



Business Intelligence

© Han, J., Kamber, M. and Pei, J., 2011. Data mining concepts and techniques 3rd edition. *The Morgan Kaufmann Series in Data Management Systems*.

Course Outline

- Data Mining Concepts
- An Overview of Business Intelligence Technology
- Getting to Know Your Data
- R Programming Basics
- Data Preprocessing
- Data Warehousing and Online Analytical Processing
- Mining Frequent Patterns, Associations, and Correlations
- Classification: Basic Concepts
- Cluster Analysis: Basic Concepts and Methods
- Outlier Detection

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

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.)
 - Application-oriented DBMS (spatial, scientific, engineering, etc.)
- 1990s:
 - Data mining, data warehousing, multimedia databases, and Web databases
- 2000s
 - Stream data management and mining
 - Data mining and its applications
 - Web technology (XML, data integration) and global information systems

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
 - Data mining: **a misnomer?**
- 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 (not) Data Mining?

■ What is not Data Mining?

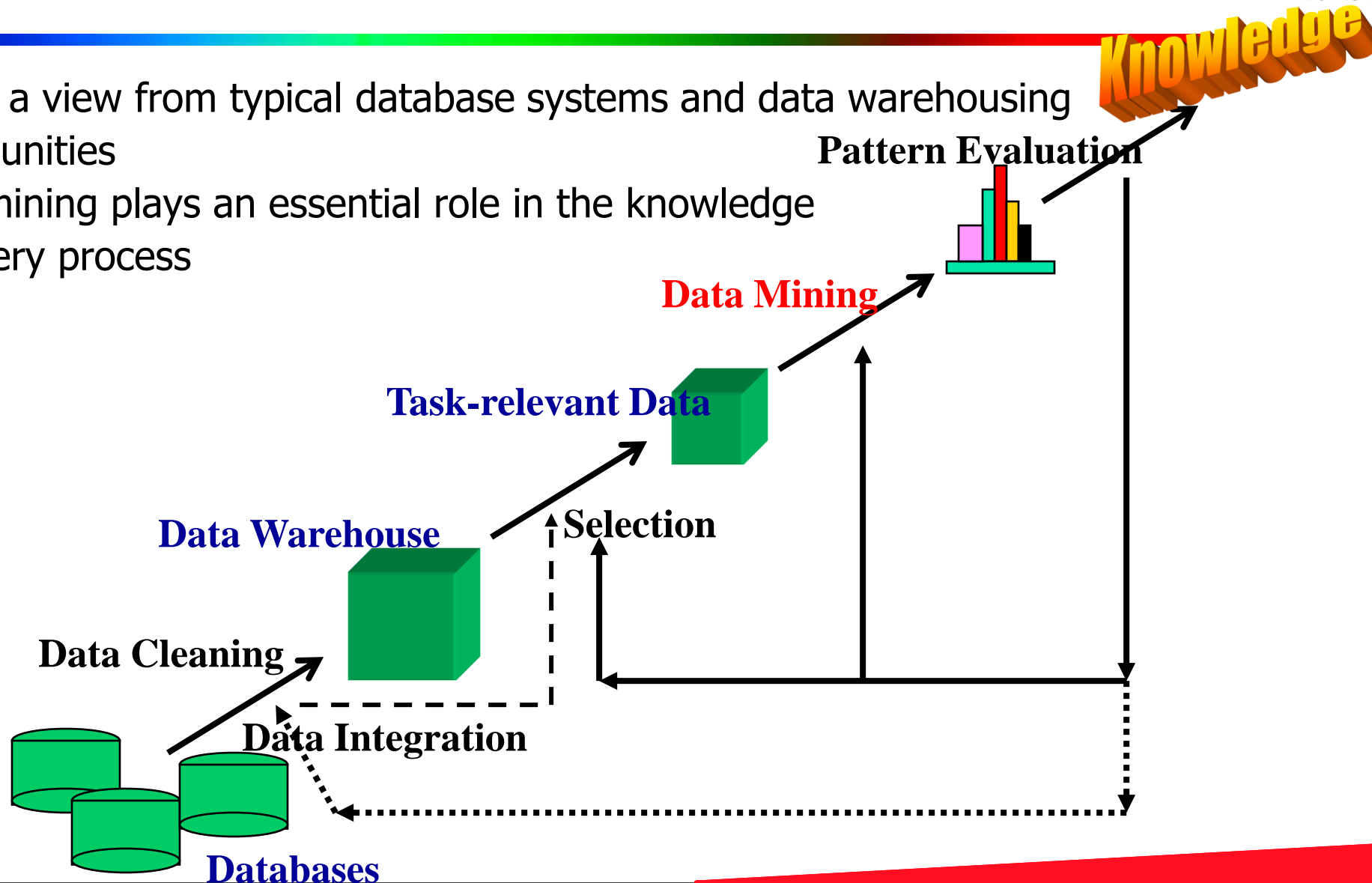
- Simple Querying
- Done with a predetermined hypothesis.
- Examples
 - Look up phone number in phone directory
 - Query a Web search engine for information about "Amazon"

