```
from sklearn import preprocessing
from sklearn.tree import DecisionTreeClassifier
Outlook = ['Rainy', 'Rainy', 'Overcast', 'Sunny', 'Sunny', 'Sunny', 'Overcast',
            'Rainy', 'Rainy', 'Sunny', 'Rainy','Overcast', 'Overcast', 'Sunny']
Temperature = ['Hot', 'Hot', 'Hot', 'Mild', 'Cool', 'Cool', 'Cool',
                'Mild', 'Cool', 'Mild', 'Mild', 'Hot', 'Mild']
Humidity = ['High', 'High', 'High', 'Normal', 'Normal', 'Normal',
            'High', 'Normal', 'Normal', 'High', 'Normal', 'High']
Wind = ['False', 'True', 'False', 'False', 'True', 'True',
            'False', 'False', 'True', 'True', 'False', 'True']
Play = ['No', 'No', 'Yes', 'Yes', 'No', 'Yes', 'No',
'Yes', 'Yes', 'Yes', 'Yes', 'No']
le = preprocessing.LabelEncoder()
Outlook_encoded = le.fit_transform(Outlook)
Outlook name mapping = dict(zip(le.classes , le.transform(le.classes )))
print("Outllok mapping:",Outlook_name_mapping)
Temperature encoded = le.fit transform(Temperature)
Temperature name mapping = dict(zip(le.classes , le.transform(le.classes )))
print("Temperature mapping:",Temperature_name_mapping)
Humidity_encoded = le.fit_transform(Humidity)
Humidity name mapping = dict(zip(le.classes , le.transform(le.classes )))
print("Humidity mapping:",Humidity name mapping)
Wind encoded = le.fit transform(Wind)
Wind_name_mapping = dict(zip(le.classes_, le.transform(le.classes_)))
print("Wind mapping:",Wind_name_mapping)
Play_encoded = le.fit_transform(Play)
Play_name_mapping = dict(zip(le.classes_, le.transform(le.classes_)))
print("Play mapping:",Play_name_mapping)
print("\n\n")
print("Weather:" ,Outlook_encoded)
print("Temerature:" ,Temperature_encoded)
print("Humidity:" ,Humidity_encoded)
print("Wind:" ,Wind_encoded)
print("Play:" ,Play_encoded)
```

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Outllok mapping: {'Overcast': 0, 'Rainy': 1, 'Sunny': 2}
     Temperature mapping: {'Cool': 0, 'Hot': 1, 'Mild': 2}
     Humidity mapping: {'High': 0, 'Normal': 1}
     Wind mapping: {'False': 0, 'True': 1}
     Play mapping: {'No': 0, 'Yes': 1}
features=tuple(zip(Outlook_encoded, Temperature_encoded, Humidity_encoded, Wind_encoded))
print("Features: ",features)
Features: ((1, 1, 0, 0), (1, 1, 0, 1), (0, 1, 0, 0), (2, 2, 0, 0), (2, 0, 1, 0), (2, 0, 1, 1), ((
from sklearn.tree import DecisionTreeClassifier
clf = DecisionTreeClassifier(criterion="entropy")
from sklearn.model_selection import train_test_split
data_train, data_test, target_train, target_test = train_test_split(features,
                        Play encoded, test size = 0.05, random state =142)
clf.fit(data_train,target_train)
print(clf.score(data_train,target_train))
predicted = clf.predict(data_test)
print(predicted)
[→ 1.0
     [1]
from sklearn import metrics
print("Accuracy:",metrics.accuracy_score(target_test, predicted))
from sklearn.metrics import confusion_matrix
confusion_matrix(target_test, predicted)

← Accuracy: 1.0

     array([[1]])
from sklearn.metrics import precision_score
from sklearn.metrics import recall_score
precision = precision_score(target_test, predicted, average='macro')
recall = recall_score(target_test, predicted, average='macro')
print('precision: {}'.format(precision))
print('recall: {}'.format(recall))

¬ precision: 1.0

     recall: 1.0
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from sklearn import tree

