

Introduction to Data Mining

Data Mining (2018, Fall Semester)

Why Mine Data? Commercial Viewpoint

- ▶ Lots of data is being collected and warehoused

- ▶ Web data, e-commerce
- ▶ purchases at department/grocery stores
- ▶ Bank/Credit Card transactions

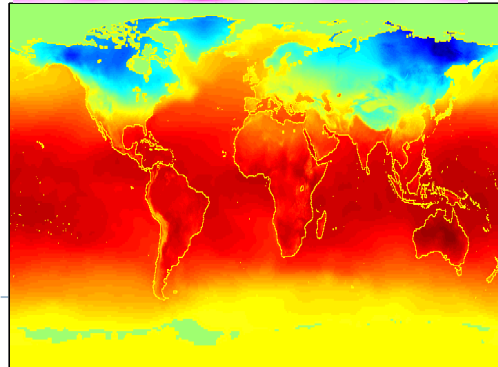
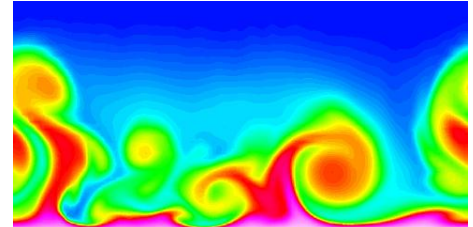
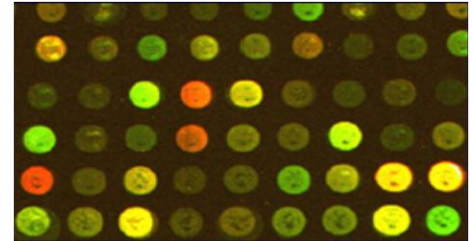
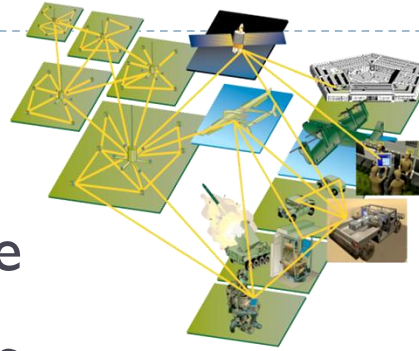


- ▶ Computers have become cheaper and more powerful
- ▶ Competitive Pressure is Strong
 - ▶ Provide better, customized services for an edge
 - ▶ e.g., Customer Relationship Management (CRM)



