

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
df = pd.read_csv("tested.csv")
```

```
df.head()
```

	PassengerId	Survived	Pclass	\
0	892	0	3	
1	893	1	3	
2	894	0	2	
3	895	0	3	
4	896	1	3	

	Name	Sex	Age	SibSp
0	Kelly, Mr. James	male	34.5	0
1	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1
2	Myles, Mr. Thomas Francis	male	62.0	0
3	Wirz, Mr. Albert	male	27.0	0
4	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1

	Ticket	Fare	Cabin	Embarked
0	330911	7.8292	NaN	Q
1	363272	7.0000	NaN	S
2	240276	9.6875	NaN	Q
3	315154	8.6625	NaN	S
4	3101298	12.2875	NaN	S

```
df.head(10)
```

	PassengerId	Survived	Pclass	\
0	892	0	3	
1	893	1	3	
2	894	0	2	
3	895	0	3	
4	896	1	3	
5	897	0	3	
6	898	1	3	
7	899	0	2	
8	900	1	3	
9	901	0	3	

	Name	Sex	Age	SibSp
--	------	-----	-----	-------

Parch \					
0		Kelly, Mr. James	male	34.5	0
0					
1		Wilkes, Mrs. James (Ellen Needs)	female	47.0	1
0					
2		Myles, Mr. Thomas Francis	male	62.0	0
0					
3		Wirz, Mr. Albert	male	27.0	0
0					
4		Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1
1					
5		Svensson, Mr. Johan Cervin	male	14.0	0
0					
6		Connolly, Miss. Kate	female	30.0	0
0					
7		Caldwell, Mr. Albert Francis	male	26.0	1
1					
8		Abraham, Mrs. Joseph (Sophie Halaut Easu)	female	18.0	0
0					
9		Davies, Mr. John Samuel	male	21.0	2
0					

	Ticket	Fare	Cabin	Embarked
0	330911	7.8292	NaN	Q
1	363272	7.0000	NaN	S
2	240276	9.6875	NaN	Q
3	315154	8.6625	NaN	S
4	3101298	12.2875	NaN	S
5	7538	9.2250	NaN	S
6	330972	7.6292	NaN	Q
7	248738	29.0000	NaN	S
8	2657	7.2292	NaN	C
9	A/4 48871	24.1500	NaN	S

df.shape

(418, 12)

df.describe()

	PassengerId	Survived	Pclass	Age	SibSp \
count	418.000000	418.000000	418.000000	332.000000	418.000000
mean	1100.500000	0.363636	2.265550	30.272590	0.447368
std	120.810458	0.481622	0.841838	14.181209	0.896760
min	892.000000	0.000000	1.000000	0.170000	0.000000
25%	996.250000	0.000000	1.000000	21.000000	0.000000
50%	1100.500000	0.000000	3.000000	27.000000	0.000000
75%	1204.750000	1.000000	3.000000	39.000000	1.000000
max	1309.000000	1.000000	3.000000	76.000000	8.000000

	Parch	Fare
count	418.000000	417.000000
mean	0.392344	35.627188
std	0.981429	55.907576
min	0.000000	0.000000
25%	0.000000	7.895800
50%	0.000000	14.454200
75%	0.000000	31.500000
max	9.000000	512.329200

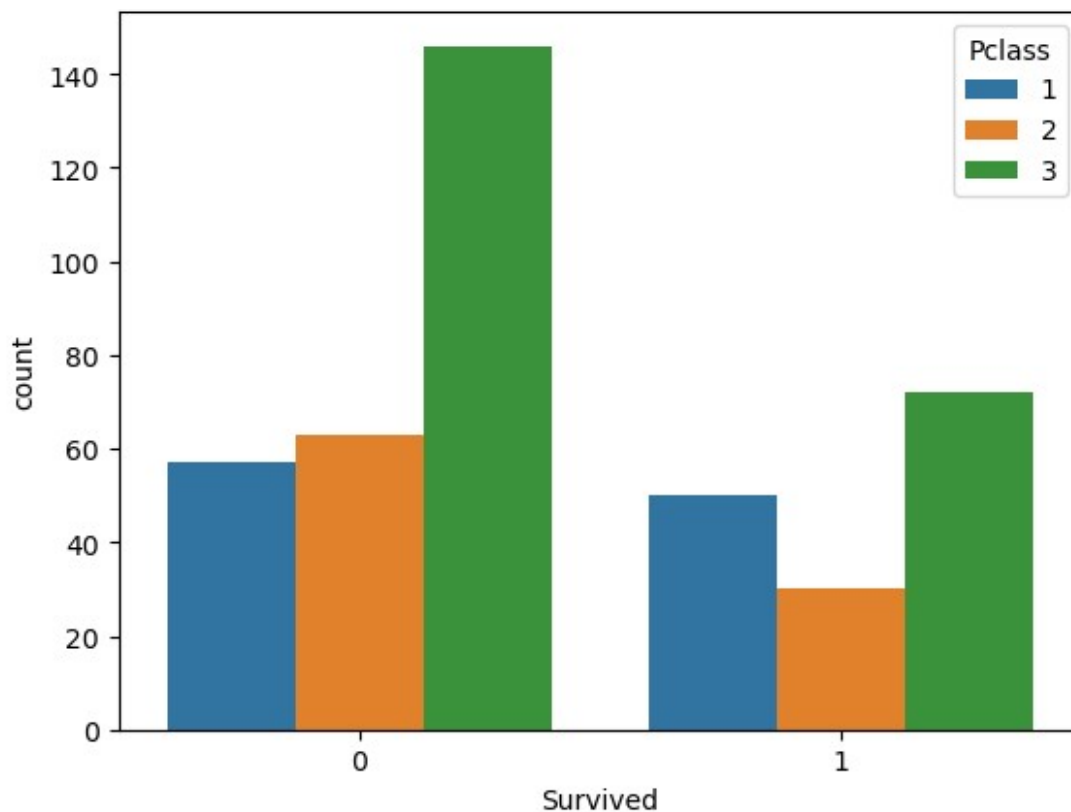
```
df['Survived'].value_counts()
```

