DATA PROCESSING

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Data Preprocessing

- Why preprocess the data?
- Descriptive data summarization
- Data cleaning
- Data integration and transformation
- Data reduction
- Discretization and concept hierarchy generation
- Summary

Why Data **Preprocessing**?

- Data in the real world is dirty
 - incomplete: lacking attribute values, lacking certain attributes of interest, or containing only aggregate data
 - e.g., occupation=""
 - noisy: containing errors or outliers
 - o e.g., Salary="-10"
 - inconsistent: containing discrepancies in codes or names
 - e.g., Age="42" Birthday="03/07/1997"
 - e.g., Was rating "1,2,3", now rating "A, B, C"
 - o e.g., discrepancy between duplicate records

Why Is Data Dirty?

- · Incomplete data may come from
 - "Not applicable" data value when collected
 - Different considerations between the time when the data was collected and when it is analyzed.
 - Human/hardware/software problems
- Noisy data (incorrect values) may come from
 - Faulty data collection instruments
 - Human or computer error at data entry
 - Errors in data transmission
- Inconsistent data may come from
 - Different data sources
 - Functional dependency violation (e.g., modify some linked data)
- Duplicate records also need data cleaning

Why Is Data Preprocessing Important?

- No quality data, no quality mining results!
 - Quality decisions must be based on quality data
 - e.g., duplicate or missing data may cause incorrect or even misleading statistics.
 - Data warehouse needs consistent integration of quality data
- Data extraction, cleaning, and transformation comprises the majority of the work of building a data warehouse

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Multi-Dimensional Measure of Data Quality

- A well-accepted multidimensional view:
 - Accuracy
 - Completeness
 - Consistency
 - Timeliness
 - Believability
 - Value added
 - Interpretability
 - Accessibility

Major Tasks in Data Preprocessing

Data cleaning

 Fill in missing values, smooth noisy data, identify or remove outliers, and resolve inconsistencies

Data integration

Integration of multiple databases, data cubes, or files

Data transformation

Normalization and aggregation

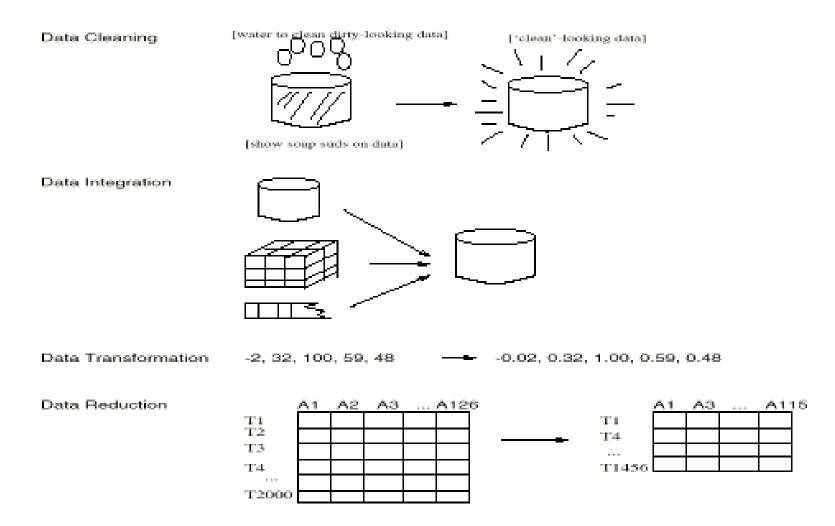
Data reduction

 Obtains reduced representation in volume but produces the same or similar analytical results

Data discretization

 Part of data reduction but with particular importance, especially for numerical data

Forms of Data Preprocessing



Mining Data Descriptive Characteristics

Motivation

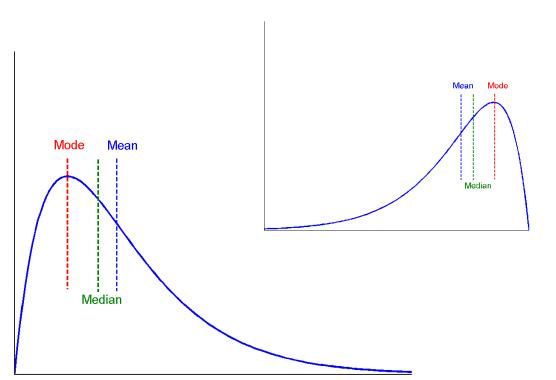
- To better understand the data: central tendency, variation and spread
- Data dispersion characteristics
 - median, max, min, quantiles, outliers, variance, etc.
- Numerical dimensions correspond to sorted intervals
 - Data dispersion: analyzed with multiple granularities of precision
 - Boxplot or quantile analysis on sorted intervals
- <u>Dispersion analysis on computed measures</u>
 - Folding measures into numerical dimensions
 - Boxplot or quantile analysis on the transformed cube

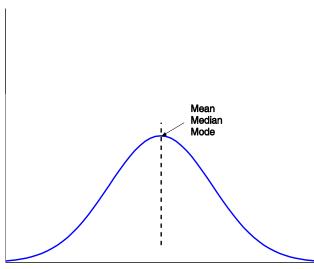
Measuring the Central Tendency

- Mean (algebraic measure) (sample vs. population):
 - Weighted arithmetic mean:
 - Trimmed mean: chopping extreme values
- Median: A holistic measure
 - Middle value if odd number of values, or average of the middle two values otherwise
- Estimated by interpolation (for *grouped data*): median = $L_1 + (\frac{n/2 - (\sum f)l}{f_{modian}})c$ **Mode**
 - Value that occurs most frequently in the data $mean-mode = 3 \times (mean-median)$
 - Unimodal, bimodal, tri-moda
 - Empirical formula:

Symmetric vs. Skewed Data

 Median, mean and mode of symmetric, positively and negatively skewed data





Granks Sylvin