ROUGH SET APPROXIMATIONS: A CONCEPT ANALYSIS POINT OF VIEW

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Summary

Rough set theory was proposed by Pawlak for analyzing data and reasoning about data. From a concept analysis point of view, we review and reformulate main results of rough set theory in the context of data processing and analysis. This enables us to see clearly the motivations for introducing rough set theory and its basic components and appropriate applications, leading to an appreciation for the theory.

1. Two Aspects of Data

In order to see the motivations for introducing rough set theory and, hence, its uniqueness and contributions, we first give a brief discussion of two important aspects of data and then present an interpretation of rough sets as a theory concerning the meaning of data from a concept analysis point of view.

In processing and analyzing data, we consider two important aspects of data, namely, the form and content of data. Consequently, there are two fundamental classes of tasks: one is the class of form-oriented tasks and the other the class of content-oriented tasks. Form-oriented tasks focus on manipulating data as uninterpreted symbols, such as communication, storage and retrieval of data, without considering their physical meaning. Content-oriented tasks concentrate on semantics of data, such as determining the meaning of data, providing an explanation of data, building models from data etc., without worrying about how data are stored, retrieved and communicated. The division of the two (i.e., separation of form and content), on the one hand, and the union (i.e., integration of form and content), on the other hand, are crucial to data processing and analysis.

Normally, the separation of form and content leads to a simple and general theory for

data processing and analysis at symbolic level. Two examples of form-oriented data processing are the information theory of communications proposed by Shannon (1948) and the relational database theory proposed by Codd (1970) for storing and retrieving data. Shannon's theory focuses on "reproducing at one point either exactly or approximately a message selected at another point." The meaning of the messages is considered to be irrelevant for purpose of transmitting the messages. Codd's theory is "concerned with the application of elementary relation theory to systems which provide shared access to large banks of formatted data." Data are represented conceptually by using "n-ary relations, a normal form for data base relations," for retrieval, independent of particular machine implementations and specific applications. The meaning of data in a database is not considered.

For content-oriented tasks, the semantics of data is of the main concern. We determine the meaning of data independent of the form or appearance of data as well as the methods for communicating, storing or retrieving data. Unlike the form-oriented tasks, it might be difficult to have a simple and general theory for modeling semantics of data, as semantics is usually domain and context dependent. Rough set analysis (Pawlak, 1982. 1991) and formal concept analysis (Ganter and Wille, 1999; Wille, 1982) are two theories, proposed at the same time, for describing and studying definable concepts and the structures of all definable concepts in data represented in a tabular form as in relational databases.

Concepts are the basic units of thought that underlie human intelligence and communication. A study of concepts involves multiple disciplines, including philosophy, psychology, cognitive science, mathematics, inductive data processing and analysis, inductive learning, and many others (Michalski et al, 1983; Smith, 1989; Sowa, 1984; van Mechelen et al, 1993). There are many views of concepts such as the classical view, the exemplar view, the frame view, and the theory view (van Mechelen et al, 1993). In the classical view, concepts have well-defined boundaries and are describable by sets of singly necessary and jointly sufficient conditions (van Mechelen et al, 1993). Every concept consists of two parts, the intension and the extension of the concept (Ogden and Richards, 1946; Smith, 1989; Sowa, 1984; van Mechelen et al, 1993). The intension of a concept consists of all properties or attributes that are valid for all those objects to which the concept applies. The extension of a concept is the set of objects or entities that are instances of the concept.

Due to the complexity and diversity of concepts, it is difficult to design a method that is general enough for describing intensions of all concepts. Instead, we build a specific model that enables us to define explicitly and precisely a certain class of concepts in a particular context. Formal concept analysis, proposed by Wille (1982), investigates a concept that is defined by and only defined by a set of attributes in a binary data/information table called a formal context. The set of all formal concepts, i.e., all definable concepts, forms a lattice, showing the hierarchical relationships between concepts. Significant contributions of formal concept analysis are an explicit and precise description of the intension and extension of a concept, and the characterization of relationships between concepts using a lattice.

Rough set theory is another theory for concept analysis using an information table.

Although earlier studies (Marek 2013; Marek and Pawlak, 1976; Marek and Truszczyński, 1999; Pawlak, 1981) aimed at formulating a mathematical foundation of information systems characterized by information tables, the main contributions of rough set theory are the introduction of the notion of definability of concepts/sets (Marek and Pawlak, 1976; Yao, 2007) and the approximations of a set by a pair of definable sets.

In this chapter, we only examine the two notions of definability and approximations. For a more complete discussion on all aspects of rough set theory and its applications, a reader may read the book by Pawlak (1991) and some recently edited books (Peters et al, 2012; Skowron and Suraj, 2013). For studies on the connections between formal concept analysis and rough set analysis, a reader may read some recent papers (for example, Lai and Zhang, 2012; Wolff, 2001; Yao, 2004; Ytow et al, 2006).

2. Definability and Approximations

Rough set analysis is based on two basic notions of the definability of concepts and the approximation of concepts. These two notions are defined with respect to an information table that describes all available information of a set of objects.

2.1. Information Tables

An information table T can be defined as a tuple as follows (Pawlak, 1981, 1991):

$$T = \left(U, AT, \left\{V_a \mid a \in AT\right\}, \left\{I_a \mid a \in AT\right\}\right)$$

$$\tag{1}$$

where U is a finite set of objects called the universe, AT is a finite set of attributes, V_a is the domain of attribute a, and $I_a: U \to V_a$ is an information function. We use $I_a(x)$ to denote the value of object x on attribute a. We can conveniently represent an information table in a tabular form, in which each row represents an object, each column represents an attribute, and each cell represents the value of an object on the corresponding attribute.

Table 1 is an information table with $U = \{o_1, o_2, o_3, o_4, o_5, o_6, o_7\}$, $AT = \{Height, Hair, Eyes\}, V_{Height} = \{short, tall\}, V_{Hair} = \{blond, red, dark\}$ and $V_{Eves} = \{blue, brown\}$. For object o_1 , we have:

$$I_{\text{Height}}(o_1) = short$$
,

 $I_{\text{Hair}}(o_1) = blond$,

 $I_{\rm Eyes}(o_1) = blue$.

In a table representation, objects are given in a sequence of rows and attributes are in a

sequence of columns. Although in the literature of rough sets one typically refers to an object by its row number or an attribute by its column number, it is important to note that semantically there is no ordering on the set of objects nor on the set of attributes. From the table, it can be seen that some objects have the same description. For example, objects o_2 and o_3 have the same description. Consequently, based only on their description, one can not distinguish objects o_2 and o_3 . This observation is in fact the basis of rough set analysis.

Object	Height	Hair	Eyes
<i>o</i> ₁	short	blond	blue
<i>o</i> ₂	short	blond	brown
03	short	blond	brown
04	tall	dark	blue
05	tall	dark	blue
<i>0</i> ₆	tall	dark	blue
07	tall	red	blue

Table 1. An information table

2.2. Concepts and Definable Concepts

In an information table, a subset of objects $X \subseteq U$ may be viewed as the extension of a concept. In order to describe formally the intension of a concept, we introduce a description language, as suggested by Marek and Pawlak (1976). A description language DL can be recursively defined as follows:

- (1) $(a = v) \in DL$, where $a \in AT$, $v \in V_a$,
- (2) if $p, q \in DL$, then $(p \land q), (p \lor q) \in DL$.

Formulas defined by (1) are called atomic formulas. For simplicity, we consider a language defined by two logic connectives \land and \lor , which is a sub-language used by Marek and Pawlak (1976) and by Pawlak (1991). By assuming that \land has a higher precedence in computation, one may remove unnecessary parentheses in a formula. This language is powerful enough for rough set analysis.

The satisfiability of a formula p by an object x, written $x \models p$, is defined as follows:

- (*i*) $x \vDash a = v$, iff $I_a(x) = v$,
- (*ii*) $x \vDash p \land q$, iff $x \vDash p$ and $x \vDash q$
- (*iii*) $x \models p \lor q$, iff $x \models p$ or $x \models q$

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Biographical Sketches

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