MEASURES OF FUZZINESS OF FUZZY CLUSTERS

Radko Mesiar, Slávka Bodjanová Bratislava, Czechoslovakia

#### Introduction

Fuzzy clustering algorithms yield fuzzy partitions as clustering solutions for partitioning finite data sets. One of the ways for evaluation of clustering results is to measure the amount of fuzziness of fuzzy partitions. Some measures of fuzziness of fuzzy clusters are given in [1], [2]. The aim of this paper is to develope an axiomatic framework for the measures of fuzziness of fuzzy partitions and to demonstrate a simple mean for their constructing.

#### 1. Preliminaries

Let  $X = \{x_1, x_2, \dots, x_n\}$  be a finite set of vectors in p-dimensional Euclidian space; let us denote the class of all fuzzy sets built on X by L(X). L(X) can be partially ordered by relation  $\leq_s$  called "sharpened" ( De Luca and Termini [3]) defined for all  $u, v \in L(X)$  as follows:

$$\mathbf{v} \leq_{\mathbf{S}} \mathbf{u} \quad \text{iff } \mathbf{v}(\mathbf{x}) \leq \mathbf{u}(\mathbf{x}) < \frac{1}{2} \quad \text{and} \quad \mathbf{v}(\mathbf{x}) \geq \mathbf{u}(\mathbf{x}) > \frac{1}{2} \quad (1.1)$$

De Luca and Termini [3] introduced for every fuzzy set  $u \in L(X)$  a measure d(u) of its "fuzziness". This measure satisfies the following conditions:

(1) 
$$d(u) = 0$$
 iff for each  $x \in X$ :  $u(x) = 0$  or  $u(x) = 1$  (1.2a)

(2) 
$$d(u)$$
 is maximum iff for each xeX:  $u(x) = \frac{1}{2}$  (1.2b)

(3) if 
$$u,v \in L(X)$$
 are such that  $v \leq s$  u then  $d(v) \leq d(u)$  (1.2c)

Let us denote the usual vector space of real  $k \times n$  matrices by  $V_{km}$ . Partitions of X are defined in [2] as follows:

$$P_{k} = \left\{ U \in V_{kn}; u_{ij} \in \{0,1\} \text{ for all } i,j; \sum_{i} u_{ij} = 1 \text{ for all } j; \sum_{j} u_{ij} > 0 \text{ for all } i \right\},$$
(1.3)

$$P_{fk} = \left\{ \begin{array}{l} U \in V_{kn}; \ u_{ij} \in \langle 0, 1 \rangle \text{ for all } i, j; \ \sum_{i} u_{ij} = 1 \text{ for all } j; \\ \sum_{j} u_{ij} > 0 \text{ for all } i \end{array} \right\}.$$
 (1.4)

We call  $P_k$  the hard k-partition space and  $P_{fk}$  the fuzzy k-partition space associated with X. Beadek [2] used so called classification entropy as a measure of fuzziness of  $U \in P_{fk}$ 

$$H(U) = -\frac{1}{n} \sum_{i} \sum_{j} h(u_{ij}) , \qquad (1.5)$$

where  $h(u_{ij}) = u_{ij} \cdot \log_a u_{ij}$  for  $u_{ij} > 0$ , h(0) = 0, a is a real constant, a > 1.

Backer [1] proposed many quantitative partitioning characterizations. Some of them are based upon the measures of fuzziness of fuzzy sets (by De Luca and Termini), e.g. for  $U \in P_{fk}$ :

$$\varphi(U) = 1 - \frac{2}{k-1} \sum_{r=1}^{k-1} \sum_{s=r+1}^{k} d(u_r \cap u_s)$$
 (1.6)

letting  $d(u_r \cap u_s)$  be a measure of fuzziness contained in intersection of fuzzy clusters  $u_r$ ,  $u_s$ .

# 2. Sharpened fuzzy partitions

Definition 2.1.

Consider  $U, W \in P_{fk}$ . W is a sharpened version of U denoted by  $W \triangleleft U$  if and only if

$$\mathbf{w}_{ij} \leq \mathbf{u}_{ij} < \frac{1}{k} \quad \text{and} \quad \mathbf{w}_{ij} \geq \mathbf{u}_{ij} > \frac{1}{k} \quad .$$
 (2.1)

It is easy to show that relation of satisfies the following properties:

- i) Pfk is partially ordered by
- ii)  $U = \begin{bmatrix} \frac{1}{k} \end{bmatrix}$  is a minimal element of  $P_{fk}$
- iii) for each  $U \in P_k$ : if  $V \notin U$  them V = U.

Theorem 2.1.

Partition  $U \in P_{fk}$  has a sharpened version  $V \in P_{fk} - \{U\}$  iff there exists  $u_{i,j} \in \{0, \frac{1}{k}\}$ .

Theorem 2.2.

