Data Mining: Introduction

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Course Outline

- Evaluation
 - Mid1 25 Marks
 - Mid2 25 Marks
 - Quiz 10 Marks
 - Term Paper 15 Marks
 - Endsem 25 Marks

Text Books

- Han, Jiawei, Jian Pei, and Micheline Kamber. Data mining: concepts and techniques.
- Pang-Ning Tan, Vipin Kumar, Michael Steinbach.
 Introduction to data mining.
- Anand Rajaraman, Jeffrey David Ullman. Mining of Massive Datasets.

Some research papers ...

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

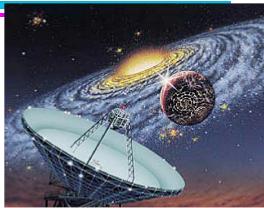


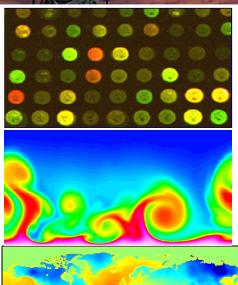
- Computers have become cheaper and more powerful
- Competitive Pressure is Strong
 - Provide better, customized services for an edge (e.g. in Customer Relationship Management)

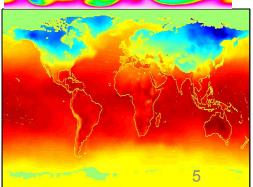
Why Mine Data? Scientific Viewpoint

- 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



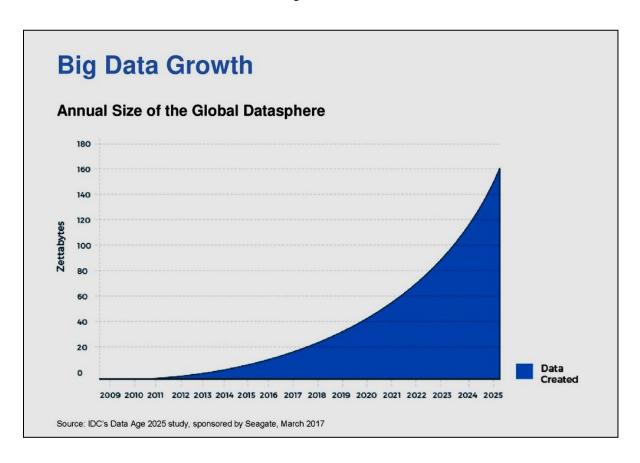






Mining Large Data Sets - Motivation

- 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





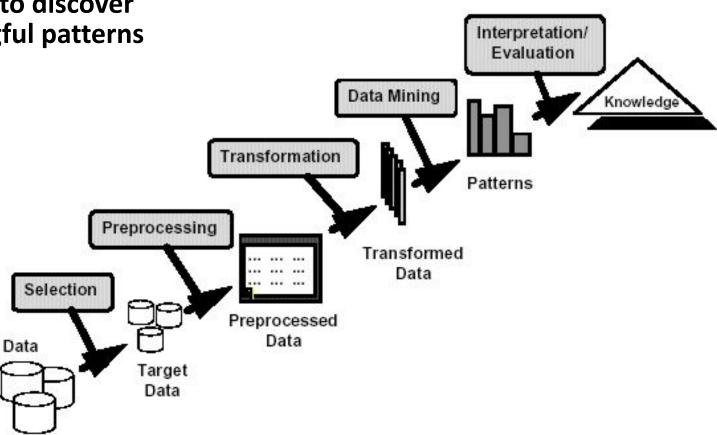
What Is Data Mining?

- Data mining (knowledge discovery in databases):
 - Extraction of interesting (<u>non-trivial</u>, <u>implicit</u>, <u>previously</u>
 <u>unknown</u> and <u>potentially useful</u>) information or patterns from data in <u>large databases</u>
- Alternative names and their "inside stories":
 - Data mining: a misnomer?
 - Knowledge discovery(mining) in databases (KDD), knowledge extraction, data/pattern analysis, data archeology, data dredging, information harvesting, business intelligence, etc.
- What is not data mining?
 - (Deductive) query processing.
 - Expert systems or small ML/statistical programs



What is Data Mining?

 Exploration & analysis, by automatic or semi-automatic means, of large quantities of data in order to discover meaningful patterns



What is (not) Data Mining?

