

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

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
dataset = pd.read_csv('/content/Mall_Customers.csv')
X = dataset.iloc[:,[3,4]].values
print(dataset)

from sklearn.cluster import KMeans
wcss = []
for i in range(1,11):
    kmeans = KMeans(n_clusters = i, init = 'k-means++', max_iter = 300,
n_init = 10, random_state = 0)
    kmeans.fit(X)
    wcss.append(kmeans.inertia_)
# Plot the graph to visualize the Elbow Method to find the optimal
number of cluster
plt.plot(range(1,11),wcss)
plt.title('The Elbow Method')
plt.xlabel('Number of clusters')
plt.ylabel('WCSS')
plt.show()

kmeans=KMeans(n_clusters= 5, init = 'k-means++', max_iter = 300,
n_init = 10, random_state = 0)
y_kmeans = kmeans.fit_predict(X)
y_kmeans
```

```
type(y_kmeans)
```

```
y_kmeans
```

```
plt.scatter(X[y_kmeans == 0, 0], X[y_kmeans == 0, 1], s = 100, c = 'red',  
label = 'Cluster 1')
```

```
plt.scatter(X[y_kmeans == 1, 0], X[y_kmeans == 1, 1], s = 100, c =  
'blue', label = 'Cluster 2')
```

```
plt.scatter(X[y_kmeans == 2, 0], X[y_kmeans == 2, 1], s = 100, c =  
'green', label = 'Cluster  
3')
```

```
plt.scatter(X[y_kmeans == 3, 0], X[y_kmeans == 3, 1], s = 100, c =  
'cyan', label = 'Cluster 4')
```

```
plt.scatter(X[y_kmeans == 4, 0], X[y_kmeans == 4, 1], s = 100, c =  
'magenta', label =  
'Cluster 5')
```

```
plt.scatter(kmeans.cluster_centers_[:, 0], kmeans.cluster_centers_[:, 1], s  
= 300, c = 'yellow',  
label = 'Centroids')
```

```
plt.title('Clusters of customers')
```

```
plt.xlabel('Annual Income (k$)')
```

```
plt.ylabel('Spending Score (1-100)')
```

```
plt.legend()
```

```
plt.show()
```

```
plt.scatter(X[y_kmeans == 0, 0], X[y_kmeans == 0, 1], s = 100, c = 'red',  
label = 'Cluster 1')
```

```
plt.scatter(X[y_kmeans == 1, 0], X[y_kmeans == 1, 1], s = 100, c =  
'blue', label = 'Cluster 2')
```

```
plt.scatter(X[y_kmeans == 2, 0], X[y_kmeans == 2, 1], s = 100, c =  
'green', label = 'Cluster 3')
```

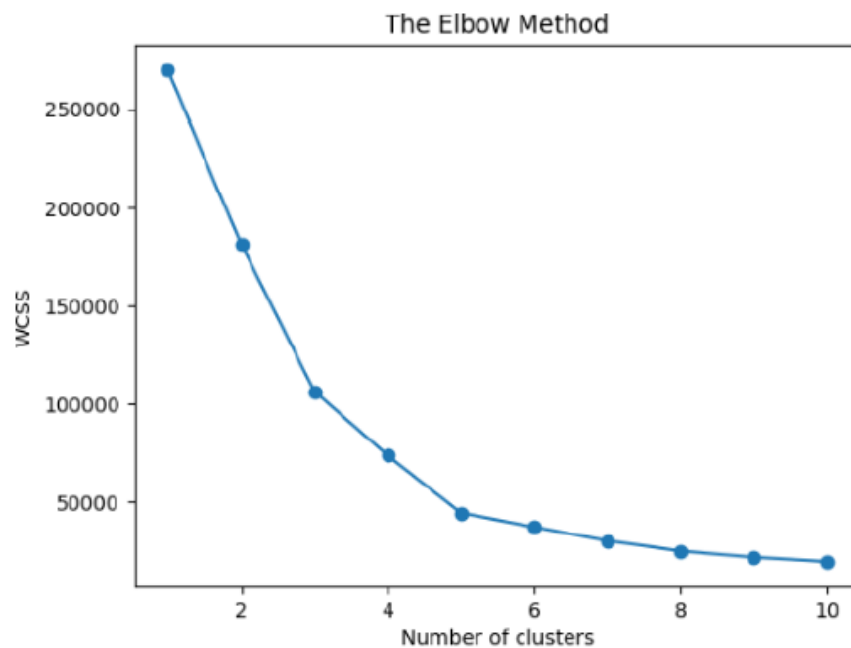
```
plt.scatter(X[y_kmeans == 3, 0], X[y_kmeans == 3, 1], s = 100, c =  
'cyan', label = 'Cluster 4')
```

