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INTERVAL VALUED PICTURE FUZZY SOFT SET IN PATTERN RECOGNITION

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ABSTRACT. This paper deals with interval valued picture fuzzy soft set (IVPFSS) as a generalization of picture fuzzy soft set. We introduce the concept of normalized Euclidean distance between IVPFSS and establish that it is a metric. An algorithm based on this distance of IVPFSS is developed and an illustration is provided.

1. INTRODUCTION

The concept of picture fuzzy set was introduced by Cuong [1,2] as extensions of fuzzy sets and intuitionistic fuzzy sets (Atanassov). They defined some Operations on picture fuzzy sets established some of their properties. Cuong et al. [3] constructed main operations for fuzzy inference processes in picture fuzzy systems. Yang et al. [4] defined picture fuzzy soft set and introduced an algorithm by using level soft set and picture fuzzy soft set to solve decision making problems. Motivated by these concepts we have developed interval valued picture fuzzy soft set. This paper deals with interval valued picture fuzzy soft set (IVPFSS) as a generalization of picture fuzzy soft set. We introduce the concept of normalized Euclidean distance between IVPFSS and establish that it is

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a metric. An algorithm based on this distance of IVPFSS is developed and an illustration is provided.

2. INTERVAL VALUED PICTURE FUZZY SOFT SET

In this section some operations on IVPFSS are defined and some of their properties.

Definition 2.1. Let U be an universe and E be a set of parameters. Let IVPFSS(U)denote the set of all interval valued picture fuzzy soft set over U and $A \subseteq E$. A pair (P_F, A) is an IVPFSS over U, where P_F is a mapping given by $P_F : A \rightarrow IVPFSS(U)$ and

$$(P_F, A) = \{x, [\underline{\mu}_{P_{F(e)}}(x), \overline{\mu}_{P_{F(e)}}(x)], [\underline{\eta}_{P_{F(e)}}(x), \overline{\eta}_{P_{F(e)}}(x)], \\ [\underline{\nu}_{P_{F(e)}}(x), \overline{\nu}_{P_{F(e)}}(x)]); x \in U, e \in A\}.$$

For any parameter $e \in A$, $P_{F(e)}$ is an IVPFSS.

Definition 2.2. The necessity operator on an $IVPFSS(P_F, A)$ denoted by

$$(P_F, A) = \{ (x, [\underline{\mu}_{P_{F(e)}}(x), \overline{\mu}_{P_{F(e)}}(x)], [1 - \underline{\eta}_{P_{F(e)}}(x), 1 - \overline{\eta}_{P_{F(e)}}(x)], \\ [1 - \underline{\nu}_{P_{F(e)}}(x), 1 - \overline{\nu}_{P_{F(e)}}(x)]; x \in U, e \in A \}.$$

Definition 2.3. The possibility operator on an $IVPFSS(P_F, A)$ denoted by

$$(P_F, A) = \{ (x, [1 - \underline{\mu}_{P_{F(e)}}(x), 1 - \overline{\mu}_{P_{F(e)}}(x)], [\underline{\eta}_{P_{F(e)}}(x), \overline{\eta}_{P_{F(e)}}(x)], [\underline{\nu}_{P_{F(e)}}(x), \overline{\nu}_{P_{F(e)}}(x)]; x \in U, e \in A \}.$$

Definition 2.4. The complement of an $IVPFSS(P_F, A)$ denoted by $(P_F, A)^C = \{(x, [\underline{\nu}_{P_{F(e)}}(x), \overline{\nu}_{P_{F(e)}}(x)], [\underline{\eta}_{P_{F(e)}}(x), \overline{\eta}_{P_{F(e)}}(x)], [\underline{\mu}_{P_{F(e)}}(x), \overline{\mu}_{P_{F(e)}}(x)]; x \in U, e \in A\}.$

Definition 2.5. Let $A, B \subseteq E, (P_F, A)$ is an interval valued picture fuzzy soft subset of (P_F, B) denoted by $(P_F, A) \in (P_F, B)$, if, and only if,

- (i) $A \subseteq B$.
- (ii) $\forall e \in A, P_{F(e)}$ is an interval valued picture fuzzy soft subset of $P_{G(e)}$, i.e, $\forall x \in U, \forall e \in A, \underline{\mu}_{P_{F(e)}}(x) \leq \underline{\mu}_{P_{G(e)}}(x), \overline{\mu}_{P_{F(e)}}(x) \leq \overline{\mu}_{P_{G(e)}}(x), \underline{\eta}_{P_{F(e)}}(x) \geq \underline{\eta}_{P_{G(e)}}(x), \overline{\eta}_{P_{F(e)}}(x) \geq \overline{\eta}_{P_{G(e)}}(x), \underline{\nu}_{P_{F(e)}}(x) \geq \underline{\nu}_{P_{G(e)}}(x), \overline{\nu}_{P_{F(e)}}(x) \geq \overline{\nu}_{P_{G(e)}}(x).$

Further (P_G, B) is called an interval valued picture fuzzy superset of (P_F, A) and is denoted by $(P_G, B) \supseteq (P_F, A)$.

Definition 2.6. The degree of non-determinacy of an element $x \in U$, $e \in A$ to the $IVPFSS(P_F, A)$ is defined as

$$\underline{\pi}_{P_{F}(e)}(x) = (1 - \underline{\mu}_{P_{F}(e)}(x) - \underline{\eta}_{P_{F}(e)}(x) - \underline{\nu}_{P_{F}(e)}(x))$$

and

$$\overline{\pi}_{P_{F(e)}}(x) = 1 - (\overline{\mu}_{P_{F(e)}}(x) - \overline{\eta}_{P_{F(e)}}(x) - \overline{\nu}_{P_{F(e)}}(x)).$$

