

Data mining fundamentals



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Data analysis

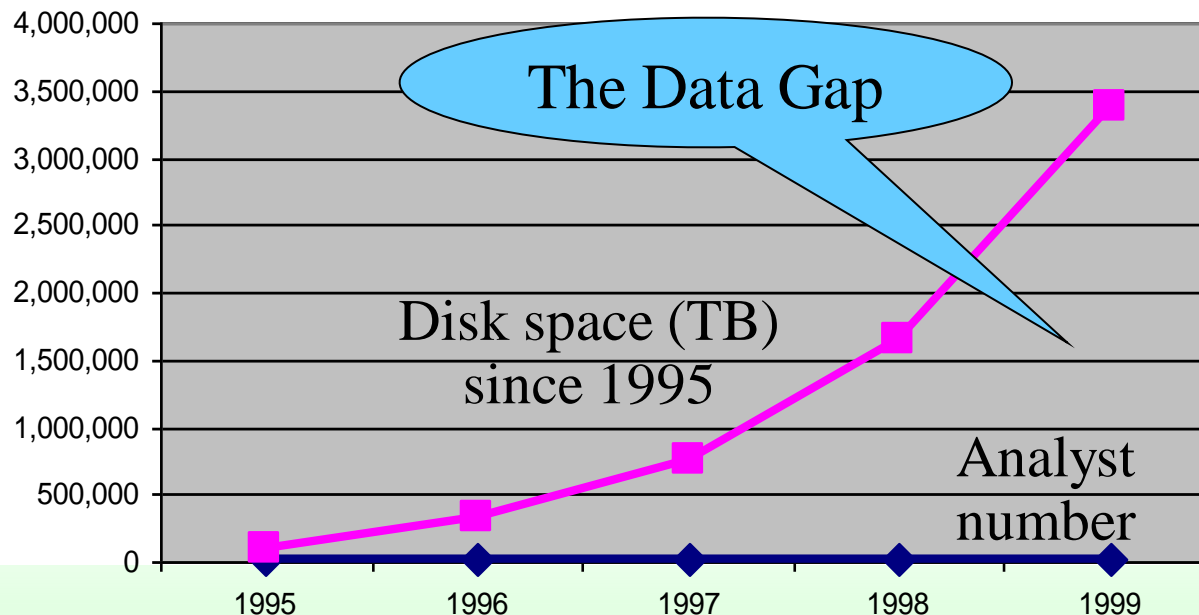
- Most companies own huge databases containing
 - operational data
 - textual documents
 - experiment results
- These databases are a potential source of useful information





Data analysis

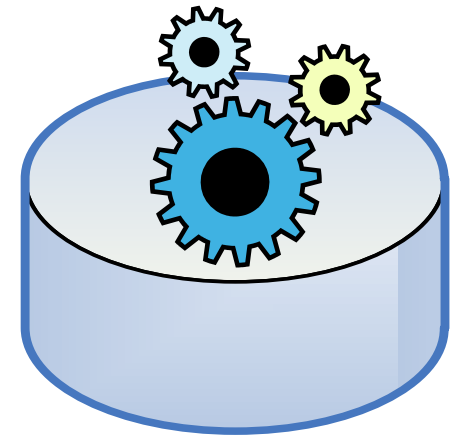
- Information is “hidden” in huge datasets
 - not immediately evident
 - human analysts need a large amount of time for the analysis
 - most data *is never analyzed at all*





Data mining

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

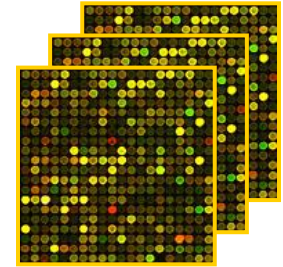




Example: biological data

■ Microarray

- expression level of genes in a cellular tissue
- various types (mRNA, DNA)



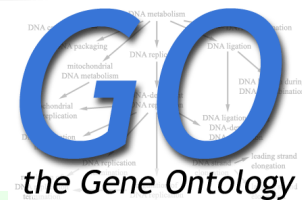
■ Patient clinical records

- personal and demographic data
- exam results

CLID	PATIENT ID	shx013: 49A34	shv060: 45A9	shq077: 52A28	shx009: 4A34	shx014: 61A31	shq082: 99A6	shq083: 46A15	shx008: 41A31
IMAGE:74	ISG20 in	-1.02	-2.34	1.44	0.57	-0.13	0.12	0.34	-0.51
IMAGE:76	TNFSF13	-0.52	-4.06	-0.29	0.71	1.03	-0.67	0.22	-0.09
IMAGE:36	LOC93343	-0.25	-4.08	0.06	0.13	0.08	0.06	-0.08	-0.05
IMAGE:23	ITGA4 in	-1.375	-1.605	0.155	-0.015	0.035	-0.035	0.505	-0.865

■ Textual data in public collections

- heterogeneous formats, different objectives
- scientific literature (PubMed)
- ontologies (Gene Ontology)





Biological analysis objectives

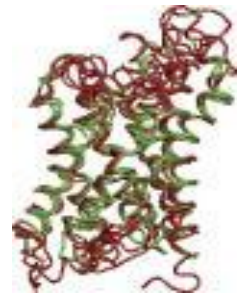
■ Clinical analysis

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

⇒ diagnosis improvement and definition of new specific therapies

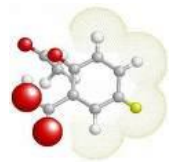
■ Bio-discovery

- gene network discovery
- analysis of multifactorial genetic pathologies



■ Pharmacogenesis

- lab design of new drugs for genic therapies

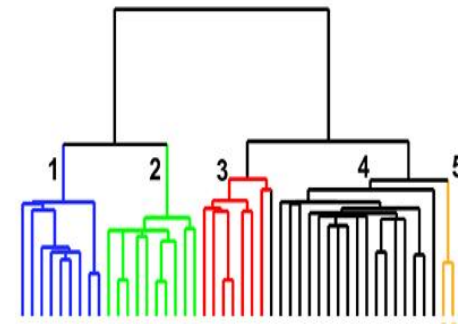
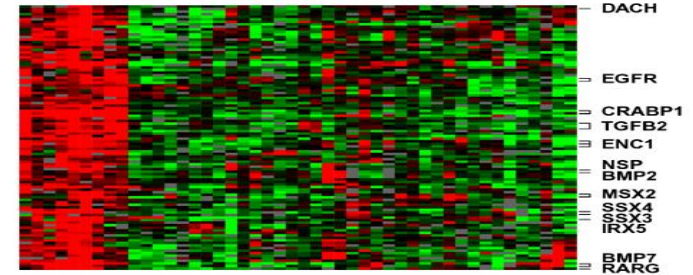


How can data mining contribute?