■ What is Data Mining?

- Extraction of patterns and Knowledge
- Examples:
 - Certain Cars are more prevalent in Colorado Springs (Tesla, Chevy)
 - Group together similar documents returned by search engine according to their context (e.g. Amazon rainforest, Amazon.com,)

Knowledge Discovery (KDD) Process

- This is a view from typical database systems and data warehousing communities
- Data mining plays an essential role in the knowledge discovery process



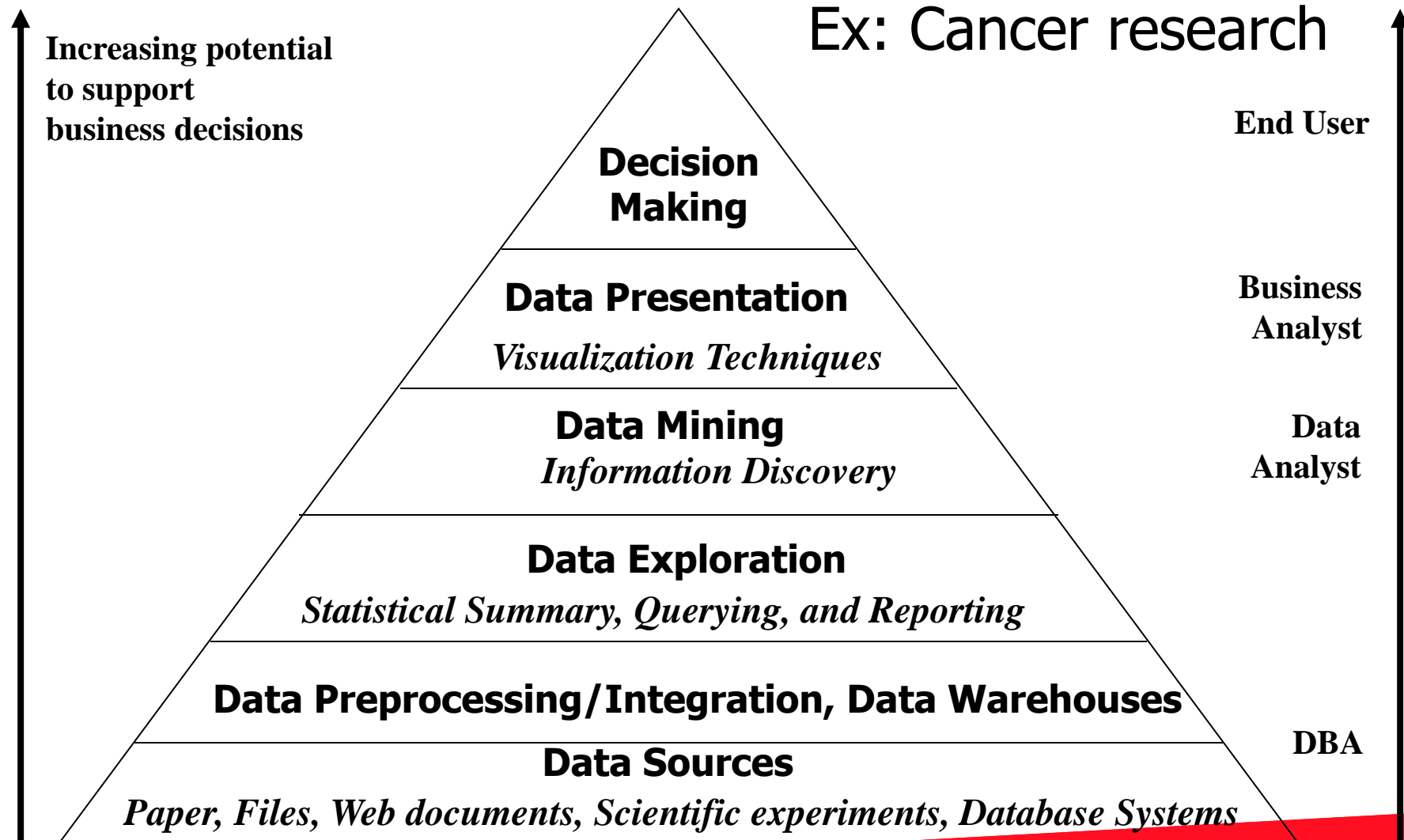
Example: A Web Mining Framework

- Web mining usually involves
 - Data cleaning
 - Data integration from multiple sources
 - Warehousing the data
 - Data cube construction
 - Data selection for data mining
 - Data mining
 - Presentation of the mining results
 - Patterns and knowledge to be used or stored into knowledge-base

