

## CS 412 Intro. to Data Mining

Chapter 3. Data Preprocessing

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### Chapter 3: Data Preprocessing

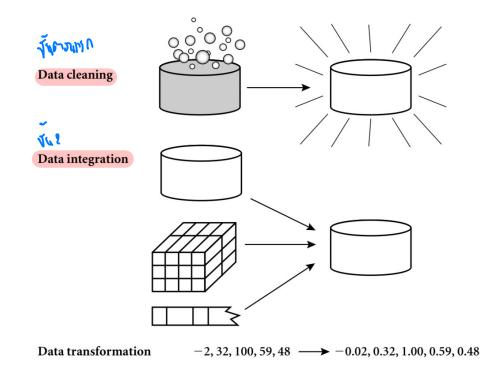
Data Preprocessing: An Overview

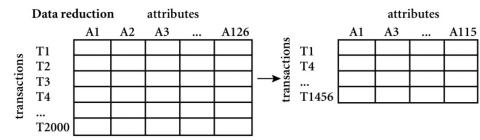


- 1 Data Cleaning with Clean won missing
- Data Integration เดิงให้เมา มามาน เพื่อไปประเทศ เลื่อในประเทศ

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- ▶ □ Data Reduction and Transformation
  - Dimensionality Reduction
  - Summary





#### What is Data Preprocessing? — Major Tasks

- Data cleaning Jamsmissing mão noisy lut outliers
  - Handle missing data, smooth noisy data, identify or remove outliers, and resolve inconsistencies | majors inconsistencies
- **Data integration** LYM MUTAMA WHO CSS
  - Integration of multiple databases, data cubes, or files
- Data reduction Managed 1222
  - Dimensionality reduction
  - Numerosity reduction
  - Data compression
- Data transformation and data discretization נאלמשוש נישאל וא ות הני לישיא שלאל
  - Normalization
  - Concept hierarchy generation

#### Why Preprocess the Data? — Data Quality Issues

- Measures for data quality: A multidimensional view
  - Accuracy: correct or wrong, accurate or not
  - Completeness: not recorded, unavailable, ...
  - Consistency: some modified but some not, dangling, ...

  - Believability: how trustable the data are correct?
  - Interpretability: how easily the data can be understood?

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- Data Preprocessing: An Overview
- Data Cleaning



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#### Data Cleaning

Data & Mary Ly Ly Session

- - Incomplete: lacking attribute values, lacking certain attributes of interest, or containing only aggregate data
    - 💶 e.g., Occupation = "" (missing data) ե กราก 🔌 ลุ
  - Noisy: containing noise, errors, or outliers
  - e.g., Salary = "-10" (an error)
  - Inconsistent: containing discrepancies in codes or names, e.g.,

  - Was rating "1, 2, 3", now rating "A, B, C"
  - discrepancy between duplicate records
  - Intentional (e.g., disguised missing data)
    - Jan. 1 as everyone's birthday?

#### Incomplete (Missing) Data

- Data is not always available กลุ่ง โประบูเพื่
  - E.g., many tuples have no recorded value for several attributes, such as customer income in sales data
- Missing data may be due to () Missing data may be due to
  - 12/10/201 Equipment malfunction

  - Data were not entered due to misunderstanding אינו על אינון אוניין אינון אינו
  - Certain data may not be considered important at the time of entry
  - Did not register history or changes of the data
- Missing data may need to be inferred

# How to Handle Missing Data?

- Ignore the tuple: usually done when class label is missing (when doing classification)—not effective when the % of missing values per attribute varies considerably
- ☐ Fill in the missing value manually: tedious + infeasible?
- □ Fill in it automatically with בְּעֹבוֹ הַא רֹבְּיִי וֹ Fill in it automatically with
  - □ a global constant : e.g., "unknown", a new class?!
  - the attribute mean
  - the attribute mean for all samples belonging to the same class: smarter
  - the most probable value: inference-based such as Bayesian formula or decision tree