

# Arsi University Dep't of Information Technology

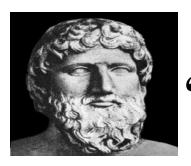
# Introduction to Data Warehousing and Data Mining

# Contents of the course

- ✓ Chapter 1: Introduction to Dm. and data warehousing
- √ Chapter 2: Data Preprocessing
- √ Chapter 3: Classification
- √ Chapter 4: Clustering
- **✓ Chapter 5: Associations**

#### Introduction

- In this chapter we will cover the following issues in brief
  - Motivation: Why data mining?
  - What is data mining?
  - Data mining vs Statistics
  - Challenges in Data Mining
  - Application of data mining
  - Data mining functionality
  - Are all the patterns interesting?
  - Classification of data mining systems



- Our capacity of generating and collecting data have been increased rapidly in the last several decades
- Huge amount of data is available at the tip of our hand
- It is predicted that more data will be produced in the next 2 years than has been generated during the entire existence of humankind!

#### **Motivation:**

### "Necessity is the Mother of Invention"

- Contributing factors include
  - Widespread use of bar code for most commercial products,
  - 40 billion RFID tags world wide
  - Billions of telephone calls are recorded daily worldwide
  - Billions of customers are using face book and other social network applications
  - 10 billions of content are shared on face book per month
  - Computerization of many business, scientific, and governmental transactions,
  - Advances in data collection tools (audio, video, satellite, remote sensing, scanning, image capturing tools)
  - Usage of WWW as a global information system
  - comprehensive application software,
  - new computing and storage technologies

#### **Motivation:**

### "Necessity is the Mother of Invention"

- All this have made it easier to create, collect, and store all types of data.
- As a result it creates a problem what is called data exposition.
- Data explosion is the problem of having huge amount of data in an enterprise stored in databases, data warehouses and other information repositories generated by automated
- data collection tools and mature database technology in large databases which has to be processed to make a decision.
- As the size of data get larger, analyzing the data becomes very difficult

- Data can be managed and stored in
  - -Data warehouse
  - -structured databases;
  - -in semi-structured file systems, such as e-mail;
  - -unstructured fixed content, like documents and graphic files.

- Companies rely on this enterprise data to improve decision-making and to gain a competitive advantage;
- Data has indeed become a highly valued business asset.
- The huge amount of data exceeds our human ability to make comprehension on the data and to put the best decision without tools
- Generating and storing of large volumes of data has reached a critical mass and appropriate tools for comprehend the data becomes vital.

- We are drowning in data, but starving for knowledge!
- The Solution: Data warehousing and data mining
- Data mining can be viewed as a result of the natural evolution of information technology.
- This can be more explained if we look at the evolution of database technology since 19<sup>th</sup> century.

#### **Motivation:**

# "Necessity is the Mother of Invention"

#### • 1960s:

- Known to be the era of primitive file processing
- There were activities such as
  - Data collection,
  - database creation,
  - Information management system (IMS), mainly using COBOL

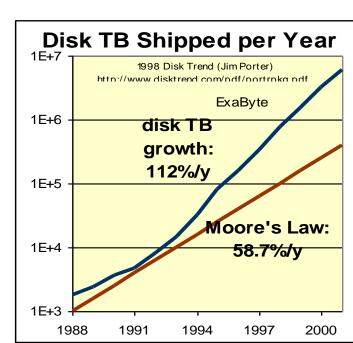
#### • 1970s:

- Relational data model, relational DBMS implementation
- Data modeling tools like ER diagram
- Indexing and data organization techniques such as B+ tree, hashing, etc
- Query language such as SQL
- User interfaces, forms and reports
- Query processing and optimization techniques
- Transaction management: recovery, concurrency control, etc
- Online Transaction processing (OLTP)

- 1980s:
  - Period of advanced DB Systems
    - advanced data models
      - extended-relational, Object Oriented, Object-Relational, deductive, etc.)
    - application-oriented DBMS
      - -spatial, temporal, multimedia, active, scientific, engineering, Knowledgebase, etc.)
- 1990s—2000s:
  - Data mining and data warehousing, Knowledge discovery,
     OLAP and Web based databases