Data Mining Applications

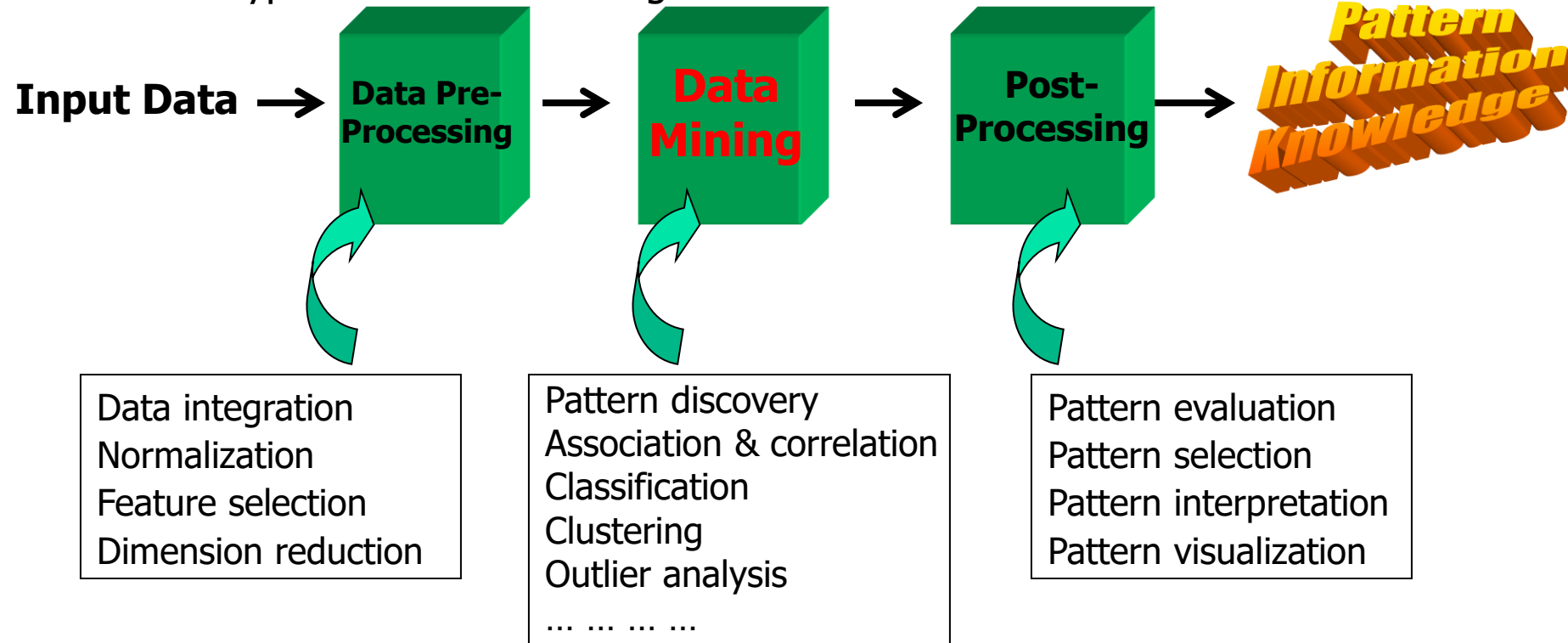
- Two highly successful and popular applications of data mining: *business intelligence* and *search engines*.
- **Business Intelligence (BI)**
 - critical for businesses to acquire a better understanding of the commercial context of their organization, such as their customers, the market, supply and resources, and competitors.
 - BI technologies provide historical, current, and predictive views of business operations.
 - data mining at the core of business intelligence
- **Web Search Engines**
 - specialized computer server that searches for information on the Web.
 - Various data mining techniques are used in all aspects of search engines, ranging from *crawling*, indexing and searching (e.g., ranking , advertising and personalizing results).

