

# Construction of a Unified Model for Formal Contexts and Formal Decision Contexts

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## Abstract

*This paper mainly constructs a unified model for formal contexts, consistent formal decision contexts and inconsistent formal decision contexts based on object oriented concept lattices, which is called a consistent approximate representation space. Congruence relations are first introduced into formal contexts and then relationships between congruence relations and the corresponding object oriented concept lattices are developed. Finally, consistent approximate representation space is defined for a formal context, which is a quadruple  $(U, A, \mathfrak{R}, R')$ . It is verified that if we give  $R'$  different meanings, then we obtain the corresponding formal context, consistent formal decision context and inconsistent formal decision context with respect to object oriented concept lattices. Therefore, the quadruple  $(U, A, \mathfrak{R}, R')$  is a unified model for formal contexts and formal decision contexts.*

**Keywords:** *Object oriented concept lattice, Formal context, Congruence relation, Approximate representation space*

## 1. Introduction

Formal concept analysis, proposed by Wille in 1982 [1], is an effective tool for data analysis, knowledge representation and information management. At present, many efforts have been made to construction of concept lattice [2-5], pruning of concept lattice [6], acquisition of rules [4, 5], relationship with rough set [7-13], and applications [6, 14]. In [12], Yao compared the theory of rough sets and formal concept analysis in a common framework based on formal contexts. Particularly object oriented concept lattices and property oriented concept lattices were also constructed in [12]. In [15], an approach to knowledge reduction in the object oriented concept lattices and the property oriented concept lattices were presented which can keep all extents and their original hierarchy in a formal context. Wang *et al.*, [16] provided another approach to attribute reduction in the object oriented concept lattices and the property oriented concept lattices, which only required preserving all extents of meet irreducible elements. Relatively knowledge reduction in classical concept lattices has much more research results. Ganter *et al.*, [2] proposed reducible attribute and reducible object from the viewpoint of shortening lines or rows. Zhang *et al.*, [17-20] presented an attribute reduction approach to find minimal attribute sets which can determine all extents and their original hierarchy in a formal context. Wang *et al.*, [21] provided an approach to attribute reduction based on meet irreducible elements. Wu *et al.*, [22] studied attribute reduction in formal contexts from the viewpoint of keeping granular structure of concept lattices. Liu *et al.* [23] showed an efficient post-processing method to prune redundant rules by virtue of the

property of Galois connection, which inherently constrains rules with respect to objects. Mi *et al.*, [24] formulated a Boolean approach to calculating all reducts of a formal context via the use of discernibility function. In [25], a rule acquisition oriented framework of knowledge reduction was proposed for real decision formal contexts and a corresponding reduction method was formulated by constructing a discernibility matrix and its associated Boolean function. In [26], based on fuzzy K-means clustering, Kumar and Srinivas proposed a method to reduce the size of the concept lattices by employing corresponding object-attribute matrix. In [27], approaches to attribute reduction in a formal context and in a consistent formal decision context were proposed by preserving congruence relation classes. Wang *et al.*, [28] studied notions and approaches to attribute reduction in an inconsistent formal decision context based on congruence relation classes.

Since databases obtained from real world are usually very complicated, knowledge reduction in formal concept analysis becomes more complex. On the other hand, through analyzing the above methods of knowledge reduction, we know that there are different places existing in knowledge reduction of different contexts even if we use the same type of method. And it makes the application of formal concept analysis more difficult. In order to solve this problem, we construct a model to unify the formal contexts, consistent formal decision contexts and inconsistent formal decision contexts, which is a quadruple  $(U, A, \mathfrak{R}, R')$ . It is verified that if we give  $R'$  different meanings, then we obtain corresponding formal context, consistent formal decision context and inconsistent formal decision context. Therefore, the quadruple  $(U, A, \mathfrak{R}, R')$  is a unified model for formal contexts and formal decision contexts. Therefore, we can consider methods of knowledge reduction or rule acquisition of consistent approximate representation space as a whole, which can reduce the complexities of different contexts.

The paper is organized as follows. Section 2 recalls preliminaries on formal concept analysis and dependence space. Section 3 constructs a model named consistent approximate representation space and then studies relationships between consistent approximate representation spaces and the corresponding formal contexts. Finally, Section 4 concludes the paper.