```
0    266
1    152
```

```
Name: Survived, dtype: int64
```

```
#let's visualize the count of survivals wrt pclass
sns.countplot(x=df['Survived'], hue=df['Pclass'])
```

```
<AxesSubplot: xlabel='Survived', ylabel='count'>
```



```
df["Sex"]
```

```
0    male
1    female
```

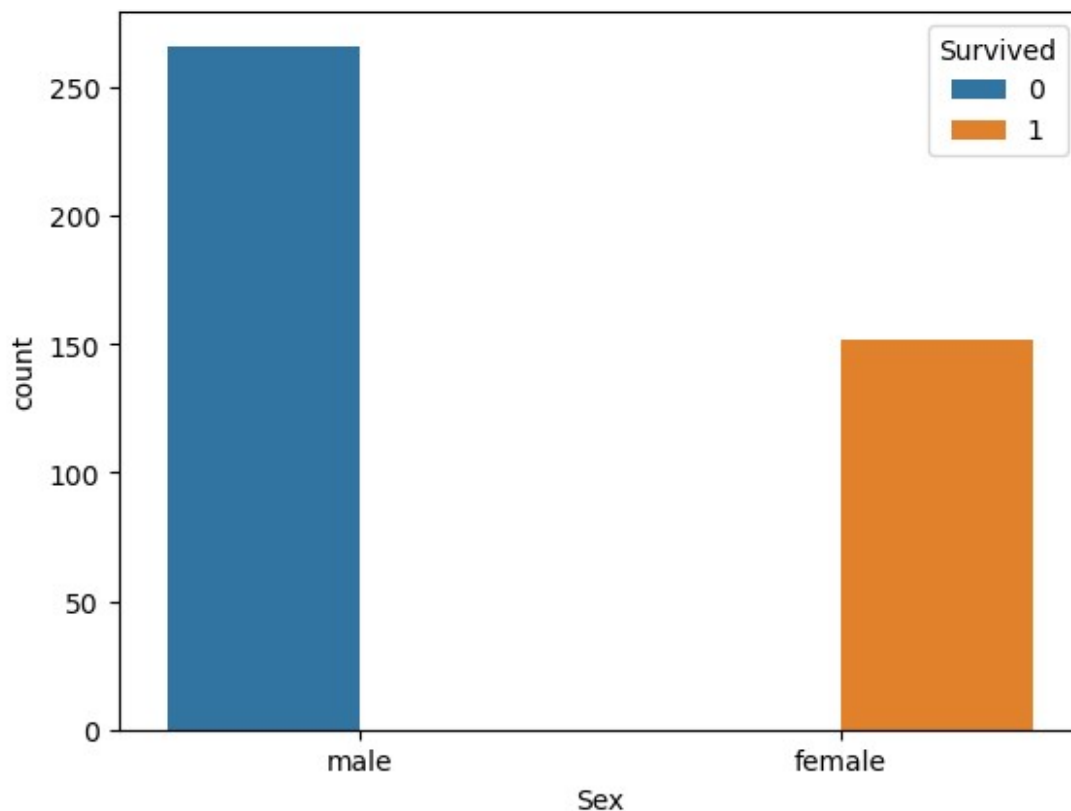
```

2      male
3      male
4      female
...
413    male
414    female
415    male
416    male
417    male
Name: Sex, Length: 418, dtype: object

#let's visualize the count of survivals wrt Gender
sns.countplot(x=df['Sex'], hue=df['Survived'])

<AxesSubplot: xlabel='Sex', ylabel='count'>

```



```

#Look at survival rate by sex
df.groupby('Sex')[['Survived']].mean()

```

Sex	Survived
female	1.0
male	0.0

```
df['Sex'].unique()
array(['male', 'female'], dtype=object)
from sklearn.preprocessing import LabelEncoder
labelencoder = LabelEncoder()
df['Sex']= labelencoder.fit_transform(df['Sex'])
df.head()
```

	PassengerId	Survived	Pclass	\
0	892	0	3	
1	893	1	3	
2	894	0	2	
3	895	0	3	
4	896	1	3	

		Name	Sex	Age	SibSp
Parch	\				
0		Kelly, Mr. James	1	34.5	0
0					
1		Wilkes, Mrs. James (Ellen Needs)	0	47.0	1
0					
2		Myles, Mr. Thomas Francis	1	62.0	0
0					
3		Wirz, Mr. Albert	1	27.0	0
0					
4		Hirvonen, Mrs. Alexander (Helga E Lindqvist)	0	22.0	1
1					

	Ticket	Fare	Cabin	Embarked
0	330911	7.8292	NaN	Q
1	363272	7.0000	NaN	S
2	240276	9.6875	NaN	Q
3	315154	8.6625	NaN	S
4	3101298	12.2875	NaN	S

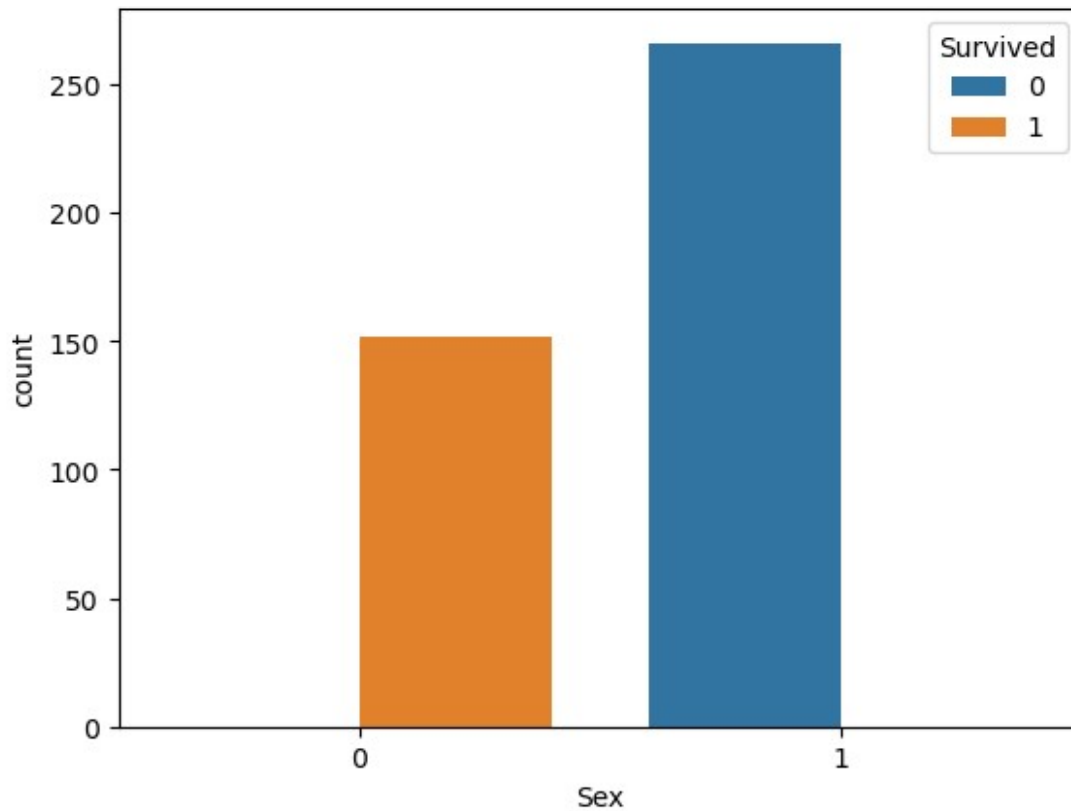
```
df['Sex'], df['Survived']
```

(0	1
1	0
2	1
3	1
4	0
	..
413	1
414	0
415	1
416	1
417	1

