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ENTROPY OF COMPLETE FUZZY PARTITIONS

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ABSTRACT. This paper deals with a fuzzy generalization of notion of a probability space. An entropy and a conditional entropy of complete fuzzy partitions are defined. The main properties of such quantities are proved.

0. Introduction

In the classical probability theory [1] probability spaces (X, \mathcal{S}, P) are studied. A σ -algebra \mathcal{S} of subsets of a set X is the main notion of the Kolmogorov classical model of probability theory. The Kolmogorov probability model may be uniquely represented by a system of characteristic functions of subsets of a set X from the given σ -algebra \mathcal{S} , which have values in the closed interval (0, 1). When an event f, say, is described vaguely, then by a fuzzy set f (fuzzy event f) we shall understand a real-valued function $f: X \to (0, 1)$, which describes the fuzziness of the event f. This is a basic idea of Z a d e h 's fuzzy sets theory [2].

In this paper we shall use a fuzzy generalization of notion of a probability space. A fuzzy generalization of a notion of measurable partition from the classical probability theory is a notion of complete fuzzy partition [3]. In this paper an entropy and a conditional entropy of complete fuzzy partitions are defined. The main properties of such quantities are stated.

1. Basic definitions and facts

Here we follow mainly [3]. Let $X \neq \emptyset$. By a soft fuzzy σ -algebra M we mean the set $M \subset (0,1)^X$ satisfying the following conditions:

- (1.1) if 1(x) = 1 for any $x \in X$, then $1 \in M$;
- (1.2) if $f \in M$, then $f' := 1 f \in M$;
- (1.3) if 1/2(x) = 1/2 for any $x \in X$, then $1/2 \notin M$;

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(1.4)
$$\bigvee_{n=1}^{\infty} f_n := \sup_n f_n \in M \text{ for any } \{f_n\}_{n=1}^{\infty} \subset M$$

In the set M we define the partial ordering relation in the following way: $f \leq g$ if and only if $f(x) \leq g(x)$ for each $x \in X$. Using the complementation ': $f \to f'$ for any fuzzy subset $f \in M$, we see that the complementation ' satisfies two conditions:

- (1.5) (f')' = f for every $f \in M$;
- (1.6) if $f \le g$, then $g' \le f'$.

So that M is a distributive σ -lattice with the complementation ', for which the de Morgan laws hold:

(1.7)
$$\left(\bigvee_{n=1}^{\infty} f_n\right)' = \bigwedge_{n=1}^{\infty} f'_n;$$

(1.8) $\left(\bigwedge_{n=1}^{\infty} f_n\right)' = \bigvee_{n=1}^{\infty} f'_n$ for any sequence $\{f_n\}_{n=1}^{\infty} \subset M$

Of course, here $\bigwedge_n f_n = \inf_n f_n$. In the fuzzy sets theory the fuzzy subset 1 is

called universum, the fuzzy subset 0 = 1' is called *empty set* and all fuzzy subsets $f, g \in M$ such that $f \wedge g = 0$ are called *separated fuzzy sets*. Analogous weak notions (W-notions) are defined in [4] as follows: Each fuzzy subset $f \in M$ such that $f \geq 1 - f$ is called a W-universum. Each fuzzy subset $f \in M$ such that $f \leq 1 - f$ is called a W-empty set. All fuzzy subsets $f, g \in M$ such that $f \leq 1 - g$ are called W-separated fuzzy sets.

LEMMA 1.1. A fuzzy subset $f \in M$ is a W-universum if and only if there exists a fuzzy subset $g \in M$ such that $f = g \lor (1 - g)$ [4].

LEMMA 1.2. Let a finite or infinite sequence $\{f_n\}$ of fuzzy subsets from M be given. Then the fuzzy subsets g_n defined by

$$g_{n} = \begin{cases} f_{1}, & \text{if } n = 1, \\ f_{n} \wedge \left(\bigvee_{i=1}^{n-1} f_{i}\right)', & \text{if } n > 1 \end{cases}$$
(1.9)

are pairwise W-separated. Furthermore, if $\bigvee_n f_n$ is a W-universum, then $\bigvee_n g_n$ is a W-universum [3].

