Research Article

Decision-Making Framework for an Effective Sanitizer to Reduce COVID-19 under Fermatean Fuzzy Environment

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1. Introduction

In decision sciences, it is an important aspect to find the ranking order of the alternatives corresponding to different attributes according to the preferences of the decision-making experts. Therefore, selection of various attributes of the alternatives is a very complex task. These decisions cannot be interpreted by the exact data so the need of a powerful model was raised to handle the ambiguous data. For that issue, Zadeh [1] initiated the innovative idea of fuzzy set (FS) which served as the backbone of the FS theory. FS permits the experts to describe their satisfaction level (membership degree) regarding performance of a member within the unit interval. Although, the FSs provide the grounds to the uncertain assessments but they were not adequate enough to describe the NMD. To overcome the limitations of FS, Atanassov [2] introduced a more dominant model, namely, IFS which has both MD $\mu$ and NMD $\nu$ with condition $\mu + \nu \leq 1$. The theory of IFS was felt to be inept and insufficient to represent the inexact data as there are a lot of problems where the sum of MD and NMD is exceeded by 1. To reduce such type of complications, Yager [3] delivered the idea of PFS with condition $\mu^2 + \nu^2 \leq 1$. However, PFS has also some limitations if MD of an element is 0.8 and NMD is 0.7, then sum of square of these values is greater than 1. Then, Yager [4] developed the theory of $q$-rung orthopair fuzzy set ($q$-ROFS) with condition $\mu^q + \nu^q \leq 1$. Recently, Senapati and Yager [5] gave the concept of FFS as a generalization of IFS and PFS. The worthwhile theory of AOs is widely applied to decision-making scenarios for the sake of data aggregation and to identify the best alternative from the possible choices. Xu [6] gave the idea of intuitionistic fuzzy (IF) AOs. The concept of generalized AOs for IFS was developed by Zhao et al. [7]. Rahman et al. [8]
introduced Pythagorean fuzzy (PF) AOs. Zhao and Wei [9] studied the Einstein hybrid AOs under IF environment. The idea of IF aggregation using Einstein operations was discussed by Wang and Liu [10]. The induced interval-valued IF Einstein AOs were developed by Cai and Han [11]. Garg [12] studied the generalized PF Einstein weighted arithmetic AOs. Garg [13] also proposed the generalized PF Einstein weighted geometric AOs. The Pythagorean Dombi fuzzy AOs with applications were discussed by Akram et al. [14]. Shahzadi et al. [15] proposed the decision-making approach using PF Yager AOs. Liu and Wang [16] expressed q-rung orthopair fuzzy (q-ROF) weighted AOs. Wei et al. [17] studied weighted Heronian mean AOs for complex -ROF aggregation operators. Akram and Shahzadi [22] developed the hybrid decision-making model under q-ROF Yager AOs. Liu et al. [23] extended the concept of prioritized weighted AOs for complex q-ROFS. Senapati and Yager [24] studied subtraction, division, and Fermatean arithmetic mean operations over FFS. The idea of Fermatean fuzzy (FF) weighted averaging/geometric operators was also given by Senapati and Yager [25]. For more information and applications, the readers can refer to [26–59].

The motivations of this article are described as follows:

1. The judgement of a perfect alternative in an FF environment is a laborious MADM problem. The prevalent model, possessing the more space than the IF model and PF model, vigorously elaborates the imprecise decisions for the selection of best alternative.

2. As Einstein AOs are the simplest and quite creative approach for dealing with DM affairs, basically, this article directs Einstein AOs in FF surroundings to face complex issues.

3. The outcomes based on conclusion are quite accurate under Einstein AOs when it is put on to the reality-based MADM problems in FF data.

4. The proposed operators are keen to provide the optimal solution not only for FF environment but also to work efficiently for IF and PF environment.

The contributions of this article are described as follows:

1. The feasibility of FFNs is merged with the aggregation skills of Einstein norms to establish more powerful, multiskilled, and practical AOs which can be deployed to aggregate FF data and to get more accuracy in decision-making scenarios.

2. The dominant properties as well as the notable results of the proposed operators are highlighted.

3. An algorithm is studied to handle complex realistic problems with FF data.

4. A MADM problem for the selection of an effective sanitizer to reduce coronavirus is discussed by using proposed operators.

5. A validity test is discussed for the approval and authenticity of proposed theory.

6. At the end, the benefits and characteristics of the proposed work are discussed by comparison analysis.

The remaining paper is as follows. In Section 2, we recall the concept of FFS and related score functions. Section 3 provides Einstein operational laws for FFNs. In Sections 4 and 5, we study the FF EFEWA and FF EFEOWA operators, respectively, and related properties to them. In Sections 6 and 7, we present the idea of GFFEWA and GFFEOWA operators, respectively. In Section 8, we propose an algorithm for our new model and discuss a MADM problem for the selection of a good sanitizer to reduce the coronavirus. Section 9 provides the validity criteria to prove the consistency of the proposed work. Section 10 gives the comparison analysis of proposed theory with the FF TOPSIS method. In Section 11, we have concluded the results related to the proposed model.

The list of acronyms in research paper is given in Table 1.

2. Preliminaries

In this section, we recall some basic definitions including IFS, PFS, FFS, and score functions related to FFS.

Definition 1. (see [2]). An IFS I on nonempty set \( \mathcal{V} \) is given by

\[ I = \{ (x, \mu_I(x), \nu_I(x)) \} \]

where \( \mu_I: \mathcal{V} \rightarrow [0, 1] \) and \( \nu_I: \mathcal{V} \rightarrow [0, 1] \) specify MD and NMD of an element \( x \in \mathcal{V} \), respectively. \( \omega_I(x) = 1 - \mu_I(x) - \nu_I(x) \) is indeterminacy degree (InD) of an element \( x \in \mathcal{V} \).

Definition 2. (see [3]). A PFS P on nonempty set \( \mathcal{V} \) is given by

\[ P = \{ (x, \mu_P(x), \nu_P(x)) \} \]

where \( \mu_P: \mathcal{V} \rightarrow [0, 1] \) and \( \nu_P: \mathcal{V} \rightarrow [0, 1] \) specify MD and NMD of an element, respectively. \( \omega_P(x) = \sqrt{1 - (\mu_P(x))^2 - (\nu_P(x))^2} \) is InD.

Definition 3. (see [5]). An FFS \( \mathcal{R} \) on nonempty set \( \mathcal{V} \) is given by

\[ \mathcal{R} = \{ (x, \mu_\mathcal{R}(x), \nu_\mathcal{R}(x)) \} \]

where \( \mu_\mathcal{R}: \mathcal{V} \rightarrow [0, 1], \nu_\mathcal{R}: \mathcal{V} \rightarrow [0, 1], \) and \( \omega_\mathcal{R}(x) = \sqrt{1 - (\mu_\mathcal{R}(x))^2 - (\nu_\mathcal{R}(x))^2} \) specify MD, NMD, and InD, respectively. FFNs are components of the FFS.

