A Measure Method for Indiscernibility in Imperfect Information System

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Abstract

Traditionally, the information system is assumed to be perfect, i.e. attribute values are not missing and supposed to be precise. In fact, imperfect information system is always existent. In this paper, based on imperfect information system (include missing data and imprecise data), the concepts of indiscernibility and discernibility are defined, their important properties are given, and the relationship between those concepts is established. These results will be helpful for measuring the indiscernibility of knowledge, and have instructive significance for studying for knowledge acquisition in imperfect information system.

Keywords: imperfect information system, tolerance relation, maximal consistent class, indiscernibility

1. Introduction

The entropy of a system as defined by Shannon gives a measure of uncertainty about its actual structure in [1]. In [2, 3], several authors have used Shannon's entropy and its variants to measure uncertainty in rough set theory. A new definition for information entropy in rough set theory is presented in [4]. Especially, M.J. Wierman presents a well justified measure of uncertainty, measure of granularity, along with an axiomatic derivation in [5]. Furthermore, the relationships among information entropy, rough entropy and knowledge granulation in complete information systems are established in [6].

However, imperfect information system is always existent. In [7], a generalization of the rough sets approach which deals with missing and imprecise data is presented. In [8], rough set and rough classification of imperfect information system are investigated. But researches on the measure of indiscernibility based on imperfect information system are relatively less.

In this paper, a measure method for indiscernibility in imperfect information system is proposed, some important properties are given, and the strict complementary relationship between indiscernibility and discernibility is established. These results have instructive significance for studying for the measure of indiscernibility in imperfect information system.

2. Imperfect Information system

Let S = (U, A) be an information system, where

- (1) U is a finite non-empty set of objects;
- (2) A is a finite non-empty set of attributes;
- (3) for every $a \in A$, there is a mapping f_a ,
- $f_a: U \rightarrow V_a$, where V_a is called the value set of a.

Each subset of attributes $P \subseteq A$ determines a binary indiscernibility relation IND(P) on U,

 $IND(P) = \{(x, y) \in U \times U \mid \forall a \in P, f_a(x) = f_a(y)\}.$

The relation IND(P), $P \subseteq A$, is an equivalence relation and constructs a partition of U, which is denoted by U/IND(P).

If V_a contains a null value for at least one attribute $a \in A$, then S is called an incomplete information system ^[9, 10].

By an incomplete information system we mean a system with missing data (null values).

If we change the mapping f_a to $f_a: U \rightarrow (2^{V_a} - \emptyset)$, where 2^{V_a} is the power set of V_a, \emptyset is empty set, then S is called an imperfect information system^[8].

By an imperfect information system we mean a system with missing data (null values) or imprecise data. Generally, a null value of attribute a may be any value in V_a , which is denoted by *.

Obviously, incomplete information system is a special instance of imperfect information system.

Let $P \subseteq A$, a tolerance relation (satisfies reflexivity and symmetry) is defined as follows ^[8], $SIM(P) = \{(x, y) \in U \times U \mid \forall a \in P, f_a(x) \cap f_a(y) \neq \emptyset \}.$ It is easily shown that $SIM(P) = \bigcap_{a \in P} SIM(\{a\})$.

Let U/SIM(P) denote the set of maximal consistent classes in terms of tolerance relation SIM(P). By the maximal consistent class $X \subseteq U$ we mean that any pair of objects in X are in SIM(P), but if get an object from U - X to X, then at least two objects in X are not in SIM(P). The maximal consistent classes in U/SIM(P) generally do not constitute a partition of U but a covering of U.

Example 2.1. Here, we employ an example in [8] to illustrate the tolerance relation. Table 2.1 is an imperfect information system, where

 $U = \{x_1, x_2, x_3, x_4, x_5, x_6\}, A = \{a, b, c, d\},\$ $V_a = \{0, 1, 2\}, V_b = \{0, 1\}, V_c = \{0, 1\},\$ $V_d = \{0, 1\}, f_a(x_4) = \{1, 2\}, \text{ and } f_d(x_3) = * = \{0, 1\}.$

U	A			
	а	b	С	d
x_1	0	1	0	1
x_2	1	0	1	0
<i>x</i> ₃	0	1	0	*
x_4	1, 2	1	1	0
x_5	1	*	1	0
x_6	0, 2	0	0	1

Table 2.1: An imperfect information system^[8].

By computing, we have that

 $U/SIM(A) = \{\{x_1, x_3\}, \{x_2, x_5\}, \{x_4, x_5\}, \{x_6\}\}.$

3. Imperfect Information System

Let S = (U, A) be an imperfect information system, $P \subseteq A$. Based on the tolerance relation defined in [8], we denote the set of all maximal consistent classes of P which includes some object x as $S_x(P)$, i.e. $S_x(P) = \{X \in U / SIM(P) | x \in X\}$.

Definition 3.1. Let S = (U, A) be an imperfect information system, $P \subseteq A$, $X \in U/SIM(P)$ and $x, y \in X$. The indiscernibility between x and y with respect to P is defined by

$$\rho_{P}(x, y) = \frac{\prod_{a \in P} |f_{a}(x) \cap f_{a}(y)|}{\prod_{a \in P} |f_{a}(x)|} \cdot \frac{\prod_{a \in P} |f_{a}(x) \cap f_{a}(y)|}{\prod_{a \in P} |f_{a}(y)|}$$
$$= \prod_{a \in P} \frac{|f_{a}(x) \cap f_{a}(y)|^{2}}{|f_{a}(x)||f_{a}(y)|}.$$

Let S = (U, A) be an imperfect information system, $P \subseteq A$, $S_x^*(P) = \bigcup \{X \in S_x(P)\}$. Then the sum of the indiscernibility between x and all objects included in $S_x^*(P)$ with respect to P is denoted by

$$\rho_x(S_x^*(P)) = \sum_{y \in S_x^*(P)} \rho_P(x, y) \, .$$

We can easily deduce the following properties.

Property 3.1. Let S = (U, A) be an imperfect information system, $P \subseteq A$, $X \in U / SIM(P)$. Then for $\forall x, y \in X$, we have that $0 < \rho_P(x, y) \le 1$.

Property 3.2. Let S = (U, A) be an imperfect information system, $P \subseteq A$, $X \in U / SIM(P)$. Then for $\forall x, y \in X$, we have that $\rho_P(x, y) = \rho_P(y, x)$.

Property 3.3. Let S = (U, A) be an imperfect information system, $P \subseteq A$. Then for $\forall x \in U$, we have that $0 < \rho_x(S_x^*(P)) \le |S_x^*(P)|$.

Theorem 3.1. Let S = (U, A) be an imperfect information system, $P, Q \subseteq A$ and $Q \subseteq P$. Then for $\forall x \in U$, we have that $\rho_x(S_x^*(P)) \le \rho_x(S_x^*(Q))$.

3.1. The Indiscernibility of Information System

In this section we will introduce the indiscernibility of the imperfect information system S based on definition 3.1.

Definition 3.2. Let S = (U, A) be an imperfect information system, $U = \{x_1, x_2, \dots, x_{|U|}\}$. Then the indiscernibility of system S with respect to A is defined as

$$I(A) = \frac{1}{|U|^2} \sum_{i=1}^{|U|} \rho_{x_i}(S_{x_i}^*(A))$$

For an imperfect information system S = (U, A), we have that $1/|U| \le I(A) \le 1$.

Theorem 3.2. Let S = (U, A) be a complete information system, $U = \{x_1, x_2, \dots, x_{|U|}\}$, U / IND(A) $= \{Z_1, Z_2, \dots, Z_m\}$. Then the indiscernibility of system S with respect to A degenerates into

$$G(A) = \frac{1}{|U|^2} \sum_{i=1}^{m} |Z_i|^2,$$

where $G(A) = \frac{1}{|U|^2} \sum_{i=1}^{m} |Z_i|^2$ is the knowledge

granulation of knowledge A defined in [6].

Theorem 3.2 states that the knowledge granulation in complete information system is a special instance of the indiscernibility of imperfect information system.

Theorem 3.3. Let S = (U, A) be an imperfect information system, $U = \{x_1, x_2, \dots, x_{|U|}\}$, $P, Q \subseteq A$. If $Q \subseteq P$, then we have that $I(P) \leq I(Q)$.

Theorem 3.3 states that the indiscernibility of system S with respect to Q decreases by adding more attributes to the attribute set Q.

3.2. The Discernibility of Information System

In this section we give a measure of discernibility in the imperfect information system.

Definition 3.3. Let S = (U, A) be an imperfect information system, $U = \{x_1, x_2, \dots, x_{|U|}\}$. Then the discernibility of system *S* with respect to *A* is defined as

$$D(A) = \sum_{i=1}^{|U|} \frac{1}{|U|} \left[1 - \frac{\rho_{x_i}(S_{x_i}^*(A))}{|U|}\right]$$
$$= \frac{1}{|U|^2} \sum_{i=1}^{|U|} \left[|U| - \rho_{x_i}(S_{x_i}^*(A))\right],$$

where $|U| - \rho_{x_i}(S^*_{x_i}(A))$ represents the sum of the discernibility between x_i and all objects within the universe U with respect to A.