# **Data Mining Enablers**

- Explosion of data
- Fast and cheap computation and storage
  - Moore's Law: processing doubles every 19 months
  - Disk storage doubles every 9 months
  - Database technology
- Competitive pressure in business
  - Data has value!
- New, successful models
- Commercial products
  - SAS, SPSS, Insightful, IBM, Oracle
  - Open Source products
    - Weka









- Data mining is extraction of interesting (non-trivial, implicit, previously unknown and potentially useful) information or patterns from data source(Han and Kamber)
- The process of discovering meaningful new correlations, patterns, and trends by sifting through large amounts of data stored in repositories and
- by using pattern recognition technologies as well as statistical and mathematical techniques (The Gartner Group)
- The exploration and analysis of large quantities of data in order to discover meaningful patterns and rules (Berry and Linoff)
- The nontrivial extraction of implicit, previously unknown, and potentially useful information from data (Frawley, Paitestsky-Shapiro and Mathews)
- The non-trivial discovery of novel, valid, comprehensible and potentially useful patterns from data (Fayyad et. al).
- Focused on hypothesis generation, not on hypothesis testing







- The term Data mining is a misnomer as it doesn't directly related to what is does.
- For exampling mining gold from rock is called Gold mining but not rock mining.
- Similarly oil mining is mining oil from the ground.
- Data mining should best describe as knowledge mining from data rather that data mining
- Any way, we will use the term with this understanding

#### • Alternative names

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

#### • *Note that:*

 query processing systems, Expert statistical data analysis or Information retrieval systems are not data mining tasks

- Sample pattern you might find
  - ■Supermarket data
    - On Thursday nights people who buy diapers also tend to buy beer
  - Insurance company data
    - People with good credit ratings are less likely to have accidents
  - Telecom data
    - ■Government lines are busy than private line

# Statistics vs. Data Mining

Statistics	Data Mining
Confirmative	Explorative
Small data sets/ File-based	Large data sets/ Databases
Small number of variables	Large number of variables
Deductive	Inductive
Numeric data	Numeric and non-numeric (including txt, networks)
Clean data	Data cleaning

# Data Mining vs. Statistics

- Statistics is known for:
  - -well defined hypotheses used to learn about a topic
  - Work on specifically chosen population
  - -Require carefully collected data for inferences well known properties.
- Data mining isn't that careful. It is:
  - -data driven discovery of pattern
  - -observational data sets is needed (data collected as side issue of other operations)

# Data Mining vs. Statistics

#### • Traditional statistics

- -first hypothesize, then collect data, then analyze
- often model-oriented (strong parametric models)

#### • Data mining:

- -few if any a priori hypotheses
- data is usually already collected a priori
- analysis is typically data-driven not hypothesisdriven
- Often algorithm-oriented rather than model-oriented

#### **Challenges in Data Mining**

- > Efficiency and scalability of data mining algorithms
- Parallel, distributed, stream, and incremental mining methods
- ➤ Handling high-dimensionality
- Handling noise, uncertainty, and incompleteness of data
- Incorporation of constraints, expert knowledge, and background knowledge
- ➤ Pattern evaluation and knowledge integration
- ➤ Mining diverse and heterogeneous kinds of data: e.g., bioinformatics, Web,
- > Application-oriented and domain-specific data mining
- ➤ Invisible data mining (embedded in other functional modules)
- Protection of security, integrity, and privacy in data mining

# **Potential Applications of Data Mining**

- Market analysis and management
  - -target marketing analysis
  - -Market basket analysis
  - -Customer cross selling Analysis
  - -customer purchase pattern analysis
  - -Market segmentation
- Fraud detection and management



# Data source for DM applications

- Where are the data sources for analysis?
  - -Credit card transactions,
  - -loyalty cards,
  - -discount coupons,
  - -customer complaint calls,
  - -Customer calls
  - -Log files
  - Transaction files etc...