Definition 2.7. Let $\alpha \in [0, 1]$ be a fixed number. Given an $IVPFSS(P_F, A)$, the operator

$$\begin{split} \mathbb{D}_{\alpha}(P_{F},A) &= \{x, [(\underline{\mu}_{P_{F(e)}}(x) + \alpha \underline{\pi}_{P_{F(e)}}(x)), (\overline{\mu}_{P_{F(e)}}(x) + \alpha \overline{\pi}_{P_{F(e)}}(x))], \\ &\quad [(\underline{\eta}_{P_{F(e)}}(x) + (1-\alpha)(\underline{\pi}_{P_{F(e)}}(x)), (\overline{\eta}_{P_{F(e)}}(x) + (1-\alpha)(\overline{\pi}_{P_{F(e)}}(x))], \\ &\quad [(\underline{\nu}_{P_{F(e)}}(x) + (1-\alpha)(\underline{\pi}_{P_{F(e)}}(x)), (\overline{\nu}_{P_{F(e)}}(x) + (1-\alpha)(\overline{\pi}_{P_{F(e)}}(x))]; \\ &\quad x \in U, \ e \in A\}. \end{split}$$

Theorem 2.1. For every $IVPFSS(P_F, A)$ and $\alpha, \beta \in [0, 1]$, if $\alpha \leq \beta$ then $\mathbb{D}_{\alpha}(P_F, A) \Subset \mathbb{D}_{\beta}(P_F, B)$.

Proof. We have

$$\mathbb{D}_{\alpha}(P_{F}, A) = \{x, [(\underline{\mu}_{P_{F(e)}}(x) + \alpha \underline{\pi}_{P_{F(e)}}(x)), (\overline{\mu}_{P_{F(e)}}(x) + \alpha \overline{\pi}_{P_{F(e)}}(x))], [(\underline{\eta}_{P_{F(e)}}(x) + (1 - \alpha)(\underline{\pi}_{P_{F(e)}}(x)), (\overline{\eta}_{P_{F(e)}}(x) + (1 - \alpha)(\overline{\pi}_{P_{F(e)}}(x))], [(\underline{\nu}_{P_{F(e)}}(x) + (1 - \alpha)(\underline{\pi}_{P_{F(e)}}(x)), (\overline{\nu}_{P_{F(e)}}(x) + (1 - \alpha)(\overline{\pi}_{P_{F(e)}}(x))]; x \in U, e \in A\}$$

and

$$\mathbb{D}_{\beta}(P_{F}, A) = \{x, [(\underline{\mu}_{P_{F(e)}}(x) + \beta \underline{\pi}_{P_{F(e)}}(x)), (\overline{\mu}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x))], [(\underline{\eta}_{P_{F(e)}}(x) + (1 - \beta)(\underline{\pi}_{P_{F(e)}}(x)), (\overline{\eta}_{P_{F(e)}}(x) + (1 - \beta)(\overline{\pi}_{P_{F(e)}}(x))], [(\underline{\nu}_{P_{F(e)}}(x) + (1 - \beta)(\underline{\pi}_{P_{F(e)}}(x)), (\overline{\nu}_{P_{F(e)}}(x) + (1 - \beta)(\overline{\pi}_{P_{F(e)}}(x))]; x \in U, e \in A\}.$$

$$\begin{split} & \text{Similarly } \overline{\mu}_{P_{F(e)}}(x) + \alpha \overline{\pi}_{P_{F(e)}}(x) \leq \overline{\mu}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x), \text{ since } \alpha < \beta, (1-\alpha) > \\ & (1-\beta) \text{ and } (\underline{\eta}_{P_{F(e)}}(x) + (1-\beta)(\underline{\pi}_{P_{F(e)}}(x)) \leq (\underline{\eta}_{P_{F(e)}}(x) + (1-\alpha)(\underline{\pi}_{P_{F(e)}}(x)).\\ & \text{Similarly } (\overline{\eta}_{P_{F(e)}}(x) + (1-\beta)\overline{\pi}_{P_{F(e)}}(x)) \leq (\overline{\eta}_{P_{F(e)}}(x) + (1-\alpha)(\overline{\pi}_{P_{F(e)}}(x)) \text{ and}\\ & (\underline{\nu}_{P_{F(e)}}(x) + (1-\beta)(\underline{\pi}_{P_{F(e)}}(x)) \leq (\underline{\nu}_{P_{F(e)}}(x) + (1-\alpha)(\underline{\pi}_{P_{F(e)}}(x)).\\ & \text{Similarly, } (\overline{\nu}_{P_{F(e)}}(x) + (1-\beta)\overline{\pi}_{P_{F(e)}}(x)) \leq (\overline{\nu}_{P_{F(e)}}(x) + (1-\alpha)(\overline{\pi}_{P_{F(e)}}(x)).\\ & \text{Hence, it follows that } \mathbb{D}_{\alpha}(P_{F}, A) \Subset \mathbb{D}_{\beta}(P_{F}, B). \end{split}$$

Definition 2.8. For $\alpha, \beta \in [0, 1]$, $\alpha + 2\beta \leq 1$ the operator $\mathbb{F}_{\alpha, \beta}$ for an $IVPFSS(P_F, A)$ is defined as

$$\mathbb{F}_{\alpha,\beta}(P_F,A) = \{x, [(\underline{\mu}_{P_{F(e)}}(x) + \alpha \underline{\pi}_{P_{F(e)}}(x)), (\overline{\mu}_{P_{F(e)}}(x) + \alpha \overline{\pi}_{P_{F(e)}}(x))], (\underline{\mu}_{P_{F(e)}}(x) + \alpha \overline{\pi}_{P_{F(e)}}(x))\}$$

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$$\begin{split} &[(\underline{\eta}_{P_{F(e)}}(x) + \beta \underline{\pi}_{P_{F(e)}}(x)), (\overline{\eta}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x))], \\ &[(\underline{\nu}_{P_{F(e)}}(x) + \beta \underline{\pi}_{P_{F(e)}}(x)), (\overline{\nu}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x))]; \\ & x \in U, \ e \in A\}. \end{split}$$

Theorem 2.2. For $IVPFSS(P_F, A)$ and $\forall \alpha, \beta, \gamma \in [0, 1]$ such that $\alpha + 2\beta \leq 1$, the following hold:

(i) F_{α,β}(P_F, A) is an IVPFSS,
(ii) If 0 ≤ γ ≤ α then F_{γ,β}(P_F, A) ∈ F_{α,β}(P_F, A),
(iii) If 0 ≤ γ ≤ β then F_{α,β}(P_F, A) ∈ F_{α,γ}(P_F, A),
(iv) (F_{α,β}(P_F, A)^c)^c = F_{α,β}(P_F, A).