## 2. Preliminaries

Some basic notions and properties about formal concept analysis and dependence space are introduced in this section.

### 2.1. Basic notions about Formal Concept Analysis

**Definition 1** ([2]) *A formal context  $(U, A, I)$  consists of two sets  $U$  and  $A$ , and a relation  $I \subseteq U \times A$ . The elements of  $U$  are called objects and the elements of  $A$  are called attributes of the formal context.*

For  $X \subseteq U$  and  $B \subseteq A$ , Yao [12] defined two closure operators as follows:

$$X^{\square} = \{a \in A \mid \forall x \in U, xRa \Rightarrow x \in X\},$$

$$X^{\diamond} = \{a \in A \mid \exists x \in U, xRa \wedge x \in X\},$$

$$B^{\square} = \{x \in U \mid \forall a \in A, xRa \Rightarrow x \in B\},$$

$$B^{\diamond} = \{x \in U \mid \exists a \in A, xRa \wedge a \in B\}.$$

**Definition 2** ([2]) *Let  $(U, A, I)$  be a formal context. The formal context  $(U, B, I_B)$  is called a subcontext of  $(U, A, I)$ , where  $I_B = I \cap (U \times B)$  for any  $B \subseteq A$ .*

Let  $\square^B, \diamond^B$  stand for the operator in the subcontext  $(U, B, I_B)$  for any  $B \subseteq A$ . Clearly, for  $X \subseteq U$ ,  $X^{\square^B} = X^{\square^A} \cap B$ ,  $X^{\diamond^B} = X^{\diamond^A} \cap B$  and  $X^{\square^A} = X^{\square}$ ,  $X^{\diamond^A} = X^{\diamond}$ .

**Definition 3** ([12]) An object oriented concept of a formal context  $(U, A, I)$  is a pair  $(X, B)$  with  $X \subseteq U, B \subseteq A$ ,  $X = B^{\diamond}$  and  $B = X^{\square}$ . We call  $X$  the extent and  $B$  the intent of the object oriented concept  $(X, B)$ .

The concepts of a formal context  $(U, A, I)$  are partially ordered by  $(X_1, B_1) \leq (X_2, B_2)$  if and only if (iff for short)  $X_1 \subseteq X_2$  (iff  $B_2 \subseteq B_1$ ), where  $(X_1, B_1)$  and  $(X_2, B_2)$  are two object oriented concepts. The set of all object oriented concepts of  $(U, A, I)$  partially ordered in this way is denoted by  $L_o(U, A, I)$  and is called the object oriented concept lattice of the formal context  $(U, A, I)$ . The infimum and supremum are given by:

$$(X_1, B_1) \wedge (X_2, B_2) = ((B_1 \cap B_2)^{\diamond}, B_1 \cap B_2)$$

$$(X_1, B_1) \vee (X_2, B_2) = (X_1 \cup X_2, (X_1 \cup X_2)^{\square})$$

We denote the extent set of  $(U, A, I)$  by  $L_{ou}(U, A, I) = \{X \mid (X, B) \in L_o(U, A, I)\}$ . It is evident that  $L_{ou}(U, D, I_D) \subseteq L_{ou}(U, A, I)$  for any  $D \subseteq A$ .

**Proposition 1** ([12]) Let  $(U, A, I)$  be a formal context,  $X, X_1, X_2$  be object sets, and  $B, B_1, B_2$  be attribute sets, then

$$(i) X_1 \subseteq X_2 \Rightarrow X_1^{\square} \subseteq X_2^{\square}, X_1^{\diamond} \subseteq X_2^{\diamond},$$

$$B_1 \subseteq B_2 \Rightarrow B_1^{\square} \subseteq B_2^{\square}, B_1^{\diamond} \subseteq B_2^{\diamond},$$

$$(ii) X^{\square\diamond} \subseteq X \subseteq X^{\diamond\square}, B^{\square\diamond} \subseteq B \subseteq B^{\diamond\square},$$

$$(iii) X^{\square\diamond\square} = X^{\square}, B^{\square\diamond\square} = B^{\square},$$

$$X^{\diamond\square\diamond} = X^{\diamond}, B^{\diamond\square\diamond} = B^{\diamond},$$

$$(iv) (X_1 \cap X_2)^{\square} = X_1^{\square} \cap X_2^{\square},$$

$$(X_1 \cup X_2)^{\diamond} = X_1^{\diamond} \cup X_2^{\diamond},$$

$$(B_1 \cap B_2)^{\square} = B_1^{\square} \cap B_2^{\square},$$

$$(B_1 \cup B_2)^{\diamond} = B_1^{\diamond} \cup B_2^{\diamond}.$$

## 2.2. Dependence Space based on a Formal Context

In [29], Novotny defined a congruence relation on the attribute power set  $P(A)$  and dependence space in information systems.

