## Leihui Li-day-3-handin-solution

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## 1 Basic Data Science in Python - Handin 3

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This handin is indiviual and mandatory to pass the course.

Hand in this .ipynb file and the compiled pdf, no later than 27/10 kl 09:30.

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import sklearn
from sklearn import datasets
from sklearn.model_selection import train_test_split
from sklearn.decomposition import PCA
```

## 1.1.1 Exercise 2: Classify Wine (Handin)

Use a decision tree to classify wine: Train on the training data, and test your fit on the test data. Print the In Sample Score and Test Score.

Try to change the hyperparameters a bit around to get the best Test score.

```
[2]: from sklearn.tree import DecisionTreeClassifier
from sklearn.model_selection import train_test_split

wine = datasets.load_wine()
X = wine.data
y = wine.target
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25)

df = pd.DataFrame(X, columns=wine.feature_names)
df.head()
```

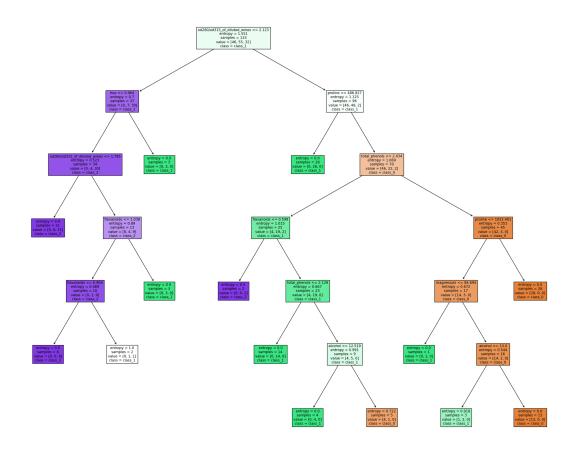
```
[2]:
        alcohol malic_acid
                              ash alcalinity_of_ash magnesium total_phenols \
     0
          14.23
                       1.71 2.43
                                                15.6
                                                           127.0
                                                                           2.80
          13.20
                       1.78 2.14
                                                11.2
     1
                                                           100.0
                                                                           2.65
     2
          13.16
                       2.36 2.67
                                                18.6
                                                           101.0
                                                                           2.80
     3
          14.37
                       1.95 2.50
                                                16.8
                                                                           3.85
                                                           113.0
          13.24
                       2.59 2.87
                                                21.0
                                                                           2.80
                                                           118.0
```

```
flavanoids nonflavanoid_phenols proanthocyanins color_intensity hue \
     0
              3.06
                                    0.28
                                                      2.29
                                                                       5.64 1.04
              2.76
                                    0.26
                                                      1.28
                                                                       4.38 1.05
     1
     2
              3.24
                                    0.30
                                                      2.81
                                                                       5.68 1.03
                                                                       7.80 0.86
     3
              3.49
                                    0.24
                                                      2.18
              2.69
                                                      1.82
                                                                       4.32 1.04
                                    0.39
        od280/od315_of_diluted_wines proline
                                3.92
                                       1065.0
     0
                                3.40
                                       1050.0
     1
     2
                                3.17
                                      1185.0
     3
                                3.45
                                      1480.0
     4
                                2.93
                                       735.0
[3]: ### YOUR CODE HERE
     from sklearn import tree
     np.set_printoptions(precision=3, suppress=True)
     new X train = X train
     new_X_{test} = X_{test}
     # Try to use PCA to reduce the dimension of the feature, but looks like notu
      ⇔qood so much
     # print (X_train.shape)
     # pca = PCA(n_components=12)# 10, 11, 12
     \# pca.fit(X_train)
     # new X train = pca.transform(X train)
     \# new_X_{test} = pca.transform(X_{test})
     print (new_X_train.shape, new_X_test.shape)
     print (new_X_train[0].tolist(),'\n', new_X_test[0].tolist())
     clf_model = tree.DecisionTreeClassifier(
         criterion='entropy',
         splitter = 'random',
         random_state=42,
         max_depth=6,
         #min_impurity_decrease = 0.01,
         min_samples_leaf=1,
         min samples split=3)
     clf_model = clf_model.fit(new_X_train, y_train)
     clf_score_sample, clf_score_test = clf_model.score(new_X_train,y_train),_
      →clf_model.score(new_X_test,y_test)
           print (clf_model.get_params())
     print (f"sampe score: {clf_score_sample} test score: {clf_score_test}")
```

(133, 13) (45, 13)

```
[11.41, 0.74, 2.5, 21.0, 88.0, 2.48, 2.01, 0.42, 1.44, 3.08, 1.1, 2.31, 434.0]
     [12.79, 2.67, 2.48, 22.0, 112.0, 1.48, 1.36, 0.24, 1.26, 10.8, 0.48, 1.47,
    480.0]
    sampe score: 0.9774436090225563 test score: 0.9555555555555556
[4]: import random
    np.set_printoptions(precision=6, suppress=True)
    x_i = random.randint(0, len(X_test))
    print (f"we select the {x_i}th among the test data")
    print (f"Ground Truth:\n{X_test[x_i].tolist()} -> {wine.
      →target_names[y_test[x_i]]}")
    print (f"Predict Result:\n{wine.target_names[clf_model.predict(X_test[x_i].
      \neg reshape(1, -1))].item()}")
    we select the 13th among the test data
    Ground Truth:
    [11.61, 1.35, 2.7, 20.0, 94.0, 2.74, 2.92, 0.29, 2.49, 2.65, 0.96, 3.26, 680.0]
    -> class 1
    Predict Result:
    class_1
[5]: # Have a test on the generated tree
    text_representation = tree.export_text(clf_model)
    print(text_representation)
    |--- feature_11 <= 2.12
        |--- feature_10 <= 0.96
            |--- feature_11 <= 1.78
               |--- class: 2
            |--- feature_11 > 1.78
            | |--- feature_6 <= 1.04
            | | |--- feature_6 <= 0.90
              | | |--- class: 2
            |--- feature_6 > 0.90
                    | |--- class: 1
                |--- feature_6 > 1.04
            | |--- class: 1
        |--- feature_{10} > 0.96
        | |--- class: 1
    |--- feature_11 > 2.12
        |--- feature_12 <= 486.83
          |--- class: 1
        |--- feature_12 > 486.83
          |--- feature 5 <= 2.43
            | |--- feature_6 <= 0.60
            | | |--- class: 2
           | |--- feature_6 > 0.60
            | | |--- feature_5 <= 2.13
```

```
|--- class: 1
       |--- feature_5 > 2.13
           |--- feature_0 <= 12.52
           | |--- class: 1
           |--- feature_0 > 12.52
           | |--- class: 0
|--- feature 5 > 2.43
   |--- feature_12 <= 1022.40
       |--- feature_4 <= 84.69
           |--- class: 1
       |--- feature_4 > 84.69
           |--- feature_0 <= 13.00
           | |--- class: 1
           |--- feature_0 > 13.00
               |--- class: 0
   |--- feature_12 > 1022.40
   | |--- class: 0
```



[]:	
[]:	