

### Program No:11

Aim:Program to implement k-means clustering techniques using any standard dataset available in the public domain.

Program:

```
import numpy as np
import matplotlib.pyplot as mtp
import pandas as pd
dataset=pd.read_csv('Mall_Customers.csv')
x=dataset.iloc[:,[3,4]].values
print(x)
from sklearn.cluster import KMeans
wcss_list = []
for i in range(1, 11):
    kmeans = KMeans(n_clusters=i,
init='k-means++')
    kmeans.fit(x)
    wcss_list.append(kmeans.inertia_)
mtp.plot(range(1,11), wcss_list)
mtp.title('The elbow method Graph')
mtp.xlabel('Number of clusters (k)')
mtp.ylabel('wcss_list')
mtp.show()
kmeans = KMeans(n_clusters=5,init='k-means++',random_state=42)
y_predict=kmeans.fit_predict(x)
print(y_predict)
mtp.scatter(x[y_predict == 0,0], x[y_predict ==0,1], s=100, c='blue', label='Cluster0')
mtp.scatter(x[y_predict == 1,0], x[y_predict ==1,1], s=100, c='green', label= 'Cluster1')
mtp.scatter(x[y_predict == 2,0], x[y_predict ==2,1], s=100, c='red', label= 'Cluster2')
mtp.scatter(x[y_predict == 3,0], x[y_predict ==3,1], s=100, c='cyan', label= 'Cluster3')
mtp.scatter(x[y_predict == 4,0], x[y_predict ==4,1], s=100, c='magenta', label= 'Cluster4')
mtp.scatter(kmeans.cluster_centers_[:,0],kmeans
```

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.cluster_centers_[ :,1], s = 300,)
mtp.title('clusters of customers')
mtp.xlabel('annual incmome (k$')
mtp.ylabel('spending score (1-100)')
mtp.legend()
mtp.show()
```

Output:

```
C:\Users\mca\PycharmProjects\pythonProject\venv\Scripts\python.exe C:/Users/mca/PycharmProjects/pythonProject/kmeansclustering.py
[[ 15  39]
 [ 15  81]
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[ 62 56]
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[ 71 35]
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[137 83]]
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```

Figure 1

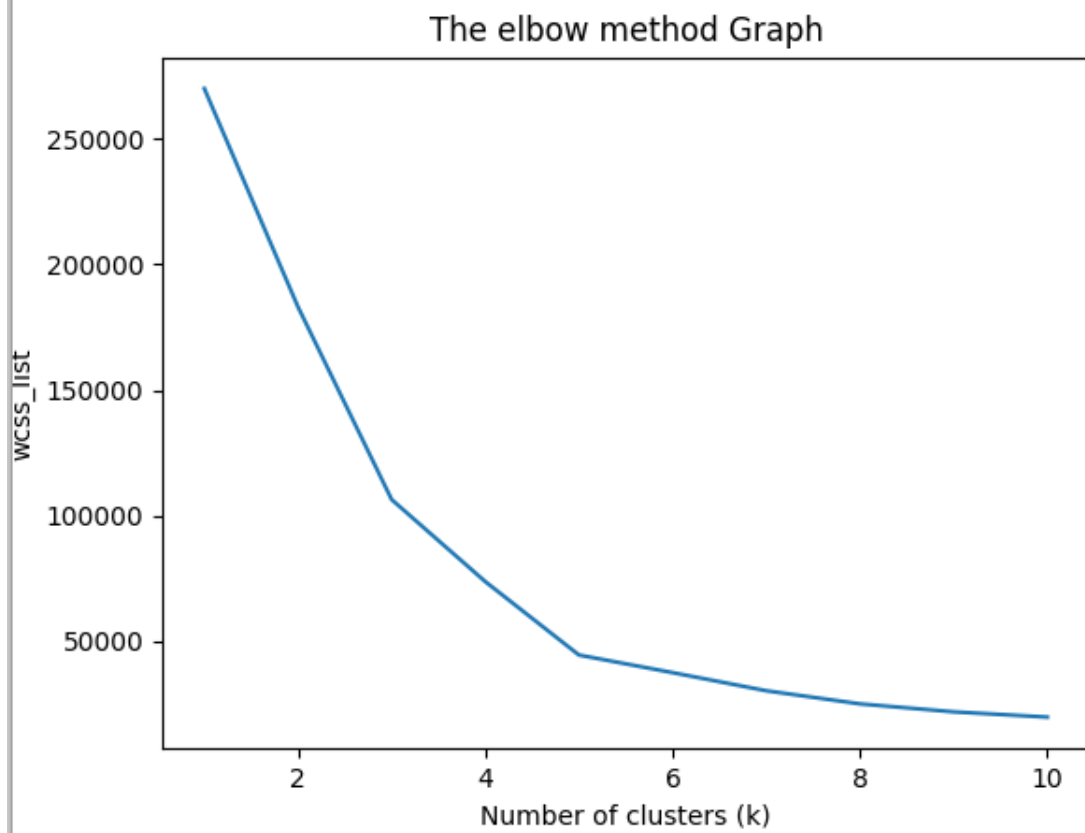


Figure 1