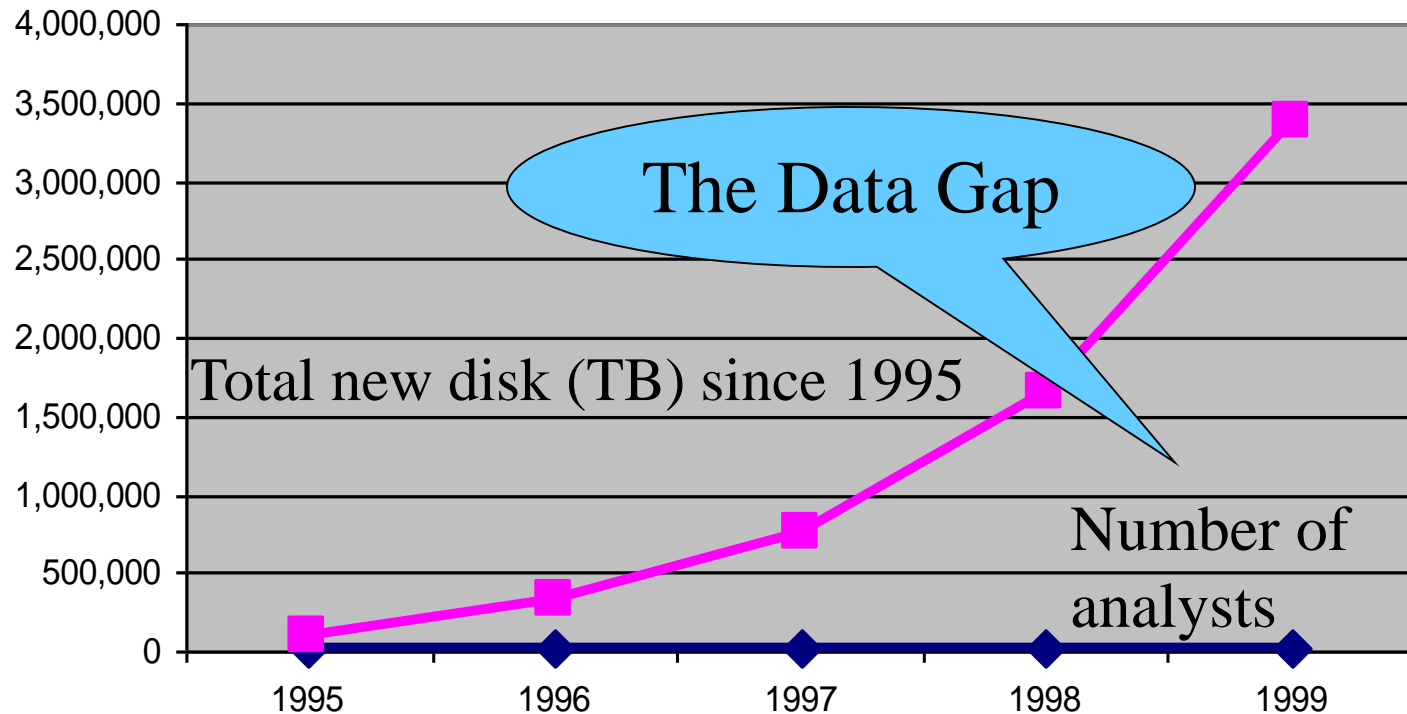
Why Mine Data? Scientific Viewpoint

- ▶ Data collected and stored at enormous speeds (GB/hour)
 - ▶ remote sensors on a satellite
 - ▶ telescopes scanning the skies
 - ▶ microarrays generating gene expression data
 - ▶ scientific simulations generating terabytes of data
- ▶ Traditional techniques infeasible for raw data
- ▶ Data mining may help scientists
 - ▶ in classifying and segmenting data
 - ▶ in Hypothesis Formation



Mining Large Data Sets - Motivation

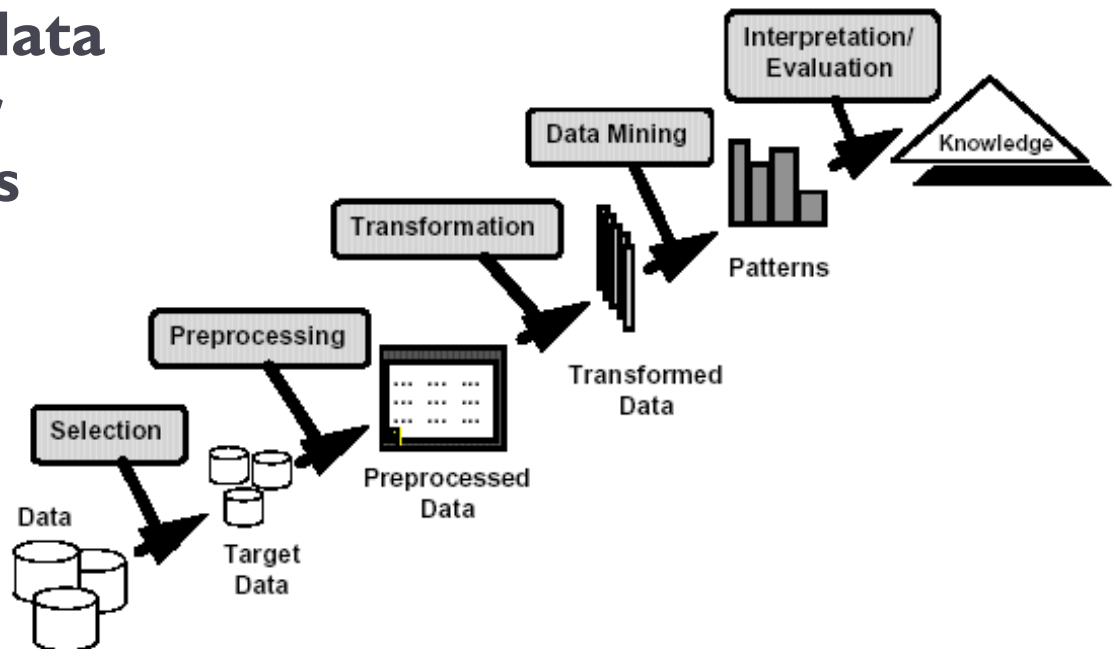
- ▶ There is often information “hidden” in the data that is not readily evident
- ▶ Human analysts may take weeks to discover useful information
- ▶ Much of the data is never analyzed at all