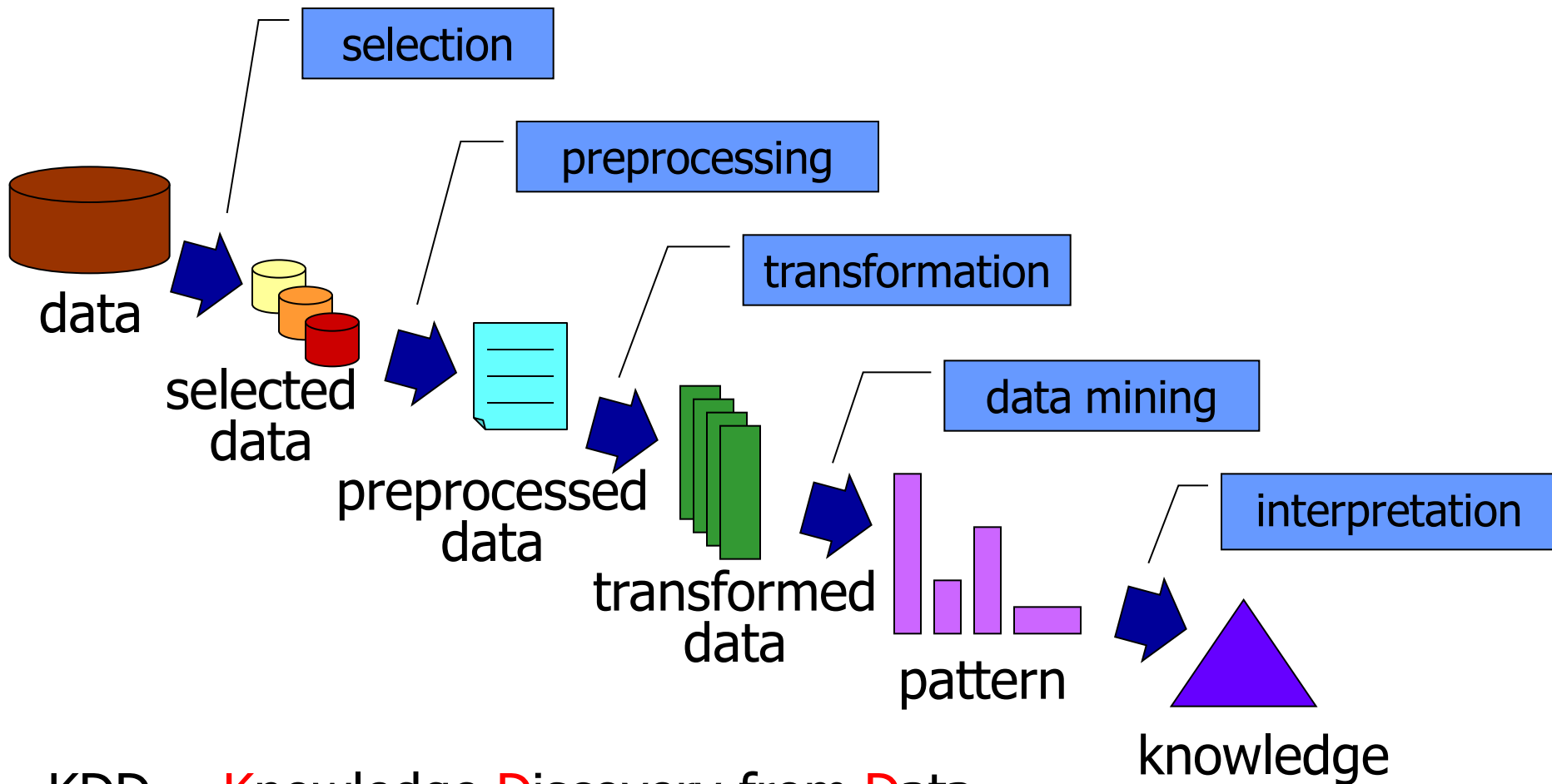
Data mining contributions

- Pathology diagnosis
 - classification
- Selecting genes involved in a specific pathology
 - feature selection
 - clustering
- Grouping genes with similar functional behavior
 - clustering
- Multifactorial pathologies analysis
 - association rules
- Detecting chemical components appropriate for specific therapies
 - classification





