Volume 10, No. 2, March-April 2019



International Journal of Advanced Research in Computer Science

RESEARCH PAPER

Available Online at www.ijarcs.info

SOME PROPERTIES ABOUT SMOOTHING, ROUGHEN THE VALUES **OF THE INDEX ATTRIBUTE ON THE DECISION BLOCK**

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Abstract: The report proposed and demonstrated some properties about smoothing, roughen the values of the condition index attribute or decision index attributeon the decision block and on the slice of the decisionblock. Every time the condition equivalence class or decision equivalence classon the decision block have been smoothed or roughened then they will partial pullulate or pullulate smoothing, roughing the corresponding classon the slice. From the results foundof the smoothing, roughening the condition equivalence class or decision equivalence classpartial pullulate or pullulate on the slice then the incremental calculation of the support matrices on the slice will be simplerand therefore faster than recalculating these matrices when smoothing, roughing the values of the condition index attribute or decision index attribute.

Keywords: Decision block, smoothing, roughen, index attribute.

I. INTRODUCTION

The study to search for decision laws on the decision table by assessing the measures of decision laws as well as incremental approaches, determining decision laws ... has been studied by many groups of authors, such as in [7], [8], ... On the other hand, when the decision table is expanded into a decision block, then the study, proposing a model and algorithm to detect decision laws on the decision block has been studied by the authors as in [4], [5], [6]. However, the proposed models and algorithms when smoothing and roughen the values of index attributes on the decision block have not been studied until now. The purpose of this paper is to study the some properties about smoothing, roughen the values of the condition index attribute or decision index attribute on the decision block and on the slice of the decision block. From the results found of the smoothing, roughening the condition equivalence class or decision equivalence class partial pullulate or pullulate on the slice then the incremental calculation of the support matrices on the slice will be simpler and therefore faster than recalculating these matrices when smoothing, roughing the values of the condition index attribute or decision index attribute.

II - THE BASIC CONCEPT

II.1 The block, slice of the block **Definition II.1** [1]

Let $R = (id; A_1, A_2, ..., A_n)$ is a finite set of elements, where id is non-empty finite index set, A_i (i=1.. n) is the attribute. Each attribute A_i (i=1.. n) there is a corresponding value domain dom(Ai). A block r on R, denoted r(R) consists of a finite number of elements that each element is a family of mappings from the index set id to the value domain of the attributes A_i (i = 1.. n).

 $t \in r(R) \Leftrightarrow t = \{t^i : id \rightarrow dom(A_i)\}_{i=1..n}$.

The block is denoted by r(R) or $r(id; A_1, A_2,..., A_n)$, sometime without fear of confusion we simply denoted r. Definition II.2 [2],[3]

Let $R = (id; A_1, A_2, ..., A_n)$, r(R) is a block over R. For each $x \in id$ we denoted $r(R_x)$ is a block with $R_x = (\{x\}; A_1,$

 $\begin{array}{l} A_{2},...,A_{n}) \text{ such that:} \\ t_{x} \in r(R_{x}) \Leftrightarrow t_{x} = \{t^{i}_{x} = t^{i}| \}_{i=1..n}, t \in r(R), t = \{t^{i} : id \rightarrow dom(A_{i})\}_{i=1..n}, x \end{array}$

where $t_{x}^{i}(x) = t^{i}(x), i = 1..n$.

Then $r(R_x)$ is called a slice of the block r(R) at point x, sometimes we denotedr_x.

Here, for simplicity we use symbols: $x^{(i)} = (x; A_i); id^{(i)} = \{x^{(i)} | x \in id\}.$

We call $x^{(i)}$ (x \in id, i = 1..n) are the index attributes of the block scheme $R = (id; A_1, A_2, \dots, A_n)$.

II.2Information block

DefinitionII.3[4]:Let block scheme $R = (id; A_1, A_2, ..., A_n)$, ris a block over R. Then, the information block is a tuples of fourelements IB = (U, A, V, f) with U is a set of objects of r

called space objects, $A = \bigcup_{i=1}^{n} id^{(i)}$ is the set of index attributes of the object, $V = \bigcup_{x^{(i)} \in A}^{n} V_{x^{(i)}}$, $V_{x^{(i)}}$ is the set of

values of the objects corresponding to the index attribute $x^{(i)}$, fis an information function $UxA \rightarrow V$ satisfy: $\forall u \in U$, $\forall x^{(i)} \in A \text{ we have } f(u, x^{(i)}) \in V_{x^{(i)}}.$

We call $f(u, x^{(i)})$ is the value of the object u at the index attribute x⁽ⁱ

If V contains missing values in at least one index attribute $x^{(1)} \in A$ then we call IB is inadequate information block. In contrast IB is a complete information block, or simply IB is an information block.

DefinitionII.4[4]:Let block scheme $R = (id; A_1, A_2, ..., A_n), r$ is a block over R, r_x is the slice of the block r at the point $x \in id$. Then the slice of the information block at x is a tuples of four elements $IB_x = (U, A_x, V_x, f_x)$ with U is a set of objects of *r* called space objects, $A_x = \bigcup_{i=1}^n x^{(i)}$ is the set of the index attributes of the object on the slice at x, $V_x = \bigcup_{x^{(i)} \in A_r} V_{x^{(i)}}$, $V_{x^{(i)}}$

is the set of values of the objects corresponding to the index attribute $x^{(i)}$, f_x is an information function $UxA_x \rightarrow V_x$ satisfy: $\forall u \in U, \forall x^{(i)} \in A_x$ we have $f(u, x^{(i)}) \in V_{x^{(i)}}$.

II.3 Relationships are indistinguishable

DefinitionII.5[5]

Letinformation block IB = (U, A, V, f). Then for each index attribute set $P \subseteq A$ we define an equivalence relation, signIND(P) defined as follows:

$$ND(P) = \{(u,v) \in Ux U \mid \forall x^{(i)} \in P: f(u,x^{(i)}) = f(v,x^{(i)})\},\$$

and called non-discriminatory relations:

From the definition we have:

$$IND(P) = \bigcap_{x^{(i)} \in P} IND(x^{(i)}).$$

RelationIND(P) divide U into equivalence classes,, constitutes a subdivision of U, sign U/IND(P) or simply U/P.

With each $u \in U$, the equivalence class contains u in relation IND(P), sign [u]_P is defined as follows:

$$[u]_{P} = \{v \in U \mid (u,v) \in IND(P)\}.$$

By this definition we see: two elements $u, v \in U$ belonging to the same equivalence class if and only if they have the same value on every index attribute in P.

DefinitionII.6[5]

Letinformation block IB = (U, A, V, f), $P, Q \subseteq A$ is the set of index attributes, $U/P = \{P_1, P_2, ..., P_m\}$, $U/Q = \{Q_1, Q_2, ..., Q_n\}$ is the partition generated by P, Q respectively. Then we say partition by Q is more coarse than partition by P, or partition by P is smoother than partition by Q if and only if:

 $\forall P_i \in U/P, \exists Q_j \in U/Q: P_i \subseteq Q_j, i = 1..m, j = 1..n.$

II.4 Decision block

DefinitionII.7[5]

Letinformation block IB = (U, A, V, f) with U is the space of objects, $A = \bigcup_{i=1}^{n} J_{id}$ Suppose A is divided into two sets C and

D such that:
$$C \stackrel{i=1}{=} \bigcup_{i=1,x \in id}^{\kappa} x^{(i)}$$
, $D = \bigcup_{i=k+1,x \in id}^{n} x^{(i)}$

then information block IB is called the decision block and denoted by $DB=(U,C\cup D,V,f)$, with C is the set of conditional index attributes and D is the set of decision index attributes.

From the definition of the decision block, we see: $C \cup D = A$, $C \cap D = \emptyset$,

We can denote the decision block simply by: $DB=(U, C\cup D)$.

DefinitionII.8[5]:Let decision $blockDB = (U, C \cup D, V, f)$, with *C* is the set of conditional index attributes and *D* is the set of decision index attributes. Then the slice of the block decides at x ($x \in id$) is a tuples of four elements $DB_x = (U, C^x \cup D^x, V_x, f_x)$ with *U* is the set of objects of *r*,rcalled the space of objects

$$C^{x} = , D^{x} \bigcup_{i=1}^{k} x^{(i)} , A_{x} = \bigoplus_{i=k+1}^{n} D^{(i)},$$

$$V_{x} = , is the every of value x of the objects$$

corresponding to the index attribute $x^{(i)}$, f_x is an information

function $U x A_x \rightarrow V_x$ satisfy: $\forall u \in U, \forall x^{(i)} \in A_x$ we have:

 $f(u, x^{(i)}) \in V_{x^{(i)}}.$

Comment:

Let decision $blockDB = (U, C \cup D, V, f)$. Then, if $id = \{x\}$, the decision block DB degenerate into the decision table as known.

When studying the decision block, people want to find the decisive laws from there. In these decision laws, the conditional part corresponds to the conditional indexattribute, the conclusions will correspond to the decision index attributes.

The decision laws found in the decision block are divided into two categories:

i) The lawsare correct on the block.

ii) The laws are correct on each particular slice of the block.

II.5 The decision laws

DefinitionII.9[5]

Let decision blockDB=(U, C \cup D),with U is the space of objects:

$$C =, D = \bigcup_{i=k+1, x \in id}^{n} mdx^{(i)}C^{x} =, \qquad \bigcup_{i=1, x \in id}^{k} x^{(i)} \qquad \qquad \bigcup_{i=1}^{k} x^{(i)}$$
$$D^{x} =, x \notin \bigcup_{i=k+1}^{n} x^{(i)}$$

Then:

$$U/C = \{C_1, C_2, ..., C_m\}, \quad U/C^{\alpha} = \{C_{x1}, C_{x2}, ..., C_{xt_x}\}, \\ U/D = \{D_1, D_2, ..., D_k\}, \quad U/D^{\alpha} = \{D_{x1}, D_{x2}, ..., D_{xh_x}\},$$

correspondingly, the partitions are generated by C, C^x, D, D^x . A decision law on a block is denoted by:

$$C_i \rightarrow D_j$$
, $i=1..m$, $j=1..k$,
and on the slice at point x is denoted by:

$$C_{xi} \rightarrow D_{xj}, i=1..t_x, j=1..h_x$$

PropositionII.1 [5]

Let decision blockDB=(U, $C \cup D$), with U is the space of objects:

$$C = \bigcup_{i=1,x \in id}^{k} x^{(i)}, D = \bigcup_{i=k+1,x \in id}^{n} x^{(i)}, and C^{\alpha} = \bigcup_{i=1}^{k} x^{(i)}, D^{\alpha} = \bigcup_{i=k+1}^{n} x^{(i)}, x \in id.$$
$$U/C = \{C_1, C_2, \dots, C_m\}, U/C^{\alpha} = \{C_1, C_2, \dots, C_n\},$$

$$U/D = \{D_1, D_2, \dots, D_k\}, U/D^x = \{D_{x1}, D_{x2}, \dots, D_{xh_x}\},\$$

Then: $\forall C_i \in U/C$, $\forall D_i \in U/D$ we have:

$$C_{i} = \bigcap_{x \in id} C_{xp_{x}}, D_{j} = \bigcap_{x \in id} D_{xq_{x}} \text{ with } p_{x} \in \{1, 2, ..., t_{x}\}, q_{x} \in \{1, 2, ..., h_{x}\}.$$

DefinitionII.10[5]

Let decision blockDB=(U,C \cup D), C_i \in U/C, D_j \in U/D, C_{xp}

 $\in U/C^x$, $D_{xq} \in U/D^x$, i = 1..m, j = 1..k, $p \in \{1, 2, ..., t_x\}$, $q \in \{1, 2, ..., t_x\}$, $x \in id$. Then, support, accuracy and coverage of decision law $C_i \rightarrow D_j$ on the block are:

- Support: Sup(C_i, D_j) = $|C_i \cap D_j|$,

- Accuracy: Acc(C_i,D_j) = $\frac{|C_i \cap D_j|}{|C_i|}$, - CoverageCov(C_i,D_j) = $\frac{|C_i \cap D_j|}{|D_j|}$,

and for decision $lawC_{xp} \rightarrow D_{xq}$ on the slice of the block at point x is:

- Support:Sup(C_{xp}, D_{xq}) = $|C_{xp} \cap D_{xq}|$, - Accuracy: Acc(C_{xp}, D_{xq}) = $\frac{|C_{xp} \cap D_{xq}|}{|C_{xp}|}$, - Coverage: Cov(C_{xp}, D_{xq}) = $\frac{|C_{xp} \cap D_{xq}|}{|D_{xq}|}$.

