

PROGRAMA DE VERÃO DO LNCC Jornada de Ciência de Dados

Introduction to Data Mining



Eduardo Ogasawara http://eic.cefet-rj.br/~eogasawara

Why Data Mining?

- Big Data scenario:
 - The explosive growth of data: from terabytes to petabytes
 - Data collection and data availability
 - Automated data collection tools, database systems, Web
 - Major sources of abundant and diverse data
 - Business: Web, e-commerce, transactions
 - Science: sensors, astronomy, bioinformatics, simulation
 - Society and everyone: news, photos, videos, open data, IoT
- We are drowning in data but starving for knowledge!
- "Need is the mother of invention"
 - Data mining Automated analysis of massive data sets

What is Data Mining?

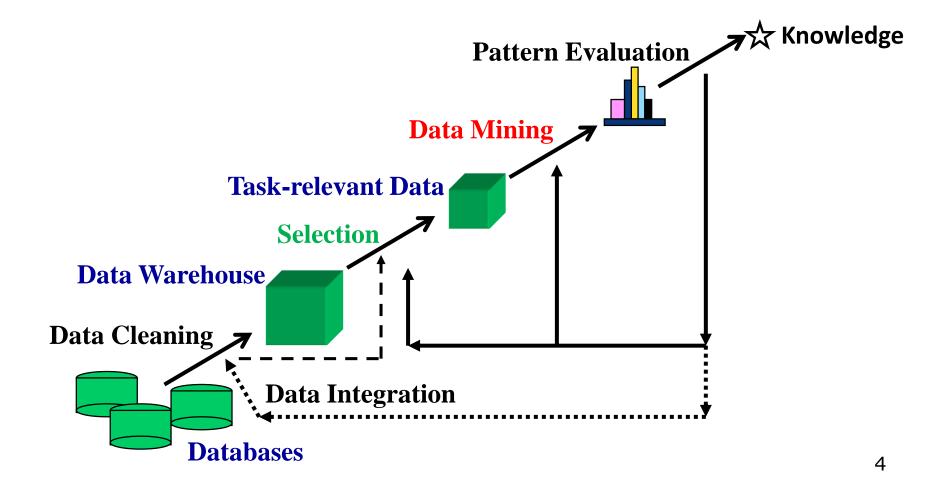


- Data mining (knowledge discovery from data)
 - Extraction of interesting (non-trivial, implicit, previously unknown and potentially useful) patterns or knowledge from a massive amount of data
- Alternative names
 - Knowledge Discovery in Databases (KDD)
 - Knowledge Extraction
 - Business Intelligence
 - Data Analysis

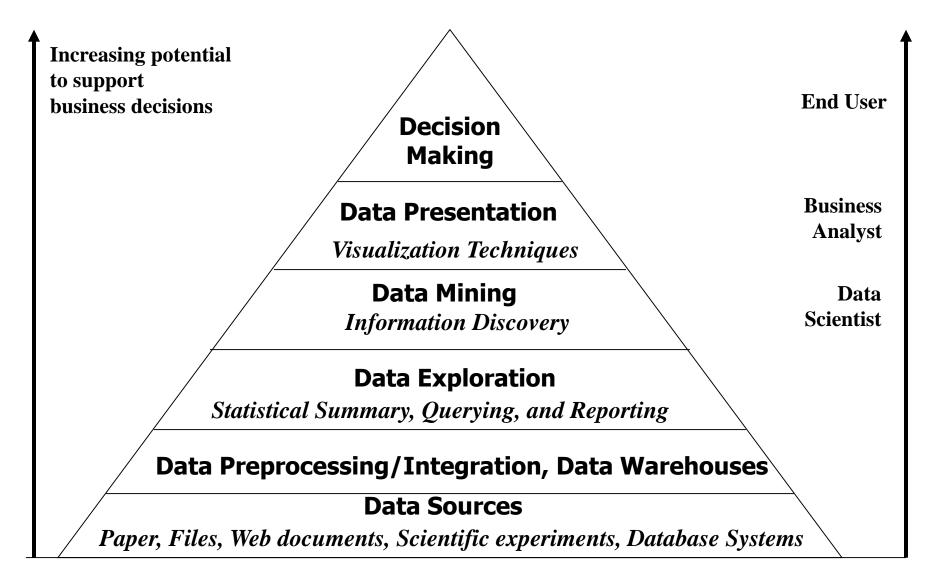


Knowledge discovery from data (KDD) process (Database perspective)

- This is a view from typical database systems
- Data mining plays an essential role in the KDD process

