20bsit154-ass-3

March 30, 2023

```
[3]: # Assignment - III
     # ID : 20BSIT154
     #Q-1 Read the seeds.csv file
     import pandas as pd
     df = pd.read_csv("D:/sem6/dma/practical/seeds.csv")
[4]: print(df)
                             Compactness
                                           Kernel.Length Kernel.Width
           Area
                 {\tt Perimeter}
                                                    5.763
    0
          15.26
                      14.84
                                   0.8710
                                                                    3.312
    1
          14.88
                      14.57
                                   0.8811
                                                     5.554
                                                                    3.333
    2
          14.29
                      14.09
                                   0.9050
                                                     5.291
                                                                    3.337
    3
          13.84
                      13.94
                                   0.8955
                                                     5.324
                                                                    3.379
    4
          16.14
                                                     5.658
                                                                    3.562
                      14.99
                                   0.9034
     . .
    194
          12.19
                      13.20
                                   0.8783
                                                    5.137
                                                                    2.981
    195
          11.23
                                   0.8511
                                                    5.140
                                                                    2.795
                      12.88
    196
          13.20
                      13.66
                                   0.8883
                                                    5.236
                                                                    3.232
          11.84
                                   0.8521
                                                                    2.836
    197
                      13.21
                                                     5.175
                                                                    2.974
    198
          12.30
                      13.34
                                   0.8684
                                                     5.243
          Asymmetry.Coeff
                            Kernel.Groove
                                             Туре
    0
                     2.221
                                     5.220
                                                1
                     1.018
                                     4.956
    1
                                                1
    2
                     2.699
                                     4.825
                                                1
    3
                     2.259
                                     4.805
                                                1
    4
                                     5.175
                     1.355
                                                1
                                                3
    194
                     3.631
                                     4.870
    195
                     4.325
                                     5.003
                                                3
    196
                     8.315
                                     5.056
                                                3
                                                3
    197
                     3.598
                                     5.044
    198
                     5.637
                                     5.063
                                                3
```

[199 rows x 8 columns]

[5]: #Q-2 Perform all data pre-processing techniques. df.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 199 entries, 0 to 198 Data columns (total 8 columns):

#	Column	Non-Null Count	Dtype
0	Area	199 non-null	float64
1	Perimeter	199 non-null	float64
2	Compactness	199 non-null	float64
3	Kernel.Length	199 non-null	float64
4	Kernel.Width	199 non-null	float64
5	Asymmetry.Coeff	199 non-null	float64
6	Kernel.Groove	199 non-null	float64
7	Туре	199 non-null	int64

dtypes: float64(7), int64(1)

memory usage: 12.6 KB

[6]: df.isnull().sum()

[6]: Area 0 Perimeter 0 Compactness 0 Kernel.Length 0 Kernel.Width 0 Asymmetry.Coeff 0 Kernel.Groove 0 Туре 0

dtype: int64

[7]: #Q-3 Perform one-hot encoding and normalization. df2 = pd.get_dummies(df,columns=['Type'])

[8]: print(df2)

	Area	Perimeter	Compactness	Kernel.Length	Kernel.Width	\
0	15.26	14.84	0.8710	5.763	3.312	
1	14.88	14.57	0.8811	5.554	3.333	
2	14.29	14.09	0.9050	5.291	3.337	
3	13.84	13.94	0.8955	5.324	3.379	
4	16.14	14.99	0.9034	5.658	3.562	
	•••	•••	•••	•••	•••	
194	12.19	13.20	0.8783	5.137	2.981	
195	11.23	12.88	0.8511	5.140	2.795	
196	13.20	13.66	0.8883	5.236	3.232	
197	11.84	13.21	0.8521	5.175	2.836	

```
198 12.30
                13.34
                          0.8684
                                      5.243
                                                 2.974
        Asymmetry.Coeff Kernel.Groove Type_1
                                      Type_2
                                             Type_3
    0
               2.221
                           5.220
                                     1
                                           0
                                           0
    1
                1.018
                           4.956
                                     1
                                                 0
    2
               2.699
                           4.825
                                     1
                                           0
                                                 0
    3
               2.259
                           4.805
                                     1
                                           0
                                                 0
    4
                1.355
                           5.175
                                     1
                                           0
                                                 0
                 •••
    . .
               3.631
                                           0
    194
                           4.870
                                     0
                                                 1
               4.325
                           5.003
                                     0
                                           0
                                                 1
    195
               8.315
                           5.056
                                     0
                                           0
                                                 1
    196
                                           0
                                                 1
    197
               3.598
                           5.044
                                     0
               5.637
                           5.063
                                           0
                                                 1
    198
                                     0
    [199 rows x 10 columns]
[9]: from sklearn import preprocessing
    df3 = preprocessing.normalize(df2)
    print(df3)
    [[0.66209382 0.64387105 0.03779055 ... 0.04338754 0.
                                                          ]
                                                  0.
     [0.66344682 0.64962501 0.03928515 ... 0.04458648 0.
                                                  0.
                                                          ]
     [0.6561907  0.64700678  0.04155721  ...  0.04591957  0.
                                                          1
                                                  0.
     [0.59318291 0.61385444 0.03991851 ... 0.
                                          0.
                                                  0.0449381 ]
     [0.59982874 0.6692346 0.04316842 ... 0.
                                          0.
                                                  0.05066121]
     [0.59681799 0.64728065 0.04213632 ... 0.
                                          0.
                                                  0.04852179]]
[10]: x=df.values[:,0:6]
    print(x)
    [[15.26]
            14.84
                   0.871
                         5.763
                               3.312
                                      2.221 ]
     [14.88
            14.57
                   0.8811
                         5.554
                               3.333
                                      1.018]
     [14.29
                                      2.699 ]
            14.09
                   0.905
                         5.291
                               3.337
     [13.2
            13.66
                   0.8883 5.236
                               3.232
                                      8.315 ]
     [11.84
            13.21
                   0.8521
                         5.175
                               2.836
                                      3.598 ]
     [12.3
            13.34
                               2.974
                   0.8684 5.243
                                      5.637 ]]
[11]: y=df.values[:,7]
    print(y)
```

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3. 3. 3. 3. 3. 3.]
[12]: #Q-4 Apply decision tree algorithm and display classification report.
     from sklearn.model_selection import train_test_split
[13]: x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.
      →2, random_state=100)
[14]: from sklearn.tree import DecisionTreeClassifier
     from sklearn.metrics import accuracy_score
     from sklearn import tree
     clf_gini = DecisionTreeClassifier(criterion = "gini", random_state = 100,
                                 max_depth=3, min_samples_leaf=5)
     clf_gini.fit(x_train, y_train)
[14]: DecisionTreeClassifier(max_depth=3, min_samples_leaf=5, random_state=100)
[15]: | y_pred = clf_gini.predict(x_test)
     y_pred
[15]: array([2., 2., 2., 2., 2., 3., 2., 1., 1., 2., 3., 1., 3., 3., 2., 3., 3.,
           3., 3., 1., 1., 2., 1., 2., 2., 2., 3., 2., 3., 2., 2., 3., 3.,
           3., 2., 1., 3., 1., 2.])
[16]: print("Accuracy is ", accuracy_score(y_test,y_pred)*100)
    Accuracy is 87.5
[17]: from sklearn.metrics import confusion matrix
     print(confusion_matrix(y_test, y_pred))
    [[5 1 1]
     [ 2 16 0]
     [ 1 0 14]]
[18]: from sklearn.metrics import classification_report
     print(classification_report(y_test, y_pred))
                            recall f1-score
                 precision
                                              support
                                                   7
                              0.71
                                       0.67
             1.0
                      0.62
                               0.89
             2.0
                      0.94
                                        0.91
                                                  18
             3.0
                      0.93
                              0.93
                                       0.93
                                                  15
                                       0.88
                                                  40
        accuracy
```

2. 2. 2. 2. 2. 2. 2. 2. 2. 2. 2. 2. 2. 3. 3. 3. 3. 3. 3. 3. 3. 3. 3.

```
weighted avg
                        0.88
                                  0.88
                                             0.88
                                                         40
[19]: clf_entropy = DecisionTreeClassifier(criterion = "entropy", random_state = 100,
      max_depth=3, min_samples_leaf=5)
      clf_entropy.fit(x_train, y_train)
[19]: DecisionTreeClassifier(criterion='entropy', max_depth=3, min_samples_leaf=5,
                             random state=100)
[20]: y_pred = clf_entropy.predict(x_test)
      y_pred
[20]: array([2., 2., 2., 2., 2., 3., 2., 1., 2., 3., 2., 3., 3., 2., 3., 3.,
             3., 3., 1., 1., 2., 1., 1., 2., 2., 3., 2., 3., 2., 2., 3., 3., 3.,
             3., 2., 1., 3., 3., 2.])
[21]: print("Accuracy is ", accuracy_score(y_test,y_pred)*100)
     Accuracy is 97.5
[22]: from sklearn.metrics import confusion_matrix
      print(confusion_matrix(y_test, y_pred))
     [[6 0 1]
      Γ 0 18 0]
      [ 0 0 15]]
[23]: from sklearn.metrics import classification report
      print(classification_report(y_test, y_pred))
                   precision
                                recall f1-score
                                                   support
              1.0
                        1.00
                                  0.86
                                            0.92
                                                         7
                        1.00
                                  1.00
                                            1.00
              2.0
                                                         18
              3.0
                        0.94
                                  1.00
                                            0.97
                                                         15
                                            0.97
                                                         40
         accuracy
        macro avg
                        0.98
                                  0.95
                                            0.96
                                                         40
                                  0.97
                                            0.97
     weighted avg
                        0.98
                                                         40
[24]: #Q -5 Apply naïve-byes algorithm and display classification report.
      from sklearn.naive_bayes import GaussianNB
      model = GaussianNB()
      # Model-training
      model.fit(x_train, y_train)
```

0.83

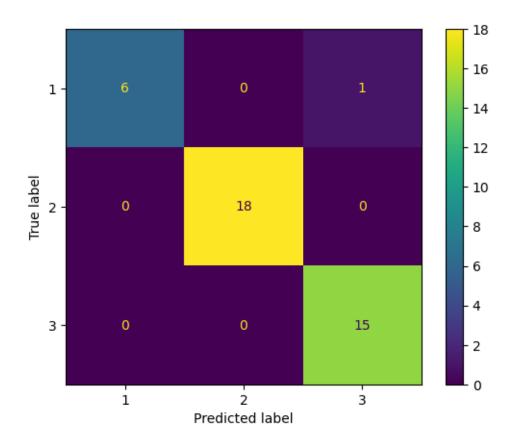
macro avg

0.85

0.84

40

```
y_pred = model.predict(x_test)
      y_pred
[24]: array([2., 2., 2., 2., 2., 3., 2., 1., 2., 3., 2., 3., 3., 2., 3., 3.,
             3., 3., 1., 1., 2., 1., 1., 2., 2., 3., 2., 3., 2., 2., 3., 3., 3.,
            3., 2., 1., 3., 3., 2.])
[25]: from sklearn.metrics import (
          accuracy_score,
          confusion_matrix,
          ConfusionMatrixDisplay,
          f1_score,
      print("Accuracy is ", accuracy_score(y_test,y_pred)*100)
      f1 = f1_score(y_pred, y_test, average="weighted")
      print("F1 Score:", f1)
     Accuracy is 97.5
     F1 Score: 0.9755583126550869
[26]: from sklearn.metrics import confusion_matrix
      print(confusion_matrix(y_test, y_pred))
     [[ 6 0 1]
      [ 0 18 0]
      [ 0 0 15]]
[27]: labels = [1,2,3]
      cm = confusion_matrix(y_test, y_pred, labels=labels)
      disp = ConfusionMatrixDisplay(confusion_matrix=cm, display_labels=labels)
      disp.plot();
```



[28]: from sklearn.metrics import classification_report print(classification_report(y_test, y_pred))

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| | | | | |
| 1.0 | 1.00 | 0.86 | 0.92 | 7 |
| 2.0 | 1.00 | 1.00 | 1.00 | 18 |
| 3.0 | 0.94 | 1.00 | 0.97 | 15 |
| | | | | |
| accuracy | | | 0.97 | 40 |
| macro avg | 0.98 | 0.95 | 0.96 | 40 |
| weighted avg | 0.98 | 0.97 | 0.97 | 40 |

[]: