



# DATA MINING

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## Lecture 1: Introduction

Dr. Doaa Elzanfaly

# Course Info.

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- **Lectures:** Monday – 10:00-12:00 / 12:00-2:00
- **Instructor:** Dr. Doaa Elzanfaly
  - email: [doaa.saad@fci.helwan.edu.eg](mailto:doaa.saad@fci.helwan.edu.eg)
  - Contact: Teams
- **Textbook:** Jiawei Han, Micheline Kamber, and Jian Pei. 2011. Data Mining: Concepts and Techniques (3rd. ed.). Morgan Kaufmann Publishers Inc., San Francisco, CA, USA.
- **Reference book:** Charu C. Aggarwal. 2015. Data Mining: The Textbook. Springer Publishing Company, Incorporated.

# Course Objectives

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- To introduce students to various data mining concepts and technologies.
  - Understanding the foundation concepts of data mining.
  - Exploring algorithms commonly used in data mining tools.
  - Ability to apply data mining tools to real-world problems.

# Tentative Syllabus

| Week | Lecture                   | Lab  |
|------|---------------------------|--|
| 1    | Introduction              | Environment Installation<br>Data Mining Introduction |
| 2    | Getting to Know Your Data | Data Preprocessing Techniques                        |
| 3    | Data Preprocessing I      | Data Preprocessing Techniques Continued.             |
| 4    | Data Preprocessing II     | Data Visualization                                   |
| 5    | Association Analysis I    | Apriori Algorithm Implementation                     |
| 6    | Association Analysis II   | A Frequent Pattern Growth Approach                   |
| 7    | Midterm Exam              |  |
| 8    | Classification            | Classification Algorithm Implementation              |
| 9    | Clustering                | Regression Algorithm                                 |
| 10   | Outlier Detection         | Outlier Detection                                    |

# Assessment Scheme

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- **Midterm:** 30 marks
  - **Lab Assignments:** 10 marks
  - **Practical Exam:** 10 marks
  - **Final Exam:** 50 marks
- Bonus points:** 5 marks - Based on participation.

# What I expect from you ...

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- Attend the lectures and lab regularly. (70% to pass)
- Study and learn the material presented in the teaching sessions *and in the textbook.*
- Perform well in the exams.
- Don't cheat.

# Lecture Outline

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- Why Data Mining?
- What Is Data Mining?
- What Kind of Data Can Be Mined?
- Data Mining Tasks
- What Kind of Applications Are Targeted?
- Major Issues in Data Mining

# 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.



*Cyber Security*



*E-Commerce*



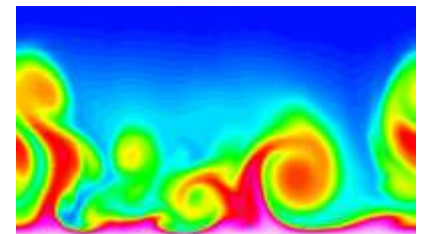
*Traffic Patterns*



*Social Networking: Twitter*



*Sensor Networks*



*Computational Simulations*



# Why Data Mining?



- The Explosive Growth of Data: from terabytes to petabytes
  - Data collection and data availability
    - Automated data collection tools, database systems, Web, computerized society
  - Major sources of abundant data
    - Business: Web, e-commerce, transactions, stocks, ...
    - Science: Remote sensing, bioinformatics, scientific simulation, ...
    - Society and everyone: news, digital cameras, YouTube
- We are drowning in data but starving for knowledge!
- “Necessity is the mother of invention”—Data mining—Automated analysis of massive data sets

