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Pattern Recognition Assignment -2
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          Question 2
          This repository contains the dataset
 In [61]: cd ./datasets/
          [Errno 2] No such file or directory: './datasets/'
          /mnt/c/Users/siddh/ML_lab/PR-assignment-2/datasets
 In [62]: ls
           Index*
                          bezdekIris.data* iris.names*
                                                              sat.trn*
           'Index (1)'* iris.data*
                                            'sat (1).doc'* sat.tst*
 In [63]: import matplotlib.pyplot as plt
          import numpy as np
          import pandas as pd
          import seaborn as sns
          from sklearn.cluster import KMeans
          from sklearn import metrics
          from sklearn.model_selection import train_test_split
          from sklearn import datasets
          from scipy.spatial.distance import cdist
          import math
          Loading the dataset
 In [64]: # load iris dataset
          iris = datasets.load_iris()
          # Since this is a bunch, create a dataframe
          iris_df=pd.DataFrame(iris.data)
          iris_df['class']=iris.target
          iris_df.columns=['sepal_length', 'sepal_width', 'petal_length', 'petal_width', 'class']
In [65]: iris_df.head()
 Out[65]:
              sepal_length sepal_width petal_length petal_width class
           0
                                                  0.2
           1
                    4.9
                              3.0
                                        1.4
                                                  0.2
                                                        0
                    4.7
                              3.2
                                        1.3
                                                  0.2
                                                        0
           3
                    4.6
                              3.1
                                        1.5
                                                  0.2
                                                        0
                    5.0
                                        1.4
                                                  0.2
                              3.6
 In [66]: iris_df.describe()
Out[66]:
                 sepal_length sepal_width petal_length petal_width
           count 150.000000 150.000000 150.000000 150.000000
                    5.843333
                             3.057333
                                       3.758000
                                                 1.199333
                                                          1.000000
                    0.828066
                             0.435866
                                       1.765298
                                                 0.762238
                                                          0.819232
                    4.300000
                             2.000000
                                        1.000000
                                                 0.100000
                                                           0.000000
            25%
                    5.100000
                             2.800000
                                       1.600000
                                                 0.300000
                                                          0.000000
                    5.800000
                             3.000000
                                        4.350000
                                                  1.300000
                                                           1.000000
            75%
                    6.400000
                             3.300000
                                       5.100000
                                                  1.800000
                                                          2.000000
                    7.900000
                             4.400000
                                        6.900000
                                                 2.500000
                                                          2.000000
 In [67]: sns.heatmap(iris_df.corr(), annot = True, linecolor='black')
 Out[67]: <AxesSubplot:>
                                   0.87
                                         0.82
                                               0.78
           sepal_length <sup>-</sup>
                                                         - 0.8
                                                        - 0.6
                                   -0.43
                                        -0.37
            sepal_width
           petal_length
                                                         - 0.2
                                                0.96
                      0.82
                            -0.37
            petal_width
                                                         - -0.2
                                   0.95
                                         0.96
                      0.78
 In [68]: iris_df.shape
 Out[68]: (150, 5)
 In [69]: X = iris_df.iloc[:,:4]
          Y = iris_df.iloc[:,4]
 In [70]: X.hist()
          plt.show()
                                          sepal_width
                  sepal_length
           20 -
                                   20
           10
                  <sup>5</sup>petal <sup>6</sup>length
                                          petal_width 4
           20 -
          Test Train split
 In [71]: xtrain, xval, ytrain, yval = train_test_split(X,Y, test_size=0.3, random_state=42)
In [100]: | plt.scatter(x=iris_df.petal_length,y=iris_df.petal_width)
          plt.title("The dataset")
Out[100]: Text(0.5, 1.0, 'The dataset')
           2.5
           2.0
           1.5
           1.0
           0.5
In [103]: color =np.array(['red','blue','green'])
          plt.scatter(x=iris_df.petal_length,y=iris_df.petal_width,c=color[iris.target])
Out[103]: <matplotlib.collections.PathCollection at 0x7f5a746395b0>
           2.5
           2.0
           1.5
           1.0
          Elbow Method
In [82]: sse=[] #sum of squared error
          # Finding inertia on various k values
          for i in range(1,8):
               kmeans=KMeans(n_clusters = i, init = 'k-means++', max_iter = 100, n_init = 10, random_state = 0).fit(xtrain)
               sse.append(kmeans.inertia_)
          plt.figure(figsize=(10, 6))
          plt.plot(range(1, 8), sse, 'bx-', color='red')
          plt.title('The elbow method- Error vs Number of Clusters')
          plt.xlabel('Number of clusters')
          plt.ylabel('SSE')
Out[82]: Text(0, 0.5, 'SSE')
                                 The elbow method- Error vs Number of Clusters
              400 -
             300
             200
             100
                                             Number of clusters
          The optimal value for number of clusters is 3.
In [105]: kmeans = KMeans(n_clusters=3,init = 'k-means++', max_iter = 100, n_init = 10, random_state = 0)
          y_kmeans = kmeans.fit_predict(x)
In [106]: kmeans.cluster_centers_
Out[106]: array([[5.79555556, 2.69555556, 4.34444444, 1.41777778],
                  [4.96451613, 3.37741935, 1.46451613, 0.2483871 ],
                  [6.85517241, 3.10344828, 5.70689655, 2.02068966]])
          Scatter Plot Visualisations
In [107]: plt.scatter(x[y_kmeans == 0, 0], x[y_kmeans == 0, 1],
                       s = 100, c = 'red', label = 'Iris-setosa')
          plt.scatter(x[y_kmeans == 2, 0], x[y_kmeans == 2, 1],
                       s = 100, c = 'green', label = 'Iris-virginica')
          # Plotting the centroids of the clusters
          plt.scatter(kmeans.cluster_centers_[:, 0], kmeans.cluster_centers_[:,1],
                       s = 100, c = 'black', label = 'Centroids')
           plt.legend()
          plt.title('Sepal Length vs Sepal Width')
Out[107]: Text(0.5, 1.0, 'Sepal Length vs Sepal Width')
                         Sepal Length vs Sepal Width
           3.5
           3.0
           2.5
           2.0
                                            7.0 7.5
                      5.0
                            5.5
                                 6.0
                                       6.5
In [125]: kmeans = KMeans(n_clusters=3,init = 'k-means++', max_iter = 100, n_init = 10, random_state = 0)
          predictedy = kmeans.fit_predict(xtrain)
In [126]: metrics.accuracy_score(predictedy, ytrain)
Out[126]: 0.24761904761904763
In [131]: con_matrix = metrics.confusion_matrix(ytrain, predictedy)
          print(con_matrix)
          [[ 0 31 0]
            [34 0 3]
           [11 0 26]]
In [134]: def draw_confusionmatrix(ytest, yhat):
           plt.figure(figsize=(10,7))
           cm = metrics.confusion_matrix(ytest, yhat)
           ax = sns.heatmap(cm, annot=True, fmt="d")
           plt.ylabel('True label')
           plt.xlabel('Predicted label')
In [135]: draw_confusionmatrix(ytrain, predictedy)
                                                            2
                                      Predicted label
In [136]: kmeans = KMeans(n_clusters=3,init = 'k-means++', max_iter = 100, n_init = 10, random_state = 0)
          predictedy = kmeans.fit_predict(xval)
In [137]: metrics.accuracy_score(predictedy, yval)
Out[137]: 0.93333333333333333
In [140]: draw_confusionmatrix(yval, predictedy)
                                                                            - 17.5
                        19
                                                                            - 15.0
                                                                            - 12.5
```

- 10.0

- 7.5

Predicted label