**Definition 4** ([29]) Let  $(U, A, F)$  be an information system.  $\kappa$  is an equivalence relation on  $P(A)$ . Then,  $\kappa$  is called a congruence relation on  $(P(A), \cup)$ , whenever it satisfies the following condition: if  $(B_1, C_1) \in \kappa, (B_2, C_2) \in \kappa$ , then  $(B_1 \cup B_2, C_1 \cup C_2) \in \kappa$ .

**Definition 5** ([29]) Let  $A$  be a finite nonempty set,  $\kappa$  a congruence relation on  $(P(A), \cup)$ . Then the ordered pair  $(A, \kappa)$  is said to be a dependence space.

### 3. Main Results

In this section, we first introduce congruence relations into formal contexts to obtain the relationships between congruence relations and the corresponding object oriented concept lattices. And then a unified model of formal contexts and formal decision contexts is constructed, which is called a consistent approximate representation space. It is proved that formal contexts and formal decision contexts are special cases of the consistent approximate representation space.

#### 3.1. Relationships between Congruence Relations and the Corresponding Object Oriented Concept Lattices

In this subsection, we first give the definition of a congruence relation based on formal contexts, and then obtain the relationships between congruence relations and the corresponding object oriented concept lattices.

Let  $(U, A, I)$  be a formal context. For  $B \subseteq A$ , we define a binary relation on the object power set  $P(U)$  as follows:

$$R^{\square B} = \{(X, Y) \in P(U) \times P(U) \mid X^{\square B} = Y^{\square B}\}.$$

It is obvious that for any  $B \subseteq A$ ,  $R^{\square B}$  is a congruence relation on  $(P(U), \cap)$  and  $(U, R^{\square B})$  is a dependence space according to Proposition 1. We then define  $[X]_{R^{\square B}} = \{Y \in P(U) \mid (X, Y) \in R^{\square B}\}$  the congruence class respect to  $X$ , and  $IN_{R^{\square B}}(X) = \cap \{Y \mid Y \in [X]_{R^{\square B}}\}$ .

**Lemma 1** *Let  $(U, A, I)$  be a formal context. For  $X, Y, Z \subseteq U$  and  $B \subseteq A$ , the following statements hold:*

- (1)  $(IN_{R^{\square B}}(X), X) \in R^{\square B}$ ,
- (2)  $IN_{R^{\square B}}(X)$  is an inner operator,
- (3) If  $X \subseteq Y \subseteq Z$  and  $(X, Z) \in R^{\square B}$ , then  $(X, Y) \in R^{\square B}$  and  $(Y, Z) \in R^{\square B}$ .

*Proof.* (1) Since  $(IN_{R^{\square B}}(X))^{\square B} = (\cap \{Y \mid Y \in [X]_{R^{\square B}}\})^{\square B} = \cap_{Y \in [X]_{R^{\square B}}} Y^{\square B} = X^{\square B}$  holds by

Proposition 1, we have  $(IN_{R^{\square B}}(X), X) \in R^{\square B}$ .

(2) In order to prove that  $IN_{R^{\square B}}(X)$  is an inner operator, we should prove that

- (a)  $IN_{R^{\square B}}(X) \subseteq X$  for any  $X \subseteq U$ ; (b) if  $X \subseteq Y$ , then  $IN_{R^{\square B}}(X) \subseteq IN_{R^{\square B}}(Y)$ ; and (c)

$$IN_{R^{\square B}}(X) = IN_{R^{\square B}}(IN_{R^{\square B}}(X))$$

Obviously, (a)  $IN_{R^{\square B}}(X) \subseteq X$  holds by the definition of  $IN_{R^{\square B}}(X)$ . (b) Since  $(IN_{R^{\square B}}(X), X) \in R^{\square B}$  and  $(IN_{R^{\square B}}(Y), Y) \in R^{\square B}$  hold for any  $X$  and  $Y$ , we have  $(IN_{R^{\square B}}(X) \cap IN_{R^{\square B}}(Y), X \cap Y) \in R^{\square B}$  according to Definition 4. Thus, if  $X \subseteq Y$ , then  $(IN_{R^{\square B}}(X) \cap IN_{R^{\square B}}(Y), X) \in R^{\square B}$  holds which shows that  $IN_{R^{\square B}}(X) \cap IN_{R^{\square B}}(Y) = IN_{R^{\square B}}(X)$ . Therefore,  $IN_{R^{\square B}}(X) \subseteq IN_{R^{\square B}}(Y)$ . (c) Finally, since  $(IN_{R^{\square B}}(X), X) \in R^{\square B}$  and  $(IN_{R^{\square B}}(X), IN_{R^{\square B}}(IN_{R^{\square B}}(X))) \in R^{\square B}$ , we have

$(X, IN_{R^{\square B}}(IN_{R^{\square B}}(X))) \in R^{\square B}$ , which leads to  $IN_{R^{\square B}}(X) \subseteq IN_{R^{\square B}}(IN_{R^{\square B}}(X))$ . So  $IN_{R^{\square B}}(X) = IN_{R^{\square B}}(IN_{R^{\square B}}(X))$  holds.

(3) Since  $(Y, Y) \in R^{\square B}$  and  $(X, Z) \in R^{\square B}$ , we have  $(X \cap Y, Z \cap Y) \in R^{\square B}$ . So if  $X \subseteq Y \subseteq Z$ , then we have  $(X, Y) \in R^{\square B}$ . Combining  $(X, Z) \in R^{\square B}$  and  $(X, Y) \in R^{\square B}$ , we have  $(Y, Z) \in R^{\square B}$ .

Lemma 1 shows that  $IN_{R^{\square B}}(X)$  is the minimum element in  $[X]_{R^{\square B}}$ . An object subset  $X \subseteq U$  is called  $IN_{R^{\square B}}$ -closed if  $IN_{R^{\square B}}(X) = X$ . The set of all  $IN_{R^{\square B}}$ -closed sets is denoted by  $\Pi_B$ .

**Lemma 2** Let  $(U, A, I)$  be a formal context. For  $X \subseteq U$  and  $B \subseteq A$ , we have

- (1)  $IN_{R^{\square B}}(X) = X^{\square B \diamond B}$ .
- (2)  $\Pi_B = L_{OU}(U, B, I_B)$ .
- (3)  $(IN_{R^{\square B}}(X), X^{\square B}) \in L_{OU}(U, B, I_B)$ .

*Proof.* (1) By Proposition 1 and Lemma 1, we have  $X^{\square B \diamond B} = (IN_{R^{\square B}}(X))^{\square B \diamond B} \subseteq IN_{R^{\square B}}(X)$ . Conversely,  $X^{\square B \diamond B \diamond B} = X^{\square B}$  by Proposition 1. So  $X^{\square B \diamond B} \in [X]_{R^{\square B}}$ . Then,  $IN_{R^{\square B}}(X) \subseteq X^{\square B \diamond B}$ . Therefore,  $IN_{R^{\square B}}(X) = X^{\square B \diamond B}$ .

(2) For any  $X \in L_{OU}(U, B, I_B)$ , it is clear that  $X = X^{\square B \diamond B}$ . Then, by (1), we have  $IN_{R^{\square B}}(X) = X^{\square B \diamond B} = X$ , i.e.,  $X$  is a  $IN_{R^{\square B}}$ -closed set. Therefore,  $L_{OU}(U, B, I_B) \subseteq \Pi_B$ . Furthermore, for any  $X \in \Pi_B$ , we have  $IN_{R^{\square B}}(X) = X$ . Using (1) again,  $IN_{R^{\square B}}(X) = X^{\square B \diamond B} = X$ , that is  $X \in L_{OU}(U, B, I_B)$ . Therefore we conclude  $\Pi_B \subseteq L_{OU}(U, B, I_B)$ .

(3) follows immediately from (1) and Lemma 1.

Lemma 2 says that all the  $IN_{R^{\square B}}$ -closed sets form the extent set of  $(U, B, I_B)$  exactly.