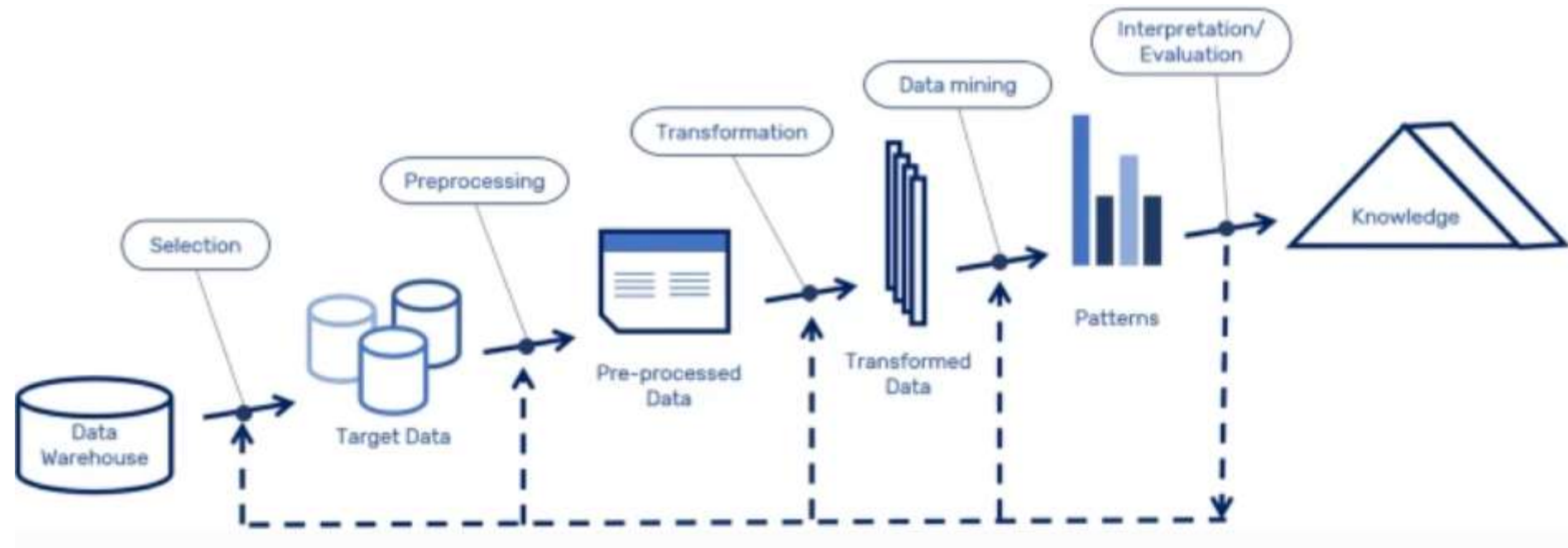
# What Is Data Mining?



- Data mining (knowledge discovery from data)
  - Extraction of interesting (non-trivial, implicit, previously unknown and potentially useful) patterns or knowledge from huge amount of data to predict future trends.
- Alternative names
  - Knowledge Discovery (mining) in Databases (KDD), knowledge extraction, data/pattern analysis, data archeology, data dredging, information harvesting, business intelligence, etc.
- Watch out: Is everything “data mining”?
  - Simple search and query processing
  - (Deductive) expert systems

What is the difference between mining and Querying??

# Knowledge Discovery (KDD) Process



<https://link.springer.com/article/10.1007/s42979-020-0117-6>

# What Kind of Data Can Be Mined?

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- Database-oriented data sets and applications
  - Relational database, data warehouse, transactional database
- Advanced data sets and advanced applications
  - Data streams and sensor data
  - Time-series data, temporal data, sequence data (incl. bio-sequences)
  - Structure data, graphs, social networks and multi-linked data
  - Object-relational databases
  - Heterogeneous databases and legacy databases
  - Spatial data and spatiotemporal data
  - Multimedia database
  - Text databases
  - The World-Wide Web

