# Assignment-1: K-Nearest Neighbors (KNN)

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#### Introduction

In this assignment, the **K-Nearest Neighbors (KNN) algorithm** is applied on the Mall Customers dataset to segment customers based on their demographics and spending behavior. The objective is to classify customers into meaningful groups such as low, medium, and high spenders, which can be useful for targeted marketing strategies.

#### **Dataset**

The dataset used in this assignment is the **Mall Customers Dataset** from Kaggle, containing 200 customer records with five features: CustomerID, Gender, Age, Annual Income (k\$), and Spending Score (1–100). A target variable, **Spending Segment (Low, Medium, High)**, was created by binning the spending scores.

#### Methodology

Preprocessing included encoding the Gender column and creating spending segments. The dataset was split into 70% training and 30% testing sets, and features were standardized for KNN. A **K-Nearest Neighbors model (k=5)** was trained, and performance was evaluated using a confusion matrix, classification report, and accuracy score. Visualization of Annual Income versus Spending Score was also performed to illustrate customer segments.

### CODE:

```
# Step 1: Upload dataset (choose 'mall customers.csv' from your computer)
from google.colab import files
uploaded = files.upload()
# Step 2: Load dataset
import pandas as pd
# If file has a space in name, handle with quotes
data = pd.read_csv("mall customers.csv")
print("Dataset Loaded Successfully!")
print("Shape:", data.shape)
print("\nFirst 5 rows:")
print(data.head())
# Step 3: Preprocessing
from sklearn.preprocessing import LabelEncoder
# Encode Gender (Male=1, Female=0)
le = LabelEncoder()
data['Gender'] = le.fit_transform(data['Gender'])
# Features
X = data[['Gender', 'Age', 'Annual Income (k$)', 'Spending Score (1-100)']]
# Target: Segment based on Spending Score
data['Segment'] = pd.cut(data['Spending Score (1-100)'],
             bins=[0,40,70,100],
             labels=['Low','Medium','High'])
```

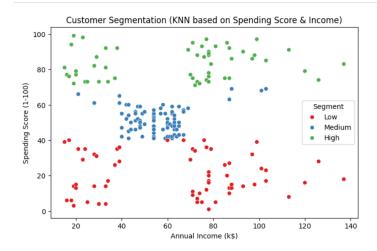
```
y = data['Segment']
print("\n Segment Distribution:")
print(y.value_counts())
# Step 4: Train-Test Split
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
# Step 5: Feature Scaling
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
# Step 6: Train KNN Model
from sklearn.neighbors import KNeighborsClassifier
knn = KNeighborsClassifier(n_neighbors=5)
knn.fit(X_train, y_train)
# Step 7: Predictions
y_pred = knn.predict(X_test)
# Step 8: Evaluation
from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
print("\n Confusion Matrix:")
print(confusion_matrix(y_test, y_pred))
print("\n Classification Report:")
print(classification_report(y_test, y_pred))
```

## **OUTPUT:**

```
Saving mall customers.csv to mall customers (1).csv
Dataset Loaded Successfully!
Shape: (200, 5)
  CustomerID Gender Age Annual Income (k$) Spending Score (1-100)

1 Male 19 15 39
                Male
              Female
                                                                  6
77
                       20
              Female
                       23
                                           16
           5 Female
                       31
                                                                   40
Segment Distribution:
Segment
Medium
High
         54
Name: count, dtype: int64
Confusion Matrix:
[[13 0 0]
[ 0 17 4]
[ 1 0 25]]
  Classification Report:
                  precision
                                 recall f1-score
                                                        support
           High
                        0.93
                                    1.00
                                                0.96
                                                              13
                        1.00
                                    0.81
                                                0.89
                                                              21
            Low
        Medium
                        0.86
                                    0.96
                                                0.91
                                                              26
      accuracy
                                                0.92
                                                              60
     macro avg
                        0.93
                                    0.92
                                                              60
                                                0.92
 weighted avg
                        0.92
                                    0.92
                                                0.92
                                                              60
```

Accuracy Score: 0.916666666666666



GITHUB LINK: https://github.com/deepanshi211/2201CS85\_APR\_Assignment1