# 1.2 Data Preparation

Where the majority of the time is spent on any data mining project

importing, manipulating, cleaning, transforming, augmenting

### **Data**

- Collection of data objects (cases) and their attributes (features)
  - Attribute: property/characteristic of an object
  - Object: collection of attributes
- Can be structured (data table) or non-structured (text)
- Can have **non-dependency** or **dependency** (time, space)

## **Types of Data Sets**

- Nondependency-oriented data: no dependencies between cases
- Dependency-oriented data: implicit/explicit relationships between cases

## **Types of Attributes**

## **Categorical/Qualitative Attributes**

- Nominal: no relationship between values
- Ordinal: order between the values, but no mathematical operation can be performed on them

## **Numeric/Quantitative Attributes**

- Discrete: finite/countably infinite set of values for which differences are meaningful
- **Continuous**: infinite set of values that represent the absolute numbers

## **Important Characteristics**

- Dimensionality
- Sparsity
- Resolution
- Size

## **Data Preparation**

Data analysis tasks use source data sets stored in tabular format

## **Data Wrangling**

Transform and map data from one "raw" data form into another format appropriate for analytics

· Goal: attain quality and useful data

### **Steps**

- 1. Discovering
- 2. Structuring
- 3. Cleaning
- 4. Enriching
- 5. Validating
- 6. Publishing

## **Data Quality**

 Raw → values may be missing, inconsistent across different data sources, erroneous → poor data quality

## **Data Quality Problems**

- · Noise and outliers
- Missing values
- Duplicate data
- Inconsistent/incorrect data

#### Noise

- Irrelevant or useless information
- Possible causes: incorrect/distorted measurements, proper variability of the domain

#### **Outliers**

- Data objects with characteristics that are considerably different from most of the other objects in the data sets
- Cases: outlier = noise or outliers = goal

## **Missing Values**

- Missing Completely at Random (MCAR): missing value is independent of observed and unobserved data; nothing systematic about it
  - e.g.: a lab value because a lab sample was processes improperly

- Missing at Random (MAR): missing value is related to observed data, not to unobserved data; may be something systematic about it
  - e.g.: missing income value may depend on the age
- Missing Not a Random (MNAR): missing value is related to unobserved data of the variable itself; informative/non-ignorable missingness
  - e.g.: a person did not enter his/weight in a survey

#### Solutions

- Remove: critical if there are meany observations with missing values
- Ignore
- Make estimates: imputation
  - the most common value of the attribute; based on other attributes; more sophisticated methods
  - might introduce bias in data and affect the results

## **Duplicates**

• Major issues when merging data from heterogeneous source

## **Inconsistent/Incorrect Data**

- The hardest type of data quality issues to detect
- Depends on expert domain knowledge

## < Go back