**Lemma 3** Let  $(U, A_1, I_1)$  and  $(U, A_2, I_2)$  be two formal contexts with the same object set. If  $L_{OU}(U, A_2, I_2) \subseteq L_{OU}(U, A_1, I_1)$ , for  $X \subseteq U$ , we have the following two statements:

- (1)  $IN_{R^{\square A_1}}(IN_{R^{\square A_2}}(X)) = IN_{R^{\square A_2}}(X)$ ,
- (2)  $IN_{R^{\square A_2}}(X) \subseteq IN_{R^{\square A_1}}(X)$ .

*Proof.* (1) Since  $L_{OU}(U, A_2, I_2) \subseteq L_{OU}(U, A_1, I_1)$  and for  $X \subseteq U$   $IN_{R^{\square A_2}}(X) \in L_{OU}(U, A_2, I_2)$  by Lemma 2, we have  $IN_{R^{\square A_2}}(X) \in L_{OU}(U, A_1, I_1)$  and  $(IN_{R^{\square A_2}}(X))^{\square A_1 \diamond A_1} = IN_{R^{\square A_2}}(X)$ . Combined with  $(IN_{R^{\square A_2}}(X))^{\square A_1 \diamond A_1} = IN_{R^{\square A_1}}(IN_{R^{\square A_2}}(X))$ , (i) is concluded.

(2) Since  $IN_{R^{\square A_1}}$  and  $IN_{R^{\square A_2}}(X)$  are two inner operators, we have  $IN_{R^{\square A_2}}(X) \subseteq X$  and  $IN_{R^{\square A_1}}(IN_{R^{\square A_2}}(X)) \subseteq IN_{R^{\square A_1}}(X)$ . Thus,  $IN_{R^{\square A_2}}(X) \subseteq IN_{R^{\square A_1}}(X)$  follows directly from (i).

**Theorem 1** Let  $(U, A_1, I_1)$  and  $(U, A_2, I_2)$  be two formal contexts with the same object set. Then we have,  $L_{OU}(U, A_2, I_2) \subseteq L_{OU}(U, A_1, I_1) \Leftrightarrow R^{\square A_1} \subseteq R^{\square A_2}$ .

*Proof.* Sufficiency. Assume  $L_{OU}(U, A_2, I_2) \subseteq L_{OU}(U, A_1, I_1)$  does not hold, then there exists  $X \in L_{OU}(U, A_2, I_2)$  such that  $X \notin L_{OU}(U, A_1, I_1)$ . Thus,  $IN_{R^{\square A_1}}(X) \subset X = IN_{R^{\square A_2}}(X)$  is concluded by Lemma 1. Since  $R^{\square A_1} \subseteq R^{\square A_2}$  implies  $[X]_{R^{\square A_1}} \subseteq [X]_{R^{\square A_2}}$ , we have  $IN_{R^{\square A_2}}(X) \subseteq IN_{R^{\square A_1}}(X)$ , which is a contradiction to  $IN_{R^{\square A_1}}(X) \subset IN_{R^{\square A_2}}(X)$ . Consequently,  $L_{OU}(U, A_2, I_2) \subseteq L_{OU}(U, A_1, I_1)$ .

Necessity. Assume  $R^{\square A_1} \subseteq R^{\square A_2}$  does not hold, then there exists  $X \subseteq U$  such that  $[X]_{R^{\square A_1}} \subseteq [X]_{R^{\square A_2}}$  does not hold. Thus, there exists  $Y \in [X]_{R^{\square A_1}}$  such that  $Y \notin [X]_{R^{\square A_2}}$ . We prove it from two cases:  $X \in L_{OU}(U, A_1, I_1)$  and  $X \notin L_{OU}(U, A_1, I_1)$ .

Firstly, we suppose  $X \in L_{OU}(U, A_1, I_1)$ . Since  $Y \in [X]_{R^{\square A_1}}$  and  $Y \notin [X]_{R^{\square A_2}}$ , we obtain  $X = IN_{R^{\square A_1}}(Y) \subset Y$ . Combining with  $IN_{R^{\square A_2}}(Y) \subseteq IN_{R^{\square A_1}}(Y)$  by Lemma 3, we have  $IN_{R^{\square A_2}}(Y) \subseteq X \subset Y$ . Due to Lemma 1,  $(Y, X) \in R^{\square A_2}$ , which is a contradiction to  $Y \notin [X]_{R^{\square A_2}}$ . Therefore,  $R^{\square A_1} \subseteq R^{\square A_2}$  holds.

