

THE INTERDEPENDENCE BETWEEN ORTHOGONAL PARTITIONS WITH FUZZY EVENTS

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ABSTRACT: In this paper an index quantifying the interdependence between orthogonal partitions with fuzzy events is defined and its properties are investigated.

Keywords: fuzzy event, orthogonal partition, index of the interdependence

RESUME: Dans cet article nous définons un index qu'évalue l'interdépendance parmi deux partitions orthogonales avec événements flous, et nous analysons ses propriétés.

Mots clés: événement flou, partition orthogonale, index de l'interdépendance.

1. INTRODUCTION

Let (X, β_X, P) be a probability space, where X is a set in the real line R , β_X is the smallest Borel σ -field on X , and P a probability distribution on (X, β_X) .

A fuzzy event on X is a fuzzy set X on X whose membership function μ_X - which associates with each element $x \in X$ a real number in the unit interval $[0,1]$ with the value $\mu_X(x)$ representing the "grade of membership" of x in X - is Borel-measurable. If $\sum_{X \in X^*} \mu_X(x) = 1 \quad \forall x \in X$ the family $X^* = \{X\}$ is called orthogonal partition with fuzzy events associated with X .

Following Zadeh (6) the probability P associated with the measurable space (X, β_X) induces a probability distribution on X^* given by

$$P(X) = \int_X \mu_X(x) dP(x)$$

and the uncertainty of X^* (Ken Kuriyama(4)) is defined by

$$H(X^*) = - \sum_{X \in X^*} P(X) \log P(X)$$

If (X, \mathcal{B}_X, P^1) and (Y, \mathcal{B}_Y, P^2) are two probability spaces, the product probability space is the probability space $(X \times Y, \mathcal{B}_{X \times Y}, P)$ such that \mathcal{B}_X is the σ -field over X induced from $\mathcal{B}_{X \times Y}$ by the projection of index $i=1$, \mathcal{B}_Y is the σ -field over Y induced from $\mathcal{B}_{X \times Y}$ by the projection of index $i=2$, P^1 is the probability measure on \mathcal{B}_X induced from P and P^2 is the probability measure on \mathcal{B}_Y induced from P .

Let X^* and Y^* be two orthogonal partitions with fuzzy events associated with X and Y , respectively. We define the combined orthogonal partition $X^* \times Y^*$ as the orthogonal partition on $X \times Y$ formed by all algebraic products of a fuzzy event $X \in X^*$ and a fuzzy event $Y \in Y^*$, denoted by (X, Y) .

Then a probability P associated with $(X \times Y, \mathcal{B}_{X \times Y})$ induces the probability distribution on $X^* \times Y^*$ given by

$$P(X, Y) = \int_{X \times Y} \mu_X(x) \mu_Y(y) dP(x, y)$$

Let X^* and Y^* be two orthogonal partitions with fuzzy events associated with X and Y respectively. X^* and Y^* are said to be independent if and only if, $\forall (X, Y) \in X^* \times Y^*$, $P(X, Y) = P(X) \cdot P(Y)$

(That is, according to Zadeh's notion of probabilistic independence, (6), X^* and Y^* are said to be independent if and only if, whatever the fuzzy events $X \in X^*$ and $Y \in Y^*$ may be, X and Y are independent).

In particular and according to preceding definition, if the probability spaces (X, \mathcal{B}_X, P^1) and (Y, \mathcal{B}_Y, P^2) are independent then X^* and Y^* are also independent.

Let $X^* \times Y^*$ be a combined orthogonal partition with fuzzy events on $X \times Y$. Then the uncertainty of $X^* \times Y^*$ is given by

$$H(X^* \times Y^*) = \sum_{X \in X^*} \sum_{Y \in Y^*} P(X, Y) \cdot \log P(X, Y)$$

Finally, the uncertainty of the orthogonal partition Y^* given the partition X^* is given by

$$H(Y^*/X^*) = \sum_{X \in X^*} P(X) \sum_{Y \in Y^*} P(Y|X) \cdot \log P(Y|X)$$

In this framework, an index of the interdependence between the orthogonal partitions with fuzzy events, X^* , Y^* is given by

$$i(X^*, y^*) = \begin{cases} 1 - \frac{H(X^*/y^*)}{H(X^*)} & \text{if } H(X^*) \geq H(y^*) \\ 1 - \frac{H(y^*/X^*)}{H(y^*)} & \text{if } H(X^*) \leq H(y^*) \end{cases}$$

2. PROPERTIES

In this section, we establish the properties of the index of interdependence.

