# Data Mining: Concepts and Techniques

Chapter-1 Introduction

# Course Schedule

- Introduction
- Data preprocessing
- Mining Frequent Patterns
- Mining Association Rules in Large Databases
- Classification
- Prediction
- Clustering analysis

# Where to Find the contents

Text Book

Jiawei Han & Micheline Kamber, Elsevier

Reference Book

Arun Ku. Pujari, "Data Mining Techniques", University Press.

Tutorial sections

http://www.cs.sfu.ca/~han/dmbook

Other conference presentation slides (.ppt):

http://db.cs.sfu.ca/ or http://www.cs.sfu.ca/~han

Research papers, DBMiner system, and other related information:

http://db.cs.sfu.ca/ or http://www.cs.sfu.ca/~han

# Chapter 1. Introduction

- Motivation: Why data mining?
- What is data mining?
- Data Mining: On what kind of data?
- Data mining functionality
- Are all the patterns interesting?
- Classification of data mining systems
- Major issues in data mining

# Motivation: "Necessity is the Mother of Invention"

### Data explosion problem

- Automated data collection tools and mature database technology lead to tremendous amounts of data stored in databases, data warehouses and other information repositories
- We are drowning in data, but starving for knowledge!
- Solution: Data warehousing and data mining
  - Data warehousing and on-line analytical processing
  - Extraction of interesting knowledge (rules, regularities, patterns, constraints) from data in large databases

# **Evolution of Database Technology**

### ▶ 1960s:

Data collection, database creation, IMS and network DBMS

#### ▶ 1970s:

Relational data model, relational DBMS implementation

### ▶ 1980s:

RDBMS, advanced data models (extended-relational, OO, deductive, etc.) and application-oriented DBMS (spatial, scientific, engineering, etc.)

### ▶ 1990s—2000s:

 Data mining and data warehousing, multimedia databases, and Web databases

# What Is Data Mining?



- Data mining (knowledge discovery in databases):
  - Extraction of interesting (non-trivial, implicit, previously unknown and potentially useful) information or patterns from data in <u>large</u> databases
- 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?
  - Query processing.
  - Expert systems or small ML/statistical programs

# Why Data Mining?Potential Applications

### Database analysis and Decision Support

- Market analysis and management
  - target marketing, customer relation management, market basket analysis, cross selling, market segmentation
- Risk analysis and management
  - Forecasting, customer retention, improved underwriting, quality control, competitive analysis
- Fraud Detection and Management

### Other Applications

- Text mining (news group, email, documents) and Web analysis.
- Intelligent query answering

# Market Analysis and Management (1)

- Where are the data sources for analysis?
  - Credit card transactions, loyalty cards, discount coupons, customer complaint calls, plus (public) lifestyle studies
- Target marketing
  - Find clusters of "model" customers who share the same characteristics: interest, income level, spending habits, etc.
- Determine customer purchasing patterns over time
  - Conversion of single to a joint bank account: marriage, etc.
- Cross-market analysis
  - Associations/co-relations between product sales
  - Prediction based on the association information

# Market Analysis and Management (2)

### Customer profiling

- data mining can tell you what types of customers buy what products (clustering or classification)
- Identifying customer requirements
  - identifying the best products for different customers
  - use prediction to find what factors will attract new customers
- Provides summary information
  - various multidimensional summary reports
  - statistical summary information (data central tendency and variation)

# Corporate Analysis and Risk Management

- Finance planning and asset evaluation
  - cash flow analysis and prediction
  - contingent claim analysis to evaluate assets
  - cross-sectional and time series analysis (financial-ratio, trend analysis, etc.)
- Resource planning:
  - summarize and compare the resources and spending
- Competition:
  - monitor competitors and market directions
  - group customers into classes and a class-based pricing procedure
  - set pricing strategy in a highly competitive market

# Fraud Detection and Management (1)

### Applications

 widely used in health care, retail, credit card services, telecommunications (phone card fraud), etc.

### Approach

 use historical data to build models of fraudulent behavior and use data mining to help identify similar instances

### Examples

- <u>auto insurance</u>: detect a group of people who stage accidents to collect on insurance
- money laundering: detect suspicious money transactions (US Treasury's Financial Crimes Enforcement Network)
- medical insurance: detect professional patients and ring of doctors and ring of references

# Fraud Detection and Management (2)

### Detecting inappropriate medical treatment

 Australian Health Insurance Commission identifies that in many cases blanket screening tests were requested (save Australian \$1 m/yr).