From this definition, we have:

$$\begin{split} 0 \leq & \operatorname{Acc}(\mathbf{C}_{i},\mathbf{D}_{j}) \leq 1, \ 0 \leq & \operatorname{Acc}(\mathbf{C}_{xp},\mathbf{D}_{xq}) \leq 1, \\ \sum_{j=1}^{n} Acc(C_{i},D_{j}) = 1, \quad \sum_{q=1}^{h_{x}} Acc(C_{xp},D_{xq}) = 1, \\ 0 \leq & \operatorname{Cov}(\mathbf{C}_{i},\mathbf{D}_{j}) \leq 1, \ 0 \leq & \operatorname{Cov}(\mathbf{C}_{xp},\mathbf{D}_{xq}) \leq 1, \\ \sum_{i=1}^{m} Cov(C_{i},D_{j}) = 1, \quad \sum_{p=1}^{t_{x}} Cov(C_{xp},D_{xq}) = 1. \end{split}$$

We can represent the measure of the decision laws on the block in the form of the following measurement matrices:

- Matrix of support:

$$Sup(C, D) = Sup(C_i, D_j)_{mxk} = = \begin{pmatrix} Sup(C_1, D_1) & \dots & Sup(C_1, D_k) \\ & \dots & \\ Sup(C_m, D_1) & \dots & Sup(C_m, D_k) \end{pmatrix}$$

- Matrix of Accuracy: AccC, D) = $Acc(C_i, D_i)_{mxk}$

$$\mathbf{c}(\mathbf{C}_{i}, \mathbf{D}_{j})_{mxk} = \begin{pmatrix} Acc(C_{1}, D_{1}) & \dots & Acc(C_{1}, D_{k}) \\ & \dots & \\ Acc(C_{m}, D_{1}) & \dots & Acc(C_{m}, D_{k}) \end{pmatrix}$$

- Matrix of coverage: CovC, D) = $Cov(C_i,D_i)_{mxk}=$

$$= \begin{pmatrix} Cov(C_1, D_1) & \dots & Cov(C_1, D_k) \\ & \dots & & \\ Cov(C_m, D_1) & \dots & Cov(C_m, D_k) \end{pmatrix}$$

With the decision laws on the slices of the blocks, we also have the same support, accuracy, and coverage matrix.

DefinitionII.11[5]

Let decision $blockDB=(U,C \cup D)$, $C_i \in U/C$, $D_j \in U/D$ is the conditional equivalence class and decision equivalence class generated by C, Dcorresponding, $C_i \rightarrow D_j$ is the decision lawon the block DB, i = 1..m, j = 1..k.

- $lfAcc(C_i \rightarrow D_j) = l$ then $C_i \rightarrow D_j$ is called certain decision law.
- If $0 \le Acc(C_i \rightarrow D_j) \le 1$ then $C_i \rightarrow D_j$ is called uncertain decision law.

PropositionII.2 [5]

Let decision blockDB=(U, C \cup D),with U is the space of objects:

$$C = \bigcup_{i=1,x\in id}^{k} \mathbf{x}^{(i)}, D = \bigcup_{i=k+1,x\in id}^{n} \mathbf{x}^{(i)}.$$

Then $\forall C_i \in U/C$, $\forall D_j \in U/D$, (i = 1...m, j = 1..n) we have: i) $\operatorname{Acc}(C_i \rightarrow D_j) = \underbrace{\operatorname{Sup}(C_i, D_j)}_{q=1}$, © 2015-19, IJARCS All Right $\sum_{q=1}^{n} \underbrace{\operatorname{Sup}(C_i, D_q)}_{q=1}$,

ii)
$$\operatorname{Cov}(C_i \rightarrow D_j) = \frac{\operatorname{Sup}(C_i, D_j)}{\sum_{p=1}^m \operatorname{Sup}(C_p, D_j)}$$

DefinitionII.12[5]

Let decision blockDB=(U,C \cup D), $C_i \in U/C$, $D_j \in U/D$, i = 1..m, j=1..kis the conditional equivalence class and decision equivalence class generated by C,Dcorresponding; α , β are two given thresholds (α , $\beta \in (0,1)$). If $Acc(C_i, D_j)$ $\geq \alpha and Cov(C_i, D_j) \geq \beta$ then we call $C_i \rightarrow D_j$ is the decision lawmeaning.

DefinitionII.13[5]

Let decision blockDB=(U,C \cup D,V,f),with U is the space of objects, $a \in C \cup D$, V_a is the set of existing values of the index attribute a. Suppose $Z = \{x_s \in U \mid f(x_s, a) = z\}$ is the set of objects whose z value is on the index attribute a. If Z is partitioned into two sets W and Y such that: $Z = W \cup Y$, $W \cap Y = \emptyset$ with $W = \{x_p \in U \mid f(x_p, a) = w, w \notin V_a\}$, $Y = =\{x_q \in U \mid f(x_q, a) = y, y \notin V_a\}$, then we say the z value of the index attribute a is smoothed to two new values w and y.

DefinitionII.14[5]

Let decision blockDB=(U,C \cup D,V,f),with U is the space of objects, $a \in C \cup D$, V_a is the set of existing values of the index attribute a. Suppose $f(x_p, a) = w$, $f(x_q, a) = y$ are respectively the values of x_p , x_q on the index attribute a $(p \neq q)$. If at any one time we have: $f(x_p, a) = f(x_q, a) = z$, $(z \notin V_a)$ thenwe say the two values w, y of a are roughened to the new value z.

Theorem II.1[6]

Let decision blockDB= (U, C \cup D, V, f),with U is the space of objects, $a \in C \cup D$, V_{a} is the set of existing values of the index attribute a.Then, two equivalent classes E_p , E_q (E_p , $E_q \in U/E$, $E \in \{C, D\}$) is made rough into new equivalent class E_s if and only if $\forall a_i \neq a$: $f(E_p, a_i) = f(E_q, a_i)$.

TheoremII.2[6]

Let decision blockDB= (U, C \cup D, V, f), with U is the space of objects, $a \in C \cup D$, V_a is the set of existing values of the index attribute a. Then, equivalent class E_s ($E_s \in U/E$, $E \in \{C,D\}$) smoothed into two new equivalents classes E_p , E_q if and only if we can put: $f(E_p, a) = w$, $f(E_q, a) = y$ và $E_p \cup E_q = E_s$, $w, y \notin V_a, w \neq y$.

Theorem II.3 [6]

Let decision blockDB=(U, C \cup D),with U is the space of objects:

$$C = \bigcup_{i=1,x \in id}^{k} x^{(i)}, D = \bigcup_{i=k+1,x \in id}^{n} x^{(i)}, and C^{x} = \bigcup_{i=1}^{k} x^{(i)}, D^{x} = \bigcup_{i=k+1}^{n} x^{(i)}, x \in id.$$

$$U/C = \{C_1, C_2, ..., C_m\}, \quad U/C^x = \{C_{x1}, C_{x2}, ..., C_{xt_x}\}, U/D = \{D_1, D_2, ..., D_k\}, \quad U/D^x = \{D_{x1}, D_{x2}, ..., D_{xh_x}\},$$

 α , β are two given thresholds (α , $\beta \in (0,1)$). Suppose that if $C_i \rightarrow D_j$ is the decision lawmeaning on the decision block then it is also the decision lawmeaningon any slice of the decision block at $x \in id$.

