

# **DMDW LAB**

## **ASSIGNMENT 3**

# TOPICS

- **Data Preprocessing:**
  - **Data Quality**
  - **Major Tasks in Data Preprocessing**
- **Data Cleaning**
- **Data Integration**
- **Data Reduction**
- **Data Transformation and Data Discretization**
- **Summary**

# Data Preprocessing

- **Data preprocessing is a data mining technique which is used to transform the raw data in a useful and efficient format.**
- **Data Preprocessing is that step in which the data gets transformed or Encoded to bring it to such a state that now the machine can easily parse it.**
- **Data preprocessing includes cleaning , instance selection, normalization, transformation, feature extraction and selection.**

# Why Preprocess the Data?

- **Accuracy:** correct or wrong, accurate or not
- **Completeness:** not recorded, unavailable, ...
- **Consistency:** some modified but some not, dangling,  
...
- **Timeliness:** timely update?
- **Believability:** how trustable the data are correct?
- **Interpretability:** how easily the data can be understood?

# Major Tasks in Data Preprocessing



# 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 reduction**
  - Dimensionality reduction
  - Numerosity reduction
  - Data compression
- **Data transformation and data discretization**
  - Normalization
  - Concept hierarchy generation

# What is Data Cleaning?

- **Data in the Real World Is Dirty:** Lots of potentially incorrect data, e.g., instrument faulty, human or computer error, transmission error

**Incomplete:** lacking attribute values, lacking certain attributes of interest, or containing only aggregated data

- e.g., *Occupation*=" " (missing data)

**Noisy:** containing noise, errors, or outliers

- e.g., *Salary*="-10" (an error)

**Inconsistent:** containing discrepancies in codes or names, e.g.,

- *Age*="42", *Birthday*="03/07/2010"
- 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?

# How to Handle Missing Data?

- Ignore the tuple: usually done when class label is missing which is not effective when the % of missing values per attribute varies considerably.
- Fill in the missing value manually
- Fill in it automatically with
  - A global constant : e.g., “unknown”, a new class
  - The attribute mean



# What is Noisy Data?

- **Noise:** Random error or variance in a measured variable
- **Incorrect attribute values** may be due to
  - Faulty data collection
  - Data entry problems
  - Data transmission problems
  - Technology limitation
  - Inconsistency in naming convention
- **Other data problems** which require data cleaning
  - Duplicate records
  - Incomplete data
  - Inconsistent data

# Assignment 3

- Upload the Toyota Dataset :

path=<https://raw.githubusercontent.com/archana1822/DMDW-Lab/main/Toyota.csv>

```
import pandas as pd  
data = pd.read_csv(path)
```

- Use the following command on Toyota dataset

1. type(data)
2. data.shape
3. data.info()
4. data.index
5. data.columns

# Assignment 3

6. `data.head()`
7. `data.tail()`
8. `data.head(5)`
9. `data[['Price','Age']].head(10)`
10. `data.isnull().sum()`
11. `data.dropna(inplace=True)`  
    `data.isnull().sum()`
12. `data.shape`
13. `data.head(10)`
14. `data['MetColor'].mean()`
15. `data['MetColor'].head()`

# Assignment 3

```
16. import numpy as np
data['MetColor'].replace(np.NaN,data['MetColor'].mean()).head()

17. data.head(10)
18. data['CC'].mean()
19. data['CC'].head()
20. data[['Age','KM']].head(20)
```