

ON SOME RELATION BETWEEN FUZZY PROBABILITY MEASURE
AND FUZZY P-MEASURE

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Let $\mathcal{G} = \{\mu : \Omega \rightarrow [0,1]\}$ be a soft fuzzy \mathcal{G} -algebra i.e. fuzzy \mathcal{G} -algebra (see [1]) uncountaining the fuzzy subset $\left[\frac{1}{2} \right]_{\Omega} : \Omega \rightarrow \left\{ \frac{1}{2} \right\}$.

Let us compare the following notions:

Definition 1: A fuzzy probability measure is a mapping

$$m: \mathcal{G} \rightarrow [0,1]$$

such that

$$m(\emptyset_{\Omega}) = 0 ,$$

$$m(\mathbb{1}_{\Omega}) = 1 ,$$

$$\forall (\mu, \nu) \in \mathcal{G}^2 \quad m(\mu \vee \nu) + m(\mu \wedge \nu) = m(\mu) + m(\nu) ,$$

$$\forall \{\mu_n\} \in \mathcal{G}^{\mathbb{N}} \quad \{\mu_n\} \uparrow \mu \Rightarrow \{m(\mu_n)\} \uparrow m(\mu) . \quad [2]$$

Definition 2: Each mapping

$$p: \mathcal{G} \rightarrow \mathbb{R}^+ \cup \{0\}$$

having the following properties:

- for any $\mu \in \mathcal{G}$

$$p(\mu \vee (1 - \mu)) = 1$$

- if $\{\mu_n\}$ is finite or infinite sequence of pairwise W-separated fuzzy subsets from \mathcal{G} (i.e. $\mu_i \leq 1 - \mu_j$ for each pair (i,j) which $i \neq j$)

then

$$p\left(\sup_n \{\mu_n\}\right) = \sum_n p(\mu_n)$$

is called a fuzzy P-measure. [6]

As we know, the fuzzy P-measure are the unique fuzzy probability measures satisfying the Bayes Formula [6,7]. The next relationship between measures mentioned above, will be presented in this paper.

Let $\mathbb{R} = [-\infty, +\infty]$. Then we have.

Definition 3: A fuzzy relation "less or equal" FLE is a mapping

$$\xi: \mathbb{R}^2 \rightarrow [0, 1]$$

such that

$$\xi(x, y) \geq 1 - \xi(y, x) ,$$

$$\xi(y, x) + \xi(z, y) \leq 1 ,$$

for each $(x, y, z) \in \mathbb{R}^3$ which $x < z$. [3]

Any FLE ξ generates a fuzzy relation "less than" given by the identity

$$\xi_s(x, y) = 1 - \xi(y, x)$$

for every $(x, y) \in \mathbb{R}^2$ [3].

Let ξ be a fixed FLE.

Definition 4: Each mapping:

$$\varphi \langle a, b \rangle : \mathbb{R} \rightarrow [0, 1]$$

defined by the identity

$$\varphi \langle a, b \rangle (x) = \psi(a, x) \wedge \eta(x, b)$$

for every $(a, b, x) \in \mathbb{R}^3$, $(\psi, \eta) \in \{\xi, \xi_s\}^2$, is called a fuzzy interval. [4]

We note, that the above definition describes all kinds of intervals on real line \mathbb{R} generalized for fuzzy case. Among other things, it defines

- if $\psi = \xi$ and $\eta = \xi$ then $\varphi \langle a, b \rangle = \varphi [a, b]$;
- if $\psi = \xi$ and $\eta = \xi_s$ then $\varphi \langle a, b \rangle = \varphi [a, b[$.

Let us suppose now, that FLE ξ is quasi-antisymmetrical, continuous from above and it unfuzzily bounds the real line (see [3]). Then there exists the smallest soft fuzzy \mathfrak{G} -algebra containing all fuzzy intervals $\varphi [-\infty, a[$, β_ξ say [4]. If $m: \beta_\xi \rightarrow [0, 1]$ is a fixed fuzzy probability measure on β_ξ , then we define:

Definition 5: The cumulative distributions function of fuzzy probability measure m is a mapping

$$F: \mathbb{R} \rightarrow [0, 1]$$

defined as

$$\forall x \in \mathbb{R} \quad F(x) = m(\varphi [-\infty, x[) . \quad [5]$$

Theorem 1: Each cumulative distribution function F fulfils the following conditions:

$$\forall (x, y) \in \mathbb{R}^2 \quad x \leq y \Rightarrow F(x) \leq F(y) , \quad (1)$$

$$\forall \{x_n\} \in \mathbb{R}^{\mathbb{N}} \quad \{x_n\} \uparrow x \Rightarrow \{F(x_n)\} \uparrow F(x) , \quad (2)$$

$$\lim_{x \uparrow +\infty} F(x) = 1 = F(+\infty) , \quad (3)$$

$$\lim_{x \uparrow -\infty} F(x) = \alpha \geq 0 = F(-\infty) . \quad [5] \quad (4)$$

Theorem 2: If a function $f: \mathbb{R} \rightarrow [0, 1]$ fulfils the properties (1), (2), (3) and (4) then there exists the unique fuzzy P-measure $p: \beta_\xi \rightarrow [0, 1]$ such that

$$\begin{aligned} \forall x \in \mathbb{R} \quad p(\varphi[-\infty, x[) &= f(x) \quad , \\ p(\varphi[+\infty, +\infty]) &= 0 \quad [8] \quad . \end{aligned} \quad (5)$$

More details about fuzzy P-measure p generated by function f we can find in [8]. If we take into account two above theorems, then we obtain finally thesis.

Theorem 3: For each fuzzy probability measure $m: \mathcal{F}_S \rightarrow [0,1]$, there exists the unique fuzzy P-measure $p: \mathcal{F}_S \rightarrow [0,1]$ satisfying the conditions (5) and

$$\forall x \in \mathbb{R} \quad p(\varphi[-\infty, x[) = m(\varphi[-\infty, x[) \quad .$$

By means of this theorem, the Bayes Formula can be applied for any fuzzy probability space defined by Klement et.al. [2].

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