Definition 4. (see [5]). The score function and accuracy function for FFN \( \mathcal{R} = (\mu_\mathcal{R}, \nu_\mathcal{R}) \) are represented by

\[ S(\mathcal{R}) = \mu_\mathcal{R}^3 - \nu_\mathcal{R}^3, \quad \text{where} \ S(\mathcal{R}) \in [-1, 1], \]

\[ A(\mathcal{R}) = \mu_\mathcal{R}^3 + \nu_\mathcal{R}^3, \quad \text{where} \ A(\mathcal{R}) \in [0, 1]. \]
Let three FFNs; then, 

\[ (1 + (μ_{\mathcal{R}}(x))^3)^{1/3} - (1 - (μ_{\mathcal{R}}(x))^3)^{1/3} \]

\[ (1 + (μ_{\mathcal{R}}(x))^3)^{1/3} + (1 - (μ_{\mathcal{R}}(x))^3)^{1/3} \]

\[ \sqrt[3]{(2 - (ν_{\mathcal{R}}(x))^3)} + (ν_{\mathcal{R}}(x))^3 \]

\[ \frac{\sqrt[3]{(2 - (ν_{\mathcal{R}}(x))^3)} + (ν_{\mathcal{R}}(x))^3}{1 + (μ_{\mathcal{R}}(x))^3 + (1 - (μ_{\mathcal{R}}(x))^3)} \leq 1. \]

Thus,

\[ \frac{\sqrt[3]{(2 - (ν_{\mathcal{R}}(x))^3)} + (ν_{\mathcal{R}}(x))^3}{1 + (μ_{\mathcal{R}}(x))^3 + (1 - (μ_{\mathcal{R}}(x))^3)} = 0, \]

iff $μ_{\mathcal{R}}(x) = ν_{\mathcal{R}}(x) = 0$ and

\[ \frac{\sqrt[3]{(2 - (ν_{\mathcal{R}}(x))^3)} + (ν_{\mathcal{R}}(x))^3}{1 + (μ_{\mathcal{R}}(x))^3 + (1 - (μ_{\mathcal{R}}(x))^3)} = 1, \]

iff $(μ_{\mathcal{R}}(x))^3 + (ν_{\mathcal{R}}(x))^3 = 1$.

Thus, $\mathcal{R}_4 = \lambda \cdot \mathcal{R}$ is an FFN for $\lambda > 0.$

### Theorem 2

Let $\lambda, λ_1, λ_2 \geq 0$; then,

(i) $\mathcal{R}_1 \oplus \mathcal{R}_2 = \mathcal{R}_1 \oplus \mathcal{R}_1$

(ii) $\mathcal{R}_1 \otimes \mathcal{R}_2 = \mathcal{R}_1 \otimes \mathcal{R}_2$

(iii) $λ \cdot (\mathcal{R}_1 \oplus \mathcal{R}_2) = λ \cdot \mathcal{R}_1 \oplus λ \cdot \mathcal{R}_2$
(iv) $\mathcal{R}_1 \otimes \mathcal{R}_2 = \mathcal{R}_1 \otimes \mathcal{R}_2$
(v) $\lambda_1 \cdot \mathcal{R}_1 \cdot \mathcal{R}_2 = (\lambda_1 + \lambda_2) \cdot \mathcal{R}$
(vi) $\mathcal{R}_1 \otimes_L \mathcal{R}_2 = \mathcal{R}_1$ is equivalent to

$$\mathcal{R}_1 \otimes_L \mathcal{R}_2 = \left\langle \left( 1 + \mu_1^3 \cdot \epsilon, \frac{\epsilon^2 - (1 - \mu_1^3) \cdot \epsilon}{1 + \mu_1^3 \cdot \epsilon} \right)^3 \cdot \epsilon \left( 1 - \mu_2^3 \cdot \epsilon \right)^3 \cdot \epsilon \left( 1 - \mu_2^3 \cdot \epsilon \right) \cdot \epsilon \right\rangle.$$

By the Einstein FF law,

$$\lambda \cdot \mathcal{R}_1 = \left\langle \left( 1 + \mu_1^3 \cdot \epsilon, \frac{\epsilon^2 - (1 - \mu_1^3) \cdot \epsilon}{1 + \mu_1^3 \cdot \epsilon} \right)^3 \cdot \epsilon \left( 1 - \mu_2^3 \cdot \epsilon \right)^3 \cdot \epsilon \left( 1 - \mu_2^3 \cdot \epsilon \right) \cdot \epsilon \right\rangle = \left\langle \frac{\epsilon^3 - (1 - \mu_1^3) \cdot \epsilon}{1 + \mu_1^3 \cdot \epsilon} \cdot \epsilon \left( 1 - \mu_2^3 \cdot \epsilon \right) \cdot \epsilon \right\rangle.$$
where \( a_1 = (1 + \mu_1^1)^3, \ b_1 = (1 - \mu_1^1)^3, \ c_1 = (\nu_1^1)^3, \)
\( d_1 = (2 - \nu_1^1)^3, \ a_2 = (1 + \mu_2^3)^3, \ b_2 = (1 - \mu_2^3)^3, \ c_2 = (\nu_2^3)^3, \)
and \( d_2 = (2 - \nu_2^3)^3; \) therefore,

\[
(\lambda \cdot \mathcal{R}_1)\Phi_2(\lambda \cdot \mathcal{R}_2) = \left\langle \frac{\sqrt{2c_1}}{\sqrt{d_1 + c_1}} \right\rangle^{\Phi_2} \left\langle \frac{\sqrt{2c_2}}{\sqrt{d_2 + c_2}} \right\rangle^{\Phi_2}
\]

\[
= \left\langle \frac{\sqrt{2} \cdot (1 - \mu_1^1)^3 \cdot (1 - \mu_2^3)^3}{\sqrt{2} \cdot (2 - \nu_1^1)^3 \cdot (2 - \nu_2^3)^3} \right\rangle^{\Phi_2}
\]

\[
= \left\langle \frac{2^{2/3} \cdot \sqrt{2} c_1 \cdot c_2 \cdot (d_1 + c_1) \cdot (d_2 + c_2)}{1 + (1 - (2c_1/(d_1 + c_1)) \cdot (1 - (2c_2/(d_2 + c_2)))}
\]

\[
(\lambda \cdot \mathcal{R}_1)\Phi_2(\lambda \cdot \mathcal{R}_2) = \lambda \cdot \mathcal{R}_1 \Phi_2 \lambda \cdot \mathcal{R}_2.
\]

Hence, \( \lambda \cdot (\mathcal{R}_1 \Phi_2 \mathcal{R}_2) = \lambda \cdot \mathcal{R}_1 \Phi_2 \lambda \cdot \mathcal{R}_2. \)