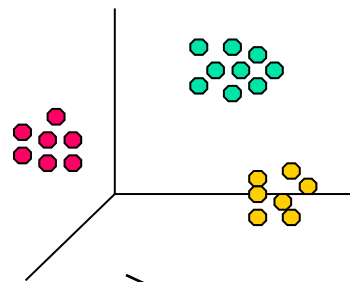
# Data Mining Tasks



- 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



Clustering

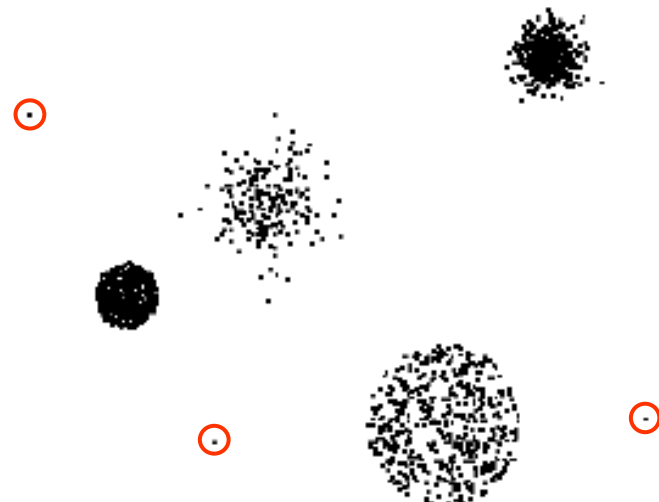
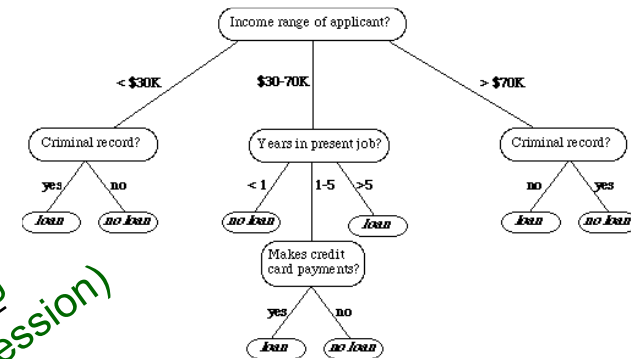
## Data

| 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   |
| 11  | No     | Married        | 60K            | No    |
| 12  | Yes    | Divorced       | 220K           | No    |
| 13  | No     | Single         | 85K            | Yes   |
| 14  | No     | Married        | 75K            | No    |
| 15  | No     | Single         | 90K            | Yes   |

Association Rules

Predictive Modeling  
(Classification & Regression)

Anomaly Detection



# Predictive Modeling: Classification

| <i>Tid</i> | 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           |
| ...        | ...      | ...                | ...                        | ...           |

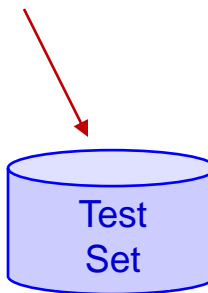
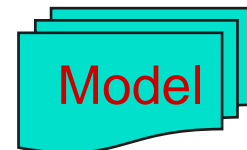
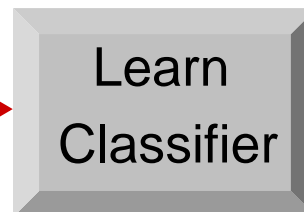
categorical

categorical

quantitative

class

| <i>Tid</i> | 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

Model

# Classification

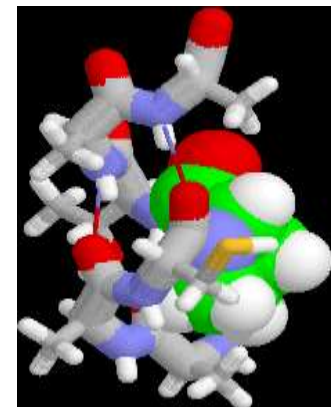


- Classification and label prediction – Supervised Learning
  - Construct models (functions) based on some training examples
  - Describe and distinguish classes or concepts for future prediction
    - E.g., classify countries based on (climate), or classify cars based on (gas mileage)
  - Predict some unknown class labels
- Typical methods
  - Decision trees, naïve Bayesian classification, support vector machines, neural networks, rule-based classification, pattern-based classification, logistic regression, ...
- Typical applications:
  - Credit card fraud detection, direct marketing, classifying stars, diseases, web-pages, ...