```

Name: Sex, Length: 418, dtype: int64,
0      0
1      1
2      0
3      0
4      1
..
413    0
414    1
415    0
416    0
417    0
Name: Survived, Length: 418, dtype: int64)
sns.countplot(x=df['Sex'], hue=df["Survived"])
<AxesSubplot: xlabel='Sex', ylabel='count'>

```



```

df.isna().sum()
PassengerId    0
Survived       0
Pclass         0
Name           0

```

```

Sex          0
Age          86
SibSp        0
Parch        0
Ticket       0
Fare         1
Cabin       327
Embarked     0
dtype: int64

```

*# After dropping non required column*

```
df=df.drop(['Age'], axis=1)
```

```
df_final = df
```

```
df_final.head(10)
```

	PassengerId	Survived	Pclass	\
0	892	0	3	
1	893	1	3	
2	894	0	2	
3	895	0	3	
4	896	1	3	
5	897	0	3	
6	898	1	3	
7	899	0	2	
8	900	1	3	
9	901	0	3	

	Ticket	\	Name	Sex	SibSp	Parch
0	330911		Kelly, Mr. James	1	0	0
1	363272		Wilkes, Mrs. James (Ellen Needs)	0	1	0
2	240276		Myles, Mr. Thomas Francis	1	0	0
3	315154		Wirz, Mr. Albert	1	0	0
4	3101298		Hirvonen, Mrs. Alexander (Helga E Lindqvist)	0	1	1
5	7538		Svensson, Mr. Johan Cervin	1	0	0
6	330972		Connolly, Miss. Kate	0	0	0
7	248738		Caldwell, Mr. Albert Francis	1	1	1
8	2657		Abraham, Mrs. Joseph (Sophie Halaut Easu)	0	0	0
9	A/4 48871		Davies, Mr. John Samuel	1	2	0

	Fare	Cabin	Embarked
0	7.8292	NaN	Q
1	7.0000	NaN	S
2	9.6875	NaN	Q
3	8.6625	NaN	S
4	12.2875	NaN	S
5	9.2250	NaN	S
6	7.6292	NaN	Q
7	29.0000	NaN	S
8	7.2292	NaN	C
9	24.1500	NaN	S

```
X= df[['Pclass', 'Sex']]
Y=df['Survived']
```

```
from sklearn.model_selection import train_test_split
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size =
0.2, random_state = 0)
```

```
from sklearn.linear_model import LogisticRegression
```

```
log = LogisticRegression(random_state = 0)
log.fit(X_train, Y_train)
```

```
LogisticRegression(random_state=0)
```

```
pred = print(log.predict(X_test))
```

```
[0 0 1 0 1 0 1 0 0 0 1 1 0 0 0 0 1 0 1 1 0 1 0 0 0 0 1 0 0 0 1 1 1 1 1
0 0
 1 1 1 1 0 1 1 0 1 0 0 0 0 0 1 1 0 0 1 0 1 0 0 0 1 1 0 0 1 1 1 1 0 0 1
1 1
 1 0 0 1 0 1 0 1 0 0]
```

```
print(Y_test)
```

```
360    0
170    0
224    1
358    0
309    1
..
100    1
7      0
22     1
68     0
328    0
```

```
Name: Survived, Length: 84, dtype: int64
```



```
import warnings
warnings.filterwarnings("ignore")

res= log.predict([[2,1]])

if(res==0):
    print("So Sorry! Not Survived")
else:
    print("Survived")

So Sorry! Not Survived
```