t=1}^{m} M_{t} \left( \sum_{i=1}^{n} N_{i} \int_{e}^{g} (M_{R_{\alpha_{ij}(t)})} \right) \right) \right) e^{2\pi i \left( z \sum_{t=1}^{m} M_{t} \left( \sum_{i=1}^{n} N_{i} \int_{e}^{g} (M_{R_{\alpha_{ij}(t)})} \right) \right)} \\
&= \left( f^{-1} \left( \sum_{t=1}^{m} M_{t} \left( \sum_{i=1}^{n} N_{i} \int_{e}^{f} (A_{R_{\alpha_{ij}(t)})} \right) \right) \right) e^{2\pi i \left( z \sum_{t=1}^{m} M_{t} \left( \sum_{i=1}^{n} N_{i} \int_{e}^{f} (A_{R_{\alpha_{ij}(t)})} \right) \right)} \\
&= \left( f^{-1} \left( \sum_{t=1}^{m} M_{t} \left( \sum_{i=1}^{n} N_{i} \int_{e}^{f} (M_{R_{\alpha_{ij}(t)})} \right) \right) \right) e^{2\pi i \left( z \sum_{t=1}^{m} M_{t} \left( \sum_{i=1}^{n} N_{i} \int_{e}^{f} (M_{R_{\alpha_{ij}(t)})} \right) \right)}
\end{align*}
\]

*(46)*

**Proof** Omitted.

**Application ("MADM Processes using Proposed Operators")**

In the consideration of diagnosed operators using CPFS information, we illustrated a MADM tool to find the best option from the family of decisions.

**Strategic decision-making processes.** For managing awkward and problematic information that occurred in genuine life dilemmas, the MADM tool plays an important role in the circumstance of FS theory. Here, we have discussed a procedure for resolving the above issues, we choose \( m \) alternatives and \( n \) attributes whose representations are described: \( \Xi_{AT} = \{ \Xi_{AT-1}, \Xi_{AT-2}, \ldots, \Xi_{AT-m} \} \) and \( \Xi_{AL} = \{ \Xi_{AL-1}, \Xi_{AL-2}, \ldots, \Xi_{AL-n} \} \), with weight vector \( \sum_{t=1}^{m} \mu_{t} = 1 \) and \( \sum_{i=1}^{n} \eta_{i} = 1 \), provided for attributes and parameters. Then, we give a CPFSNs

\[ \Xi_{CIF-it} = \left( \prod_{i=1}^{m} e^{2\pi i (M_{R_{\alpha_{ij}(t)})}, A_{R_{\alpha_{ij}(t)}}, R_{\alpha_{ij}(t)} e^{2\pi i (M_{R_{\alpha_{ij}(t)})} \right), i = 1, 2, \ldots, n, \] with \( 0 < M_{R_{\alpha_{ij}(t)}}, A_{R_{\alpha_{ij}(t)}} + R_{\alpha_{ij}(t)} \leq 1 \) for each alternative in the shape of a matrix. To handle the above scenario, we described various stages for resolving the above issues.

**Stage 1:** Deliberated information in the form of a closed matrix that contains CPFSNs for each alternative and their attributes, the mathematical term \( \Xi_{CIF-it}^{(b)} \), \( b = 1, 2, \ldots, z \), stated the matrix.

**Stage 2:** Summarized the \( \Xi_{it} \), stated the support for the intellectuals, we have

\[ \Xi_{it}^{(b)} = \sum_{k=1}^{n} \text{Sup} \left( \Xi_{pf-it}^{(b)} \Xi_{pf-kt}^{(b)} \right) i \neq k \]

where

\[ \text{Sup} \left( \Xi_{pf-it}^{(b)} \Xi_{pf-kt}^{(b)} \right) = 1 - d \left( \Xi_{pf-it}^{(b)} \Xi_{pf-kt}^{(b)} \right), d \left( \Xi_{pf-it}^{(b)} \Xi_{pf-kt}^{(b)} \right) = \text{score of current entry}. \]

**Stage 3:** Considering the weighted averaging/geometric operators, we initiate \( \Xi_{t}^{(b)} \), stated the support for the intellectuals, we have

\[ \Xi_{t}^{(b)} = \sum_{k=1}^{m} \text{Sup} \left( \Xi_{pf-kt}^{(b)} \Xi_{pf-it}^{(b)} \right) i \neq k \]

where \( \text{Sup} \left( \Xi_{pf-kt}^{(b)} \Xi_{pf-it}^{(b)} \right) = 1 - d \left( \Xi_{pf-kt}^{(b)} \Xi_{pf-it}^{(b)} \right), d \left( \Xi_{pf-kt}^{(b)} \Xi_{pf-it}^{(b)} \right) = \text{sum of all elements in order of number in row}. \]

