

Lab 13 19-10-2022

October 19, 2022

1 Decision Tree

```
[1]: import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
```

1.1 Load Dataset

```
[2]: df = pd.read_csv('iris.csv')
print(df)
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa
..
145	6.7	3.0	5.2	2.3	virginica
146	6.3	2.5	5.0	1.9	virginica
147	6.5	3.0	5.2	2.0	virginica
148	6.2	3.4	5.4	2.3	virginica
149	5.9	3.0	5.1	1.8	virginica

[150 rows x 5 columns]

```
[3]: df.species.value_counts()
```

```
[3]: setosa      50
versicolor  50
virginica    50
Name: species, dtype: int64
```

1.2 Preprocessing

```
[4]: y=df['species']  
print(y)  
df=df.drop(['species'],axis=1)  
x= df  
print(x)
```

```
0      setosa  
1      setosa  
2      setosa  
3      setosa  
4      setosa  
...  
145     virginica  
146     virginica  
147     virginica  
148     virginica  
149     virginica  
Name: species, Length: 150, dtype: object  
   sepal_length  sepal_width  petal_length  petal_width  
0           5.1           3.5           1.4           0.2  
1           4.9           3.0           1.4           0.2  
2           4.7           3.2           1.3           0.2  
3           4.6           3.1           1.5           0.2  
4           5.0           3.6           1.4           0.2  
..          ...           ...           ...           ...  
145          6.7           3.0           5.2           2.3  
146          6.3           2.5           5.0           1.9  
147          6.5           3.0           5.2           2.0  
148          6.2           3.4           5.4           2.3  
149          5.9           3.0           5.1           1.8
```

[150 rows x 4 columns]

```
[5]: x.describe()
```

```
[5]:
```

	sepal_length	sepal_width	petal_length	petal_width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.054000	3.758667	1.198667
std	0.828066	0.433594	1.764420	0.763161
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.350000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000

1.3 Test & Train

1.3.1 Split Dataset

```
[6]: from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test= train_test_split(x, y, test_size= 0.25,
↪random_state=0)
```

1.3.2 Feature Scaling

```
[7]: from sklearn.preprocessing import MinMaxScaler
st_x= MinMaxScaler()
x_train= st_x.fit_transform(x_train)
x_test=st_x.fit_transform(x_test)
print(x_train)
```