(v) For \( \lambda_1, \lambda_2 > 0, \)

\[
\lambda_1 \cdot \mathcal{R} = \left\langle \frac{\sqrt{2} \cdot (1 - \mu_1^1)^3}{\sqrt{2} \cdot (2 - \nu_1^1)^3} \right\rangle^{\Phi_2} = \left\langle \frac{\sqrt{2c_1}}{\sqrt{d_1 + c_1}} \right\rangle^{\Phi_2}
\]

\[
\lambda_2 \cdot \mathcal{R} = \left\langle \frac{\sqrt{2} \cdot (1 - \mu_2^3)^3}{\sqrt{2} \cdot (2 - \nu_2^3)^3} \right\rangle^{\Phi_2} = \left\langle \frac{\sqrt{2c_2}}{\sqrt{d_2 + c_2}} \right\rangle^{\Phi_2}
\]

where \( a_j = (1 + \mu_j^3)^3, \ b_j = (1 - \mu_j^3)^3, \ c_j = (\nu_j^3)^3, \) and 
\( d_j = (2 - \nu_j^3)^3, \) for \( j = 1, 2. \)

\[
(\lambda_1 \cdot \mathcal{R})\Phi_2(\lambda_2 \cdot \mathcal{R}) = \left\langle \frac{\sqrt{2c_1}}{\sqrt{d_1 + c_1}} \right\rangle^{\Phi_2} \left\langle \frac{\sqrt{2c_2}}{\sqrt{d_2 + c_2}} \right\rangle^{\Phi_2}
\]

\[
= \left\langle \frac{\sqrt{2} \cdot (1 - \mu_1^1)^3 \cdot (1 - \mu_2^3)^3}{\sqrt{2} \cdot (2 - \nu_1^1)^3 \cdot (2 - \nu_2^3)^3} \right\rangle^{\Phi_2}
\]

\[
= \left\langle \frac{2^{2/3} \cdot \sqrt{2} c_1 \cdot c_2 \cdot (d_1 + c_1) \cdot (d_2 + c_2)}{1 + (1 - (2c_1/(d_1 + c_1)) \cdot (1 - (2c_2/(d_2 + c_2)))}
\]

\[
= (\lambda_1 + \lambda_2) \cdot \mathcal{R}.
\]
Hence, \( \lambda_1 \cdot \mathcal{F}_\lambda \lambda_2 \cdot \mathcal{F} = (\lambda_1 + \lambda_2) \cdot \mathcal{F} \).
Similarly, others can be verified. \(\square\)

**Theorem 3.** Let \( \mathcal{R}_1 = \langle \mu_1, v_1 \rangle \) and \( \mathcal{R}_2 = \langle \mu_2, v_2 \rangle \) be FFNs; then,

1. \( \mathcal{R}_1 \wedge \mathcal{R}_2 = (\mathcal{R}_1 \vee \mathcal{R}_2)^\epsilon \)
2. \( \mathcal{R}_1 \vee \mathcal{R}_2 = (\mathcal{R}_1 \wedge \mathcal{R}_2)^\epsilon \)
3. \( \mathcal{R}_1 \oplus \mathcal{R}_2 = (\mathcal{R}_1 \otimes \mathcal{R}_2)^\epsilon \)
4. \( \mathcal{R}_1 \otimes \mathcal{R}_2 = (\mathcal{R}_1 \oplus \mathcal{R}_2)^\epsilon \)
5. \( (\mathcal{R}_1 \lor \mathcal{R}_2) \otimes (\mathcal{R}_1 \land \mathcal{R}_2) = \mathcal{R}_1 \otimes \mathcal{R}_2 \)
6. \( (\mathcal{R}_1 \lor \mathcal{R}_2) \otimes (\mathcal{R}_1 \land \mathcal{R}_2) = \mathcal{R}_1 \otimes \mathcal{R}_2 \)

**Proof.** It is obvious. \(\square\)

**Theorem 4.** Let \( \mathcal{R}_1 = \langle \mu_1, v_1 \rangle \), \( \mathcal{R}_2 = \langle \mu_2, v_2 \rangle \), and \( \mathcal{R}_3 = \langle \mu_3, v_3 \rangle \) be three FFNs; then,

1. \( \mathcal{R}_1 \lor \mathcal{R}_2 \land \mathcal{R}_3 = (\mathcal{R}_1 \land \mathcal{R}_2) \lor (\mathcal{R}_2 \land \mathcal{R}_3) \)
2. \( \mathcal{R}_1 \land \mathcal{R}_2 \lor \mathcal{R}_3 = (\mathcal{R}_1 \lor \mathcal{R}_2) \land (\mathcal{R}_2 \lor \mathcal{R}_3) \)
3. \( \mathcal{R}_1 \oplus \mathcal{R}_2 \otimes \mathcal{R}_3 = (\mathcal{R}_1 \otimes \mathcal{R}_2) \oplus (\mathcal{R}_2 \otimes \mathcal{R}_3) \)
4. \( \mathcal{R}_1 \otimes \mathcal{R}_2 \oplus \mathcal{R}_3 = (\mathcal{R}_1 \oplus \mathcal{R}_2) \otimes (\mathcal{R}_2 \oplus \mathcal{R}_3) \)
5. \( (\mathcal{R}_1 \lor \mathcal{R}_2) \otimes (\mathcal{R}_1 \land \mathcal{R}_2) \lor \mathcal{R}_3 = (\mathcal{R}_1 \lor \mathcal{R}_2) \land (\mathcal{R}_2 \lor \mathcal{R}_3) \)
6. \( (\mathcal{R}_1 \lor \mathcal{R}_2) \otimes (\mathcal{R}_1 \land \mathcal{R}_2) \lor \mathcal{R}_3 = (\mathcal{R}_1 \land \mathcal{R}_2) \lor (\mathcal{R}_2 \land \mathcal{R}_3) \)

**Proof.** Use the mathematical induction to prove equation (20).

When \( \sigma = 2 \),

\[
\text{FFEW}(\mathcal{R}_1, \mathcal{R}_2) = w_1 \cdot \mathcal{R}_1 \otimes w_2 \cdot \mathcal{R}_2.
\]

By Theorem 1, both \( w_1 \cdot \mathcal{R}_1 \) and \( w_2 \cdot \mathcal{R}_2 \) are FFNs and value of \( w_1 \cdot \mathcal{R}_1 \otimes w_2 \cdot \mathcal{R}_2 \) is an FFN. By using (vi) in Definition 6,

\[
\begin{align*}
w_1 \cdot \mathcal{R}_1 &= \sqrt{\frac{(1 + \mu_1)^{w_1} - (1 - \mu_1)^{w_1}}{(1 + \mu_1)^{w_1} + (1 - \mu_1)^{w_1}}}, \\
w_2 \cdot \mathcal{R}_2 &= \sqrt{\frac{(1 + \mu_2)^{w_2} - (1 - \mu_2)^{w_2}}{(1 + \mu_2)^{w_2} + (1 - \mu_2)^{w_2}}},
\end{align*}
\]

**4. Fermatean Fuzzy Einstein Weighted Averaging Operators**

The Einstein weighted averaging operators under FF environment are defined here.

