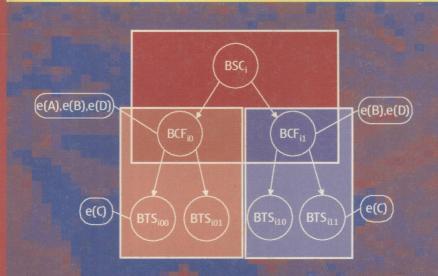
Rosa Meo Pier Luca Lanzi Mika Klemettinen (Eds.)

Database Support for Data Mining Applications

Discovering Knowledge with Inductive Queries





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Discovering Knowledge with Inductive Queries





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Library of Congress Control Number: 2004095564

CR Subject Classification (1998): I.2, H.2, H.3

ISSN 0302-9743 ISBN 3-540-22479-3 Springer-Verlag Berlin Heidelberg New York

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Typesetting: Camera-ready by author, data conversion by Scientific Publishing Services, Chennai, India Printed on acid-free paper SPIN: 11301172 06/3142 5 4 3 2 1 0

Lecture Notes in Artificial Intelligence

2682

Edited by J. G. Carbonell and J. Siekmann

Subseries of Lecture Notes in Computer Science

Preface

Data mining from traditional relational databases as well as from non-traditional ones such as semi-structured data, Web data and scientific databases such as biological, linguistic and sensor data has recently become a popular way of discovering hidden knowledge. In the context of relational and traditional data, methods such as association rules, chi square rules, ratio rules, implication rules, etc. have been proposed in multiple, varied contexts. In the context of non-traditional data, newer, more experimental yet novel techniques are being proposed. There is an agreement among the researchers across communities that data mining is a key ingredient for success in their respective areas of research and development. Consequently, interest in developing new techniques for data mining has peaked and a tremendous stride is being made to answer interesting and fundamental questions in various disciplines using data mining.

In the past, researchers mainly focused on algorithmic issues in data mining and placed much emphasis on scalability. Recently, the focus has shifted towards a more declarative way of answering questions using data mining that has given rise to the concept of mining queries.

Data mining has recently been applied with success to discovering hidden knowledge from relational databases. Methods such as association rules, chi square rules, ratio rules, implication rules, etc. have been proposed in several and very different contexts. To cite just the most frequent and famous ones: the market basket analysis, failures in telecommunication networks, text analysis for information retrieval, Web content mining, Web usage, log analysis, graph mining, information security and privacy, and finally analysis of objects traversal by queries in distributed information systems.

From these widespread and various application domains it results that data mining rules constitute a successful and intuitive descriptive paradigm able to offer complementary choices in rule induction. Other than inductive and abductive logic programming, research into data mining from knowledge bases has been almost non-existent, because contemporary methods place the emphasis on the scalability and efficiency of algorithmic solutions, whose inherent procedurality is difficult to cast into the declarativity of knowledge base systems.

In particular, researchers convincingly argue that the ability to declaratively mine and analyze relational databases for decision support is a critical requirement for the success of the acclaimed data mining technology. Indeed, DBMSs constitute today one of the most advanced and sophisticated achievements that applied computer science has made in the past years. Unfortunately, almost all the most powerful DBMSs we have today have been developed with a focus on On-Line Transaction-Processing tasks. Instead, database technology for On-Line Analytical-Processing tasks, such as data mining, is more recent and in need of further research.

Although there have been several encouraging attempts at developing methods for data mining using SQL, simplicity and efficiency still remain significant prerequisites for further development. It is well known that today database technology is mature enough: popular DBMSs, such as Oracle, DB2 and SQL-Server, provide interfaces, services, packages and APIs that embed data mining algorithms for classification, clustering, association rules extraction and temporal sequences, such that they are directly available to programmers and ready to be called by applications.

Therefore, it is envisioned that we should be able now to mine relational databases for interesting rules directly from database query languages, without any data restructuring or preprocessing steps. Hence no additional machineries with respect to database languages would be necessary. This vision entails that the optimization issues should be addressed at the system level for which we have now a significant body of research, while the analyst could concentrate better on the declarative and conceptual level, in which the difficult task of interpretation of the extracted knowledge occurs. Therefore, it is now time to develop declarative paradigms for data mining so that these developments can be exploited at the lower and system level, for query optimization.

With this aim we planned this book on "Data Mining" with an emphasis on approaches that exploit the available database technology, declarative data mining, intelligent querying and associated issues such as optimization, indexing, query processing, languages and constraints. Attention is also paid to solution of data preprocessing problems, such as data cleaning, discretization and sampling, developed using database tools and declarative approaches, etc.

Most of this book resulted also as a consequence of the work we conducted during the development of the cInQ project (consortium on discovering knowledge with Inductive Queries) an EU funded project (IST 2000-26469) aiming at developing database technology for leveraging decision support systems by means of query languages and inductive approaches to knowledge extraction from databases. It presents new and invited contributions, plus the best papers, extensively revised and enlarged, presented during workshops on the topics of database technology, data mining and inductive databases at international conferences such as EDBT and PKDD/ECML, in 2002.

May 2004

Rosa Meo Pier Luca Lanzi Mika Klemettinen

Volume Organization

This volume is organized in two main sections. The former focuses on *Database Languages and Query Execution*, while the latter focuses on methodologies, techniques and new approaches that provide *Support for Knowledge Discovery Process*. Here, we briefly overview each contribution.

Database Languages and Query Execution

The first contribution is *Inductive Databases and Multiple Uses of Frequent Itemsets: The cInQ Approach* which presents the main contributions of theoretical and applied nature, in the field of inductive databases obtained in the cInQ project.

In Query Languages Supporting Descriptive Rule Mining: A Comparative Study we provide a comparison of features of available relational query languages for data mining, such as DMQL, MSQL, MINE RULE, and standardization efforts for coupling database technology and data mining systems, such as OLEDB-DM and PMML.

Declarative Data Mining Using SQL-3 shows a new approach, compared to existing SQL approaches, to mine association rules from an object-relational database: it uses a recursive join in SQL-3 that allows no restructuring or preprocessing of the data. It proposes a new mine by SQL-3 operator for capturing the functionality of the proposed approach.

Towards a Logic Query Language for Data Mining presents a logic database language with elementary data mining mechanisms, such as user-defined aggregates that provide a model, powerful and general as well, of the relevant aspects and tasks of knowledge discovery.

Data Mining Query Language for Knowledge Discovery in a Geographical Information System presents SDMOQL a spatial data mining query language for knowledge extraction from GIS. The language supports the extraction of classification rules and association rules, the use of background models, various interestingness measures and the visualization.

Towards Query Evaluation in Inductive Databases Using Version Spaces studies inductive queries. These ones specify constraints that should be satisfied by the data mining patterns in which the user is interested. This work investigates the properties of solution spaces of queries with monotonic and anti-monotonic constraints and their boolean combinations.

The GUHA Method, Data Preprocessing and Mining surveys the basic principles and foundations of the GUHA method, the available systems and related works. This method originated in the Czechoslovak Academy of Sciences of Prague in

the mid 1960s with strong logical and statistical foundations. Its main principle is to let the computer generate and evaluate all the hypotheses that may be interesting given the available data and the domain problem. This work discusses also the relationships between the GUHA method and relational data mining and discovery science.

Constraint Based Mining of First Order Sequences in SeqLog presents a logical language, SeqLog, for mining and querying sequential data and databases. This language is used as a representation language for an inductive database system. In this system, variants of level-wise algorithms for computing the version space of the solutions are proposed and experimented in the user-modeling domain.

Support for Knowledge Discovery Process

Interactivity, Scalability and Resource Control for Efficient KDD Support in DBMS proposes a new approach for combining preprocessing and data mining operators in a KDD-aware implementation algebra. In this way data mining operators can be integrated smoothly into a database system, thus allowing interactivity, scalability and resource control. This framework is based on the extensive use of pipelining and is built upon an extended version of a specialized database index.

Frequent Itemset Discovery with SQL Using Universal Quantification investigates the integration of data analysis functionalities into two basic components of a database management system: query execution and optimization. It employs universal and existential quantifications in queries and a vertical layout to ease the set containment operations needed for frequent itemsets discovery.

Deducing Bounds on the Support of Itemsets provides a complete set of rules for deducing tight bounds on the support of an itemset if the support of all its subsets are known. These bounds can be used by the data mining system to choose the best access path to data and provide a better representation of the collection of frequent itemsets.

Model-Independent Bounding of the Supports of Boolean Formulae in Binary Data considers frequencies of arbitrary boolean formulas, a new class of aggregates: the summaries. These ones are computed for descriptive purposes on a sparse binary data set. This work considers the problem of finding tight upper bounds on these frequencies and gives a general formulation of the problem with a linear programming solution.

Condensed Representations for Sets of Mining Queries proposes a general framework for condensed representations of sets of mining queries, defined by monotonic and anti-monotonic selection predicates. This work proves important for inductive and database systems for data mining since it deals with sets of queries, whereas previous work in maximal, closed and condensed representations treated so far the representation of a single query only.

Acknowledgments

For reviewing the contributions included in this volume we relied on a group of internationally well-known experts who we wish to thank here.

Roberto Bayardo

IBM Almaden Research Center, USA

Elena Baralis

Politecnico di Torino, Italy

Christian Böhm

University for Health Informatics and Technology,

Austria

Saso Dzeroski Kris Koperski Jozef Stefan Institute, Slovenia Insightful Corporation, USA

Stefan Kramer

Technische Universität München, Germany

Dominique Laurent

Université F. Rabelais, Tours, France

Giuseppe Manco

ICAR-CNR, Italy

Stefano Paraboschi

Università di Bergamo, Italy

Jian Pei

The State University of New York, USA

Giuseppe Psaila

Università di Bergamo, Italy

Lorenza Saitta

Università del Piemonte Orientale, Italy

Maria Luisa Sapino Hannu Toivonen Università di Torino, Italy University of Helsinki, Finland

Jiong Yang

University of Illinois at Urbana Champaign, USA

Mohammed Zaki

Rensselaer Polytechnic Institute, USA

One-Sided Instance-Based Boundary Sets introduces a family of version-space representations that are important for their applicability to inductive databases. They correspond to the task of concept learning from a database of examples when this database is updated. One-sided instance-based boundary sets are shown to be correctly and efficiently computable.

Domain Structures in Filtering Irrelevant Frequent Patterns introduces a notion of domain constraints, based on distance measures and in terms of domain structure and concept taxonomies. Domain structures are useful in the analysis of communications networks and complex systems. Indeed they allow irrelevant combinations of events that reflect the simultaneous information of independent processes in the same database to be pruned.

Integrity Constraints over Association Rules investigates the notion of integrity constraints in inductive databases. This concept is useful in detecting inconsistencies in the results of common data mining tasks. This work proposes a form of integrity constraints called association map constraints that specifies the allowed variations in confidence and support of association rules.

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Table of Contents

I Database Languages and Query Execution	
Inductive Databases and Multiple Uses of Frequent Itemsets: the cInQ Approach	1
Query Languages Supporting Descriptive Rule Mining: A Comparative Study	24
Declarative Data Mining Using SQL35 Hasan M. Jamil	52
Towards a Logic Query Language for Data Mining	76
A Data Mining Query Language for Knowledge Discovery in a Geographical Information System	95
Towards Query Evaluation in Inductive Databases Using Version Spaces . $11Luc~De~Raedt$	17
The GUHA Method, Data Preprocessing and Mining	35
Constraint Based Mining of First Order Sequences in SeqLog	54
II Support for KDD-Process	
Interactivity, Scalability and Resource Control for Efficient KDD Support in DBMS	74
Frequent Itemset Discovery with SQL Using Universal Quantification 19 Ralf Rantzau	94
Deducing Bounds on the Support of Itemsets	14