```
[[0.44444444 0.41666667 0.53448276 0.58333333]
 [0.41666667 0.25      0.5        0.45833333]
 [0.69444444 0.41666667 0.75862069 0.83333333]
 [0.11111111 0.5        0.03448276 0.04166667]
 [0.72222222 0.45833333 0.68965517 0.91666667]
 [0.19444444 0.625      0.0862069  0.20833333]
 [0.30555556 0.70833333 0.06896552 0.04166667]
 [0.19444444 0.         0.4137931  0.375      ]
 [0.61111111 0.41666667 0.75862069 0.70833333]
 [0.66666667 0.54166667 0.79310345 1.         ]
 [0.47222222 0.08333333 0.67241379 0.58333333]
 [0.66666667 0.20833333 0.81034483 0.70833333]
 [0.36111111 0.20833333 0.48275862 0.41666667]
 [0.94444444 0.41666667 0.86206897 0.91666667]
 [0.55555556 0.54166667 0.62068966 0.625      ]
 [0.33333333 0.16666667 0.46551724 0.41666667]
 [0.55555556 0.29166667 0.65517241 0.70833333]
 [0.55555556 0.33333333 0.68965517 0.58333333]
 [0.16666667 0.20833333 0.5862069  0.66666667]
 [0.55555556 0.20833333 0.67241379 0.75      ]
 [0.75        0.5        0.62068966 0.54166667]
 [0.61111111 0.41666667 0.70689655 0.79166667]
 [0.47222222 0.58333333 0.5862069  0.625      ]
 [0.13888889 0.45833333 0.0862069  0.04166667]
 [0.41666667 0.29166667 0.68965517 0.75      ]
 [0.36111111 0.29166667 0.53448276 0.5        ]
 [0.36111111 0.375      0.43103448 0.5        ]
 [0.33333333 0.20833333 0.5         0.5        ]
 [0.5         0.41666667 0.60344828 0.54166667]
 [0.80555556 0.5         0.84482759 0.70833333]
 [0.27777778 0.70833333 0.06896552 0.04166667]
 [0.         0.41666667 0.         0.         ]]
```

[0.58333333 0.29166667 0.72413793 0.75]
 [0.38888889 0.41666667 0.53448276 0.45833333]
 [0.30555556 0.58333333 0.10344828 0.04166667]
 [0.38888889 1. 0.06896552 0.125]
 [0.72222222 0.45833333 0.65517241 0.58333333]
 [0.08333333 0.45833333 0.06896552 0.04166667]
 [0.44444444 0.41666667 0.68965517 0.70833333]
 [0.22222222 0.20833333 0.32758621 0.41666667]
 [0.08333333 0.58333333 0.05172414 0.08333333]
 [0.52777778 0.08333333 0.5862069 0.58333333]
 [0.80555556 0.66666667 0.86206897 1.]
 [0.38888889 0.375 0.53448276 0.5]
 [0.13888889 0.41666667 0.05172414 0.]
 [0.77777778 0.41666667 0.82758621 0.83333333]
 [0.72222222 0.5 0.79310345 0.91666667]
 [0.61111111 0.41666667 0.81034483 0.875]
 [0.58333333 0.33333333 0.77586207 0.83333333]
 [0.22222222 0.75 0.0862069 0.04166667]
 [0.13888889 0.58333333 0.0862069 0.04166667]
 [0.61111111 0.5 0.68965517 0.79166667]
 [0.66666667 0.54166667 0.79310345 0.83333333]
 [0.05555556 0.125 0.03448276 0.08333333]
 [0.52777778 0.58333333 0.74137931 0.91666667]
 [0.16666667 0.41666667 0.05172414 0.04166667]
 [0.38888889 0.20833333 0.67241379 0.79166667]
 [0.72222222 0.45833333 0.74137931 0.83333333]
 [0.02777778 0.5 0.03448276 0.04166667]
 [0.19444444 0.66666667 0.05172414 0.04166667]
 [0.80555556 0.41666667 0.81034483 0.625]
 [0.22222222 0.625 0.05172414 0.08333333]
 [0.02777778 0.41666667 0.03448276 0.04166667]
 [0.30555556 0.79166667 0.10344828 0.125]
 [0.33333333 0.125 0.5 0.5]
 [0.69444444 0.5 0.82758621 0.91666667]
 [0.91666667 0.41666667 0.94827586 0.83333333]
 [0.22222222 0.625 0.05172414 0.04166667]
 [0.16666667 0.45833333 0.06896552 0.]
 [0.25 0.58333333 0.05172414 0.04166667]
 [0.38888889 0.33333333 0.5862069 0.5]
 [0.63888889 0.41666667 0.56896552 0.54166667]
 [0.19444444 0.5 0.01724138 0.04166667]
 [0.22222222 0.54166667 0.10344828 0.16666667]
 [0.58333333 0.375 0.55172414 0.5]
 [0.30555556 0.58333333 0.06896552 0.125]
 [0.94444444 0.25 1. 0.91666667]
 [0.16666667 0.16666667 0.37931034 0.375]
 [1. 0.75 0.9137931 0.79166667]
 [0.66666667 0.45833333 0.56896552 0.54166667]