Secondly, we suppose  $X \notin L_{OU}(U, A_1, I_1)$ . According to the above discussions, we have  $[IN_{R^{\square A_1}}(X)]_{R^{\square A_1}} \subseteq [IN_{R^{\square A_1}}(X)]_{R^{\square A_2}}$  due to  $IN_{R^{\square A_1}}(X) \in L_{OU}(U, A_1, I_1)$ . Since  $Y \in [X]_{R^{\square A_1}}$ , it is evident that  $IN_{R^{\square A_1}}(X) \subseteq Y$  and  $Y \in [IN_{R^{\square A_1}}(X)]_{R^{\square A_2}}$ . Combining with  $IN_{R^{\square A_2}}(X) \subseteq IN_{R^{\square A_1}}(X)$  we have  $IN_{R^{\square A_2}}(X) \subseteq Y$ . Since  $Y \in [IN_{R^{\square A_1}}(X)]_{R^{\square A_2}}$ , we obtain  $IN_{R^{\square A_2}}(IN_{R^{\square A_1}}(X)) = IN_{R^{\square A_2}}(Y)$ . According to  $IN_{R^{\square A_1}}(X) \subseteq X$ , we have  $IN_{R^{\square A_2}}(IN_{R^{\square A_1}}(X)) \subseteq IN_{R^{\square A_2}}(X)$ . Thus,  $IN_{R^{\square A_2}}(Y) \subseteq IN_{R^{\square A_2}}(X) \subseteq Y$ . By Lemma 1,  $(IN_{R^{\square A_2}}(X), Y) \in R^{\square A_2}$  holds. That is,  $(X, Y) \in R^{\square A_2}$ , which is a contradiction to  $Y \notin [X]_{R^{\square A_2}}$ . Therefore,  $R^{\square A_1} \subseteq R^{\square A_2}$  is concluded.

Consequently, if  $L_{OU}(U, A_2, I_2) \subseteq L_{OU}(U, A_1, I_1)$ , then  $R^{\square A_1} \subseteq R^{\square A_2}$  holds.

### 3.2. Consistent Approximate Representation Space for Formal Contexts and Formal Decision Contexts

In this subsection, we define the consistent approximate representation space to unify formal contexts and formal decision contexts.

**Definition 5** Let  $(U, A, I)$  be a formal context.  $\mathfrak{R}^{\square} = \{R^{\square(a)} \subseteq P(U) \times P(U) \mid a \in A\}$  is a family of equivalence relations on  $P(U)$  and  $R'$  is an equivalence relation on  $P(U)$ . A quadruple  $(U, A, \mathfrak{R}^{\square}, R')$  is said to be an approximate representation space of the context  $(U, A, I)$ .

**Definition 6** Let  $S = (U, A, \mathfrak{R}^\square, R')$  be a consistent approximate representation space and  $R^{\square B} = \bigcap_{a \in B} R^{\square \{a\}}$  for any  $B \subseteq A$ . If  $R^{\square A} \subseteq R'$ , then  $S$  is called a consistent approximate representation space of the context  $(U, A, I)$ .

Obviously,  $S = (U, A, \mathfrak{R}^\square, R^{\square A})$  is a consistent approximate representation space of  $(U, A, I)$ . In fact, for any  $a \in A$ ,  $R^{\square \{a\}}$  and  $R^{\square A}$  are both equivalence relations on  $P(U)$  and  $R^{\square A} \subseteq R^{\square \{a\}}$ , therefore  $S = (U, A, \mathfrak{R}^\square, R^{\square A})$  is a consistent approximate representation space of  $(U, A, I)$  according to Definition 6.

**Definition 7** Let  $(U, A, I)$  and  $(U, C, J)$  be two formal contexts with the same object set.  $(U, A, I, C, J)$  is called a formal decision context, where  $I \subseteq U \times A, J \subseteq U \times C$  and  $A \cap C = \emptyset$ .  $A$  and  $C$  are called condition attribute set and decision attribute set respectively.

