Data Mining: Introduction

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Introduction to Data Mining, 2nd Edition Chapter I



What is Data Mining?

It is the use of **efficient** techniques for the analysis of **very large collections of data** and the **extraction** of useful and possibly unexpected patterns in data (hidden knowledge).



Large-scale Data is Everywhere!

Enormous data growth in both commercial and scientific databases

 due to advances in data generation and collection technologies

New mantra

 Gather whatever data you can whenever and wherever possible

Expectations

 Gathered data will have value either for the purpose collected or for a purpose not envisioned.



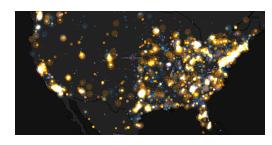
Cyber Security



E-Commerce



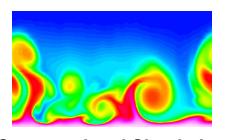
Traffic Patterns



Social Networking: Twitter



Sensor Networks



Computational Simulations



Why Data Mining? Commercial Viewpoint

- Lots of data is being collected and warehoused
 - Web data
 - Yahoo has Peta Bytes of web data
 - Facebook has billions of active users
 - purchases at department/ grocery stores, e-commerce
 - Amazon handles millions of visits/day
 - 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)











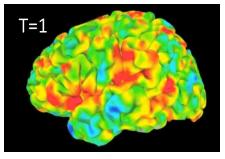
Why Data Mining? Scientific Viewpoint

Data collected and stored at enormous speeds

- remote sensors on a satellite
 - NASA EOSDIS archives over petabytes of earth science data / year
- telescopes scanning the skies
 - Sky survey data
- High-throughput biological data
- scientific simulations
 - terabytes of data generated in a few hours

Data mining helps scientists

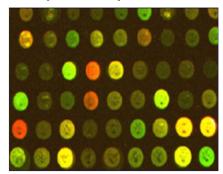
- in automated analysis of massive datasets
- In hypothesis formation



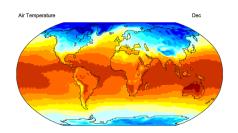
fMRI Data from Brain



Sky Survey Data



Gene Expression Data



Surface Temperature of Earth



Big data proxies of social life

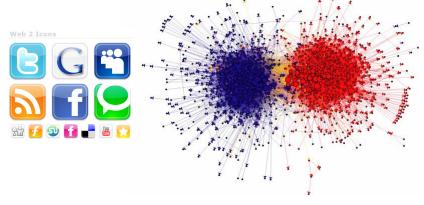
Shopping patterns & lyfestyle



DESIRES, OPINIONS, SENTIMENTS



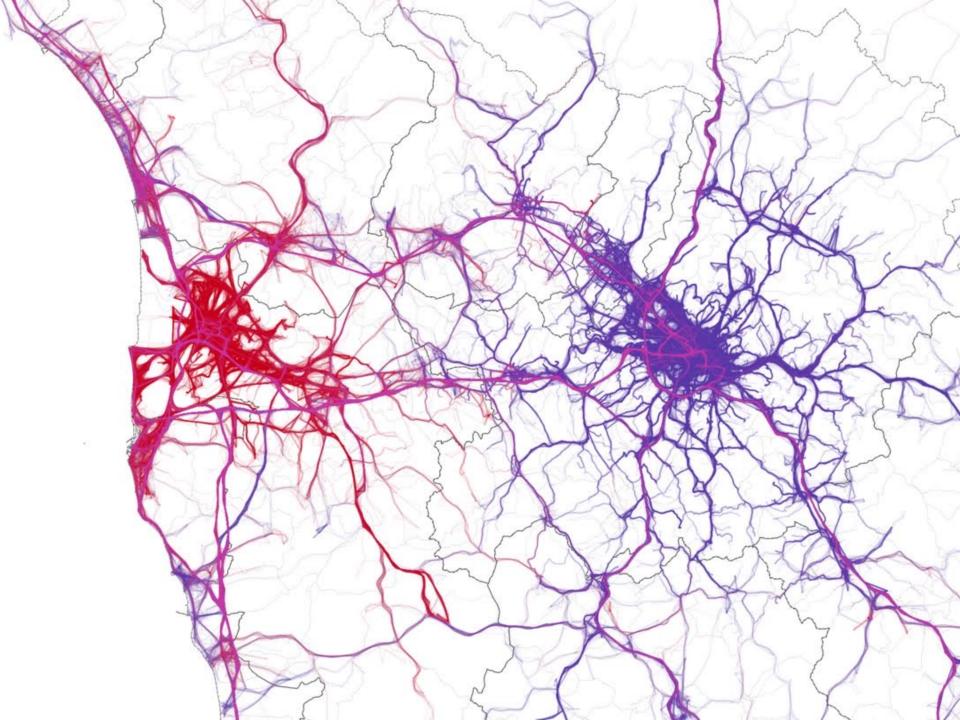
RELATIONSHIPS & SOCIAL TIES

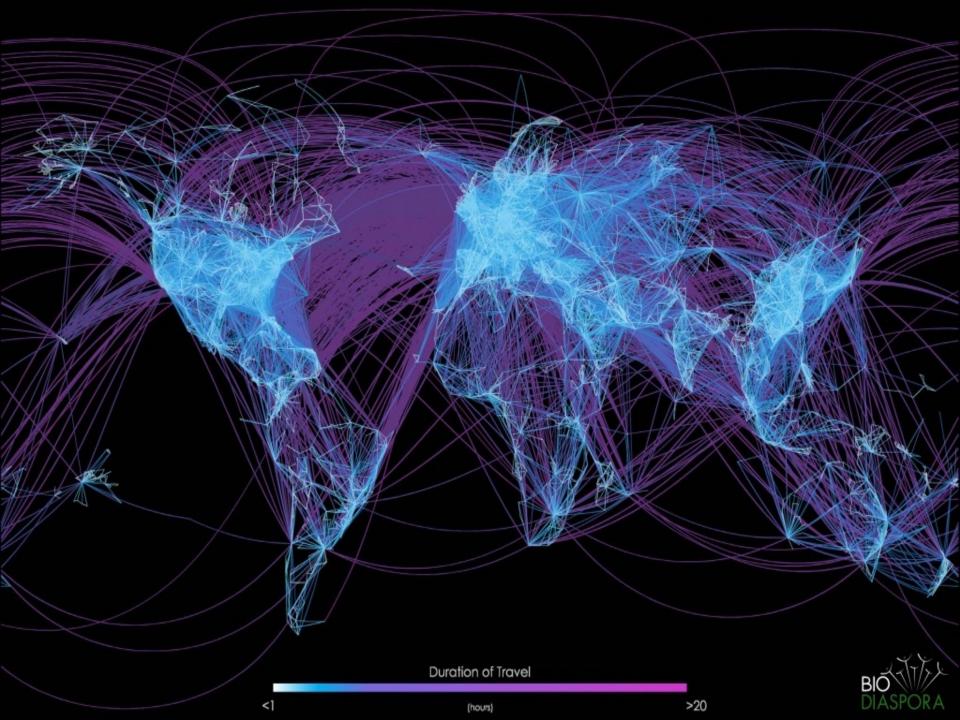


MOVEMENTS





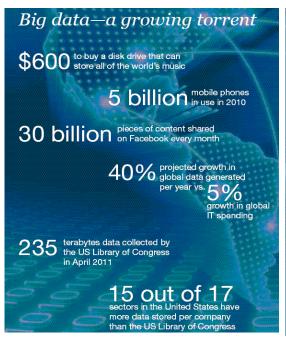


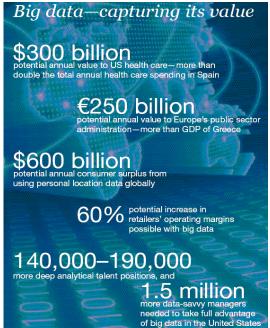


Great opportunities to improve productivity in all walks of life

McKinsey Global Institute

Big data: The next frontier for innovation, competition, and productivity.







Great Opportunities to Solve Society's Major Problems

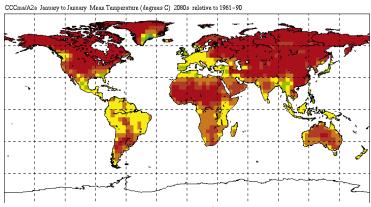


Improving health care and reducing costs



Finding alternative/ green energy sources

01/17/2018



Predicting the impact of climate change



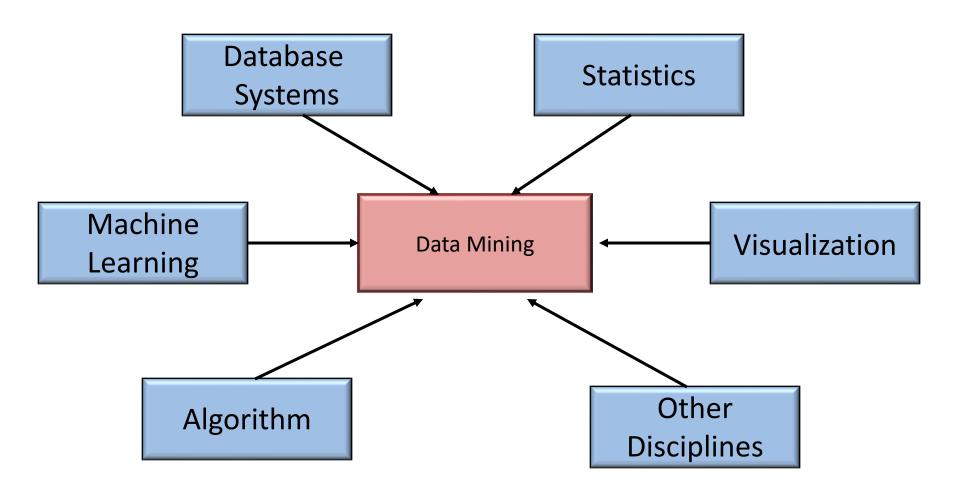
Reducing hunger and poverty by increasing agriculture production

