

# ML7

May 23, 2022

```
[2]: # Apply Decision Tree Classification technique to solve given problem:
# A dataset collected in a cosmetics shop showing details of
    ↳ customers and whether or not they responded to a special offer to
    ↳ buy a new lip-stick is shown in table below.
# Use this dataset to build a decision tree, with Buys as the target variable,
    ↳ to help in buying lip-sticks in the future. Find the root node of
    ↳ decision tree.
# According to the decision tree you have made from previous training
    ↳ data set, what is the decision for the test data:
# [Age < 21, Income = Low, Gender = Female, Marital Status = Married]?
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```
[35]: import numpy as np
import pandas as pd
from sklearn.preprocessing import LabelEncoder
from sklearn.tree import DecisionTreeClassifier
from sklearn.tree import plot_tree
```

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[36]: data = pd.read_csv("ML7.csv")
data
```

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[36]:
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	Age	Income	Gender	MaritalStatus	Buys
0	<21	High	Male	Single	No
1	<21	High	Male	Married	No
2	21-35	High	Male	Single	Yes
3	>35	Medium	Male	Single	Yes
4	>35	Low	Female	Single	Yes
5	>35	Low	Female	Married	No
6	21-35	Low	Female	Married	Yes
7	<21	Medium	Male	Single	No
8	<21	Low	Female	Married	Yes
9	>35	Medium	Female	Single	Yes
10	<21	Medium	Female	Married	Yes
11	21-35	Medium	Male	Married	Yes
12	21-35	High	Female	Single	Yes

```
[37]: le = LabelEncoder()
x = data.iloc[:, :-1]
x = x.apply(le.fit_transform)
```

```
[38]: print("Age: ",list(zip(data.iloc[:,0],x.iloc[:,0])))
      print("Income: ",list(zip(data.iloc[:,1],x.iloc[:,1])))
      print("Gender: ",list(zip(data.iloc[:,2],x.iloc[:,2])))
      print("MaritalStatus: ",list(zip(data.iloc[:,3],x.iloc[:,3])))
```

```
Age:  [('<21', 1), ('<21', 1), ('21-35', 0), ('>35', 2), ('>35', 2), ('>35', 2),
      ('21-35', 0), ('<21', 1), ('<21', 1), ('>35', 2), ('<21', 1), ('21-35', 0),
      ('21-35', 0)]
Income:  [('High', 0), ('High', 0), ('High', 0), ('Medium', 2), ('Low', 1),
      ('Low', 1), ('Low', 1), ('Medium', 2), ('Low', 1), ('Medium', 2), ('Medium', 2),
      ('Medium', 2), ('High', 0)]
Gender:  [('Male', 1), ('Male', 1), ('Male', 1), ('Male', 1), ('Female', 0),
      ('Female', 0), ('Female', 0), ('Male', 1), ('Female', 0), ('Female', 0),
      ('Female', 0), ('Male', 1), ('Female', 0)]
MaritalStatus:  [('Single', 1), ('Married', 0), ('Single', 1), ('Single', 1),
      ('Single', 1), ('Married', 0), ('Married', 0), ('Single', 1), ('Married', 0),
      ('Single', 1), ('Married', 0), ('Married', 0), ('Single', 1)]
```

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[39]: x
```

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[39]:
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	Age	Income	Gender	MaritalStatus
0	1	0	1	1
1	1	0	1	0
2	0	0	1	1
3	2	2	1	1
4	2	1	0	1
5	2	1	0	0
6	0	1	0	0
7	1	2	1	1
8	1	1	0	0
9	2	2	0	1
10	1	2	0	0
11	0	2	1	0
12	0	0	0	1

```
[40]: y = data.iloc[:,-1]
      y
```

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[40]:
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0	No
1	No
2	Yes
3	Yes
4	Yes
5	No
6	Yes
7	No
8	Yes
9	Yes

```
10    Yes
11    Yes
12    Yes
Name: Buys, dtype: object
```

```
[41]: dt = DecisionTreeClassifier()
      dt.fit(x,y)
```

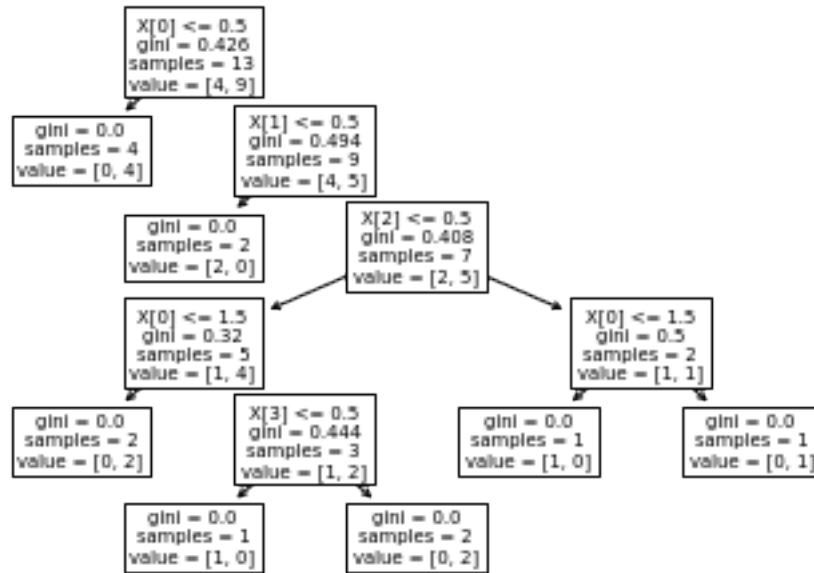
```
[41]: DecisionTreeClassifier()
```

```
[42]: #[Age < 21, Income = Low, Gender = Female, Marital Status = Married]
      query = np.array([1,1,0,0])
      pred = dt.predict([query])
      pred
```

```
[42]: array(['Yes'], dtype=object)
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[43]: plot_tree(dt)
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```
[43]: [Text(83.7, 199.32, 'X[0] <= 0.5\ngini = 0.426\nsamples = 13\nvalue = [4, 9]'),
      Text(41.85, 163.07999999999998, 'gini = 0.0\nsamples = 4\nvalue = [0, 4]'),
      Text(125.55000000000001, 163.07999999999998, 'X[1] <= 0.5\ngini =
0.494\nsamples = 9\nvalue = [4, 5]'),
      Text(83.7, 126.83999999999999, 'gini = 0.0\nsamples = 2\nvalue = [2, 0]'),
      Text(167.4, 126.83999999999999, 'X[2] <= 0.5\ngini = 0.408\nsamples = 7\nvalue
= [2, 5]'),
      Text(83.7, 90.6, 'X[0] <= 1.5\ngini = 0.32\nsamples = 5\nvalue = [1, 4]'),
      Text(41.85, 54.359999999999985, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
      Text(125.55000000000001, 54.359999999999985, 'X[3] <= 0.5\ngini =
0.444\nsamples = 3\nvalue = [1, 2]'),
      Text(83.7, 18.119999999999976, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
      Text(167.4, 18.119999999999976, 'gini = 0.0\nsamples = 2\nvalue = [0, 2]'),
      Text(251.10000000000002, 90.6, 'X[0] <= 1.5\ngini = 0.5\nsamples = 2\nvalue =
[1, 1]'),
      Text(209.25, 54.359999999999985, 'gini = 0.0\nsamples = 1\nvalue = [1, 0]'),
      Text(292.95, 54.359999999999985, 'gini = 0.0\nsamples = 1\nvalue = [0, 1]')]
```



[ ]: