```
In [168]: import numpy as np
    import pandas as pd
    from sklearn import datasets
    from matplotlib import pyplot as plt
    from sklearn.preprocessing import LabelEncoder
    from sklearn import tree
    import six
    import sys
    sys.modules['sklearn.externals.six'] = six
    from id3 import Id3Estimator, export_text
    from sklearn.model_selection import train_test_split, cross_val_score
    from sklearn.metrics import accuracy_score, classification_report
    from sklearn.datasets import load_iris
```

```
In [149]:    iris = datasets.load_iris()
    X = iris.data
    y = iris.target
```

```
In [150]: df = pd.read_csv('PredictingSignupsTrain.csv')
    df
```

Out[150]:

	Referrer	Location	Read_FAQ	Pages_Viewed	Service_Chosen
0	Slashdot	USA	Yes	Mid	None
1	Google	France	Yes	High	Premium
2	Digg	USA	Yes	High	Basic
3	Kiwitobes	France	Yes	High	Basic
4	Google	UK	No	Mid	Premium
5	(direct)	'New Zealand'	No	Low	None
6	(direct)	UK	No	Mid	Basic
7	Google	USA	No	High	Premium
8	Slashdot	France	Yes	Mid	None
9	Digg	USA	No	Mid	None
10	Google	UK	No	Mid	None
11	Kiwitobes	UK	No	Mid	None
12	Digg	'New Zealand'	Yes	Low	Basic
13	Google	UK	Yes	Mid	Basic
14	Kiwitobes	France	Yes	Mid	Basic
15	Google	UK	No	Mid	Premium
16	Digg	USA	No	Low	Basic
17	Slashdot	'New Zealand'	Yes	High	None

```
In [151]: lb_enc = LabelEncoder()
    df['Referrer'] = lb_enc.fit_transform(df['Referrer'])
    df['Location'] = lb_enc.fit_transform(df['Location'])
    df['Read_FAQ'] = lb_enc.fit_transform(df['Read_FAQ'])
    df['Pages_Viewed'] = lb_enc.fit_transform(df['Pages_Viewed'])
    df['Service_Chosen'] = lb_enc.fit_transform(df['Service_Chosen'])
    df.to_csv("PredictingSignupsTrain_label.csv")
    df
```

Out[151]:

	Referrer	Location	Read_FAQ	Pages_Viewed	Service_Chosen
0	4	3	1	2	1
1	2	1	1	0	2
2	1	3	1	0	0
3	3	1	1	0	0
4	2	2	0	2	2
5	0	0	0	1	1
6	0	2	0	2	0
7	2	3	0	0	2
8	4	1	1	2	1
9	1	3	0	2	1
10	2	2	0	2	1
11	3	2	0	2	1
12	1	0	1	1	0
13	2	2	1	2	0
14	3	1	1	2	0
15	2	2	0	2	2
16	1	3	0	1	0
17	4	0	1	0	1

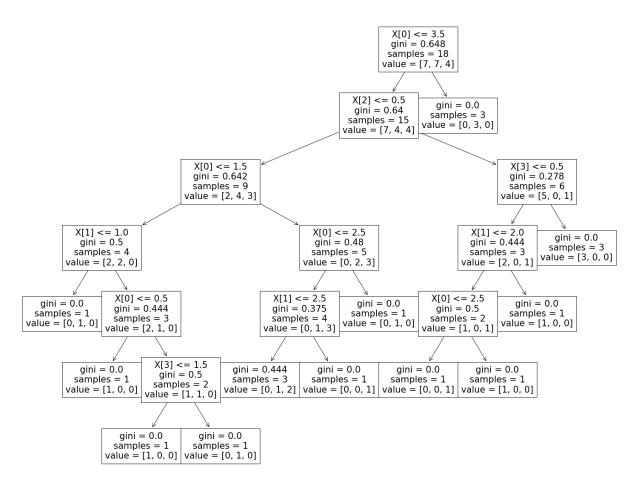
```
In [152]: %time
    clf = tree.DecisionTreeClassifier()
    X = df[['Referrer', 'Location', 'Read_FAQ', 'Pages_Viewed']]
    y = df['Service_Chosen']
    clf = clf.fit(X, y)
    y_pred = clf.predict(X)
    y_pred
```

Wall time: 0 ns

Out[152]: array([1, 2, 0, 0, 2, 1, 0, 2, 1, 1, 2, 1, 0, 0, 0, 2, 0, 1])