A fuzzy *P*-measure on *M* is a mapping $m: M \to (0, \infty)$ fulfilling the following conditions:

- (1.10) $m(f \lor (1-f)) = 1$ for every $f \in M$;
- (1.11) if $\{f_n\}_{n=1}^{\infty}$ is a finite or infinite sequence of pairwise W-separated fuzzy subsets from M, then $m\left(\bigvee_{n=1}^{\infty} f_n\right) = \sum_{n=1}^{\infty} m(f_n)$.

Each above described triplet (X, M, m) is called in the fuzzy theory a soft fuzzy probability space.

Example 1.1. Let (X, \mathcal{S}, P) be a probability space in the sense of the classical probability theory. Put $M = \{\chi_A; A \in \mathcal{S}\}$ (χ_A) is the characteristic function of the set $A \in \mathcal{S}$. If we define the mapping $m: M \to \langle 0, 1 \rangle$ by the equality $m(\chi_A) = P(A)$, then the triplet (X, M, m) is a soft fuzzy probability space.

Example 1.2. Let $X = \langle 0, 1 \rangle$, $M = \{f, f', f \lor f', f \land f', 0, 1\}$, where $f: X \to \langle 0, 1 \rangle$, f(x) = x for each $x \in X$. If we define the mapping $m: M \to \langle 0, 1 \rangle$ by the equalities m(f) = m(f') = 1/2, $m(1) = m(f \lor f') = 1$, $m(0) = m(f \land f') = 0$, then the triplet (X, M, m) is a soft fuzzy probability space.

It is easy to see that any fuzzy P-measure m has the following properties:

- (1.12) m(f') = 1 m(f) for every $f \in M$.
- (1.13) m is a nondecreasing function, i.e. if $f,g \in M$, $f \leq g$, then $m(f) \leq m(g)$.
- (1.14) Let $g \in M$ be given. Then $m(f \wedge g) = m(f)$ for all $f \in M$ if and only if m(g) = 1.
- (1.15) If $f, g \in M$ are W-separated, then $m(f \wedge g) = 0$.

The mapping $m(\cdot/g): M \to (0, \infty)$ defined for each $g \in M$, m(g) > 0, by the equality $m(f/g) = \frac{m(f \wedge g)}{m(g)}$, is a *P*-measure on *M* (see [3]).

The monotonicity of fuzzy *P*-measure implies that this measure transforms M into the interval (0,1).

2. Entropy of complete fuzzy partitions

Let any soft fuzzy probability space (X, M, m) be given. Kabala and Wrocinski [5] mean by a complete partition each finite or infinite sequence of pairwise separated fuzzy subsets $\{f_n\}$ such that $\bigvee_n f_n$ is a universum. If $\{f_n\} \subset M$ is a complete partition, then for every $x \in X$ there exists i_0 such that $f_{i_0}(x) = 1$ and for every $j \neq i_0$, $f_j(x) = 0$ holds. This means that $\{f_n\}$ contains only crisp subsets of the set X and hence the mentioned definition is not useful for considerations on fuzzy subsets. Therefore in this contribution we shall work with the following notion:

DEFINITION 2.1. [3] Each finite or infinite sequence of pairwise W-separated fuzzy subsets $\{f_n\} \subset M$ such that $\bigvee f_n$ is a W-universum is called a complete fuzzy partition.

It is easy to see that a partition described in Definition 2.1 contains uncrisp subsets, in general. Namely, if we have any sequence $\{f_n\}$ such that $\bigvee_n f_n$ is a *W*-universum, then we can find the complete fuzzy partition $\{g_n\}$ defined by (1.9). The sequence $\{f_n\}$ described above always exists. So, if it does not contain crisp subsets only, then the generated partition $\{g_n\}$ contains uncrisp subsets.