### Detecting telephone fraud

- Telephone call model: destination of the call, duration, time of day or week. Analyze patterns that deviate from an expected norm.
- British Telecom identified discrete groups of callers with frequent intra-group calls, especially mobile phones, and broke a multimillion dollar fraud.

### Retail

 Analysts estimate that 38% of retail shrink is due to dishonest employees.

# Other Applications

### Sports

 IBM Advanced Scout analyzed NBA game statistics (shots blocked, assists, and fouls) to gain competitive advantage for New York Knicks and Miami Heat

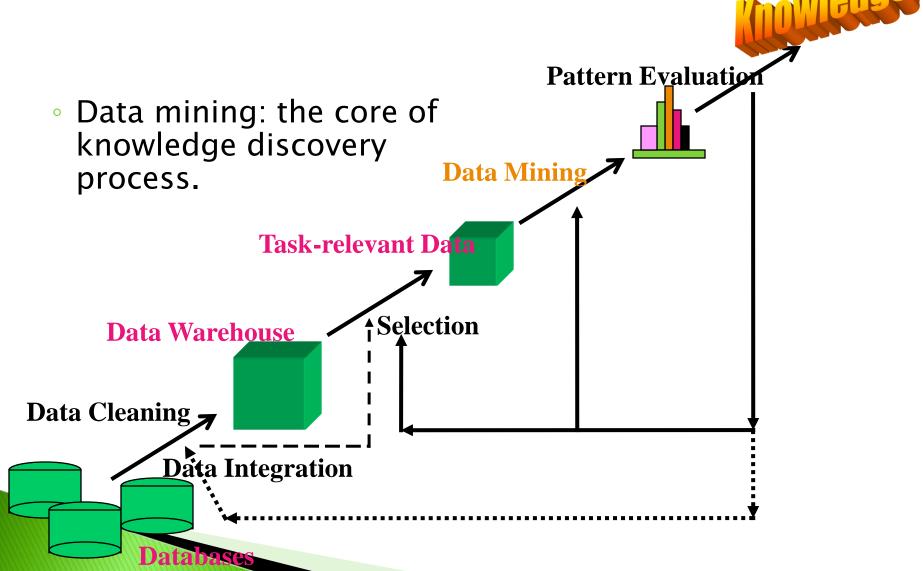
### Astronomy

 JPL and the Palomar Observatory discovered 22 quasars with the help of data mining

#### Internet Web Surf-Aid

 IBM Surf-Aid applies data mining algorithms to Web access logs for market-related pages to discover customer preference and behavior pages, analyzing effectiveness of Web marketing, improving Web site organization, etc.