What is Data Mining?

▶ Many Definitions

- ▶ Non-trivial extraction of **implicit, previously unknown** and **potentially useful** information from data
- ▶ Exploration & analysis, by automatic or semi-automatic means, of large quantities of data in order to discover meaningful patterns



What is (not) Data Mining?

- What is not Data Mining?

- Look up phone number in phone directory
- Query a Web search engine for information about “Amazon”

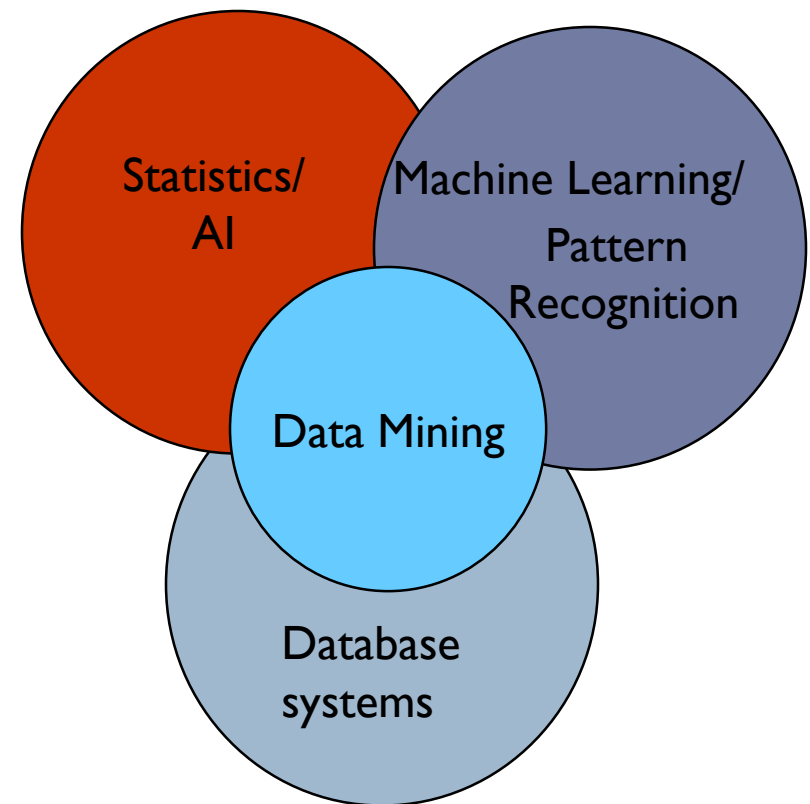
- What is Data Mining?

- Certain names are more prevalent in certain US locations (O’Brien, O’Rourke, O’Reilly... in Boston area)
- Group together similar documents returned by search engine according to their context (e.g. Amazon rainforest, Amazon.com,)



Origins of Data Mining

- ▶ **Draws ideas from machine learning/AI, pattern recognition, statistics, and database systems**
- ▶ **Traditional Techniques may be unsuitable due to**
 - ▶ **Enormity of data**
 - ▶ **High dimensionality of data**
 - ▶ **Heterogeneous, distributed nature of data**



Data Mining Tasks

- ▶ **Prediction Methods**

- ▶ Use some variables to predict unknown or future values of other variables.