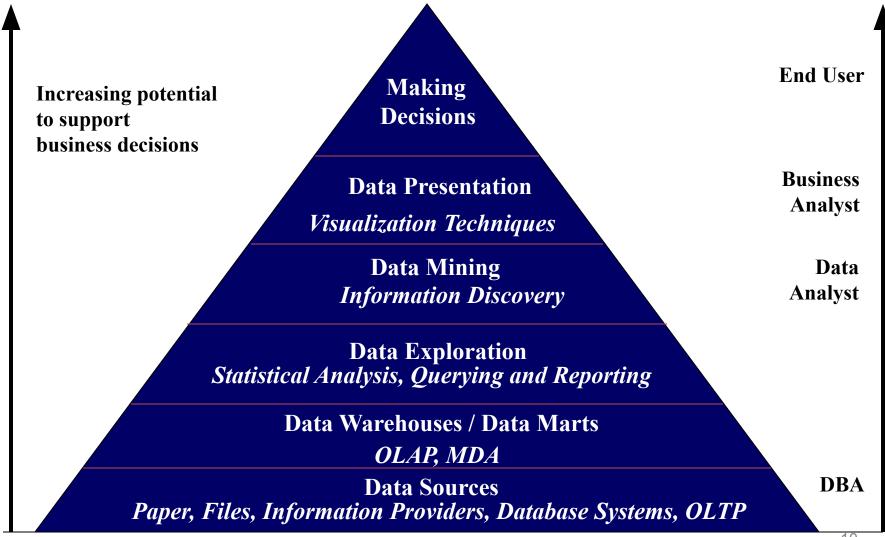
What is not Data Mining?

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

• What is Data Mining?

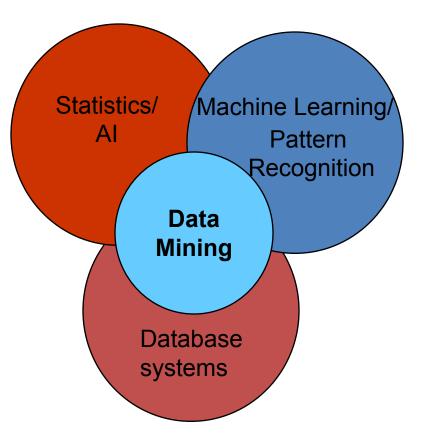
- Certain names are more prevalent in certain US locations (O'Brien, O'Rurke, O'Reilly... in Boston area)
- Group together similar documents returned by search engine according to their context (e.g. Amazon rainforest, Amazon.com,)

Data Mining and Business Intelligence



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.

Data Mining Tasks...

- Classification [Predictive]
- Clustering [Descriptive]
- Association Rule Discovery [Descriptive]
- Sequential Pattern Discovery [Descriptive]
- Regression [Predictive]
- Deviation Detection [Predictive] of the given data element
- Anomaly or novelty detection within a given data set [Descriptive]

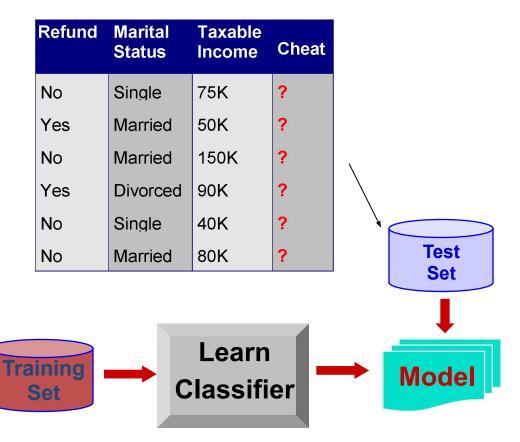
Classification: Definition

- Given a collection of records (training set)
 - Each record contains a set of attributes, one of the attributes is the class or decision.
- Find a *model* for class attribute as a function of the values of other attributes.
- Goal: <u>previously unseen</u> 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