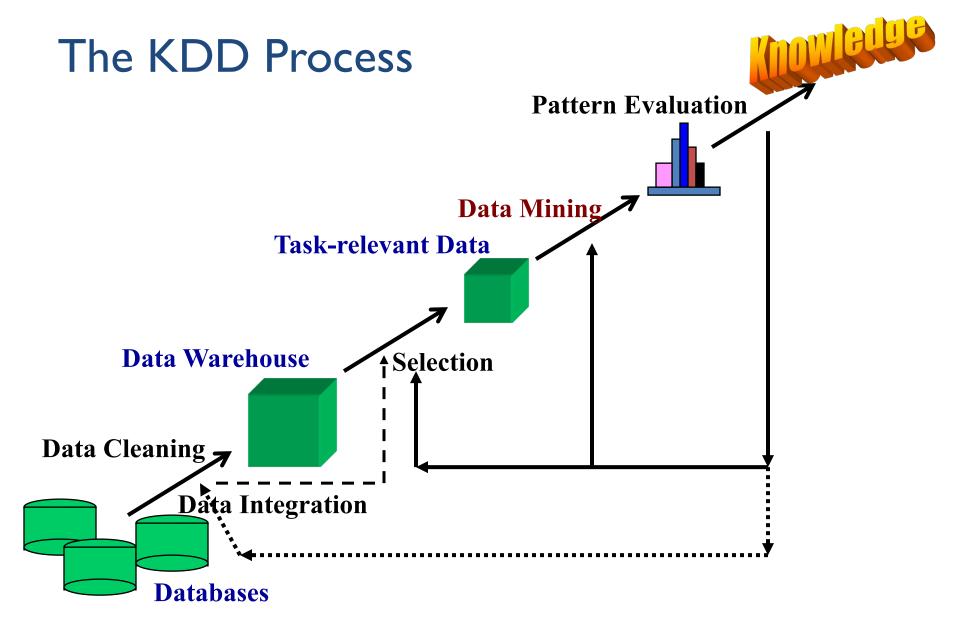
Università di Pisa

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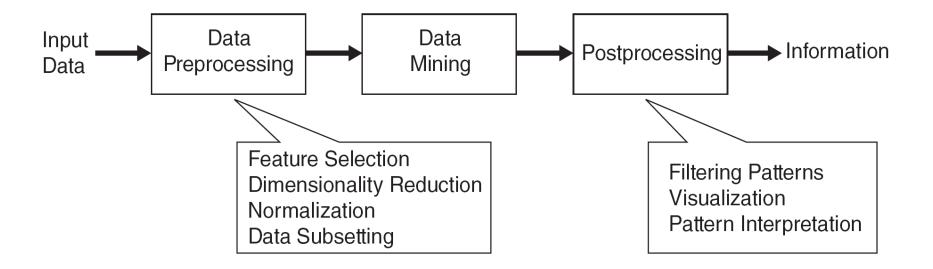
Data Mining: Confluence of Multiple Disciplines







What is Data Mining?



Primary & Secondary Data

Primary Data

- Original data that has been collected for a specific purpose
- Primary data is not altered by humans

Secondary Data

- Data that has been already collected and made available for other purposes
- Secondary data may be obtained from many sources

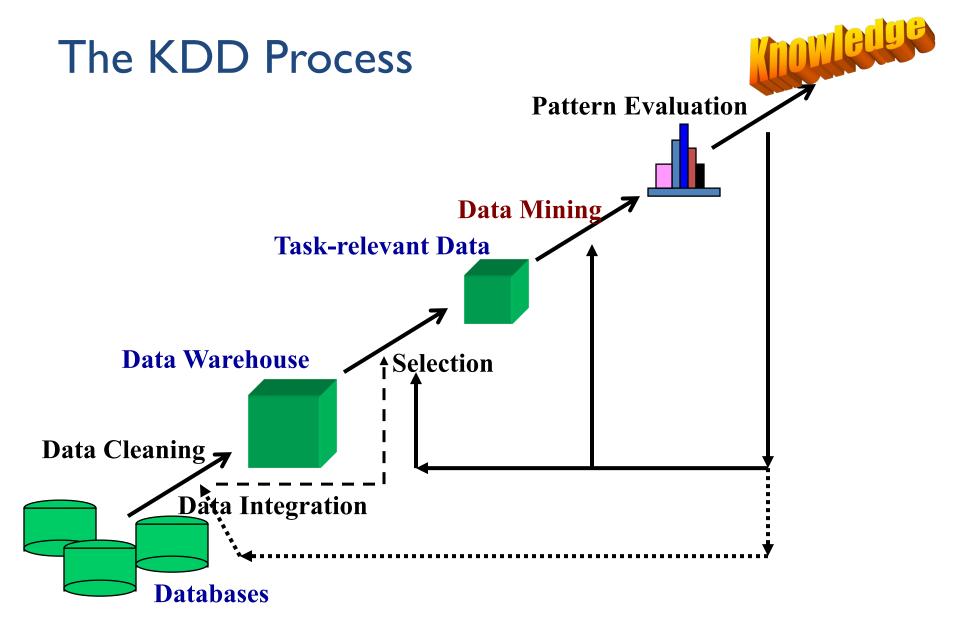




Varierty of Data Sources

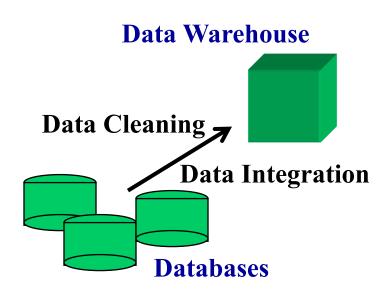








Data Integration and Preparation

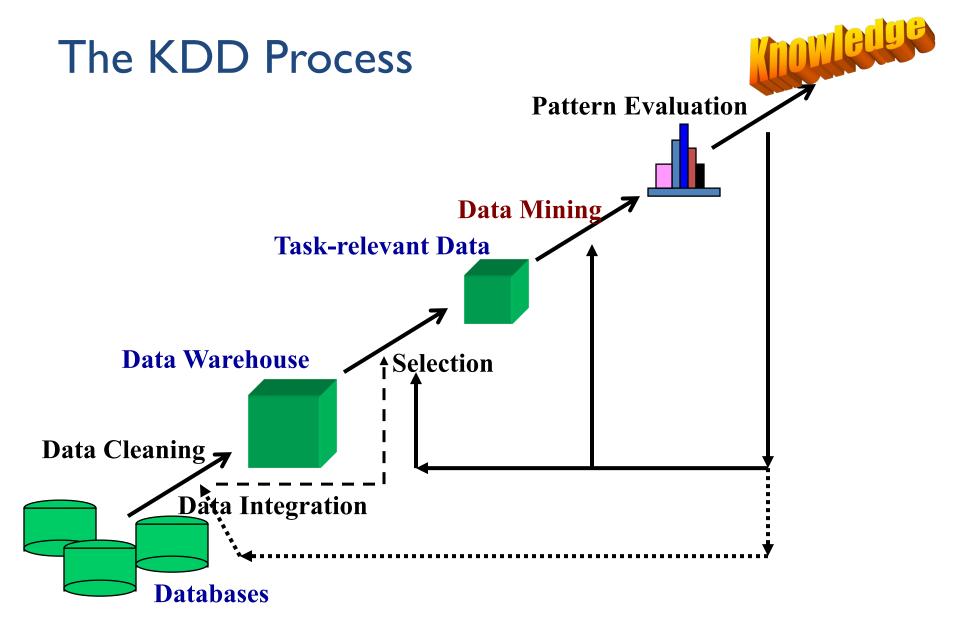


Data Integration involves the process of data understanding, data cleaning, merging data coming from multiple sources and transfoming them to load them into a Data Warehouse

Data Warehouse is a database targeted to answer specific business questions

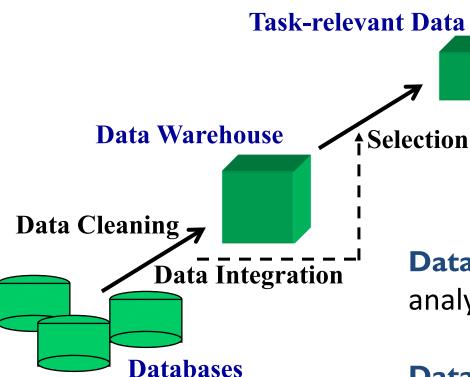
Developing a data analytics project requires the **BUSINESS UNDERSTANDING**







Data Selection and Transformation



Data Selection: Data relevant to analysis tasks are retrieved from data

Data transformation: Transform data into appropriate form for mining (summary, aggregation, etc.)



The KDD Process **Pattern Evaluation Data Mining Task-relevant Data Data Warehouse Selection Pattern Evaluation:** Identify truly interesting patterns Data Cleaning **Data Integration** Knowledge representation: Use visualization and knowledge

Databases



the mined data to the user

representation tools to present

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

Prediction Methods

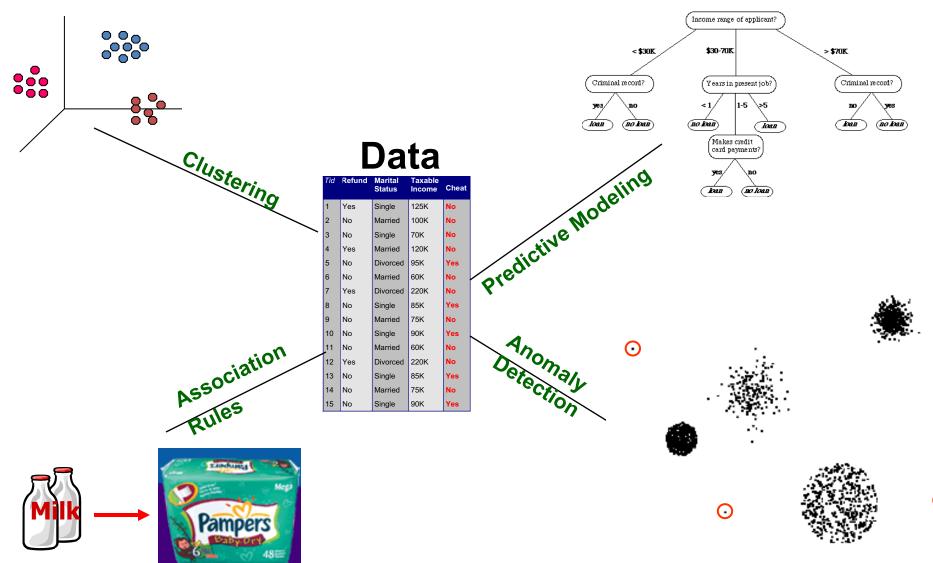
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Description Methods

Find human-interpretable patterns that describe the data.