Knowledge Discovery Process

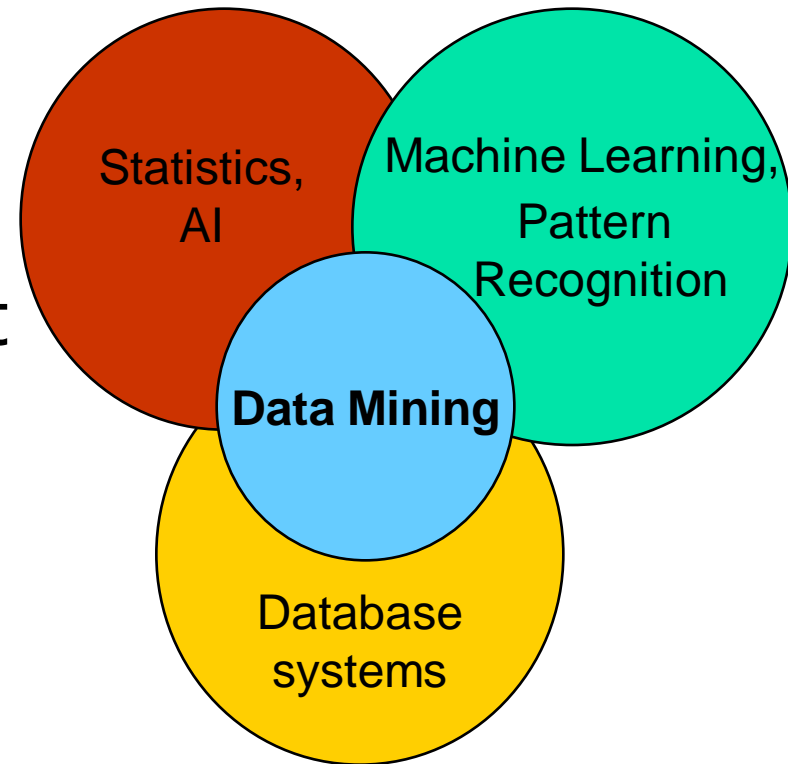


KDD = Knowledge Discovery from Data



Data mining origins

- Draws from
 - statistics, artificial intelligence (AI)
 - pattern recognition, machine learning
 - database systems
- Traditional techniques are not appropriate because of
 - significant data volume
 - large data dimensionality
 - heterogeneous and distributed nature of data



From: P. Tan, M. Steinbach, V. Kumar,
"Introduction to Data Mining"



Analysis techniques

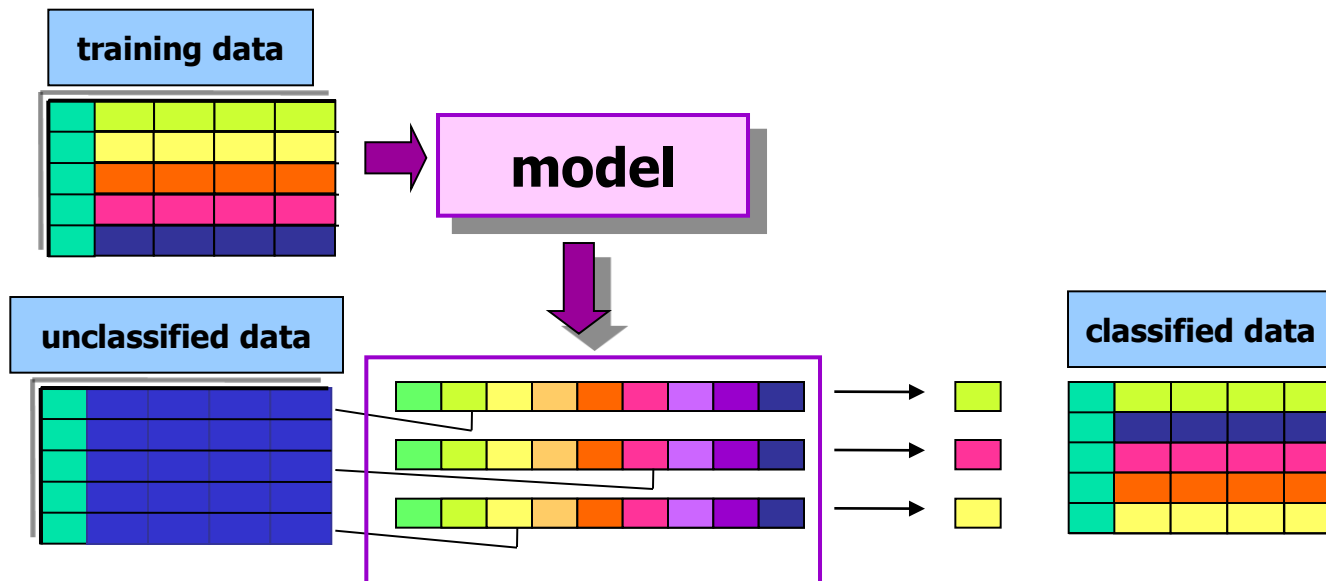
- Descriptive methods
 - Extract interpretable models describing data
 - Example: client segmentation
- Predictive methods
 - Exploit some known variables to predict unknown or future values of (other) variables
 - Example: “spam” email detection