Data Mining in Business Intelligence



KDD Process: A Typical View from ML and Statistics

- This is a view from typical machine learning and statistics communities



- **Data to be mined**

- Database data (extended-relational, object-oriented, heterogeneous, legacy), data warehouse, transactional data, stream, spatiotemporal, time-series, sequence, text and web, multi-media, graphs & social and information networks

- **Knowledge to be mined (or: Data mining functions)**

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

- **Techniques utilized**

- Data-intensive, data warehouse (OLAP), machine learning, statistics, pattern recognition, visualization, high-performance, etc.

- **Applications adapted**

- Retail, telecommunication, banking, fraud analysis, bio-data mining, stock market analysis, text mining, Web mining, etc.

Big Data Applications



Smarter Healthcare



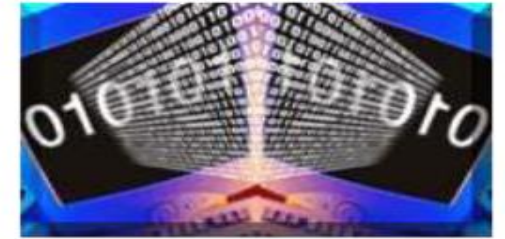
Multi-channel



Finance



Log Analysis



Homeland Security



Traffic Control



Telecom



Search Quality



Manufacturing



Trading Analytics



Fraud and Risk



Retail: Churn, NBO



Data Mining: On What Kinds of Data?

- 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(Web Mining)**

Multi-Dimensional View of Data Mining

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Data Mining Function: (1) Generalization

- Information integration and data warehouse construction
 - Data cleaning, transformation, integration, and multidimensional data model
- Data cube technology
 - Scalable methods for computing (i.e., materializing) multidimensional aggregates
 - OLAP (online analytical processing)
- Multidimensional concept description: Characterization and discrimination
 - Generalize, **summarize**, and **contrast data** characteristics, e.g., dry vs. wet region

Multi-Dimensional View of Data Mining

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Data Mining Function: (2) Association and Correlation Analysis

- Frequent patterns (or frequent itemsets)
 - *What items are frequently purchased together in your Walmart?*
- Association, correlation vs. causality
 - A typical association rule
 - *Diaper* \rightarrow *Beer* [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

