

# Decision\_Tree

October 14, 2021

```
[54]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn import tree
import seaborn as sns
sns.set(style="white")
sns.set(style="whitegrid", color_codes=True)
from sklearn import metrics
from sklearn.metrics import classification_report, confusion_matrix
```

```
[55]: # load the SANK dataset:
sank_data=pd.read_csv("Sank.csv",header=0)
```

```
[56]: print(sank_data.shape)
print(list(sank_data.columns))
```

```
(891, 5)
['Alive', 'Class', 'Sex', 'Age', 'Fare']
```

```
[57]: sank_data
```

```
[57]:
```

	Alive	Class	Sex	Age	Fare
0	0	3	male	22.0	7.2500
1	1	1	female	38.0	71.2833
2	1	3	female	26.0	7.9250
3	1	1	female	35.0	53.1000
4	0	3	male	35.0	8.0500
..	...	...	...	...	...
886	0	2	male	27.0	13.0000
887	1	1	female	19.0	30.0000
888	0	3	female	NaN	23.4500
889	1	1	male	26.0	30.0000
890	0	3	male	32.0	7.7500

```
[891 rows x 5 columns]
```

```
[58]: sank_data = sank_data.dropna()
```

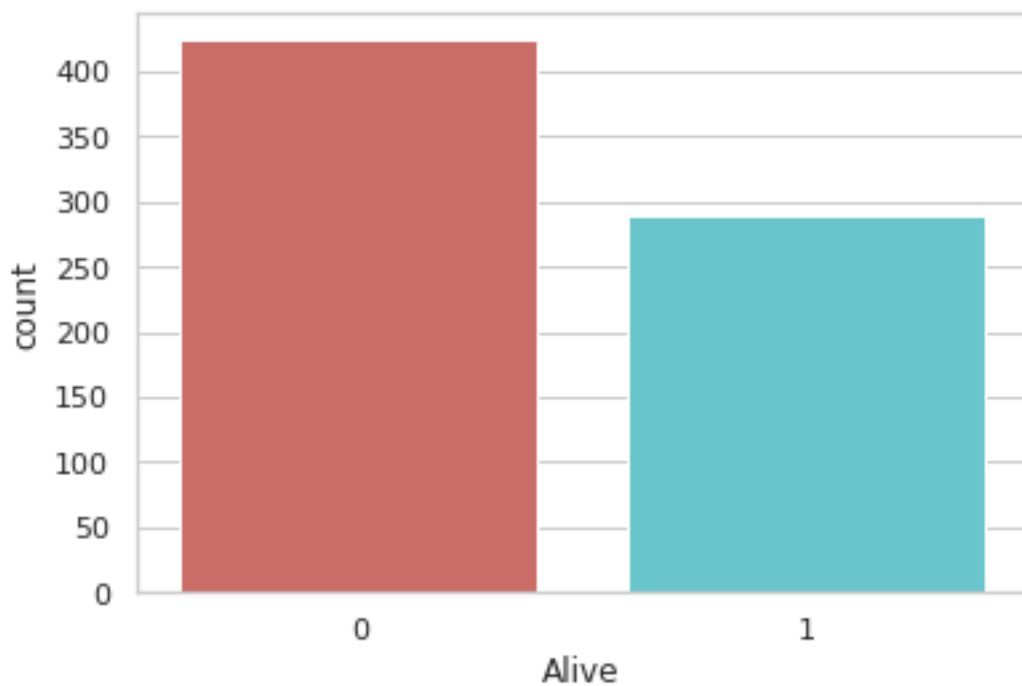
```
[59]: sank_data
```

```
[59]:
```

	Alive	Class	Sex	Age	Fare
0	0	3	male	22.0	7.2500
1	1	1	female	38.0	71.2833
2	1	3	female	26.0	7.9250
3	1	1	female	35.0	53.1000
4	0	3	male	35.0	8.0500
..	...	...	...	...	...
885	0	3	female	39.0	29.1250
886	0	2	male	27.0	13.0000
887	1	1	female	19.0	30.0000
889	1	1	male	26.0	30.0000
890	0	3	male	32.0	7.7500

[714 rows x 5 columns]

```
[60]: sns.countplot(x="Alive",data=sank_data,palette='hls')
plt.show()
```



```
[61]: X = sank_data.loc[:, sank_data.columns != 'Alive']
y = sank_data.loc[:, sank_data.columns == 'Alive']

X
```

```
[61]:
```

	Class	Sex	Age	Fare
0	3	male	22.0	7.2500
1	1	female	38.0	71.2833
2	3	female	26.0	7.9250
3	1	female	35.0	53.1000
4	3	male	35.0	8.0500
..	...	...	...	...
885	3	female	39.0	29.1250
886	2	male	27.0	13.0000
887	1	female	19.0	30.0000
889	1	male	26.0	30.0000
890	3	male	32.0	7.7500

[714 rows x 4 columns]

```
[62]: d = {'male': 1, 'female': 0}
X = X.replace({"Sex": d})
```

```
[63]: X
```

```
[63]:
```

	Class	Sex	Age	Fare
0	3	1	22.0	7.2500
1	1	0	38.0	71.2833
2	3	0	26.0	7.9250
3	1	0	35.0	53.1000
4	3	1	35.0	8.0500
..	...	...	...	...
885	3	0	39.0	29.1250
886	2	1	27.0	13.0000
887	1	0	19.0	30.0000
889	1	1	26.0	30.0000
890	3	1	32.0	7.7500

[714 rows x 4 columns]

```
[64]: X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.2,random_state=0)
```

```
[65]: dtree = tree.DecisionTreeClassifier()
dtree = dtree.fit(X_train,y_train)

y_pred=dtree.predict(X_test)
```

```
[66]: print("Model Score:",dtree.score(X_test,y_test))
```

Model Score: 0.7552447552447552

```
[67]: cnfusion_matrix = metrics.confusion_matrix(y_test, y_pred)
cnfusion_matrix
```

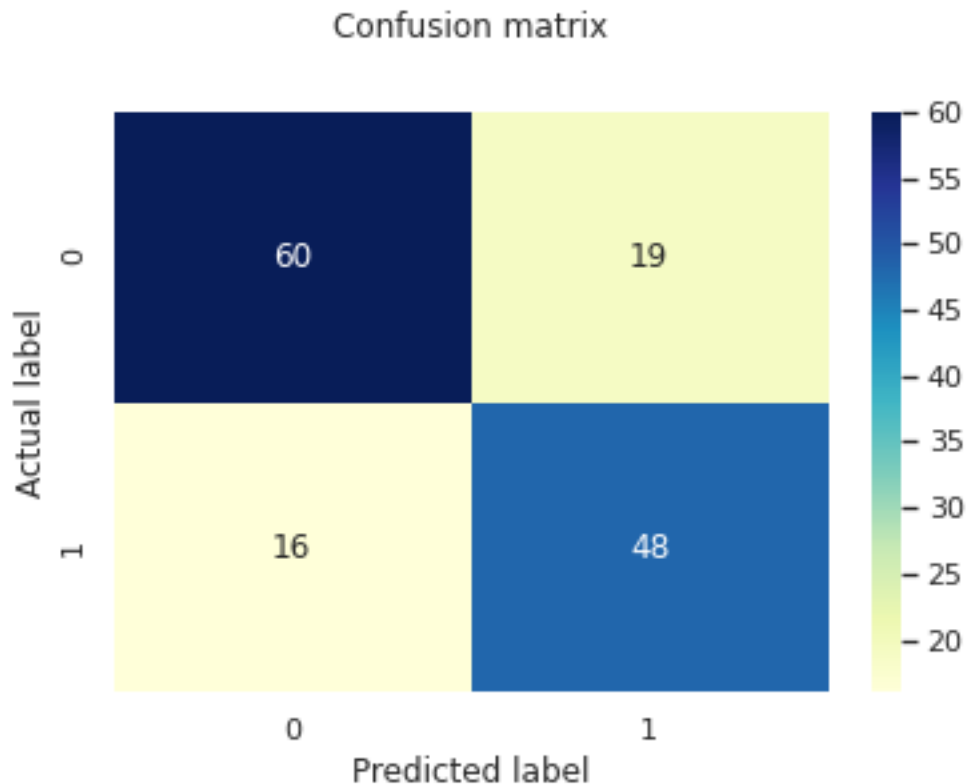
```
[67]: array([[60, 19],
        [16, 48]])
```

```
[68]: # Heat Map Visualization

# fig, ax = plt.subplots()
tick_marks = np.arange(len(sank_data.Alive))
plt.xticks(tick_marks, sank_data.Alive)
plt.yticks(tick_marks, sank_data.Alive)

sns.heatmap(pd.DataFrame(cnfusion_matrix), annot=True, cmap="YlGnBu" ,fmt='g')
plt.title('Confusion matrix', y=1.1)
plt.ylabel('Actual label')
plt.xlabel('Predicted label')
```

```
[68]: Text(0.5, 12.5, 'Predicted label')
```



```
[69]: print(classification_report(y_test, dtree.predict(X_test)))
```

	precision	recall	f1-score	support
0	0.79	0.76	0.77	79
1	0.72	0.75	0.73	64
accuracy			0.76	143
macro avg	0.75	0.75	0.75	143
weighted avg	0.76	0.76	0.76	143

```
[70]: # Male = 1, Female = 0 ; As per our defined Dictionary d = {'male': 1, 'female':
      ↪ 0}
```

```
# Class = 1, Male, Age: 28, Fare: 20.5
dtree.predict((np.array([1, 1, 28, 20.5]).reshape(1, -1)))
```

```
[70]: array([1])
```

```
[71]: # Class = 2, Male, Age: 70, Fare: 7.5
dtree.predict((np.array([2, 1, 70, 7.5]).reshape(1, -1)))
```

```
[71]: array([0])
```

```
[72]: # Class = 3, Female, Age: 25, Fare: 6.76
dtree.predict((np.array([3, 0, 25, 6.76]).reshape(1, -1)))
```

```
[72]: array([0])
```

```
[73]: # Class = 2, Female, Age: 43, Fare: 12.88
dtree.predict((np.array([2, 0, 43, 12.88]).reshape(1, -1)))
```

```
[73]: array([1])
```

```
[74]: print("Accuracy:", metrics.accuracy_score(y_test, y_pred))
      print("Precision:", metrics.precision_score(y_test, y_pred))
```

```
Accuracy: 0.7552447552447552
Precision: 0.7164179104477612
```