- ▶ **Description Methods**

- ▶ Find human-interpretable patterns that describe the data.

From [Fayyad, et.al.] Advances in Knowledge Discovery and Data Mining, 1996



Data Mining Tasks...

- ▶ Classification [Predictive]
- ▶ Clustering [Descriptive]
- ▶ Association Rule Discovery [Descriptive]
- ▶ Regression [Predictive]
- ▶ Deviation Detection [Predictive]



Classification: Definition

- ▶ Given a collection of records (*training set*)
 - ▶ Each record contains a set of *attributes*, one of the attributes is the *class*.
- ▶ Find a *model* for class attribute as a function of the values of other attributes.
- ▶ Goal: previously unseen records should be assigned a class as accurately as possible.
 - ▶ A *test set* is used to determine the accuracy of the model. Usually, the given data set is divided into training and test sets, with training set used to build the model and test set used to validate it.

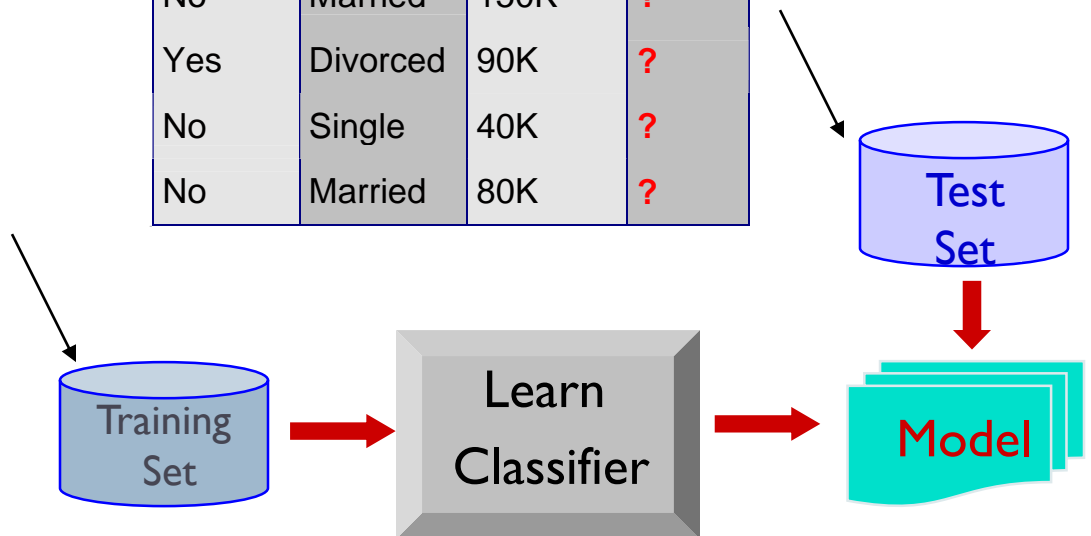


Classification Example

categorical categorical continuous class

<i>Tid</i>	Refund	Marital Status	Taxable Income	Cheat
1	Yes	Single	125K	No
2	No	Married	100K	No
3	No	Single	70K	No
4	Yes	Married	120K	No
5	No	Divorced	95K	Yes
6	No	Married	60K	No
7	Yes	Divorced	220K	No
8	No	Single	85K	Yes
9	No	Married	75K	No
10	No	Single	90K	Yes

Refund	Marital Status	Taxable Income	Cheat
No	Single	75K	?
Yes	Married	50K	?
No	Married	150K	?
Yes	Divorced	90K	?
No	Single	40K	?
No	Married	80K	?