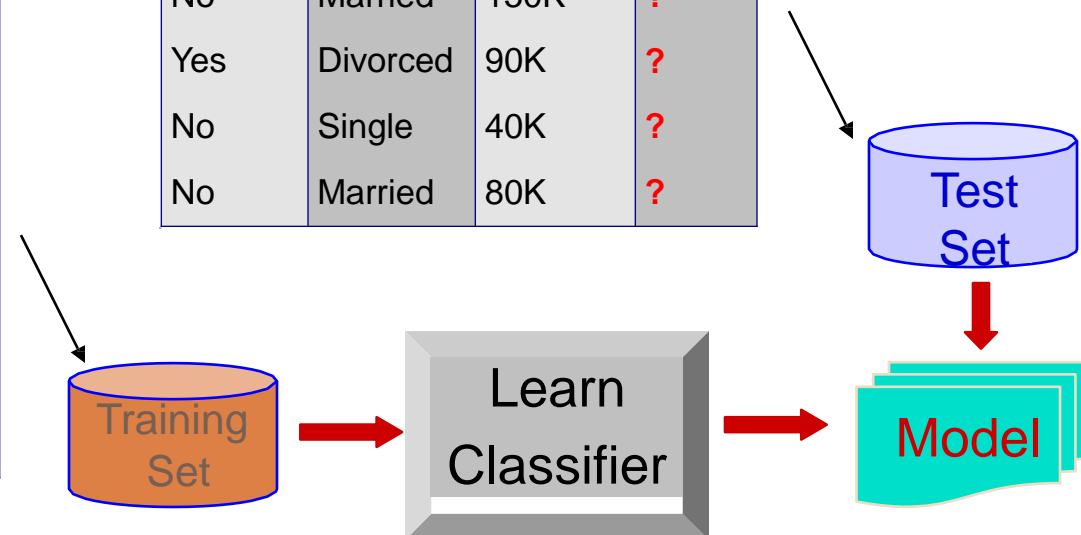
Data Mining Function: (3) Classification

- Classification and label prediction
 - 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, ...

Classification Example

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

Refund	Marital Status	Taxable Income	Cheat
No	Single	75K	?
Yes	Married	50K	?
No	Married	150K	?
Yes	Divorced	90K	?
No	Single	40K	?
No	Married	80K	?



Data Mining Function: (4) Cluster Analysis

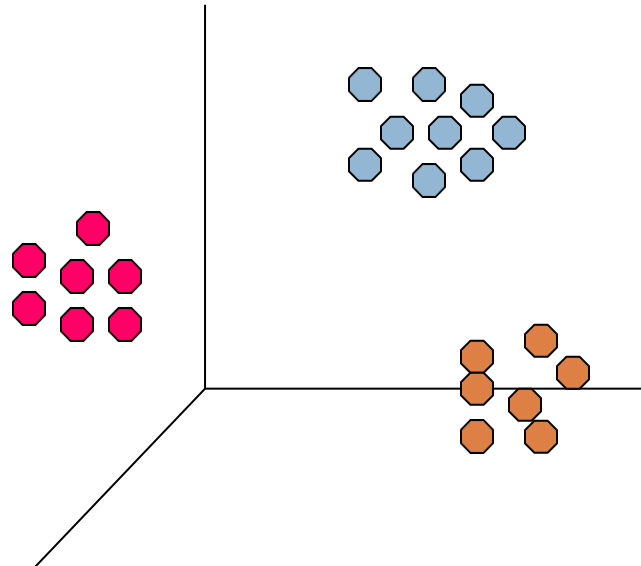
- Unsupervised learning (i.e., Class label is unknown)
- Group data to form new categories (i.e., clusters), e.g., cluster houses to find distribution patterns
- Principle: Maximizing intra-class similarity & minimizing interclass similarity
- Many methods and applications

Illustrating Clustering

- Euclidean Distance Based Clustering in 3-D space.