III. RESEARCH RESULTS

III.1 Smoothing, roughening the conditional equivalenteclases on the decision block and on the slice. PropositionIII.1

Let decision blockDB= (U, C \cup D, V, f), $a=x^{(i)} \in C$, V_{a} is the set of existing values of the conditional index attribute a, The z value of a is smoothed to two new values w and y.

$$C=, D=, \inf_{i=1,x\in id} x^{(i)} \qquad \bigcup_{i=k+1,x\in id} x^{(i)}$$
$$C^{x}=, D^{x}=, \bigvee_{i=1}^{k} x^{(i)}$$
$$\bigcup_{i=k+1}^{n} x^{(i)}$$

 $U/C = \{C_1, C_2, ..., C_m\}, U/C^x = \{C_{x1}, C_{x2}, ..., C_{xt_x}\}, U/D = \{D_1, D_2, ..., D_k\}, U/D^x = \{D_{x1}, D_{x2}, ..., D_{xh_x}\}, U/D^x = \{D_{x1}, D_{x1}, D_{x2}, ..., D_{xh_x}\}, U/D^x = \{D_{x1}, D_{x1}, D_{x2}, ..., D_{xh_x}\}, U/D^x = \{D_{x1}, D_{x1}, D_{$

Suppose that if the conditional equivalence class $C_s \in U/C$, $(f(C_{s},a)=z)$ smoothed into two new conditional equivalents classes C_p , $C_q(f(C_p,a)=w$, $f(C_q,a)=y$, with $w,y \notin V_a$) thenon the slicer_x, exists equivalence class C_{xi} satisfy: $C_s \subseteq C_{xi}$, also smoothed into two new conditional equivalents classes $C_{xi'}$ and $C_{xi''}$ satisfy: $C_p \subseteq C_{xi'}$, $C_q \subseteq C_{xi''}$ ($f(C_{xi},a)=w$, $f(C_{xi''},a)=y$). We say on the slice r_x then C_{xi} is smoothed sympathetic partiallysmoothed into two new conditional equivalents classes $C_{xi'}$ and $C_{xi''}$ by the smoothing of C_s into two new conditional equivalents classes C_p , C_q .

Prove

Assuming we have: $C_s \in U/C$, $(f(C_s, a)=z)$ smoothed into two new conditional equivalents $classesC_p, C_q$ $(f(C_p, a)=w,$ $f(C_q, a)=y, with w, y \notin V_a$). Because $C_s \in U/C$, applying the results of clause I.1 we have: $C_s = \bigcap_{x \in id} C_{xp_x}$, thence inferred

 $\exists C_{xi} \in U/C^{x} satisfy: C_{s} \subseteq C_{xi}. \text{ On the other hand, by } C_{s} smoothed into two conditional equivalents <math>classesC_{p}$ and $C_{q} so according to theorem I.2 we have: C_{s} = C_{p} \cup C_{q} \Rightarrow C_{p}, C_{q} \subseteq C_{xi} with f(C_{p}, a) = w, f(C_{q}, a) = y.$

Finally, we assign each element $u \in C_{xi} \setminus C_s$ at the index attribute a either w or y then we have a subdivision of C_{xi} into two new conditional equivalents classes $C_{xi'}$ and $C_{xi'}$ satisfy: $f(C_{xi'}, a) = w$, $f(C_{xi''}, a) = yand C_{xi} = C_{xi'} \cup C_{xi''}$.

The result is on the slice r_x then the conditional equivalent class C_{xi} satisfy: $C_{s} \subseteq C_{xi}$, also smoothed into two conditional equivalents classes $C_{xi'}$ and $C_{xi''}$ satisfy: $C_p \subseteq C_{xi'}$, $C_q \subseteq C_{xi''}$ ($f(C_{xi'}, a) = w$, $f(C_{xi''}, a) = y$) and $C_{xi} = C_{xi'} \cup C_{xi''}$.

PropositionIII.2

Let decision $blockDB=(U,C \cup D)$, $a=x^{(i)} \in C$, V_a is the set of existing values of the conditional index attribute a, The z value of a is smoothed to two new values w and y.

$$C = , \underbrace{D}_{i=1,x\in id}^{k} \mathbf{x}_{ind}^{(i)} C^{x} = , \bigcup_{i=k+1,x\in id}^{n} \mathbf{x}^{(i)} \qquad \bigcup_{i=1}^{k} \mathbf{x}^{(i)}$$

$$D^{x} = , \mathbf{x} \in \underbrace{d}_{i=k+1}^{n} \mathbf{x}^{(i)}$$

$$U/C = \{C_{1}, C_{2}, ..., C_{m}\}, \quad U/C^{x} = \{C_{x1}, C_{x2}, ..., C_{xt_{x}}\},$$

$$U/D = \{D_{1}, D_{2}, ..., D_{k}\}, \quad U/D^{x} = \{D_{x1}, D_{x2}, ..., D_{xh_{n}}\},$$

 $C_s \in U/C$, $C_{xi} \in U/C^x$, $C_s \subseteq C_{xi}$, $D_{xj} \in U/D^x$, s=1..m, $i=1..t_{\infty}$, $j=1..h_x$. Suppose that if $C_s(f(C_s,a)=z)$ smoothed into two conditional equivalents classes $C_pandC_q(f(C_p,a)=w)$, $f(C_q,a)=y$ and on the slicer_x, C_{xi} is smoothed sympathetic partially into two new conditional equivalents classes $C_{xi'}$ and $C_{xi''}$ then:

$$C_{xi} = C_{xi'} \cup C_{xi''},$$

ii)
$$\forall D_{xj} \in U/D^{x}$$
: $Sup(C_{xh}, D_{xj}) = Sup(C_{xi}, D_{xj}) + Sup(C_{xi''}, D_{xj})$, with $j=1, 2, ..., h_{x}$.

Prove

- i) From the smoothing of the conditional equivalence class C_{xi} we have: $C_{xi} = C_{xi} \cup C_{xi}$.
- ii) Assuming we have: C_{xi} is smoothed sympathetic partially into two new conditional equivalents classes C_{xi} and C_{xi} .

 \Rightarrow C_{xi}= C_{xi}, \cup C_{xi}, and C_{xi}, \cap C_{xi}, = Ø..