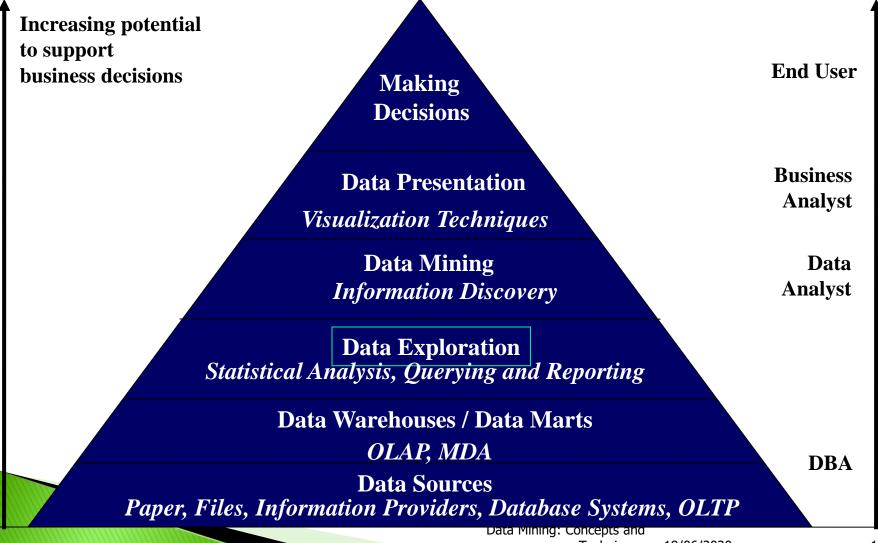
# Data Mining: A KDD Process



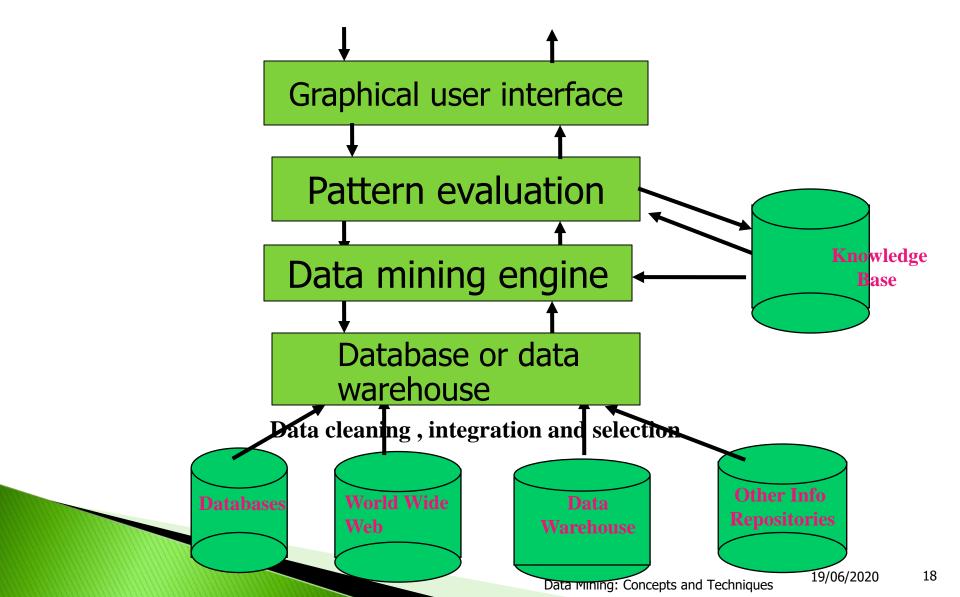
# Steps of a KDD Process

- Learning the application domain:
  - relevant prior knowledge and goals of application
- Creating a target data set: data selection
- Data cleaning and preprocessing: (may take 60% of effort!)
- Data reduction and transformation:
  - Find useful features, dimensionality/variable reduction, invariant representation.
- Choosing functions of data mining
  - summarization, classification, regression, association, clustering.
- Choosing the mining algorithm(s)
- Data mining: search for patterns of interest
- Pattern evaluation and knowledge presentation
  - visualization, transformation, removing redundant patterns, etc.
- Use of discovered knowledge

# Data Mining and Business Intelligence



# Architecture of a Typical Data Mining System



# Data Mining: On What Kind of Data?

- Relational databases
- Data warehouses
- Transactional databases
- Advanced DB and information repositories
  - Object-oriented and object-relational databases
  - Spatial databases
  - Time-series data and temporal data
  - Text databases and multimedia databases
  - Heterogeneous and legacy databases
  - <u>WWW</u>

# Data Mining Functionalities (1)

- Concept description: Characterization and discrimination
  - Generalize, summarize, and contrast data characteristics, e.g., dry vs. wet regions
- Association (correlation)
  - Multi-dimensional vs. single-dimensional association
  - age(X, "20..29") ^ income(X, "20..29K") → buys(X, "PC") [support = 2%, confidence = 60%]
  - contains(T, "computer") → contains(x, "software")
    [1%, 75%]

# Data Mining Functionalities (2)

### Classification and Prediction

- Finding models (functions) that describe and distinguish classes or concepts for future prediction
- E.g., classify countries based on climate, or classify cars based on gas mileage
- Presentation: decision-tree, classification rule, neural network
- Prediction: Predict some unknown or missing numerical values

### Cluster analysis

- Class label is unknown: Group data to form new classes, e.g., cluster houses to find distribution patterns
- Clustering based on the principle: maximizing the intraclass similarity and minimizing the interclass similarity