Proof.

(i) Consider,

$$\frac{\overline{\mu}_{P_{F(e)}}(x) + \alpha(\overline{\pi}_{P_{F(e)}}(x))}{2} + \frac{\overline{\eta}_{P_{F(e)}}(x) + \beta(\overline{\pi}_{P_{F(e)}}(x))}{2} + \frac{\overline{\nu}_{P_{F(e)}}(x) + \beta(\overline{\pi}_{P_{F(e)}}(x))}{2} \\
= \frac{\overline{\mu}_{P_{F(e)}}(x)}{2} + \frac{\overline{\eta}_{P_{F(e)}}(x)}{2} + \frac{\overline{\nu}_{P_{F(e)}}(x)}{2} + (\alpha + 2\beta)(\frac{\overline{\pi}_{P_{F(e)}}(x)}{2}) \\
\leq \frac{\overline{\mu}_{P_{F(e)}}(x)}{2} + \frac{\overline{\eta}_{P_{F(e)}}(x)}{2} + \frac{\overline{\nu}_{P_{F(e)}}(x)}{2} + \frac{1 - \overline{\mu}_{P_{F(e)}}(x) - \overline{\eta}_{P_{F(e)}}(x) - \overline{\nu}_{P_{F(e)}}(x)}{2} \\
\leq \frac{1}{2} < 1.$$

Hence $\mathbb{F}_{\alpha,\beta}(P_F, A)$ is an IVPFSS.

(ii)
$$\mathbb{F}_{\gamma,\beta}(P_F, A) = \{ x, [(\underline{\mu}_{P_{F(e)}}(x) + \gamma \underline{\pi}_{P_{F(e)}}(x)), (\overline{\mu}_{P_{F(e)}}(x) + \gamma \overline{\pi}_{P_{F(e)}}(x))], \\ [(\underline{\eta}_{P_{F(e)}}(x) + \beta \underline{\pi}_{P_{F(e)}}(x)), (\overline{\eta}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x))], \\ [(\underline{\nu}_{P_{F(e)}}(x) + \beta \underline{\pi}_{P_{F(e)}}(x)), (\overline{\nu}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x))], \\ x \in U, e \in A \} \\ \mathbb{F}_{\alpha,\beta}(P_F, A) = \{ x, [(\underline{\mu}_{P_{F(e)}}(x) + \alpha \underline{\pi}_{P_{F(e)}}(x)), (\overline{\mu}_{P_{F(e)}}(x) + \alpha \overline{\pi}_{P_{F(e)}}(x))], \\ (\overline{\mu}_{P_{F(e)}}(x) + \alpha \overline{\pi}_{P_{F(e)}}(x)), (\overline{\mu}_{P_{F(e)}}(x) + \alpha \overline{\pi}_{P_{F(e)}}(x))], \\ \end{bmatrix}$$

$$[(\underline{\eta}_{P_{F(e)}}(x) + \beta \underline{\pi}_{P_{F(e)}}(x)), (\overline{\eta}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x))], \\ [(\underline{\nu}_{P_{F(e)}}(x) + \beta \underline{\pi}_{P_{F(e)}}(x)), (\overline{\nu}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x))], \\ x \in U, e \in A\}$$

Now, $\underline{\mu}_{P_{F(e)}}(x) + \gamma \underline{\pi}_{P_{F(e)}}(x) \leq \underline{\mu}_{P_{F(e)}}(x) + \alpha \underline{\pi}_{P_{F(e)}}(x)$, since $\gamma \leq \alpha$. Similarly, we have $\overline{\mu}_{P_{F(e)}}(x) + \gamma \overline{\pi}_{P_{F(e)}}(x) \leq \overline{\mu}_{P_{F(e)}}(x) + \alpha \overline{\pi}_{P_{F(e)}}(x)$. Hence it follows that $\mathbb{F}_{\gamma,\beta}(P_F, A) \in \mathbb{F}_{\alpha,\beta}(P_F, A)$.

(iii)
$$\mathbb{F}_{\alpha,\beta}(P_{F}, A) = \{x, [(\underline{\mu}_{P_{F(e)}}(x) + \alpha \underline{\pi}_{P_{F(e)}}(x)), (\overline{\mu}_{P_{F(e)}}(x) + \alpha \overline{\pi}_{P_{F(e)}}(x))], [(\underline{\eta}_{P_{F(e)}}(x) + \beta \underline{\pi}_{P_{F(e)}}(x)), (\overline{\eta}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x))], [(\underline{\nu}_{P_{F(e)}}(x) + \beta \underline{\pi}_{P_{F(e)}}(x)), (\overline{\nu}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x))], x \in U, e \in A\}$$

$$\mathbb{F}_{\alpha,\gamma}(P_{F}, A) = \{x, [(\underline{\mu}_{P_{F(e)}}(x) + \alpha \underline{\pi}_{P_{F(e)}}(x)), (\overline{\mu}_{P_{F(e)}}(x) + \alpha \overline{\pi}_{P_{F(e)}}(x))], [(\underline{\eta}_{P_{F(e)}}(x) + \gamma \underline{\pi}_{P_{F(e)}}(x)), (\overline{\eta}_{P_{F(e)}}(x) + \gamma \overline{\pi}_{P_{F(e)}}(x))], [(\underline{\nu}_{P_{F(e)}}(x) + \gamma \underline{\pi}_{P_{F(e)}}(x)), (\overline{\nu}_{P_{F(e)}}(x) + \gamma \overline{\pi}_{P_{F(e)}}(x))], [(\underline{\nu}_{P_{F(e)}}(x) + \gamma \underline{\pi}_{P_{F(e)}}(x)), (\overline{\nu}_{P_{F(e)}}(x) + \gamma \overline{\pi}_{P_{F(e)}}(x))], x \in U, e \in A\}$$

Now, $\underline{\mu}_{P_{F(e)}}(x) + \gamma \underline{\pi}_{P_{F(e)}}(x) \leq \underline{\mu}_{P_{F(e)}}(x) + \beta \underline{\pi}_{P_{F(e)}}(x)$, since $\gamma \leq \beta$. Similarly, we have $\overline{\mu}_{P_{F(e)}}(x) + \gamma \overline{\pi}_{P_{F(e)}}(x) \leq \overline{\mu}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x)$. Hence, it follows that $\mathbb{F}_{\alpha,\beta}(P_F, A) \in \mathbb{F}_{\alpha,\gamma}(P_F, A)$.