Theorem 1

Let X^* and y^* be two orthogonal partitions with fuzzy events. Then

$$0 \leq i(X^*, y^*) \leq 1$$

Proof.

Assume $H(X^*) \geq H(y^*)$; following Okuda, Tanaka and Asai (5) we have

$$H(X^*/y^*) \leq H(X^*)$$

whence

$$\frac{H(X^*/y^*)}{H(X^*)} \leq 1$$

consequently

$$1 - \frac{H(X^*/y^*)}{H(X^*)} \geq 0$$

Therefore

$$i(X^*, y^*) \geq 0$$

In addition we obtain

$$H(X^*) \geq 0 \quad \text{and} \quad H(X^*/y^*) \geq 0$$

and hence

$$\frac{H(X^*/y^*)}{H(X^*)} \geq 0$$

Then

$$i(X^*, y^*) \leq 1$$

On the other hand if $H(X^*) \leq H(y^*)$, it can be verified with similar arguments that

$$0 \leq i(X^*, y^*) \leq 1$$

Theorem 2

Let X^* and y^* be two orthogonal partition with fuzzy events associated with X and Y respectively. Then $i(X^*, y^*) = 0$ if and only if X^* and y^* are independent orthogonal partitions.

Proof.

Assume that $H(X^*) \geq H(Y^*)$. In this case, the condition $i(X^*, Y^*) = 0$ is equivalent to $H(X^*/Y^*) = H(X^*)$

We are now going to verify that $H(X^*/Y^*) = H(X^*)$ if and only if X^* and Y^* are independent. Indeed, $\sum_{X \in X^*} P(X/Y) = \sum_{X \in X^*} P(X) = 1$

If we apply Gibbs' Lemma we have,

$$-\sum_{X \in X^*} P(X/Y) \cdot \log P(X/Y) - \sum_{X \in X^*} P(X/Y) \cdot \log P(X) \quad (1)$$

with equality if and only if $P(X/Y) = P(X)$

Multiplying the inequality (1) by $P(Y)$ and summing over Y^* we obtain

$$-\sum_{X \in X^*} \sum_{Y \in Y^*} P(Y) \cdot P(X/Y) \log P(X/Y) - \sum_{X \in X^*} \sum_{Y \in Y^*} P(Y) \cdot P(X/Y) \cdot \log P(X)$$

that is,

$$\sum_{X \in X^*} \sum_{Y \in Y^*} P(X, Y) \cdot \log P(X/Y) - \sum_{X \in X^*} \sum_{Y \in Y^*} P(X, Y) \cdot \log P(X) = - \sum_{X \in X^*} P(X) \log P(X)$$

Therefore $H(X^*/Y^*) \leq H(X^*)$ with equality if and only if $P(X/Y) = P(X)$ i.e. if and only if X^* and Y^* are independent orthogonal partitions.

When $H(Y^*) \geq H(X^*)$ an argument identical to the preceding one may be used to prove the corresponding result.

Theorem 3

Let X^* and Y^* be two orthogonal partitions with fuzzy events. Then $i(X^*, Y^*) = 1$ if and only if there exists an invertible mapping g such that $Y^* = g(X^*)$.

Proof.

If there exists an invertible mapping g such that $Y^* = g(X^*)$ it follows that for all $Y \in Y^*$ there exists a fuzzy event X^i in X^* and essentially one, such that

$$P(Y/X^i) = 1 \quad \text{and} \quad P(Y/X^j) = 0 \quad \forall X^j \neq X^i, X^j \in X^*$$

Consequently,

$$H(Y^*/X^*) = 0$$

Therefore

$$i(X^*, Y^*) = 1$$

On the other hand, if $i(X^*, Y^*) = 1$ it is easy to prove that there exists an invertible mapping g such that $Y^* = g(X^*)$

Obviously the index $i(X^*, Y^*) = 1$ is a symmetric function of its arguments.

