Lecture Notes for Chapter 1

Large-scale Data is Everywhere!

- There has been 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.





E-Commerce

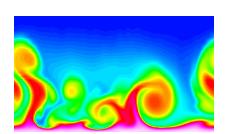


Traffic Patterns

Social Networking: Twitter



Sensor Networks



Computational Simulations

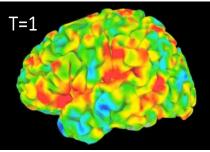
Why Data Mining? Commercial Viewpoint

- Lots of data is being collected and warehoused
 - Web data
 - Google 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

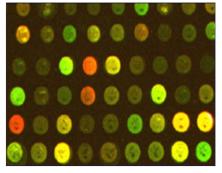
- Data collected and stored at enormous speeds
 - remote sensors on a satellite
 NASA EOSDIS archives over petabytes of earth science data / year
 - telescopes scanning the skies
 Sky survey data
 - High-throughput biological data
 - scientific simulations
 - terabytes of data generated in a few hours
- Data mining helps scientists
 - in automated analysis of massive datasets
 - In hypothesis formation



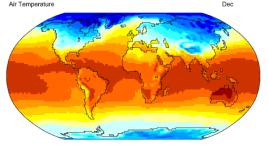
fMRI Data from Brain



Sky Survey Data



Gene Expression Data



Surface Temperature of Earth

Great opportunities to improve productivity in all walks of life

McKinsey Global Institute

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

Big data—a growing torrent

\$600 to buy a disk drive that can store all of the world's music

5 billion mobile phones in use in 2010

30 billion pieces of content shared on Facebook every month

40% projected growth in global data generated per year vs. 5%

15 out of 17 sectors in the United States have more data stored per company

than the US Library of Congress

growth in global IT spending

235 terabytes data collected by the US Library of Congress in April 2011 Big data—capturing its value \$300 billion potential annual value to US health care – more than double the total annual health care spending in Spain €250 billion potential annual value to Europe's public sector administration – more than GDP of Greece \$600 billion potential annual consumer surplus from

using personal location data globally

140,000–190,000 more deep analytical talent positions, and

60% potential increase in retailers' operating margins possible with big data

needed to take full advantage

of big data in the United States

09/09/2020

Great Opportunities to Solve Society's Major Problems



Improving health care and reducing costs



Finding alternative/ green energy sources

CCCms/A2a January to January Mean Temperature (degrees C) 2080s relative to 1961-90



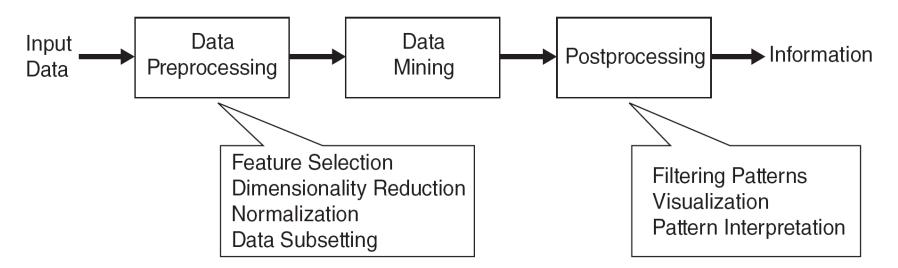
Reducing hunger and poverty by increasing agriculture production

09/09/2020

What is Data Mining?

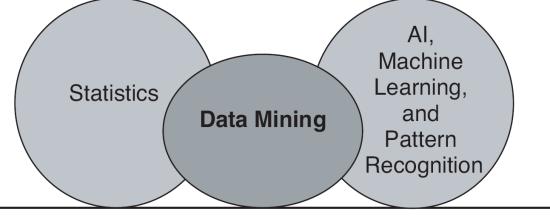
Many Definitions

- 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



Origins of Data Mining

- Draws ideas from machine learning/AI, pattern recognition, statistics, and database systems
- Traditional techniques may be unsuitable due to data that is
 - Large-scale
 - High dimensional
 - Heterogeneous
 - Complex
 - Distributed