**Definition 7.** Let \( \mathcal{R}_j = \langle \mu_j, v_j \rangle \) be a collection of FFNs and \( w_j \) be the weight vector (VW) of \( \mathcal{R}_j \) with \( w_j > 0 \) and \( \sum_{j=1}^{J} w_j = 1 \); then, FFEWA operator is a mapping \( \mathcal{Q}^J \rightarrow \mathcal{Q} \) such that

\[
\text{FFEW}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_J) = w_1 \cdot \mathcal{R}_1 \otimes w_2 \cdot \mathcal{R}_2 \otimes \cdots \otimes w_J \cdot \mathcal{R}_J.
\]

(18)

If \( w_j = (1/\sigma), \forall j \), then FFEWA operator becomes FFA operator:

\[
\text{FFA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_J) = \frac{1}{\sigma} \left( \mathcal{R}_1 \oplus \mathcal{R}_2 \oplus \cdots \oplus \mathcal{R}_J \right).
\]

(19)

**Theorem 5.** Let \( \mathcal{R}_j = \langle \mu_j, v_j \rangle \) be FFNs; then, the aggregated value by using equation (18) is

\[
\text{FFEW}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_J) = \sqrt{\frac{\prod_{j=1}^{J} \left( (1 + \mu_j)^{w_j} - (1 - \mu_j)^{w_j} \right)^{w_j} + \prod_{j=1}^{J} \left( (1 + \mu_j)^{w_j} + (1 - \mu_j)^{w_j} \right)^{w_j}}{\prod_{j=1}^{J} \left( (2 - \mu_j)^{w_j} - (1 - \mu_j)^{w_j} \right)^{w_j} + \prod_{j=1}^{J} \left( (2 - \mu_j)^{w_j} + (1 - \mu_j)^{w_j} \right)^{w_j}}}.
\]

(20)
Then,

\[
\text{FFEWA}(R_1, R_2) = w_1 \cdot R_1 \Phi_i w_2 \cdot R_2
\]

\[
= \left( \frac{\sqrt{2} v_1^{w_1} \sqrt{2} v_1^{w_1} + (v_1^{w_1})^2}}{\sqrt{2} v_2^{w_2} + (v_2^{w_2})^2} \right) \cdot \left( \frac{\sqrt{2} v_2^{w_2} \sqrt{2} v_2^{w_2} + (v_2^{w_2})^2}}{\sqrt{2} v_3^{w_3} + (v_3^{w_3})^2} \right)
\]

\[
= \left( \frac{(1 + \mu_1^{w_1}) v_1^{w_1} (1 + \mu_2^{w_2}) v_2^{w_2} (1 + \mu_3^{w_3}) v_3^{w_3}}{(1 + \mu_1^{w_1}) v_1^{w_1} (1 + \mu_2^{w_2}) v_2^{w_2} (1 + \mu_3^{w_3}) v_3^{w_3}} \right)
\]

Thus, equation (20) is true when \( \delta = 2 \).

Suppose result is true for \( \delta = k \):

\[
\text{FFEWA}(R_1, R_2, \ldots, R_k) = \left( \prod_{j=1}^{k} (1 + \mu_j^{w_j}) v_j^{w_j} - \prod_{j=1}^{k} (1 - \mu_j^{w_j}) v_j^{w_j} \right) \cdot \frac{\sqrt{2} \prod_{j=1}^{k} v_j^{w_j}}{\sqrt{2} \prod_{j=1}^{k} (2 - v_j^{w_j}) + \prod_{j=1}^{k} (v_j^{w_j})}
\]

Now, for \( \delta = k + 1 \),

\[
\text{FFEWA}(R_1, R_2, \ldots, R_{k+1}) = \left( \prod_{j=1}^{k+1} (1 + \mu_j^{w_j}) v_j^{w_j} - \prod_{j=1}^{k+1} (1 - \mu_j^{w_j}) v_j^{w_j} \right) \cdot \frac{\sqrt{2} \prod_{j=1}^{k+1} v_j^{w_j}}{\sqrt{2} \prod_{j=1}^{k+1} (2 - v_j^{w_j}) + \prod_{j=1}^{k+1} (v_j^{w_j})}
\]

Thus, the result is true for \( \delta = k + 1 \). Hence, equation (20) holds, \( \forall \delta \).

**Lemma 1.** Let \( \mathcal{R}_j = \langle \mu_j, v_j \rangle \), \( w_j > 0 \), and \( \sum_{j=1}^{\delta} w_j = 1 \); then,

\[
\prod_{j=1}^{\delta} w_j R_j \leq \sum_{j=1}^{\delta} w_j R_j
\]

where equality holds iff \( R_1 = R_2 = \cdots = R_j \).

**Theorem 6.** If \( \mathcal{R}_j = \langle \mu_j, v_j \rangle \) are FFNs, then FFEWA \((\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_\delta)\) is also an FFN.

**Proof.** Since \( \mathcal{R}_j = \langle \mu_j, v_j \rangle \) are FFNs, so \( 0 \leq \mu_j, v_j \leq 1 \) and \( 0 \leq \mu_j + v_j \leq 1 \). Therefore,
\[
\prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} - \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} = 1
\]
\[
\prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j}
\]
\[
\leq 1 - \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \leq 1.
\]

Also,
\[
(1 + \mu_j^3) \geq (1 - \mu_j^3) \Rightarrow \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right) - \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right) \\
\prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} - \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \\
\prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \geq 0.
\]

Thus, \(0 \leq \mu_{\text{FFWA}} \leq 1. \)

Moreover,
\[
\mu_{\text{FFWA}} + \nu_{\text{FFWA}} = \frac{\prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} - \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j}}{\prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j}} + \frac{2 \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j}}{\prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j}}
\]
\[
\leq 1 - \frac{2 \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j}}{\prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j}} + \frac{2 \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j}}{\prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j}}
\]
\[
= 1.
\]

\[ \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} - \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \leq \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \]
\[ \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \geq 0. \]

Thus, \(0 \leq \nu_{\text{FFWA}} \leq 1. \) Moreover,
\[
\prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} - \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \leq \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \]
\[ \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \geq 0. \]

Hence, \( \text{FFWA} \in [0, 1]. \) Therefore, \( \text{FFEA} (\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_\delta) \in \mathcal{F}. \)

\[ \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} - \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \leq \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \]
\[ \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \geq 0. \]

Corollary 1. The FFEWA and FFWA operators have the relationship:
\[ \text{FFEWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_\delta) \leq \text{FFWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_\delta). \]

Proof. Let \( \text{FFEWA} (\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_\delta) = (\mu_{\text{R}}, \nu_{\text{R}}) = \mathcal{R}_\delta \)
and \( \text{FFWA} (\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_\delta) = (\mu_{\text{R}}, \nu_{\text{R}}) = \mathcal{R}. \) Since
\[ \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \leq \sum_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \sum_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} = 2, \]
then from equation (27), we obtain
\[
\prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} - \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \leq \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \]
\[ \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \geq 0. \]

equality holds iff \( \mu_1 = \mu_2 = \cdots = \mu_j. \)

Also,
\[
\prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} - \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \leq \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \]
\[ \prod_{j=1}^{\delta} \left(1 + \mu_j^3 \right)^{-w_j} + \prod_{j=1}^{\delta} \left(1 - \mu_j^3 \right)^{-w_j} \geq 0. \]
equality holds iff \( \nu_1 = \nu_2 = \cdots = \nu_j \).

Thus,
\[
\delta(\mathcal{R}) = (\mu^\beta) - (\nu^\beta) \leq (\mu^\beta) - (\nu^\beta) = \delta(\mathcal{R}).
\] (35)

If \( \delta(\mathcal{R}^\beta) < \delta(\mathcal{R}) \), then
\[
\text{FFEWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_j) < \text{FFEWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_j).
\] (36)

If \( \delta(\mathcal{R}^\beta) = \delta(\mathcal{R}) \), that is, \( (\mu^\beta) - (\nu^\beta) = (\mu^\beta) - (\nu^\beta) \), then by condition \( \mu^\beta \leq \mu^\beta \) and \( \nu^\beta \geq \nu^\beta \), thus, the accuracy function \( \mathcal{A}(\mathcal{R}) = (\mu^\beta) - (\nu^\beta) = (\mu^\beta) - (\nu^\beta) = \mathcal{A}(\mathcal{R}) \). Thus,

\[
\text{FFEWA}(\mathcal{R}_1, \mathcal{R}_2, \mathcal{R}_3, \mathcal{R}_4) = \left\langle \frac{\sum_{j=1}^{4} (1 + \mu_j)^{w_j} - \sum_{j=1}^{4} (1 - \mu_j)^{w_j}}{\prod_{j=1}^{4} (1 + \mu_j^{w_j}) + \prod_{j=1}^{4} (1 - \mu_j^{w_j})} \right\rangle
\] (39)

Now,

\[
\text{FFEWA}(\mathcal{R}_1, \mathcal{R}_2, \mathcal{R}_3, \mathcal{R}_4)
= \left\langle \frac{1}{\sqrt{1 - \prod_{j=1}^{4} (1 - \mu_j^{w_j})}} \prod_{j=1}^{4} (\nu_j^{w_j}) \right\rangle = \left\langle \frac{1.49 \times 0.55}{1.80} \right\rangle = \left\langle 0.80, 0.55 \right\rangle.
\] (40)

**Proposition 1.** Let \( \mathcal{R}_j = \langle \mu_j, \nu_j \rangle \) be FFNs and \( w_j \) be the WV of \( \mathcal{R}_j \), such that \( w_j \in [0, 1] \) and \( \sum_{j=1}^{4} w_j = 1 \).

(i) Idempotency: if \( \mathcal{R}_j = \mathcal{R}_o = \langle \mu_o, \nu_o \rangle \) for all \( j \), then
\[
\text{FFEWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_j) = \mathcal{R}_o.
\] (41)

(ii) Boundedness: let \( \mathcal{R}^- = (\min_j (\mu_j), \max_j (\nu_j)) \) and \( \mathcal{R}^+ = (\max_j (\mu_j), \min_j (\nu_j)) \); then,

\[
\text{FFEWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_j) \leq \text{FFEWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_j).
\] (42)

Example 1. Let \( \mathcal{R}_1 = (0.8, 0.5), \mathcal{R}_2 = (0.9, 0.4), \mathcal{R}_3 = (0.6, 0.7), \) and \( \mathcal{R}_4 = (0.8, 0.7) \) be four FFNs and \( w = (0.4, 0.2, 0.2, 0.2) \); then,

\[
\text{FFEWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_j) = \text{FFEWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_j).
\] (37)

Hence,

\[
\text{FFEWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_j) < \text{FFEWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_j).
\] (38)

equality holds iff \( \mathcal{R}_1 = \mathcal{R}_2 = \cdots = \mathcal{R}_j \), \( \square \)

Proof. (i) As \( \mathcal{R}_j = \langle \mu_o, \nu_o \rangle \) are FFNs, \( \forall \mathcal{R}_j \), then
(ii) Consider \( f(x) = ((1-x)/(1+x)), \ x \in [0,1] \), then \( f'(x) = -2/(1+x)^2 < 0 \), so \( f(x) \) is a decreasing function (DF). As \( \mu_{j,\min}^3 \leq \mu_j^3 \leq \mu_{j,\max}^3, \forall j = 1,2,\ldots, \delta \), then \( f(\mu_{j,\max}^3) \leq f(\mu_j^3) \leq f(\mu_{j,\min}^3), \forall j \), that is, \( ((1-\mu_j^3)/1+\mu_j^3) \leq ((1-\mu_j^3)/(1+\mu_j^3)) \leq ((1-\mu_{j,\min}^3)/(1+\mu_{j,\min}^3)), \) for all \( j \). Let \( w_j \in [0,1] \) and \( \sum_{j=1}^\delta w_j = 1 \), and we have

\[
\prod_{j=1}^\delta \left( \frac{1-\mu_{j,\max}^3}{1+\mu_j^3} \right)^{w_j} \leq \prod_{j=1}^\delta \left( \frac{1-\mu_{j,\min}^3}{1+\mu_j^3} \right)^{w_j} \leq \prod_{j=1}^\delta \left( \frac{1-\mu_{j,\max}^3}{1+\mu_j^3} \right)^{w_j} \, ,
\]

\[
\prod_{j=1}^\delta \left( \frac{1-\mu_j^3}{1+\mu_j^3} \right)^{w_j} \leq \prod_{j=1}^\delta \left( \frac{1-\mu_{j,\min}^3}{1+\mu_j^3} \right)^{w_j} \leq \prod_{j=1}^\delta \left( \frac{1-\mu_{j,\max}^3}{1+\mu_j^3} \right)^{w_j} \leq \left( \frac{1}{2} \right)^{\sum_{j=1}^\delta w_j} \leq \left( \frac{1}{2} \right)^{\sum_{j=1}^\delta w_j} \, .
\]

\[
\prod_{j=1}^\delta \left( \frac{1-\mu_j^3}{1+\mu_j^3} \right)^{w_j} \leq \prod_{j=1}^\delta \left( \frac{1-\mu_{j,\min}^3}{1+\mu_j^3} \right)^{w_j} \leq \prod_{j=1}^\delta \left( \frac{1-\mu_{j,\max}^3}{1+\mu_j^3} \right)^{w_j} \leq \frac{1}{2} \leq \prod_{j=1}^\delta \left( \frac{1-\mu_j^3}{1+\mu_j^3} \right)^{w_j} \leq \frac{1}{2} \, .
\]

\[
\prod_{j=1}^\delta \left( 1+\mu_j^3 \right)^{w_j} \leq \prod_{j=1}^\delta \left( 1+\mu_{j,\min}^3 \right)^{w_j} \leq \prod_{j=1}^\delta \left( 1+\mu_{j,\max}^3 \right)^{w_j} \, ,
\]

\[
\prod_{j=1}^\delta \left( 1+\mu_j^3 \right)^{w_j} \leq \prod_{j=1}^\delta \left( 1+\mu_{j,\min}^3 \right)^{w_j} \leq \prod_{j=1}^\delta \left( 1+\mu_{j,\max}^3 \right)^{w_j} \, .
\]

