

Data Mining

Elena Baralis, Paolo Garza
Politecnico di Torino

Data analysis

- Most companies own huge databases containing
 - operational data
 - textual documents
 - experiment results
- These databases are a potential source of useful information

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Data analysis

- Information is "hidden" in huge datasets
 - not immediately evident
 - human analysts need a large amount of time for the analysis
 - most data *is never analyzed at all*

From R. Grossman, C. Kamath, V. Kumar, "Data Mining for Scientific and Engineering Applications"

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Data science

"Extracting meaning from very large quantities of data"

D.J. Patil coined the word *data scientist*

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Data mining

- Non trivial extraction of
 - implicit
 - previously unknown
 - potentially useful information from available data
- Extraction is automatic
 - performed by appropriate algorithms
- Extracted information is represented by means of abstract models
 - denoted as *pattern*

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Example: profiling

- Consumer behavior in e-commerce sites
 - Selected products, requested information, ...
- Search engines and portals
 - Query keywords, searched topics and objects
- Social network data
 - Facebook, google+ profiles
 - Dynamic data: posts on blogs, FB, tweets
- Maps and georeferenced data
 - Localization, interesting locations for users

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Example: profiling

- User/service profiling
 - Recommendation systems
 - Advertisements
- Market basket analysis
 - Correlated objects for cross selling
 - User registration, fidelity cards
- Context-aware data analysis
 - Integration of different dimensions
 - E.g., location, time of the day, user interest
- Text mining
 - Brand reputation, sentiment analysis, topic trends

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Example: biological data

- Microarray
 - expression level of genes in a cellular tissue
 - various types (mRNA, DNA)
- Patient clinical records
 - personal and demographic data
 - exam results
- Textual data in public collections
 - heterogeneous formats, different objectives
 - scientific literature (PubMed)
 - ontologies (Gene Ontology)

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Biological analysis objectives

- Clinical analysis
 - detecting the causes of a pathology
 - monitoring the effect of a therapy

⇒ diagnosis improvement and definition of new specific therapies
- Bio-discovery
 - gene network discovery
 - analysis of multifactorial genetic pathologies
- Pharmacogenesis
 - lab design of new drugs for genic therapies

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Knowledge Discovery Process

KDD = Knowledge Discovery from Data

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Preprocessing

- data cleaning
 - reduces the effect of noise
 - identifies or removes outliers
 - solves inconsistencies
- data integration
 - reconciles data extracted from different sources
 - integrates metadata
 - identifies and solves data value conflicts
 - manages redundancy

Real world data is "dirty"
Without good quality data, no good quality pattern

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Analysis techniques

- Descriptive methods
 - Extract interpretable models describing data
 - Example: client segmentation
- Predictive methods
 - Exploit some known variables to predict unknown or future values of (other) variables
 - Example: "spam" email detection

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Classification

- Objectives
 - prediction of a class label
 - definition of an interpretable model of a given phenomenon

training data → model → unclassified data → classified data

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Classification

- Approaches
 - decision trees
 - bayesian classification
 - classification rules
 - neural networks
 - k-nearest neighbours
 - SVM

training data → model → unclassified data → classified data

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Classification

- Requirements
 - accuracy
 - interpretability
 - scalability
 - noise and outlier management

training data → model → unclassified data → classified data

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Classification

- Applications
 - detection of customer propensity to leave a company (churn or attrition)
 - fraud detection
 - classification of different pathology types
 - ...

training data → model → unclassified data → classified data

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Clustering

- Objectives
 - detecting groups of similar data objects
 - identifying exceptions and outliers

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Clustering

- Approaches
 - partitional (K-means)
 - hierarchical
 - density-based (DBSCAN)
 - SOM
- Requirements
 - scalability
 - management of
 - noise and outliers
 - large dimensionality
 - interpretability

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Clustering

- Applications
 - customer segmentation
 - clustering of documents containing similar information
 - grouping genes with similar expression pattern
 - ...

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Association rules

- Objective
 - extraction of frequent correlations or pattern from a transactional database

Tickets at a supermarket counter

TID	Items
1	Bread, Coke, Milk
2	Beer, Bread
3	Beer, Coke, Diapers, Milk
4	Beer, Bread, Diapers, Milk
5	Coke, Diapers, Milk
...	...

- Association rule
 - diapers \Rightarrow beer
 - 2% of transactions contains both items
 - 30% of transactions containing diapers also contain beer

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Association rules

- Applications
 - market basket analysis
 - cross-selling
 - shop layout or catalogue design

Tickets at a supermarket counter

TID	Items
1	Bread, Coca Cola, Milk
2	Beer, Bread
3	Beer, Coca Cola, Diapers, Milk
4	Beer, Bread, Diapers, Milk
5	Coca Cola, Diapers, Milk
...	...

- Association rule
 - diapers \Rightarrow beer
 - 2% of transactions contains both items
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Other data mining techniques

- Sequence mining
 - ordering criteria on analyzed data are taken into account
 - example: motif detection in proteins
- Time series and geospatial data
 - temporal and spatial information are considered
 - example: sensor network data
- Regression
 - prediction of a continuous value
 - example: prediction of stock quotes
- Outlier detection
 - example: intrusion detection in network traffic analysis

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Open issues

- Scalability to **huge** data volumes
 - Big data
- Data dimensionality
- Complex data structures, heterogeneous data formats
- Data quality
- Privacy preservation
- Streaming data

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