Classification

■ Objectives

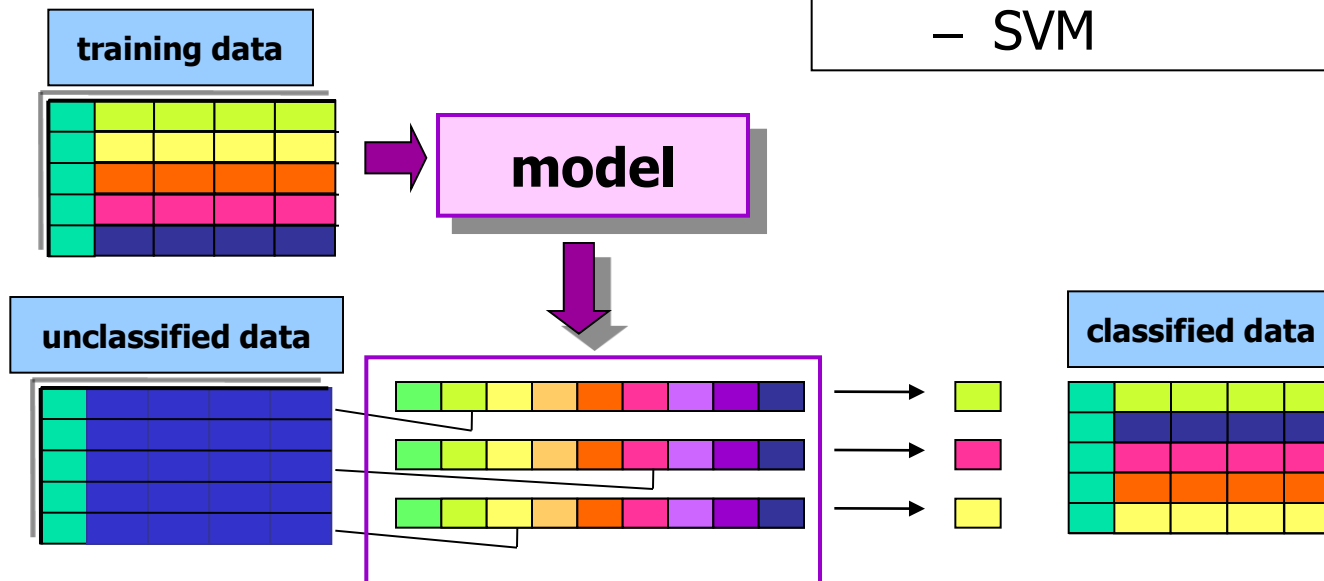
- prediction of a class label
- definition of an interpretable model of a given phenomenon





Classification

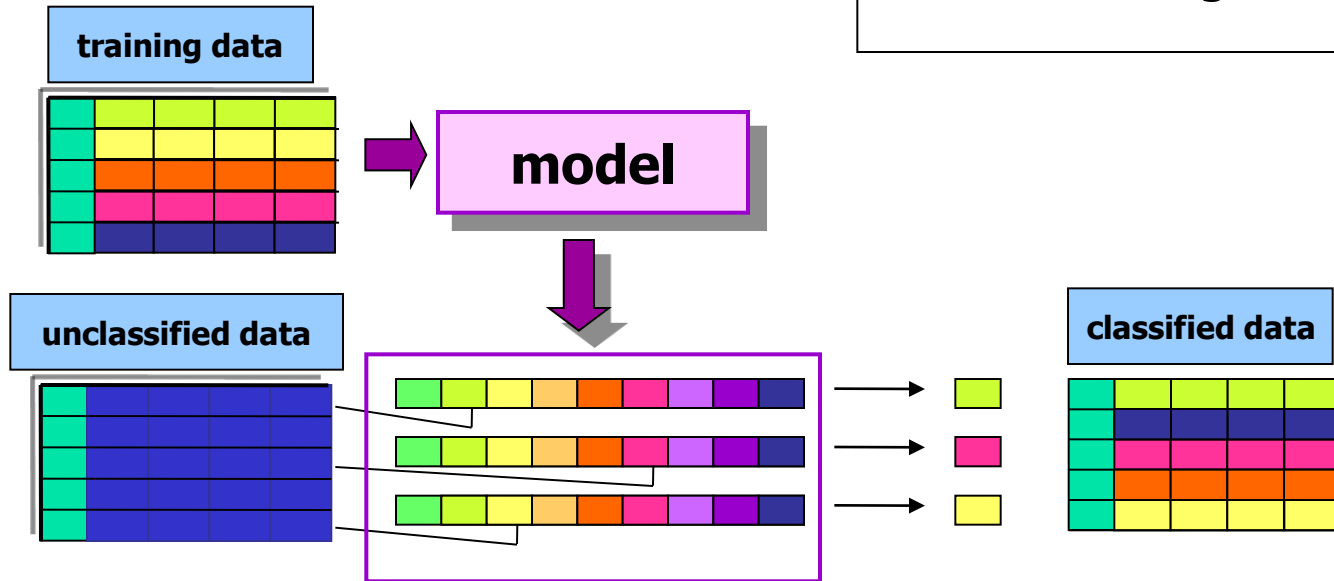
- Approaches
 - decision trees
 - bayesian classification
 - classification rules
 - neural networks
 - k-nearest neighbours
 - SVM





