## **EXPERIMENT-11**

AIM: Implement K Means Clustering in python on any dataset.

## **CODE AND OUTPUT:**

```
import numpy as nm
import matplotlib.pyplot as mtp
import pandas as pd
                                                                            # Importing the dataset
                                                                          dataset = pd.read_csv('Mall_Customers.csv')
                                                                          x = dataset.iloc[:, [3, 4]].values
                                                                                                                                           15, 39], 15, 81], 16, 6], 16, 77], 17, 40], 17, 76], 18, 6], 18, 94], 19, 72], 19, 72], 20, 15], 20, 77], 20, 13], 20, 79], 21, 35], 21, 35], 21, 36], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 66], 21, 6
                                                                                                                                             23,
                                                                                                                                                                               29],
                                                                            #finding optimal number of clusters using the elbow method
from sklearn.cluster import KNeans
wcss_list=[] #Initializing the list for the values of WCSS
                                                                      #Using for loop for iterations from 1 to 10.

for i in range(1, 11):
    kmeans = KNeans(n_clusters=i, init='k-means++', random_state= 42)
    kmeans.fit(x)
    wcss_list.append(kmeans.inertia_)
    mtp.plot(range(1, 11), wcss_list)
    mtp.title('The Elobw Method Graph')
    mtp.xlabel('Number of clusters(k)')
    mtp.ylabel('Number of clusters(k)')
    mtp.show()
                                                                                                                                                                                                                                      The Elobw Method Graph
                                                                                                      50000
   #training the K-means model on a dataset
kmeans = KMeans(n_clusters=5, init='k-means++', random_state= 42)
y_predict= kmeans.fit_predict(x)
#visulaizing the clusters
mtp.scatter(x[y_predict == 0, 0], x[y_predict == 0, 1], s = 100, c = 'blue', label = 'Cluster 1') #for first cluster
mtp.scatter(x[y_predict == 1, 0], x[y_predict == 1, 1], s = 100, c = 'green', label = 'cluster 2') #for second cluster
mtp.scatter(x[y_predict == 2, 0], x[y_predict == 2, 1], s = 100, c = 'red', label = 'cluster 3') #for third cluster
mtp.scatter(x[y_predict == 3, 0], x[y_predict == 3, 1], s = 100, c = 'cyan', label = 'cluster 4') #for fourth cluster
mtp.scatter(x[y_predict == 4, 0], x[y_predict == 4, 1], s = 100, c = 'magenta', label = 'cluster 5') #for fifth cluster
mtp.scatter(x[weans.cluster_centers[:, 0], kmeans.cluster_centers[:, 1], s = 300, c = 'yellow', label = 'Centroid')
mtp.xlabel('Annual Income (k$)')
mtp.xlabel('Apnual Income (k$)')
mtp.ylabel('Spending Score (1-100)')
mtp.slow()
```