### Market Basket Analysis

- It is a processes of modeling item-set that consumers will put into his/her basket in one shopping
- This permits seller to arrange item-set so that consumers will find them easily

#### **Customer Cross-Selling Analysis**

- It is a processes of modeling item-set that consumers will purchase them at different time so that if customer buys item X them the business will recommend item Y which goes together
- This permits seller to maximize their profit, motivate their customers and improve their business strategy

# **Target Market Analysis**

- It is the process of identifying cluster of customers who will buy your service
- These customers share the same characteristics
- Target market analysis is the process of identifying (modeling) such groups of individuals

### **Market Segmentation**

- It is the process of dividing the market into different homogeneous groups of consumers
- This better satisfy customers as they can choose the appropriate market for their need

### Customer purchase pattern Analysis

- It is the process of identifying the behaviors of consumers on their purchase pattern which includes
  - -Why consumers make purchase and when?
  - -What factors influence their purchase behavior
- This allows business to make selective promotion of good

#### Fraud Detection and Management

#### Applications

- widely used in health care, retail, credit card services, banking, insurance company, 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
- medical insurance: detect professional patients and ring of doctors and ring of references

### Fraud Detection and Management

#### • Detecting inappropriate medical treatment

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

#### **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

-WWW

- Data mining can be performed on various types of data stores and Databases
- Data mining functionalities are used to specify the kind of patterns to be found in data mining task
- Data mining task can be broadly classified into two as
  - Descriptive
  - Predictive

- Descriptive data mining task characterize the general properties of the data in a database.
  - For example one can say
    - Ethiopia's weather is selected to leave in for many birds
    - The past 10 years rainfall of Ethiopia is appropriate for the agriculturalist in southern Shewa
    - All mobile callers make few calls to wired lines than mobile receipents

- Predictive data mining task perform inference on the current data in order to make prediction to the future reference
- For example one can say
  - A person loves to leave in Ethiopia if he/she was in ASIA for the last two years
  - It will rain in Addis with in two days if there is a wind from Mediterranean see in west east direction and average current temperature at Addis is bellow 20°c

- The kind of pattern to be mined form a given data is not known for the user (hence it is hypothesis generation not hypothesis proving)
- Techniques should be implemented to extract various pattern from the available data so that user can choose what they need to use.
- There are different kinds of data mining functionalities that can be used to extract various types of pattern from data

- This are
  - Concept /class description: Characterization and discrimination
  - -Association Analysis
  - Classification and prediction
  - Clustering analysis
  - Outlier analysis
  - Evolution analysis

# Are All the "Discovered" Patterns Interesting?

- A data mining system/query may generate thousands of patterns, not all of them are interesting.
- Questions
  - 1. What makes a pattern interesting?
  - 2. Can a data mining system generate all of the interesting patterns?
  - 3. Can a data mining system generate only interesting patterns?

#### **Question 1**

- 1. What makes a pattern interesting?
- A pattern is interesting if it is <u>easily understood</u> by humans, <u>valid on</u>
   <u>new or test data</u> with some degree of certainty, <u>potentially useful</u>, <u>novel</u>,
   <u>or validates some hypothesis</u> that a user seeks to confirm
- An interesting pattern represents knowledge
- Measure of Interestingness measures
  - Two types (Objective vs. subjective)
    - <u>Objective:</u> based on statistics and structures of patterns, e.g., support, confidence, FP, FN, TN, TP, Recall, Precision, etc.
    - <u>Subjective</u>: based on user's belief in the data, e.g., unexpectedness (contradicting a user's belief), novelty, actionability, etc.

#### **Question 2**

- 2. Can a data mining system generate all of the interesting patterns?
- Referred as Completeness of the data mining algorithm
- No single data mining system is complete but users can set a
  constraint on the type of pattern they are looking for in which the data
  mining function generate all the pattern with the specified constraints
- Association algorithms don't find classification pattern and others for example

#### **Question 3**

- 3. Can a data mining system generate only interesting patterns?
- This is an Optimization problem in data mining system
- it remain an challenging issue
- Usually data mining system generate pattern from the data set which may or may not relevant at the point
- So first generate all the patterns and then filter out the uninteresting ones.

# **Data Mining: Classification Schemes**

- 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, object-relational, 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

#### • <u>Techniques utilized</u>

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

#### • Applications adapted

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