Data Mining
Introduction

Organization

- Lectures
  - Mondays and Thursdays from 10:30 to 12:30
  - Lecturer: Mouna Kacimi
  - Office hours: appointment by email
- Labs
  - Thursdays from 14:00 to 16:00
  - Teaching Assistant: Mouna Kacimi
- Course Webpage: http://www.inf.unibz.it/~mkacimi/teaching.shtml
- Textbooks
  - Jiawei Han and Micheline Kamber, "Data Mining: Concepts and Techniques", Second Edition, 2006
  - Pang-Ning Tan, Michael Steinbach, and Vipin Kumar, "Introduction to Data Mining", Pearson Addison Wesley, 2008, ISBN: 0-32-134136-7

Project

- During Lab hours
- The project will be divided into small tasks, a new task every week
- The project can be done individually
- Groups of no more than 2 students are allowed
- You need to know how to program. If you do not know, team up with someone who knows
- You have the option to do a free project on your own: your proposal needs to be approved by the teacher

Exam Procedure

- Requirement: obtain 18 credit points in each of the following:
  - Project
  - Exam

$$\text{Final Grade} = 0.5 \times \text{Project Grade} + 0.5 \times \text{Exam Grade}$$

- Exams
  - Midterm Exam (optional) : 15 points
  - Final Exam
    - Full: 30 points
    - Partial: 15 points

$$\text{Exam Grade} = \begin{cases} \text{Midteram Grade+Partial Exam Grade} & \text{if student took the midtem exam} \\ \text{Full Exam Grade} & \text{if student did not take the midterm exam} \\ & \text{or decided not to consider the midterm exam} \end{cases}$$

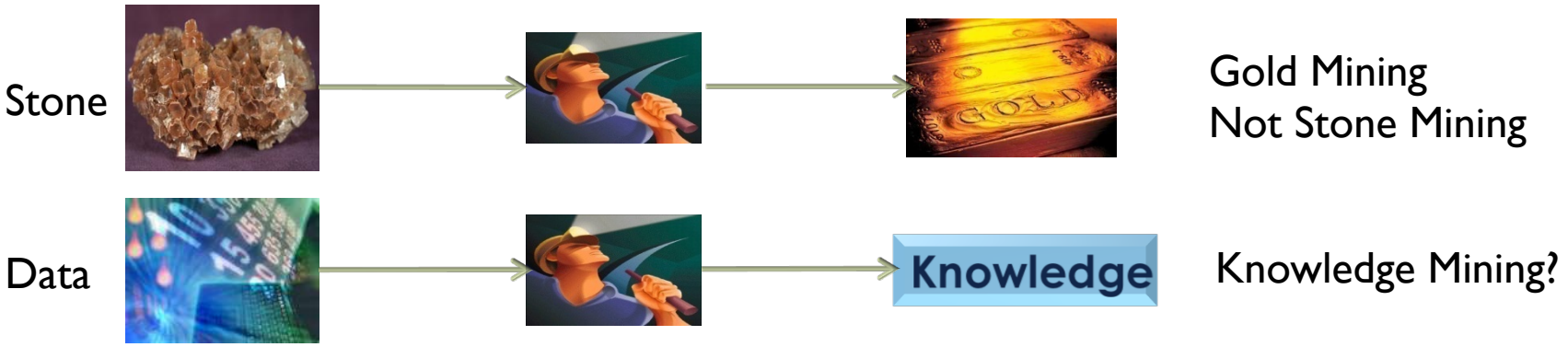
Exam Procedure

- Students must have a successful project to be able to take the final exam
- A successful project remains valid even when the student fails the exam
- If a project is unsuccessful until the day of the exam, its validity expires
- Students can do a new project until the next exam session. In this case, the teaching assistant does not guarantee support for supervising the students.

Road Map

1. Definitions & Motivations
2. Data to be mined
3. Knowledge to be discovered
4. Major Issues in Data Mining

Data Mining: what does it?

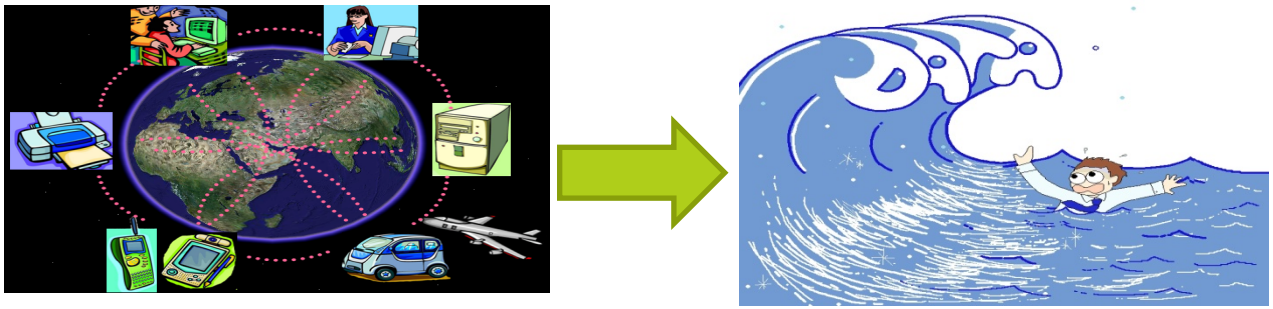


- Data mining (knowledge discovery from data)
  - Extraction of interesting (non-trivial, implicit, previously unknown and potentially useful) patterns or knowledge from huge amount of data

- Alternative names
  - Knowledge discovery (mining) in databases (KDD), knowledge extraction, data/pattern analysis, data archeology, data dredging, information harvesting, business intelligence, etc.