**Definition 8** Let  $(U, A, I, C, J)$  be a formal decision context.  $(U, A, I, C, J)$  is said to be consistent if  $R^{\square A} \subseteq R^{\square C}$ , otherwise, it is said to be inconsistent. Where  $R^{\square C} = \{(X, Y) \in P(U) \times P(U) \mid X^{\square C} = Y^{\square C}\}$ .

Combining Theorem 1 and Definition 8, we have  $(U, A, I, C, J)$  is consistent if and only if  $L_{O_U}(U, C, J) \subseteq L_{O_U}(U, A, I)$ . So we have the following result directly.

**Theorem 2** Let  $(U, A, I, C, J)$  be a formal decision context and  $B \subseteq A$ . Then  $S = (U, A, \mathfrak{R}^\square, R^{\square C})$  is a consistent approximate representation space of  $(U, A, I, C, J)$  iff the formal decision context  $(U, A, I, C, J)$  is consistent.

In the following text, we will develop the notions of consistent approximate representation space based on inconsistent formal decision context.

Let  $(U, A, I, C, J)$  be an inconsistent formal decision context.  $R^{\square C}$  partitions  $U$  into a family of disjoint subsets  $U / R^{\square C}$ , we denote  $U / R^{\square C} = \{D_1, D_2, \dots, D_t\}$ , where  $D_j, 1 \leq j \leq t$  is the decision congruence class. For any  $X \in P(U)$ ,  $B \subseteq A$  and  $D_j \in U / R^{\square C}, 1 \leq j \leq t$ , we

define  $P(D_j / [X]_{R^{\square B}}) = \frac{|D_j \cap [X]_{R^{\square B}}|}{|[X]_{R^{\square B}}|}$ , the degree in which the condition congruence class

$[X]_{R^{\square B}}$  belongs to the decision congruence class  $D_j$ .

A membership distribution function  $\mu_B : P(U) \rightarrow [0,1]$  is defined as follows:

$$\mu_B(X) = (P(D_1 / [X]_{R^{\square B}}), P(D_2 / [X]_{R^{\square B}}), \dots, P(D_t / [X]_{R^{\square B}})).$$

Evidently,  $\mu_B(X)$  is a conditional probability distribution on  $U / R^{\square C}$ . For any  $X \in P(U)$ , we denote the maximum decision function by

$$\eta_B(X) = \{D_{j_0} \mid P(D_{j_0} / [X]_{R^{\square B}}) = \max_{1 \leq j \leq t} P(D_j / [X]_{R^{\square B}})\}.$$

**Theorem 3** Let  $(U, A, I, C, J)$  be an inconsistent formal decision context and  $B \subseteq A$ . And we denote  $R^{\square \mu} = \{(X, Y) \in P(U) \times P(U) \mid \mu_A(X) = \mu_A(Y)\}$  and

$$R^{\square \eta} = \{ (X, Y) \in P(U) \times P(U) \mid \eta_A(X) = \eta_A(Y) \} .$$

Then  $S_\mu = (U, A, \mathfrak{R}^\square, R^{\square \mu})$  and  $S_\eta = (U, A, \mathfrak{R}^\square, R^{\square \eta})$  are both consistent approximate representation spaces of  $(U, A, I, C, J)$ . And we call  $S_\mu = (U, A, \mathfrak{R}^\square, R^{\square \mu})$  the distribution consistent approximate representation spaces, and  $S_\eta = (U, A, \mathfrak{R}^\square, R^{\square \eta})$  the maximum decision consistent approximate representation spaces of  $(U, A, I, C, J)$ .

For any  $B \subseteq A$ , the lower and upper approximation distribution functions of inconsistent formal decision context  $(U, A, I, C, J)$  are defined as follows:

$$\begin{aligned} \underline{R}^{\square B} &= (\underline{R}^{\square B}(D_1), \underline{R}^{\square B}(D_2), \dots, \underline{R}^{\square B}(D_l)) , \\ \overline{R}^{\square B} &= (\overline{R}^{\square B}(D_1), \overline{R}^{\square B}(D_2), \dots, \overline{R}^{\square B}(D_l)) , \end{aligned}$$

Where  $\underline{R}^{\square B}(D_j) = \{ [X]_{R^{\square B}} \mid [X]_{R^{\square B}} \subseteq D_j \}$ ,  $\overline{R}^{\square B}(D_j) = \{ [X]_{R^{\square B}} \mid [X]_{R^{\square B}} \cap D_j \neq \emptyset \}$ .

