



INFORMATION MEASURE ON BIPOLAR INTUITIONISTIC FUZZY SOFT SETS

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Abstract

The aim of this paper is to introduce operators on bipolar intuitionistic fuzzy soft sets (BIFSS). Properties of operators on BIFSS are established. An information measure on BIFSS is proposed. Making use of the proposed information measure a decision making concept is constructed. Finally, this concept is illustrated by an example.

1. Introduction

Zadeh [6] introduced fuzzy sets and Atanassov [2] introduced the IFS. Chiranjibe et.al., [3] developed the notion of BIFS set. Information measure in fuzzy sets is certainly a measure of fuzziness, while for intuitionistic fuzzy sets information measure measures both the fuzziness and intuitionism. Srinivasan et.al., [5] introduced some operations on intuitionistic fuzzy sets of root type. Anita shanthi et.al., [1] introduced information measure of IVIFSSRT. Mishra [4] developed the intuitionistic fuzzy information measure.

Now, we define operators on BIFSS and study some of its properties. We propose the concept of an information measure of BIFSS. Further decision making concept on information measure of BIFSS is developed.

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2. Operations on BIFSS

In this section, we define operators on BIFSS. Some properties of these operators are discussed.

Definition 2.1. Let

$$(BF, E) = \{x, ((\mu_{BF(e)}^n(x), \mu_{BF(e)}^p(x)), (v_{BF(e)}^n(x), v_{BF(e)}^p(x)) : x \in U, e \in E\}$$

be BIFSS. The hesitant degree of BIFSS is denoted by $\pi_{BF(e)}^n$ and $\pi_{BF(e)}^p$ is defined as

$$\pi_{BF(e)}^n(x) = -1 - \mu_{BF(e)}^n(x) - v_{BF(e)}^n(x) \text{ and } \pi_{BF(e)}^p(x) = 1 - \mu_{BF(e)}^p(x) - v_{BF(e)}^p(x)$$

where $\pi_{BF(e)}^n \in [-1, 0]$ and $\pi_{BF(e)}^p \in [1, 0]$ respectively.

Definition 2.2. Let $\eta \in [0, 1]$ be a fixed number. Given an BIFSS (F, E) the operator D_η is defined as

$$\begin{aligned} D_\eta(BF, E) = \{x, ((\mu_{BF(e)}^n(x) + \eta\pi_{BF(e)}^n(x)), ((\mu_{BF(e)}^p(x) + \eta\pi_{BF(e)}^p(x)), \\ ((v_{BF(e)}^n(x) + (1 - \eta)\pi_{BF(e)}^n(x)), ((v_{BF(e)}^p(x) + (1 - \eta)\pi_{BF(e)}^p(x)) \\ : x \in U, e \in E\}. \end{aligned}$$

Definition 2.3. For $\eta, \delta \in [0, 1]$, $\eta + \delta \leq 1$ the operator $F_{\eta, \delta}$ for a BIFSS (BF, E) is defined as

$$\begin{aligned} F_{\eta, \delta}(BF, E) = \{x, ((\mu_{BF(e)}^n(x) + \eta\pi_{BF(e)}^n(x)), ((\mu_{BF(e)}^p(x) + \eta\pi_{BF(e)}^p(x)), \\ ((v_{BF(e)}^n(x) + \delta\pi_{BF(e)}^n(x)), ((v_{BF(e)}^p(x) \\ + \delta\pi_{BF(e)}^p(x)) : x \in U, e \in E\}. \end{aligned}$$

Definition 2.4. Let $\eta, \delta \in [0, 1]$. Given a BIFSS (BF, E) , the operator $G_{\eta, \delta}$ is defined as

$$G_{\eta, \delta}(BF, E) = \{x, ((\eta\mu_{BF(e)}^n(x), \eta\mu_{BF(e)}^n(x)), ((\delta v_{BF(e)}^n(x), \delta v_{BF(e)}^n(x)) : x \in U, e \in E\}.$$

Obviously $G_{1,1}(BF, E) = (BF, E)$ and $G_{0,0}(BF, E) = \Phi$.