# Data Mining Functionalities (3)

### Outlier analysis

- Outlier: a data object that does not comply with the general behavior of the data
- It can be considered as noise or exception but is quite useful in fraud detection, rare events analysis

### Trend and evolution analysis

- Trend and deviation: regression analysis
- Sequential pattern mining, periodicity analysis
- Similarity-based analysis
- Other pattern-directed or statistical analyses

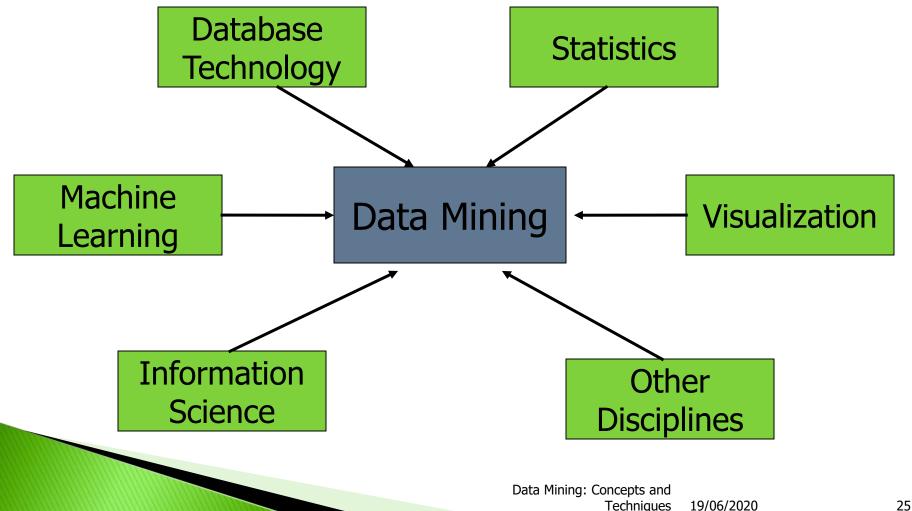
### Are All the "Discovered" Patterns Interesting?

- A data mining system/query may generate thousands of patterns, not all of them are interesting.
- Interestingness measures: A pattern is interesting if it is easily understood by humans, valid on new or test data with some degree of certainty, potentially useful, novel, or validates some hypothesis that a user seeks to confirm
- Objective vs. subjective interestingness measures:
  - Objective: based on statistics and structures of patterns, e.g., support, confidence, etc.
  - <u>Subjective:</u> based on user's belief in the data, e.g., unexpectedness, novelty, action ability, etc.

# Can We Find All and Only Interesting Patterns?

- Find all the interesting patterns: Completeness
  - Can a data mining system find <u>all</u> the interesting patterns?
- Search for only interesting patterns: Optimization
  - Can a data mining system find <u>only</u> the interesting patterns?
  - Approaches
    - First general all the patterns and then filter out the uninteresting ones.
    - Generate only the interesting patterns—mining query optimization

### Data Mining: Confluence of Multiple Disciplines



# Data Mining: Classification Schemes

- General functionality
  - Descriptive data mining
  - Predictive data mining
- Different views, different classifications
  - Kinds of databases to be mined
  - Kinds of knowledge to be discovered
  - Kinds of techniques utilized
  - Kinds of applications adapted

# A Multi-Dimensional View of Data Mining Classification

### Databases to be mined

 Relational, transactional, object-oriented, objectrelational, active, spatial, time-series, text, multi-media, heterogeneous, legacy, WWW, etc.

### Knowledge to be mined

- Characterization, discrimination, association, classification, clustering, trend, deviation and outlier analysis, etc.
- Multiple/integrated functions and mining at multiple levels

### Techniques utilized

 Database-oriented, data warehouse (OLAP), machine learning, statistics, visualization, neural network, etc.

### Applications adapted

 Retail, telecommunication, banking, fraud analysis, DNA mining, stock market analysis, Web mining, Weblog analysis, etc.