Consider  $U, W \in P_{fk}$ ,  $W \leq U$ . Then for j=1,...,n:

i) 
$$\sum_{i} \left( u_{i,j} - \frac{1}{k} \right)^{2} \leq \sum_{i} \left( w_{i,j} - \frac{1}{k} \right)^{2}$$
 (2.2)

$$ii) \sum_{\mathbf{r}=1}^{\mathbf{k}-1} \sum_{\mathbf{s}=\mathbf{r}+1}^{\mathbf{k}} |\mathbf{u}_{\mathbf{r}j} - \mathbf{u}_{\mathbf{s}j}| \sum_{\mathbf{r}=1}^{\mathbf{k}-1} \sum_{\mathbf{s}=\mathbf{r}+1}^{\mathbf{k}} |\mathbf{w}_{\mathbf{r}j} - \mathbf{w}_{\mathbf{s}j}| \qquad (2.3)$$

## 3. Measures of fuzziness of fuzzy partitions

## Definition 3.1.

Consider a real nonnegative function  $\mathcal{L}: P_{fk} \to \mathbb{R}$ . This function is called a measure of fuzziness of partitions from Pfk iff the fellowing properties held true:

Pl 
$$\psi(U) = \psi(U^{(p)})$$
, where  $U^{(p)}$  is any permutation of U, (3.1a)

$$2 \varphi(\mathbf{U}) = \mathbf{0} \quad \text{iff } \mathbf{U} \in \mathbf{P}_{\mathbf{k}}, \tag{3.1b}$$

P2 
$$\varphi(U) = \emptyset$$
 iff  $U \in P_k$ , (3.1b)  
P3  $\psi(U)$  is maximum iff  $U = \begin{bmatrix} 1 \\ k \end{bmatrix}$ , (3.1c)  
P4 if  $W \not\setminus U$  then  $\varphi(W) \leq \varphi(U)$ .

P4 if 
$$W \not\setminus U$$
 then  $\varphi(W) \leq \varphi(U)$ . (3.1d)

# Theorem 3.1.

Let f be a real function on  $\langle 0,1 \rangle$  satisfying one of two next conditions i), ii) and the condition iii):

i) 
$$A = \inf \frac{f(y) - f(x)}{y - x} \ge \sup \frac{f(s) - f(r)}{s - r} = B$$
,  
 $0 \le x \le y \le \frac{1}{k}$   $\frac{1}{k} \le r \le s \le 1$ 

ii) 
$$C = \sup \frac{f(y) - f(x)}{y - x} \le \inf \frac{f(s) - f(r)}{s - r} = D$$
,  
 $0 \le x \le y \le \frac{1}{k}$   $\frac{1}{k} \le r \le s \le 1$ 

iii) 
$$(A - B)^2 + (C - D)^2 \neq 0$$
.

Then there exist constants 
$$\mathcal{L}_{j}$$
 so that
$$\mathcal{L}_{j} = \mathcal{L}_{j} \sum_{j} f(u_{ij}) + \mathcal{L}_{j} \qquad (3.2)$$

is a measure of fuzziness of fuzzy partitions from  $P_{fk}$  and  $\varphi(U)=1$  for  $U=\begin{bmatrix} \frac{1}{k} \end{bmatrix}$ .

### Proof:

We prove the case  $A \ge B$ .

Let V \( U. Then for j=1,...,n we have

$$\sum_{\mathbf{i}} (\mathbf{f}(\mathbf{u}_{\mathbf{i}\mathbf{j}}) - \mathbf{f}(\mathbf{v}_{\mathbf{i}\mathbf{j}})) = \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \geq \frac{1}{k}} (\mathbf{f}(\mathbf{u}_{\mathbf{i}\mathbf{j}}) - \mathbf{f}(\mathbf{v}_{\mathbf{i}\mathbf{j}})) + \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} < \frac{1}{k}} (\mathbf{f}(\mathbf{u}_{\mathbf{i}\mathbf{j}}) - \mathbf{f}(\mathbf{v}_{\mathbf{i}\mathbf{j}})) \ge \\
\ge A \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + B \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \geq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) = \\
= \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + B \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \geq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) = \\
= \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + B \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \geq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) = \\
= \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + B \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \geq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) = \\
= \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + B \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) = \\
= \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + B \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) = \\
= \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + B \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) = \\
= \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + B \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) = \\
= \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + B \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) = \\
= \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + B \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) = \\
= \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + B \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + B \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} \leq \frac{1}{k}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) = \\
= \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + B \cdot \sum_{\mathbf{v}_{\mathbf{i}\mathbf{j}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) = \\
= \sum_$$

$$= B \cdot \sum_{\mathbf{i}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) + (A - B) \cdot \sum_{\mathbf{i}\mathbf{j}} (\mathbf{u}_{\mathbf{i}\mathbf{j}} - \mathbf{v}_{\mathbf{i}\mathbf{j}}) \ge 0,$$

$$\mathbf{v}_{\mathbf{i}\mathbf{j}} < \frac{1}{k}$$
so that 
$$\sum_{\mathbf{i}} \sum_{\mathbf{j}} (\mathbf{v}_{\mathbf{i}\mathbf{j}}) \le \sum_{\mathbf{i}} \sum_{\mathbf{j}} f(\mathbf{u}_{\mathbf{i}\mathbf{j}}).$$