The set of the condition congruence classes which belong to the decision congruence class  $D_j$  is determined by  $\underline{R}^{\square B}(D_j)$ , while  $\overline{R}^{\square B}(D_j)$  is the set of the condition congruence classes which possibly belong to  $D_j$ .

For any  $B \subseteq A$  and  $X \in P(U)$ , we denote

$$\begin{aligned} G_B(X) &= \{ D_j \in U / R^{\square C} \mid [X]_{R^{\square B}} \subseteq D_j \} , \\ M_B(X) &= \{ D_j \in U / R^{\square C} \mid [X]_{R^{\square B}} \cap D_j \neq \emptyset \} . \end{aligned}$$

**Theorem 4** Let  $(U, A, I, C, J)$  be an inconsistent formal decision context and  $B \subseteq A$ . And we denote

$$\begin{aligned} \underline{R}^\square &= \{ (X, Y) \in P(U) \times P(U) \mid G_A(X) = G_A(Y) \} \\ \overline{R}^\square &= \{ (X, Y) \in P(U) \times P(U) \mid M_A(X) = M_A(Y) \} \end{aligned}$$

Then  $\underline{S} = (U, A, \mathfrak{R}^\square, \underline{R}^\square)$  and  $\overline{S} = (U, A, \mathfrak{R}^\square, \overline{R}^\square)$  are both consistent approximate representation spaces of  $(U, A, I, C, J)$ . And we call  $\underline{S} = (U, A, \mathfrak{R}^\square, \underline{R}^\square)$  the lower consistent approximate representation spaces, and  $\overline{S} = (U, A, \mathfrak{R}^\square, \overline{R}^\square)$  the upper consistent approximate representation spaces of  $(U, A, I, C, J)$ .

Suppose  $S = (U, A, \mathfrak{R}^\square, R')$  is a consistent approximate representation space. According to Theorem 2-4, we obtain the following results:

- (i) If  $R' = R^{\square A}$ , then  $S = (U, A, \mathfrak{R}^\square, R')$  can be regarded as the consistent approximate representation space of the context  $(U, A, I)$ .
- (ii) If  $R' = R^{\square C}$ , then  $S = (U, A, \mathfrak{R}^\square, R')$  can be regarded as the consistent approximate representation space of the consistent formal decision context  $(U, A, I, C, J)$ .
- (iii) If  $R' = R^{\square \mu}$ , then  $S = (U, A, \mathfrak{R}^\square, R')$  can be regarded as the distribution consistent approximate representation space of the inconsistent formal decision context  $(U, A, I, C, J)$ .



(iv) If  $R' = R^{\square n}$ , then  $S = (U, A, \mathfrak{R}^{\square}, R')$  can be regarded as the maximum decision consistent approximate representation space of the inconsistent formal decision context  $(U, A, I, C, J)$ .

(v) If  $R' = \underline{R}^{\square}$ , then  $S = (U, A, \mathfrak{R}^{\square}, R')$  can be regarded as the lower consistent approximate representation space of the inconsistent formal decision context  $(U, A, I, C, J)$ .

(vi) If  $R' = \overline{R}^{\square}$ , then  $S = (U, A, \mathfrak{R}^{\square}, R')$  can be regarded as the upper consistent approximate representation space of the inconsistent formal decision context  $(U, A, I, C, J)$ .

Therefore, formal contexts and formal decision contexts have the unified form consistent approximate representation space. Furthermore, we can obtain knowledge on reduction and rule acquisition of formal contexts and formal decision contexts through discussing the corresponding results in consistent approximate representation space.

## 4. Conclusion

This paper has developed the notion of consistent approximate representation space in order to construct the unified model of formal contexts and formal decision contexts based on object oriented concept lattices. It is shown that formal contexts and formal decision contexts can be regarded as special cases of consistent approximate representation space. In further research, we will study knowledge reduction and rule acquisition of consistent approximate representation space based on object oriented concept lattices in order to obtain the corresponding results in formal contexts and formal decision contexts, which can reduce the complexities of different contexts.

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