CSL7670 : Fundamentals of Machine Learning

Lab Report



Name: **SOUHITYA KUNDU**

Roll Number: M20PH209

Program: MSc-MTech(Physics & Materials Sc.)

2		

Chapter 1

Lab-9

1.1 Objective

The objective of this whole assignment is to learn about K-means clustering techniques

1.2 Problem-1

The main objective is to understand K-means clustering approach by proper application in a dataset. with two features, Feature - 1 and Feature - 2.

- \bullet First, I have computed using different values of K for the clustering technique ranging from 2 to 5 as 2,3,4,5
- Second, I have plotted the scatter plot for the various K values in the order by forming loop.

Solution 1:

```
#!/usr/bin/env python
   \# coding: utf-8
   # In [26]:
6
   import numpy as np
   import pandas as pd
   # In [27]:
11
12
   df = pd.read_csv("DATA.csv")
15
16
   # In[28]:
18
19
   df.head(10)
21
22
   # In[29]:
23
24
25
```

```
from sklearn.cluster import KMeans
26
27
   # In [30]:
29
30
31
   kmeans = KMeans(n_clusters = 5, random_state = 0,n_init="auto").fit(df)
32
   kmeans.labels_
33
34
35
   # In [31]:
36
37
38
  kmeans.cluster_centers_
39
40
41
   # In [32]:
42
43
44
   import seaborn as sns
45
   sns.scatterplot(df,x="Feature-1",y="Feature-2",hue=kmeans.labels_)
46
47
   # In [33]:
49
50
51
   # Try different values of k=2,3,4, and 5 and show clustering using
      \hookrightarrow appropriate colors in a scatter
   # plot. Also, show cluster centers.
   # BLACK DOT represents the cluster centers.
54
   import matplotlib.pyplot as plt
56
57
   cluster_vals= [2, 3, 4, 5]
   for i in cluster_vals:
60
       kmeans = KMeans(n_clusters=i,random_state = 0,n_init="auto")
61
       kmeans.fit(df)
62
       cluster_centers = kmeans.cluster_centers_
64
       plt.scatter(df["Feature-1"], df["Feature-2"], c=kmeans.labels_, cmap='
66
          → autumn')
       plt.scatter(cluster_centers[:, 0], cluster_centers[:, 1], c='black',
67
          \hookrightarrow marker='o',s=50)
       plt.title(f'K-Means_Clustering_(k={i})')
       plt.xlabel('Feature 1')
69
       plt.ylabel('Feature<sub>□</sub>2')
70
       plt.show()
71
       print(cluster_centers)
72
73
74
75
76 # In[]:
```

78 79 80 81	1.2. PROBLEM-1		5
	77 78 79 80		
	81		

The K-means cluster based scatter plots are given below for different values of k such as 2,3,4,5. The Black Dots in the pictures represent the center of the corresponding neighborhood

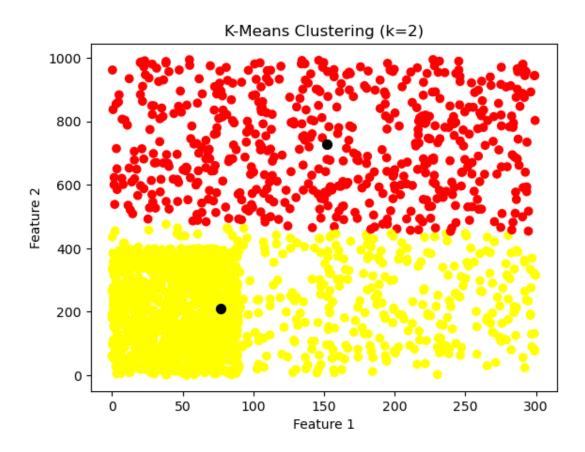


Figure 1.1: K-means cluster with k = 2.