Data Mining Tasks ...





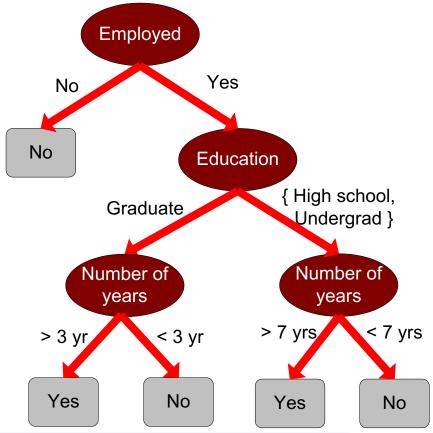
Predictive Modeling: Classification

 Find a model for class attribute as a function of the values of other attributes

Model for predicting credit worthiness

Class

Tid	Employed	Level of Education	# years at present address	Credit Worthy
1	Yes	Graduate	5	Yes
2	Yes	High School	2	No
3	No	Undergrad	1	No
4	Yes	High School	10	Yes
			•••	



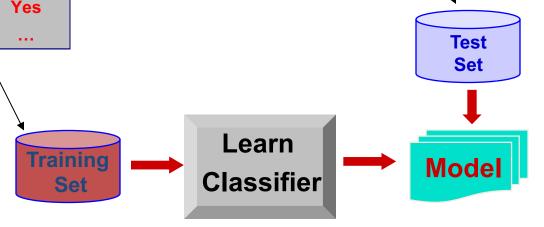


Classification Example



Tid	Employed	Level of Education	# years at present address	Credit Worthy
1	Yes	Graduate	5	Yes
2	Yes	High School	2	No
3	No	Undergrad	1	No
4	Yes	High School	10	Yes
	•••		•••	

Tid	Employed	Level of Education	# years at present address	Credit Worthy
1	Yes	Undergrad	7	?
2	No	Graduate	3	?
3	Yes	High School	2	?



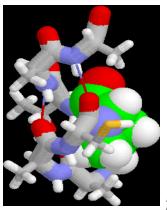


Examples of Classification Task

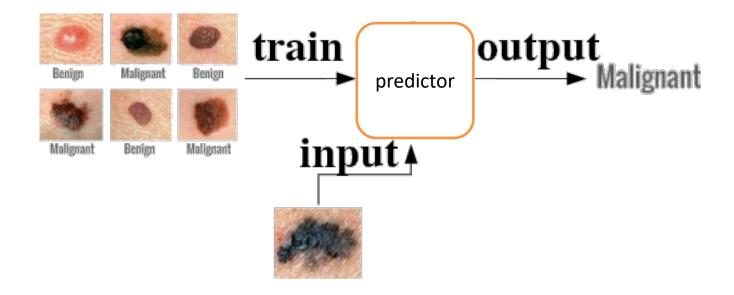
- Classifying credit card transactions as legitimate or fraudulent
- Classifying land covers (water bodies, urban areas, forests, etc.) using satellite data
- Categorizing news stories as finance, weather, entertainment, sports, etc
- Identifying intruders in the cyberspace
- Predicting tumor cells as benign or malignant
- Classifying secondary structures of protein as alpha-helix, beta-sheet, or random coil







Al = Machine Learning + Big Data



Classification: Application I

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 2

Churn prediction for telephone customers

 Goal: To predict whether a customer is likely to be lost to a competitor.