Intracuster distances
are minimized

Intercluster distances
are maximized



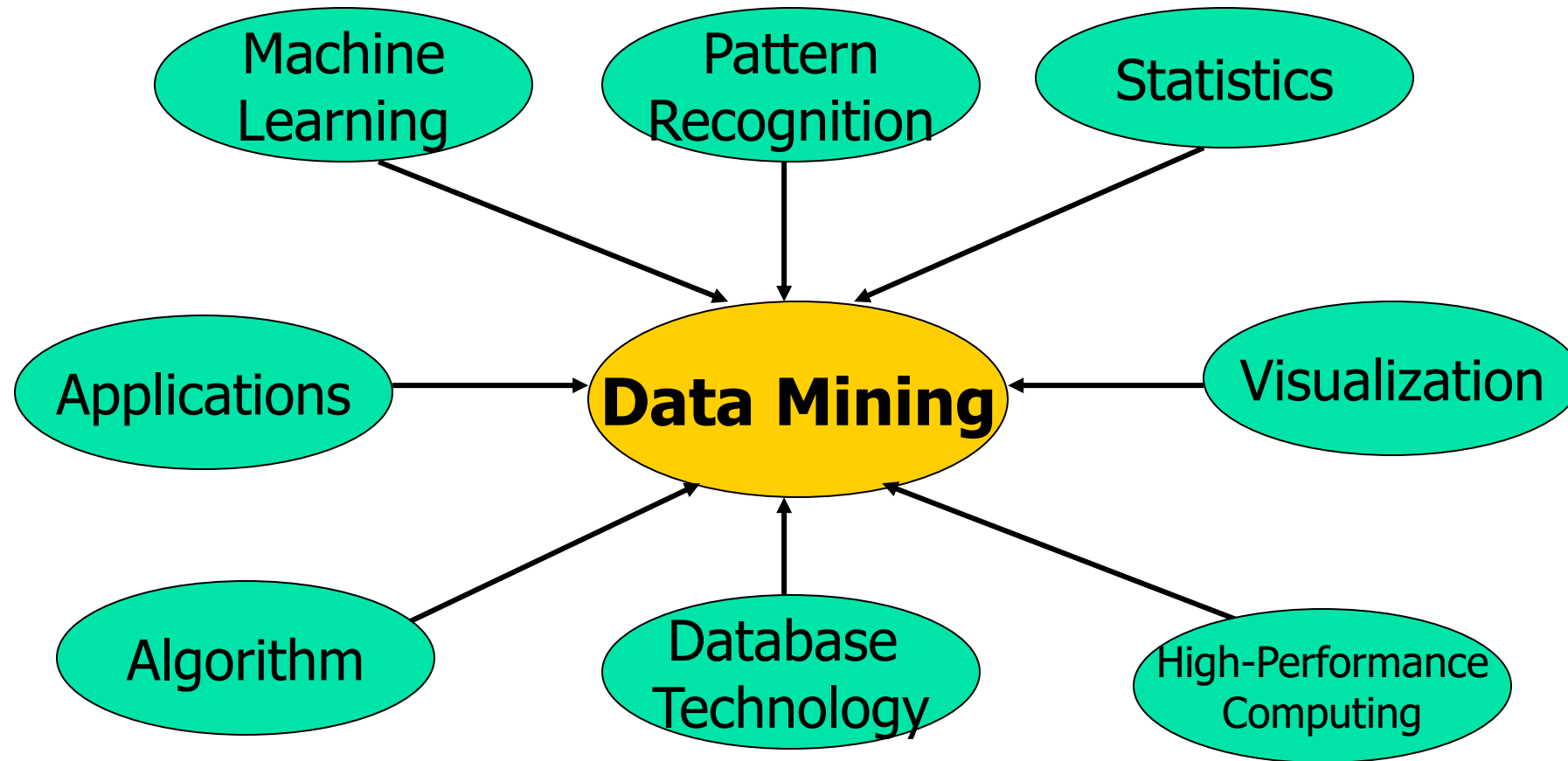
Data Mining Function: (5) Outlier Analysis

- Outlier analysis
 - Outlier: A data object that does not comply with the general behavior of the data
 - 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

Evaluation of Knowledge

- Are all mined knowledge interesting?
 - One can mine tremendous amount of “patterns” and knowledge
 - Some may fit only certain dimension space (time, location, ...)
 - Some may not be representative, may be transient, ...
- Evaluation of mined knowledge → directly mine only interesting knowledge?
 - Descriptive vs. predictive
 - Coverage
 - Typicality vs. novelty
 - Accuracy
 - Timeliness
 - ...

Data Mining: Confluence of Multiple Disciplines



Why Confluence of Multiple Disciplines?

- Tremendous amount of data
 - Algorithms must be highly scalable to handle such as tera-bytes of data
- High-dimensionality of data
 - Micro-array may have tens of thousands of dimensions
- High complexity of data
 - Data streams and sensor data
 - Time-series data, temporal data, sequence data
 - Structure data, graphs, social networks and multi-linked data
 - Heterogeneous databases and legacy databases
 - Spatial, spatiotemporal, multimedia, text and Web data
 - Software programs, scientific simulations
- New and sophisticated applications