```
[0.25      0.875      0.06896552 0.         ]
[0.47222222 0.41666667 0.63793103 0.70833333]
[0.41666667 0.83333333 0.01724138 0.04166667]
[0.94444444 0.33333333 0.96551724 0.79166667]
[0.22222222 0.75      0.06896552 0.08333333]
[0.11111111 0.5       0.0862069  0.04166667]
[0.86111111 0.33333333 0.86206897 0.75      ]
[0.19444444 0.54166667 0.05172414 0.04166667]
[0.55555556 0.58333333 0.77586207 0.95833333]
[0.38888889 0.33333333 0.51724138 0.5       ]
[0.41666667 0.29166667 0.48275862 0.45833333]
[0.38888889 0.25      0.4137931  0.375     ]
[0.58333333 0.5       0.72413793 0.91666667]
[0.66666667 0.41666667 0.70689655 0.91666667]
[0.55555556 0.20833333 0.65517241 0.58333333]
[0.66666667 0.41666667 0.67241379 0.66666667]
[0.19444444 0.41666667 0.0862069  0.04166667]
[0.33333333 0.16666667 0.44827586 0.375     ]
[0.66666667 0.45833333 0.77586207 0.95833333]
[0.41666667 0.29166667 0.68965517 0.75      ]
[0.22222222 0.58333333 0.06896552 0.04166667]
[0.63888889 0.375     0.60344828 0.5       ]
[0.36111111 0.41666667 0.51724138 0.5       ]
[0.44444444 0.5       0.63793103 0.70833333]
[0.55555556 0.125     0.56896552 0.5       ]
[0.33333333 0.625     0.03448276 0.04166667]
[0.22222222 0.70833333 0.06896552 0.125     ]
[0.16666667 0.45833333 0.06896552 0.         ]
[0.55555556 0.375     0.77586207 0.70833333]
[0.41666667 0.29166667 0.51724138 0.375     ]
[0.94444444 0.75      0.96551724 0.875     ]
[0.08333333 0.5       0.05172414 0.04166667]]
```

1.3.3 Fitting to Decision Tree

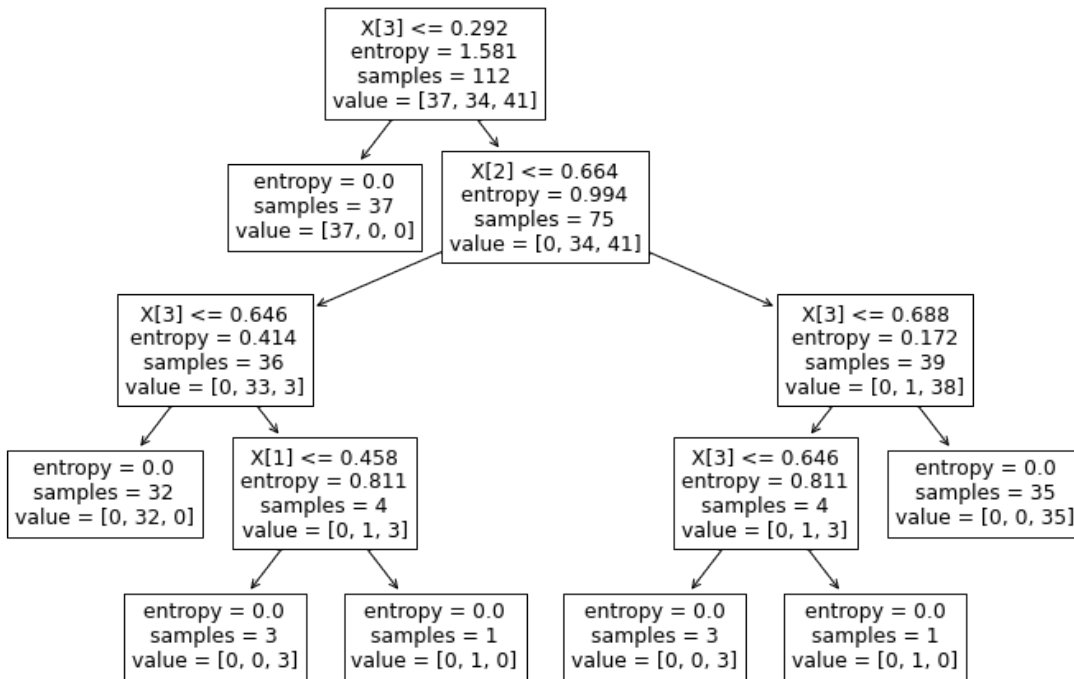
```
[8]: from sklearn import tree
      classifier = tree.DecisionTreeClassifier(criterion = 'entropy', random_state = 0)
      classifier.fit(x_train, y_train)
```

```
[8]: DecisionTreeClassifier(criterion='entropy', random_state=0)
```

1.4 Visualisation

```
[9]: plt.figure(figsize = (12,8))
      tree.plot_tree(classifier)
```

```
[9]: [Text(267.84000000000003, 391.392, 'X[3] <= 0.292\nentropy = 1.581\nsamples = 112\nvalue = [37, 34, 41]'),
      Text(200.88000000000002, 304.416, 'entropy = 0.0\nsamples = 37\nvalue = [37, 0, 0]'),
      Text(334.80000000000007, 304.416, 'X[2] <= 0.664\nentropy = 0.994\nsamples = 75\nvalue = [0, 34, 41]'),
      Text(133.92000000000002, 217.44, 'X[3] <= 0.646\nentropy = 0.414\nsamples = 36\nvalue = [0, 33, 3]'),
      Text(66.96000000000001, 130.464, 'entropy = 0.0\nsamples = 32\nvalue = [0, 32, 0]'),
      Text(200.88000000000002, 130.464, 'X[1] <= 0.458\nentropy = 0.811\nsamples = 4\nvalue = [0, 1, 3]'),
      Text(133.92000000000002, 43.488, 'entropy = 0.0\nsamples = 3\nvalue = [0, 0, 3]'),
      Text(267.84000000000003, 43.488, 'entropy = 0.0\nsamples = 1\nvalue = [0, 1, 0]'),
      Text(535.68000000000001, 217.44, 'X[3] <= 0.688\nentropy = 0.172\nsamples = 39\nvalue = [0, 1, 38]'),
      Text(468.72, 130.464, 'X[3] <= 0.646\nentropy = 0.811\nsamples = 4\nvalue = [0, 1, 3]'),
      Text(401.76000000000005, 43.488, 'entropy = 0.0\nsamples = 3\nvalue = [0, 0, 3]'),
      Text(535.68000000000001, 43.488, 'entropy = 0.0\nsamples = 1\nvalue = [0, 1, 0]'),
      Text(602.64000000000001, 130.464, 'entropy = 0.0\nsamples = 35\nvalue = [0, 0, 35]')] ]
```



1.4.1 Visualisation using Graphviz

```
[10]: !pip install graphviz
import graphviz
```

Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: graphviz in
/home/student/.local/lib/python3.9/site-packages (0.20.1)

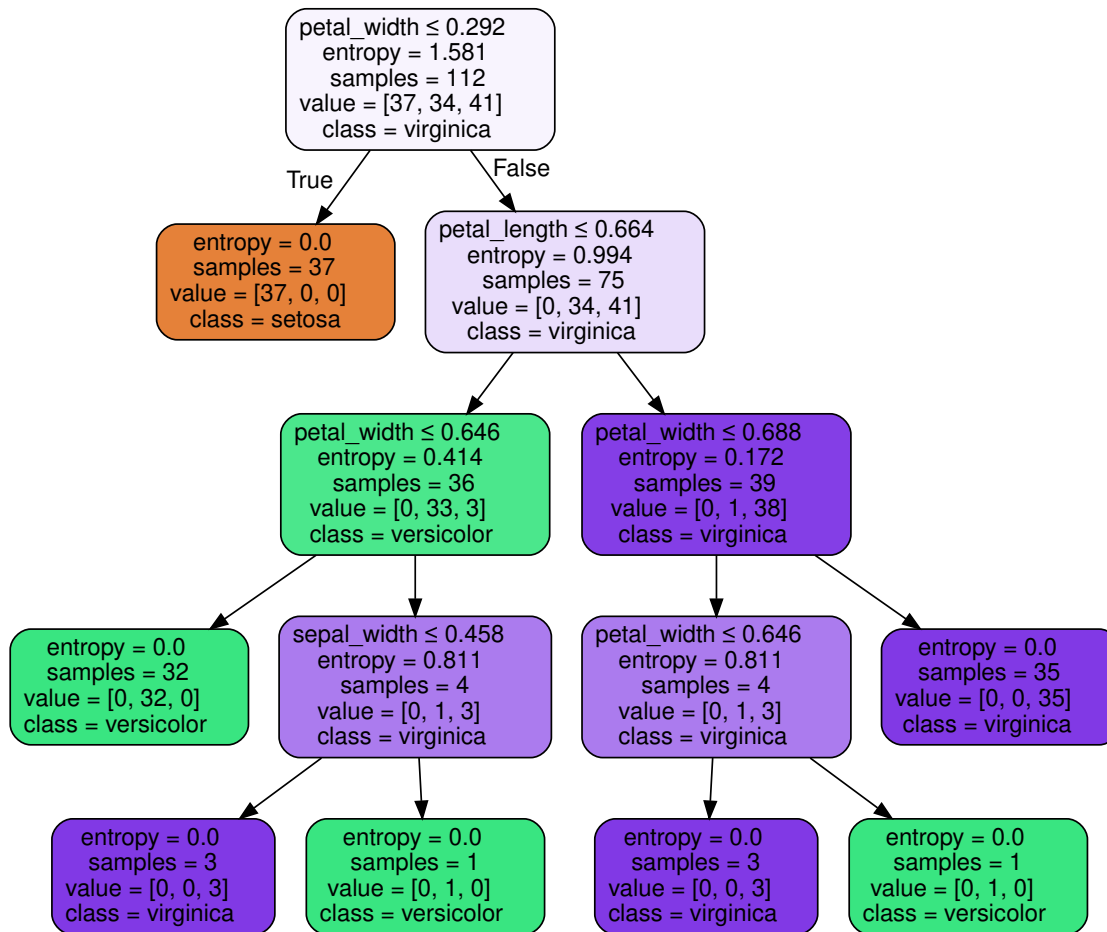
WARNING: You are using pip version 20.3.3; however, version 22.2.2 is
available.

You should consider upgrading via the '/opt/anaconda3/bin/python -m pip install
--upgrade pip' command.