Definition 2.5. (BF, E) and (BG, E) are two BIFSS. The operation @ on (BF, E) and (BG, E) is defined as

$$(BF, E) @ (BG, E) = \left\{ x, \left(\frac{\mu_{BF(e)}^n(x) + \mu_{BG(e)}^n(x)}{2}, \frac{\mu_{BF(e)}^p(x) + \mu_{BG(e)}^p(x)}{2} \right), \left(\frac{v_{BF(e)}^n(x) + v_{BG(e)}^n(x)}{2}, \frac{v_{BF(e)}^p(x) + v_{BG(e)}^p(x)}{2} \right) \right\}_{x \in U, e \in E}.$$

Theorem 2.6. Let (BF, E) and (BG, E) are two BIFSS and $\eta, \delta \in [0, 1]$, then the following conditions holds:

- (i) $D_{\eta}((BF, E) @ (BG, E)) = D_{\eta}(BF, E) @ D_{\eta}(BG, E)$
- (ii) $F_{\eta, \delta}((BF, E) @ (BG, E)) = F_{\eta, \delta}(BF, E) @ F_{\eta, \delta}(BG, E)$
- (ii) $G_{\eta, \delta}((BF, E) @ (BG, E)) = G_{\eta, \delta}(BF, E) @ G_{\eta, \delta}(BG, E)$.

Proof.

$$(i) \quad (BF, E) @ (BG, E) = \left\{ x, \left(\frac{1}{2}, (\mu_{BF(e)}^n(x) + \mu_{BG(e)}^n(x), \mu_{BF(e)}^p(x) + \mu_{BG(e)}^p(x)) \right), \left(\frac{1}{2}, (v_{BF(e)}^n(x) + v_{BG(e)}^n(x), v_{BF(e)}^p(x) + v_{BG(e)}^p(x)) \right) \right\}$$

$$D_{\eta}((BF, E) @ (BG, E)) = \left\{ x, \left(\frac{1}{2}(\mu_{BF(e)}^n(x) + \eta\pi_{BF(e)}^n(x) + \mu_{BG(e)}^n(x) + \eta\pi_{BG(e)}^n(x), \mu_{BF(e)}^p(x) + \eta\pi_{BF(e)}^p(x) + \mu_{BG(e)}^p(x) + \eta\pi_{BG(e)}^p(x)) \right) \right\}$$

$$\left(\frac{1}{2}(v_{BF(e)}^n(x) + (1 - \eta)\pi_{BF(e)}^n(x) + v_{BG(e)}^n(x) + (1 - \eta)\pi_{BG(e)}^n(x), v_{BF(e)}^p(x) \right)$$

$$\begin{aligned}
 F_{\eta, \delta}(BF, E) @ F_{\eta, \delta}(BG, E) &= \{x, (\frac{1}{2}(\mu_{BF(e)}^n(x) + \eta\pi_{BF(e)}^n(x) + \mu_{BG(e)}^n(x) \\
 &+ \eta\pi_{BG(e)}^n(x)), \mu_{BF(e)}^p(x) + \eta\pi_{BF(e)}^p(x) + \mu_{BG(e)}^p(x) + \eta\pi_{BG(e)}^p(x)), \\
 &(\frac{1}{2}(v_{BF(e)}^n(x) + \delta\pi_{BF(e)}^n(x) + v_{BG(e)}^n(x) + \delta\pi_{BG(e)}^n(x)), \\
 &v_{BF(e)}^p(x) + \delta\pi_{BF(e)}^p(x) + v_{BG(e)}^p(x) + \delta\pi_{BG(e)}^p(x))\}.
 \end{aligned}$$

Therefore $F_{\gamma, \delta}((BF, E) @ (BG, E)) = F_{\gamma, \delta}(BF, E) @ F_{\gamma, \delta}(BG, E)$

(iii) $G_{\eta, \delta}((BF, E) @ (BG, E))$

$$= \{x, \frac{1}{2}(\eta\mu_{BF(e)}^n(x) + \eta\mu_{BG(e)}^n(x) + \eta\mu_{BF(e)}^p(x) + \eta\pi_{BG(e)}^p(x)),$$

$$(\frac{1}{2}(\delta v_{BF(e)}^n(x) + \delta v_{BG(e)}^n(x) + \delta v_{BF(e)}^p(x) + \delta v_{BG(e)}^p(x))\}$$

$$G_{\eta, \delta}(BF, E) = \{x, (\eta\mu_{BF(e)}^p(x) + \eta\pi_{BF(e)}^p(x)), (\delta\mu_{BF(e)}^p(x) + \delta v_{BF(e)}^p(x))\}$$