Theorem 4

Let X^* , Y^* and Z^* be three orthogonal partitions with fuzzy events associated with X , Y and Z respectively. Then ,
 $d(X^*, Y^*) + d(Y^*, Z^*) \geq d(X^*, Z^*)$ (2) where

$$d(X^*, Y^*) = \begin{cases} H(X^*/Y^*)/H(X^*) & \text{if } H(X^*) \geq H(Y^*) \\ H(Y^*/X^*)/H(Y^*) & \text{if } H(X^*) \leq H(Y^*) \end{cases}$$

(that is, $i(X^*, Y^*) = 1 - d(X^*, Y^*)$)

Proof.

Assume $H(X^*) \geq H(Z^*)$. In order to prove the inequality (2) we have to verify it for each of the three cases:

- a) $H(X^*) \geq H(Z^*) \geq H(Y^*)$
- b) $H(X^*) \geq H(Y^*) \geq H(Z^*)$
- c) $H(Y^*) \geq H(X^*) \geq H(Z^*)$

Case a)

$$d(X^*, Y^*) + d(Y^*, Z^*) = \frac{H(X^*/Y^*)}{H(X^*)} + \frac{H(Z^*/Y^*)}{H(Z^*)}$$

As $H(Y^*) + H(Z^*/Y^*) = H(Z^*) + H(Y^*/Z^*)$ and

$$H(Y^*) \leq H(Z^*)$$

we have

$$H(Z^*/Y^*) \geq H(Y^*/Z^*)$$

and hence

$$d(X^*, Y^*) + d(Y^*, Z^*) \geq \frac{H(X^*/Y^*)}{H(X^*)} + \frac{H(Y^*/Z^*)}{H(Z^*)} \geq \frac{H(X^*/Y^*)}{H(X^*)} + \frac{H(Y^*/Z^*)}{H(X^*)}$$

Furthermore

$$H(X^*/Y^*) + H(Y^*/Z^*) \geq H(X^*/Y^*, Z^*) + H(Y^*/Z^*) = H(X^*, Y^*/Z^*) \geq H(X^*/Z^*)$$

Therefore

$$d(X^*, Y^*) + d(Y^*, Z^*) \geq \frac{H(X^*/Z^*)}{H(X^*)} = d(X^*, Z^*)$$

Case b)

Following a similar argument, we have

$$\begin{aligned}
 d(X^*, Y^*) + d(Y^*, Z^*) &= \frac{H(X^*/Z^*)}{H(X^*)} + \frac{H(Y^*/Z^*)}{H(Y^*)} \geq \frac{H(X^*/Y^*)}{H(X^*)} + \frac{H(Y^*/Z^*)}{H(X^*)} \geq \\
 &\geq \frac{H(X^*/Z^*)}{H(X^*)} = d(X^*, Z^*)
 \end{aligned}$$

Case c)

As in the previous cases, we can immediately verify that

$$\begin{aligned}
 d(X^*, Y^*) + d(Y^*, Z^*) &= \frac{H(Y^*/X^*)}{H(Y^*)} + \frac{H(Y^*/Z^*)}{H(Y^*)} = \frac{H(Y^*/X^*) + H(Y^*/Z^*)}{H(Y^*)} \geq \\
 &\geq \frac{H(X^*/Z^*) + H(Y^*) - H(X^*)}{H(Y^*)} = 1 - \frac{H(X^*)}{H(Y^*)} \left(1 - \frac{H(X^*/Z^*)}{H(X^*)}\right) \geq 1 - \left(1 - \frac{H(X^*/Z^*)}{H(X^*)}\right) = \\
 &= \frac{H(X^*/Z^*)}{H(X^*)} = d(X^*, Z^*)
 \end{aligned}$$

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