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tree.plot_tree(clf,filled=True,class_names=['Class_0','Class_1','Class_2'])
                     [Text(111.60000000000001, 195.696, 'X[0] <= 0.5\nentropy = 0.961\nsamples = 13\nvalue = [5, 8]\nc.
                        Text(74.4, 152.208, 'entropy = 0.0 \times 10^{-2} = 3 \times 10^{-2} Text(74.4, 152.208, 'entropy = 0.0 \times 10^{-2} = 0.0 \times 10^{-2} Text(74.4, 152.208, 'entropy = 0.0 \times 10^{-2} = 0.0 \times 10^{-2} Text(74.4, 152.208, 'entropy = 0.0 \times 10^{-2} = 0.0 \times 10^{-2} Text(74.4, 152.208, 'entropy = 0.0 \times 10^{-2} T
                         Text(148.8, 152.208, 'X[2] <= 0.5\nentropy = 1.0\nsamples = 10\nvalue = [5, 5]\nclass = Class_0'</pre>
                         Text(74.4, 108.72, 'X[0] <= 1.5\nentropy = 0.722\nsamples = 5\nvalue = [4, 1]\nclass = Class 0')</pre>
                         Text(37.2, 65.232, 'entropy = 0.0\nsamples = 3\nvalue = [3, 0]\nclass = Class_0'),
                         Text(111.6000000000001, 65.232, 'X[3] \le 0.5 \le 1.0 \le 2 \le 2 \le 1.0 
                         Text(74.4, 21.744, 'entropy = 0.0 \nsamples = 1 \nvalue = [0, 1] \nclass = Class_1'),
                         Text(148.8, 21.744, 'entropy = 0.0 \times 1 = 1 \times 1 = [1, 0] \times 1 = [1, 0]
                         Text(223.2000000000000, 108.72, X[3] \le 0.5 = 0.722 = 0.722 = 5 = 5 = 1, 4] = 1, 4
                         Text(186.0, 65.232, 'entropy = 0.0\nsamples = 3\nvalue = [0, 3]\nclass = Class_1'),
                         Text(260.40000000000003, 65.232, 'X[1] <= 1.0\nentropy = 1.0\nsamples = 2\nvalue = [1, 1]\nclass
                         Text(223.2000000000000, 21.744, 'entropy = 0.0\nsamples = 1\nvalue = [1, 0]\nclass = Class_0'),
                         Text(297.6, 21.744, 'entropy = 0.0\nsamples = 1\nvalue = [0, 1]\nclass = Class_1')]
from sklearn.tree import export graphviz
export_graphviz(clf,out_file='tree_entropy.dot',
                                                                   feature names=['outlook','temperature','humidity','wind'],
                                                                   class_names=['play_no','play_yes'],
                                                                  filled=True)
from subprocess import call
```

call(['dot', '-Tpng', 'tree entropy.dot', '-o', 'tree entropy.png', '-Gdpi=600'])

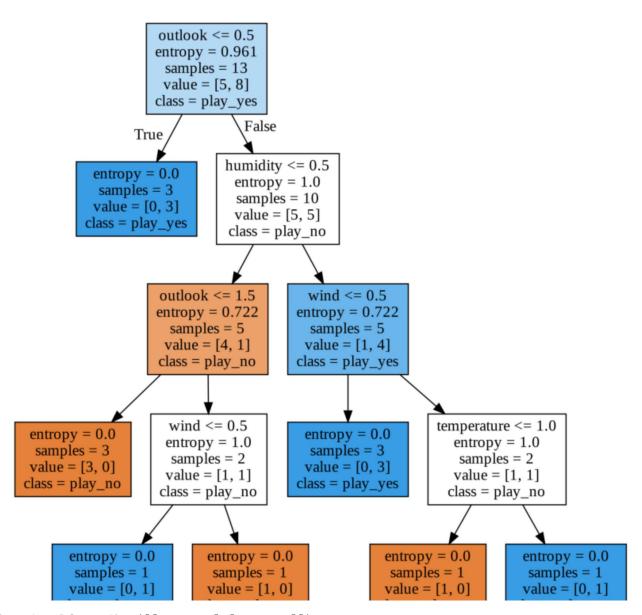
import matplotlib.pyplot as plt
plt.figure(figsize = (14, 18))

plt.axis('off');

plt.show();

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plt.imshow(plt.imread('tree_entropy.png'))



predicted= clf.predict([[1,2,1,0],[2,0,0,1]])
print("Predicted Value:", predicted)

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