Lab9- K Nearest Neighbour

In []: Harini Rajarathinam N01419700

Import Libraries

In [1]: import pandas as pd
 import seaborn as sns
 import matplotlib.pyplot as plt
 import numpy as np
 %matplotlib inline

Get the Data(Iris dataset)

In [2]: df = pd.read_csv(r'C:\Users\rcher\Documents\Humber work\Semester 2\Intro to Data
df.head()

Out[2]:

	sepal.length	sepal.width	petal.length	petal.width	variety
0	5.1	3.5	1.4	0.2	0
1	4.9	3.0	1.4	0.2	0
2	4.7	3.2	1.3	0.2	0
3	4.6	3.1	1.5	0.2	0
4	5.0	3.6	1.4	0.2	0

In [56]:

Out[56]:

	sepal.length	sepal.width	petal.length	petal.width	variety
0	5.1	3.5	1.4	0.2	0
1	4.9	3.0	1.4	0.2	0
2	4.7	3.2	1.3	0.2	0
3	4.6	3.1	1.5	0.2	0
4	5.0	3.6	1.4	0.2	0

Standardize the Variables

In [4]: from sklearn.preprocessing import StandardScaler

In [5]: | scaler = StandardScaler()

```
In [6]: | scaler.fit(df.drop('variety',axis=1))
Out[6]: StandardScaler()
In [7]:
        scaled_features = scaler.transform(df.drop('variety',axis=1))
        scaled features
Out[7]: array([[-5.81065904e-01,
                                   8.41837140e-01, -1.01297765e+00,
                 -1.04211089e+00],
                [-8.94308978e-01, -2.07835104e-01, -1.01297765e+00,
                 -1.04211089e+00],
                [-1.20755205e+00,
                                   2.12033793e-01, -1.08231219e+00,
                 -1.04211089e+00],
                [-1.36417359e+00,
                                   2.09934449e-03, -9.43643106e-01,
                 -1.04211089e+00],
                [-7.37687441e-01,
                                   1.05177159e+00, -1.01297765e+00,
                 -1.04211089e+00],
                [-1.11201292e-01,
                                   1.68157493e+00, -8.04974023e-01,
                 -6.86441647e-01],
                [-1.36417359e+00,
                                   6.31902691e-01, -1.01297765e+00,
                 -8.64276271e-01],
                [-7.37687441e-01,
                                   6.31902691e-01, -9.43643106e-01,
                 -1.04211089e+00],
                [-1.67741667e+00, -4.17769553e-01, -1.01297765e+00,
                 -1.04211089e+00],
                [-8.94308978e-01,
                                   2.09934449e-03, -9.43643106e-01,
In [8]: | df_feat = pd.DataFrame(scaled_features,columns=df.columns[:-1])
        df feat.head()
```

Out[8]:

	sepal.length	sepal.width	petal.length	petal.width
0	-0.581066	0.841837	-1.012978	-1.042111
1	-0.894309	-0.207835	-1.012978	-1.042111
2	-1.207552	0.212034	-1.082312	-1.042111
3	-1.364174	0.002099	-0.943643	-1.042111
4	-0.737687	1.051772	-1.012978	-1.042111

Train Test Split

```
In [25]: from sklearn.model_selection import train_test_split
In [29]: X_train, X_test, y_train, y_test = train_test_split(scaled_features, df['variety' random_state=50)
```

Using KNN

```
In [31]: from sklearn.neighbors import KNeighborsClassifier
```

```
In [32]: knn = KNeighborsClassifier(n_neighbors=1)
In [33]: knn.fit(X_train,y_train)
Out[33]: KNeighborsClassifier(n_neighbors=1)
In [34]: pred = knn.predict(X_test)
```

Predictions and Evaluations

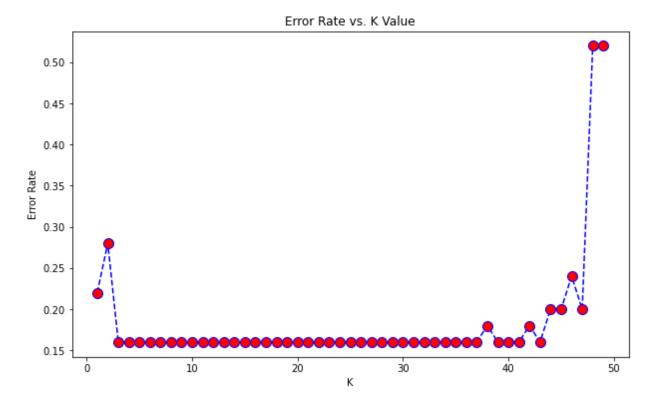
```
In [35]: from sklearn.metrics import classification report, confusion matrix
In [36]: |print(confusion_matrix(y_test,pred))
         [[18 6]
          [ 5 21]]
In [37]: print(classification_report(y_test,pred))
                        precision
                                     recall f1-score
                                                         support
                     0
                             0.78
                                       0.75
                                                  0.77
                                                              24
                                                  0.79
                     1
                             0.78
                                       0.81
                                                              26
                                                  0.78
                                                              50
              accuracy
            macro avg
                             0.78
                                       0.78
                                                  0.78
                                                              50
         weighted avg
                             0.78
                                       0.78
                                                  0.78
                                                              50
```

Choosing a K Value

```
In [38]: error_rate = []
    for i in range (1,50):
        knn = KNeighborsClassifier(n_neighbors = i)
        knn.fit(X_train,y_train)
        pred_i = knn.predict(X_test)
        error_rate.append(np.mean(pred_i != y_test))
In [21]: error_rate
```

```
In [39]: plt.figure(figsize=(10,6))
    plt.plot(range(1,50),error_rate,color='blue', linestyle='dashed', marker='o',
    markerfacecolor='red', markersize=10)
    plt.title('Error Rate vs. K Value')
    plt.xlabel('K')
    plt.ylabel('Error Rate')
```

Out[39]: Text(0, 0.5, 'Error Rate')



```
In [40]: knn = KNeighborsClassifier(n_neighbors=1)
         knn.fit(X_train,y_train)
         pred = knn.predict(X_test)
         print('WITH K=1')
         print('\n')
         print(confusion_matrix(y_test,pred))
         print('\n')
         print(classification_report(y_test,pred))
         WITH K=1
         [[18 6]
          [ 5 21]]
                        precision
                                     recall f1-score
                                                         support
                             0.78
                                        0.75
                                                  0.77
                     0
                                                              24
                     1
                             0.78
                                        0.81
                                                  0.79
                                                              26
                                                  0.78
                                                              50
              accuracy
            macro avg
                             0.78
                                        0.78
                                                  0.78
                                                              50
                                                  0.78
         weighted avg
                             0.78
                                        0.78
                                                              50
In [41]:
         knn = KNeighborsClassifier(n neighbors=3)
         knn.fit(X_train,y_train)
         pred = knn.predict(X_test)
         print('WITH K=3')
         print('\n')
         print(confusion_matrix(y_test,pred))
         print('\n')
         print(classification_report(y_test,pred))
         WITH K=3
         [[20 4]
          [ 4 22]]
                        precision
                                     recall f1-score
                                                         support
                     0
                             0.83
                                        0.83
                                                  0.83
                                                              24
                     1
                             0.85
                                        0.85
                                                  0.85
                                                              26
                                                  0.84
                                                              50
              accuracy
                             0.84
                                        0.84
                                                  0.84
                                                              50
            macro avg
         weighted avg
                             0.84
                                        0.84
                                                  0.84
                                                              50
```

In []: