

On Some New Measures of Intuitionistic Fuzzy Entropy and Directed Divergence

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Abstract

In the present paper, generalized measures of intuitionistic fuzzy directed divergence with the proof of their validity are introduced. Particular case of corresponding directed divergence and symmetric divergence have also been discussed.

Keywords: Intuitionistic Fuzzy Set, Entropy, Directed Divergence, Measures of Information.

Introduction

Uncertainty and fuzziness are the basic nature of human thinking and of many real world objectives. Fuzziness is found in our decision, in our language and in the way we process information. The main use of information is to remove uncertainty and fuzziness in fact we measure information supplied by the amount of probabilistic uncertainty removed in experiment and the measure of uncertainty removed is also called as measure of information while measure of fuzziness is the measure of vagueness and ambiguity of uncertainties.

Shannon (1948) used "entropy" to measure uncertain degree of the randomness in a probability distribution. Let X is a discrete random variable with pattern recognition $P = (P_1, P_2, P_3, \dots, P_n)$ in an experiment. The information contained in this experiment is given by

$$H(P) = - \sum_{i=1}^n p_i \log p_i$$

Which is known as Shannon entropy.

The concept of entropy has been widely used in different areas, e.g. communication theory,

Generalized measure of intuitionistic fuzzy directed divergence corresponding to Havard & Charvat measure:-

Havard & Charvat (1967) defined the directed divergence measure of a probability distribution $P=(P_1, P_2, \dots, P_n)$ from another probability distribution $Q=(q_1, q_2, \dots, q_n)$ as

$$D^\beta(P:Q) = \frac{1}{\beta-1} \sum_{i=1}^n (p_i^\beta q_i^{1-\beta} - 1); \beta > 0, \beta \neq 1 \dots \dots \dots (1)$$

Which is called generalized directed divergence of degree β .

The following measure of symmetric divergence was proposed by Kulbark (1959):

$$\begin{aligned} J^\beta(P:Q) &= D^\beta(P:Q) + D^\beta(Q:P) \\ &= \frac{1}{\beta-1} \sum_{i=1}^n (p_i^\beta q_i^{1-\beta} + q_i^\beta p_i^{1-\beta} - 2) \dots \dots \dots (2) \end{aligned}$$

Which is also called a distance measure of degree β .

Corresponding (1) and (2) we get following measure of intuitionistic fuzzy directed divergence:

$$I^\beta(A:B) = \frac{1}{\beta-1} \sum_{i=1}^n [\mu_A^\beta(x_i) \mu_B^{1-\beta}(x_i) + V_A^\beta(x_i) V_B^{1-\beta}(x_i) - 1] \dots \dots \dots (3)$$

$$\text{And } J^\beta(A:B) = I^\beta(A:B) + I^\beta(B:A)$$

Now we show that $I^\beta(A:B)$ is a valid measure of intuitionistic fuzzy directed divergence.

$I^\beta(A:B)$ is defined in the range $0 \leq \mu_A(x) + V_A(x) \leq 1$ Further it is proved that $I^\beta(A:B) \geq 0$ for all $\beta \neq 1$ and $\beta > 0$. $I^\beta(A:B)$ is continuous function of $\mu_A(x)$ and $V_A(x)$. It is easy to see that $I^\beta(A:B) = 0$ when $\mu_A(x) = 0$ and $V_A(x) = 1$. $I^\beta(A:B)$ is increasing function of $\mu_A(x)$ in the range $0 \leq \mu_A(x) \leq 0.5$ and decreasing function of $V_A(x)$ in the range $0 \leq V_A(x) \leq 0.5$. $I^\beta(A:B)$ does not changed on changing $\mu_A(x)$ to $V_A(x)$. To verify that $I^\beta(A:B)$ is convex function of $\mu_A(x)$, Let us consider $\mu_A(x) = s$ and $V_A(x) = t$ then

$$\begin{aligned} &\frac{1}{\beta-1} \sum_{i=1}^n [\mu_A^\beta(x_i) \mu_B^{1-\beta}(x_i) + V_A^\beta(x_i) V_B^{1-\beta}(x_i) - 1] \\ &\leq \frac{1}{\beta-1} \sum_{i=1}^n [s^\beta t^{1-\beta} + (1-s)^\beta (1-t)^{1-\beta} - 1] \end{aligned}$$

$$\text{Consider } f(s) = \frac{1}{\beta-1} [s^\beta t^{1-\beta} + (1-s)^\beta (1-t)^{1-\beta} - 1]$$

Therefore $f'(s) = \frac{1}{\beta-1} [\beta s^{\beta-1} t^{1-\beta} + \beta(1-s)^{\beta-1}(-1)(1-t)^{1-\beta}]$

And $f''(s) = \frac{1}{\beta-1} [\beta(\beta-1)s^{\beta-2}t^{1-\beta} + \beta(\beta-1)(1-s)^{\beta-2}(1-t)^{1-\beta}] \geq 0$

Therefore $I^\beta(A: B)$ is convex function of $\mu_A(x_i)$ Hence $I^\beta(A: B)$ is valid measure of intuitionistic fuzzy entropy.

Corresponding to Renyis measure of directed divergence:

$$D_\alpha(P: Q) = \frac{1}{\alpha-1} \log \sum_{i=1}^n p_i^\alpha q_i^{1-\alpha}$$

$\alpha \neq 0, \alpha > 0$ (4)

We define the measure of intuitionistic fuzzy directed divergence

$$I_\alpha(A: B) = \frac{1}{\alpha-1} \log \sum_{i=1}^n [\mu_A^\alpha(x_i)\mu_B^{1-\alpha}(x_i) + V_A^\alpha(x_i)V_B^{1-\alpha}(x_i)] \quad (5)$$

Where $\alpha \neq 0, \alpha > 0$ and measure of intuitionistic fuzzy symmetric divergence

$$J_\alpha(A: B) = I_\alpha(A: B) + J_\alpha(B: A) \dots\dots\dots(6)$$

Now we show that $I_\alpha(A: B)$ is a valid measure of intuitionistic fuzzy directed divergence.

It is obvious that $I_\alpha(A: B) \geq 0$ and is defined in the range $0 \leq \mu_A(x) + V_A(x) \leq 1$

$I_\alpha(A: B)$ is continuous in this range.it is easy to see that if $\mu_A(x) = 0$ and $V_A(x) = 1$ than $I_\alpha(A: B) = 0$.

$I_\alpha(A: B)$ is increasing function of $\mu_A(x)$ in the range $0 \leq \mu_A(x) \leq 0.5$ and decreasing function of $V_A(x)$ in the range $0 \leq V_A(x) \leq 0.5$. $I_\alpha(A: B)$ does not changed on changing $\mu_A(x)$ to $V_A(x)$. To verify that $I_\alpha(A: B)$ is convex function of $\mu_A(x)$, Let us consider $\mu_A(x) = s$ and $V_A(x) = t$ then

$$\begin{aligned} & \frac{1}{\alpha-1} \log \sum_{i=1}^n [\mu_A^\alpha(x_i)\mu_B^{1-\alpha}(x_i) + V_A^\alpha(x_i)V_B^{1-\alpha}(x_i)] \\ & \leq \frac{1}{\alpha-1} \log [s^\alpha t^{1-\alpha} + (1-s)^\alpha(1-t)^{1-\alpha}] \end{aligned}$$

Let

$$\begin{aligned} f(s, t) &= \frac{1}{\alpha-1} \log [s^\alpha t^{1-\alpha} + (1-s)^\alpha(1-t)^{1-\alpha}] \\ f'(s, t) &= \frac{1}{\alpha-1} \frac{1}{s^\alpha t^{1-\alpha} + (1-s)^\alpha(1-t)^{1-\alpha}} [\alpha s^{\alpha-1} t^{1-\alpha} - \\ & \quad \alpha (1-s)^{\alpha-1} (1-t)^{1-\alpha}] \end{aligned}$$

And

$$f''(s, t) = \frac{1}{\alpha - 1} \left[\frac{1}{[s^\alpha t^{1-\alpha} + (1-s)^\alpha (1-t)^{1-\alpha}]^2} \{ \alpha (\alpha - 1) s^{\alpha-2} t^{1-\alpha} + \alpha (\alpha - 1) (1-s)^{\alpha-2} (1-t)^{1-\alpha} \} - \frac{1}{[s^\alpha t^{1-\alpha} + (1-s)^\alpha (1-t)^{1-\alpha}]^2} (\alpha s^{\alpha-1} t^{1-\alpha} - \alpha (1-s)^{\alpha-1} (1-t)^{1-\alpha}) \right]$$

Therefore $I_\alpha(A: B)$ is convex function of $\mu_A(x_i)$ Hence $I_\alpha(A: B)$ is valid measure of intuitionistic fuzzy directed divergence.

Particular cases

$$\lim_{\alpha \rightarrow 1} I_\alpha(A: B) = I(A: B) \text{ and } \lim_{\alpha \rightarrow 1} J_\alpha(A: B) = J(A: B)$$

Where $I(A: B)$ and $J(A: B)$ are intuitionistic fuzzy directed divergence intuitionistic fuzzy symmetry divergence.

Let $B=A_{IF}$ the most intuitionistic fuzzy set i.e. $\mu_B(x_i) = 0.5$ and $V_B(x_i) = 0.5 \forall i$
Then

$$\begin{aligned} I_\alpha(A: A_{IF}) &= \frac{1}{\alpha-1} \sum_{i=1}^n \log[\mu_A^\alpha(x_i)(0.5)^{1-\alpha} + V_A^\alpha(x_i)(0.5)^{1-\alpha}] \\ &= \frac{1}{\alpha-1} \left[\sum_{i=1}^n \log(0.5)^{1-\alpha} + \sum_{i=1}^n \log\{\mu_A^\alpha(x_i) + V_A^\alpha(x_i)\} \right] \\ &= \frac{1}{\alpha-1} \sum_{i=1}^n (\alpha-1) \log 2 + \frac{1}{\alpha-1} \sum_{i=1}^n \log\{\mu_A^\alpha(x_i) + V_A^\alpha(x_i)\} \\ &= n \log 2 - \frac{1}{\alpha-1} \sum_{i=1}^n \log\{\mu_A^\alpha(x_i) + V_A^\alpha(x_i)\} \end{aligned}$$

$$\begin{aligned} \text{Thus } I_\alpha(A: A_{IF}) &= n \log 2 - \frac{1}{\alpha-1} \sum_{i=1}^n \log\{\mu_A^\alpha(x_i) + V_A^\alpha(x_i)\} \\ &= n \log 2 - (\text{Entropy of the intuitionistic fuzzy set}) \end{aligned}$$

Intuitionistic fuzzy entropy corresponding to Sharma and Mittals measure

Sharma and mittals (1975) characterized non additive entropy of discrete probability distribution given by

$$H_\alpha^\beta(p) = \frac{1}{2^{1-\beta} - 1} \left[\left(\sum_{i=1}^n p_i^\alpha \right)^{\frac{\beta-1}{\alpha-1}} - 1 \right]$$

Where $\alpha \neq 1, \alpha > 0, \beta > 0, \beta \neq 1$

Corresponding measure of instutionistic fuzzy entropy is given by

$$H_\alpha^\beta(A) = \frac{1}{2^{1-\beta} - 1} \left[\sum_{i=1}^n (\mu_A^\alpha(x_i) + V_A^\alpha(x_i))^{\beta-1/\alpha-1} - 1 \right]$$

Where $\alpha \neq 1, \alpha > 0, \beta > 0, \beta \neq 1$

Next, we prove $H_\alpha^\beta(A)$ is valid measure $H_\alpha^\beta(A)$ is defined in the range $0 \leq \mu_A(x) + V_A(x) \leq 1$ $H_\alpha^\beta(A)$ is continuous in the range.

$H_\alpha^\beta(A)$ is zero when $\mu_A(x) = 0$ and $V_A(x)=1$.

$H_\alpha^\beta(A)$ is increasing function of $\mu_A(x)$ in the range $0 \leq \mu_A(x) \leq 0.5$ and decreasing function of $V_A(x)$ in the range $0 \leq \mu_A(x) \leq 0.5$. $H_\alpha^\beta(A)$ does not change on changing $\mu_A(x)$ to $V_A(x)$.

Also, $H_\alpha^\beta(A)$ is concave function of $\mu_A(x)$.

Let $\mu_A(x_i) = s, V_A(x_i) = t$

$$f(s, t) = \frac{1}{2^{1-\beta} - 1} \left[(s^\alpha + t^\alpha)^{\frac{\beta-1}{\alpha-1}} - 1 \right]$$

$$f'(s, t) = \frac{1}{2^{1-\beta} - 1} \frac{\beta-1}{\alpha-1} (s^\alpha + t^\alpha)^{\frac{\beta-\alpha}{\alpha-1}} \alpha s^{\alpha-1}$$

$$f''(s, t) = \frac{\alpha}{2^{1-\beta} - 1} \frac{\beta-1}{\alpha-1} \left[\frac{\beta-1}{\alpha-1} (s^\alpha + t^\alpha)^{\frac{\beta-2\alpha+1}{\alpha-1}} s^{\alpha-1} + (s^\alpha + t^\alpha)^{\frac{\beta-\alpha}{\alpha-1}} (\alpha-1) s^{\alpha-2} \right]$$

Intuitionistic fuzzy directed divergence corresponding to Sharma & Mittals measure.

Sharma & Mittals (1977) also studied the following generalized measure of directed divergence

$$I_\alpha^\beta(P:Q) = \frac{1}{1 - 2^{1-\beta}} \left[\left(\sum_{i=1}^n p_i^\alpha q_i^{1-\alpha} \right)^{\frac{\beta-1}{\alpha-1}} - 1 \right]$$

Where $\alpha \neq 1, \alpha > 0, \beta > 0, \beta \neq 1$

Corresponding measure of intuitionistic fuzzy directed divergence is

$$I_\alpha^\beta(A:B) = \left[\left(\sum_{i=1}^n \mu_A^\alpha(x_i) \mu_B^{1-\alpha}(x_i) \right)^{\frac{\beta-1}{\alpha-1}} + \left(\sum_{i=1}^n V_A^\alpha(x_i) V_B^{1-\alpha}(x_i) \right)^{\frac{\beta-1}{\alpha-1}} - 1 \right]$$

Where $\alpha \neq 1, \alpha > 0, \beta > 0, \beta \neq 1$

Now we define following measure of symmetric intuitionistic fuzzy directed divergence

$$I_\alpha^\beta(A:B) = I_\alpha^\beta(A:B) + I_\alpha^\beta(B:A)$$

Next, we prove $I_\alpha^\beta(A:B)$ is valid measure it is obvious that $I_\alpha^\beta(A:B)$ is defined in the range $0 \leq \mu_A(x) + V_A(x) \leq 1$ and is continuous in the range. $I_\alpha^\beta(A:B)$ is zero when $\mu_A(x) = 0$ and $V_A(x)=1$. $I_\alpha^\beta(A:B)$ is increasing function of $\mu_A(x)$ in the range $0 \leq \mu_A(x) \leq 0.5$ and decreasing function of $V_A(x)$ in the range $0 \leq V_A(x) \leq 0.5$. it

does not change on changing $\mu_A(x)$ to $V_A(x)$.

Next we show that $I_\alpha^\beta(A:B)$ is convex function of $\mu_A(x)$ for this we consider $s = \mu_A(x)$ and $t = \mu_B(x)$ then

$$I_\alpha^\beta(A:B) = \frac{1}{2^{\beta-1}-1} \left[\left(\sum_{i=1}^n \mu_A^\alpha(x_i) \mu_B^{1-\alpha}(x_i) \right)^{\frac{\beta-1}{\alpha-1}} + \left(\sum_{i=1}^n V_A^\alpha(x_i) V_B^{1-\alpha}(x_i) \right)^{\frac{\beta-1}{\alpha-1}} - 1 \right] \\ \leq \frac{1}{2^{\beta-1}-1} \left[\sum_{i=1}^n (s^\alpha t^{1-\alpha})^{\frac{\beta-1}{\alpha-1}} + \sum_{i=1}^n ((1-s)^\alpha (1-t)^{1-\alpha})^{\frac{\beta-1}{\alpha-1}} - 1 \right]$$

Consider

$$f(s,t) = \frac{1}{2^{\beta-1}-1} \left[(s^\alpha t^{1-\alpha})^{\frac{\beta-1}{\alpha-1}} + ((1-s)^\alpha (1-t)^{1-\alpha})^{\frac{\beta-1}{\alpha-1}} - 1 \right] \\ f'(s,t) = \frac{1}{2^{\beta-1}-1} \left[\left(\frac{\beta-1}{\alpha-1} \right) (s^\alpha t^{1-\alpha})^{\frac{\beta-\alpha}{\alpha-1}} \alpha s^{\alpha-1} t^{1-\alpha} \right. \\ \left. + \left(\frac{\beta-1}{\alpha-1} \right) ((1-s)^\alpha (1-t)^{1-\alpha})^{\frac{\beta-\alpha}{\alpha-1}} \alpha (-1) (1-s)^{\alpha-1} (1-t)^{1-\alpha} \right]$$

Or

$$f'(s,t) = \frac{1}{2^{\beta-1}-1} \left(\frac{\beta-1}{\alpha-1} \right) \alpha \left[(s^\alpha t^{1-\alpha})^{\frac{\beta-\alpha}{\alpha-1}} s^{\alpha-1} t^{1-\alpha} \right. \\ \left. + ((1-s)^\alpha (1-t)^{1-\alpha})^{\frac{\beta-\alpha}{\alpha-1}} (-1) (1-s)^{\alpha-1} (1-t)^{1-\alpha} \right] \\ f''(s,t) = \frac{1}{2^{\beta-1}-1} \left(\frac{\beta-1}{\alpha-1} \right) \alpha \left[\left(\frac{\beta-\alpha}{\alpha-1} \right) (s^\alpha t^{1-\alpha})^{\frac{\beta-2\alpha+1}{\alpha-1}} s^{\alpha-1} t^{1-\alpha} \alpha s^{\alpha-1} \right. \\ \left. + (s^\alpha t^{1-\alpha})^{\frac{\beta-\alpha}{\alpha-1}} (\alpha-1) s^{\alpha-2} t^{1-\alpha} \right. \\ \left. + \left(\frac{\beta-\alpha}{\alpha-1} \right) ((1-s)^\alpha (1-t)^{1-\alpha})^{\frac{\beta-2\alpha+1}{\alpha-1}} \alpha (1-s)^{\alpha-1} (1-t)^{1-\alpha} (1-s)^{\alpha-1} (1-t)^{1-\alpha} \right. \\ \left. + ((1-s)^\alpha (1-t)^{1-\alpha})^{\frac{\beta-\alpha}{\alpha-1}} (\alpha-1) (1-s)^{\alpha-2} (1-t)^{1-\alpha} \right]$$

We have $f''(s,t) \geq 0$.

So $I_\alpha^\beta(A:B)$ is convex function of $\mu_A(x)$ therefore $I_\alpha^\beta(A:B)$ is valid measure of intuitionistic fuzzy diverted divergence.

Particular cases

$$\lim_{\beta \rightarrow 1} I_\alpha^\beta(A:B) = I_\alpha(A:B) \text{ and } \lim_{\beta \rightarrow 1} J_\alpha^\beta(A:B) = J_\alpha(A:B)$$

Let $B=A_{IFS}$, the most fuzzy set i.e. $\mu_B(x_i) = 0.5$ and $V_B(x_i) = 0.5 \forall x_i$ then

$$I_\alpha^\beta(A:A_{IFS}) = \frac{1}{2^{\beta-1}-1} \left[\left(\sum_{i=1}^n \mu_A^\alpha(x_i) (0.5)^{1-\alpha} \right)^{\frac{\beta-1}{\alpha-1}} + \left(\sum_{i=1}^n \mu_A^\alpha(x_i) (0.5)^{1-\alpha} \right)^{\frac{\beta-1}{\alpha-1}} - 1 \right]$$

$$\begin{aligned}
 &= \frac{1}{2^{\beta-1}} \frac{1}{2^{1-\beta}} \sum_{i=1}^n \left[\left(\mu_A^\alpha(x_i) + V_A^\alpha(x_i) \right)^{\frac{\beta-1}{\alpha-1}} - 2^{1-\beta} \right] \\
 &= \frac{1}{1-2^{1-\beta}} \sum_{i=1}^n \left[\left(\mu_A^\alpha(x_i) + V_A^\alpha(x_i) \right)^{\frac{\beta-1}{\alpha-1}} - 2^{1-\beta} \right] \\
 &= \frac{-n2^{1-\beta}}{1-2^{1-\beta}} + \frac{1}{1-2^{1-\beta}} \sum_{i=1}^n \left[\left(\mu_A^\alpha(x_i) + V_A^\alpha(x_i) \right)^{\frac{\beta-1}{\alpha-1}} - 2^{1-\beta} \right] + \frac{n}{1-2^{1-\beta}} \\
 &= n - H_\alpha^\beta(A)
 \end{aligned}$$

Thus $I_\alpha^\beta(A; A_\beta) = n - (\text{Entropy of intuitionistic fuzzy set})$

Conclusion

We have proposed some new measures of intuitionistic fuzzy directed divergence measures and proved their validity. Total ambiguity measures and fuzzy information improvement measures have also been introduced. Further comparative investigations for the amount of total ambiguity in different measures suggested for different pairs of fuzzy sets with different possible values of α and β can be computationally made and similar investigation can be done for the corresponding fuzzy information improvement measures suggested.

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