LEMMA 2.1. Let $\mathcal{A} = \{f_i\}$ and $\mathcal{B} = \{g_j\}$ be two complete fuzzy partitions. Then the set $\mathcal{A} \lor \mathcal{B} := \{f_i \land g_j; f_i \in \mathcal{A}, g_j \in \mathcal{B}\}$ is a complete fuzzy partition, too.

Proof. It is easy to see that $\mathcal{A} \lor \mathcal{B}$ is a set of pairwise W-separated elements (see [6]). Moreover, $\bigvee_i \bigvee_j (f_i \land g_j) = (\bigvee_i f_i) \land (\bigvee_j g_j) \ge 1/2$, so that $\bigvee_i \bigvee_j (f_i \land g_j)$ is a W-universum.

In the set \mathcal{F} of all complete fuzzy partitions we can define the relation \leq in the following way: for every $\mathcal{A}, \mathcal{B} \in \mathcal{F}, \ \mathcal{A} \leq \mathcal{B}$ if and only if for every $g \in \mathcal{B}$ there exists $f \in \mathcal{A}$ such that $g \leq f$. In this case we say that \mathcal{B} is the refinement of \mathcal{A} . Since $\mathcal{A} \leq \mathcal{A} \lor \mathcal{B}, \ \mathcal{B} \leq \mathcal{A} \lor \mathcal{B}$, we shall read the symbol $\mathcal{A} \lor \mathcal{B}$ a common refinement of \mathcal{A} and \mathcal{B} . Each $\mathcal{A} = \{f_1, f_2, \ldots\} \in \mathcal{F}$ represents in the sense of classical probability theory the random experiment with finite or countable number of outcomes with the probability distribution $p_i = m(f_i), \ f_i \in \mathcal{A}$, since $p_i \geq 0$ and $\sum_i p_i = \sum_i m(f_i) = m(\bigvee_i f_i) = 1$ (see Lemma 1.1 and (1.10)).

We define an entropy of any experiment $\mathcal{A} = \{f_1, f_2, ...\} \in \mathcal{F}$ by Shannon's formula:

$$H_m(\mathcal{A}) = -\sum_i F(m(f_i)), \quad \text{where} \quad F: \langle 0, \infty) \to \mathbb{R}, \quad (2.1)$$

$$F(x) = \left\{egin{array}{ccc} x\log x\,, & ext{if} & x>0\,, \ 0\,, & ext{if} & x=0\,, \end{array}
ight.$$

 $H_m(\mathcal{A})$ is not necessarily finite.

If $\mathcal{A}, \mathcal{B} \in \mathcal{F}, \ \mathcal{A} = \{f_i\}, \ \mathcal{B} = \{g_j\}$, we define a *conditional entropy*

$$H_m(\mathcal{B}/_{\mathcal{A}}) = -\sum_i \sum_j m(f_i) F(\mathring{m}(g_j/f_i)), \qquad (2.2)$$

where

$$\mathring{\mathrm{m}}(g_j/f_i) = \left\{ egin{array}{cc} m(g_j/f_i)\,, & \mathrm{if} & m(f_i) > 0\,, \ 0\,, & \mathrm{if} & m(f_i) = 0\,. \end{array}
ight.$$

The following example shows that the notion of entropy of complete fuzzy partition is a generalization of S h a n n o n's entropy of a measurable partition [7].

E x a m p l e 2.1. Let (X, \mathcal{S}, P) be a probability space in the sense of the classical probability theory. Let us consider the soft fuzzy probability space (X, M, m) from Example 1.1. Then the system \mathcal{F} contains all partitions of the type $\{\chi_{A_1}, \ldots, \chi_{A_k}\}$, where $A_i \in \mathcal{S}$ $(i = 1, \ldots, k)$, $A_i \cap A_j = \emptyset$ $(i \neq j)$ and $\bigcup_{i=1}^k A_i = X$. The entropy of a complete fuzzy partition $\mathcal{A} = \{\chi_{A_1}, \ldots, \chi_{A_k}\}$ is the number $H_m(\mathcal{A}) = -\sum_{i=1}^k F(m(\chi_{A_i})) = -\sum_{i=1}^k F(P(A_i))$, which is the Shan-

non entropy of measurable partition $\{A_1, \ldots, A_k\}$ of a space (X, \mathcal{S}, P) .