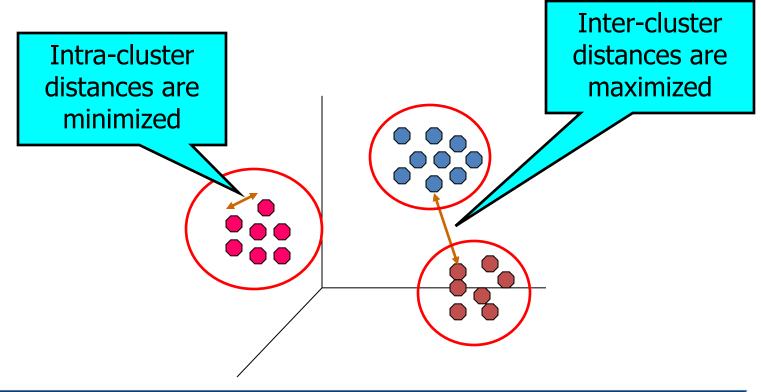
– Approach:

- Use detailed record of transactions with each of the past and present customers, to find attributes.
 - How often the customer calls, where he calls, what time-ofthe day he calls most, his financial status, marital status, etc.
- Label the customers as loyal or disloyal.
- Find a model for loyalty.



Clustering

 Finding groups of objects such that the objects in a group will be similar (or related) to one another and different from (or unrelated to) the objects in other groups





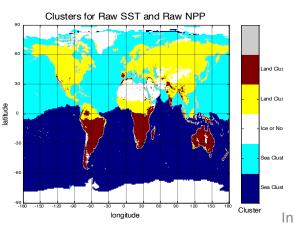
Applications of Cluster Analysis

Understanding

- Custom profiling for targeted marketing
- Group related documents for browsing
- Group genes and proteins that have similar functionality
- Group stocks with similar price fluctuations

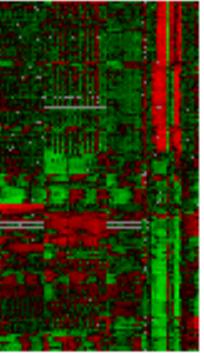
Summarization

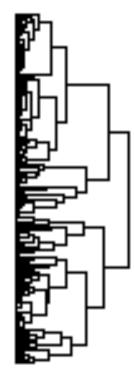
Reduce the size of large data sets



Use of K-means to partition Sea Surface Temperature (SST) and Net Primary Production (NPP) into clusters that reflect the Northern and Southern Hemispheres.

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Courtesy: Michael Eisen





Clustering: Application I

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.



A Behavior Based Segmentation

Using unsupervised clustering segmentation for a grocery chain which would like better product assortment for its high profitable customers

Potential Inputs

Value

- Basket Size
- Visit Frequency

Basket

- Spend by category
- · Type of category
- Brand spend (i.e. private label)

Promotions

- % bought on targeted promotion
- % bought from flyer

Time

- · Time of day
- Day of week

Location

- Store format
- Area population density

Clustering approach

Deal Seeking Mom

Key Differentiators



- Full store shop
- · High avg. basket size / # trips



- · High % purchased on promotion
- · Rewards seeker



- High spend categories
 - Fresh produce
 - Organic food
 - Multipack juice, snack



Clustering: Application 2

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.

Enron email dataset





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 Analysis: Applications

Market-basket analysis

 Rules are used for sales promotion, shelf management, and inventory management

Telecommunication alarm diagnosis

 Rules are used to find combination of alarms that occur together frequently in the same time period

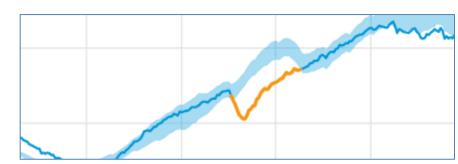
Medical Informatics

 Rules are used to find combination of patient symptoms and test results associated with certain diseases



Deviation/Anomaly/Change Detection

 Detect significant deviations from normal behavior



Applications:

- Credit Card Fraud Detection
- Network IntrusionDetection
- Identify anomalous behavior from sensor networks for monitoring and surveillance.
- Detecting changes in the global forest cover.





Motivating Challenges

Traditional techniques may be unsuitable due to some challenges:

- Scalability
- High Dimensionality
- Heterogeneous and Complex Data
- Data Ownership and Distribution
- Non-traditional Analysis