**Stage 4:** Considering the CPFSPA and CPFSOG operators, we deliberated the values from the matrices into exact values in the availability of TN and TCN \( g \left( \Xi_{it} \right) = -\log (\Xi_{it}). \)

**Stage 5:** Averaged the SV of the deliberated values.

**Stage 6:** Ranking all results and diagnosed the best optimal.

**Illustrated example.** The existing shortage of energy in various constructing countries over a decade is causing the economic growth of the country. Between the distinct energy supplies, electrical energy is the main energy supplier which is valuable in the market by the distinct area of the economy. For various years, Pakistan is suffering from various electricity issues. The intensification of the electrical shortage has become a main political dilemma in Pakistan which is causing the strong construction of the economy. The problem of electricity in Pakistan does not only affect the people and the economy but also affects many other dilemmas, because of some brain-dead employees and unsuccessful planning and policy. Therefore, the management of the government of Pakistan must give special attention to these issues and try to increase the quality of the electricity demand. For
### Table 1. Information is given for Scholars 1.

| e₁ | e₂ | e₃ | e₄ | e₅ |
|----|----|----|----|----|
| 0.41<sub>2</sub>(0.31, 0.11) | 0.31<sub>2</sub>(0.11, 0.11) | 0.32<sub>2</sub>(0.21, 0.12) | 0.33<sub>2</sub>(0.21, 0.12) | 0.43<sub>2</sub>(0.33, 0.13) |
| 0.32<sub>2</sub>(0.21, 0.12) | 0.33<sub>2</sub>(0.21, 0.12) | 0.32<sub>2</sub>(0.21, 0.12) | 0.33<sub>2</sub>(0.21, 0.12) | 0.43<sub>2</sub>(0.33, 0.13) |
| 0.32<sub>2</sub>(0.21, 0.12) | 0.33<sub>2</sub>(0.21, 0.12) | 0.33<sub>2</sub>(0.21, 0.12) | 0.33<sub>2</sub>(0.21, 0.12) | 0.43<sub>2</sub>(0.33, 0.13) |
| 0.32<sub>2</sub>(0.21, 0.12) | 0.33<sub>2</sub>(0.21, 0.12) | 0.33<sub>2</sub>(0.21, 0.12) | 0.33<sub>2</sub>(0.21, 0.12) | 0.43<sub>2</sub>(0.33, 0.13) |
| 0.32<sub>2</sub>(0.21, 0.12) | 0.33<sub>2</sub>(0.21, 0.12) | 0.33<sub>2</sub>(0.21, 0.12) | 0.33<sub>2</sub>(0.21, 0.12) | 0.43<sub>2</sub>(0.33, 0.13) |

### Table 2. Information is given for Scholars 2.

| e₁ | e₂ | e₃ | e₄ | e₅ |
|----|----|----|----|----|
| 0.51<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) |
| 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) |
| 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) |
| 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) |
| 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) | 0.61<sub>2</sub>(0.21, 0.11) |

### Table 3. Information is given for Scholars 3.

| e₁ | e₂ | e₃ | e₄ | e₅ |
|----|----|----|----|----|
| 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) |
| 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) |
| 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) |
| 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) |
| 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) | 0.31<sub>2</sub>(0.21, 0.11) |
Table 4. Information is given for Scholars 4.

| ei | e1 | e2 | e3 | e4 | e5 |
|----|----|----|----|----|----|
| 0.61e^{2r} (0.21) | 0.11e^{2r} (0.11) | 0.11e^{2r} (0.11) | 0.62e^{2r} (0.21) | 0.12e^{2r} (0.12) | 0.63e^{2r} (0.23) |
| 0.12e^{2r} (0.12) | 0.12e^{2r} (0.12) | 0.13e^{2r} (0.13) | 0.13e^{2r} (0.13) | 0.14e^{2r} (0.14) |
| 0.64e^{2r} (0.24) | 0.14e^{2r} (0.14) | 0.14e^{2r} (0.14) | 0.14e^{2r} (0.14) | 0.14e^{2r} (0.14) |
| 0.41e^{2r} (0.31) | 0.31e^{2r} (0.21) | 0.11e^{2r} (0.11) | 0.42e^{2r} (0.21) | 0.32e^{2r} (0.12) | 0.43e^{2r} (0.33) |
| 0.32e^{2r} (0.22) | 0.12e^{2r} (0.12) | 0.12e^{2r} (0.12) | 0.33e^{2r} (0.22) | 0.33e^{2r} (0.22) | 0.34e^{2r} (0.24) |
| 0.34e^{2r} (0.24) | 0.14e^{2r} (0.14) | 0.14e^{2r} (0.14) | 0.14e^{2r} (0.14) | 0.14e^{2r} (0.14) |
| 0.51e^{2r} (0.21) | 0.11e^{2r} (0.11) | 0.21e^{2r} (0.11) | 0.52e^{2r} (0.21) | 0.22e^{2r} (0.12) | 0.53e^{2r} (0.33) |
| 0.22e^{2r} (0.12) | 0.12e^{2r} (0.12) | 0.12e^{2r} (0.12) | 0.33e^{2r} (0.22) | 0.33e^{2r} (0.22) | 0.34e^{2r} (0.24) |
| 0.34e^{2r} (0.24) | 0.14e^{2r} (0.14) | 0.14e^{2r} (0.14) | 0.14e^{2r} (0.14) | 0.14e^{2r} (0.14) |

