```
In [38]: import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
         from sklearn import datasets
         from sklearn import tree
         import warnings
         warnings.filterwarnings('ignore')
In [63]: | iris=datasets.load_iris()
         print(iris.target_names)
         print("data set str: ",dir(iris))
         ['setosa' 'versicolor' 'virginica']
         data set str: ['DESCR', 'data', 'data_module', 'feature_names', 'filename',
         'frame', 'target', 'target_names']
In [60]: | df=pd.DataFrame(iris.data,columns=iris.feature_names)
         df
         df["target"]=iris.target
         df.target.unique()
         df['flower_species']=df.target.apply(lambda x : iris.target_names[x])
         print("unique target values=",df['target'].unique())
```

unique target values= [0 1 2]

df.sample(5)

## Out[60]:

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target	flower_species
140	6.7	3.1	5.6	2.4	2	virginica
116	6.5	3.0	5.5	1.8	2	virginica
85	6.0	3.4	4.5	1.6	1	versicolor
40	5.0	3.5	1.3	0.3	0	setosa
147	6.5	3.0	5.2	2.0	2	virginica

```
In [4]: #df = pd.DataFrame(X, columns=iris.feature_names)
    #df['species'] = iris.target
    #df['species'] = df['species'].replace(to_replace= [0, 1, 2], value = ['setosa
```

In	[80]	:

Out[80]:

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target
0	5.1	3.5	1.4	0.2	0
1	4.9	3.0	1.4	0.2	0
2	4.7	3.2	1.3	0.2	0
3	4.6	3.1	1.5	0.2	0
4	5.0	3.6	1.4	0.2	0
145	6.7	3.0	5.2	2.3	2
146	6.3	2.5	5.0	1.9	2
147	6.5	3.0	5.2	2.0	2
148	6.2	3.4	5.4	2.3	2
149	5.9	3.0	5.1	1.8	2

150 rows × 5 columns

## 0=setosa,1=versicolor,2=virginica

In [5]: df[df.target==0].head(3)

Out[5]:

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target	flower_species
0	5.1	3.5	1.4	0.2	0	setosa
1	4.9	3.0	1.4	0.2	0	setosa
2	4.7	3.2	1.3	0.2	0	setosa

In [6]: df[df.target==1].head(3)

Out[6]:

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target	flower_species
50	7.0	3.2	4.7	1.4	1	versicolor
51	6.4	3.2	4.5	1.5	1	versicolor
52	6.9	3.1	4.9	1.5	1	versicolor

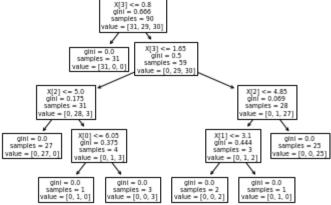
```
df[df.target==2].head(3)
 In [7]:
 Out[7]:
                   sepal length
                                   sepal width
                                                  petal length
                                                                 petal width
                                                                           target flower_species
                          (cm)
                                        (cm)
                                                        (cm)
                                                                      (cm)
           100
                           6.3
                                                                               2
                                          3.3
                                                         6.0
                                                                       2.5
                                                                                        virginica
           101
                                                                               2
                           5.8
                                          2.7
                                                         5.1
                                                                       1.9
                                                                                        virginica
           102
                           7.1
                                          3.0
                                                         5.9
                                                                       2.1
                                                                               2
                                                                                        virginica
 In [8]:
          df.columns
 Out[8]: Index(['sepal length (cm)', 'sepal width (cm)', 'petal length (cm)',
                  'petal width (cm)', 'target', 'flower_species'],
                 dtype='object')
In [12]: x=df[['sepal length (cm)', 'sepal width (cm)', 'petal length (cm)',
                  'petal width (cm)']]
 In [9]: y=df['target']
In [38]: x.shape
Out[38]: (150, 4)
In [39]: | y.shape
Out[39]: (150,)
In [13]: | from sklearn.model_selection import train_test_split
In [14]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.40,random_state
```

it[15]:	sepal	length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	
	11	4.8	3.4	1.6	0.2	
	113	5.7	2.5	5.0	2.0	
	123	6.3	2.7	4.9	1.8	
	12	4.8	3.0	1.4	0.1	
	2	4.7	3.2	1.3	0.2	
	133	6.3	2.8	5.1	1.5	
	137	6.4	3.1	5.5	1.8	
	72	6.3	2.5	4.9	1.5	
	140	6.7	3.1	5.6	2.4	
	37	4.9	3.6	1.4	0.1	
	90 rows × 4	columns				
n [16]:	y_train.sh	nape				
ut[16]:	(90,)					
1 [15]:	x_test.sha	ipe				
ut[15]:	(60, 4)					
n [16]:	y_test.sha	ipe				
ut[16]:	(60,)					
in [17]:	<pre>from sklearn.tree import DecisionTreeClassifier dtc= tree.DecisionTreeClassifier(random_state=1) dtc.fit(x_train,y_train)</pre>					
out[17]:	DecisionTr	reeClassi	fier(random_st	ate=1)		
[112]:						
[112]:	DecisionTr	reeClassi	fier(random_st	ate=1)		
n [18]:	y_pred=dto y_pred	.predict(	(x_test)			
ut[18]:	2,	0, 2, 1,	0, 0, 1, 2, 1		1, 1, 0, 1, 1, 0, 1, 0, 1, 2,	

```
In [ ]: #checking the predicted x_test with y_test
         a=iris.target_names[y_test.iloc[12]]
         b=iris.target_names[dtc.predict([x_test.iloc[12]])]
         print("iris y_test:" ,a)
         print("iris y_pred:", b)
In [20]: from sklearn.metrics import classification_report
         classification_report(y_test,y_pred)
Out[20]:
                        precision
                                     recall f1-score
                                                        support\n\n
                                                                              0
                   1.00
                            1.00
                                                                          0.95
                                                                                    0.
         1.00
                                         19∖n
                                                                0.95
         95
                   21\n
                                  2
                                          0.95
                                                    0.95
                                                              0.95
                                                                          20\n\n
                                                                                    ac
                                          0.97
                                                      60\n
                                                             macro avg
                                                                             0.97
         curacy
         0.97
                   0.97
                               60\nweighted avg
                                                      0.97
                                                                0.97
                                                                          0.97
         60\n'
In [21]: from sklearn.metrics import confusion_matrix
         confusion_matrix(y_test,y_pred)
Out[21]: array([[19, 0, 0],
                [ 0, 20, 1],
                [ 0, 1, 19]], dtype=int64)
In [22]: from sklearn import metrics
         accuracy=metrics.accuracy_score(y_test,y_pred)*100
         accuracy
Out[22]: 96.6666666666667
In [23]: #dtc.score(x_test,y_test)
Out[23]: 0.966666666666667
```

. .

```
In [24]: tree.plot_tree(dtc)
Out[24]: [Text(0.4, 0.9, 'X[3] \le 0.8 \text{ ngini} = 0.666 \text{ nsamples} = 90 \text{ nvalue} = [31, 29, 3]
                              0]'),
                                 Text(0.3, 0.7, 'gini = 0.0\nsamples = 31\nvalue = [31, 0, 0]'),
                                 Text(0.5, 0.7, 'X[3] \le 1.65 \text{ ngini} = 0.5 \text{ nsamples} = 59 \text{ nvalue} = [0, 29, 3]
                              0]'),
                                Text(0.2, 0.5, 'X[2] \le 5.0 \text{ ngini} = 0.175 \text{ nsamples} = 31 \text{ nvalue} = [0, 28, 1.5]
                              3]'),
                                 Text(0.1, 0.3, 'gini = 0.0 \land samples = 27 \land value = [0, 27, 0]'),
                                 Text(0.3, 0.3, X[0] <= 6.05 = 0.375 = 4 = 4
                              3]'),
                                 Text(0.2, 0.1, 'gini = 0.0 \land samples = 1 \land u = [0, 1, 0]'),
                                 Text(0.4, 0.1, 'gini = 0.0\nsamples = 3\nvalue = [0, 0, 3]'),
                                Text(0.8, 0.5, X[2] \le 4.85 = 0.069 = 28 = 28 = [0, 1, 2]
                              7]'),
                                 Text(0.7, 0.3, X[1] <= 3.1 = 0.444 = 3 = 3 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 = 0.444 =
                                 Text(0.6, 0.1, 'gini = 0.0 \land samples = 2 \land u = [0, 0, 2]'),
                                 Text(0.8, 0.1, 'gini = 0.0\nsamples = 1\nvalue = [0, 1, 0]'),
                                 Text(0.9, 0.3, 'gini = 0.0\nsamples = 25\nvalue = [0, 0, 25]')]
                                                                                 X[3] <= 0.8
gini = 0.666
samples = 90
                                                                               samples = 30
ilue = [31, 29, 30]
```



```
In [25]: import graphviz
```

Out[42]: <bound method Render.render of <graphviz.sources.Source object at 0x0000020CC DBACB50>>

In [65]: view=tree.export\_graphviz(dtc,out\_file=None,feature\_names=iris.feature\_names, class\_names=iris.target\_names,rounded=True,filled=T graph=graphviz.Source(view) graph Out[65]: petal width (cm) <= 0.8 gini = 0.666samples = 90 value = [31, 29, 30]class = setosa False True petal width (cm) <= 1.65 gini = 0.0gini = 0.5samples = 31 samples = 59 value = [31, 0, 0]value = [0, 29, 30]class = setosa class = virginica petal length (cm) <= 5.0 petal length (cm) < gini = 0.175gini = 0.069samples = 31 samples = 28 value = [0, 28, 3]value = [0, 1, 2 class = versicolor class = virginio sepal length (cm) <= 6.05 sepal width (cm) < gini = 0.0gini = 0.375gini = 0.444samples = 27 samples = 3samples = 4 value = [0, 27, 0]value = [0, 1, 3]value = [0, 1, 2]class = versicolor class = virginica class = virginic gini = 0.0gini = 0.0gini = 0.0samples = 1 samples = 3samples = 2value = [0, 1, 0] value = [0, 0, 3]value = [0, 0, 2]class = virginica class = virginica class = versicolor

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