Classification

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

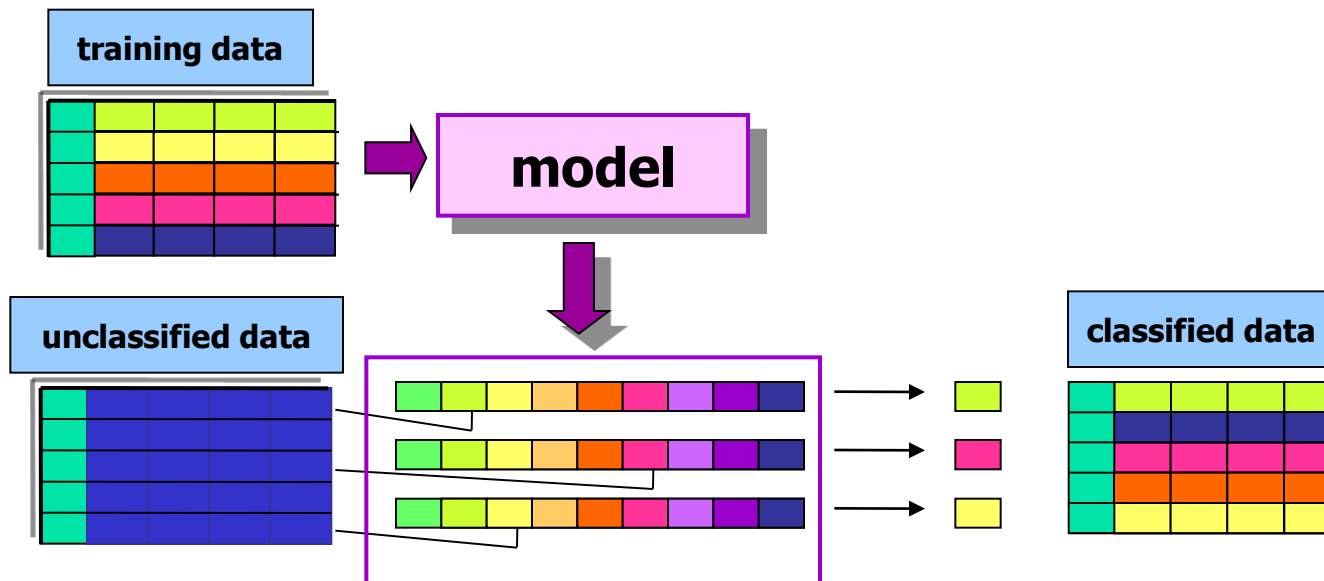




Classification

■ Applications

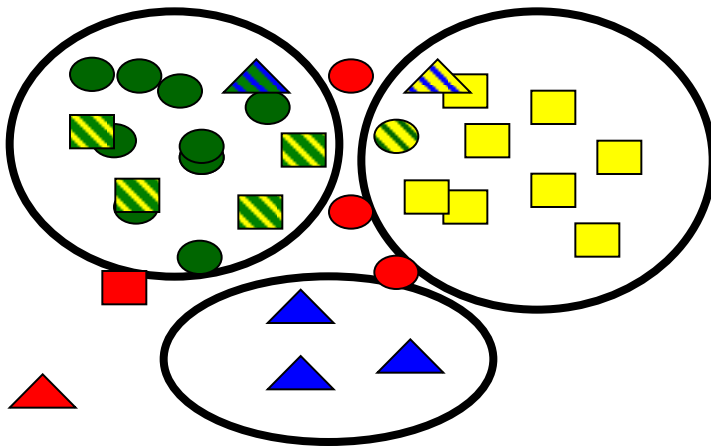
- detection of customer propension to leave a company (churn or attrition)
- fraud detection
- classification of different pathology types
- ...





Clustering

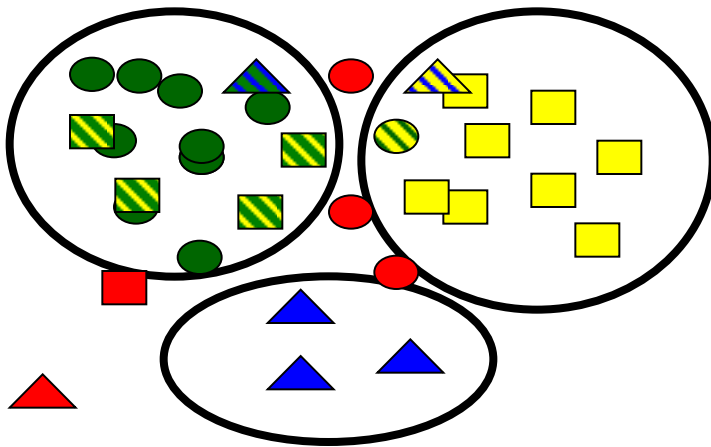
- Objectives
 - detecting groups of similar data objects
 - identifying exceptions and outliers





Clustering

- Approaches
 - partitional (K-means)
 - hierarchical
 - density-based (DBSCAN)
 - SOM



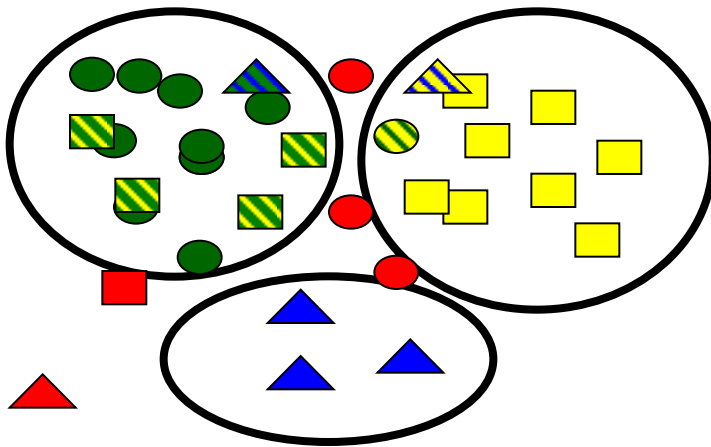
- Requirements
 - scalability
 - management of
 - noise and outliers
 - large dimensionality
 - interpretability



Clustering

■ Applications

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





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



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
...	...

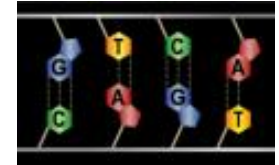
- Association rule
 - diapers \Rightarrow beer
 - 2% of transactions contains both items
 - 30% of transactions containing diapers also contain beer



Other data mining techniques

■ Sequence mining

- ordering criteria on analyzed data are taken into account
- example: motif detection in proteins



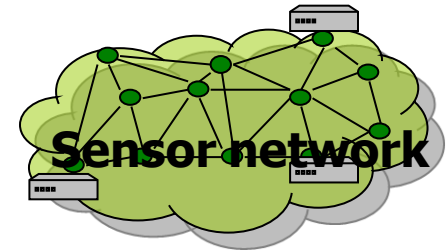
■ Time series and geospatial data

- 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





Open issues

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