For an imperfect information system S = (U, A), we have that $0 \le D(A) \le 1 - 1/|U|$. **Theorem 3.4.** Let S = (U, A) be a complete information system, $U = \{x_1, x_2, \dots, x_{|U|}\}$, U / IND(A) $= \{Z_1, Z_2, \dots, Z_m\}$. Then the discernibility of system S with respect to A degenerates into

$$E(A) = \sum_{i=1}^{m} \frac{|Z_i|}{|U|} (1 - \frac{|Z_i|}{|U|}),$$

where $E(A) = \sum_{i=1}^{m} \frac{|Z_i|}{|U|} (1 - \frac{|Z_i|}{|U|})$ is the information

entropy of knowledge A defined in [6].

Theorem 3.4 states that the information entropy in complete information system is a special instance of the discernibility of imperfect information system.

Theorem 3.5. Let S = (U, A) be an imperfect information system, $U = \{x_1, x_2, \dots, x_{|U|}\}$, $P, Q \subseteq A$. If $Q \subseteq P$, then we have that $D(Q) \leq D(P)$.

Theorem 3.5 states that the discernibility of system S with respect to Q increases by adding more attributes to the attribute set Q.

3.3. Relationship Between the Indiscernibility and the Discernibility of Information System

In this section the relationship between the indiscernibility and the discernibility of system S is established, which is proved to be a strict complementary relationship.

Theorem 3.6. For an imperfect information system S = (U, A), we have that

$$I(A) + D(A) = 1.$$

Proof. Let S = (U, A) be an imperfect information system, $U = \{x_1, x_2, \dots, x_{|U|}\}$. By the definition 3.2 and 3.3, we have that

$$D(A) = \sum_{i=1}^{|U|} \frac{1}{|U|} \left[1 - \frac{\rho_{x_i}(S_{x_i}^*(A))}{|U|}\right]$$
$$= \sum_{i=1}^{|U|} \frac{1}{|U|} - \sum_{i=1}^{|U|} \frac{\rho_{x_i}(S_{x_i}^*(A))}{|U|^2}$$
$$= 1 - I(A).$$

It follows that I(A) + D(A) = 1. This completes the proof.

Example 3.1. Continued from example 2.1, by computing, we have that

$$\begin{split} \rho_{x_{1}}(S_{x_{1}}^{*}(A)) &= \rho_{x_{1}}(\{x_{1}, x_{3}\}) = \frac{3}{2}, \\ \rho_{x_{2}}(S_{x_{2}}^{*}(A)) &= \rho_{x_{2}}(\{x_{2}, x_{5}\}) = \frac{3}{2}, \\ \rho_{x_{3}}(S_{x_{3}}^{*}(A)) &= \rho_{x_{3}}(\{x_{1}, x_{3}\}) = \frac{3}{2}, \\ \rho_{x_{4}}(S_{x_{4}}^{*}(A)) &= \rho_{x_{5}}(\{x_{1}, x_{3}\}) = \frac{5}{4}, \\ \rho_{x_{5}}(S_{x_{5}}^{*}(A)) &= \rho_{x_{5}}(\{x_{2}, x_{4}, x_{5}\}) = \frac{7}{4}, \\ \rho_{x_{6}}(S_{x_{6}}^{*}(A)) &= \rho_{x_{6}}(\{x_{6}\}) = 1. \\ I(A) &= \frac{1}{|U|^{2}} \sum_{i=1}^{|U|} \rho_{x_{i}}(S_{x_{i}}^{*}(A)) \\ &= \frac{1}{36}(\frac{3}{2} + \frac{3}{2} + \frac{3}{2} + \frac{5}{4} + \frac{7}{4} + 1) \\ &= \frac{17}{72}. \\ D(A) &= \sum_{i=1}^{|U|} \frac{1}{|U|} [1 - \frac{\rho_{x_{i}}(S_{x_{i}}^{*}(A))] \\ &= \frac{1}{6} [(1 - \frac{3}{12}) + (1 - \frac{3}{12}) + (1 - \frac{3}{12})] \\ &+ (1 - \frac{5}{24}) + (1 - \frac{7}{24}) + (1 - \frac{1}{6})] \\ &= \frac{55}{72}. \end{split}$$

It is clear that I(A) + D(A) = 1.

4. Conclusions

In this paper, the concepts of the indiscernibility and the discernibility in imperfect information system are proposed, their important properties are given, and the relationship between the indiscernibility and the discernibility of information system is established, it can be expressed as I(A) + D(A) = 1. These results provide new study view for the measure of the indiscernibility of knowledge, and can be further used in many fields such as knowledge discovery and data mining etc.

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