(iv)
$$(\mathbb{F}_{\alpha,\beta}(P_F, A)^c = \{x, [(\underline{\nu}_{P_{F(e)}}(x) + \alpha \underline{\pi}_{P_{F(e)}}(x)), (\overline{\nu}_{P_{F(e)}}(x) + \alpha \overline{\pi}_{P_{F(e)}}(x))], [(\underline{\eta}_{P_{F(e)}}(x) + \beta \underline{\pi}_{P_{F(e)}}(x)), (\overline{\eta}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x))], [(\underline{\mu}_{FP(e)}(x) + \beta \underline{\pi}_{FP(e)}(x)), (\overline{\mu}_{FP(e)}(x) + \beta \overline{\pi}_{FP(e)}(x))], x \in U, e \in A\}.$$

Then

$$(\mathbb{F}_{\alpha,\beta}(P_{F},A)^{c})^{c} = \{ x, [(\underline{\mu}_{P_{F(e)}}(x) + \alpha \underline{\pi}_{P_{F(e)}}(x)), (\overline{\mu}_{P_{F(e)}}(x) + \alpha \overline{\pi}_{P_{F(e)}}(x))], \\ [(\underline{\eta}_{P_{F(e)}}(x) + \beta \underline{\pi}_{P_{F(e)}}(x)), (\overline{\eta}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x))], \\ [(\underline{\nu}_{P_{F(e)}}(x) + \beta \underline{\pi}_{P_{F(e)}}(x)), (\overline{\nu}_{P_{F(e)}}(x) + \beta \overline{\pi}_{P_{F(e)}}(x))], \\ x \in U, e \in A \}.$$

$$= \mathbb{F}_{\alpha,\beta}(P_{F},A). \qquad \Box$$

Definition 2.9. Let $\alpha, \beta \in [0, 1]$. Given an $IVPFSS(P_F, A)$, the operator $\mathbb{G}_{\alpha, \beta}$ is defined as

$$\begin{split} \mathbb{G}_{\alpha,\beta}(P_F,A) &= \{(x, [\alpha \underline{\mu}_{P_{F(e)}}(x), \alpha \overline{\mu}_{P_{F(e)}}(x)], [\beta \underline{\eta}_{P_{F(e)}}(x), \beta \overline{\eta}_{P_{F(e)}}(x)], \\ & [\beta \underline{\nu}_{P_{F(e)}}(x), \beta \overline{\nu}_{P_{F(e)}}(x)]); x \in U, e \in A\}. \\ Obviously, \ \mathbb{G}_{1,1}(P_F,A) &= (P_F,A) \text{ and } \mathbb{G}_{0,0}(P_F,A) = \phi. \end{split}$$

Theorem 2.3. For every $IVPFSS(P_F, A)$ and $\alpha, \beta, \gamma \in [0, 1]$,

(i) G_{α,β}(P_F, A) is an IVPFSS.
(ii) If α ≤ γ then G_{α,β}(P_F, A) ∈ G_{γ,β}(P_F, A),
(iii) If β ≤ γ then G_{α,β}(P_F, A) ∋ G_{α,γ}(P_F, A),
(iv) If δ ∈ [0, 1] then G_{α,β}(G_{γ,δ}(P_F, A)) = G_{αγ,βδ}(P_F, A) = G_{γ,δ}(G_{α,β}(P_F, A)),
(v) (G_{α,β}(P_F, A)^c)^c = G_{β,α}(P_F, A).

Proof.

(i) Clearly $\mathbb{G}_{\alpha,\beta}(P_F, A)$ is an IVPFSS.

$$\begin{aligned} \text{(ii)} \ & \mathbb{G}_{\alpha,\beta}(P_F, A) = \{ (x, [\alpha \underline{\mu}_{P_F(e)}(x), \alpha \overline{\mu}_{P_F(e)}(x)]; [\beta \underline{\eta}_{P_F(e)}(x), \beta \overline{\eta}_{P_F(e)}(x)]; x \in U, e \in A \}, \\ & [\beta \underline{\nu}_{P_F(e)}(x), \beta \overline{\nu}_{P_F(e)}(x)]; x \in U, e \in A \}, \\ & \mathbb{G}_{\gamma,\beta}(P_F, A) = \{ (x, [\gamma \underline{\mu}_{P_F(e)}(x), \beta \overline{\nu}_{P_F(e)}(x)]; x \in U, e \in A \}. \\ & \text{Since } \alpha \leq \gamma, \alpha \underline{\mu}_{P_F(e)}(x) \leq \gamma \underline{\mu}_{P_F(e)}(x) \text{ and } \alpha \overline{\mu}_{P_F(e)}(x) \leq \gamma \overline{\mu}_{P_F(e)}(x). \text{ Then } \mathbb{G}_{\alpha,\beta}(P_F, A) \in \\ & \mathbb{G}_{\gamma,\beta}(P_F, A). \\ \end{aligned}$$

