

Decision Tree Classification (Titanic Data Set)

August 13, 2022

0.1 Questions & Answers

-

0.1.1 Loading Libraries

```
[1]: import pandas as pd
```

```
[2]: import numpy as np
```

```
[3]: from matplotlib import pyplot as plt
```

```
[4]: %matplotlib inline
```

```
[5]: df = pd.read_csv("../..../datasets/titanic.csv")
```

```
[6]: df.head()
```

```
[6]:
```

	PassengerId	Survived	Pclass	\
0	1	0	3	
1	2	1	1	
2	3	1	3	
3	4	1	1	
4	5	0	3	

	Name	Sex	Age	SibSp	\
0	Braund, Mr. Owen Harris	male	22.0	1	
1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	
2	Heikkinen, Miss. Laina	female	26.0	0	
3	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	
4	Allen, Mr. William Henry	male	35.0	0	

	Parch	Ticket	Fare	Cabin	Embarked
0	0	A/5 21171	7.2500	NaN	S
1	0	PC 17599	71.2833	C85	C
2	0	STON/O2. 3101282	7.9250	NaN	S

3	0	113803	53.1000	C123	S
4	0	373450	8.0500	NaN	S

0.1.2 Exploring Data

```
[7]: #Dropping colum because we do not need id column
df = df.drop(['PassengerId'], axis=1)
```

```
[8]: df.head()
```

```
[8]:
```

	Survived	Pclass	Name \
0	0	3	Braund, Mr. Owen Harris
1	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...
2	1	3	Heikkinen, Miss. Laina
3	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)
4	0	3	Allen, Mr. William Henry

	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	female	38.0	1	0	PC 17599	71.2833	C85	C
2	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	female	35.0	1	0	113803	53.1000	C123	S
4	male	35.0	0	0	373450	8.0500	NaN	S

```
[9]: len(df.columns) #columns
```

```
[9]: 11
```

```
[10]: len(df) #Rows
```

```
[10]: 891
```

```
[11]: #lets check the data set
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 11 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Survived    891 non-null    int64
1   Pclass      891 non-null    int64
2   Name        891 non-null    object
3   Sex         891 non-null    object
4   Age         714 non-null    float64
5   SibSp       891 non-null    int64
```

```

6   Parch      891 non-null   int64
7   Ticket     891 non-null   object
8   Fare       891 non-null   float64
9   Cabin      204 non-null   object
10  Embarked   889 non-null   object
dtypes: float64(2), int64(4), object(5)
memory usage: 76.7+ KB

```

```

[12]: #lets see the statistic overview
df.describe()

```

```

[12]:
count    Survived      Pclass      Age      SibSp      Parch      Fare
mean      0.383838      2.308642    29.699118    0.523008    0.381594    32.204208
std       0.486592      0.836071    14.526497    1.102743    0.806057    49.693429
min       0.000000      1.000000     0.420000    0.000000    0.000000     0.000000
25%      0.000000      2.000000    20.125000    0.000000    0.000000     7.910400
50%      0.000000      3.000000    28.000000    0.000000    0.000000    14.454200
75%      1.000000      3.000000    38.000000    1.000000    0.000000    31.000000
max       1.000000      3.000000    80.000000    8.000000    6.000000   512.329200

```

```

[13]: #finding null values
df.isnull().sum() #we saw there are null values in it

```

```

[13]: Survived      0
Pclass      0
Name        0
Sex         0
Age        177
SibSp       0
Parch       0
Ticket      0
Fare        0
Cabin      687
Embarked    2
dtype: int64

```

```

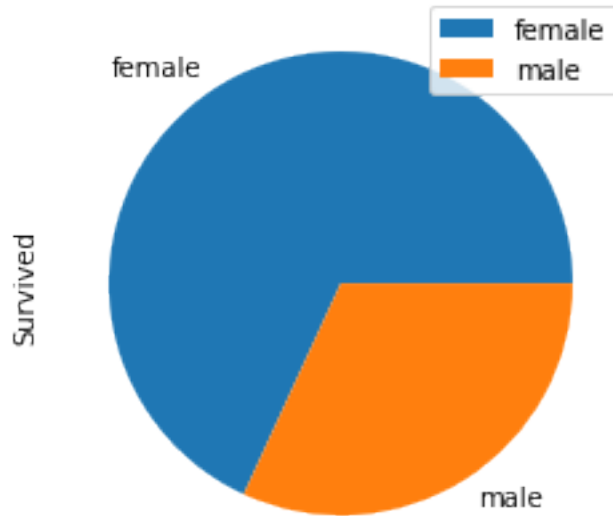
[14]: df.groupby(['Sex']).sum().plot(kind='pie', y='Survived')