Classification: Application

▶ Fraud Detection

- ▶ Goal: Predict fraudulent cases in credit card transactions.
- ▶ Approach:
 - ▶ Use credit card transactions and the information on its account-holder as attributes.
 - When does a customer buy, what does he buy, how often he pays on time, etc
 - ▶ Label past transactions as fraud or fair transactions. This forms the class attribute.
 - ▶ Learn a model for the class of the transactions.
 - ▶ Use this model to detect fraud by observing credit card transactions on an account.



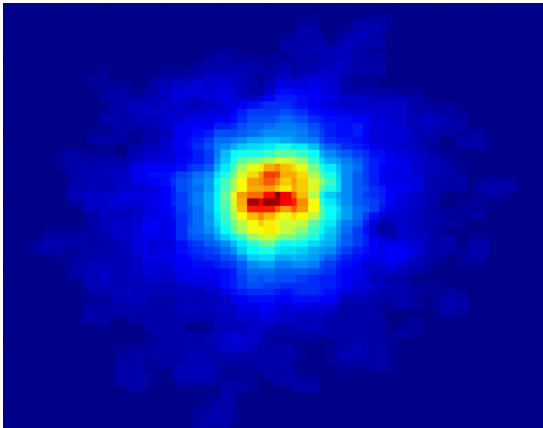
Classification: Application

- ▶ Sky Survey Cataloging
 - ▶ Goal: To predict class (star or galaxy) of sky objects, especially visually faint ones, based on the telescopic survey images (from Palomar Observatory).
 - 3000 images with 23,040 x 23,040 pixels per image.
 - ▶ Approach:
 - ▶ Segment the image.
 - ▶ Measure image attributes (features) - 40 of them per object.
 - ▶ Model the class based on these features.
 - ▶ Success Story: Could find 16 new high red-shift quasars, some of the farthest objects that are difficult to find!

Classifying Galaxies

Courtesy: <http://aps.umn.edu>

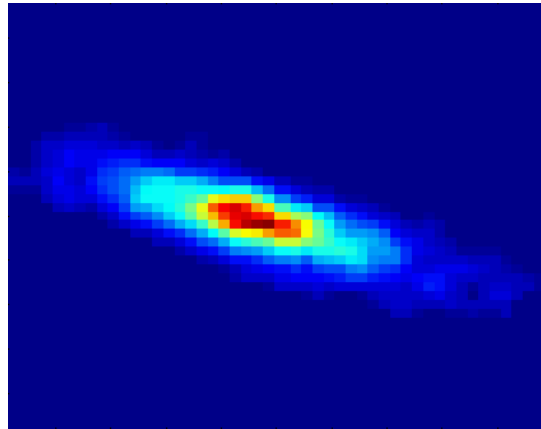
Early



Class:

- Stages of Formation

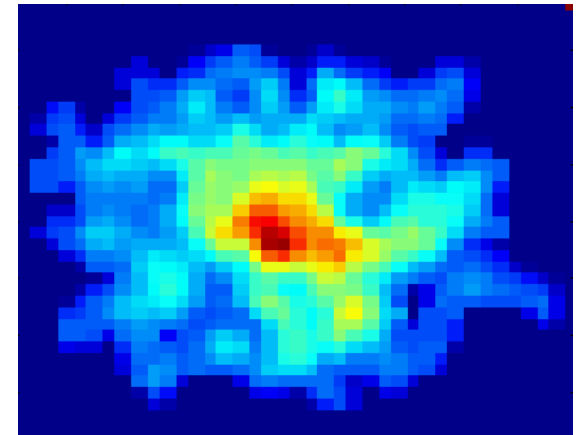
Intermediate



Attributes:

- Image features,
- Characteristics of light waves received, etc.

Late



Data Size:

- 72 million stars, 20 million galaxies
- Object Catalog: 9 GB
- Image Database: 150 GB

Clustering Definition

- ▶ Given a set of data points, each having a set of attributes, and a similarity measure among them, find clusters such that
 - ▶ Data points in one cluster are more similar to one another.
 - ▶ Data points in separate clusters are less similar to one another.
- ▶ **Similarity Measures:**
 - ▶ Euclidean Distance if attributes are continuous.
 - ▶ Other Problem-specific Measures.