$$\begin{aligned} \text{(iii)} \ & \mathbb{G}_{\alpha,\beta}(P_F, A) = \{ (x, [\alpha \underline{\mu}_{P_F(e)}(x), \alpha \overline{\mu}_{P_F(e)}(x)]; [\beta \underline{\eta}_{P_F(e)}(x), \beta \overline{\eta}_{P_F(e)}(x)], \\ & [\beta \underline{\nu}_{P_F(e)}(x), \alpha \overline{\mu}_{P_F(e)}(x)]; x \in U, e \in A \}, \\ & \mathbb{G}_{\alpha,\gamma}(P_F, A) = \{ (x, [\alpha \underline{\mu}_{P_F(e)}(x), \alpha \overline{\mu}_{P_F(e)}(x)]; x \in U, e \in A \}, \\ & \mathbb{G}_{\alpha,\gamma}(P_F, A) = \{ (x, [\alpha \underline{\mu}_{P_F(e)}(x), \alpha \overline{\mu}_{P_F(e)}(x)]; x \in U, e \in A \}, \\ & \text{Since } \beta \leq \gamma, \beta \underline{\eta}_{P_F(e)}(x) \leq \gamma \overline{\mu}_{P_F(e)}(x), \alpha \overline{\mu}_{P_F(e)}(x)]; x \in U, e \in A \}. \\ & \text{Since } \beta \leq \gamma, \beta \underline{\eta}_{P_F(e)}(x) \leq \gamma \overline{\eta}_{P_F(e)}(x), \beta \overline{\eta}_{P_F(e)}(x)]; x \in U, e \in A \}. \\ & \text{Since } \beta \leq \gamma, \beta \underline{\eta}_{P_F(e)}(x) \leq \gamma \overline{\mu}_{P_F(e)}(x), \beta \overline{\eta}_{P_F(e)}(x)], x \in U, e \in A \}. \\ & \text{Since } \beta \leq \gamma, \beta \underline{\eta}_{P_F(e)}(x) \leq \gamma \overline{\mu}_{P_F(e)}(x), \beta \overline{\eta}_{P_F(e)}(x)]; x \in U, e \in A \}. \\ & \text{Since } \beta \leq \gamma, \beta \underline{\eta}_{P_F(e)}(x) \leq \gamma \overline{\eta}_{P_F(e)}(x), \beta \overline{\eta}_{P_F(e)}(x)], x \in U, e \in A \}. \\ & \text{Since } \beta \leq \gamma, \beta \underline{\eta}_{P_F(e)}(x) \leq \gamma \overline{\mu}_{P_F(e)}(x), \beta \overline{\eta}_{P_F(e)}(x)], \\ & [\gamma \mathcal{U}_{P_F(e)}(x), \beta \overline{\eta}_{P_F(e)}(x), \beta \overline{\eta}_{P_F(e)}(x)], \\ & [\beta \delta \underline{\mu}_{P_F(e)}(x), \beta \delta \overline{\eta}_{P_F(e)}(x)], \\ & [(\delta \beta) \underline{\mu}_{P_F(e)}(x), (\beta \delta) \overline{\eta}_{P_F(e)}(x)], \\ & [(\delta \beta) \underline{\mu}_{P_F(e)}(x), (\delta \beta \overline{\eta}_{P_F(e)}(x)], \\ & [(\delta \beta) \underline{\mu}_{P_F(e)}(x), (\delta \beta) \overline{\eta}_{P_F$$

From the above, it follows that

$$\mathbb{G}_{\alpha,\beta}(\mathbb{G}_{\gamma,\delta}(P_F,A)) = \mathbb{G}_{\alpha\gamma,\beta\delta}(P_F,A) = \mathbb{G}_{\gamma,\delta}(\mathbb{G}_{\alpha,\beta}(P_F,A)).$$

3. NORMALIZED EUCLIDEAN DISTANCE BETWEEN IVPFSS

In this section we define normalized Euclidean distance between IVPFSS and establish that it is a metric.

Definition 3.1. Let $X = \{x_1, x_2, ..., x_i\}$ be an universal set, $E = \{e_1, e_2, ..., e_j\}$ be a set of parameters and $(P_F, A), (P_G, B)$ two IVPFSS on X. Then the normalized Euclidean distance between (P_F, A) and (P_G, B) is defined as

$$\mathcal{D}_{\epsilon} \langle (P_{F}, A), (P_{G}, B) \rangle = \left\{ \frac{1}{6mn} \sum_{m=1}^{i} \sum_{n=1}^{j} [|\underline{\mu}_{P_{F(e_{i})}}(x_{j}) - \underline{\mu}_{P_{G(e_{i})}}(x_{j})| + |\overline{\mu}_{P_{F(e_{i})}}(x_{j}) - \overline{\mu}_{P_{G(e_{i})}}(x_{j})| \\ + |\underline{\eta}_{P_{F(e_{i})}}(x_{j}) - \underline{\eta}_{P_{G(e_{i})}}(x_{j})| + |\overline{\eta}_{P_{F(e_{i})}}(x_{j}) - \overline{\eta}_{P_{G(e_{i})}}(x_{j})| \\ + |\underline{\nu}_{P_{F(e_{i})}}(x_{j}) - \underline{\nu}_{P_{G(e_{i})}}(x_{j})| + |\overline{\nu}_{P_{F(e_{i})}}(x_{j}) - \overline{\nu}_{P_{G(e_{i})}}(x_{j})| \\ + |\underline{\pi}_{P_{F(e_{i})}}(x_{j}) - \underline{\pi}_{P_{G(e_{i})}}(x_{j})| + |\overline{\pi}_{P_{F(e_{i})}}(x_{j}) - \overline{\pi}_{P_{G(e_{i})}}(x_{j})| \right\}.$$

Theorem 3.1. Let IVPFSS(U) be the set of all IVPFSS over U. Then the distance function \mathcal{D}_{ϵ} from IVPFSS(U) to the set of non-negative real numbers is a metric.

Proof. Let $(P_F, A), (P_G, B)$ and (P_H, C) be three IVPFSS(U) over U.

(i) $\mathcal{D}_{\epsilon}\langle (P_F, A), (P_G, B) \rangle > 0$ follows from Definition 3.1.

