

Knowledge Discovery from Data Bases

Proposal for a MAP-I UC

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1 Knowledge Discovery from Data Bases

“We are deluged by data: scientific data, medical data, demographic data, financial data, and marketing data. People have no time to look at this data. Human attention has become the precious resource. So, we must find ways to automatically analyze the data, characterize trends in it, and automatically flag anomalies.” ([Han and Kamber, 2006](#)).

The development of information and communication technologies make possible collect data with high degree of detail that might be automatically transmitted at high-speed. Some examples of real-world applications include: TCP/IP traffic, queries in search engines over Internet, records of telecommunication calls, SMS, emails, stock market, sensors in electrical grid, etc. For illustrative purposes, we present some numbers: The number of daily phone calls is around 3 billion; the number of SMS is 1 billion daily, the number of sent emails is around 30 billion.

Most of this information will never be seen by a human being. Taking this into account, tools for automatic real-time data analysis are of increasing importance. The computer processes, analyze, and filter the data, selecting the most promising hypothesis. Some typical applications include: user modeling, activity monitoring, sensor networks, classification, intrusion detection, etc.

Scientific areas: Data Mining, Machine Learning, Computer Science.

1.1 Main Goals

At the end of the semester the students should be able to:

1. Formulate a decision problem as a data mining problem;

2. Identify the basic tasks in knowledge discovery from data bases;
3. Identify and use the main methods in solving data mining problems;
4. Apply the main methods and algorithms for each mining task;
5. Apply the main methods and algorithms in real-world problems and adapt to new contexts.

1.2 Team

- Alipio Jorge, U. Porto
- João Gama, U.Porto (coordinator)
- Paulo Azevedo, U. Minho
- Rita P. Ribeiro, U. Porto

1.3 Syllabus

- Introductory Concepts
 - Data Mining tasks;
 - Cluster Analysis
 - * Cluster Analysis: concepts and methods;
 - * Partitioning and Hierarchical Methods;
 - Association Analysis (Hand et al., 2001)
 - * Frequent pattern mining;
 - * Frequent Sequence mining;
 - Predictive Data Mining: Classification and Regression.
 - * Optimization Methods: Artificial Neural Networks; Support Vector Machines. Bishop (1995)
 - * Probabilistic Methods: Bayesian Classifiers;
 - * Search based Methods: Decision Trees and Rules.
 - Evaluation in Predictive Data Mining (Japkowicz and Shah, 2011).
 - * Evaluation: goals and perspectives;
 - * Loss Functions and Cost-benefit analysis;
 - * Bias-Variance analysis;

- Ensembles and Multiple Models ([Gama et al., 2017](#))
 - * Concepts and methods;
 - * Combining Homogeneous Models;
 - * Combining Heterogeneous models;
- Advanced Topics
 - Social Network Analysis
 - * Concepts and methods;
 - * Evolution of Networks;
 - Text Mining
 - * Concepts and methods;
 - * Information retrieval;
 - * Document classification;
 - Web Mining and Link Analysis ([Han et al., 2012](#))
 - * Concepts and methods;
 - * Web and Structure mining;
 - * Link analysis;
 - Big Data and Data stream Mining ([Gama, 2010](#))
 - * Big Data: Applications and tools
 - * Concepts and methods;
 - * Summarizing data streams;
 - * Knowledge discovery from data streams;
 - Advanced Topics in classification ([Branco et al., 2016](#))
 - * outliers, rare cases, novelty detection,
 - * structured output prediction.
 - Data Mining Standards and Processes

1.4 Teaching Methods and Evaluation

The teaching method consists of theoretical-practical classes. The evaluation consists of home-works.

1.5 Bibliography

Recommended books:

- *Extração de Conhecimento de Dados - Data Mining*, J. Gama, A. Carvalho, K. Faceli, A. Lorena, M. Oliveira; Sílabo, 2012.
- *Data Mining: Concepts and Techniques*, Jiawei Han, Micheline Kamber, Jean Pei; Morgan Kaufmann, 2014.
- *Knowledge Discovery from Data Streams*, J. Gama, CRC Press, 2010
- *Principles of Data Mining*, D. Hand, H. Mannila, P. Smyth; The MIT Press, 2002
- *Guide to Intelligent Data Analysis: How to Intelligently Make Sense of Real Data*, Michael R. Berthold, Frank Klawonn, Frank Hoppner, Christian Borgelt, Springer, 2010

1.6 Software

The use of software tools has the main goal of solving practical problems, the study, analysis, and evaluation in small-scale applied problems as a formative perspective. We choose two software tools, frequently used in data mining teaching:

- R (Ihaka and Gentleman, 1996) - a statistical oriented programming language. The interface is command line.
- WEKA (Witten and Frank, 2005) is a machine-learning oriented software. It uses a graphical interface, with the possibility to develop sequences of tasks. The Knowledge Explorer allows decomposing a complex problem into sub-problems in a graphical environment.
- Knime or Rapid-Miner - new generation of data mining tools.

Referências

Bishop, C. (1995). *Neural Networks for Pattern Recognition*. Oxford University Press.

Branco, P., Torgo, L., and Ribeiro, R. P. (2016). A survey of predictive modeling on imbalanced domains. *ACM Comput. Surv.*, 49(2):31:1–31:50.

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- Gama, J., Carvalho, A., Faceli, K., and Lorena, A. (2017). *Extração de Conhecimento de Dados - Data Mining*. Sílabo, 3 edition.
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- Japkowicz, N. and Shah, M., editors (2011). *Evaluating Learning Algorithms: A Classification Perspective*. Cambridge University Press.
- Witten, I. H. and Frank, E. (2005). *Data Mining: Practical machine learning tools and techniques*. Morgan Kaufmann.