$$G_{\eta, \delta}(BG, E) = \{x, (\eta\mu_{BG(e)}^p(x) + \eta\pi_{BG(e)}^p(x)), (\delta\mu_{BG(e)}^p(x) + \delta v_{BG(e)}^p(x))\}$$

$$G_{\eta, \delta}(BF, E) @ G_{\eta, \delta}(BG, E)$$

$$= \{x, \frac{1}{2}(\eta\mu_{BF(e)}^n(x) + \eta\mu_{BG(e)}^n(x), \eta\mu_{BF(e)}^p(x) + \eta\pi_{BG(e)}^p(x)),$$

$$(\frac{1}{2}(\delta v_{BF(e)}^n(x) + \delta v_{BG(e)}^n(x) + \delta v_{BF(e)}^p(x) + \delta v_{BG(e)}^p(x))\}.$$

Therefore $G_{\eta, \delta}((BF, E) @ (BG, E)) = G_{\eta, \delta}(BF, E) @ G_{\eta, \delta}(BG, E)$.

3. Information measure on BIFSS

In this section, we define information measure on BIFSS and prove that it is an entropy.

Definition 3.1. Let the universal set be $U = \{A_1, A_2, \dots, A_m\}$ and the set of parameters be $E = \{e_1, e_2, \dots, e_n\}$. For any BIFSS (BF, E) an

information measure to indicate the degree of fuzziness of (BF, E) is defined as

$$BI_m(BF, E_p) = \frac{1}{n} \sum_{p=1}^n \frac{\wedge \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \} - \wedge \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \}}{\vee \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \} - \vee \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \}}$$

where, \wedge, \vee denotes the minimum and maximum, respectively.

Theorem 3.2. *The information measure $BI_m(BF, E)$ for BIFSS is an entropy.*

Proof. If $(BF, E) \in P(U)$, then for each $x_p \in A_p$ and $e \in E$ we have:

Case (i). If $(\mu_{BF(e)}^n(x_p), \mu_{BF(e)}^p(x_p)) = (\nu_{BF(e)}^n(x_p), \nu_{BF(e)}^p(x_p))$, then

$$\begin{aligned} \mu_{BF(e)}^n(x_p) &= \nu_{BF(e)}^n(x_p), \mu_{BF(e)}^p(x_p) = \nu_{BF(e)}^p(x_p) \\ \wedge \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \} &- \wedge \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \} \\ &= \vee \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \} - \vee \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \}. \end{aligned}$$

So, $BI_m(BF, E) = 1$.

Conversely, suppose that $BI_m(BF, E) = 1$.

This implies that

$$\wedge \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \} = \wedge \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \}$$

and

$$\wedge \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \} - \wedge \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \}.$$

Hence $\mu_{F(e)}^n(x_p) = \nu_{F(e)}^n(x_p)$ and $\mu_{F(e)}^p(x_p) = \nu_{F(e)}^p(x_p)$. Therefore, we have

$$(\mu_{BF(e)}^n(x_p), \mu_{BF(e)}^p(x_p)) = (\nu_{BF(e)}^n(x_p), \nu_{BF(e)}^p(x_p)).$$

Case (ii). Suppose that (BF, E) is less than (BG, E) then,

$$(\mu_{BF(e)}^n(x_p), \mu_{BF(e)}^p(x_p)) \leq (\mu_{BG(e)}^n(x_p), \mu_{BG(e)}^p(x_p)),$$

$$(\nu_{BF(e)}^n(x_p), \nu_{BF(e)}^p(x_p)) \geq (\nu_{BG(e)}^n(x_p), \nu_{BG(e)}^p(x_p)).$$

For $(\mu_{BG(e)}^n(x_p), \mu_{BG(e)}^p(x_p)) \leq (\nu_{BG(e)}^n(x_p), \nu_{BG(e)}^p(x_p)),$

$$\mu_{BF(e)}^n(x_p) \leq \mu_{BG(e)}^n(x_p) \leq \nu_{BG(e)}^n(x_p) \leq \nu_{BF(e)}^n(x_p)$$

and

$$\mu_{BF(e)}^p(x_p) \leq \mu_{BG(e)}^p(x_p) \leq \nu_{BG(e)}^p(x_p) \leq \nu_{BF(e)}^p(x_p)$$