(ii)
$$\mathcal{D}_{\epsilon} \langle (P_F, A), (P_G, B) \rangle = 0$$

 $\Leftrightarrow (\underline{\mu}_{P_{F(e_i)}}(x_j) - \underline{\mu}_{P_{G(e_i)}}(x_j) + \overline{\mu}_{P_{F(e_i)}}(x_j) - \overline{\mu}_{P_{G(e_i)}}(x_j))$
 $+ (\underline{\eta}_{P_{F(e_i)}}(x_j) - \underline{\eta}_{P_{G(e_i)}}(x_j) + \overline{\eta}_{P_{F(e_i)}}(x_j) - \overline{\eta}_{P_{G(e_i)}}(x_j))$
 $+ (\underline{\nu}_{P_{F(e_i)}}(x_j) - \underline{\nu}_{P_{G(e_i)}}(x_j) + \overline{\nu}_{P_{F(e_i)}}(x_j) - \overline{\nu}_{P_{G(e_i)}}(x_j))$

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$$\begin{split} &+(\underline{\pi}_{P_{F(e_{i})}}(x_{j})-\underline{\pi}_{P_{G(e_{i})}}(x_{j})+\overline{\pi}_{P_{F(e_{i})}}(x_{j})-\overline{\pi}_{P_{G(e_{i})}}(x_{j}))=0\\ \Leftrightarrow & \underline{\mu}_{P_{F(e_{i})}}(x_{j})=\underline{\mu}_{P_{G(e_{i})}}(x_{j}), \overline{\mu}_{P_{F(e_{i})}}(x_{j})=\overline{\mu}_{P_{G(e_{i})}}(x_{j}),\\ & \underline{\eta}_{P_{F(e_{i})}}(x_{j})=\underline{\eta}_{P_{G(e_{i})}}(x_{j}), \overline{\eta}_{P_{F(e_{i})}}(x_{j})=\overline{\eta}_{P_{G(e_{i})}}(x_{j}),\\ & \underline{\nu}_{P_{F(e_{i})}}(x_{j})=\underline{\nu}_{P_{G(e_{i})}}(x_{j}), \overline{\nu}_{P_{F(e_{i})}}(x_{j})=\overline{\nu}_{P_{G(e_{i})}}(x_{j}),\\ & \underline{\pi}_{P_{F(e_{i})}}(x_{j})=\underline{\pi}_{P_{G(e_{i})}}(x_{j}) \text{ and } \overline{\pi}_{P_{F(e_{i})}}(x_{j})=\overline{\pi}_{P_{G(e_{i})}}(x_{j})\\ \Leftrightarrow (P_{F},A)=(P_{G},B). \end{split}$$

(iii) Clearly, $\mathcal{D}_{\epsilon}\langle (P_F, A), (P_G, B) \rangle = \mathcal{D}_{\epsilon}\langle (P_G, B), (P_F, A) \rangle$.

(iv) Assume that $(P_F, A), (P_G, B)$ and (P_H, C) are IVPFSS over U. Then for all $i \in \{1, 2, \dots, m\}, j \in \{1, 2, \dots, n\}, j \in \{1, 2, \dots, n$

$$\begin{split} & (\underline{\mu}_{P_{F(e_i)}}(x_j) - \underline{\mu}_{P_{G(e_i)}}(x_j) + \overline{\mu}_{P_{F(e_i)}}(x_j) - \overline{\mu}_{P_{G(e_i)}}(x_j)) \\ & + (\underline{\eta}_{P_{F(e_i)}}(x_j) - \underline{\eta}_{P_{G(e_i)}}(x_j) + \overline{\eta}_{P_{F(e_i)}}(x_j) - \overline{\eta}_{P_{G(e_i)}}(x_j)) \\ & + (\underline{\nu}_{P_{F(e_i)}}(x_j) - \underline{\nu}_{P_{G(e_i)}}(x_j) + \overline{\pi}_{P_{F(e_i)}}(x_j) - \overline{n}_{P_{G(e_i)}}(x_j)) \\ & + (\underline{\pi}_{P_{F(e_i)}}(x_j) - \underline{\pi}_{P_{G(e_i)}}(x_j) + \overline{\mu}_{P_{H(e_i)}}(x_j) - \underline{\mu}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\mu}_{P_{F(e_i)}}(x_j) - \underline{\mu}_{P_{H(e_i)}}(x_j) + \underline{\mu}_{P_{H(e_i)}}(x_j) - \underline{\mu}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\mu}_{P_{F(e_i)}}(x_j) - \underline{\eta}_{P_{H(e_i)}}(x_j) + \underline{\eta}_{P_{H(e_i)}}(x_j) - \overline{\eta}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\eta}_{P_{F(e_i)}}(x_j) - \overline{\eta}_{P_{H(e_i)}}(x_j) + \overline{\eta}_{P_{H(e_i)}}(x_j) - \overline{\eta}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\eta}_{P_{F(e_i)}}(x_j) - \overline{\eta}_{P_{H(e_i)}}(x_j) + \overline{\eta}_{P_{H(e_i)}}(x_j) - \overline{\eta}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\mu}_{P_{F(e_i)}}(x_j) - \overline{\mu}_{P_{H(e_i)}}(x_j) + \overline{\mu}_{P_{H(e_i)}}(x_j) - \overline{\mu}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\mu}_{P_{F(e_i)}}(x_j) - \overline{\pi}_{P_{H(e_i)}}(x_j) + \overline{\mu}_{P_{H(e_i)}}(x_j) - \overline{\pi}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\pi}_{P_{F(e_i)}}(x_j) - \overline{\pi}_{P_{H(e_i)}}(x_j)) + (\overline{\mu}_{P_{H(e_i)}}(x_j) - \overline{\mu}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\mu}_{P_{F(e_i)}}(x_j) - \overline{\mu}_{P_{H(e_i)}}(x_j)) + (\overline{\mu}_{P_{H(e_i)}}(x_j) - \overline{\mu}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\mu}_{P_{F(e_i)}}(x_j) - \overline{\eta}_{P_{H(e_i)}}(x_j)) + (\overline{\mu}_{P_{H(e_i)}}(x_j) - \overline{\eta}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\mu}_{P_{F(e_i)}}(x_j) - \overline{\eta}_{P_{H(e_i)}}(x_j)) + (\overline{\eta}_{P_{H(e_i)}}(x_j) - \overline{\eta}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\mu}_{P_{F(e_i)}}(x_j) - \overline{\eta}_{P_{H(e_i)}}(x_j)) + (\overline{\mu}_{P_{H(e_i)}}(x_j) - \overline{\mu}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\mu}_{P_{F(e_i)}}(x_j) - \overline{\eta}_{P_{H(e_i)}}(x_j)) + (\overline{\mu}_{P_{H(e_i)}}(x_j) - \overline{\mu}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\mu}_{P_{F(e_i)}}(x_j) - \overline{\eta}_{P_{H(e_i)}}(x_j)) + (\overline{\mu}_{P_{H(e_i)}}(x_j) - \overline{\mu}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\mu}_{P_{F(e_i)}}(x_j) - \overline{\eta}_{P_{H(e_i)}}(x_j)) + (\overline{\mu}_{P_{H(e_i)}}(x_j) - \overline{\mu}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\mu}_{P_{F(e_i)}}(x_j) - \overline{\eta}_{P_{H(e_i)}}(x_j)) + (\overline{\mu}_{P_{H(e_i)}}(x_j) - \overline{\mu}_{P_{G(e_i)}}(x_j)) \\ & + (\overline{\mu}_{P$$