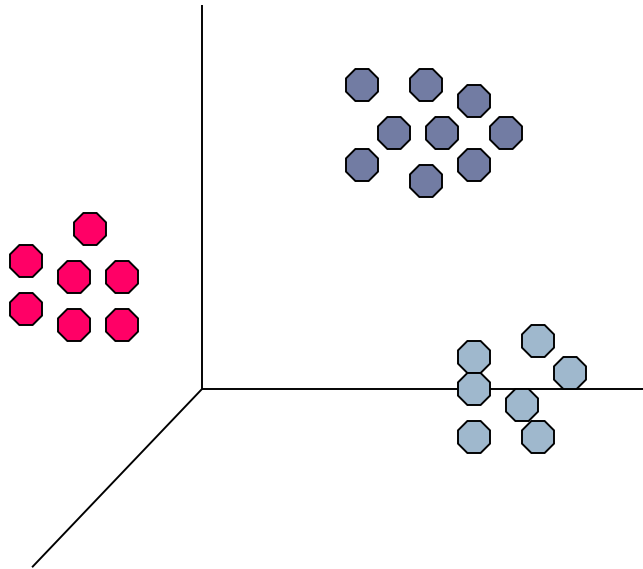


Illustrating Clustering

☒ Euclidean Distance Based Clustering in 3-D space.

Intracuster distances
are minimized

Intercluster distances
are maximized



Clustering: Application

▶ Document Clustering:

- ▶ Goal: To find groups of documents that are similar to each other based on the important terms appearing in them.
- ▶ Approach: To identify frequently occurring terms in each document. Form a similarity measure based on the frequencies of different terms. Use it to cluster.
- ▶ Gain: Information Retrieval can utilize the clusters to relate a new document or search term to clustered documents.



Illustrating Document Clustering

- ▶ Clustering Points: 3204 Articles of Los Angeles Times.
- ▶ Similarity Measure: How many words are common in these documents (after some word filtering).

<i>Category</i>	<i>Total Articles</i>	<i>Correctly Placed</i>
<i>Financial</i>	555	364
<i>Foreign</i>	341	260
<i>National</i>	273	36
<i>Metro</i>	943	746
<i>Sports</i>	738	573
<i>Entertainment</i>	354	278



Clustering of S&P 500 Stock Data

- ⌘ Observe Stock Movements every day.
- ⌘ Clustering points: Stock-{UP/DOWN}
- ⌘ Similarity Measure: Two points are more similar if the events described by them frequently happen together on the same day.
 - ⌘ We used association rules to quantify a similarity measure.

	<i>Discovered Clusters</i>	<i>Industry Group</i>
1	Applied-Matl-DOWN, Bay-Network-Down, 3-COM-DOWN, Cabletron-Sys-DOWN, CISCO-DOWN, HP-DOWN, DSC-Comm-DOWN, INTEL-DOWN, LSI-Logic-DOWN, Micron-Tech-DOWN, Texas-Inst-Down, Tellabs-Inc-Down, Natl-Semiconduct-DOWN, OracI-DOWN, SGI-DOWN, Sun-DOWN	Technology1-DOWN
2	Apple-Comp-DOWN, Autodesk-DOWN, DEC-DOWN, ADV-Micro-Device-DOWN, Andrew-Corp-DOWN, Computer-Assoc-DOWN, Circuit-City-DOWN, Compaq-DOWN, EMC-Corp-DOWN, Gen-Inst-DOWN, Motorola-DOWN, Microsoft-DOWN, Scientific-Atl-DOWN	Technology2-DOWN
3	Fannie-Mae-DOWN, Fed-Home-Loan-DOWN, MBNA-Corp-DOWN, Morgan-Stanley-DOWN	Financial-DOWN
4	Baker-Hughes-UP, Dresser-Inds-UP, Halliburton-HLD-UP, Louisiana-Land-UP, Phillips-Petro-UP, Unocal-UP, Schlumberger-UP	Oil-UP

Association Rule Discovery: Definition

- ▶ Given a set of records each of which contain some number of items from a given collection;
- ▶ Produce dependency rules which will predict occurrence of an item based on occurrences of other items.