Why Data Mining?

- Explosive Growth of Data: from terabytes to petabytes
- Data Collections and Data Availability
  - Crawlers, database systems, Web, etc.



  - Sources
    - Business: Web, e-commerce, transactions, etc.
    - Science: Remote sensing, bioinformatics, etc.
    - Society and everyone: news, YouTube, etc.
- **Problem**: We are drowning in data, but starving for knowledge!
- **Solution**: Use Data Mining tools for Automated Analysis of massive data sets

What Data Mining is Used For?

Financial Data Analysis

- Banks and Institutions offer a wide variety of banking services
  - Checking and saving accounts for business or individual customers
  - Credit business, mortgage, and automobile loans
  - Investment services (mutual funds)
  - Insurance services and stock investment services
- Financial data is relatively complete, reliable, and of high quality
- What to do with this data?

What Data Mining is Used For?

Financial Data Analysis

- Loan Payment Prediction and costumer credit policy analysis
  - Attribute selection and attribute relevance ranking may help identifying important factors and eliminate irrelevant ones
  - Example of factors related to the risk of loan payment
    - Term of the loan
    - Debt ratio
    - Payment to income ratio
    - Customer level income
    - Education level
    - Residence region
  - The bank can adjust its decisions according to the subset of factors selected

What Data Mining is Used For?

Retail Industry

- Collect huge amount of data on sales, customer shopping history, goods transportation, consumption and service, etc.
- Many stores have web sites where you can buy online. Some of them exist only online (e.g., Amazon)
- Data mining helps to
  - Identify costumer buying behaviors
  - Discover customers shopping patterns and trends
  - Improve the quality of costumer service
  - Achieve better costumer satisfaction
  - Design more effective good transportation
  - Reduce the cost of business

What Data Mining is Used For?

- Many different ways of communicating
  - Fax, cellular phone, Internet messenger, images, e-mail, computer and Web data transmission, etc.

- Great demand of data mining to help
  - Understanding the business involved
  - Identifying telecommunication patterns
  - Catching fraudulent activities
  - Making better use of resources
  - Improve the quality of service

Example of a Data Mining Problem

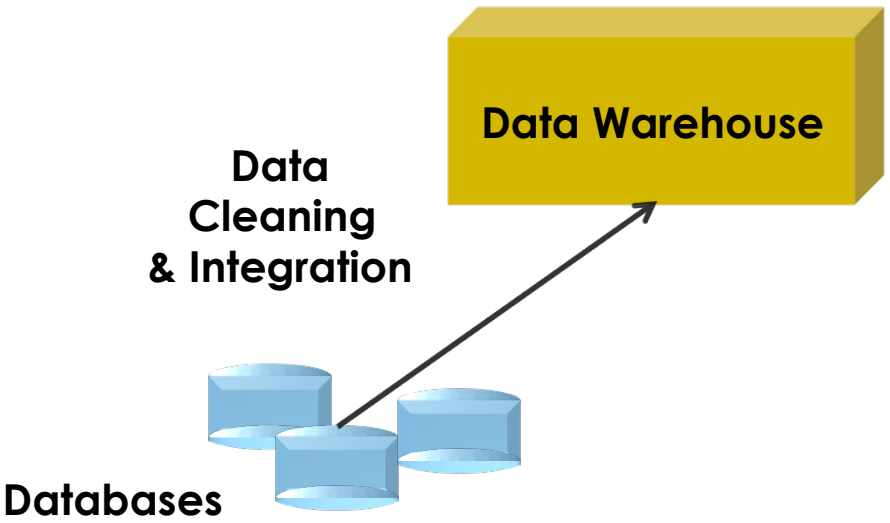
- You want to do advertisement of sport activities for a set of new users on Facebook

- These people do not have explicit information about what they like and no history of past activities

- What you know is:
  - Their age, gender, and location
  - Their friends (not necessarily new users)
  - Messages they exchange with friends

Knowledge Discovery (KDD) Process

- Data Mining as a step in the knowledge discovery process

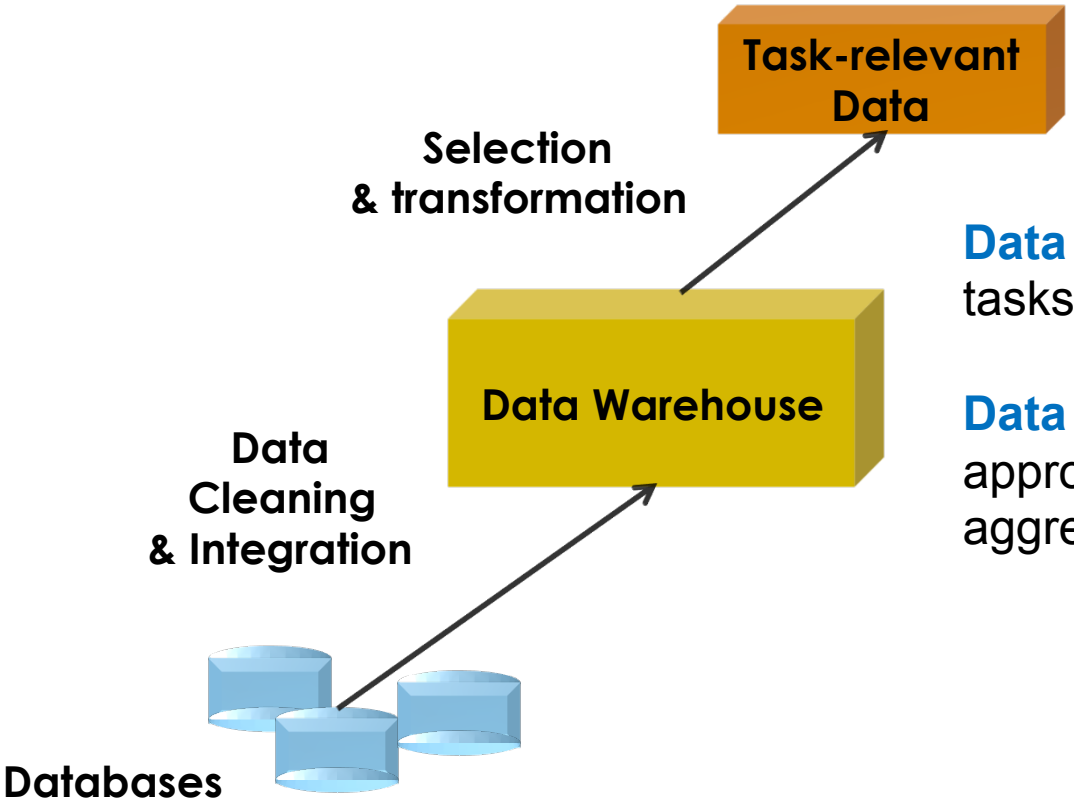


Data Cleaning: Remove noise and inconsistent data

Data Integration: Combine multiple data sources

Knowledge Discovery (KDD) Process

- Data Mining as a step in the knowledge discovery process

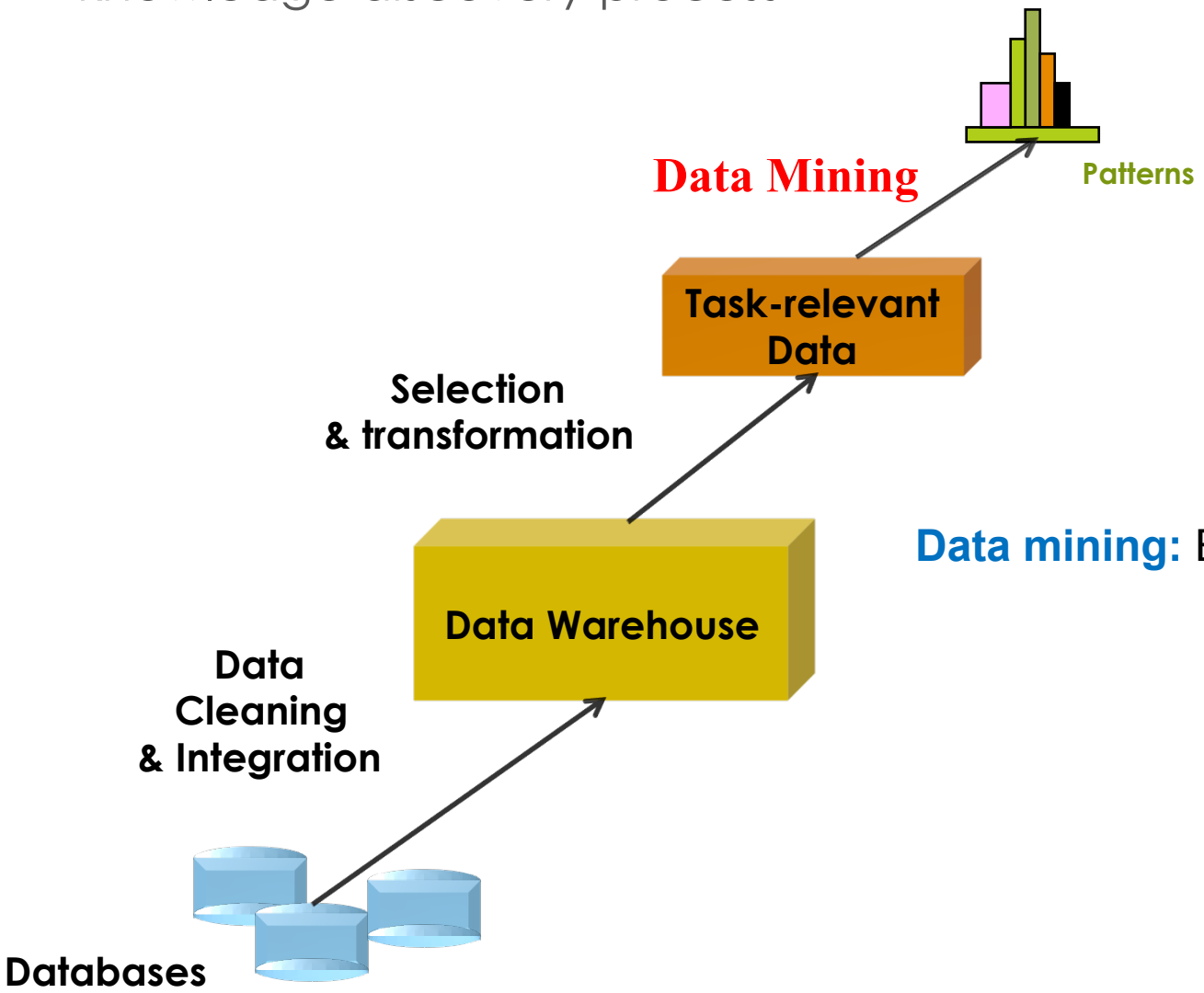


Data Selection: Data relevant to analysis tasks are retrieved form the data

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

Knowledge Discovery (KDD) Process

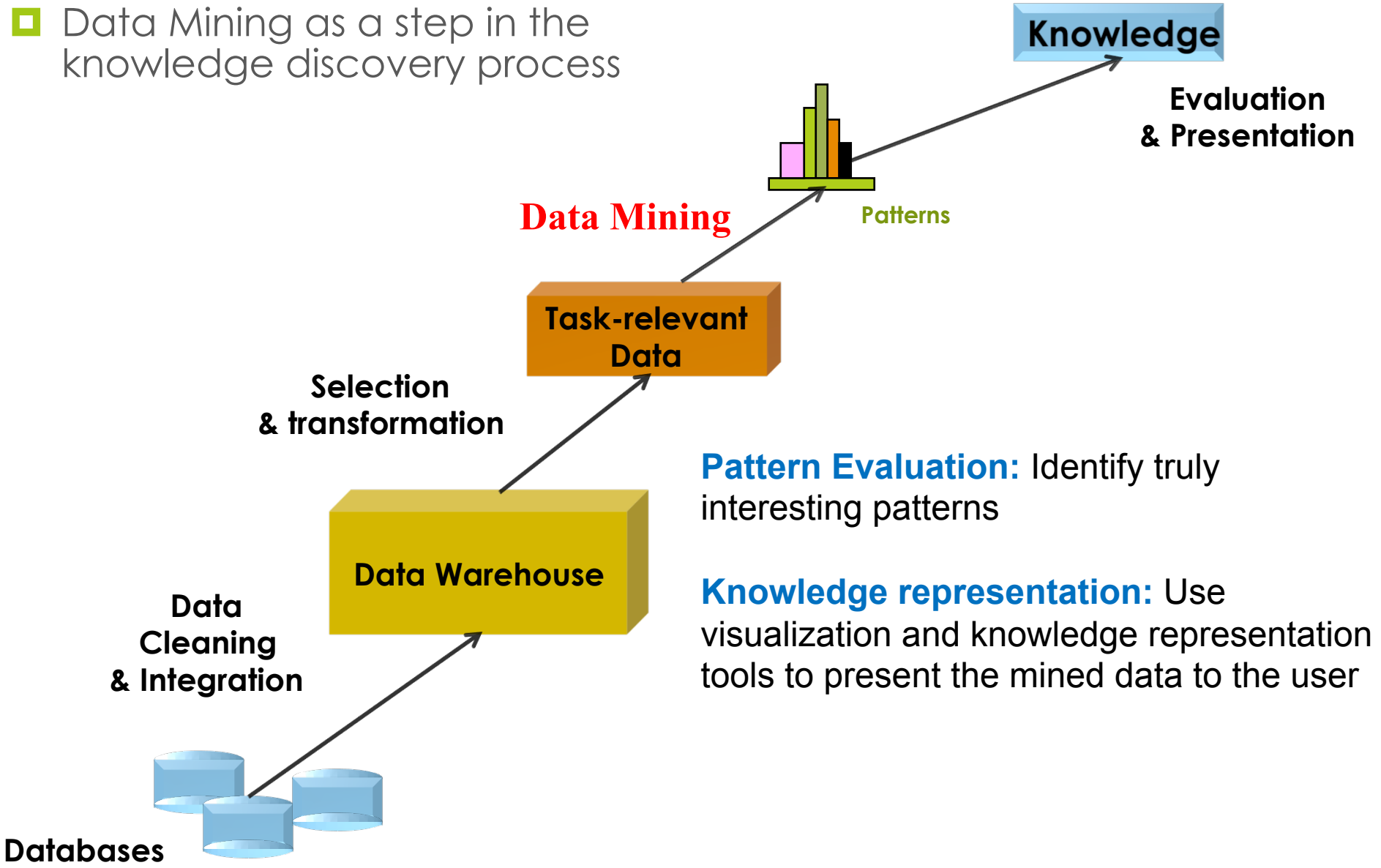
- Data Mining as a step in the knowledge discovery process