Stage 1: Deliberated information in the form of a closed matrix that contains CPFSNs for each alternative and their attributes in the form of Tables 1, 2, 3, and 4, the mathematical term: Social-Political with parameters, described in the form: e1: Client Facilities, e2: Bandwidth, e3: Package, e4: Total Cost, and e5 Internet Speed. Here, we choose the values (0.3, 0.2, 0.3, 0.2) and (0.3, 0.2, 0.25, 0.15, 0.1), stated the weighty vector for Σi. To handle the above scenario, we described various stages for resolving the above issues.

Stage 2: Summarized the Σi, stated the support for the intellectuals, we have

\[ \overline{\Sigma}_i^{(1)} = \begin{bmatrix} 5.55E -17 & 0.02 & 0.04 & 0.06 & 0.08 \\ 0.3 & 0.32 & 0.34 & 0.36 & 0.38 \\ 0.1 & 0.12 & 0.14 & 0.16 & 0.18 \\ 0.1 & 0.12 & 0.14 & 0.16 & 0.18 \\ 0.4 & 0.38 & 0.36 & 0.34 & 0.32 \end{bmatrix} , \]

\[ \overline{\Sigma}_i^{(2)} = \begin{bmatrix} 0.1 & 0.12 & 0.14 & 0.16 & 0.18 \\ 0.4 & 0.38 & 0.36 & 0.34 & 0.32 \\ 0.1 & 0.12 & 0.14 & 0.16 & 0.18 \\ 0.3 & 0.32 & 0.34 & 0.36 & 0.38 \end{bmatrix} , \]

\[ \overline{\Sigma}_i^{(3)} = \begin{bmatrix} 0.3 & 0.32 & 0.34 & 0.36 & 0.38 \\ 0.1 & 0.12 & 0.14 & 0.16 & 0.18 \\ 0.04 & 0.38 & 0.36 & 0.34 & 0.32 \\ 0.04 & 0.38 & 0.36 & 0.34 & 0.32 \\ 5.55E -17 & 0.02 & 0.04 & 0.06 & 0.08 \\ 0.3 & 0.32 & 0.34 & 0.36 & 0.38 \end{bmatrix} , \]

\[ \overline{\Sigma}_i^{(4)} = \begin{bmatrix} 0.04 & 0.38 & 0.36 & 0.34 & 0.32 \\ 0.1 & 0.12 & 0.14 & 0.16 & 0.18 \\ 0.1 & 0.12 & 0.14 & 0.16 & 0.18 \\ 0.4 & 0.38 & 0.36 & 0.34 & 0.32 \end{bmatrix} , \]

Stage 3: Considering the weighted averaging/geometric operators, we initiate \( \overline{T}_f \), stated the support for the intellectuals, we have

\[ \overline{T}_f^{(1)} = \begin{bmatrix} 0.04 \\ 0.34 \\ 0.14 \\ 0.14 \\ 0.36 \end{bmatrix} , \]

\[ \overline{T}_f^{(2)} = \begin{bmatrix} 0.14 \\ 0.36 \\ 0.04 \\ 0.34 \\ 0.14 \end{bmatrix} , \]

\[ \overline{T}_f^{(3)} = \begin{bmatrix} 0.14 \\ 0.36 \\ 0.04 \\ 0.34 \\ 0.14 \end{bmatrix} , \]

\[ \overline{T}_f^{(4)} = \begin{bmatrix} 0.14 \\ 0.36 \\ 0.04 \\ 0.34 \\ 0.14 \end{bmatrix} , \]

