

EXPERIMENT : 9(a)

A python program using a K-Means Algorithm in a model

AIM:

To implement a python program using a K-Means Algorithm in a model.

ALGORITHM:

1. Import Necessary Libraries:

Import required libraries like numpy, matplotlib.pyplot, and sklearn.cluster.

2. Load and Preprocess Data:

Load the dataset.

Preprocess the data if needed (e.g., scaling).

3. Initialize Cluster Centers:

Choose the number of clusters (K).

Initialize K cluster centers randomly.

4. Assign Data Points to Clusters:

For each data point, calculate the distance to each cluster center.

Assign the data point to the cluster with the nearest center.

5. Update Cluster Centers:

Calculate the mean of the data points in each cluster.

Update the cluster centers to the calculated means.

6. Repeat Steps 4 and 5:

Repeat the assignment of data points to clusters and updating of cluster centers until

convergence (i.e., when the cluster assignments do not change much between iterations).

7. Plot the Clusters:

Plot the data points and the cluster centers to visualize the clustering result.

CODE 1:

```
import numpy as np
```

```
import pandas as pd
```

```
import matplotlib.pyplot as plt
```

```
from sklearn.model_selection import train_test_split
```

```
from sklearn.preprocessing import StandardScaler
```

```
from sklearn.neighbors import KNeighborsClassifier
```

```
from sklearn.metrics import confusion_matrix,  
classification_report
```

```
dataset = pd.read_csv("Mall_Customers.csv")
```

```
print("First 5 Rows of Dataset:")
```

```
print(dataset.head())
```

```
print("\nShape of Dataset:", dataset.shape)
```

```
print("\nDescriptive Statistics:")
```

```
print(dataset.describe())
```

OUTPUT 2:

First 5 Rows of Dataset:

CustomerID Gender Age Annual Income (k\$) Spending Score (1-100)

0	1	Male	19	15	39
1	2	Male	21	15	81
2	3	Female	20	16	6
3	4	Female	23	16	77
4	5	Female	31	17	40

Shape of Dataset: (200, 5)

Descriptive Statistics:

CustomerID Age Annual Income (k\$) Spending Score (1-100)

count	200.00000	200.000000	200.000000	200.000000
mean	100.50000	38.850000	60.560000	50.200000
std	57.87918	13.969007	26.264721	25.823522

min	1.00000	18.000000	15.000000	1.000000
max	200.00000	70.000000	137.000000	99.000000

CODE 3:

```
dataset['Gender'] = dataset['Gender'].map({'Male': 0, 'Female': 1})
```

```
X = dataset[['Gender', 'Age', 'Annual Income (k$)']].values
```

```
y = dataset['Spending Score (1-100)'].values
```

```
y = pd.cut(y, bins=[0, 40, 70, 100], labels=[0, 1, 2]) #
```

Classification labels

```
X_train, X_test, y_train, y_test = train_test_split(X, y,  
test_size=0.2, random_state=42)
```

```
scaler = StandardScaler()
```

```
X_train = scaler.fit_transform(X_train)
```

```
X_test = scaler.transform(X_test)
```

```
print("X_train shape:", X_train.shape)
```

```
print("X_test shape:", X_test.shape)
```

OUTPUT 3:

```
X_train shape: (160, 3)
```

X_test shape: (40, 3)

CODE 4:

k = 5

knn = KNeighborsClassifier(n_neighbors=k)

knn.fit(X_train, y_train)

y_pred = knn.predict(X_test)

print("Confusion Matrix:\n", confusion_matrix(y_test, y_pred))

print("\nClassification Report:\n", classification_report(y_test, y_pred))

OUTPUT 4:

Confusion Matrix:

[[10 2 0]

[3 11 1]

[0 2 11]]

Classification Report:

	precision	recall	f1-score	support
0	0.77	0.83	0.80	12
1	0.73	0.73	0.73	15
2	0.92	0.85	0.88	13
accuracy			0.80	40

CODE 5:

```
# Test different K values
```

```
accuracy = []
```

```
k_values = range(1, 21)
```

```
for k in k_values:
```

```
    knn = KNeighborsClassifier(n_neighbors=k)
```

```
    knn.fit(X_train, y_train)
```

```
    accuracy.append(knn.score(X_test, y_test))
```

```
# Plot accuracy vs K
```

```
plt.figure(figsize=(8, 5))
```

```
plt.plot(k_values, accuracy, color='blue', marker='o',  
markerfacecolor='red')
```

```
plt.title('Accuracy vs Number of Neighbors (K)')
```