Applications of Data Mining

- Web page analysis: from web page classification, clustering to PageRank & HITS algorithms
- Collaborative analysis & recommender systems
- Basket data analysis to targeted marketing
- Biological and medical data analysis: classification, cluster analysis (microarray data analysis), biological sequence analysis, biological network analysis
- Data mining and software engineering (e.g., IEEE Computer, Aug. 2009 issue)
- From major dedicated data mining systems/tools (e.g., SAS, MS SQL-Server Analysis Manager, Oracle Data Mining Tools) to invisible data mining

Why Data Mining?—Potential Applications

- **Data analysis and decision support**

- Market analysis and management
 - Target marketing, customer relationship management (CRM), market basket analysis, cross selling, market segmentation
- Risk analysis and management
 - Forecasting, customer retention, improved underwriting, quality control, competitive analysis
- Fraud detection and detection of unusual patterns (outliers)

- **Other Applications**

- Text mining (news group, email, documents) and Web mining
- Stream data mining
- Bioinformatics and bio-data analysis

Ex. 1: Market Analysis and Management

- Where does the data come from?—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
- Cross-market analysis—Find associations/co-relations between product sales, & predict based on such association
- Customer profiling—What types of customers buy what products (clustering or classification)
- Customer requirement analysis
 - Identify the best products for different groups of customers
 - Predict what factors will attract new customers
- Provision of summary information
 - Multidimensional summary reports
 - Statistical summary information (data central tendency and variation)

Ex. 2: Corporate Analysis & 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

Ex. 3: Fraud Detection & Mining Unusual Patterns

- Approaches: Clustering & model construction for frauds, outlier analysis
- Applications: Health care, retail, credit card service, telecomm.
 - Auto insurance: ring of collisions
 - Money laundering: suspicious monetary transactions
 - Medical insurance
 - Professional patients, ring of doctors, and ring of references
 - Unnecessary or correlated screening tests
 - Telecommunications: phone-call fraud
 - Phone call model: destination of the call, duration, time of day or week. Analyze patterns that deviate from an expected norm
 - Retail industry
 - Analysts estimate that 38% of retail shrink is due to dishonest employees
 - Anti-terrorism

Major Issues in Data Mining

- Mining methodology
 - Mining different kinds of knowledge from diverse data types, e.g., bio, stream, Web
 - Performance: efficiency, effectiveness, and scalability
 - Pattern evaluation: the interestingness problem
 - Incorporation of background knowledge
 - Handling noise and incomplete data
 - Parallel, distributed and incremental mining methods
 - Integration of the discovered knowledge with existing one: knowledge fusion
- User interaction
 - Data mining query languages and ad-hoc mining
 - Expression and visualization of data mining results
 - Interactive mining of knowledge at multiple levels of abstraction

Major Issues in Data Mining (2)

- 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 Books

Main Book

- **Han, J., Kamber, M. and Pei, J., 2011. Data mining concepts and techniques 3rd edition. *The Morgan Kaufmann Series in Data Management Systems.***

Reference Books

- Andrea Cirillo (2017) *R Data Mining : Mine Valuable Insights From Your Data Using Popular Tools and Techniques in R.* Birmingham, UK: Packt Publishing. **(available on ebscohost)**
- Zhao, Y., 2012. *R and Data Mining: Examples and Case Studies.* Academic Press.
- Torgo, L., 2011. *Data Mining with R: Learning with Case Studies.* Chapman and Hall/CRC.
- Layton, R., 2017. *Learning Data Mining with Python.* Packt Publishing Ltd.
- Madhavan, S., 2015. *Mastering Python for Data Science.* Packt Publishing Ltd.
- Hastie, T., Tibshirani, R. and Friedman, J., 2009. *The elements of statistical learning: data mining, inference, and prediction.* Springer Science & Business Media.
- Tan, P.N., Steinbach, M. and Kumar, V., 2016. *Introduction to data mining.* Pearson Education India.
- Weiss, S.M. and Indurkha, N., 1998. *Predictive data mining: a practical guide.* Morgan Kaufmann.

Summary

- Data mining: Discovering interesting patterns and knowledge from massive amount 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 data
- Data mining functionalities: characterization, discrimination, association, classification, clustering, outlier and trend analysis, etc.
- Data mining technologies and applications
- Major issues in data mining