1.2. PROBLEM-1 7

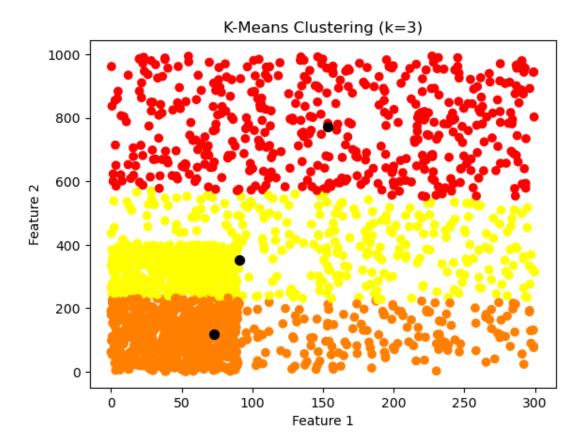


Figure 1.2: K-means cluster with $\mathbf{k}=3.$

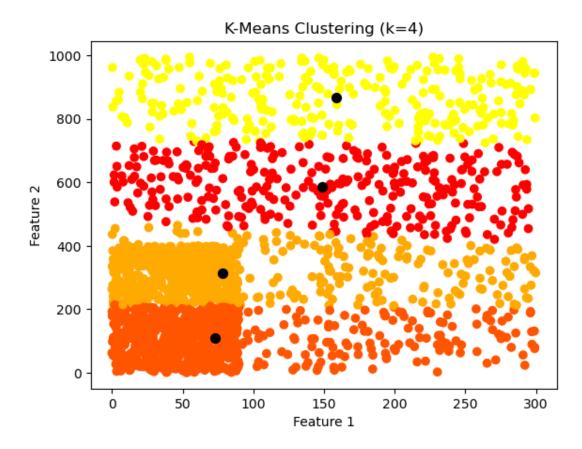


Figure 1.3: K-means cluster with $\mathbf{k}=4.$

1.2. PROBLEM-1

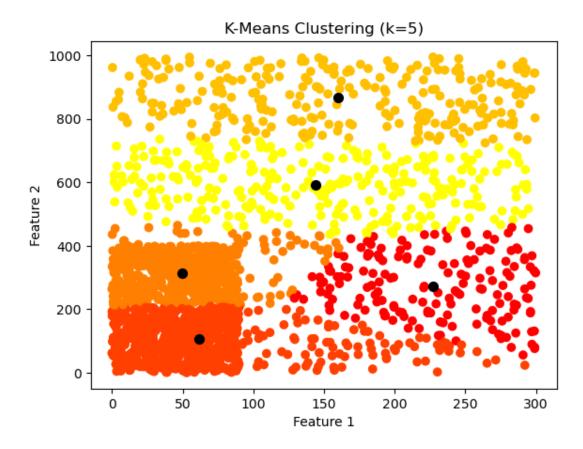


Figure 1.4: K-means cluster with $\mathbf{k}=5.$

1.3 Problem-2

The Wikipedia article document has been provided. I had to manually identify 5 keywords in these documents and use them to represent documents using Bag of Words(BOW). I have clustered them in 2 clusters and shared my observations.

Solution 2:

```
#!/usr/bin/env python
                      # coding: utf-8
     2
     3
                      # In [9]:
     6
                     from sklearn.feature_extraction.text import CountVectorizer
                      import pandas as pd
                      import matplotlib.pyplot as plt
                      import seaborn as sns
                      from sklearn.cluster import KMeans
                     \tt f1='C:/Users/user/FML_{\bot}+_{\bot}IMAGE\_PROCESSING+AI\_B/FML\_Assignments/A\_9/dataset/ASSING+AI\_B/FML\_Assignments/ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI\_B/FML\_ASSING+AI_B/FML\_ASSING+AI_B/FML\_ASSING+AI_B/FML\_ASSING+AI_B/FML\_ASSING+AI_B/FML\_ASSING+AI_B/FML\_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FML_ASSING+AI_B/FM
 13
                     \texttt{f2='C:/Users/user/FML}_{\bot} \bot \texttt{IMAGE\_PROCESSING+AI\_B/FML\_Assignments/A\_9/dataset/Algariantering} + \texttt{AIRARIANTERING} + \texttt{A
 14
                                              \hookrightarrow n2.txt'
                      f3='C:/Users/user/FML_+_IMAGE_PROCESSING+AI_B/FML_Assignments/A_9/dataset/
                                              \hookrightarrow n3.txt,
                      f4='C:/Users/user/FML_{\sqcup}+_{\sqcup}IMAGE\_PROCESSING+AI\_B/FML\_Assignments/A\_9/dataset/BerryColored
                      \texttt{f5='C:/Users/user/FML}_{\sqcup} + _{\sqcup} \texttt{IMAGE\_PROCESSING+AI\_B/FML\_Assignments/A\_9/dataset/BML}_{\sqcup} + _{\sqcup} \texttt{IMAGE\_PROCESSING+AI\_B/FML\_Assignments/A\_9/dataset/BML_Assignments/A\_9/dataset/BML}_{\sqcup} + _{\sqcup} \texttt{IMAGE\_PROCESSING+AI\_B/FML\_Assignments/A\_9/dataset/BML}_{\sqcup} + _{\sqcup} \texttt{IMAGE\_PROCESSING+AI\_B/FML}_{\sqcup} + _{\sqcup} + _{\sqcup} \texttt{IMAGE\_PROCESSING+AI\_B/FML}_{\sqcup} + _{\sqcup} + _
                                              \hookrightarrow n5.txt'
 18
                      paths = [f1,f2,f3,f4,f5]
 19
 20
                      output = []
 21
                      for path in paths:
 23
                                                        with open(path, 'r', encoding='utf-8') as file:
 24
                                                                                        content = file.read()
 25
                                                                                        output.append(content.lower())
 26
 27
                      # The 5-Keywords
                      keywords = ["tendulkar", "politician", "australian", "minister", "economic
 30
                                             \hookrightarrow "]
31
                      #BOW
 32
 33
                      countvectorizer = CountVectorizer(vocabulary=keywords)
 34
                      X = countvectorizer.fit_transform(output)
 35
                      kmeans = KMeans(n_clusters=2, random_state=0).fit(X)
 37
                      print(kmeans.labels_)
 38
 40
 41
```

	1.3. PROBLEM-2	11
42		1
43	# In[]:	

CONCLUSION:

|--|

Based on the above result I have made two clusters and fitted them accordingly in 1 and 0 labels. The manually chosen keywords are tendulkar, politician, australian, minister, economic. Hence the conclusion observed as [1-0-0-0]