# 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



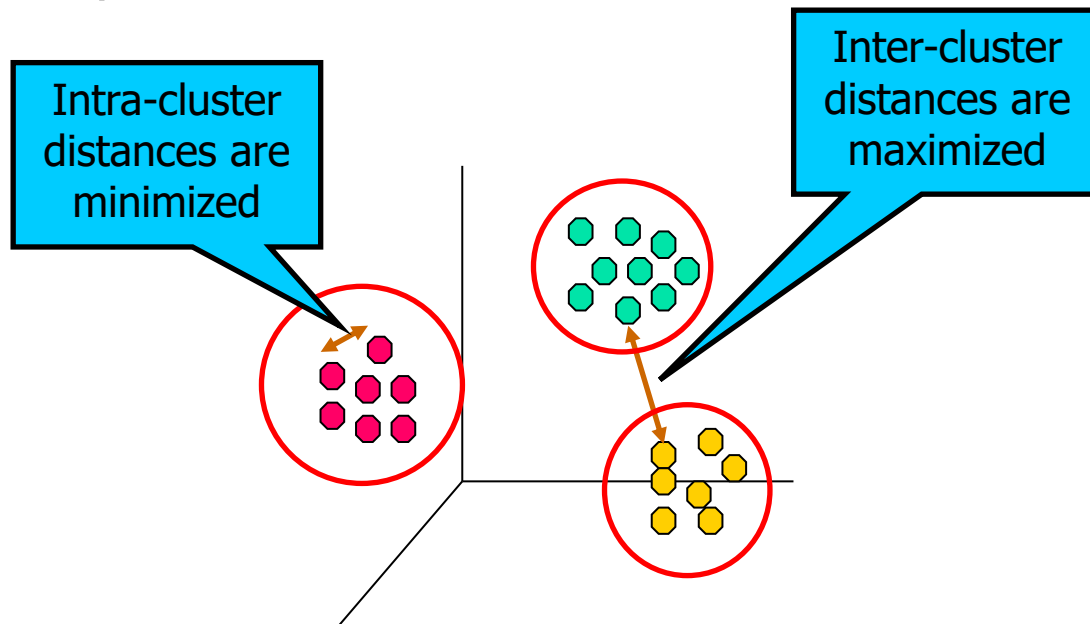
# Regression



- 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 advertising 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
- Unsupervised learning (i.e., Class label is unknown)
- Principle: Maximizing intra-class similarity & minimizing interclass similarity



# Applications of Cluster Analysis

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## ■ **Market Segmentation:**

- **Goal:** subdivide a market into distinct subsets of customers where any subset may possibly 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.

# Applications of Cluster Analysis

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- **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 and use it to cluster.

# Association and Correlation Analysis

- 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          | Tea, Bread              |
| 3          | Tea, Coke, Sugar, Milk  |
| 4          | Tea, Bread, Sugar, Milk |
| 5          | Coke, Sugar, Milk       |

Rules Discovered:

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

$\{\text{Sugar, Milk}\} \rightarrow \{\text{Tea}\}$

# Association and Correlation Analysis

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- Frequent patterns (or frequent itemsets)
  - What items are frequently purchased together?
- Association, correlation vs. causality
  - A typical association rule
    - Tea  $\rightarrow$  Sugar [0.5%, 75%] (support, confidence)
  - Are strongly associated items also strongly correlated?
- How to mine such patterns and rules efficiently in large datasets?
- How to use such patterns for classification, clustering, and other applications?

# Applications of Association Analysis

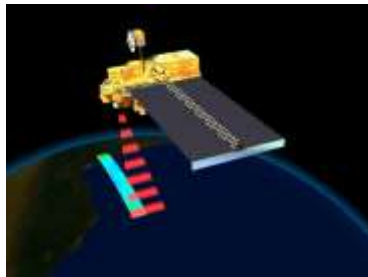
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- Market-basket analysis
  - Rules are used for sales promotion, shelf management, and inventory management
- Telecommunication alarm diagnosis
  - Rules are used to find combination of alarms that occur together frequently in the same time period
- Medical Informatics
  - Rules are used to find combination of patient symptoms and test results associated with certain diseases



# Deviation/Anomaly/Change Detection

- Also known as outlier analysis
- 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.