E x a m p l e 2.2. Let (X, M, m) be a soft fuzzy probability space from Example 1.2. Then the set $\mathcal{A} = \{f, f'\}$ is a complete fuzzy partition with the non-zero entropy $H_m(\mathcal{A}) = \log 2$.

THEOREM 2.1. The entropy H_m has the following properties:

(2.3) $H_m(\mathcal{A}) \geq 0$ for each $\mathcal{A} \in \mathcal{F}$;

(2.4) if $\mathcal{A}, \mathcal{B} \in \mathcal{F}$, $\mathcal{A} \leq \mathcal{B}$, then $H_m(\mathcal{A}) \leq H_m(\mathcal{B})$;

(2.5) $H_m(\mathcal{A}) \leq H_m(\mathcal{A} \vee \mathcal{B})$, for every $\mathcal{A}, \mathcal{B} \in \mathcal{F}$.

Proof. The property (2.3) is evident. Let $\mathcal{A}, \mathcal{B} \in \mathcal{F}, \mathcal{A} = \{f_i\}, \mathcal{B} = \{g_j\}, \mathcal{A} \leq \mathcal{B}$. Then for every $g_j \in \mathcal{B}$ there exists $f_{i_0} \in \mathcal{A}$ such that $g_j \leq f_{i_0}$. Since \mathcal{A} is a system of pairwise W-separated elements we have $g_j = g_j \wedge f_{i_0} \leq f_{i_0} \leq 1 - f_i$ for every $i \neq i_0$. Therefore by (1.15) we obtain $F(m(g_j)) = \sum F(m(g_j \wedge f_i))$.

 $\text{Put } \alpha = \left\{ (i,j)\,; \ m(f_i \wedge g_j) > 0 \right\}, \ \beta = \left\{ i\,; \ m(f_i) > 0 \right\}. \text{ Then we have }$

$$H_m(\mathcal{B}) = -\sum_j F(m(g_j)) = -\sum_j \sum_i F(m(g_j \wedge f_i))$$

= $-\sum_{(i,j)\in\alpha} m(f_i \wedge g_j) \cdot \log m(f_i \wedge g_j)$
= $-\sum_{(i,j)\in\alpha} m(f_i \wedge g_j) \cdot \log m(g_j/f_i) - \sum_{(i,j)\in\alpha} m(f_i \wedge g_j) \cdot \log m(f_i)$
 $\geq -\sum_{i\in\beta} \log m(f_i) \sum_j m(f_i \wedge g_j)$
= $-\sum_{i\in\beta} m(f_i) \log m(f_i) = -\sum_i F(m(f_i)) = H_m(\mathcal{A}).$

Since $\mathcal{A} \leq \mathcal{A} \vee \mathcal{B}$, the inequality (2.5) is a simple consequence of (2.4).

THEOREM 2.2. $H_m(\mathcal{B} \vee \mathcal{C}/\mathcal{A}) = H_m(\mathcal{C}/\mathcal{A} \vee \mathcal{B}) + H_m(\mathcal{B}/\mathcal{A})$ for every $\mathcal{A}, \mathcal{B}, \mathcal{C} \in \mathcal{F}$.

Proof. Let $\mathcal{A} = \{f_i\}$, $\mathcal{B} = \{g_j\}$, $\mathcal{C} = \{h_k\}$. If $m(f_i \wedge g_j) > 0$, then we have

$$m(g_j \wedge h_k/f_i) = \frac{m(g_j \wedge h_k \wedge f_i)}{m(f_i)} = \frac{m(g_j \wedge h_k \wedge f_i)}{m(f_i \wedge g_j)} \frac{m(f_i \wedge g_j)}{m(f_i)}$$
$$= m(h_k/f_i \wedge g_j) \cdot m(g_j/f_i).$$

Moreover, it is easy to see that the function F satisfies the condition

 $F(x \cdot y) = x \cdot F(y) + y \cdot F(x)$ for each $x, y \in (0, \infty)$. (2.6)