Data mining: Extract data patterns

Knowledge Discovery (KDD) Process

- Data Mining as a step in the knowledge discovery process

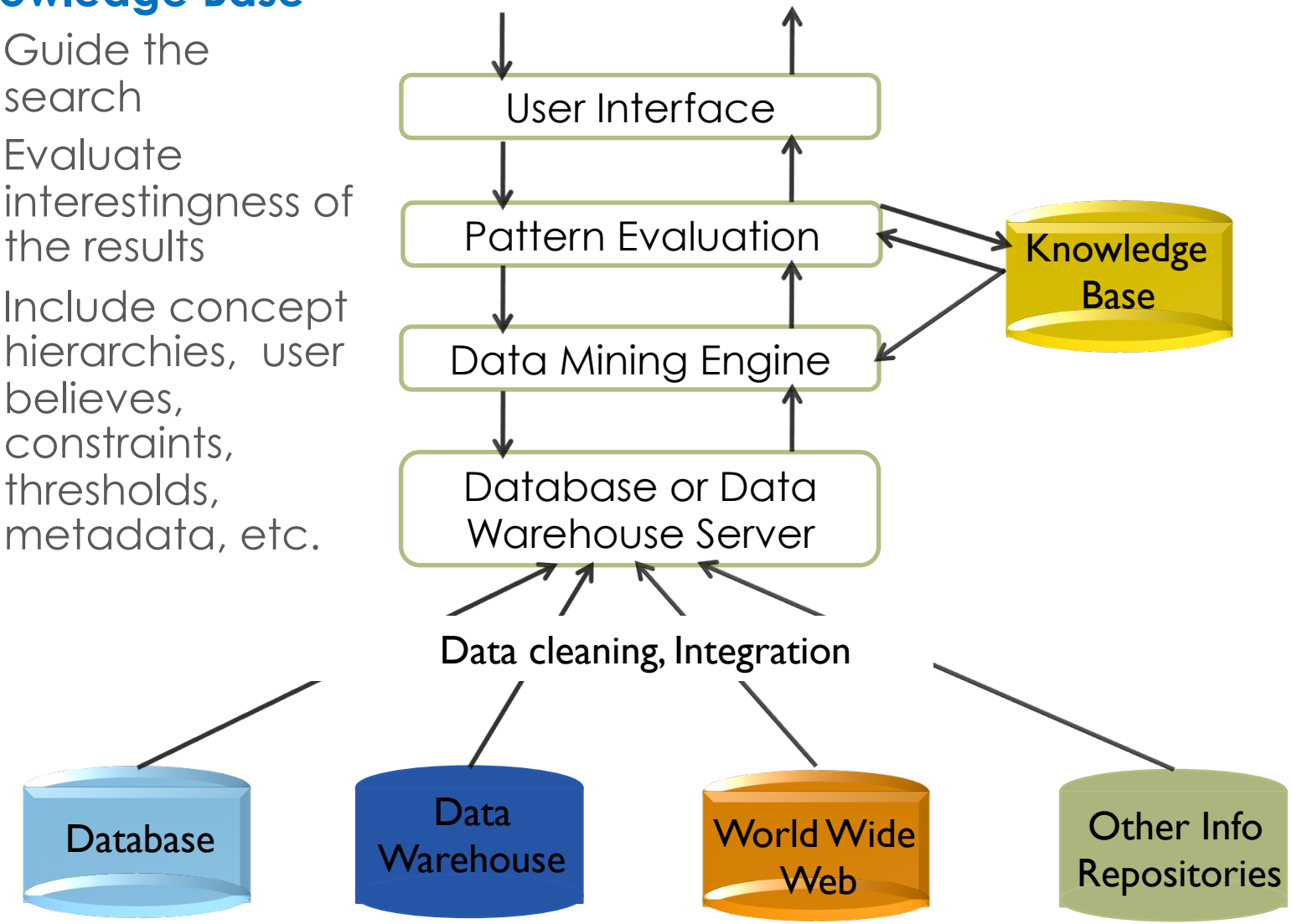


Pattern Evaluation: Identify truly interesting patterns

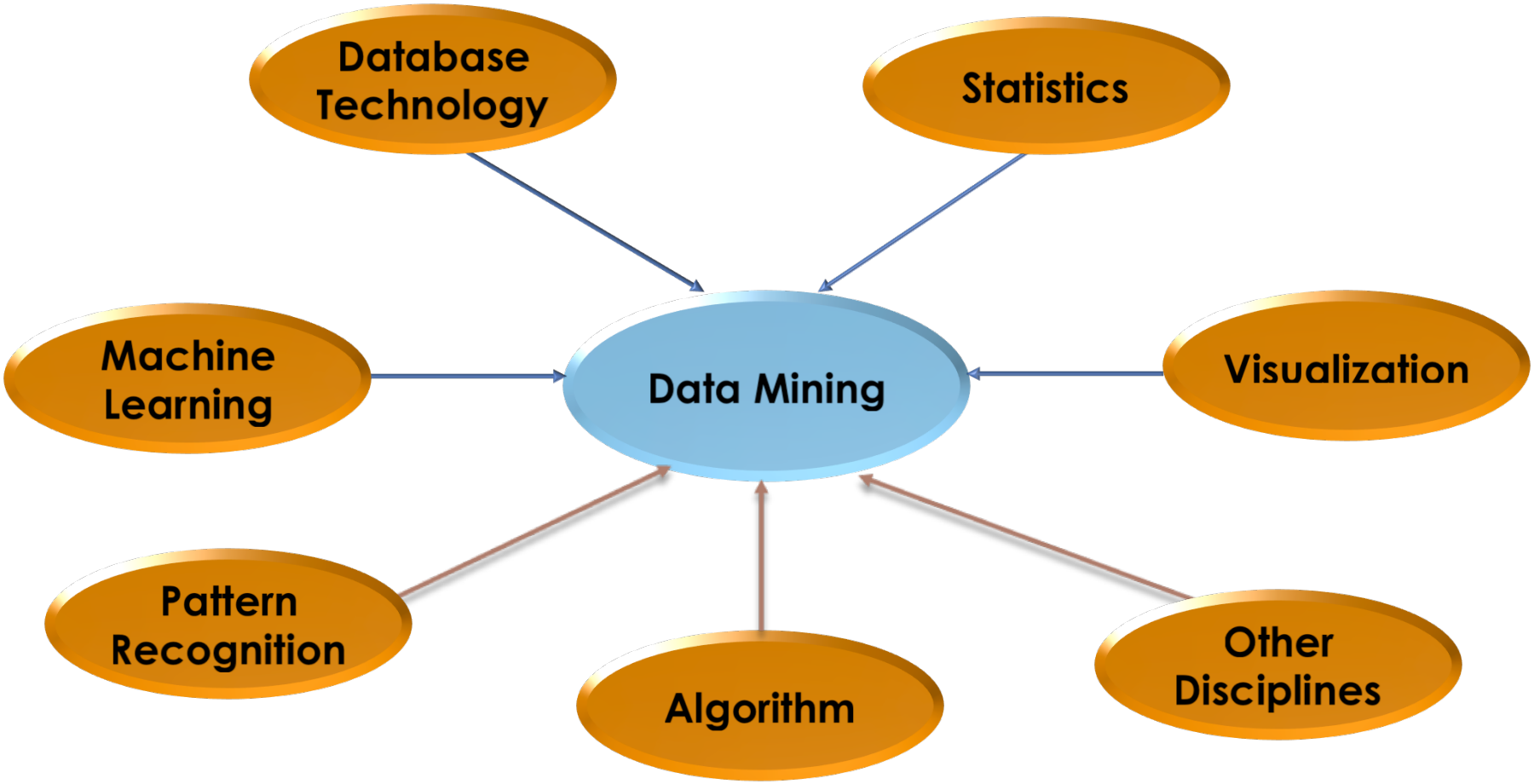
Knowledge representation: Use visualization and knowledge representation tools to present the mined data to the user