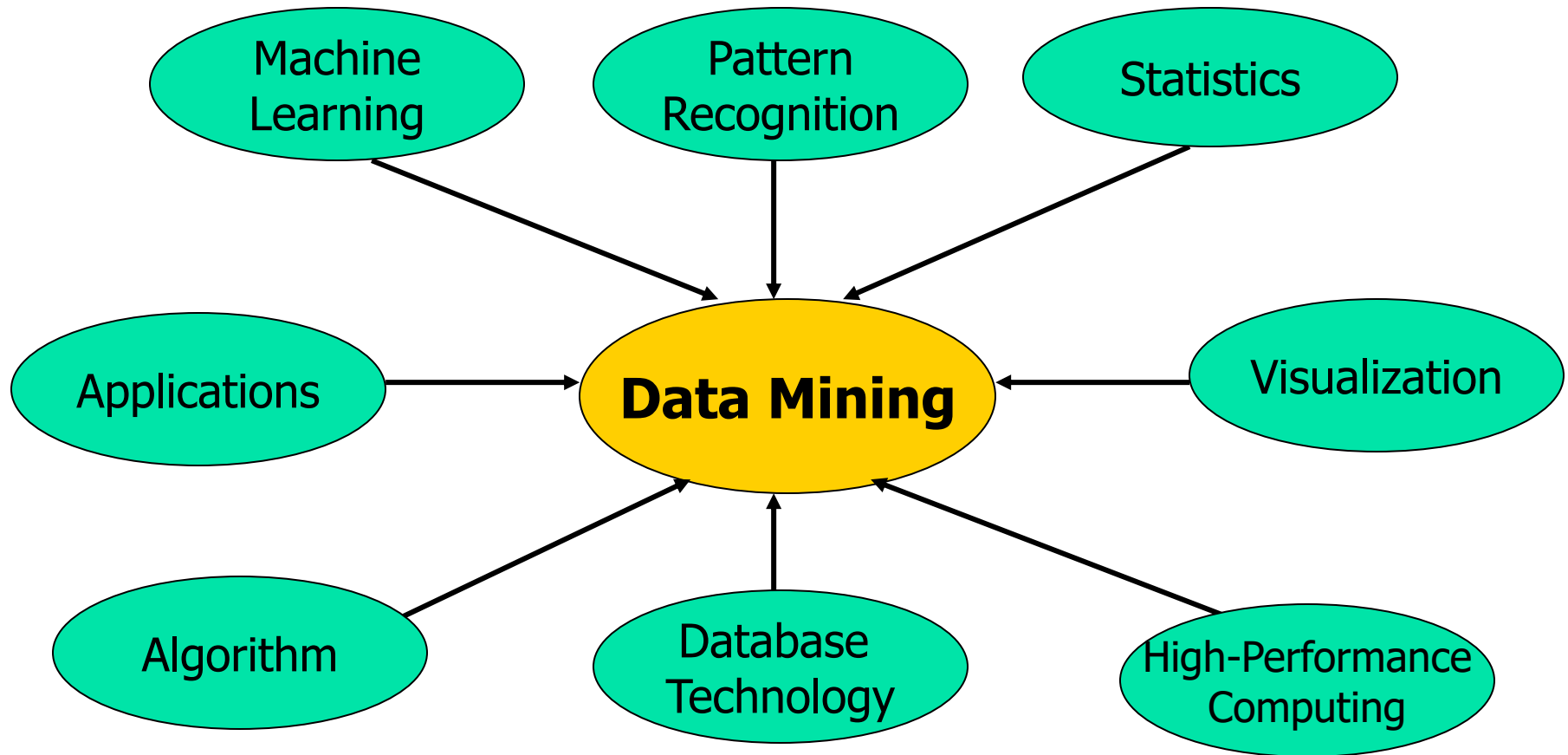


# Outlier Analysis



- Outlier analysis
  - Outlier: A data object that does not comply with the general behavior of the data.
  - Outlier analysis is to detect significant deviations from normal behavior
    - Noise or exception? — One person's garbage could be another person's treasure
    - Methods: by product of clustering or regression analysis, ...
    - Useful in fraud detection, rare events analysis, Network Intrusion Detection, Identify anomalous behavior from sensor networks for monitoring and surveillance

