



[Introduction to Data Mining]

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Data Mining for Big Data

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Course Objectives

- To know the range of problems that can be solved with data mining.
- To understand what the data is.
- To learn the variation of data mining approaches.



Agenda

- Data Mining and Related Subjects
- Data
- Data Mining Approaches

Data Mining

- **Data Mining:** the **process** of discovering **hidden** and **actionable** patterns from data
- It utilizes methods at the intersection of artificial intelligence, machine learning, statistics, and database systems
- Extracting/“mining” knowledge from large-scale data (big data)
- Data-driven discovery and modeling of hidden patterns in big data
- Extracting information/knowledge from data that is
 - implicit,
 - previously unknown,
 - unexpected, and
 - potentially useful

Data Mining vs Database

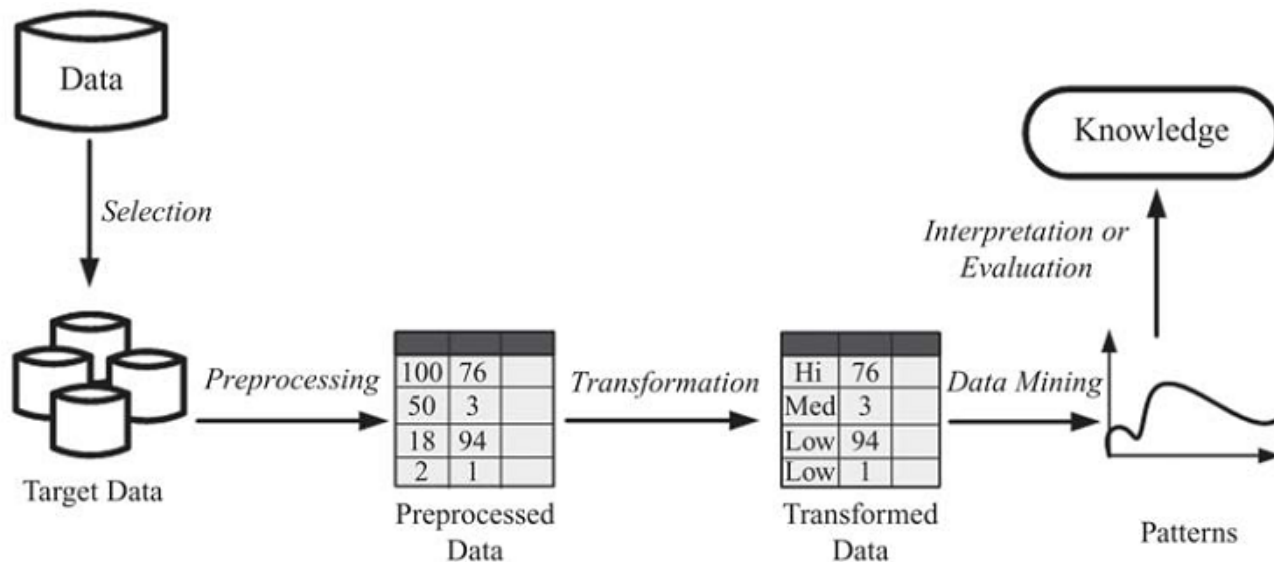
- **Data mining** is the *process* of extracting hidden and actionable patterns from data
- **Database systems** store and manage data
 - Queries return part of stored data
 - Queries do not extract hidden patterns
- Examples of querying databases
 - Find all employees with income more than \$250K
 - Find top spending customers in last month
 - Find all students from *engineering college* with GPA more than average

Examples of Data Mining Applications

- **Fraud/Spam Detections:** Identifying fraudulent transactions of a credit card or spam emails
 - You are given a user's purchase history and a new transaction, identify whether the transaction is fraud or not;
 - Determine whether a given email is spam or not
- **Frequent Patterns:** Extracting purchase patterns from existing records
 - beer \Rightarrow dippers (80%)
- **Forecasting:** Forecasting future sales and needs according to some given samples
- **Finding Like-Minded Individuals:** Extracting groups of like-minded people in a given network

Knowledge Discovery

- The process of extracting **useful patterns** from raw data is known as **Knowledge** discovery in databases (KDD)



Data

- In the KDD process, data is represented in a **tabular** format.
- Consider the example of predicting whether an individual who visits an online book seller is going to buy a specific book

| Attributes | | | | Class |
|------------|-------------|----------------|------------|----------|
| Name | Money Spent | Bought Similar | Visits | Will Buy |
| John | High | Yes | Frequently | ? |
| Mary | High | Yes | Rarely | Yes |

John is an example of an **instance**.

A **dataset** consists of one or more instances.

Features

- A dataset is represented using a set of **features (or attributes)**.
- an instance is represented using values assigned to these features.

Features

Class Attributes

| Attributes | | | | Class |
|------------|-------------|----------------|------------|----------|
| Name | Money Spent | Bought Similar | Visits | Will Buy |
| John | High | Yes | Frequently | ? |
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Feature Values

Type of Feature Values (levels of measurement)

- Nominal (categorical)
 - For instance, a customer's **name** is a nominal feature
- Ordinal
 - the feature values have an intrinsic order to them
 - In our example, **Money Spent** is an ordinal feature because a High value for Money Spent is more than a Low one
- Interval
 - differences are meaningful whereas ratios are meaningless
 - Example: **time** (e.g. 6:16, 6:45, etc)
- Ratio
 - add the additional properties of multiplication and division

Sample Data – Twitter User

| <i>Activity</i> | <i>Date Joined</i> | <i>Number of Followers</i> | <i>Verified Account?</i> | <i>Has Profile Picture?</i> |
|-----------------|--------------------|----------------------------|--------------------------|-----------------------------|
| High | 2015 | 50 | FALSE | no |
| High | 2013 | 300 | TRUE | no |
| Average | 2011 | 860000 | FALSE | yes |
| Low | 2012 | 96 | FALSE | yes |
| High | 2008 | 8,000 | FALSE | yes |
| Average | 2009 | 5 | TRUE | no |
| Very High | 2010 | 650,000 | TRUE | yes |
| Low | 2010 | 95 | FALSE | no |
| Average | 2011 | 70 | FALSE | yes |
| Very High | 2013 | 80,000 | FALSE | yes |
| Low | 2014 | 70 | TRUE | yes |
| Average | 2013 | 900 | TRUE | yes |
| High | 2011 | 7500 | FALSE | yes |
| Low | 2010 | 910 | TRUE | no |

If the data is not Tabular?

- In social media, individuals generate many types of non-tabular data, such as **text, voice, or video**.
- These types of data are first converted to tabular data and then processed using data mining algorithms.
 - voice can be converted to feature values using approximation techniques such as the **fast Fourier transform (FFT)**.
 - To convert **text** into the tabular format, we can use a process denoted as **vectorization**.

Types of Algorithms

- Supervised Learning
 - We have labeled dataset, that is, instances in this set are tuples in the format (\mathbf{x}, \mathbf{y}) , where \mathbf{x} is a feature vector and \mathbf{y} is the class attribute.
 - We learn a mapping $\mathbf{f}(\cdot)$, such that $\mathbf{f}(\mathbf{x}) = \mathbf{y}$
 - Example: *Classification* and *Regression*
- Unsupervised Learning
 - the dataset has **NO** class attribute, and our task is to find similar instances in the dataset and group them.
 - Example: Clustering



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Thank You