Thus, \( \mu_{j,\min} \leq \frac{1}{\prod_{j=1}^\delta \left( 1+\mu_j^3 \right)^{w_j} - \prod_{j=1}^\delta \left( 1-\mu_j^3 \right)^{w_j}} \leq \mu_{j,\max} \).
\[
\left( \frac{2 - v_{j,\text{max}}^3}{v_{j,\text{max}}^3} \right)^{w_j} \leq \left( \frac{2 - v_{j,\text{max}}^3}{v_{j,\text{max}}^3} \right)^{w_j} \leq \left( \frac{2 - v_{j,\text{min}}^3}{v_{j,\text{min}}^3} \right)^{w_j}
\]

\[
\prod_{j=1}^{\beta} \left( \frac{2 - v_{j,\text{max}}^3}{v_{j,\text{max}}^3} \right)^{w_j} \leq \prod_{j=1}^{\beta} \left( \frac{2 - v_{j,\text{max}}^3}{v_{j,\text{min}}^3} \right)^{w_j} \leq \prod_{j=1}^{\beta} \left( \frac{2 - v_{j,\text{min}}^3}{v_{j,\text{min}}^3} \right)^{w_j}
\]

\[
(2 - v_{j,\text{max}}^3) \sum_{j=1}^{\beta} w_j \leq (2 - v_{j,\text{max}}^3) \sum_{j=1}^{\beta} w_j \leq (2 - v_{j,\text{min}}^3) \sum_{j=1}^{\beta} w_j
\]

\[
\Rightarrow \left( \frac{2 - v_{j,\text{max}}^3}{v_{j,\text{max}}^3} \right)^{w_j} \leq \left( \frac{2 - v_{j,\text{max}}^3}{v_{j,\text{max}}^3} \right)^{w_j} \leq \left( \frac{2 - v_{j,\text{min}}^3}{v_{j,\text{min}}^3} \right)^{w_j}
\]

\[
\Rightarrow (2 - v_{j,\text{max}}^3) \leq 1 + \prod_{j=1}^{\beta} \left( \frac{2 - v_{j,\text{max}}^3}{v_{j,\text{max}}^3} \right)^{w_j} \leq \left( \frac{2 - v_{j,\text{max}}^3}{v_{j,\text{max}}^3} \right)^{w_j}
\]

\[
\Rightarrow \left( \frac{v_{j,\text{min}}^3}{2} \right)^{w_j} \leq \left( \frac{v_{j,\text{min}}^3}{2} \right)^{w_j} \leq \left( \frac{v_{j,\text{max}}^3}{2} \right)^{w_j}
\]

\[
\Rightarrow \left( \frac{v_{j,\text{max}}^3}{2} \right)^{w_j} \leq \left( \frac{v_{j,\text{max}}^3}{2} \right)^{w_j} \leq \left( \frac{v_{j,\text{max}}^3}{2} \right)^{w_j}
\]

\[
\Rightarrow v_{j,\text{max}}^3 \leq 1 + \prod_{j=1}^{\beta} \left( \frac{2 - v_{j,\text{max}}^3}{v_{j,\text{max}}^3} \right)^{w_j} \leq v_{j,\text{max}}^3
\]

\[
\Rightarrow v_{j,\text{min}}^3 \leq 1 + \prod_{j=1}^{\beta} \left( \frac{2 - v_{j,\text{max}}^3}{v_{j,\text{max}}^3} \right)^{w_j} \leq v_{j,\text{max}}^3
\]

\[
\Rightarrow \frac{\sqrt{2} \prod_{j=1}^{\beta} v_{j,\text{max}}^3}{\prod_{j=1}^{\beta} (2 - v_{j,\text{max}}^3)^{w_j}} \leq v_{j,\text{max}}^3
\]

Let FFEWA \( (\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_j) = \mathcal{R} = \langle \mu_{\mathcal{R}}, v_{\mathcal{R}} \rangle \); then, from equations (46) and (47),

\[
\begin{align*}
\mu_{\text{min}} & \leq \mu_{\mathcal{R}} \leq \mu_{\text{max}}, \\
v_{\text{min}} & \leq v_{\mathcal{R}} \leq v_{\text{max}},
\end{align*}
\]

where \( \mu_{\text{min}} = \min \{ \mu_j \} \), \( \mu_{\text{max}} = \max \{ \mu_j \} \), \( v_{\text{min}} = \min \{ v_j \} \), and \( v_{\text{max}} = \max \{ v_j \} \). So, \( s(\mathcal{R}) = \mu_{\mathcal{R}} - \mu_{\text{min}} \leq \mu_{\text{max}} - \mu_{\text{min}} = s(\mathcal{R}^+) \) and \( s(\mathcal{R}) = \mu_{\mathcal{R}} - v_{\text{min}} \geq \mu_{\text{max}} - \mu_{\text{min}} = s(\mathcal{R}^-) \). As \( s(\mathcal{R}) < s(\mathcal{R}^+) \) and \( s(\mathcal{R}) > s(\mathcal{R}^-) \), so \( \mathcal{R}^- \leq \text{FFEWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_j) \leq \mathcal{R}^+ \). (49)

(iii) It is similar to (ii), so we omit it.

5. Fermatean Fuzzy Einstein Ordered Weighted Averaging Operators

**Definition 8.** Let \( \mathcal{R}_j = \langle \mu_j, v_j \rangle \) be a family of FFNs and \( w_j \) be the WV of \( \mathcal{R}_j \) with \( w_j > 0 \) and \( \sum_{j=1}^{\beta} w_j = 1 \); then, FFEOWA operator is a mapping \( \mathcal{Q}^\beta \rightarrow \mathcal{Q} \) such that

\[
\text{FFEOWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_j)
\]

\[
= w_1 \cdot \mathcal{R}_1(\mathcal{R}_2 \oplus \mathcal{R}_3 \oplus \cdots \oplus \mathcal{R}_k(\mathcal{R}_j))
\]

where \( (q(1), q(2), \ldots, q(j)) \) is the permutation of \( (f_1, f_2, \ldots, f_j) \) such that \( \mathcal{R}_{q(j-1)} \geq \mathcal{R}_{q(j)}, \forall j = 1, 2, \ldots, k \).

**Theorem 7.** Let \( \mathcal{R}_j = \langle \mu_j, v_j \rangle \) be FFNs; then, the aggregated value by using FFEOWA is an FFN and

\[
\text{FFEOWA}(\mathcal{R}_1, \mathcal{R}_2, \ldots, \mathcal{R}_j)
\]

\[
= \left( \sqrt{2} \prod_{j=1}^{\beta} v_{\mathcal{R}_j}^{w_j} \right) \prod_{j=1}^{\beta} (2 - v_{\mathcal{R}_j}^{w_j} + \prod_{j=1}^{\beta} (v_{\mathcal{R}_j}^{w_j})^{w_j})
\]

(51)
Proof. It is similar to Theorem 4.
We give some properties without their proofs. □

Corollary 2. The FFEOWA and FOWA operators have the relation:

\[ \text{FFEOWA}(R_1, R_2, \ldots, R_3) \leq \text{FFEOWA}(R_1, R_2, \ldots, R_4). \]

(52)

Example 2. Let \( R_1 = (0.6, 0.7), R_2 = (0.8, 0.7), R_3 = (0.6, 0.9), \) and \( R_4 = (0.9, 0.4) \) be four FFNs and \( w = (0.3, 0.3, 0.2, 0.2)^T \) as

\[
\begin{align*}
\delta(R_1) &= (0.6)^3 - (0.7)^3 = -0.13, \\
\delta(R_2) &= (0.8)^3 - (0.7)^3 = 0.17, \\
\delta(R_3) &= (0.6)^3 - (0.9)^3 = -0.51, \\
\delta(R_4) &= (0.9)^3 - (0.4)^3 = 0.67.
\end{align*}
\]

(53)

