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**19BCE2074**

## **Digital Assignment 5**

### **Question 1**

#### **Problem Statement :**

K-means clustering

#### **Procedure :**

Dataset:

Link : <https://archive.ics.uci.edu/ml/machine-learning-databases/>

- 1) Import Dataset
- 2) Import necessary libraries - sklearn, numpy, pandas, etc.
- 3) Initialize k points, called means, randomly
- 4) Categorize each item to its closest mean, and we update the mean's coordinates
- 5) Repeat the process for a given number of iterations and at the end

#### **Code :**

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd

dataset = pd.read_csv('iris.csv')
X = dataset.iloc[:, [0, 1]].values

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_)
```

```

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)

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')
plt.xlabel('sepal length')
plt.ylabel('sepal width')
plt.legend()
plt.show()

```

## Code Screenshot:

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
dataset = pd.read_csv('iris.csv')
X = dataset.iloc[:,[0,1]].values
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_)
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)
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_[0,0], kmeans.cluster_centers_[0,1], s = 300, c = 'yellow', label = 'Centroids')
plt.title('Clusters')
plt.xlabel('sepal length')
plt.ylabel('sepal width')
plt.legend()
plt.show()
```

### Output Screenshots :

