

```
In [ ]: Task - 1 TITANIC SURVIVAL PREDICTION

In [ ]: CODESOFT
```

Importing Important Libraries

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

In [2]: df = pd.read_csv('Titanic-Dataset.csv')
df.head(10)
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	0	PC 17599	71.2833	C85	C
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S
5	6	0	3	Moran, Mr. James	male	NaN	0	0	330877	8.4583	NaN	Q
6	7	0	1	McCarthy, Mr. Timothy J	male	54.0	0	0	17463	51.8625	E46	S
7	8	0	3	Palsson, Master. Gosta Leonard	male	2.0	3	1	349909	21.0750	NaN	S
8	9	1	3	Johnson, Mrs. Oscar W (Elisabeth Vilhelmina Berg)	female	27.0	0	2	347742	11.1333	NaN	S
9	10	1	2	Nasser, Mrs. Nicholas (Adele Achem)	female	14.0	1	0	237736	30.0708	NaN	C

```
In [3]: df.shape
Out[3]: (891, 12)
```

```
In [4]: df.describe()

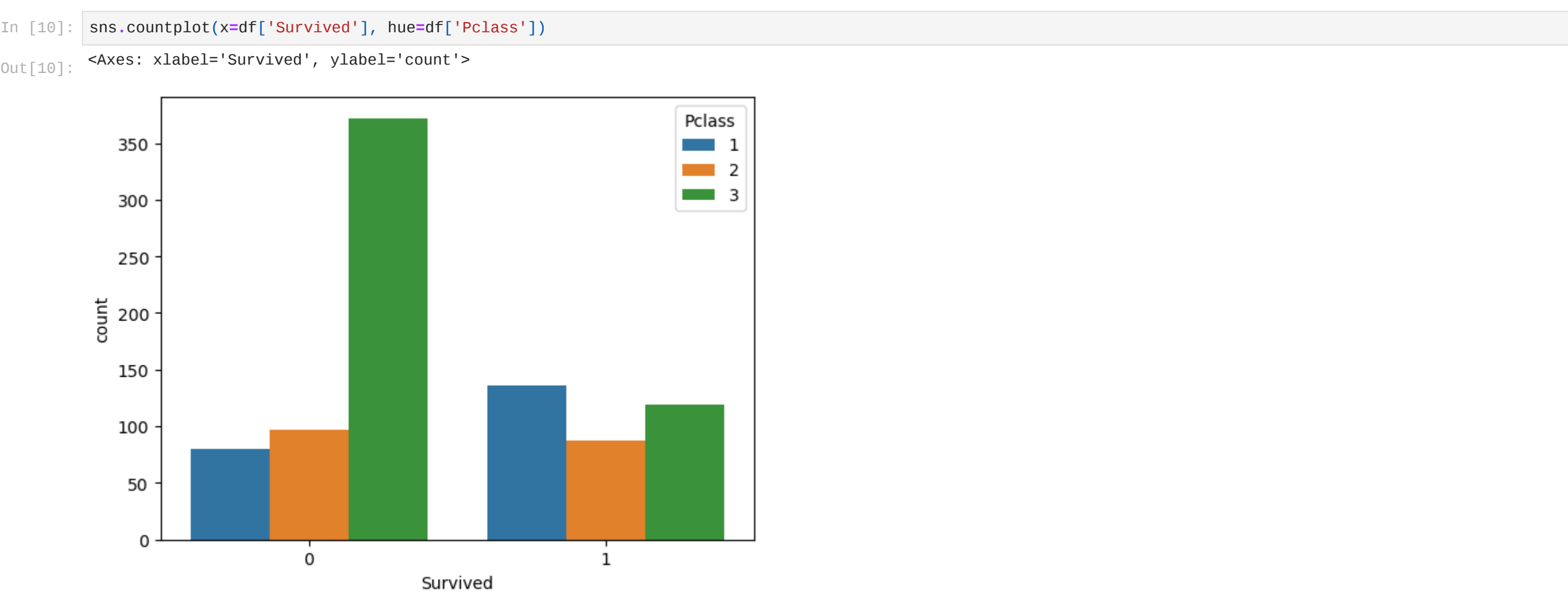
Out[4]:
```