So,

$$\begin{aligned} & \frac{\wedge \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \} - \wedge \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \}}{\vee \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \} - \vee \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \}} \\ &= \frac{\mu_{BF(e)}^n(x_p) + \mu_{BF(e)}^p(x_p)}{\nu \mu_{BF(e)}^n(x_p) - \nu \mu_{BF(e)}^p(x_p)} \leq \frac{\mu_{BG(e)}^n(x_p) + \mu_{BG(e)}^p(x_p)}{\nu \mu_{BG(e)}^n(x_p) - \nu \mu_{BG(e)}^p(x_p)} \\ &= \frac{\wedge \{ \mu_{BG(e)}^n(x_p), \nu_{BG(e)}^n(x_p) \} - \wedge \{ \mu_{BG(e)}^p(x_p), \nu_{BG(e)}^p(x_p) \}}{\vee \{ \mu_{BG(e)}^n(x_p), \nu_{BG(e)}^n(x_p) \} - \vee \{ \mu_{BG(e)}^p(x_p), \nu_{BG(e)}^p(x_p) \}}. \end{aligned}$$

By taking summation on both sides and dividing it by n , we get

$$\begin{aligned} & \frac{1}{n} \sum_{p=1}^n \frac{\wedge \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \} - \wedge \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \}}{\vee \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \} - \vee \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \}} \\ & \leq \frac{1}{n} \sum_{p=1}^n \frac{\wedge \{ \mu_{BG(e)}^n(x_p), \nu_{BG(e)}^n(x_p) \} - \wedge \{ \mu_{BG(e)}^p(x_p), \nu_{BG(e)}^p(x_p) \}}{\vee \{ \mu_{BG(e)}^n(x_p), \nu_{BG(e)}^n(x_p) \} - \vee \{ \mu_{BG(e)}^p(x_p), \nu_{BG(e)}^p(x_p) \}} \end{aligned}$$

i.e., $BI_m(BF, E) \leq BI_m(BG, E).$

Case (iii). Suppose that (BF, E) is greater than (BG, E) then,

$$(\mu_{BF(e)}^n(x_p), \mu_{BF(e)}^p(x_p)) \geq (\mu_{BG(e)}^n(x_p), \mu_{BG(e)}^p(x_p)),$$

$$(\nu_{BF(e)}^n(x_p), \nu_{BF(e)}^p(x_p)) \geq (\nu_{BG(e)}^n(x_p), \nu_{BG(e)}^p(x_p)).$$

For $(\mu_{BF(e)}^n(x_p), \mu_{BF(e)}^p(x_p)) \geq (\nu_{BG(e)}^n(x_p), \nu_{BG(e)}^p(x_p))$,

$$\mu_{BF(e)}^n(x_p) \geq \mu_{BF(e)}^n(x_p) \geq \nu_{BG(e)}^n(x_p), \nu_{BG(e)}^n(x_p)$$

and

$$\mu_{BF(e)}^p(x_p) \geq \mu_{BG(e)}^p(x_p) \geq \nu_{BG(e)}^p(x_p) \geq \nu_{BF(e)}^p(x_p).$$

So,

$$\begin{aligned} & \frac{\wedge \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \} - \wedge \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \}}{\vee \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \} - \vee \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \}}, \\ &= \frac{\mu_{BF(e)}^n(x_p) + \mu_{BF(e)}^p(x_p)}{\nu_{BF(e)}^n(x_p) - \nu_{BF(e)}^p(x_p)} \geq \frac{\mu_{BG(e)}^n(x_p) - \mu_{BG(e)}^p(x_p)}{\nu_{BG(e)}^n(x_p) - \nu_{BG(e)}^p(x_p)}, \\ &= \frac{\wedge \{ \mu_{BG(e)}^n(x_p), \nu_{BG(e)}^n(x_p) \} - \wedge \{ \mu_{BG(e)}^p(x_p), \nu_{BG(e)}^p(x_p) \}}{\vee \{ \mu_{BG(e)}^n(x_p), \nu_{BG(e)}^n(x_p) \} - \vee \{ \mu_{BG(e)}^p(x_p), \nu_{BG(e)}^p(x_p) \}}. \end{aligned}$$

By taking summation on both sides and dividing it by n , we get

$$\begin{aligned} & \frac{1}{n} \sum_{p=1}^n \frac{\wedge \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \} - \wedge \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \}}{\vee \{ \mu_{BF(e)}^n(x_p), \nu_{BF(e)}^n(x_p) \} - \vee \{ \mu_{BF(e)}^p(x_p), \nu_{BF(e)}^p(x_p) \}} \\ & \leq \frac{1}{n} \sum_{p=1}^n \frac{\wedge \{ \mu_{BG(e)}^n(x_p), \nu_{BG(e)}^n(x_p) \} - \wedge \{ \mu_{BG(e)}^p(x_p), \nu_{BG(e)}^p(x_p) \}}{\vee \{ \mu_{BG(e)}^n(x_p), \nu_{BG(e)}^n(x_p) \} - \vee \{ \mu_{BG(e)}^p(x_p), \nu_{BG(e)}^p(x_p) \}} \end{aligned}$$

i.e., $BI_m(BF, E) \geq BI_m(BG, E)$.