Now it is enough to assure  $\mathcal{L}(U) = 1$  for  $U = \left[\frac{1}{k}\right]$  and  $\mathcal{L}(W) = 0$  for  $W \in P_k$ , i.e.

$$\angle \cdot \mathbf{n} \cdot \mathbf{k} \cdot \mathbf{f}(\frac{1}{\mathbf{k}}) + \beta = 1$$

and  $\alpha \cdot m \cdot (f(1) + (k-1) \cdot f(0)) + \beta = 0$ .

This system has a solution iff  $k \cdot f(\frac{1}{k}) \neq f(1) + (k-1) \cdot f(0)$ . If A > B, then we have:

$$k \cdot f(\frac{1}{k}) - f(1) - (k-1) \cdot f(0) \ge (k-1) \cdot A \cdot \frac{1}{k} - B \cdot (1 - \frac{1}{k}) = \frac{k-1}{k} \cdot (A-B) > 0.$$

If A = B, then C  $\neq$  D. From C  $\geq$  A = B  $\geq$  D we get C > A or B > D. Let e.g. C > A. Then  $f(\frac{1}{k})$ - f(0) > A.  $\frac{1}{k}$ , so that

 $k \cdot f(\frac{1}{k}) - (k-1) \cdot f(0) - f(1) > (k-1) \cdot A \cdot \frac{1}{k} - B \cdot (1 - \frac{1}{k}) = 0$ 

Then the constants  $\omega$ ,  $\beta$  are as follows:

$$\beta = [-f(1) - (k-1) \cdot f(0)] \cdot \lambda \cdot n$$

## Corollary 1.

Let  $f:\langle 0,1\rangle \longrightarrow \mathbb{R}$  be any nonconstant increasing on  $\langle 0,\frac{1}{k}\rangle$  and decreasing on  $\langle \frac{1}{k},1\rangle$  function. Then there exist constants  $\mathscr{L}$ ,  $\mathscr{L}$  so that the function defined by equation (3.2) satisfies the properties Pl - P4 in Definition 3.1.

## Cerollary 2.

Let  $f:(0,1) \rightarrow \mathbb{R}$  be any convex or any concave nonlinear function. Then there exist constants  $\omega$ , 3 so that the function defined by equation (3.2) satisfies the properties Pl - P4 in Definition 3.1.

#### Examples:

The following functions are measures of fuzziness of  $U \in P_{pk}$ :

$$\mathbf{a}) \mathcal{L}_{1}(\mathbf{U}) = 1 - \frac{\mathbf{k}}{\mathbf{n}(\mathbf{k}-1)} \sum_{i} \sum_{j} (\mathbf{u}_{i,j} - \frac{1}{\mathbf{k}})^{2}$$
 (3.3)

$$\mathbf{b}$$
) $\frac{\varphi}{2}(\mathbf{U}) = 1 - \frac{1}{n} \sum_{i} \sum_{j} \mathbf{u}_{ij}^{2}$  (3.4)

e) 
$$\mathcal{L}_3(U) = H(U)$$
 (see (1.5)) (3.5)

d) 
$$\mathcal{G}_{4}(\mathbf{U}) = 1 - \frac{\mathbf{k}}{2\mathbf{k}(\mathbf{k}-1)} \sum_{\mathbf{i}} \sum_{\mathbf{j}} |\mathbf{u}_{\mathbf{i}\mathbf{j}} - \frac{1}{\mathbf{k}}|$$
 (3.6)

e) 
$$f_5(U) = n - \sum_{j=1}^{max} u_{ij}$$
 (3.7)

The functions  $\frac{1}{1} - \frac{1}{4}$  are constructed by Theorem 3.1. Function  $\frac{1}{5}$  shows that there are also another means for constructing of measures of fuzziness of fuzzy partitions satisfying the properties Pl - P4 in Definition 3.1.

### References:

- [1] Backer, E.: Cluster analysis by optimal decomposition of induced fuzzy sets, Delftse Universitaire Press, Delft 1978
- [2] Bezdek, J.C.: Patterm Recognition with Fuzzy Objective Function Algorithms, Plenum Press, New York 1981
- [3] De Luca, A., Termini, S.: A definition of a nonprobabilistic entropy in the setting of fuzzy sets theory, Inform. and Control 20 (1972) 301 312