Now,

\[
\begin{align*}
\text{FFEOWA}(R_1, R_2, R_3, R_4) &= \left( \prod_{j=1}^4 (1 + \mu_{0,j}(R_j))^w_j - \prod_{j=1}^4 (1 - \mu_{0,j}(R_j))^w_j \right) \\
&= \left( \prod_{j=1}^4 (1 + \mu_{0,j}(R_j))^w_j - \prod_{j=1}^4 (1 - \mu_{0,j}(R_j))^w_j \right) \\
&= \left( \frac{\sqrt{2} \prod_{j=1}^4 \mu_{0,j}(R_j)^w_j}{\sqrt{2} \prod_{j=1}^4 (1 - \mu_{0,j}(R_j))^w_j + \prod_{j=1}^4 \mu_{0,j}(R_j)^w_j} \right) \\
&= \left( \frac{\sqrt{2} \prod_{j=1}^4 \mu_{0,j}(R_j)^w_j}{\sqrt{2} \prod_{j=1}^4 (1 - \mu_{0,j}(R_j))^w_j + \prod_{j=1}^4 \mu_{0,j}(R_j)^w_j} \right) \\
&= \langle 0.79, 0.55 \rangle.
\end{align*}
\]

(55)

Thus, by applying the FFEOWA operator, we obtain

\[ \text{FFEOWA}(R_1, R_2, \ldots, R_4) \leq \text{FFEOWA}(P_1, P_2, \ldots, P_3). \]

(59)

6. Generalized Fermatean Fuzzy Einstein Weighted Averaging Operators

Definition 9. Let \( R_j = \langle \mu_j, \nu_j \rangle \) be a collection of FFNs and \( w_j \) be the WV of \( R_j \) with \( w_j > 0 \) and \( \sum_{j=1}^J w_j = 1 \); then, GFFEWA operator is a mapping \( \xi : \mathcal{F} \rightarrow \mathcal{Q} \) such that

\[
\text{GFFEWA}(R_1, R_2, \ldots, R_j) = \left( w_j = (\mu_j, \nu_j) \right)^{1/\lambda},
\]

(60)

where \( \lambda > 0 \).

Particularly,

(i) If \( \lambda = 1 \), then GFFEWA becomes FFEWA

(ii) If \( w = ((1/3), (1/3), \ldots, (1/3))^T \), then GFFEWA

\[ (R_1, R_2, \ldots, R_j) = ((1/3) \cdot \xi_{j=1}^j(1/3))^{1/3} \]

Theorem 8. Let \( R_j = \langle \mu_j, \nu_j \rangle \) be FFNs and \( w_j \) be the WV of \( R_j \) with \( w_j > 0 \) and \( \sum_{j=1}^J w_j = 1 \); then, the aggregated value by applying the GFFEWA operator is an FFN and
Proof. Since

$$\mathcal{R}_x = \left\langle \frac{\sqrt{2} \mu_j}{\sqrt{(2 - \mu_j)^4 + (\mu_j^3)^4}}, \left\langle \frac{(1 + \nu_j)^4 - (1 - \nu_j)^4}{(1 + \nu_j)^4 + (1 - \nu_j)^4} \right\rangle \right\rangle$$

$$\Rightarrow \phi_{j^1 \cdot \mathcal{R}_x}, \mathcal{R}_x = \left\langle \frac{\sqrt{2} \mu_j}{\sqrt{(2 - \mu_j)^4 + (\mu_j^3)^4}}, \left\langle \frac{(1 + \nu_j)^4 - (1 - \nu_j)^4}{(1 + \nu_j)^4 + (1 - \nu_j)^4} \right\rangle \right\rangle$$

$$= \left\langle \frac{\sqrt{2} \mu_j}{\sqrt{(2 - \mu_j)^4 + (\mu_j^3)^4}}, \left\langle \frac{(1 + \nu_j)^4 - (1 - \nu_j)^4}{(1 + \nu_j)^4 + (1 - \nu_j)^4} \right\rangle \right\rangle$$

Therefore,

$$\delta_{j^1 \cdot \mathcal{R}_x}, \mathcal{R}_x' = \left\langle \frac{\sqrt{2} \mu_j}{\sqrt{(2 - \mu_j)^4 + (\mu_j^3)^4}}, \left\langle \frac{(1 + \nu_j)^4 - (1 - \nu_j)^4}{(1 + \nu_j)^4 + (1 - \nu_j)^4} \right\rangle \right\rangle$$
When $\lambda = 1$, then

$$
\text{GFEEOWA}(R_1, R_2, \ldots, R_s) = \left\langle \sqrt{2} \left[ \prod_{j=1}^{s} \left( 1 + \mu^3_{j} \right)^{w_j} - \prod_{j=1}^{s} (1 - \mu^3_{j})^{w_j} \prod_{j=1}^{s} (1 - \mu^3_{j})^{w_j} \right] \right\rangle.
$$

(64)

7. Generalized Fermatean Fuzzy Einstein Ordered Weighted Averaging Operators

**Definition 10.** Let $R_j = \{\mu_j, \nu_j\}$ be a collection of FFNs and $w_j$ be the WV of $R_j$ with $w_j > 0$ and $\sum_{j=1}^{s} w_j = 1$; then, the GFFEOWA operator is a mapping $G \rightarrow R$ such that

$$
\text{GFEEOWA}(R_1, R_2, \ldots, R_s) = \left\langle \sqrt{2} \left[ \prod_{j=1}^{s} \left( 1 - \nu^3_{j} \right)^{w_j} - \prod_{j=1}^{s} \left( \nu^3_{j} \right)^{w_j} \right] \right\rangle.
$$

(65)

Proof. It is similar to Theorem 6, and we can prove it. □

When $\lambda = 1$, then

$$
\text{GFEEOWA}(R_1, R_2, \ldots, R_s) = \left\langle \sqrt{2} \prod_{j=1}^{s} \left( 1 + \mu^3_{j} \right)^{w_j} - \prod_{j=1}^{s} (1 - \mu^3_{j})^{w_j} \prod_{j=1}^{s} (1 - \mu^3_{j})^{w_j} \right\rangle.
$$

(66)

8. MADM Problem Using FF Information

To handle an MADM problem under FF environment, let $\mathcal{K} = \{K_1, K_2, \ldots, K_m\}$ be a set of possible alternatives and $\mathcal{J} = \{J_1, J_2, \ldots, J_j\}$ be a set of possible attributes chosen by the decision maker. Let $w = (w_1, w_2, \ldots, w_j)^T$ be the WV with $w_j > 0$ and $\sum_{j=1}^{j} w_j = 1$. Suppose that $\mathcal{K} = (\mu_{ji}, \nu_{ji})_{i,j}$ is the FF decision matrix (FFDM), where $\mu_{ji}$ and $\nu_{ji}$ are the MD and NMD of the alternative $K_i$ for the attribute $J_j$, respectively, where $0 \leq \mu_{ji} + \nu_{ji} \leq 1$.