Wall time: 0 ns

```
Out[153]: [Text(930.0, 1009.5428571428572, 'X[0] <= 3.5\ngini = 0.648\nsamples = 18\nva
                                                                     lue = [7, 7, 4]'),
                                                                           Text(837.0, 854.2285714285715, X[2] \le 0.5  o.5\ngini = 0.64\nsamples = 15\nvalu
                                                                     e = [7, 4, 4]'),
                                                                          Text(465.0, 698.9142857142858, 'X[0] <= 1.5 \setminus i = 0.642 \setminus i = 0.
                                                                     e = [2, 4, 3]'),
                                                                           Text(186.0, 543.6, 'X[1] \le 1.0 \le 0.5 \le 4 \le 4 \le 1.0 \le
                                                                     0]'),
                                                                           Text(93.0, 388.28571428571433, 'gini = 0.0\nsamples = 1\nvalue = [0, 1,
                                                                     0]'),
                                                                           Text(279.0, 388.28571428571433, 'X[0] <= 0.5\ngini = 0.444\nsamples = 3\nval
                                                                     ue = [2, 1, 0]'),
                                                                           Text(186.0, 232.97142857142865, 'gini = 0.0\nsamples = 1\nvalue = [1, 0, 1]
                                                                     0]'),
                                                                           Text(372.0, 232.97142857142865, 'X[3] \le 1.5 \le 0.5 \le 
                                                                     = [1, 1, 0]'),
                                                                           Text(279.0, 77.65714285714284, 'gini = 0.0 \times 1 = 1 \times 1 | 1, 0,
                                                                     0]'),
                                                                           Text(465.0, 77.65714285714284, 'gini = 0.0\nsamples = 1\nvalue = [0, 1, 1]
                                                                     0]'),
                                                                           Text(744.0, 543.6, 'X[0] \le 2.5 \ngini = 0.48\nsamples = 5\nvalue = [0, 2,
                                                                     3]'),
                                                                           Text(651.0, 388.28571428571433, X[1] <= 2.5 = 0.375 = 4 
                                                                     ue = [0, 1, 3]),
                                                                           Text(558.0, 232.97142857142865, 'gini = 0.444\nsamples = 3\nvalue = [0, 1, 1]
                                                                           Text(744.0, 232.97142857142865, 'gini = 0.0\nsamples = 1\nvalue = [0, 0, 0]
                                                                     1]'),
                                                                           Text(837.0, 388.28571428571433, 'gini = 0.0\nsamples = 1\nvalue = [0, 1, 1]
                                                                     0]'),
                                                                           Text(1209.0, 698.9142857142858, 'X[3] <= 0.5 \text{ ngini} = 0.278 \text{ nsamples} = 6 \text{ nval}
                                                                     ue = [5, 0, 1]'),
                                                                           Text(1116.0, 543.6, 'X[1] \le 2.0 \text{ ngini} = 0.444 \text{ nsamples} = 3 \text{ nvalue} = [2, 0, 1]
                                                                     1]'),
                                                                           Text(1023.0, 388.28571428571433, 'X[0] <= 2.5\ngini = 0.5\nsamples = 2\nvalu
                                                                     e = [1, 0, 1]'),
                                                                           Text(930.0, 232.97142857142865, 'gini = 0.0\nsamples = 1\nvalue = [0, 0, 0]
                                                                     1]'),
                                                                           Text(1116.0, 232.97142857142865, 'gini = 0.0 \times 10^{-1} = 1 \times 10^{-1} Text(1116.0, 232.97142857142865, 'gini = 0.0 \times 10^{-1} = 1 \times 10^{-1} Text(1116.0, 232.97142857142865, 'gini = 0.0 \times 10^{-1} Text(1116.0, 232.97142865, 'gini = 0.0 \times 10^{-1} Text(1116.0, 232.9714286, 'gini = 0.0 \times 10^{-1} Text(1116.0, 'g
                                                                     0]'),
                                                                           Text(1209.0, 388.28571428571433, 'gini = 0.0\nsamples = 1\nvalue = [1, 0,
                                                                     0]'),
                                                                           Text(1302.0, 543.6, 'gini = 0.0 \times 10^{-1}),
                                                                           Text(1023.0, 854.2285714285715, 'gini = 0.0\nsamples = 3\nvalue = [0, 3, 1]
                                                                     0]')]
```



Accuracy: 0.9444444444444444

Confusion:

[[7 0 0]

[0 6 1]

[0 0 4]]

Classification Report:

	precision	recall	f1-score	support
None	1.00	1.00	1.00	7
Basic	1.00	0.86	0.92	7
Premium	0.80	1.00	0.89	4
accuracy			0.94	18
macro avg	0.93	0.95	0.94	18
weighted avg	0.96	0.94	0.95	18

```
In [155]: df_test = pd.read_csv('PredictingSignupsTest.csv')
    df_test
```

Out[155]:

Referrer Lo		Location	cation Read_FAQ Pages_Viewed		Service_Chosen	
0	Google	UK	No	Mid	Premium	
1	Digg	USA	No	Low	Basic	
2	Slashdot	'New Zealand'	Yes	High	None	

```
In [156]: lb_enc = LabelEncoder()
    df_test['Referrer'] = lb_enc.fit_transform(df_test['Referrer'])
    df_test['Location'] = lb_enc.fit_transform(df_test['Location'])
    df_test['Read_FAQ'] = lb_enc.fit_transform(df_test['Read_FAQ'])
    df_test['Pages_Viewed'] = lb_enc.fit_transform(df_test['Pages_Viewed'])
    df_test['Service_Chosen'] = lb_enc.fit_transform(df_test['Service_Chosen'])
    df_test
```

Out[156]:

	Referrer	Location	Read_FAQ	Pages_Viewed	Service_Chosen
0	1	1	0	2	2
1	0	2	0	1	0
2	2	0	1	0	1

```
In [157]: %time
    clf = tree.DecisionTreeClassifier()
    X = df_test[['Referrer', 'Location', 'Read_FAQ', 'Pages_Viewed']]
    y = df_test['Service_Chosen']
    clf = clf.fit(X, y)
    y_pred = clf.predict(X)
    y_pred
```

Wall time: 0 ns

Out[157]: array([2, 0, 1])

```
In [158]: | %time
                                          fig = plt.figure(figsize=(25,20))
                                           _ = tree.plot_tree(clf, filled=True)
                                          fig.savefig("decistion tree.png")
                                          tree.plot_tree(clf)
                                         Wall time: 0 ns
Out[158]: [Text(558.0, 906.0, 'X[0] <= 0.5\ngini = 0.667\nsamples = 3\nvalue = [1, 1,</pre>
                                         1]'),
                                            Text(279.0, 543.6, 'gini = 0.0 \times 10^{-1}),
                                            Text(837.0, 543.6, 'X[1] \le 0.5 \le 0
                                            Text(558.0, 181.199999999999, 'gini = 0.0 \times 1 | Text(558.0, 181.199999999999, 'gini = 0.0 \times 1 | Text(558.0, 181.199999999999, 'gini = 0.0 \times 1 | Text(558.0, 181.19999999999)
                                         0]'),
                                            Text(1116.0, 181.199999999999, 'gini = 0.0\nsamples = 1\nvalue = [0, 0, 0]
                                         1]')]
                                                                                                                           X[0] <= 0.5
                                                                                                                           gini = 0.667
                                                                                                                          samples = 3
                                                                                                             value = [1, 1, 1]
                                                                                                                                                                                         X[1] <= 0.5
                                                                      gini = 0.0
                                                                                                                                                                                          gini = 0.5
                                                             samples = 1
                                                                                                                                                                                        samples = 2
                                                value = [1, 0, 0]
                                                                                                                                                                           value = [0, 1, 1]
                                                                                                                                   gini = 0.0
                                                                                                                                                                                                                                                              gini = 0.0
                                                                                                                          samples = 1
                                                                                                                                                                                                                                      samples = 1
```

```
In [159]:
          from sklearn.metrics import accuracy_score,confusion_matrix,classification_rep
           print("Accuracy: ", accuracy_score(y,y_pred))
          print("Confusion: \n", confusion_matrix(y,y_pred))
           print("Classification Report: \n", classification_report(y,y_pred,target_names
           = ['None', 'Basic', 'Premium']))
          Accuracy: 1.0
          Confusion:
           [[1 0 0]
           [0 1 0]
           [0 0 1]]
          Classification Report:
                          precision
                                       recall f1-score
                                                           support
                              1.00
                                        1.00
                                                   1.00
                                                                1
                  None
                 Basic
                              1.00
                                        1.00
                                                   1.00
                                                                1
               Premium
                              1.00
                                        1.00
                                                   1.00
                                                                1
                                                                3
              accuracy
                                                   1.00
                              1.00
                                        1.00
                                                   1.00
                                                                3
             macro avg
          weighted avg
                              1.00
                                                   1.00
                                                                3
                                        1.00
```

```
In [160]: X.shape, y.shape
```

Out[160]: ((3, 4), (3,))