```
[11]: g_tree = tree.export_graphviz(classifier, out_file=None,
                                   feature_names=df.columns,
                                   class_names=['setosa', 'versicolor', 'virginica'],
                                   filled=True, rounded=True,
                                   special_characters=True)
```

```
[12]: graph = graphviz.Source(g_tree)
graph
```

```
[12]:
```



```
[13]: y_pred= classifier.predict(x_test)
      print(y_pred)
```

```
['virginica' 'versicolor' 'setosa' 'virginica' 'setosa' 'virginica'
 'setosa' 'virginica' 'virginica' 'versicolor' 'virginica' 'versicolor'
 'virginica' 'virginica' 'virginica' 'setosa' 'versicolor' 'versicolor'
 'setosa' 'setosa' 'virginica' 'versicolor' 'setosa' 'setosa' 'virginica'
 'setosa' 'setosa' 'versicolor' 'versicolor' 'setosa' 'virginica'
 'versicolor' 'setosa' 'virginica' 'virginica' 'versicolor' 'setosa'
 'virginica']
```

1.5 Confusion Matrix

```
[14]: from sklearn.metrics import confusion_matrix
      cm= confusion_matrix (y_test, y_pred,labels=classifier.classes_)
      print(cm)
```

```
[[13  0  0]
 [ 0 10  6]]
```

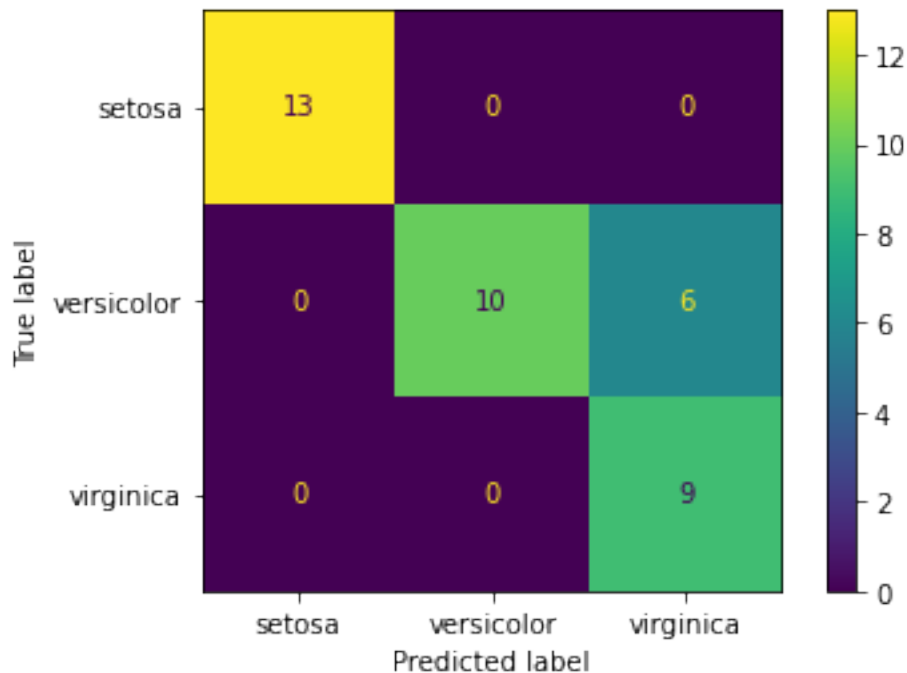


```
[ 0  0  9]]
```

1.5.1 CM Display

```
[15]: from sklearn.metrics import ConfusionMatrixDisplay
disp = ConfusionMatrixDisplay(confusion_matrix=cm,
display_labels=classifier.classes_)
disp.plot()
```

```
[15]: <sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at
0x7ff610213a00>
```



1.6 Accuracy

1.6.1 Test

```
[16]: from sklearn.metrics import accuracy_score
print('Test accuracy score with criterion entropy: {0:0.4f}'.format(
    accuracy_score(y_test,y_pred)))
```

```
Test accuracy score with criterion entropy: 0.8421
```

1.6.2 Train

```
[17]: y_pred_train= classifier.predict(x_train)
      print('Training-set accuracy score: {0:0.4f}'.format(accuracy_score(y_train,
      ↪y_pred_train)))
```

Training-set accuracy score: 1.0000