The following Algorithm 1 is used to solve the MADM problem with FFN based on using the GFFEWA operator.
the novel coronavirus. And some viruses—from functioning normally. Hand sanitizers work to prevent the proteins of microbes—including bacteria—remove the novel coronavirus. Alcohol-based hand sanitizer disease and also spread by touching a sick person. We cannot situate the situation of coronavirus, good hygiene and physical distancing spread of infections and decrease the endanger of getting sick. Hand sanitizer is a liquid or gel mostly used to reduce in-

8.1. Selection of an Effective Sanitizer to Reduce Coronavirus. Hand sanitizer is a liquid or gel mostly used to reduce infectious agents on the hands. Alcohol-based hand sanitizers are preferred for hand washing in most healthcare settings. The Centers for Disease Control and Prevention (CDC) advise the people to wash hands with soap and water to restrain the spread of infections and decrease the endanger of getting sick. In shortage of soap and water, CDC suggests people to use an alcohol-based (at least 60 percent) hand sanitizer. According to the World Health Organization (WHO), in this pandemic situation of coronavirus, good hygiene and physical distancing are the best ways to protect ourselves and everyone around us from coronavirus. This virus spreads by a person who has the disease and also spread by touching a sick person. We cannot isolate ourselves entirely to prudent from coronavirus. So, good hand hygiene can be the final barrier between us and the disease. WHO recommends alcohol-based hand sanitizers to remove the novel coronavirus. Alcohol-based hand sanitizer works to prevent the proteins of microbes—including bacteria and some viruses—from functioning normally. Hand sanitizers must contain ethanol, isopropanol, n-propanol, or a combination of these alcohols. All are effective against viruses such as the novel coronavirus.

Demand of a hand sanitizer is increased in such critical situation of COVID-19. Due to increasing demand, it is difficult to get good and effective hand sanitizers in local markets. Increasing demand has also led to low quality hand sanitizers entering the market. The main motive of this application is to select an effective sanitizer to mitigate transmission of coronavirus by applying the GFFYWA operator. Let $K = \{K_1, K_2, K_3, K_4\}$ be a set of sanitizers. Let $I = \{I_1, I_2, I_3\}$ be a set of three attributes for the evaluation of an effective sanitizer, where

$J_1$: represents quantity of ethanol,
$J_2$: represents quantity of glycerol,
$J_3$: represents quantity of hydrogenperoxide.

We use the GFFYWA operator for the selection of an effective sanitizer.

Step 1. For performance values $B_l$ of sanitizers, use the GFFYWA operator for $\lambda = 1$:

$B_1 = (0.64, 0.48), B_2 = (0.46, 0.28), B_3 = (0.72, 0.31), B_4 = (0.77, 0.34).$

Step 2. Calculate the scores $s(B_l)$ of FFNs $B_l$ and rank the sanitizers:
Step 3. Therefore, $\mathcal{H}_4$ is the best sanitizer.

### 9. Validity Test

For the validity and authenticity of MADM methods, Wang and Triantaphyllou [53] developed testing criteria, given as follows:

(i) Criterion 1: a MADM technique is valid if the most desirable alternative remains same on changes a nonoptimal alternative with some other poor or weak alternative, without changing the respective decision criteria

(ii) Criterion 2: the transitive property should be followed by a valid MADM technique

(iii) Criterion 3: the ranking result of alternatives should not change on splitting the problem into the smaller subproblems and by applying the same MADM technique on subproblems

Now, we discuss the validity of our proposed MADM technique by testing the above criteria.

(1) Validity test by criterion 1: if we replace the decision values of a nonoptimal alternative $\mathcal{H}_2$ by $\mathcal{H}_2$, then the new DM is given in Table 3.

| $\mathcal{H}$ | $\mathcal{J}_1$ | $\mathcal{J}_2$ | $\mathcal{J}_3$ |
|---------------|---------------|---------------|---------------|
| $\mathcal{H}_1$ | (0.6, 0.7)    | (0.6, 0.2)    | (0.8, 0.4)    |
| $\mathcal{H}_2$ | (0.4, 0.2)    | (0.6, 0.4)    | (0.3, 0.6)    |
| $\mathcal{H}_3$ | (0.8, 0.3)    | (0.5, 0.3)    | (0.6, 0.4)    |
| $\mathcal{H}_4$ | (0.7, 0.5)    | (0.8, 0.1)    | (0.9, 0.6)    |

By applying the GFFEWA operator for $\lambda = 1$ and score function, the score values of alternatives are

$$\delta(\mathcal{B}_1) = 0.15,$$
$$\delta(\mathcal{B}_2) = 0.08,$$
$$\delta(\mathcal{B}_3) = 0.34,$$
$$\delta(\mathcal{B}_4) = 0.42. \quad (72)$$

The ranking of sanitizers is $\mathcal{H}_4 > \mathcal{H}_3 > \mathcal{H}_1 > \mathcal{H}_2$. \quad (73)

(2) Validity test by criteria 2 and 3: for the validity of the FF TOPSIS method are

(3) FQ_he distance between the alternative

(4) FQ_he revised closeness degree of each alternative is

(5) We get the following ranking list by arranging the alternatives in the decreasing order with respect to $\xi(\mathcal{H})$:

$$\mathcal{H}_4 > \mathcal{H}_3 > \mathcal{H}_1 > \mathcal{H}_2. \quad (77)$$

(6) $\mathcal{H}_4$ is the best alternative.
From the outcomes of proposed operators and FF TOPSIS method, as shown in Figure 1, we conclude that ranking lists obtained from both compared methods are the same and the best alternative from both approaches is $K_4$. FF TOPSIS method is a good approach to solve DM problems but there are many hindrances which can be solved by using our proposed theory. FF Einstein AOs are more flexible and easy approach. A best alternative can be obtained by a short process. FF results from proposed theory are more accurate and closest to original results. The steps to solve any MADM problem by FF Einstein AOs and FF TOPSIS method are shown in Figure 2.

10.1. Advantages and Limitations of Proposed Model. The proposed model is superior than the IF and PF models because it contains the space of IF and PF models. The cubic sum of membership and nonmembership degrees is bounded by 1 in the proposed model. The MADM approaches discussed in [10, 12, 13, 15] failed to handle the proposed application because $0.9 + 0.6 > 1$ and $0.9^2 + 0.6^2 > 1$ but proposed approach covers all such situations. The results are more precise and accurate by using the proposed model. However, there are some limitations of this model. It cannot be applied in the situations where we take the parameters for the evaluation of anything. It means this theory lacks parametrization property.

11. Conclusions

An FFS is an extension of IFS and PFS which has more flexible structure to solve decision-making problems owing to the condition $\mu^3 + \nu^3 \leq 1$. Moreover, Einstein’s $t$-norm and $t$-conorm have more generalized structure that operates efficiently to integrate the intricate information. The limitations of existing operators and beneficial characteristics of Einstein AOs motivated us to endeavor for the development of a fruitful combination of Einstein AOs with FFNs.

A major contribution of the study is the development new tremendous AOs, called, FFEWA, FFEOWA, GFFEWA, and GFFEOWA operators. Some captivating properties of these operators have been discussed. Another achievement of this study is the establishment of a MADM technique on the basis of the proposed operators to manifest the application of the proposed operators. A MADM problem for the selection of an effective sanitizer to reduce the coronavirus has been presented to demonstrate the potency of the proposed strategy.

The validity test has been discussed to unfold the consistency of proposed work. A comparison analysis of our proposed theory with the FF TOPSIS method has been presented to exhibit the dominance of our proposed operators over the FF TOPSIS method. In short, this article builds up a tool that has the rich properties of Einstein AOs and flexibility of the FF model. In future, our aim is to develop some worthwhile AOs using the theoretical foundations of Einstein norms for the Fermatean fuzzy soft set.

Data Availability

No data were used to support this study.

Conflicts of Interest

The authors declare that they have no conflicts of interest regarding the publication of the research article.

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