Stage 4: Considering the CPFSPA and CPFSFG operators, we deliberated the values from the matrices into exact values in the availability of TN and TCN

\[ \overline{\Sigma}_i^{(1)} = \begin{bmatrix} 0.4384e^{2r} (0.2268) & 0.172e^{2r} (0.1571) & 0.128e^{2r} (0.1515) \\ 0.4485e^{2r} (0.2267) & 0.1836e^{2r} (0.1869) & 0.1394e^{2r} (0.1666) \end{bmatrix} , \]

\[ \overline{\Sigma}_i^{(2)} = \begin{bmatrix} 0.4587e^{2r} (0.2466) & 0.195e^{2r} (0.18) & 0.15e^{2r} (0.1775) \\ 0.4689e^{2r} (0.2564) & 0.2063e^{2r} (0.1912) & 0.1606e^{2r} (0.1893) \end{bmatrix} , \]

\[ \overline{\Sigma}_i^{(3)} = \begin{bmatrix} 0.4791e^{2r} (0.2663) & 0.2174e^{2r} (0.2035) & 0.1711e^{2r} (0.2001) \end{bmatrix} , \]

\[ \overline{\Sigma}_i^{(4)} = \begin{bmatrix} 0.4238e^{2r} (0.2332) & 0.194e^{2r} (0.1846) & 0.128e^{2r} (0.1747) \\ 0.4342e^{2r} (0.2434) & 0.2039e^{2r} (0.1946) & 0.1383e^{2r} (0.1846) \end{bmatrix} , \]

\[ \overline{\Sigma}_i^{(5)} = \begin{bmatrix} 0.4455e^{2r} (0.2537) & 0.2138e^{2r} (0.2046) & 0.148e^{2r} (0.1945) \\ 0.4548e^{2r} (0.2639) & 0.2237e^{2r} (0.2143) & 0.1579e^{2r} (0.2044) \end{bmatrix} , \]

\[ \overline{\Sigma}_i^{(6)} = \begin{bmatrix} 0.4651e^{2r} (0.2741) & 0.2336e^{2r} (0.2245) & 0.1678e^{2r} (0.2143) \end{bmatrix} , \]

Stage 5: Averaged the SV of the deliberated values, such that
SPG operators. To further enhance the standard of the invented approaches, we discuss the sensitive analysis of
existing theories is described in Refs.32,40–44.

Stage 6: Ranking all results and diagnosed the best optimal, such that

\[ L_1 \geq L_3 \geq L_2 \geq L_4 \geq L_5, \]

Hence, we obtained two different sorts of ranking results in the shape of \( L_1 \) and \( L_5 \), using CPFSPA and CPFSPG operators. To further enhance the standard of the invented approaches, we discuss the sensitive analysis of diagnosed work with various suggested approaches.

**Sensitive analysis.** The main idea of this analysis is to prove the invented work is more beneficial and realistic than the existing operators with the help of some comparison. Many scholars have diagnosed this procedure as demonstrating the proposed approaches are more utilized than the existing operators. For this, we consider various prevailing theories are tried to compare them with our diagnosed operators. The information related to existing theories is described in Refs.32,40–44.

1. Information given in Ref.40 contained the power aggregation operators for CIFSs, the invested power aggregation operators based on CPFS information are more able to resolve intuitionistic, picture, complex intuitionistic, and complex picture fuzzy information. But one deficiency that occurred in the existing operator is that it contained two grades in the shape of truth and falsity in the form of polar coordinates, which is not suitable because the proposed work contained information in the shape of truth, abstinence, falsity grades with parameters. Therefore, the prevailing operators in Ref.40 failed.
2. Information given in Ref.32 contained the power aggregation operators for CPFSs, the invested power aggregation operators based on CPFS information are more able to resolve intuitionistic, picture, complex intuitionistic, and complex picture fuzzy information. But one deficiency that occurred in the existing operator is that it contained three grades in the shape of truth, abstinence, and falsity in the form of polar coordinates, which is not suitable because the proposed work contained information in the shape of truth, abstinence, falsity grades with parameters. Therefore, the prevailing operators in Ref.32 failed.
3. Information given in Ref.41 contained the aggregation operators for PFSSs, the invested power aggregation operators based on CPFS information are more able to resolve intuitionistic, picture, complex intuitionistic, and complex picture fuzzy information. But one deficiency that occurred in the existing operator is that it contained three grades in the shape of truth, abstinence, and falsity with parameters in the form of one dimension, which is not suitable because the proposed work contained information in the shape of truth, abstinence, falsity grades with parameters in the shape of polar coordinates. Therefore, the prevailing operators in Ref.41 failed. The comparative analysis is described in Table 5.
4. The theory diagnosed by Garg and Arora42, is called generalized Maclaurin symmetric information based on IFSSs. Noticed that the theory presented by Garg and Arora42 based on IFSSs is the special case of the diagnosed CPFS information and because of this reason, the theory of Garg and Arora42 is not able to use for evaluating information discussed in section “Illustrated example”. Therefore, the theory of diagnosed information is massively powerful, and dominant compared to the prevailing theory discussed by Garg and Arora42.
5. Robust aggregation operators diagnosed in Ref.43 contained the robust aggregation operators for intuitionistic hypersoft sets (IHSS), the invested robust aggregation operators based on IHSS information are more able to resolve intuitionistic information. But one deficiency that occurred in the existing operator is that it contained three grades in the shape of truth and falsity in the form of one-dimension information, which is not suitable because the proposed work contained information in the shape of truth, abstinence, falsity grades with parameters. Therefore, the prevailing operators in Ref.43 failed.