# Major Issues in Data Mining (1)

### Mining methodology and user interaction

- Mining different kinds of knowledge in databases
- Interactive mining of knowledge at multiple levels of abstraction
- Incorporation of background knowledge
- Data mining query languages and ad-hoc data mining
- Expression and visualization of data mining results
- Handling noise and incomplete data
- Pattern evaluation: the interestingness problem
- Performance and scalability
  - Efficiency and scalability of data mining algorithms
  - Parallel, distributed and incremental mining methods

# Major Issues in Data Mining (2)

- Issues relating to the diversity of data types
  - Handling relational and complex types of data
  - Mining information from heterogeneous databases and global information systems (WWW)
- Issues related to applications and social impacts
  - Application of discovered knowledge
    - Domain-specific data mining tools
    - Intelligent query answering
    - Process control and decision making
  - Integration of the discovered knowledge with existing knowledge: A knowledge fusion problem
  - Protection of data security, integrity, and privacy

# Summary

- Data mining: discovering interesting patterns from large amounts of data
- A natural evolution of database technology, in great demand, with wide applications
- A KDD process includes data cleaning, data integration, data selection, transformation, data mining, pattern evaluation, and knowledge presentation
- Mining can be performed in a variety of information repositories
- Data mining functionalities: characterization, discrimination, association, classification, clustering, outlier and trend analysis, etc.
- Classification of data mining systems
  - Major issues in data mining

# A Brief History of Data Mining Society

- 1989 IJCAI Workshop on Knowledge Discovery in Databases (Piatetsky-Shapiro)
  - Knowledge Discovery in Databases (G. Piatetsky-Shapiro and W. Frawley, 1991)
- ▶ 1991–1994 Workshops on Knowledge Discovery in Databases
  - Advances in Knowledge Discovery and Data Mining (U. Fayyad, G. Piatetsky-Shapiro, P. Smyth, and R. Uthurusamy, 1996)
- 1995-1998 International Conferences on Knowledge Discovery in Databases and Data Mining (KDD'95-98)
  - Journal of Data Mining and Knowledge Discovery (1997)
- 1998 ACM SIGKDD, SIGKDD'1999-2001 conferences, and SIGKDD Explorations
- More conferences on data mining
  - PAKDD, PKDD, SIAM-Data Mining, (IEEE) ICDM, etc.

### Where to Find References?

- Data mining and KDD (SIGKDD member CDROM):
  - Conference proceedings: KDD, and others, such as PKDD, PAKDD, etc.
  - Journal: Data Mining and Knowledge Discovery
- Database field (SIGMOD member CD ROM):
  - Conference proceedings: ACM-SIGMOD, ACM-PODS, VLDB, ICDE, EDBT, DASFAA
  - Journals: ACM-TODS, J. ACM, IEEE-TKDE, JIIS, etc.
- Al and Machine Learning:
  - Conference proceedings: Machine learning, AAAI, IJCAI, etc.
  - Journals: Machine Learning, Artificial Intelligence, etc.
- Statistics:
  - Conference proceedings: Joint Stat. Meeting, etc.
  - Journals: Annals of statistics, etc.
- Visualization:
  - Conference proceedings: CHI, etc.
    - Journals JEEE Trans. visualization and computer graphics, etc.

### References

- U. M. Fayyad, G. Piatetsky-Shapiro, P. Smyth, and R. Uthurusamy. Advances in Knowledge Discovery and Data Mining. AAAI/MIT Press, 1996.
- J. Han and M. Kamber. Data Mining: Concepts and Techniques. Morgan Kaufmann, 2000.
- T. Imielinski and H. Mannila. A database perspective on knowledge discovery. Communications of ACM, 39:58-64, 1996.
- ▶ G. Piatetsky-Shapiro, U. Fayyad, and P. Smith. From data mining to knowledge discovery: An overview. In U.M. Fayyad, et al. (eds.), Advances in Knowledge Discovery and Data Mining, 1-35. AAAI/MIT Press, 1996.
- G. Piatetsky-Shapiro and W. J. Frawley. Knowledge Discovery in Databases. AAAI/MIT Press, 1991.



# Thank you !!!

# Difference between data mining and Data warehousing?

- Data mining is intended for users who are statistically inclined. These analysts look for patterns hidden in data, which they are able to extract using statistical models. Data miners engage in question formulation based primarily on the "law of large numbers" to identify potentially useful relationships between data elements, which can be profitable to companies.
- Data warehouse users, tend to be data experts who analyze by business dimensions directly. Data warehousing analysts are concerned with what kinds of purchases their customers make, and if the analyst can help the customer by improving the customer experience.