Database Technology, Parallel Computing, Distributed Computing

 A key component of the emerging field of data science and datadriven discovery

Data Mining Tasks

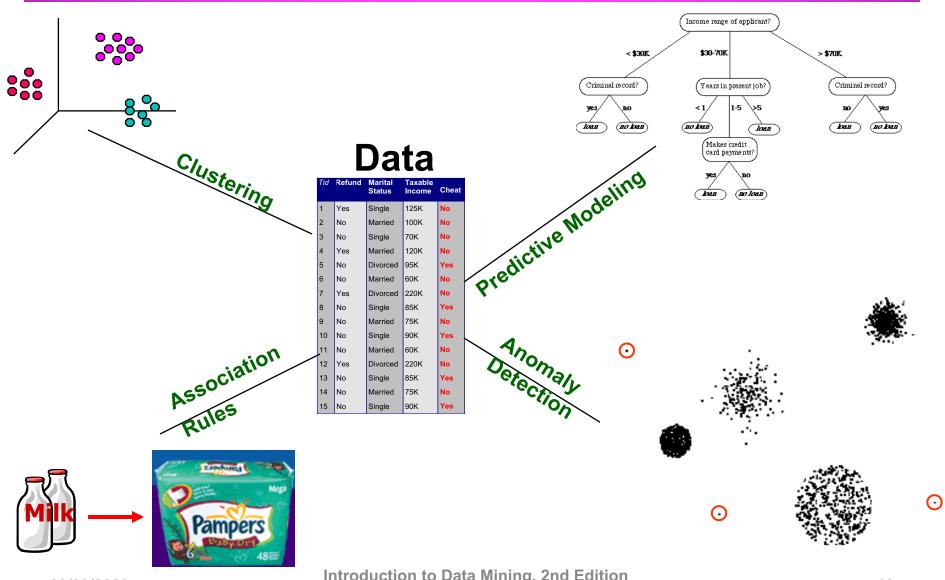
Prediction Methods

 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 ...



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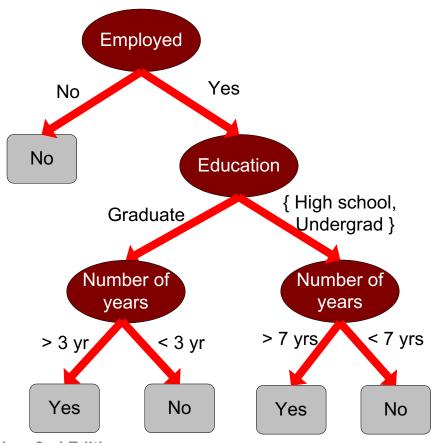
Predictive Modeling: Classification

Class

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

Model for predicting credit worthiness

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



Tan, Steinbach, Karpatne, Kumar

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	?	
Test Set Learn Classifier Model					

Introduction to Data Mining, 2nd Edition Tan, Steinbach, Karpatne, Kumar

Traini

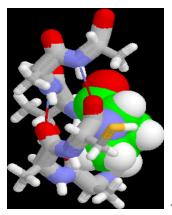
Set

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





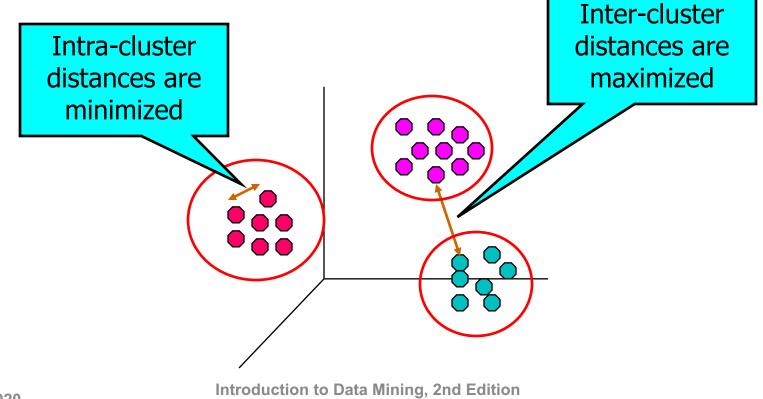


Regression

- Predict a value of a given continuous valued variable based on the values of other variables, assuming a linear or nonlinear model of dependency.
- Extensively 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.

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



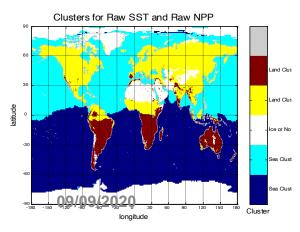
Applications of Cluster Analysis

Understanding

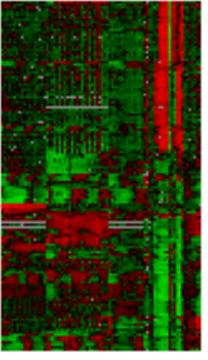
- Custom profiling for targeted marketing
- Group related documents for browsing
- Group genes and proteins that have similar functionality
- Group stocks with similar price fluctuations

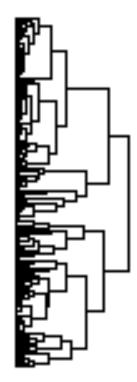
Summarization

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.





Courtesy: Michael Eisen



Clustering: Application 1

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.

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

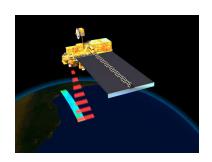
- Rules are used for sales promotion, shelf management, and inventory management
- Telecommunication alarm diagnosis
 - Rules are used to find combination of alarms that occur together frequently in the same time period

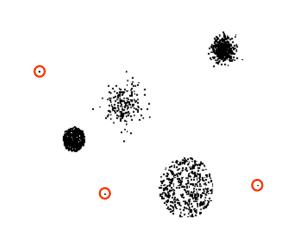
Medical Informatics

 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 Intrusion
 Detection
 - Identify anomalous behavior from sensor networks for monitoring and surveillance.
 - Detecting changes in the global forest cover.







Motivating Challenges

- Scalability
- High Dimensionality
- Heterogeneous and Complex Data
- Data Ownership and Distribution

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    Non-traditional Analysis
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