# Data Mining: Confluence of Multiple Disciplines



# Major Issues in Data Mining



- Efficiency and Scalability
  - Efficiency and scalability of data mining algorithms
  - Parallel, distributed, stream, and incremental mining methods
- Diversity of data types
  - Handling complex types of data
  - Mining dynamic, networked, and global data repositories
- Data mining and society
  - Social impacts of data mining
  - Privacy-preserving data mining
  - Invisible data mining

# Where to Find References? DBLP, CiteSeer, Google

- Data mining and KDD (SIGKDD: CDROM)
  - Conferences: ACM-SIGKDD, IEEE-ICDM, SIAM-DM, PKDD, PAKDD, etc.
  - Journal: Data Mining and Knowledge Discovery, KDD Explorations, ACM TKDD
- Database systems (SIGMOD: ACM SIGMOD Anthology—CD ROM)
  - Conferences: ACM-SIGMOD, ACM-PODS, VLDB, IEEE-ICDE, EDBT, ICDT, DASFAA
  - Journals: IEEE-TKDE, ACM-TODS/TOIS, JIIS, J. ACM, VLDB J., Info. Sys., etc.
- AI & Machine Learning
  - Conferences: Machine learning (ML), AAAI, IJCAI, COLT (Learning Theory), CVPR, NIPS, etc.
  - Journals: Machine Learning, Artificial Intelligence, Knowledge and Information Systems, IEEE-PAMI, etc.
- Web and IR
  - Conferences: SIGIR, WWW, CIKM, etc.
  - Journals: WWW: Internet and Web Information Systems,
- Statistics
  - Conferences: Joint Stat. Meeting, etc.
  - Journals: Annals of statistics, etc.
- Visualization
  - Conference proceedings: CHI, ACM-SIGGraph, etc.
  - Journals: IEEE Trans. visualization and computer graphics, etc.

# Recommended Reference Books

- S. Chakrabarti. Mining the Web: Statistical Analysis of Hypertext and Semi-Structured Data. Morgan Kaufmann, 2002
- R. O. Duda, P. E. Hart, and D. G. Stork, Pattern Classification, 2ed., Wiley-Interscience, 2000
- T. Dasu and T. Johnson. Exploratory Data Mining and Data Cleaning. John Wiley & Sons, 2003
- U. M. Fayyad, G. Piatetsky-Shapiro, P. Smyth, and R. Uthurusamy. Advances in Knowledge Discovery and Data Mining. AAAI/MIT Press, 1996
- U. Fayyad, G. Grinstein, and A. Wierse, Information Visualization in Data Mining and Knowledge Discovery, Morgan Kaufmann, 2001
- J. Han and M. Kamber. Data Mining: Concepts and Techniques. Morgan Kaufmann, 3<sup>rd</sup> ed., 2011
- D. J. Hand, H. Mannila, and P. Smyth, Principles of Data Mining, MIT Press, 2001
- T. Hastie, R. Tibshirani, and J. Friedman, The Elements of Statistical Learning: Data Mining, Inference, and Prediction, 2<sup>nd</sup> ed., Springer-Verlag, 2009
- B. Liu, Web Data Mining, Springer 2006.
- T. M. Mitchell, Machine Learning, McGraw Hill, 1997
- G. Piatetsky-Shapiro and W. J. Frawley. Knowledge Discovery in Databases. AAAI/MIT Press, 1991
- P.-N. Tan, M. Steinbach and V. Kumar, Introduction to Data Mining, Wiley, 2005
- S. M. Weiss and N. Indurkha, Predictive Data Mining, Morgan Kaufmann, 1998
- I. H. Witten and E. Frank, Data Mining: Practical Machine Learning Tools and Techniques with Java Implementations, Morgan Kaufmann, 2<sup>nd</sup> ed. 2005