Thus, $\mathcal{D}_{\epsilon}\langle (P_F, A), (P_G, B) \rangle \leq \mathcal{D}_{\epsilon}\langle (P_F, A), (P_H, C) \rangle + \mathcal{D}_{\epsilon}\langle (P_H, C), (P_G, B) \rangle$. This shows that \mathcal{D}_{ϵ} satisfies the triangle inequality. So \mathcal{D}_{ϵ} is a metric.

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4. PATTERN RECOGNITION PROBLEM

- **Step 1:** Construct an $IVPFSS(P_S, E)$, over U based on expert evaluation and this can be considered as the known pattern.
- **Step 2:** Construct an $IVPFSS(P_{F_i}, E), \forall i = 1, 2, 3, ...$ over U based on the data available for the unknown pattern.
- **Step 3:** Calculate the normalized Euclidean distance between (P_S, E) and (P_{F_i}, E) .
- **Step 4:** The pattern with less normalized Euclidean distance between (P_S, E) and (P_{F_i}, E) the pattern in the best suitable pattern.

Example 1. Let three picture fuzzy soft sets $(P_{F_1}, E), (P_{F_2}, E)$ and (P_{F_3}, E) denote microwave oven of three different companies depending on the parameters $E = \{e_1, e_2, e_3, e_4, e_5\}$ where e_1 = range of price e_2 = solo, grill or convection, e_3 = ordinary or inverter, e_4 = electromagnetic radiation and e_5 = dielectric material which are rated by four customers C_1 , C_2 , C_3 and C_4 on the five parameters. Depending on the rating of the customers the best microwave oven is to be selected depending upon the above parameters.

U	C_1	C_2
e_1	[0.21,0.33],[0.45,0.52],[0.08,0.13]	[0.09,0.23],[0.29,0.31],[0.18,0.26]
e_2	[0.15,0.26],[0.35,0.43],[0.25,0.31]	[0.17,0.30],[0.12,0.29],[0.26,0.35]
e_3	[0.30,0.41],[0.23,0.35],[0.17,0.22]	[0.28,0.42],[0.19,0.36],[0.08,0.13]
e_4	[0.11,0.28],[0.20,0.37],[0.22,0.34]	[0.06,0.15],[0.25,0.45],[0.22,0.38]
e_5	[0.03,0.16],[0.32,0.44],[0.27,0.38]	[0.14,0.21],[0.16,0.50],[0.10,0.27]
U	C_3	C_4
e_1	[0.18,0.25],[0.07,0.36],[0.29,0.39]	[0.12,0.24],[0.09,0.42],[0.25,0.31]
e_2	[0.02,0.11],[0.23,0.49],[0.14,0.28]	[0.10,0.32],[0.18,0.22],[0.11,0.45]
e_3	[0.20,0.31],[0.13,0.27],[0.35,0.41]	[0.21,0.41],[0.03,0.30],[0.07,0.28]
e_4	[0.08,0.25],[0.21,0.30],[0.22,0.45]	[0.06,0.16],[0.29,0.49],[0.14,0.33]
e_5	[0.26,0.34],[0.16,0.42],[0.10,0.19]	[0.13,0.20],[0.08,0.52],[0.17,0.23]

Step 1. The construction of known pattern is as follows (Table 1).

Table 1. $IVPFSS(P_S, E)$ over U is the data from the previous records of the customer for the best microwave oven.

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U	C_1	C_2	
e_1	[0.05,0.26],[0.13,0.33],[0.22,0.40]	[0.23,0.35],[0.07,0.28],[0.16,0.31]	
e_2	[0.18,0.21],[0.08,0.29],[0.31,0.48]	[0.11,0.27],[0.18,0.32],[0.05,0.22]	
e_3	[0.37,0.43],[0.17,0.25],[0.03,0.32]	[0.09,0.20],[0.38,0.43],[0.10,0.30]	
e_4	[0.15,0.23],[0.29,0.35],[0.19,0.40]	[0.14,0.33],[0.26,0.53],[0.02,0.13]	
e_5	[0.07,0.34],[0.14,0.42],[0.06,0.20]	[0.08,0.25],[0.12,0.40],[0.29,0.34]	
U	C_3	C_4	
e_1	[0.08,0.21],[0.13,0.39],[0.23,0.40]	[0.15,0.32],[0.24,0.48],[0.07,0.19]	
e_2	[0.17,0.30],[0.05,0.27],[0.30,0.43]	[0.20,0.28],[0.11,0.30],[0.38,0.42]	
e_3	[0.24,0.38],[0.19,0.48],[0.03,0.14]	[0.08,0.14],[0.25,0.51],[0.21,0.34]	
e_4	[0.06,0.26],[0.31,0.41],[0.28,0.32]	[0.35,0.45],[0.18,0.39],[0.02,0.10]	
e_5	[0.15,0.46],[0.02,0.12],[0.35,0.42]	[0.12,0.53],[0.05,0.16],[0.23,0.31]	

Step 2. The construction of unknown pattern are given in Tables 2,3 and 4.