```

```

[14]: <AxesSubplot:ylabel='Survived'>

```



```
[15]: df.head()
```

```
[15]:
```

	Survived	Pclass	Name \
0	0	3	Braund, Mr. Owen Harris
1	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...
2	1	3	Heikkinen, Miss. Laina
3	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)
4	0	3	Allen, Mr. William Henry

	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	female	38.0	1	0	PC 17599	71.2833	C85	C
2	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	female	35.0	1	0	113803	53.1000	C123	S
4	male	35.0	0	0	373450	8.0500	NaN	S

```
[17]: #we will drop the target column as inputs this is what we will find
df.drop("Survived",axis="columns")
```

```
[17]:
```

	Pclass	Name	Sex	Age \
0	3	Braund, Mr. Owen Harris	male	22.0
1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0
2	3	Heikkinen, Miss. Laina	female	26.0
3	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0
4	3	Allen, Mr. William Henry	male	35.0
..
886	2	Montvila, Rev. Juozas	male	27.0

887	1	Graham, Miss. Margaret Edith	female	19.0
888	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN
889	1	Behr, Mr. Karl Howell	male	26.0
890	3	Dooley, Mr. Patrick	male	32.0

	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	A/5 21171	7.2500	NaN	S
1	1	0	PC 17599	71.2833	C85	C
2	0	0	STON/O2. 3101282	7.9250	NaN	S
3	1	0	113803	53.1000	C123	S
4	0	0	373450	8.0500	NaN	S
..
886	0	0	211536	13.0000	NaN	S
887	0	0	112053	30.0000	B42	S
888	1	2	W./C. 6607	23.4500	NaN	S
889	0	0	111369	30.0000	C148	C
890	0	0	370376	7.7500	NaN	Q

[891 rows x 10 columns]

```
[19]: #now lets check our target
target = df["Survived"]
```

```
[20]: target #it come in shape of numpy form
```

```
[20]: 0      0
1      1
2      1
3      1
4      0
..
886    0
887    1
888    0
889    1
890    0
Name: Survived, Length: 891, dtype: int64
```

```
[82]: df.head()
```

```
[82]:   Pclass  Sex  Parch
0       3    1      0
1       1    0      0
2       3    0      0
3       1    0      0
4       3    1      0
```

```
[83]: df.head()
```

```
[83]:
```

	Pclass	Sex	Parch
0	3	1	0
1	1	0	0
2	3	0	0
3	1	0	0
4	3	1	0

```
[87]: from sklearn.preprocessing import LabelEncoder
```

```
[88]: t_Sex = LabelEncoder()
```

```
[89]: t_Pclass = LabelEncoder()
```

```
[90]: t_Parch = LabelEncoder()
```

```
[92]: df['sex'] = t_Name.fit_transform(df['Sex'])
```

```
[93]: df['Parch'] = t_fare.fit_transform(df['Parch'])
```

```
[37]: df['Age'] = t_age.fit_transform(df['Age'])
```

```
[94]: df.head()
```

```
[94]:
```

	Pclass	Sex	Parch	sex
0	3	1	0	1
1	1	0	0	0
2	3	0	0	0
3	1	0	0	0
4	3	1	0	1

```
[102]: df.drop("Sex",axis="columns")
```

```
[102]:
```

	Pclass	Parch	sex
0	3	0	1
1	1	0	0
2	3	0	0
3	1	0	0
4	3	0	1
..
886	2	0	1
887	1	0	0
888	3	2	0
889	1	0	1
890	3	0	1

[891 rows x 3 columns]

```
[103]: from sklearn import tree
```

```
[104]: model = tree.DecisionTreeClassifier()
```

```
[105]: model.fit(df,target)
```

```
[105]: DecisionTreeClassifier()
```

```
[110]: model.score(df,target)
```

```
[110]: 0.8058361391694725
```

```
[112]: model.predict([[890,1,0,1]])
```

```
/home/muhammadsardardaudkhan/.local/lib/python3.8/site-  
packages/sklearn/base.py:450: UserWarning: X does not have valid feature names,  
but DecisionTreeClassifier was fitted with feature names  
warnings.warn(
```

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

```
[113]: df.tail()
```

```
[113]:
```

	Pclass	Sex	Parch	sex
886	2	1	0	1
887	1	0	0	0
888	3	0	2	0
889	1	1	0	1
890	3	1	0	1