Pre-processing: Data Cleansing

DSA 8102: Data mining, data storage and retrieval







Objectives

- Describe data pre-processing and the concept of data quality.
- Differentiate data exploration from EDA.
- Discuss ways to fix missing value problems.
- Determine what an outlier is and ways to detect and remove it.



Data Quality

Data Quality



Definition (standard-based perspective):

- "degree to which the inherent characteristics of data fulfills requirements" (how well does it represent the real-world construct to which it refers).
- "measure of the condition of data (i.e. accuracy, completeness, consistency, reliability, up-to-date etc.) for its application" (fit for its intended use).

Data governance:

- used to form agreed upon definitions and standards for data quality.
- to achieve this: data pre-processing among other techniques are used.



Data Exploration

Data Exploration



- Originally known as Exploratory Data Analysis (EDA), created by statistician John Tukey.
 - Currently EDA techniques (outlier detection, clustering) have evolved into independent areas of research.

Description:

 visualization and computation to better understand characteristics of data.

Objectives:

- aims at creating a mental understanding of the data (relationship between variables)
- quickly identify faulty points in data (errors, missing values etc.)

Data Exploration



Motivations:

- Help select right tool for preprocessing
- Make use of human's ability to recognize patterns (e.g., it takes a lot of work for an algorithm to recognize faces than humans)

Technique & Tools:

 Statistical graphics and data visualizations (i.e., Box plot, Histogram, Run chart, Scatter plot etc.)



Class Exercise

1. Differentiate data exploration from EDA.



Processing vs Pre-processing





Definition:

task of collecting data and translating it into usable information.

Stages:

- Data extraction: pulling from numerous sources.
- Data pre-processing: preparation stage.
- Data input: clean data is uploaded into its destination (i.e., data warehouse like Redshift).
- Processing: actual processing for interpretation/results.
- Data output: Interpretations simplified for non-data scientists (i.e., charts, graphs etc.).
- Data storage: for future use.



Data Pre-processing

Definition 1:

 technique used to convert raw data into a clean data set, preparing the data set for modelling (i.e., ML, DM).

Definition 2:

 a technique used to transform raw data into a useful and efficient format.

Definition 3:

 step in which data gets transformed/encoded, in order to bring it to a state where its <u>features can</u> <u>be easily interpreted</u> by an algorithm.

Pre-processing Significance



- Analogy: data is like <u>crude oil</u>
 - cannot be used directly from its source; so,
 - must be processed before being used for diff. purposes.

Real-world data issues:

- comes from different sources with different structures and data types.
- Often incomplete, inconsistent, noisy (i.e., errors, outliers) and/or lacking in certain behaviors.

Pre-processing:

- used to improve data quality (purify data).
- prepare data set for meaningful analysis.
- product is a <u>clean data/training set.</u>

Pre-processing steps



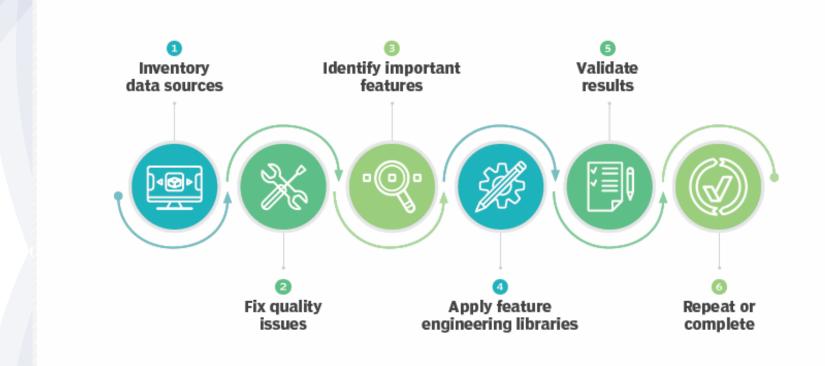


ILLUSTRATION: FIREOHEART/GETTY IMAGES; ALEXDNDZ/ADOBE STOCK

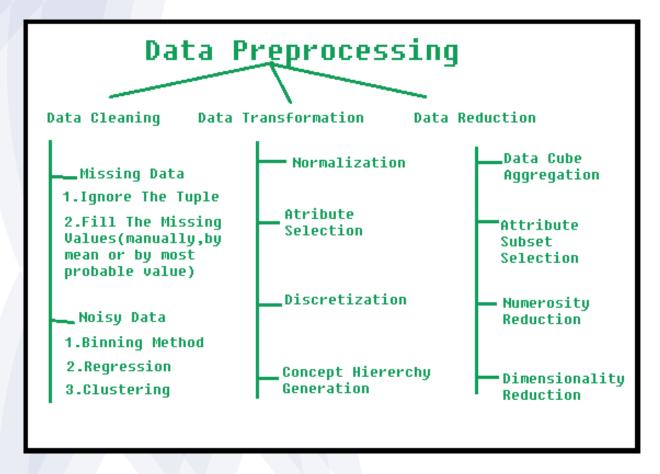
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Source: https://searchsqlserver.techtarget.com/definition/data-preprocessing

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Pre-processing Tasks



Source: https://www.geeksforgeeks.org/data-preprocessing-in-data-mining/

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Class Exercise

- 1. What could be the possible reasons for data to be inconsistent?
- 2. Why is data quality important? Or what could be the benefits of pre-processing data before using it?





- 1. Inventory (data sources)
- 2. Data Cleansing
- 3. Feature Selection
- 4. Feature Transformation
- 5. Validation (2, 3, 4)



Pre-processing (step 2): data cleansing



Data cleansing (or cleaning)

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Definition:

- process of detecting and correcting the corrupt parts of a data set.
- Benefits: (purify data)
 - correct errors;
 - detect and analyze outliers.
- Data cleansing tasks (examples):
 - cleaning missing values
 - reducing noise (i.e., class and attribute noise)



Fixing missing values

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Missing Values



Description:

- also known as data holes
- incomplete data point (s) in records
- most data mining and machine learning techniques do not support data with missing values

Possible Reasons:

- No information is provided;
- Incorrect data types for attributes during extraction;
- Integrating data from multiple sources
- List goes on ...

Dealing with Missing Values



- Ignore incomplete records
 - Not practical for data sets that contain large portions of missing values (poor rep. of initial pop.)
- Update with descriptive/sensible values
 - e.g., mean, mode, median etc.
 - Introduces inconsistencies
- Investigate an identify patterns of missing values. E.g.
 - Data is truly unavailable
 - Duplicate records (one updates another)
 - Simple human errors (like wrong input types)
 - List goes on ...



Detecting and handling outliers

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Noise



Description:

- Refers to erroneous values (i.e., outliers)
- Random error or abnormal variance in a measured variable
- Adversely influences model/data mining performance
- Requires detection and removal

Types:

- Class noise (categorical variables): erroneous class labels
- Attribute noise (numeric variables): corruptive values and outliers

Possible Reasons:

- Data entry errors (i.e., human mistakes)
- Faulty data collection (i.e., automated record collection)

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Handling Outliers



Binning

 Segmenting data into small bins (or buckets) and smoothing (or replacing) bin values by mean, mode etc.

Clustering

Group (numeric) data by similarity {detects outliers}

Classification

 Arrange (categorical) data according to shared classes {erroneous classes have very few members}

Regression

Fit data into regression functions (i.e., linear) {remove random variance}

y = mx + b: where y is predicted value, x is actual value



Class Exercise

- 1. Which collection errors may lead to:
 - a. missing values
 - b. noisy data

Research Perspectives



Pre-processing

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Thank you!

Any Questions?