

# Data Mining for Cryptographers

## T-79.514, September 17th 2003

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## About this presentation

We will discuss

- what is data mining (DM) in general
- some examples of application areas

We will not actually discuss

- specific relationships between DM and cryptography — hopefully revealed later
- specific methods and algorithms in DM — hopefully studied in context of their applications into privacy preserving DM; the choice of methods/algorithms is largely problem dependent

# Data mining is ...

(Material borrowed from H. Mannila's course notes)

“Data mining is the analysis of (often large) observational data sets to find **unsuspected** relationships and to **summarize** the data in novel ways that are both **understandable** and **useful** to the data analyst”

In short, the goal is to **obtain useful knowledge from large masses of data**

At the intersection of machine learning, statistics and databases

An umbrella covering several types of methods, algorithms, and practices. No clear borders enforced.

An applied area, driven by successful applications

# General characteristics of DM

## Warnings

- terminology is overlapping: DM/KDD? algorithm? etc.
- hype
- heuristics that work in practice and are thus useful
- no universal methods
- no correct answers

## Examples of discovered knowledge:

- association rules: “80% of customers who buy beer and sausage buy also mustard”
- rules: “if Age < 40 then Income < 10”
- models:  $Y = aX + b$
- clusterings: similar data points grouped together

# Data presentation

- as a matrix:
  - rows present different **observations**: customers in a transaction database, text documents, web pages etc.
  - columns present different **attributes** (also called **variables**) of the above observations: products, words, linked pages
- as a graph (can be written as a matrix)
- as a sequence/string such as telecom alarm log or DNA where events take place in time

## Aims of DM

- classification or clustering
- regression
- summarization
- dependency modeling



In each DM algorithm/method one must decide

- what to look for in the data (see the above list)
- how to measure the goodness of the estimated pattern or model
- what is the optimization method to maximise the goodness
- data handling; is the data in main memory during computation, or where

# Dichotomies

DM is about finding

- **patterns:** interesting (small) local features that
  - occur frequently, or
  - are outliers: rare and different.
  - patterns may be groups of attributes or observations
  - do not describe the whole data
  - fast algorithms exist
- **models:** global descriptions of the whole data
  - rules, clusters, joint density of attributes, prediction. . .
  - algorithms often slower

The algorithms are typically

- combinatorial (fast; for discrete data)
- probabilistic (gives a continuous description)

## DM versus KDD

Knowledge discovery in databases (KDD) is a more general term that includes DM as a subprocess: KDD is the total process of  
data selection → data preprocessing/transformation  
→ data mining → interpretation/evaluation.

Several steps of the above may be of interest to privacy preservation

For our purposes, we can use the terms DM and KDD interchangeably (this happens elsewhere, too)

# Current methodological challenges

- formulating the problem and hence choosing the method → algorithmic details not important
- large databases (number of observations and attributes) → computational efficiency is a must
- did we find anything useful? does our description generalize to other data sets or did we overfit?
- nonstationary data: possible tasks include
  - updating the description
  - detecting what is new in the data
- breaking the dichotomies presented earlier

# Hot applications (for PPDM, too?)

- link analysis of Web pages
  - hubs and authorities
  - spectral clustering of the link matrix
- personalization
- recommendation systems, collaborative filtering:  
e.g. movie ratings given by a large set of people  
are used to give recommendations
- fraud/anomaly detection
  - in telecommunication logs: is a client planning  
to avoid the phone bill?

- industrial etc. alarm logs: problems not voluntarily revealed in public
- bioinformatics

# Privacy preserving DM?

Privacy not an issue in several applications.

On the other hand, methods developed for one such application could be used in another where privacy is an issue