XII Table of Contents

Model-Independent Bounding of the Supports of Boolean Formulae in	
Binary Data	234
Artur Bykowski, Jouni K. Seppänen, Jaakko Hollmén	
Condensed Representations for Sets of Mining Queries	250
Arnaud Giacometti, Dominique Laurent, Cheikh Talibouya Diop	-00
One-Sided Instance-Based Boundary Sets	270
Evgueni N. Smirnov, Ida G. Sprinkhuizen-Kuyper, H. Japp van den Herik	
Domain Structures in Filtering Irrelevant Frequent Patterns	289
Kimmo Hätönen, Mika Klemettinen	
Integrity Constraints over Association Rules	306
Artur Bykowski, Thomas Daurel, Nicolas Méger, Christophe Rigotti	
Author Index	204
Addition index	324

Inductive Databases and Multiple Uses of Frequent Itemsets: The CINQ Approach

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Abstract. Inductive databases (IDBs) have been proposed to afford the problem of knowledge discovery from huge databases. With an IDB the user/analyst performs a set of very different operations on data using a query language, powerful enough to perform all the required elaborations, such as data preprocessing, pattern discovery and pattern post-processing. We present a synthetic view on important concepts that have been studied within the cINQ European project when considering the pattern domain of itemsets. Mining itemsets has been proved useful not only for association rule mining but also feature construction, classification, clustering, etc. We introduce the concepts of pattern domain, evaluation functions, primitive constraints, inductive queries and solvers for itemsets. We focus on simple high-level definitions that enable to forget about technical details that the interested reader will find, among others, in cINQ publications.

1 Introduction

Knowledge Discovery in Databases (KDD) is a complex interactive process which involves many steps that must be done sequentially. In the CINQ project¹, we want to develop a new generation of databases, called "inductive databases" (IDBs), suggested by Imielinski and Mannila in [42] and for which a simple formalization has been proposed in [20]. This kind of databases integrate raw data with knowledge extracted from raw data, materialized under the form of patterns into a common framework that supports the knowledge discovery process within a database framework. In this way, the process of KDD consists essentially in a querying process, enabled by a query language that can deal either with raw data or patterns and that can be used throughout the whole KDD process across many different applications. A few query languages can be considered as candidates for inductive databases. For instance, considering the prototypical case of assoc-

¹ This research is partially funded by the Future and Emerging Technologies arm of the IST Programme FET-Open scheme (cInQ project IST-2000-26469). The author warmly acknowledges all the contributors to the cInQ project and more particularly Luc De Raedt, Baptiste Jeudy, Mika Klemettinen, and Rosa Meo.

R. Meo et al. (Eds.): Database Support for Data Mining Applications, LNAI 2682, pp. 1–23, 2004. © Springer-Verlag Berlin Heidelberg 2004

iation rule mining, [10] is a comparative evaluation of three proposals (MSQL [43], DMQL [38], and MINE RULE [59]) in the light of the IDBs' requirements.

In this paper, we focus on mining queries, the so-called *inductive queries*, i.e., queries that return patterns from a given database. More precisely, we consider the pattern domain of itemsets and databases that are transactional databases. Doing so, we can provide examples of concepts that have emerged as important within the CINQ project after 18 months of work.

It is useful to abstract the meaning of mining queries. A simple model has been introduced in [55] that considers a data mining process as a sequence of queries over the data but also the so-called theory of the data. Given a language \mathcal{L} of patterns (e.g., itemsets, sequences, association rules), the theory of a database r with respect to \mathcal{L} and a selection predicate q is the set $Th(r, \mathcal{L}, q) = \{\phi \in \mathcal{L} \mid q(r, \phi) \text{ is true}\}.$ The predicate q indicates whether a pattern ϕ is considered interesting (e.g., ϕ denotes a property that is "frequent" in r). The selection predicate can be defined as a combination (boolean expression) of primitive constraints that have to be satisfied by the patterns. Some of them refer to the "behavior" of a pattern in the data, e.g., its "frequency" in a given data set is above or below a user-given threshold, some others define syntactical restrictions on desired patterns, e.g., its "length" is below a user-given threshold. Preprocessing concerns the definition of the database r, the mining phase is often the computation of the specified theory while post-processing can be considered as a querying activity on a materialized theory or the computation of a new theory.

This formalization however does not reflect the context of many classical data mining processes. Quite often, the user is interested not only in a collection of patterns that satisfy some constraints (e.g., frequent patterns, strong rules, approximate inclusion or functional dependencies) but also to some properties of these patterns in the selected database (e.g., their frequencies, the error for approximate dependencies). In that case, we will consider the so-called extended theories. For instance, when mining frequent itemsets or frequent and valid association rules [2], the user needs for the frequency of the specified patterns or rules. Indeed, during the needed post-processing phase, the user/analyst often uses various objective interestingness measures like the confidence [2], the conviction [23] or the J-mesure [72] that are computed efficiently provided that the frequency of each frequent itemset is available. Otherwise, it might be extremely expensive to look at the data again.

Designing solvers for more or less primitive constraints concerns the core of data mining algorithmic research. We must have solvers that can compute the (extended) theories and that have good properties in practice (e.g., scalability w.r.t. the size of the database or the size of the search space). A "generate and test" approach that would enumerate the sentences of $\mathcal L$ and then test the selection predicate q is generally impossible. A huge effort has concerned a clever use of the constraints occurring in q to have a tractable evaluation of useful inductive queries. This is the research area of constraint-based data mining. Most of the algorithmic research in pattern discovery tackles the design

of complete algorithms for computing (extended) theories given more or less specific conjunctions of primitive constraints. Typically, many researchers have considered the computation of frequent patterns, i.e., patterns that satisfy a minimal frequency constraint. An important paper on a generic algorithm for such a typical mining task is [55]. However, if the active use of the so-called *anti-monotonic constraints* (e.g., the minimal frequency) is now well-understood, the situation is far less clear for non anti-monotonic constraints [64, 51, 18].

A second major issue is the possibility to approximate the results of (extended) inductive queries. This approximation can concern a collection of patterns that is a superset or a subset of the desired collection. This is the typical case when the theories are computed from a sample of the data (see, e.g., [74]) or when a relaxed constraint is used. Another important case of approximation for extended theories is the exact computation of the underlying theory while the evaluation functions are only approximated. This has lead to an important research area, the computation of the so-called *condensed representations* [54], a domain in which we have been playing a major role since the study of frequent closed itemsets as an ϵ -adequate representation for frequency queries [12].

This paper is organized as follows. Section 2 introduces notations and definitions that are needed for discussing inductive queries that return itemsets. It contains an instance of the definition of a pattern domain. Section 3 identifies several important open problems. Section 4 provides elements of solution that are currently studied within the cINQ project. Section 5 is a short conclusion.

2 A Pattern Domain for Itemsets

The definition of a pattern domain is made of the definition of a language of patterns \mathcal{L} , evaluation functions that assign a semantics to each pattern in a given database \mathbf{r} , languages for primitive constraints that specify the desired patterns, and inductive query languages that provide a language for combining the primitive constraints.

We do not claim that this paper is an exhaustive description of the itemset pattern domain. Even though we selected representative examples of evaluation functions and primitive constraints, many others have been or might be defined and used.

2.1 Language of Patterns and Terminology

We introduce some notations that are used for defining the pattern domain of itemsets. In that context, we consider that:

- A so-called transactional database contains the data,
- Patterns are the so-called *itemsets* and one kind of descriptive rule that can be derived from them, i.e., the association rules.

Definition 1 (Transactional Databases). Assume that Items is a finite set of symbols denoted by capital letters, e.g., Items= {A,B,C,...}. A transaction