knn-classifier-1

April 5, 2024

0.1 To predict whether a person will purchase from Social Network Ads

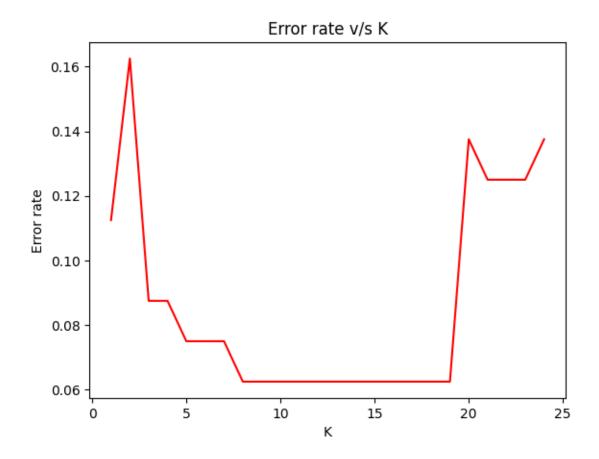
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[1]: ## Importing Necessary Libraries
    import pandas as pd
    from sklearn.model_selection import train_test_split
    from sklearn.neighbors import KNeighborsClassifier
    from sklearn.metrics import accuracy_score, classification_report,_
      from sklearn.preprocessing import StandardScaler
    from sklearn.preprocessing import LabelEncoder
    import matplotlib.pyplot as plt
[2]: ## Loading dataset
    data = pd.read_csv(r'C:\Users\ntpc\Desktop\Social_Network_Ads.csv')
    data.head()
[2]:
        User ID
                Gender Age EstimatedSalary
                                              Purchased
                   Male
                                        19000
    0 15624510
    1 15810944
                   Male
                          35
                                        20000
                                                       0
    2 15668575 Female
                                        43000
                                                       0
                          26
    3 15603246 Female
                                        57000
                                                       0
                          27
    4 15804002
                                                       0
                   Male
                                        76000
                          19
[3]: data.dtypes
[3]: User ID
                        int64
    Gender
                       object
    Age
                        int64
    EstimatedSalary
                        int64
    Purchased
                        int64
    dtype: object
[4]: #checking each value counts
    data['Purchased'].value_counts()
[4]: Purchased
    0
         257
```

1

143

```
Name: count, dtype: int64
```

```
[5]: #dropping User ID column
     data = data.drop('User ID',axis = 1)
     data.head()
[5]:
        Gender Age EstimatedSalary Purchased
     0
          Male
                 19
                                19000
          Male
                 35
                                20000
                                               0
     1
     2 Female
                26
                                43000
                                               0
     3 Female
                 27
                                               0
                                57000
                                               0
          Male
                 19
                                76000
[6]: # Encoding categorical variables
     le = LabelEncoder()
     data['Gender'] = le.fit_transform(data['Gender'])
[7]: # Splitting the data into training and testing sets
     X = data.drop('Purchased', axis=1)
     y = data['Purchased']
     X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,_
       →random_state=42)
[8]: ## Doing Feature Scaling for better accuracy
     from sklearn.preprocessing import StandardScaler
     sc = StandardScaler()
     X_train = sc.fit_transform(X_train)
     X_test = sc.transform(X_test)
[9]: ## Selecting the k value
     acc_list = []
     err_list = []
     for i in range(1,25):
         model = KNeighborsClassifier(n_neighbors=i)
         model.fit(X_train,y_train)
         y_pred = model.predict(X_test)
         acc = accuracy_score(y_test,y_pred)
         acc_list.append(acc)
          err_list.append(1-acc)
[10]: plt.plot(list(range(1,25)),err_list,c= 'r')
     plt.title('Error rate v/s K')
     plt.xlabel('K')
     plt.ylabel('Error rate')
     plt.show()
```



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[11]: model = KNeighborsClassifier(n_neighbors=5)
      model.fit(X_train,y_train)
      y_pred = model.predict(X_test)
[12]: # Evaluating the model
      accuracy = accuracy_score(y_test, y_pred)
      print(f'Accuracy: {accuracy}')
     Accuracy: 0.925
[13]: # Printing classification report and confusion matrix
      print('Classification Report:')
      print(classification_report(y_test, y_pred))
     Classification Report:
                   precision
                                recall f1-score
                                                    support
                0
                        0.96
                                   0.92
                                             0.94
                                                         52
                        0.87
                                   0.93
                                             0.90
                                                         28
```

```
0.93
                                                       80
         accuracy
                                           0.92
                                                       80
        macro avg
                        0.91
                                 0.93
     weighted avg
                        0.93
                                 0.93
                                           0.93
                                                       80
[14]: print('Confusion Matrix:')
     print(confusion_matrix(y_test, y_pred))
     Confusion Matrix:
     [[48 4]
      [ 2 26]]
[]:
```