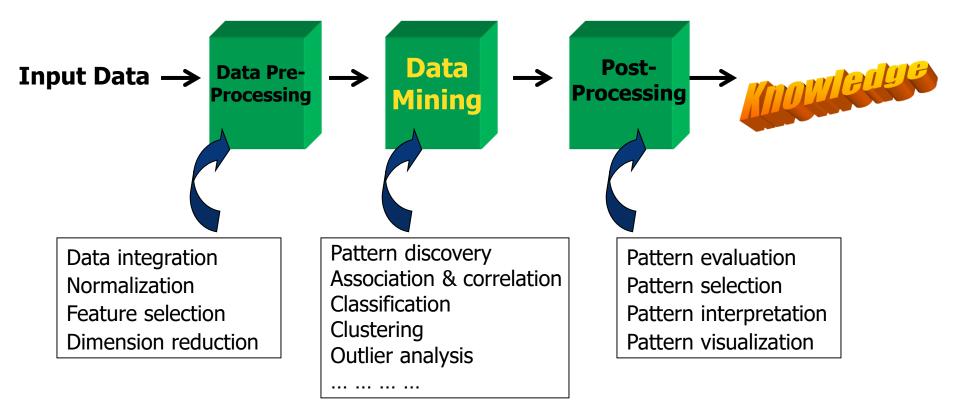


Data Mining (Business Intelligence perspective)



KDD process Machine/Statistical Learning perspective

This is a view from typical machine learning and statistics communities



Is everything "data mining"?

- Watch out:
 - Simple search and query processing
 - Continue (Deductive) expert systems
 - Data Mining vs. data exploration
 - Mining vs. OLAP vs. presentation tools
 - Warehouse, data cube, reporting but not much mining



Multi-Dimensional View of Data Mining

- Data models to be mined
- Knowledge to be mined
 - Data mining functions
- Techniques used
- Applications

Data Mining: on what kinds of data models?

- Database-oriented data sets and applications
 - A Relational database, data warehouse, transactional database
 - Object-relational databases, Heterogeneous databases, and legacy databases
- Advanced data sets and advanced applications
 - Data streams and sensor data
 - Time-series data, temporal data, sequence data (incl. bio-sequences)
 - Structure data, graphs, social networks, and information networks
 - Spatial data and spatiotemporal data
 - Multimedia database
 - Text databases (text mining)
 - The World-Wide Web

Data Mining Function: (1) Generalization

- Information integration and data warehouse construction
 - Data cleaning, transformation, integration, and multidimensional data model
- Data cube technology
 - Scalable methods for computing (i.e., materializing) multidimensional aggregates
 - OLAP (online analytical processing)
- Multidimensional concept description: characterization and discrimination
 - Generalize, summarize, and contrast data characteristics,
 - E.g.: dry vs. wet region (instead of pluviometry measures)

Data Mining Function: (2) Association and Correlation Analysis

- Frequent patterns (or frequent item sets)
 - What items are frequently purchased together in your supermarket?
- Association, correlation vs. causality
 - A typical association rule
 - Diaper → Beer [0.5%, 75%] (support, confidence)
 - Are strongly associated items also strongly correlated?
- How to mine such patterns and identify such rules efficiently in large datasets?
- How to use (rank) such patterns?

Data Mining Function: (3) Prediction

- Classification and label prediction
 - Construct models based on some training examples
 - Data-driven models
 - Describe and distinguish classes or concepts for future prediction
 - E.g., classify countries based on (climate)
- Typical methods:
 - Decision trees, naïve Bayesian classification, support vector machines, neural networks, rule-based classification, patternbased classification, logistic regression
- Typical applications:
 - classifying stars, predicting delays, forecasting diseases

Data Mining Function: (4) Cluster Analysis

- Unsupervised learning (i.e., Class label is unknown)
- Group data to form new categories (i.e., clusters), e.g., cluster houses to find distribution patterns
- Principle: Maximizing intra-class similarity & minimizing interclass similarity
- Many methods and applications

Data Mining Function: (5) Outlier Analysis

- Outlier analysis
 - Outlier: A data object that does not comply with the general behavior of the data
 - Noise or exception? One person's garbage could be another person's treasure
 - Methods: by-product of clustering or regression analysis
 - Useful in fraud detection, rare events analysis

Data Mining Function: (6) Sequential Pattern, Trend and Evolution Analysis

- Sequential pattern mining
 - e.g., buy a digital camera, then buy a large SD memory cards
- Time-series, trend, and deviation analysis:
 - e.g., regression and value prediction
- Periodicity analysis
- Motifs and biological sequence analysis
 - Approximate and consecutive motifs
- Similarity-based analysis

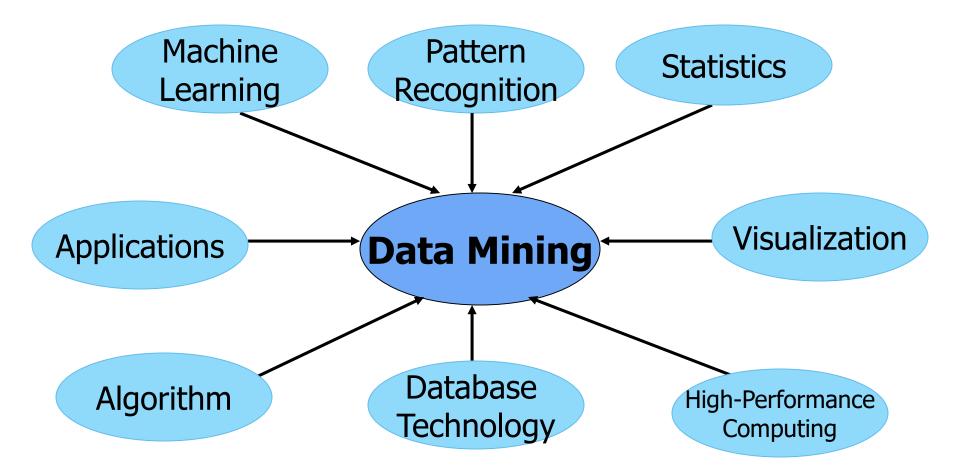
Data Mining Function: (7) Structure and Network Analysis

- Graph mining
 - Finding frequent subgraphs (e.g., chemical compounds), trees (XML), substructures (web fragments)
- Information network analysis
 - Social networks: actors (objects, nodes) and relationships (edges)
 - e.g., Web community discovery
- Web mining
 - Opinion mining, topic detection, sentiment analysis

Evaluation of Knowledge

- Are all mined knowledge interesting?
 - One can mine the tremendous amount of "patterns"
 - Some may fit only certain dimension space (time, location)
 - Some may not be representative, may be transient
- Evaluation of mined knowledge → directly mine only interesting knowledge?
 - Descriptive vs. Predictive vs. Prescriptive
 - Accuracy
 - Coverage
 - Typicality vs. novelty vs. rarity

Data Mining: Confluence of Multiple Disciplines



Why Confluence of Multiple Disciplines?

- Tremendous amount of data
 - Algorithms must be scalable to handle big data
- High-dimensionality of data
 - Micro-array may have tens of thousands of dimensions
- High complexity of data
 - Data streams, sensor data, spatial-temporal, text, multimedia
- New and sophisticated applications

Availability of data

- Do you have access to the data?
- Can you use the data?
- Can you publish your results?
- Is it big enough to be considered data mining?

Applications of Data Mining

- Consider all data mining functions
- The question might be:
 - What areas are difficult to do Data Mining?

Major Issues in Data Mining (1)

- Mining Methodology
 - Mining knowledge in multi-dimensional space
 - Boosting the power of discovery in a networked environment
 - Handling noise, uncertainty, and incompleteness of data
 - Pattern evaluation and constraint-guided pattern mining
- User Interaction
 - Interactive mining
 - Incorporation of background knowledge
 - Presentation and visualization of data mining results

Major Issues in Data Mining (2)

- Efficiency and Scalability
 - Efficiency and scalability of data mining algorithms
 - Parallel, distributed, stream, and incremental mining methods
- Diversity of data types
 - Handling complex types of data
 - Mining dynamic, networked, and global data repositories
- Data mining and society
 - Social impacts of data mining
 - Privacy-preserving data mining

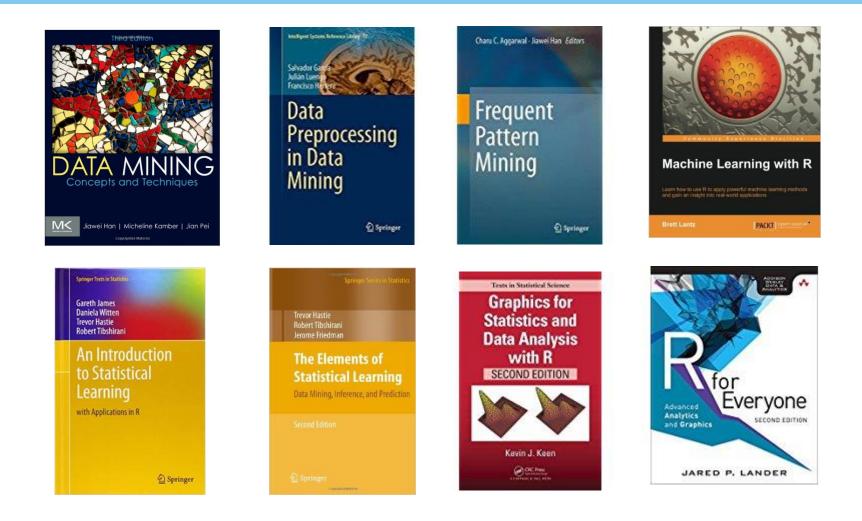
Where to publish?

- Data mining and KDD
 - Conferences: SIGKDD, IEEE-ICDM, SIAM-DM, PKDD.
 - Journal: Data Mining and Knowledge Discovery, KDD Explorations, ACM TKDD
- Database systems
 - Conferences: SIGMOD, PODS, VLDB, IEEE-ICDE, EDBT, ICDT, SSDBM
 - Journals: IEEE-TKDE, VLDB J., Info. Sys.
- AI & Machine Learning
 - Conferences: IJCNN, ML, AAAI, IJCAI, NIPS
 - Journals: Machine Learning, Artificial Intelligence, Knowledge and Information Systems
- Statistics
 - Journals: Journal of Applied Statistics, Annals of Data Science

Trending Data Mining Languages/Frameworks

- Python (Machine Learning Course)
- R (Data Mining Course)
- Spark (Parallel and Distributed Computing)

Main References



Most of the slides were extracted from Data Mining Concepts and Techniques