

**Name:** Prathamesh Khoje

**Roll No:** 281046

**Batch:** A2

## **Assignment 4**

### **Statement**

In this assignment, we aim to:

Apply appropriate ML algorithm on a dataset collected in a cosmetics shop showing details of

customers to predict customer response for special offer.

Create confusion matrix based on above data and find

a)Accuracy

b)Precision

c)Recall

d)F-1 score

### **Objective**

1. Understand how to compute summary statistics for a dataset using Python.
2. Learn how to visualize data distributions using histograms.
3. Develop skills in data preprocessing, transformation, and integration to prepare data for machine learning models.
4. Build a classification model using the cleaned and transformed dataset.
5. Evaluate classification model performance using a confusion matrix and other metrics.

### **Resources Used**

- **Software:** Visual Studio Code
- **Libraries:** Pandas, Matplotlib, Seaborn, Scikit-Learn

### **Introduction to Pandas and Data Analysis**

Pandas is an open-source Python library designed for data manipulation and analysis. It provides data structures like Series (1D) and DataFrame (2D), making it easier to analyze and process structured data efficiently.

## **Key Capabilities**

- Importing data from various formats (CSV, Excel, SQL).
- Data preprocessing (cleaning, transforming, handling missing values).
- Statistical and analytical operations (descriptive statistics, visualization).
- Data modeling (classification, regression, clustering).

## **Basic Functions Used**

1. `pd.read_csv()` – Reads data from a CSV file into a DataFrame.
2. `describe()` – Generates descriptive statistics for numerical columns.
3. `hist()` – Creates histograms to visualize data distributions.
4. `fillna()` – Handles missing values using strategies like mean, median, or mode imputation.
5. `LabelEncoder()` – Converts categorical variables into numerical representations for machine learning models.
6. `train_test_split()` – Splits data into training and testing sets.
7. `LogisticRegression()` – Builds a regression-based classification model.
8. `confusion_matrix()` – Computes the confusion matrix.
9. `accuracy_score()`, `precision_score()`, `recall_score()`, `f1_score()` – Evaluate model performance.

## **Methodology**

### **1. Data Collection and Exploration**

- **Collect Data:** The dataset includes customer details from a cosmetics shop.
- **Explore Data:** Load the dataset into a Pandas DataFrame and examine its structure, including feature types, missing values, and unique categories.

### **2. Data Preprocessing**

- Handle Missing Values: Use techniques like mean, median, or mode imputation, or remove records with excessive missing data.
- Data Cleaning: Remove duplicates, correct formatting inconsistencies, and handle outliers.

### 3. Summary Statistics Computation

- Compute the minimum, maximum, mean, range, standard deviation, variance, and percentiles for each numerical feature using `describe()` and additional statistical functions.

### 4. Feature Visualization using Histograms

- Use `hist()` and Seaborn's `sns.histplot()` to visualize feature distributions.

### 5. Data Transformation and Feature Engineering

- Feature Encoding: Convert categorical features into numeric using `LabelEncoder()` or `OneHotEncoder()`.
- Feature Selection: Identify and retain relevant features based on correlation analysis.

### 6. Data Integration

- Merge data from multiple sources if required, ensuring consistency across datasets.

### 7. Model Building (Classification)

- Split the dataset into training and testing sets using `train_test_split()`.
- Train a classification model using Logistic Regression.
- Evaluate the model's performance using a confusion matrix and compute accuracy, precision, recall, and F1-score.

### Evaluation Metrics

After training the model, the following metrics were computed from the confusion matrix:

- **Accuracy** =  $(TP + TN) / (TP + TN + FP + FN)$
- **Precision** =  $TP / (TP + FP)$
- **Recall** =  $TP / (TP + FN)$

- **F1-score** =  $2 \times (\text{Precision} \times \text{Recall}) / (\text{Precision} + \text{Recall})$

### **Advantages of Pandas and Machine Learning in Data Analysis**

1. Pandas simplifies data handling and provides powerful functions for data cleaning and transformation.
2. Visualization tools help in understanding feature distributions and data patterns.
3. Machine learning models allow for automated classification and prediction.

### **Disadvantages of Pandas and Data Processing**

1. Memory-intensive operations may slow down processing for large datasets.
2. Preprocessing complexity increases with unstructured data.

### **Conclusion**

This assignment provided insights into the use of Pandas for structured data analysis, including reading, preprocessing, and summarizing datasets. We explored feature distributions using histograms and implemented classification models to make predictions. Additionally, we evaluated model performance using a confusion matrix and key classification metrics.