Therefore we obtain

$$H_{m}(\mathcal{B} \vee \mathcal{C}/\mathcal{A}) = -\sum_{i} \sum_{j} \sum_{k} m(f_{i}) \cdot F(\mathring{m}(g_{j} \wedge h_{k}/f_{i}))$$

$$= -\sum_{i} \sum_{j} \sum_{k} m(f_{i}) \cdot F(\mathring{m}(h_{k}/f_{i} \wedge g_{j}) \cdot \mathring{m}(g_{j}/f_{i}))$$

$$= -\sum_{i} \sum_{j} \sum_{k} m(f_{i}) (\mathring{m}(g_{j}/f_{i}) \cdot F(\mathring{m}(h_{k}/f_{i} \wedge g_{j})))$$

$$+ \mathring{m}(h_{k}/f_{i} \wedge g_{j}) \cdot F(\mathring{m}(g_{j}/f_{i})))$$

$$= -\sum_{i} \sum_{j} \sum_{k} m(f_{i}) \cdot \mathring{m}(g_{j}/f_{i}) \cdot F(\mathring{m}(h_{k}/f_{i} \wedge g_{j}))$$

$$-\sum_{i} \sum_{j} \sum_{k} m(f_{i}) \cdot \mathring{m}(h_{k}/f_{i} \wedge g_{j}) \cdot F(\mathring{m}(g_{j}/f_{i}))$$

$$= -\sum_{i}\sum_{j}\sum_{k}m(f_{i} \wedge g_{j}) \cdot F(\mathring{m}(h_{k}/f_{i} \wedge g_{j}))$$
$$-\sum_{i}\sum_{j}m(f_{i})\sum_{k}\mathring{m}(h_{k}/f_{i} \wedge g_{j}) \cdot F(\mathring{m}(g_{j}/f_{i}))$$
$$= H_{m}(\mathcal{C}/\mathcal{A} \vee \mathcal{B}) + H_{m}(\mathcal{B}/\mathcal{A}).$$

THEOREM 2.3. Let $\mathcal{A}, \mathcal{B} \in \mathcal{F}, \mathcal{A} \leq \mathcal{B}$. Then $H_m(\mathcal{A}/\mathcal{C}) \leq H_m(\mathcal{B}/\mathcal{C})$ for each $\mathcal{C} \in \mathcal{F}$.

Proof. Put $\mathcal{A} = \{f_i\}, \ \mathcal{B} = \{g_j\}, \ \mathcal{C} = \{h_k\}.$ Since $\mathcal{A} \leq \mathcal{B}$ we have $F(\mathring{m}(g_j/h_k)) = \sum_i F(\mathring{m}(g_j \wedge f_i/h_k)).$ This fact along with (2.6) implies

$$H_{m}(\mathcal{B}/_{\mathcal{C}}) = -\sum_{j} \sum_{k} m(h_{k}) \cdot F(\mathring{m}(g_{j}/h_{k}))$$

$$= -\sum_{j} \sum_{k} m(h_{k}) \sum_{i} F(\mathring{m}(g_{j} \wedge f_{i}/h_{k}))$$

$$= -\sum_{j} \sum_{k} m(h_{k}) \sum_{i} F(\mathring{m}(g_{j}/f_{i} \wedge h_{k}) \cdot \mathring{m}(f_{i}/h_{k}))$$

$$= -\sum_{j} \sum_{k} m(h_{k}) \sum_{i} \mathring{m}(f_{i}/h_{k}) \cdot F(\mathring{m}(g_{j}/f_{i} \wedge h_{k}))$$

$$-\sum_{j} \sum_{k} m(h_{k}) \sum_{i} \mathring{m}(g_{j}/f_{i} \wedge h_{k}) \cdot F(\mathring{m}(f_{i}/h_{k}))$$

$$\geq -\sum_{i} \sum_{k} m(h_{k}) \sum_{j} \mathring{m}(g_{j}/f_{i} \wedge h_{k}) \cdot F(\mathring{m}(f_{i}/h_{k})) = H_{m}(\mathcal{A}/_{\mathcal{C}}).$$