Table 2. $IVPFSS(P_{F_1}, E)$ over U gives the data rated by the customers for the microwave oven-1.

U	C_1	C_2
e_1	[0.11,0.36],[0.22,0.44],[0.03,0.17]	[0.27,0.38],[0.09,0.27],[0.16,0.33]
e_2	[0.25,0.37],[0.08,0.15],[0.20,0.38]	[0.12,0.35],[0.30,0.43],[0.05,0.21]
e_3	[0.05,0.40],[0.28,0.34],[0.12,0.25]	[0.08,0.17],[0.25,0.32],[0.47,0.50]
e_4	[0.18,0.32],[0.43,0.50],[0.06,0.14]	[0.24,0.40],[0.06,0.20],[0.13,0.36]
e_5	[0.09,0.21],[0.13,0.46],[0.26,0.33]	[0.19,0.29],[0.31,0.45],[0.03,0.18]
U	C_3	C_4
$\begin{array}{ c c } U \\ e_1 \end{array}$	C ₃ [0.22,0.35],[0.15,0.27],[0.08,0.32]	C ₄ [0.07,0.22],[0.28,0.37],[0.12,0.41]
$\begin{array}{c} U\\ e_1\\ e_2 \end{array}$	C3 [0.22,0.35],[0.15,0.27],[0.08,0.32] [0.15,0.29],[0.33,0.42],[0.19,0.26]	C4 [0.07,0.22],[0.28,0.37],[0.12,0.41] [0.15,0.30],[0.09,0.20],[0.27,0.35]
$\begin{array}{c} U\\ e_1\\ e_2\\ e_3 \end{array}$	C3 [0.22,0.35],[0.15,0.27],[0.08,0.32] [0.15,0.29],[0.33,0.42],[0.19,0.26] [0.09,0.36],[0.20,0.51],[0.05,0.12]	C4 [0.07,0.22],[0.28,0.37],[0.12,0.41] [0.15,0.30],[0.09,0.20],[0.27,0.35] [0.23,0.36],[0.18,0.45],[0.05,0.16]
$\begin{array}{c} U\\ e_1\\ e_2\\ e_3\\ e_4 \end{array}$	$\begin{array}{c} C_{3} \\ \hline [0.22, 0.35], [0.15, 0.27], [0.08, 0.32] \\ [0.15, 0.29], [0.33, 0.42], [0.19, 0.26] \\ [0.09, 0.36], [0.20, 0.51], [0.05, 0.12] \\ [0.11, 0.45], [0.02, 0.18], [0.25, 0.30] \end{array}$	$\begin{array}{c} C_4 \\ \hline [0.07, 0.22], [0.28, 0.37], [0.12, 0.41] \\ \hline [0.15, 0.30], [0.09, 0.20], [0.27, 0.35] \\ \hline [0.23, 0.36], [0.18, 0.45], [0.05, 0.16] \\ \hline [0.12, 0.29], [0.31, 0.50], [0.11, 0.21] \end{array}$

Table 3. $IVPFSS(P_{F_2}, E)$ over U gives the data rated by the customers for the microwave oven-2.

U	C_1	C_2
e_1	[0.09,0.28],[0.15,0.30],[0.22,0.41]	[0.15,0.21],[0.35,0.44],[0.07,0.25]
e_2	[0.11,0.32],[0.05,0.42],[0.12,0.25]	[0.09,0.19],[0.28,0.39],[0.13,0.20]
e_3	[0.21,0.34],[0.18,0.27],[0.08,0.36]	[0.26,0.30],[0.14,0.32],[0.03,0.34]
e_4	[0.07,0.29],[0.33,0.51],[0.10,0.20]	[0.17,0.33],[0.05,0.42],[0.18,0.22]
e_5	[0.14,0.21],[0.03,0.19],[0.37,0.48]	[0.06,0.24],[0.29,0.31],[0.16,0.43]

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U	C_3	C_4
e_1	[0.19,0.23],[0.32,0.41],[0.09,0.36]	[0.14,0.25],[0.08,0.22],[0.38,0.42]
e_2	[0.06,0.12],[0.21,0.39],[0.16,0.47]	[0.18,0.29],[0.16,0.30],[0.28,0.40]
e_3	[0.28,0.35],[0.03,0.18],[0.29,0.43]	[0.09,0.19],[0.23,0.44],[0.15,0.37]
e_4	[0.15,0.38],[0.24,0.31],[0.05,0.20]	[0.13,0.33],[0.05,0.17],[0.26,0.43]
e_5	[0.04,0.49],[0.17,0.29],[0.13,0.22]	[0.10,0.21],[0.48,0.54],[0.03,0.20]

Table 4. $IVPFSS(P_{F_3}, E)$ over U gives the data rated by the customers for the microwave oven-3.

Step 3. The normalized Euclidean distance between $\mathcal{D}_{\epsilon}\langle (P_S, E), (P_{F_i}, E) \rangle$ iS calculated using Definition 3.1. The values evaluated are as follows:

 $\mathcal{D}_{\epsilon} \langle (P_S, E), (P_{F_1}, E) \rangle = 0.1523,$ $\mathcal{D}_{\epsilon} \langle (P_S, E), (P_{F_2}, E) \rangle = 0.1465,$ $\mathcal{D}_{\epsilon} \langle (P_S, E), (P_{F_3}, E) \rangle = 0.1375.$

Step 4. We observe that $\mathcal{D}_{\epsilon}\langle (P_S, E), (P_{F_3}, E) \rangle$ is the least distance. Hence microwave oven-3 is the best.

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