Case (iv).

$$(BF, E)^c = \{x, (\nu_{BF(e)}^n(x_p), \nu_{BF(e)}^p(x_p))$$

$$(\mu_{BF(e)}^n(x_p), \mu_{BF(e)}^p(x_p)) : x_p \in A_p, e \in E\}.$$

Therefore,

$$\begin{aligned}
 BI_m(BF, E)^C &= \frac{1}{n} \sum_{p=1}^n \frac{\wedge \{v_{BF(e)}^n(x_p), \mu_{BF(e)}^n(x_p)\} - \wedge \{v_{BF(e)}^p(x_p), \mu_{BF(e)}^p(x_p)\}}{\vee \{v_{BF(e)}^n(x_p), \mu_{BF(e)}^n(x_p)\} - \vee \{v_{BF(e)}^p(x_p), \mu_{BF(e)}^p(x_p)\}} \\
 &= BI_m(BF, E).
 \end{aligned}$$

Hence an information measure is an entropy.

4. Information measure on BIFSS

In this section, we develop a decision making concept based on information measure of BIFSS. A procedure for decision making method is framed.

4.1. Procedure:

The following are the steps to be followed for this method:

Step 1: Construct the BIFS sets (BF, E_q) over U .

Step 2. Determine the information measure of (BF, E_q) by using

Definition 3.1.

Step 3. Compare the values of $BI_m(BF, E_q)$ and conclude. Thus the better choice is the alternative for which the information measure is the least.

Example 4.2. A customer decides to buy a television. Television (alternatives) of three different companies $(BF, E_1), (BF, E_2), (BF, E_3)$ are evaluated over six factors $\{t_1, t_2, t_3, t_4, t_5, t_6\}$, where $t_1 =$ price range, $t_2 =$ LED or OLED, $t_3 =$ cables and accessories, $t_4 =$ audio upgrades, $t_5 =$ screen size and $t_6 =$ smart television. Depending on the six factors, the best alternative is determined.

Step 1. BIFSS over U to assess the alternatives is as follows:

U	(BF, E_1)	(BF, E_2)	(BF, E_3)
t_1	$((-0.13, 0.47), (-0.25, 0.1))$	$((-0.2, 0.68), (-0.32, 0.15))$	$((0.18, 0.42), (0.28, 0.12))$
t_2	$((-0.34, 0.62), (-0.16, 0.28))$	$((-0.18, 0.83), (-0.61, 0.1))$	$((0.17, 0.22), (0.18, 0.06))$
t_3	$((-0.25, 0.73), (-0.4, 0.14))$	$((-0.32, 0.45), (-0.27, 0.13))$	$((-0.14, 0.73), (0.3, 0.12))$
t_4	$((-0.1, 0.51), (-0.38, 0.08))$	$((-0.41, 0.64), (-0.5, 0.3))$	$((-0.24, 0.6), (0.04, 0.11))$
t_5	$((-0.12, 0.38), (-0.26, 0.04))$	$((-0.24, 0.76), (-0.61, 0.15))$	$((0.31, 0.52), (0.17, 0.28))$
t_6	$((-0.27, 0.47), (-0.3, 0.13))$	$((-0.3, 0.54), (-0.43, 0.28))$	$((0.16, 0.72), (0.22, 0.12))$

Step 2. The $BI_m(BF, E_q)$ are estimated as follows:

$$BI_m(BF, E_q) = 0.6987$$

$$BI_m(BF, E_2) = 0.6969$$

$$BI_m(BF, E_3) = 0.5922 .$$

Step 3. The best alternative is the one which has least information measure with respect to six factors. We have $BI_m(BF, E_3) < BI_m(BF, E_2) < BI_m(BF, E_1)$. Therefore, the information measure $BI_m(BF, E_3)$ has the least value and its corresponding alternative is the best alternative. So, the television (BF, E_3) is the best one among the others.

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