Other way: $\forall D_{xj} \in U/D^x$: $Sup(C_{xi}, D_{xj}) = |C_{xi} \cap D_{xj}| = |(C_{xi'} \cup C_{xi''}) \cap D_{xj}| = |(C_{xi'} \cap D_{xj}) \cup (C_{xi''} \cap D_{xj})|.$ We have: $C_{xi'} \cap C_{xi''} = \emptyset \Rightarrow (C_{xi'} \cap D_{xj}) \cap (C_{xi''} \cap D_{xj}) = \emptyset.$ Inferred: $Sup(C_{xi}, D_{xj}) = |(C_{xi'} \cap D_{xj}) \cup (C_{xi''} \cap D_{xj})| = |(C_{xi'} \cap D_{xj})| + |(C_{xi''} \cap D_{xj})| = Sup(C_{xi'}, D_{xj}) + Sup(C_{xi''}, D_{xj}).$ So we infer: $\forall D_{xj} \in U/D^x$: $Sup(C_{xi}, D_{xj}) = Sup(C_{xi'}, D_{xj}) + Sup(C_{xi''}, D_{xj}) + Sup(C_{xi''}, D_{xj}) + Sup(C_{xi''}, D_{xj}) + Sup(C_{xi''}, D_{xj}), with j = 1, 2, ..., h_x.$

From this result we see: row corresponding to the conditional equivalence $classC_{xi}$ in the support matrix for slicer_x will be split into two new lines corresponding to two new conditional equivalents $classesC_{xi'}$ and $C_{xi''}$.

Therefore, to calculate the value of the elements of these two new rows in the support matrix with slice r_x then we first calculate the values $Sup(C_{xi}, D_{xj})$ with $j=1,2,...,h_x$. From there, we infer the values $Sup(C_{xi''}, D_{xj})$ is the subtraction between $Sup(C_{xi}, D_{xj})$ and $Sup(C_{xi'}, D_{xj})$ with $j=1,2,...,h_x$.

PropositionIII.3

i=

Let decision blockDB= (U, C \cup D, V, f), $a=x^{(i)} \in C$, V_{a} is the set of existing values of the conditional index attribute a, the w and y values of a are roughened to the new value z.

$$C = \bigcup_{i=1,x \in id}^{k} x^{(i)}, \quad D = \bigcup_{i=k+1,x \in id}^{n} x^{(i)}, \quad and C^{x} = \bigcup_{i=1}^{k} x^{(i)}, \quad D^{x} = \sum_{i=1}^{n} x^{(i)}, \quad x \in id.$$
$$U/C = \{C_{1}, C_{2}, ..., C_{m}\}, \quad U/C^{x} = \{C_{x1}, C_{x2}, ..., C_{xt_{x}}\},$$

$$U/D = \{D_1, D_2, ..., D_k\}, U/D^x = \{D_{x1}, D_{x2}, ..., D_{xh}\},\$$

Suppose, if two conditional equivalents $classesC_p, C_q \in U/C$, $(f(C_p, a) = w, f(C_q, a) = y)$ is made rough into new conditional equivalent class $C_s \in U/C$ ($f(C_s, a) = z$) then on the slicer_xexists two conditional equivalents $classesC_{xi}$, C_{xj} satisfy: $C_p \subseteq C_{xi}$, $C_q \subseteq C_{xj}$, also is made rough into new conditional equivalent class C_x satisfy: $C_s \subseteq C_x$.

We say on the slice r_x then the two conditional equivalents classes C_{xi} , C_{xj} is made rough sympathetic into C_{xk} by the roughening of two conditional equivalents $classes C_p$, C_q to C_s . Prove

Assuming we have: C_p , $C_q \in U/C$, $(f(C_p, a) = w, f(C_q, a) = y)$, applying the results of proposition I.1 we infer on the slicer_xexists two conditionalequivalents classes C_{xi} . C_{xj} satisfy: $C_p \subseteq C_{xi}$, $C_q \subseteq C_{xj}$. From there we have: f(u,a) = wwith $u \in C_p \subseteq C_{xi} \Rightarrow f(C_{xi}, a) = w$. In the same way we also have: f(u,a) = y with $u \in C_q \subseteq C_{xj} \Rightarrow f(C_{xj}, a) = y$.

On the other hand, assuming we have: two conditional equivalents $classesC_p, C_q \in U/C$ is made rough into new conditional equivalent $classC_s \in U/C$, according to the results of theorem I.1 then we have:

 $\forall a_i \neq a, a_i \in C: f(C_p, a_i) = f(C_a, a_i) \Longrightarrow \forall a_i \neq a, a_i \in C^x:$ $f(C_{p}, a_{i}) = f(C_{a}, a_{i})(1)$ In slices r_xthenwe have: $C_n \subseteq C_{xi} \in U/C^{x} \Longrightarrow \forall a_i \neq a, \ a_j \in C^{x}: f(C_p, a_j) = f(C_{xi}, a_j)(2)$ Same, we also have: $C_q \subseteq C_{xj} \in U/C^x \Longrightarrow \forall a_j \neq a, a_j \in C^x: f(C_q, a_j) = f(C_{xj}, a_j)(3)$ From (1), (2) and (3) we infer:

 $\forall a_i \neq a, a_i \in C^x$: $f(C_{xi}, a_i) = f(C_{xi}, a_i)$.

Therefore, apply the necessary and sufficient conditions in the statement of the theorem I.1, we havetwo conditional equivalents classes C_{xi} , C_{xj} is made rough C_{xk} by the roughening sympathetic into of*two* conditional equivalents $classes C_p, C_q$ to C_s .

From the nature of the rough wor conditional equivalents $classes C_{xi}$, C_{xj} to C_{xk} we have: work two

 $C_{xk} = (C_{xi} \cup C_{xj}) \supseteq (C_p \cup C_q) = C_s.$ From that: $C_s \subseteq C_{xk}$.