```
plt.scatter(X[y_kmeans == 4, 0], X[y_kmeans == 4, 1], s = 100, c =  
'magenta', label =  
'Cluster 5')
```

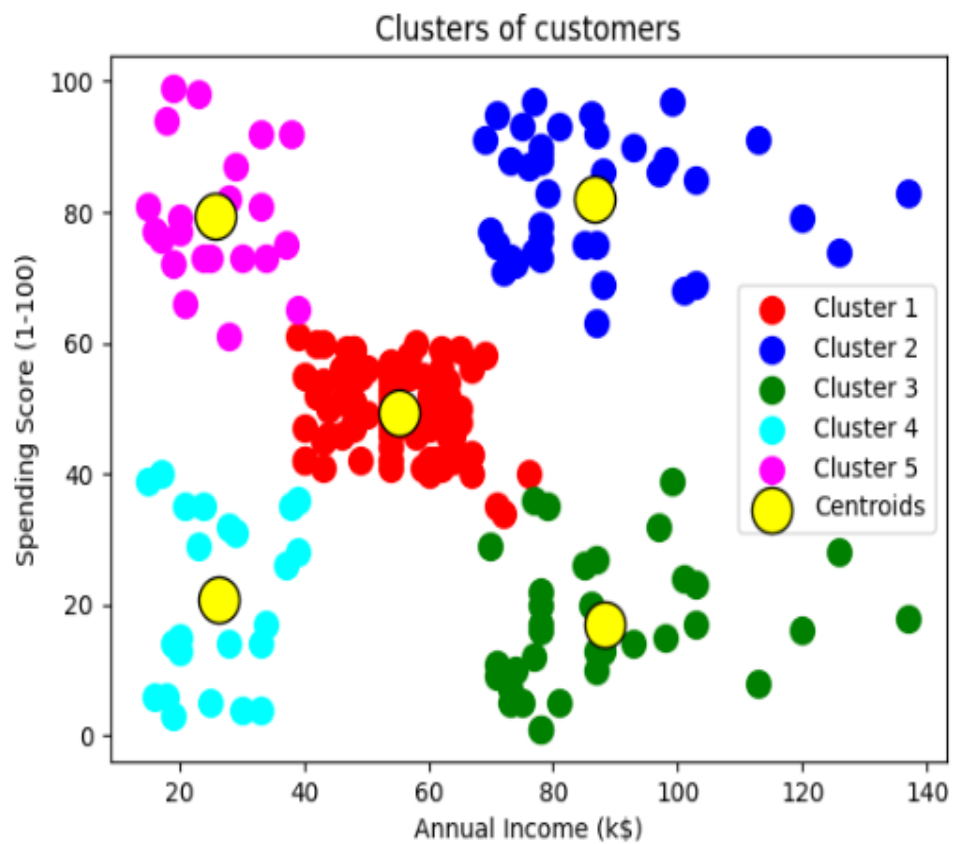
```
plt.scatter(kmeans.cluster_centers_[:, 0], kmeans.cluster_centers_[:, 1], s
= 300, c = 'yellow',
label = 'Centroids')
plt.title('Clusters of customers')
plt.xlabel('Annual Income (k$)')
plt.ylabel('Spending Score (1-100)')
plt.legend()
plt.show()
```

OUTPUT

	CustomerID	Genre	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

[illegible]

Type of labels: <class 'numpy.ndarray'>



RESULT:

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