LEMMA 2.2. Let $\mathcal{A}, \mathcal{B} \in \mathcal{F}, \mathcal{A} = \{f_i\}, \mathcal{B} = \{g_j\}, \mathcal{A} \leq \mathcal{B}$. Then for every $h \in M$ it holds that $m\left(h \wedge \left(\bigvee_{j \in \delta_i} g_j\right)\right) = m(h \wedge f_i)$, where $\delta_i = \{j; g_j \leq f_i\}$, $i = 1, 2, \ldots$.

Proof. Since $\bigvee_{j\in\delta_i}g_j\leq f_i$, the monotonicity of fuzzy P-measure implies the inequality

$$m\left(h\wedge \left(\bigvee_{j\in\delta_i}g_j
ight)
ight)\leq m(h\wedge f_i)$$
 $(i=1,2,\ldots)$

Let us suppose that the assertion of the proved lemma is not true. This means that there exists i_0 such that $m\left(h \wedge \left(\bigvee_{j \in \delta_{i_0}} g_j\right)\right) < m(h \wedge f_{i_0})$. Then we get

$$\sum_{i} m\left(h \wedge \left(\bigvee_{j \in \delta_{i}} g_{j}\right)\right) < \sum_{i} m(h \wedge f_{i}).$$

This conclusion is contradictory, because by (1.14) we have

$$\sum_{i} m\left(h \land \left(\bigvee_{j \in \delta_{i}} g_{j}\right)\right) = m\left(h \land \left(\bigvee_{j} g_{j}\right)\right) = m(h)$$

 and

$$\sum_{i} m(h \wedge f_{i}) = m\left(h \wedge \left(\bigvee_{i} f_{i}\right)\right) = m(h).$$

THEOREM 2.4. Let \mathcal{A}, \mathcal{B} be two complete fuzzy partitions, $\mathcal{A} \leq \mathcal{B}$. Then $H_m(\mathcal{C}/\mathcal{A}) \geq H_m(\mathcal{C}/\mathcal{B})$ for each $\mathcal{C} \in \mathcal{F}$.

Proof. The function F is convex and therefore for any convex combination $\sum_{j} \alpha_{j} x_{j}$ (i.e. such that $\alpha_{j} \geq 0, \ j = 1, 2, ..., \ \sum_{j} \alpha_{j} = 1$) of elements $x_{j} \in \langle 0, 1 \rangle$ there holds

$$F\left(\sum_{j} \alpha_{j} x_{j}\right) \leq \sum_{j} \alpha_{j} F(x_{j}).$$
(2.7)

Let $\mathcal{A} = \{f_i\}, \ \mathcal{B} = \{g_j\}, \ \mathcal{C} = \{h_k\}$. Denote by

$$\alpha = \{i; m(f_i) > 0\}, \qquad \beta = \{j; m(g_j) > 0\}, \qquad \gamma = \{k; m(h_k) > 0\},$$

 $\begin{array}{ll} \delta_i = \{j \; ; \; g_j \leq f_i\}, & i = 1, 2, \dots & \text{Put } \alpha_j = \mathring{\mathrm{m}}(g_j/f_i), & x_j = \mathring{\mathrm{m}}(h_k/g_j), \\ i, k - \text{fixed}, \; j = 1, 2, \dots & \text{Let } i \in \alpha . \text{ Then} \end{array}$

$$\sum_{j \in \beta} \alpha_j = \sum_{j \in \beta} m(g_j/f_i) = \sum_j m(g_j/f_i) = \sum_j \frac{m(g_j \wedge f_i)}{m(f_i)}$$
$$= \frac{m\left(\left(\bigvee_j g_j\right) \wedge f_i\right)}{m(f_i)} = 1.$$