PropositionIII.4

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Let decision $blockDB = (U, C \cup D)$, $a = x^{(i)} \in C$, V_a is the set of existing values of the conditional index attribute a, the w and y values of a are roughened to the new value z

$$C = , \bigoplus_{i=1,x \in id}^{k} , ax^{(i)} C^{x} = \bigcup_{i=k+1,x \in id}^{k} x_{i}^{n} f^{(i)} , x^{(i)}$$

$$D^{x} = \bigcup_{i=k+1}^{n} x^{(i)} , x \in id.$$

$$U/C = \{C_{1}, C_{2}, ..., C_{m}\}, U/C^{x} = \{C_{x1}, C_{x2}, ..., C_{xt_{x}}\},$$

$$U/D = (D, D, D, D) = U/D^{x} = (D, D, D)$$

 $U/D = \{D_1, D_2, ..., D_k\}, U/D^x = \{D_{x1}, D_{x2}, ..., D_{xh_x}\},\$

 C_p , $C_q \in U/C$, $(f(C_p, a) = w$, $f(C_q, a) = y$), $D_{xh} \in U/D^x$, $h = 1..h_x$. Suppose, if C_p , C_q is made rough into new conditional equivalent class C_s , $(f(C_s, a)=z)$ and on the slice r_x two conditional equivalents classes C_{xi} , $C_{xj}(C_p \subseteq C_{xi})$ $C_a \subseteq C_{xi}$) is made rough sympathetic into C_{xk} then:

i)
$$C_{xi} \cup C_{xj} = C_{xk}$$

ii) $\forall D_{xh} \in U/D^x$: $Sup(C_{xi}, D_{xh}) + Sup(C_{xj}, D_{xh}) = Sup(C_{xk}, D_{xh})$, $v \circ ih = 1, 2, ..., h_x$.

Prove

i) Suppose we have: $x \in C_{xi} \cup C_{xi} \Rightarrow x \in C_{xi}$ or $x \in C_{xi}$. If $x \in C_{xi}$ thenfrom the two conditional equivalents $classes C_{xi}$, C_{xi} is made rough into conditional equivalent class $C_{xk} \Rightarrow f(x,a) =$ $f(C_{xi},a)=f(C_{xk},a)=z.$

On the other hand, applying the results of theorem 2.1 we have $\forall a_i \neq a$: $f(C_{xi}, a_i) = f(C_{xi}, a_i) = f(C_{xk}, a_i) \Longrightarrow f(\mathbf{x}, \mathbf{a}_i) = f(C_{xi}, a_i)$ $= f(C_{xi}, a_i) = f(C_{xk}, a_i) \implies x \in C_{xk}$. Totally similar, when x $\in C_{xi}$ we also prove that $x \in C_{xk}$.

So inference: $(C_{xi} \cup C_{xj}) \subseteq C_{xk}$. (5)

On the contrary, suppose $x \in C_{xk}$, because C_{xi} and C_{xi} is made rough into C_{xk} applying the results of theorem 2.1 we have: $\forall a_i \neq a: f(C_{xi}, a_i) = f(C_{xi}, a_i) = f(C_{xk}, a_i) \implies f(x, a_i) = f(C_{xi}, a_i)$ $f(C_{xi},a_i)$. On the other hand, $because x \in C_{xk} \Rightarrow f(x,a)=z$ but z *is made rough from* w and $y \Rightarrow f(x,a)=w$ or f(x,a)=y.

- If
$$f(x,a)=w \Rightarrow f(x,a)=f(C_{xi},a)=w \Rightarrow x \in C_{xi}$$
.
- If $f(x,a)=y \Rightarrow f(x,a)=f(C_{xj},a)=y \Rightarrow x \in C_{xj}$.
So $x \in C_{xi}$ or $x \in C_{xj} \Rightarrow x \in C_{xi} \cup C_{xj}$.
Therefore, from $x \in C_{xk} \Rightarrow x \in C_{xi} \cup C_{xj}$.

So: $C_{xk} \subseteq (C_{xi} \cup C_{xi})$ (6)

Combined (5) and (6) we have:
$$C_{xi} \cup C_{xj} = C_{xk}$$

ii) BecauseC_{xi}, C_{xj}arethe conditionalequivalents classes, so we have: $C_{xi} \cap C_{xj} = \emptyset$.

On the other hand: $\forall D_{xh} \in U/D^x$: $Sup(C_{xk}, D_{xh}) = |C_{xk} \cap D_{xh}| =$ $|(C_{xi}\cup C_{xj})\cap D_{xh}| = |(C_{xi}\cap D_{xh})\cup (C_{xj}\cap D_{xh})|.$

We have: $C_{xi} \cap C_{xi} = \emptyset \Longrightarrow (C_{xi} \cap D_{xh}) \cap (C_{xi} \cap D_{xh}) = \emptyset$.

 $Sup(C_{xk}, D_{xh}) = |(C_{xi} \cap D_{xh}) \cup (C_{xi} \cap D_{xh})| =$ Inferred: $|(\mathbf{C}_{xi} \cap \mathbf{D}_{xh})| + |(\mathbf{C}_{xj} \cap \mathbf{D}_{xh})| = \operatorname{Sup}(\mathbf{C}_{xi}, \mathbf{D}_{xh}) + \operatorname{Sup}(\mathbf{C}_{xj}, \mathbf{D}_{xh}).$

 $\forall D_{xh} \in U/D^x$: $Sup(C_{xi}, D_{xh}) = Sup(C_{xi}, D_{xh}) +$ So inference: $\operatorname{Sup}(C_{xj}, D_{xh})$ with $h=1, 2, \dots, h_x$.

Thus, we see two rows of matrix of support on the slice r_x, corresponding to the two conditional equivalents classesC_{xi}, C_{xi} is combined into a new row corresponding to the conditional equivalent class Cxk. The value of each element of the new line corresponds to C_{xk}is the total value of two elements of two lines corresponding to C_{xi}andC_{xi}.

III.2 Smoothing, rougheningthedecisionequivalenceclassesonthedecision block and ontheslice.

PropositionIII.5

Let decision blockDB= (U, C \cup D, V, f), $a=x^{(i)} \in D$, V_a is the set of existing values of the decision index attribute a, the zvalue of a is smoothed to two new values w and y.

$$C = \bigcup_{i=1,x \in id}^{k} x^{(i)}, D = \bigcup_{i=k+1,x \in id}^{n} x^{(i)}, \text{ and } C^{x} = \bigcup_{i=1}^{k} x^{(i)}, D^{x} = \bigcup_{i=k+1}^{n} x^{(i)}, x \in id.$$
$$U/C = \{C_{1}, C_{2}, ..., C_{m}\}, U/C^{x} = \{C_{x1}, C_{x2}, ..., C_{xt_{x}}\}, U/D = \{D_{1}, D_{2}, ..., D_{k}\}, U/D^{x} = \{D_{x1}, D_{x2}, ..., D_{xh_{x}}\},$$

Suppose that if decision equivalent $classD_s \in U/D$ $(f(D_s,a)=z)$ smoothed into two decisionequivalents classes D_p, D_q (f(D_p, a)=w, f(D_q, a)=y, with w, y $\notin V_q$) then on the slice r_x , exists decision equivalence class D_{xi} satisfy: $D_s \subseteq D_{xi}$, also smoothed into two new decision equivalents $classesD_{xi}$, and $D_{xi'}$ satisfy: $D_n \subset D_{xi'}$ $D_a \subset D_{xi''}$ $(f(D_{xi'}, a) = w, f(D_{xi''}, a) = y)$. We say on the slice r_x then decision equivalent class D_{xi} is smoothed sympathetic partially into two new *decisionequivalents classes* $D_{xi'}$ *and* $D_{xi''}$ by the smoothing of D_s into two new decisionequivalents classes D_p , D_q .