| Methods | Score values | Ranking values |
|---------|--------------|----------------|
| Rani and Garg43 | Not justified | Not justified |
| Liu et al.32 | Not justified | Not justified |
| Khan et al.14 | Not justified | Not justified |
| Garg and Arora42 | Not justified | Not justified |
| Zulqarnain et al.43 | Not justified | Not justified |
| Wang et al.44 | Not justified | Not justified |

Table 5. Represented the sensitivity analysis.

\[ L_1 = 0.0522, L_2 = 0.0272, L_3 = 0.0024, L_4 = 0.0217, L_5 = 0.0454. \]

\[ L_1 = 0.0247, L_2 = 0.0437, L_3 = 0.0628, L_4 = 0.0819, L_5 = 0.1011. \]
6. The theory diagnosed by Wang et al.⁴⁴, called Hamy means information based on Pythagorean uncertain linguistic information. Noticed that the theory presented by Wang et al.⁴⁴ based on Pythagorean uncertain linguistic information is the special case of the diagnosed CPFSS information and because of this reason, the theory of Wang et al.⁴⁴ is not able to use for evaluating information discussed in section "Illustrated example". Therefore, the theory of diagnosed information is massively powerful, and dominant compared to the prevailing theory discussed by Wang et al.⁴².

From the above-cited information, we obtained the final result in the shape of $\mathcal{L}_1$ and $\mathcal{L}_5$. Therefore, the invented operators based on CPFSS are very effective and dominant as compared to prevailing work⁴⁵,⁴⁰,⁴¹.

Conclusion
The major theme of this analysis is diagnosed below:

1. We invented the new theory in the form of CPFS information and invented their major algebraic laws, score value, and accuracy values. The mathematical form of the CPFS set includes three main functions, called supporting, abstinence, and supporting against terms with a prominent characteristic that is the sum of the triplet will lie in the unit interval.

2. In the consideration of the power aggregation operator using generalized t-norm and t-conorm and CPFS information, we diagnosed the mathematical concept of CPFSWA, CPFSWPA, CPFSOWPA, CPFSWG, CPFSOWPG.

3. The major results and their particular investigation of the invented approaches are also deliberated with the help of t-norm and t-conorm $\mu((\bar{\psi}) \neg \log (\bar{\psi})) = -\log (\bar{\psi})$, $\mu((\bar{\psi}) \log (\frac{2-\bar{\psi}}{\bar{\psi}}), \bar{\psi} \neq 0$, and $\mu((\bar{\psi}) = \log \left(\frac{1}{\sigma} + \frac{1}{\bar{\psi}}\right)$, $\sigma \in (0, \infty), \bar{\psi} \neq 0$.

4. In the consideration of diagnosed operators using CPFS information, we illustrated a MADM tool to find the best option from the family of decisions.

5. Finally, we showed the supremacy and feasibility of the diagnosed operators with the help of sensitive analysis and geometrical representations.

Limitations of the proposed approaches. No doubt, the theory of CPFS information has a lot of benefits, but in some cases, the theory of CPFS information, if someone provided such type of information whose sum is exceeded from the unit interval, then for managing with such sort of situation, we need to propose the theory of complex spherical fuzzy soft sets and complexT-spherical fuzzy soft sets.

Future work. In the upcoming times, we will try to modify the principle of complex q-rung orthopair fuzzy sets⁵¹, complex spherical fuzzy sets⁵⁶–⁵⁷, T-spherical fuzzy sets⁵⁸, and decision-making⁵⁹–⁶⁷ to enhance the excellence and capacity of the exploration performs.

Ethics declaration statement. The authors state that this is their original work, and it is neither submitted nor under consideration in any other journal simultaneously.

Data availability
The data utilized in this manuscript are hypothetical and artificial, and one can use these data before prior permission by just citing this manuscript.

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