

Exp-7 Decision Tree Random forest

November 1, 2022

```
[1]: import pandas
      from sklearn import tree
      from sklearn.tree import DecisionTreeClassifier
      import matplotlib.pyplot as plt
```

```
[2]: from sklearn.model_selection import train_test_split
      from sklearn.datasets import load_iris
```

```
[3]: irisData = load_iris()
```

```
[4]: X = irisData.data
      y = irisData.target
```

```
[5]: X
```

```
[5]: array([[5.1, 3.5, 1.4, 0.2],
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            [4.6, 3.6, 1. , 0.2],
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 [6.4, 3.2, 5.3, 2.3],
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```

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[6.5, 3. , 5.2, 2. ],
[6.2, 3.4, 5.4, 2.3],
[5.9, 3. , 5.1, 1.8]])

```

```
[6]: y
```

```

[6]: array([0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
          0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
          0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
          1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
          1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
          2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
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          2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2])

```

```

[7]: X_train, X_test, y_train, y_test = train_test_split(
      X, y, test_size = 0.3, random_state=50)

```

```
[8]: dtree = DecisionTreeClassifier()
dtree.fit(X_train, y_train)
```

```
[8]: DecisionTreeClassifier()
```

```
[9]: y_pred = dtree.predict(X_test)
```

```
[10]: y_pred
```

```
[10]: array([1, 1, 0, 0, 2, 2, 2, 0, 0, 1, 0, 2, 0, 2, 1, 0, 1, 0, 1, 2, 2, 1,
        0, 2, 1, 2, 1, 1, 1, 2, 2, 1, 1, 2, 0, 0, 1, 1, 1, 0, 0, 1, 2, 0,
        2])
```

```
[11]: from sklearn.metrics import confusion_matrix, accuracy_score, \
      ↪ classification_report

cm_DT = confusion_matrix(y_test, y_pred)

print(f"Confusion Matrix for DT:\n{cm_DT}\n")

acc_DT = accuracy_score(y_test, y_pred)
print(f"Accuracy Score: {acc_DT}")
```

Confusion Matrix for DT:

```
[[14  0  0]
 [ 0 16  1]
 [ 0  1 13]]
```

Accuracy Score: 0.9555555555555556

```
[12]: print(f"Classification Report for DT:\n{classification_report(y_test, \
      ↪ y_pred)}\n")
```

Classification Report for DT:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	14
1	0.94	0.94	0.94	17
2	0.93	0.93	0.93	14
accuracy			0.96	45
macro avg	0.96	0.96	0.96	45
weighted avg	0.96	0.96	0.96	45

```
[13]: from sklearn import tree
```

```
[14]: plt.figure(figsize=(10, 10)) # Resize figure
tree.plot_tree(dtrees, filled=True)
plt.show()
```



1 Random Forest

```
[15]: from sklearn.ensemble import RandomForestClassifier
```

```
[16]: classifier_rf = RandomForestClassifier(random_state=0, n_jobs=-1, max_depth=10,
n_estimators=100, oob_score=True)
```

```
[17]: classifier_rf.fit(X_train, y_train)
```

```
[17]: RandomForestClassifier(max_depth=10, n_jobs=-1, oob_score=True, random_state=0)
```

```
[18]: y_pred_RF = classifier_rf.predict(X_test)
```

```
[19]: y_pred_RF
```

```
[19]: array([1, 1, 0, 0, 2, 2, 2, 0, 0, 1, 0, 2, 0, 2, 1, 0, 1, 0, 1, 2, 2, 1,
         0, 2, 1, 2, 1, 1, 1, 2, 2, 1, 1, 2, 0, 0, 1, 1, 1, 0, 0, 1, 2, 0,
         2])
```

```
[20]: # Confusion Matrix
cm_RF= confusion_matrix(y_test, y_pred_RF)
print(f"Confusion Matrix for RF:\n{cm_RF}\n")
```

Confusion Matrix for RF:

```
[[14  0  0]
 [ 0 16  1]
 [ 0  1 13]]
```

```
[21]: # Accuracy Score
acc_RF = accuracy_score(y_test, y_pred_RF)
print(f"Accuracy Score: {acc_RF}")
```

Accuracy Score: 0.9555555555555556

```
[22]: # Classification Report
print(f"Classification Report for RF:\n{classification_report(y_test,
↪y_pred_RF)}\n")
```

Classification Report for RF:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	14
1	0.94	0.94	0.94	17
2	0.93	0.93	0.93	14
accuracy			0.96	45
macro avg	0.96	0.96	0.96	45
weighted avg	0.96	0.96	0.96	45