Typical Architecture of a DM System

- **Knowledge Base**
  - Guide the search
  - Evaluate interestingness of the results
  - Include concept hierarchies, user believes, constraints, thresholds, metadata, etc.



Confluence of Multiple Disciplines



Why Confluence of Multiple Disciplines?

- **Tremendous amount of data**
  - Scalable algorithms to handle terabytes of data (e.g., Flickr hits 6 billion images but facebook does that every 2 months http://thenextweb.com/socialmedia/2011/08/05/flickr-hits-6-billion-total-photos-but-facebook-does-that-every-2-months//)
- **High dimensionality of data**
  - Data can have tens of thousands of features (e,g., DNA microarray)
- **High complexity of data**
  - Data can be highly complex, can be of different types, and can include different descriptors
    - Images can be described using text and visual features such as color, texture, contours, etc.
    - Videos can be described using text, images and their descriptors, audio phonemes, etc.
    - Social networks can have a complex structure…
- **New and sophisticated applications**

Different Views of Data Mining

- **Data View**
  - Kinds of data to be mined
- **Knowledge view**
  - Kinds of knowledge to be discovered
- **Method view**
  - Kinds of techniques utilized
- **Application view (seen before)**

Road Map

1. Definitions & Motivations
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Data to be Mined

- In principle, data mining should be applicable to any data repository

- This lecture includes examples about:

  - Relational databases
  - Data warehouses
  - Transactional databases
  - Advanced database systems

Relational Databases

- **Database System**
  - Collection of interrelated data, known as **database**
  - A set of software programs that manage and access the data
- **Relational Databases (RD)**
  - A collection of tables. Each one has a unique name
  - A table contains a set of attributes (columns) & tuples (rows).
  - Each object in a relational table has a unique key and is described by a set of attribute values.
  - Data are accessed using database queries (SQL): projection, join, etc.

Costumers

cust_Id	Name	age	income
152	Anna	27	24000 €
...	...	...	...

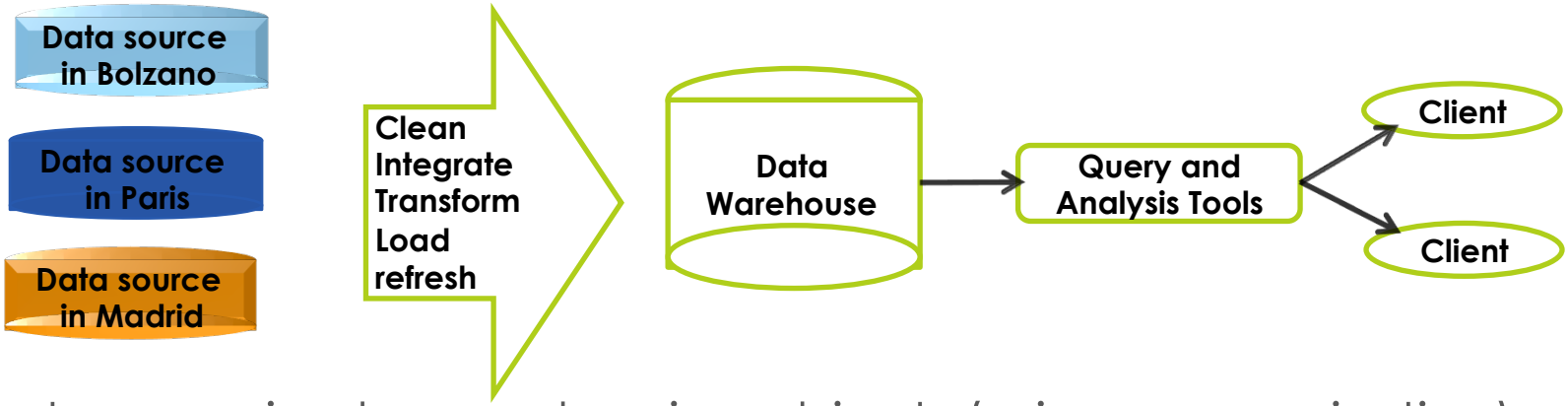
Purchases

trans_Id	cust_Id	method	Amount
T156	152	Visa	1357 €
...	...	...	...

- **Data Mining applied to RD**
  - Search for trends or data patterns
  - Example: predict the credit risk of costumers based on their income, age and expenses.

Data Warehouses

- A data warehouse (DW) is a repository of information collected from multiple sources, stored under a unified schema.



- Data organized around major subjects (using summarization)
- Multidimensional database structure (e.g., data Cube)
  - Dimension = one attribute or a set of attributes
  - Cell = stores the value of some aggregated measures.
- **Data Mining applied to DW**
  - Data warehouse tools help data analysis
  - Data Mining tools are required to allow more in-depth and automated analysis

Transactional Databases

- A transactional database (TD) consists of a file where each record represents a transaction.
- A transaction includes a unique transaction identifier (trans_id) and a list of the items making the transaction.
- A transaction database may include other tables containing other information regarding the sale (customer_Id, location, etc.)
- Basic analysis (examples)
  - Show me all the items purchased by David Winston?
  - How many transactions include item number 5?
- **Data Mining on TD**
  - Perform a deeper analysis
  - Example: Which items sold well together?
  - Basically, data mining systems can identify frequent sets in transactional databases and perform *market basket data analysis*.

trans_Id	List of items_IDs
T100	I1,I3,I8,I16
T200	I2,I8
...	...

Advanced Database Systems (1)

- Advanced database systems provide tools for handling **complex** data
  - Spatial data (e.g., maps)
  - Engineering design data (e.g., buildings, system components)
  - Hypertext and multimedia data (text, image, audio, and video)
  - Time-related data (e.g., historical records)
  - Stream data (e.g., video surveillance and sensor data)
  - World Wide Web, a huge, widely distributed information repository made available by Internet
- Require **efficient** data structures and **scalable** methods to handle
  - Complex object structures and variable length records
  - Semi structured or unstructured data
  - Multimedia and spatiotemporal data
  - Database schema with complex and dynamic structures

Advanced Database Systems (2)

- **Example: World Wide Web**
  - Provide rich, worldwide, online and distributed information services.
  - Data objects are linked together
  - Problems
    - Data can be highly unstructured
    - Understand the semantic of web pages

- **Data Mining on WWW**
  - Web usage Mining (user access pattern)
    - Improve efficiency and make better marketing decisions
  - Authoritative Web page Analysis
    - Ranking web pages based on their importance
  - Automated Web page clustering and classification
    - Group and arrange web pages based on their content
  - Web community analysis
    - Identify hidden web social networks and observe their evolution

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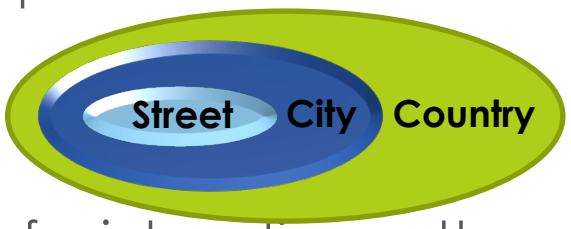
Knowledge to be Discovered

- Data mining functionalities are used to specify the kind of patterns to be found in data mining tasks
- Data mining tasks can be classified into two categories
  - **Descriptive :** Characterize the general properties of the data
  - **Predictive :** Perform inference on the current data to make predictions
- **What to extract?**
  - Users may not have an idea about what kinds of patterns in their data can be interesting
- **What to do?**
  - Have a data mining system that can mine multiple types of patterns to handle different user and application needs.
  - Discover patterns at various granularities (levels of abstraction)
  - Allow users to guide the search for interesting patterns



Example of different granularities

Characterization and Discrimination (1)

- Data can be associated with classes or concepts

Example of data from a store

Classes

printers

computers

Concepts

Big-Spenders

Budget-Spenders

- **Class/Concept descriptions:** describe individual classes and concepts in summarized, concise, and precise way.
  - Data characterization
    - Summarize the data of the class under study (**target class**)
  - Data Discrimination
    - Compare the target class with a set of comparative classes (**contrasting classes**)
  - Data characterization & Discrimination
    - Perform both analysis

Characterization and Discrimination (2)

Data Characterization

- Output: charts, curves, multidimensional data cubes, etc.
- Example

Summarize the characteristics of costumers who spend more than 1000€ → **Costumers profile**

- **40-50 years old**
- **Employed**
- **excellent credit ratings**

Data Discrimination

- Output: similar to characterization + comparative measures
- Example

Compare customers who shop for computer products regularly(more than 2 times a month) with those who rarely shop for such products(less then three times a year) →