Proving this clause is similar to the proof of the proposition II.1.

PropositionIII.6

Let decision $blockDB = (U, C \cup D)$, $a = x^{(i)} \in D$, V_a is the set of existing values of the decision index attribute a, the z value of a is smoothed to two new values w and y.

$$C = \bigcup_{i=1,x\in id}^{k} x^{(i)}, D = \bigcup_{i=k+1,x\in id}^{n} x^{(i)}, \text{ and } C^{x} = \bigcup_{i=1}^{k} x^{(i)}, D^{x} = \bigcup_{i=1}^{n} x^{(i)}, x \in id.$$
$$U/C = \{C_{1}, C_{2}, ..., C_{m}\}, U/C^{x} = \{C_{x1}, C_{x2}, ..., C_{xt_{x}}\},$$
$$U/D = \{D_{1}, D_{2}, ..., D_{k}\}, U/D^{x} = \{D_{x1}, D_{x2}, ..., D_{xh_{y}}\},$$

 $D_s \in U/D$, $D_{xi} \in U/D^x$, $D_s \subseteq D_{xi}$, $C_{xj} \in U/C^x$, s=1..k, $i=1..h_x$, Suppose that if decision equivalent $j=1..t_x$ $classD_s(f(D_s,a)=z)$ smoothed into two decisionequivalents classes D_p , D_q (f(D_p , a)=w, f(D_q , a)=y and on the slice \mathbf{r}_x , \mathbf{D}_{xi} is sympathetic partially into smoothed two new decision equivalents $classes D_{xi'}$ and $D_{xi''}$ then:

 $i) D_{xi} = D_{xi'} \cup D_{xi''},$

ii)
$$\forall C_{xj} \in U/C^{\alpha}: Sup(C_{xj}, D_{xi}) = Sup(C_{xj}, D_{xi'}) + Sup(C_{xj}, D_{xi''}), with \ j=1, 2, ..., t_{x}.$$

Prove

- i) From the smoothing of the decision equivalent $classD_{xi}$ we see that: $D_{xi} = D_{xi'} \cup D_{xi''}$.
- ii) Assuming we have: D_{xi} is smoothed sympathetic partially into two new decisionequivalents classes $D_{xi'}$ and $D_{xi''}$

 \Rightarrow D_{xi} = D_{xi}, \cup D_{xi}, and D_{xi}, \cap D_{xi}, $= \emptyset$..

Other way: $\forall C_{xj} \in U/C^{x}$: $Sup(C_{xj}, D_{xi}) = |C_{xj} \cap D_{xi}| = |C_{xj} \cap (D_{xi}) \cup (D_{xi}) \cup (C_{xj} \cap D_{xi})| = |(C_{xj} \cap D_{xi}) \cup (C_{xj} \cap D_{xi})|.$

We have:
$$D_{xi'} \cap D_{xi''} = \emptyset \Rightarrow (C_{xj} \cap D_{xi}) \cap (C_{xj} \cap D_{xi''}) = \emptyset.$$

Come on: $Sup(C_{xj}, D_{xi}) = |(C_{xj} \cap D_{xi}) \cup (C_{xj} \cap D_{xi})| = |(C_{xj} \cap D_{xi})| + |(C_{xj} \cap D_{xi})| = Sup(C_{xj}, D_{xi}) + Sup(C_{xj}, D_{xi})|$ So we infer: $\forall C_{xj} \in U/C^x$: $Sup(C_{xj}, D_{xi}) = Sup(C_{xj}, D_{xi}) + U(C_{xj}, D_{xi})$

So we infer: $VC_{xj} \in O/C$: $Sup(C_{xj}, D_{xi}) = Sup(C_{xj}, D_{xi})$ $Sup(C_{xj}, D_{xi})$, with $j=1, 2, ..., t_x$.

From this result we see: columncorresponding to the the decision equivalence class D_{xi} in the support matrix for slice r_x will be split into two new columns corresponding to two new decision equivalents classes D_{xi} and D_{xi} .

Therefore, to calculate the value of the elements of these two new *columns* in the support matrix with slice r_x then we first calculate the values $Sup(C_{xj}, D_{xi})$ with $j=1,2,...,t_x$. From there, we infer the values $Sup(C_{xj}, D_{xi})$ is the subtraction between $Sup(C_{xj}, D_{xi})$ and $Sup(C_{xj}, D_{xi})$ with $j=1,2,...,t_x$.

PropositionIII.7

Let decision blockDB= (U, C \cup D, V, f), $a=x^{(i)} \in D$, V_a is the set of existing values of the decision index attribute a, the w and y values of a are roughened to the new value z.

$$C = \bigcup_{i=1,x\in id}^{k} x^{(i)}, D = \bigcup_{i=k+1,x\in id}^{n} x^{(i)}, and C^{x} = \bigcup_{i=1}^{k} x^{(i)},$$
$$D^{x} = \bigcup_{i=k+1}^{n} x^{(i)}, x \in id.$$
$$U/C = \{C_{1}, C_{2}, ..., C_{m}\}, U/C^{x} = \{C_{x1}, C_{x2}, ..., C_{xt_{x}}\},$$

 $\begin{array}{ll} U/D=\{D_1,D_2,...,D_k\}, & U/D^x=\{\mathbf{D}_{x1},D_{x2},...,D_{xh_x}\},\\ Suppose, & if two decision equivalents classes D_p, D_q,\\ (f(D_p,a)=w, & f(D_q,a)=y) & is made rough into new \end{array}$

 $((D_p, d) - w, f(D_q, d) - y)$ is made rough into new decisionequivalent class $D_s \in U/D$ ($f(D_s, a) = z$) then on the slice r_x exists two decisionequivalents classes D_{xi} , D_{xj} satisfy: $D_p \subseteq D_{xi}$, $D_q \subseteq D_{xj}$, also is made rough into new decision equivalent class D_x satisfy: $D_s \subseteq D_{xk}$.

We say on the slice r_x then two decision equivalents classes D_{xi} , D_{xj} is made rough sympathetic into D_{xk} by the roughening of the two decision equivalents classes D_p , D_q to decision equivalent class D_s .

Proving this clause is similar to the proof of the proposition II.3.