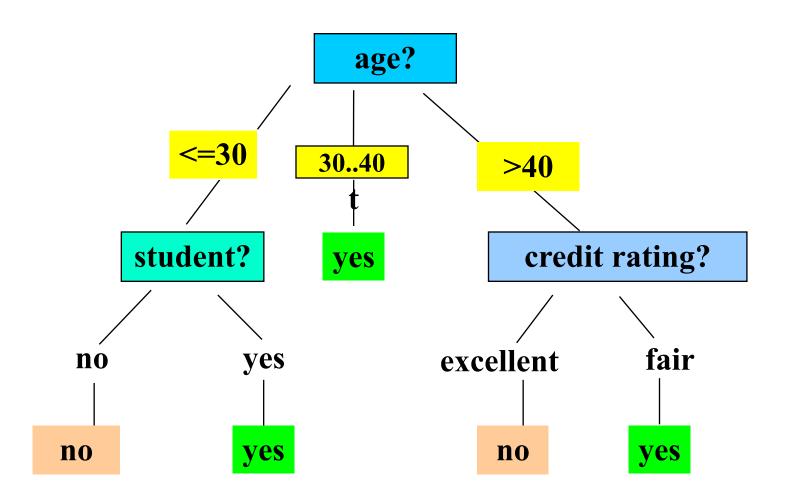
Tid	Refund	Marital Status	Taxable Income	Cheat
1	Yes	Single	125K	No
2	No	Married	100K	No
3	No	Single	70K	No
4	Yes	Married	120K	No
5	No	Divorced	95K	Yes
6	No	Married	60K	No
7	Yes	Divorced	220K	No
8	No	Single	85K	Yes
9	No	Married	75K	No
10	No	Single	90K	Yes



Training Dataset

age	income	student	cred rati	buys comp
<=30	high	no	fair	no
<=30	high	no	excellent	no
3140	high	no	fair	yes
>40	medium	no	fair	yes
>40	low	yes	fair	yes
>40	low	yes	excellent	no
3140	low	yes	excellent	yes
<=30	medium	no	fair	no
<=30	low	yes	fair	yes
>40	medium	yes	fair	yes
<=30	medium	yes	excellent	yes
3140	medium	no	excellent	yes
3140	high	yes	fair	yes
>40	medium	no	excellent	no

Output: A Decision Tree for "buys_computer"



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.

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 advetising 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 IntrusionDetection



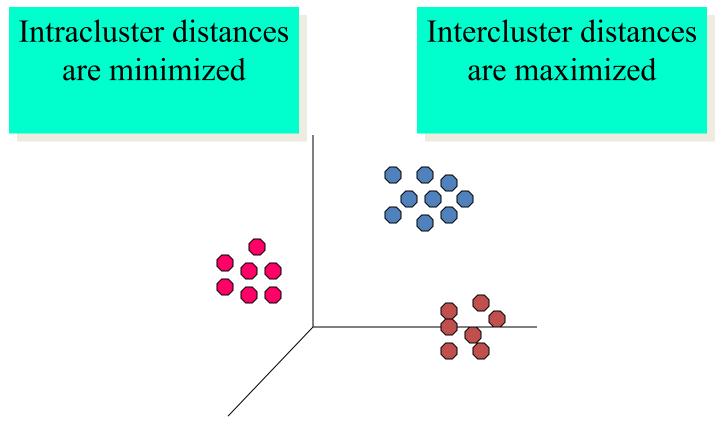
Typical network traffic at University level may reach over 100 million connections per day

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.



Clustering: Application

Market Segmentation:

 Goal: subdivide a market into distinct subsets of customers where any subset may conceivably be selected as a market target to be reached with a distinct marketing mix.

– Approach:

- Collect different attributes of customers based on their geographical and lifestyle related information.
- Find clusters of similar customers.
- Measure the clustering quality by observing buying patterns of customers in same cluster vs. those from different clusters.

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.

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.

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

```
Rules Discovered:

{Milk} --> {Coke}

{Diaper, Milk} --> {Beer}
```

Association Rule Discovery: Application

- Marketing and Sales Promotion:
 - Let the rule discovered be {Bagels, ...} --> {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!

Association Rule Discovery: Application

- Supermarket shelf management.
 - Goal: To identify items that are bought together by sufficiently many customers.
 - Approach: Process the point-of-sale data collected with barcode scanners to find dependencies among items.
 - A classic rule ---
 - If a customer buys diaper and milk, then he is very likely to buy beer.
 - So, don't be surprised if you find six-packs stacked next to diapers!

Association Rule Discovery: Application

- Inventory Management:
 - Goal: A consumer appliance repair company wants to anticipate the nature of repairs on its consumer products and keep the service vehicles equipped with right parts to reduce on number of visits to consumer households.
 - Approach: Process the data on tools and parts required in previous repairs at different consumer locations and discover the co-occurrence patterns.

Challenges of Data Mining

- Scalability
- Dimensionality
- Complex and Heterogeneous Data
- Data Quality
- Data Ownership and Distribution
- Privacy Preservation
- Streaming Data
- WWW
- Social Media