Comparative profile

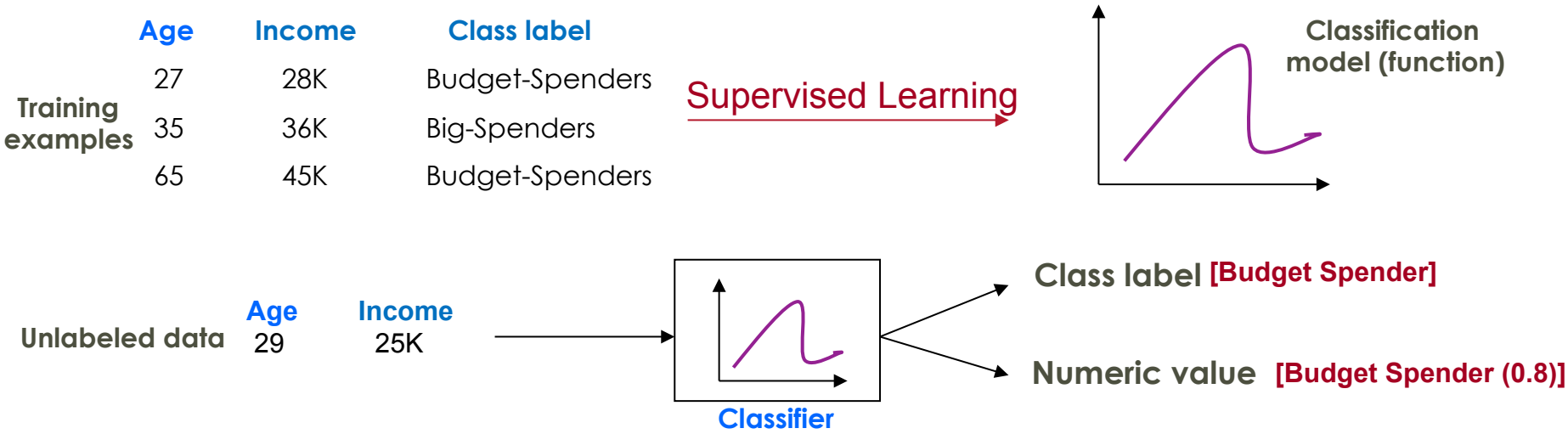
Frequent costumers	**Rare costumers**
80%	60%
•Are between 20 and 40	•Are senior or youths
•Have university education	•Have no university degree

Frequent Patterns, Associations, Correlations

- **Frequent patterns** are patterns occurring frequently in the data (e.g., item-sets, sub-sequences, and substructures)
  - Frequent item-sets: items that frequently appear together
    - Example in a transactional data set: **bread** and **milk**
  - Frequent Sequential pattern: a frequently occurring subsequence
    - Example in a transactional data set: buy **first PC**, **second digital camera**, **third memory card**
- **Association Analysis**
  - Derive some association rules
    - *buys(X, "computer") ⇒ buys (X, "software") [support =1%, confidence=50%]*
    - *age(X, "20...29") ∧ income(X, "20K...29K") ⇒ buys (X, "CD player") [support =2%, confidence=60%]*
- **Correlation Analysis**
  - Uncover interesting statistical correlations between associated attribute-value pairs

Classification & Prediction

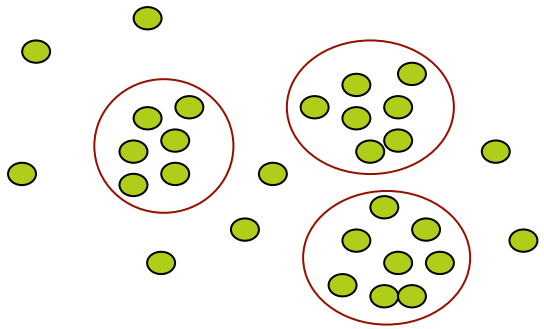
- Construct models (functions) based on some training examples
- Describe and distinguish classes or concepts for future prediction
- Predict some unknown class labels



- **Typical Models**: Decision trees, Bayesian classifiers, Regression, etc.
- **Typical Application**s: Credit card fraud detection, classifying web pages, stars, diseases, etc

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



- **Typical methods**: Hierarchical, density-based, Grid-based, Model-Based, constraint-based , etc.
- **Typical Applications**: WWW, social networks, Marketing, Biology, Library, etc.

Outlier Analysis

- **Outlier**: A data object that does not comply with the general behavior of the data learning (i.e., Class label is unknown)
- Noise or exception? — One person's garbage could be another person's treasure


Or

?

- **Typical methods:** Product of clustering or regression analysis, etc
- **Typical Applications:** Useful in fraud detection:
  - How to uncover fraudulent usage of credit card?
  - Detect purchases of extremely large amounts for a given account number in comparison to regular charges incurred by the same account
  - Outliers may also be detected with respect to the location and type of purchase, or the frequency.

Road Map

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Major Challenges of Data Mining

- Efficiency and scalability of data mining algorithms
- Parallel, distributed, stream, and incremental mining methods
- Handling high-dimensionality, noise, uncertainty, and incompleteness of data
- Incorporation of constraints, expert knowledge, and background knowledge in data mining
- Pattern evaluation and knowledge integration
- Mining diverse and heterogeneous kinds of data
- Application-oriented and domain-specific data mining
- Protection of security, integrity, and privacy in data mining

Summary

- Data Mining is a process of extracting knowledge from data
- Data to be mined can be of any type
  - Relational Databases, Advanced databases, etc.
- Knowledge to be discovered
  - Frequent patterns, correlations, associations, classification, prediction, clustering
- Data Mining is interdisciplinary
  - Large amount of complex data and sophisticated applications
- Challenges of data Mining
  - Efficiency, scalability, parallel and distributed mining, handling high dimensionality, handling noisy data, mining heterogeneous data, etc.