<i>TID</i>	<i>Items</i>
1	Bread, Coke, Milk
2	Beer, Bread
3	Beer, Coke, Diaper, Milk
4	Beer, Bread, Diaper, Milk
5	Coke, Diaper, Milk

Rules Discovered:

$\{\text{Milk}\} \rightarrow \{\text{Coke}\}$

$\{\text{Diaper, Milk}\} \rightarrow \{\text{Beer}\}$

Association Rule Discovery: Application 1

- ▶ Marketing and Sales Promotion:
 - ▶ Let the rule discovered be
 $\{Bagels, \dots\} \rightarrow \{Potato\ Chips\}$
 - ▶ Potato Chips as consequent => Can be used to determine what should be done to boost its sales.
 - ▶ Bagels in the antecedent => Can be used to see which products would be affected if the store discontinues selling bagels.
 - ▶ Bagels in antecedent and Potato chips in consequent => Can be used to see what products should be sold with Bagels to promote sale of Potato chips!



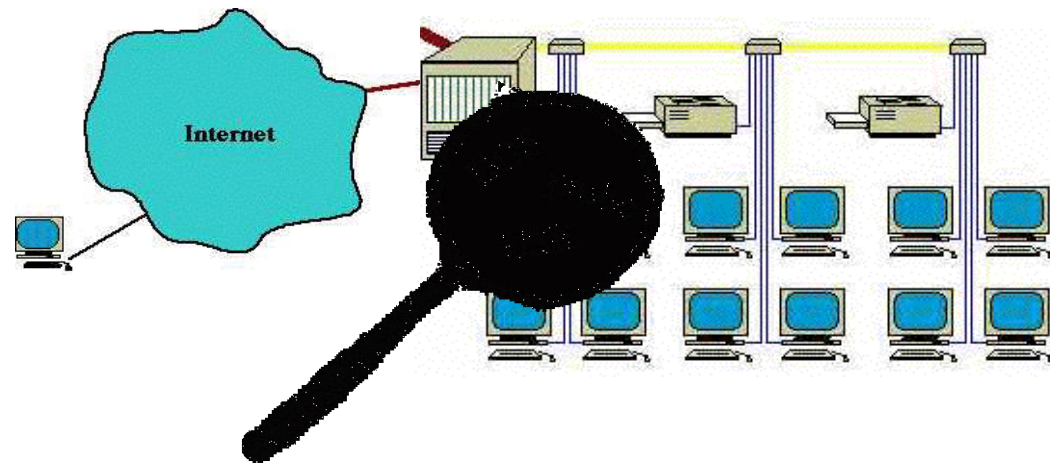
Regression

- ▶ Predict a value of a given continuous valued variable based on the values of other variables, assuming a linear or nonlinear model of dependency.
- ▶ Greatly studied in statistics, neural network fields.
- ▶ Examples:
 - ▶ Predicting sales amounts of new product based on advertising expenditure.
 - ▶ Predicting wind velocities as a function of temperature, humidity, air pressure, etc.
 - ▶ Time series prediction of stock market indices.



Deviation/Anomaly Detection

- ▶ Detect significant deviations from normal behavior
- ▶ Applications:
 - ▶ Credit Card Fraud Detection
 - ▶ Network Intrusion Detection



Typical network traffic at University level may reach over 100 million connections per day

Topics of Data Mining

- ▶ Types of Data, Exploring Data
- ▶ Concept Learning, Decision Tree Learning
- ▶ Linear Models for Classification and Regression
- ▶ Learning Algorithms, Artificial Neural Networks
- ▶ Evaluation Methods
- ▶ Clustering Methods
- ▶ Association Analysis
- ▶ Probabilistic Models
- ▶ Committee Machines, Deep Networks