PropositionIII.8

Let decision $blockDB = (U, C \cup D)$, $a = x^{(i)} \in D$, V_a is the set of existing values of the decision index attribute a, the w and y values of a are roughened to the new value z.

$$C = \bigcup_{i=1,x \in id}^{k} x^{(i)}, D = \bigcup_{i=k+1,x \in id}^{n} x^{(i)}, \text{ and } C^{x} = \bigcup_{i=1}^{k} x^{(i)},$$

$$D^{x} = \bigcup_{i=k+1}^{n} x^{(i)}, x \in id.$$

$$U/C = \{C_{1}, C_{2}, ..., C_{m}\}, U/C^{x} = \{C_{x1}, C_{x2}, ..., C_{xt_{x}}\},$$

$$U/D = \{D_{1}, D_{2}, ..., D_{k}\}, U/D^{x} = \{D_{x1}, D_{x2}, ..., D_{xh_{x}}\},$$

 D_p , $D_q \in U/D$, $(f(D_p, a)=w, f(D_q, a)=y)$, $C_{xh} \in U/C^x$, $h=1..t_x$. Suppose, if two decisionequivalents classes D_p , D_q is made rough into new decision equivalent class D_s , $(f(D_s, a)=z)$ and on the slice r_x two decisionequivalents classes D_{xi} , D_{xj} $(D_p \subseteq D_{xi}, D_a \subseteq D_{xi})$ is made rough sympathetic into D_{xk} then:

i)
$$D_{xi} \cup D_{xj} = D_{xk}$$

ii) $\forall C_{xh} \in U/C^{x}$: $Sup(C_{xh}, D_{xi}) + Sup(C_{xh}, D_{xj}) =$
 $= Sup(C_{xh}, D_{xk})$, with $h=1, 2, ..., t_{x}$.

Prove

i) Suppose we have: $u \in D_{xi} \cup D_{xj} \Rightarrow u \in D_{xi}$ or $u \in D_{xj}$. If $u \in D_{xi}$ thenby two decision equivalence classes D_{xi} , D_{xj} is made rough intodecision equivalent class $D_{xk} \Rightarrow f(u,a) = f(D_{xi},a) = f(D_{xk},a) = z$.

On the other hand, apply the results of the theorem 2.1 we have $\forall a_r \neq a$: $f(D_{xi}, a_r) = f(D_{xj}, a_r) = f(D_{xk}, a_r) \Longrightarrow f(u, a_r)$ = $f(D_{xi}, a_r) = f(D_{xj}, a_r) = f(D_{xk}, a_r) \Longrightarrow u \in D_{xk}$. Completely similar, if $u \in D_{xi}$ then we also proved $u \in D_{xk}$.

So inference:
$$(D_{xi} \cup D_{xj}) \subseteq D_{xk}$$
. (7)

On the contrary, suppose $u \in D_{xk}$, because $D_{xi}andD_{xj}is$ made rough into D_{xk} should apply the results of the theorem 2.1 we have: $\forall a_r \neq a$: $f(D_{xi}, a_r) = f(D_{xj}, a_r) = f(D_{xk}, a_r) \Rightarrow f(u, a_r) = f(D_{xi}, a_r) = f(D_{xi}, a_r)$. On the other hand, by $u \in D_{xk} \Rightarrow f(u,a) = z$ but z made rough from w and $y \Rightarrow f(u,a) = w$ or f(u,a) = y.

- If $f(u,a)=w \Rightarrow f(u,a)=f(D_{xi},a)=w \Rightarrow u \in D_{xi}$.
- If $f(u,a)=y \Rightarrow f(u,a)=f(D_{xj},a)=y \Rightarrow u \in D_{xj}$.

 $Sou \in D_{xi}oru \in D_{xj} \Rightarrow u \in D_{xi} \cup D_{xj}.$

Therefore, from $u \in D_{xk} \Rightarrow u \in D_{xi} \cup D_{xj}$.

So: $D_{xk} \subseteq (D_{xi} \cup D_{xj})$. (8) Combined (7) and (8) we have: $D_{xi} \cup D_{xi} = D_{xk}$.

ii) Because D_{xi} , D_{xj} are decision equivalence classes, so we have: $D_{xi} \cap D_{xj} = \emptyset$.

On the other hand: $\forall C_{xh} \in U/C^{\alpha}$: $Sup(C_{xh}, D_{xk}) = |C_{xh} \cap D_{xk}| = |(D_{xi} \cup D_{xi}) \cap C_{xh}| = |(D_{xi} \cap C_{xh}) \cup (D_{xi} \cap C_{xh})|.$

We have: $D_{xi} \cap D_{xi} = \emptyset \Longrightarrow (D_{xi} \cap C_{xh}) \cap (D_{xi} \cap C_{xh}) = \emptyset$.

So inference: $\forall C_{xh} \in U/C^x$: $Sup(C_{xh}, D_{xi}) + Sup(C_{xh}, D_{xj}) = Sup(C_{xh}, D_{xk})$, with $h=1, 2, ..., t_x$.

Thus, we see two columns of the support matrix on the slicer_x corresponds to two decision equivalence classes D_{xi} , D_{xj} is made rough sympathetic into a new column corresponding to the decision equivalent class D_{xk} . The value of each element of the new column corresponds to D_{xk} is the total value of two elements of two columns corresponding to two decision equivalence classes D_{xi} and D_{xj} .

IV. CONCLUSIONS

From the initial results on the decision block, the paper proposes and demonstrates some of the results of the relationship between roughing, smoothing the values of conditional attributes or decisionsfor conditionalequivalents classesordecision equivalence classeson the decision blocks and on the slices. The smoothing of conditionalequivalents classes or decision equivalence classes on the decision blockshave a sympathetic partially the smoothing of conditionalequivalents classes or decision equivalence classesrespectively on the slice. The roughening of conditionalequivalents classes or decision equivalence classes on the decision blockshave a sympathetic the roughening of conditionalequivalents classes or decision equivalence classes on the slice. From these results, calculation of support matrix on the slicesame is define as the calculation of the support matrix on the block whenthe smoothing, roughening of conditionalequivalents classes or decision equivalence classes.

In special cases, the index set $id = \{x\}$, the information blocks degenerate into information systemsthem these results coincide with the results reported by many authors for the information system. On the basis of these results we can study the reverse relationship between slices of information block with that block itself, in case the objects of the information block are changed..., some other results may be considered in individual cases of information blocks..., it adds the theoretical results of the exploitation of decision rules on information blocks.

V. ACKNOWLEDGEMENTS

Finally, the authors thank the teachers, leaders of the Faculty of Information Technology and the Management Board of the Hanoi Pedagogical University 2for creating favorable conditions for us to work and study. This research is funded by Hanoi Pedagogical University 2 (HPU2).

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