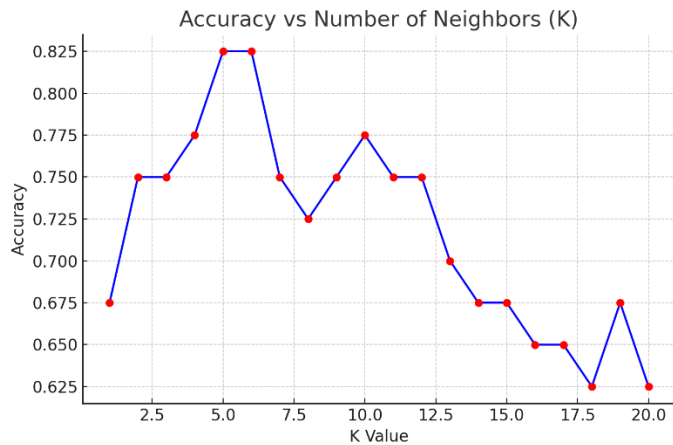
```
plt.xlabel('K Value')
```

```
plt.ylabel('Accuracy')
```

```
plt.grid(True)
```

```
plt.show()
```

OUTPUT 5:



CODE 6:

Use only 2 features for plotting

X_plot = dataset[['Age', 'Annual Income (k\$)']].values

y_plot = pd.cut(dataset['Spending Score (1-100)'], bins=[0, 40, 70, 100], labels=[0, 1, 2])

scaler2 = StandardScaler()

X_plot = scaler2.fit_transform(X_plot)

knn2 = KNeighborsClassifier(n_neighbors=5)

knn2.fit(X_plot, y_plot)

Plot points by category

plt.figure(figsize=(7, 6))

**plt.scatter(X_plot[y_plot == 0][:, 0], X_plot[y_plot == 0][:, 1],
c='red', label='Low Spenders')**

```
plt.scatter(X_plot[y_plot == 1][:, 0], X_plot[y_plot == 1][:, 1],  
c='blue', label='Medium Spenders')
```

```
plt.scatter(X_plot[y_plot == 2][:, 0], X_plot[y_plot == 2][:, 1],  
c='green', label='High Spenders')
```

```
plt.title('KNN Visualization of Spending Categories')
```

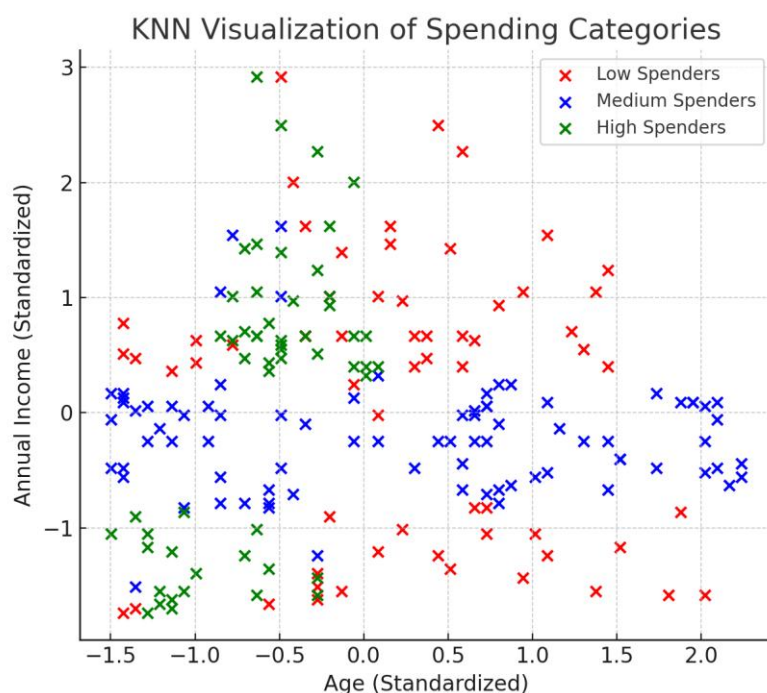
```
plt.xlabel('Age (Standardized)')
```

```
plt.ylabel('Annual Income (Standardized)')
```

```
plt.legend()
```

```
plt.show()
```

OUTPUT 6:



CODE 6:

```
print("Final Model Accuracy: {:.2f}%".format(knn.score(X_test,  
y_test) * 100))
```


OUTPUT 6:

Final Model Accuracy: 80.00%

RESULT:

Thus a python program using a K-Means Algorithm in a model is written and the output is verified.