By the preceding lemma we get

$$\sum_{j \in \beta} \alpha_j x_j = \sum_{j \in \beta} m(g_j/f_i) \cdot m(h_k/g_j) = \sum_{j \in \beta} \frac{m(g_j \wedge f_i)}{m(f_i)} \frac{m(h_k \wedge g_j)}{m(g_j)}$$
$$= \sum_{j \in \delta_i} \frac{m(g_j)}{m(f_i)} \frac{m(h_k \wedge g_j)}{m(g_j)} = \frac{m\left(h_k \wedge \left(\bigvee_{j \in \delta_i} g_j\right)\right)}{m(f_i)}$$
$$= \frac{m(h_k \wedge f_i)}{m(f_i)} = \mathring{m}(h_k/f_i).$$

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Evidently $\sum_{j \in \beta} \alpha_j x_j = \mathring{m}(h_k/f_i)$ also for $i \notin \alpha$. By (2.7) we obtain $F(\mathring{m}(h_k/f_i))$ $\leq \sum_j \mathring{m}(g_j/f_i) \cdot F(\mathring{m}(h_k/g_j))$. If we multiply this inequality with $-m(f_i)$, we get

$$-m(f_i) \cdot F(\mathring{\mathrm{m}}(h_k/f_i)) \geq -m(f_i) \sum_j \mathring{\mathrm{m}}(g_j/f_i) \cdot F(\mathring{\mathrm{m}}(h_k/g_j)), \quad i, k = 1, 2, \dots$$

Hence

$$H_m(\mathcal{C}/_{\mathcal{A}}) = -\sum_i \sum_k m(f_i) \cdot F(\mathring{m}(h_k/f_i))$$

$$\geq -\sum_i \sum_k \sum_j m(f_i) \cdot \mathring{m}(g_j/f_i) \cdot F(\mathring{m}(h_k/g_j))$$

$$= -\sum_j \sum_k m(g_j) \cdot F(\mathring{m}(h_k/g_j)) = H_m(\mathcal{C}/_{\mathcal{B}}).$$

THEOREM 2.5. $H_m(\mathcal{A}/\mathcal{B}) \leq H_m(\mathcal{A})$ for each $\mathcal{A}, \mathcal{B} \in \mathcal{F}$.

Proof. Put $\mathcal{E} = \{1\}, \ \mathcal{A} = \{f_i\}$. Then

$$H_m(\mathcal{A}/\mathcal{E}) = -\sum_i m(1) \cdot F(m(f_i/1)) = -\sum_i F(m(f_i)) = H_m(\mathcal{A}).$$

Since any complete fuzzy partition \mathcal{B} is a refinement of the partition $\mathcal{E} = \{1\}$, by means of Theorem 2.4 we obtain $H_m(\mathcal{A}) = H_m(\mathcal{A}/\mathcal{E}) \geq H_m(\mathcal{A}/\mathcal{B})$.

THEOREM 2.6. For each $\mathcal{A}, \mathcal{B}, \mathcal{C} \in \mathcal{F}$, we have:

$$H_m(\mathcal{B} \vee \mathcal{C}/_{\mathcal{A}}) \leq H_m(\mathcal{B}/_{\mathcal{A}}) + H_m(\mathcal{C}/_{\mathcal{A}}).$$

Proof. Since $\mathcal{A} \leq \mathcal{A} \vee \mathcal{B}$ for each $\mathcal{A}, \mathcal{B} \in \mathcal{F}$, according to Theorem 2.4 we have the inequality

$$H_m(\mathcal{C}/_{\mathcal{A}\vee\mathcal{B}}) \leq H_m(\mathcal{C}/_{\mathcal{A}}).$$

This along with Theorem 2.2 implies

$$H_m(\mathcal{B} \vee \mathcal{C}/_{\mathcal{A}}) = H_m(\mathcal{C}/_{\mathcal{A} \vee \mathcal{B}}) + H_m(\mathcal{B}/_{\mathcal{A}}) \leq H_m(\mathcal{C}/_{\mathcal{A}}) + H_m(\mathcal{B}/_{\mathcal{A}}).$$

We have seen that the conditional entropy of complete fuzzy partitions defined here fulfils all properties analogous to the properties of entropy of measurable partitions in the crisp case.

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