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare
count	891.000000	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000
mean	446.000000	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208
std	257.353842	0.486592	0.836071	14.526497	1.102743	0.806057	49.693429
min	1.000000	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000
25%	223.500000	0.000000	2.000000	20.125000	0.000000	0.000000	7.910400
50%	446.000000	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200
75%	668.500000	1.000000	3.000000	38.000000	1.000000	0.000000	31.000000
max	891.000000	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200

```
In [6]: df['Survived'].value_counts()

Out[6]:
0    549
1    342
Name: Survived, dtype: int64
```

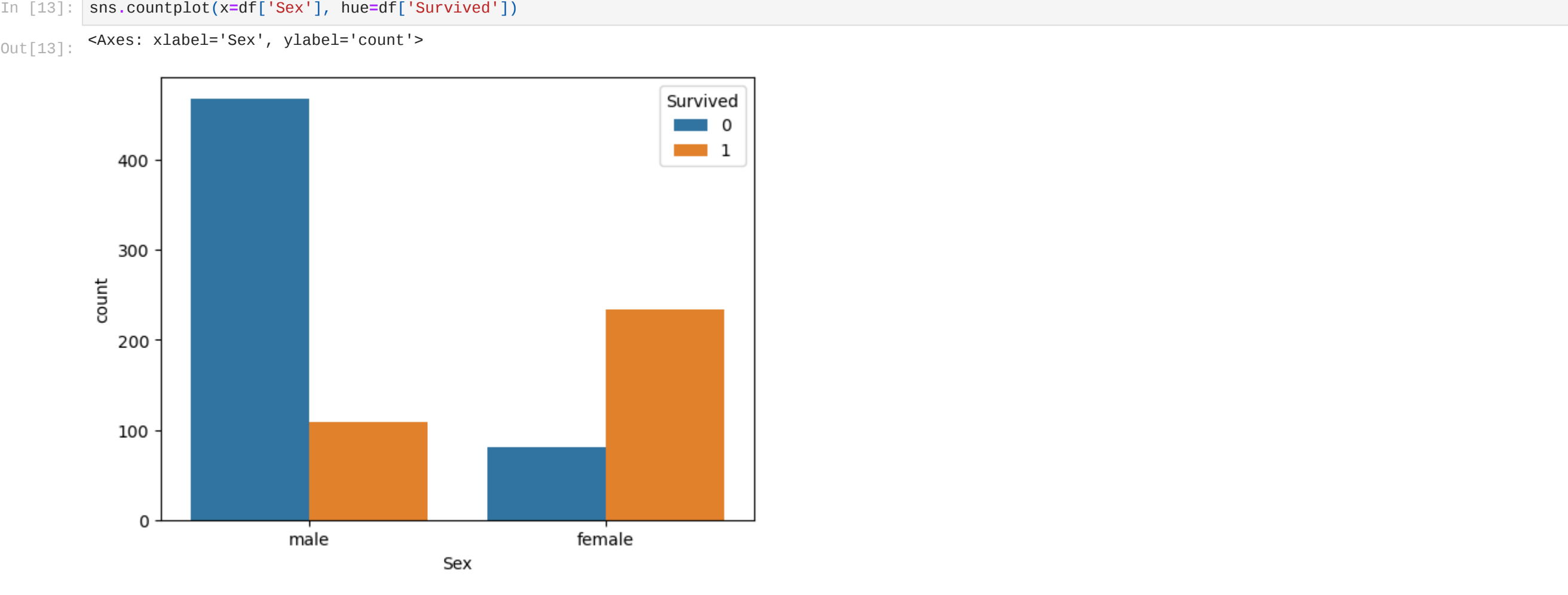
let visualize the count of Survived wrt pclass



```
In [12]: df['Sex']

Out[12]:
0    male
1    female
2    female
3    female
4    male
...
886    male
887    female
888    female
889    male
890    male
Name: Sex, Length: 891, dtype: object
```

let visualize the count of Survived wrt Gender



Look at Survived rate by Sex

```
In [14]: df.groupby('Sex')[['Survived']].mean()

Out[14]:
```

	Survived
female	0.742038
male	0.188908

```
In [15]: df['Sex'].unique()

Out[15]: array(['male', 'female'], dtype=object)
```

```
In [16]: from sklearn.preprocessing import LabelEncoder
labelencoder = LabelEncoder()

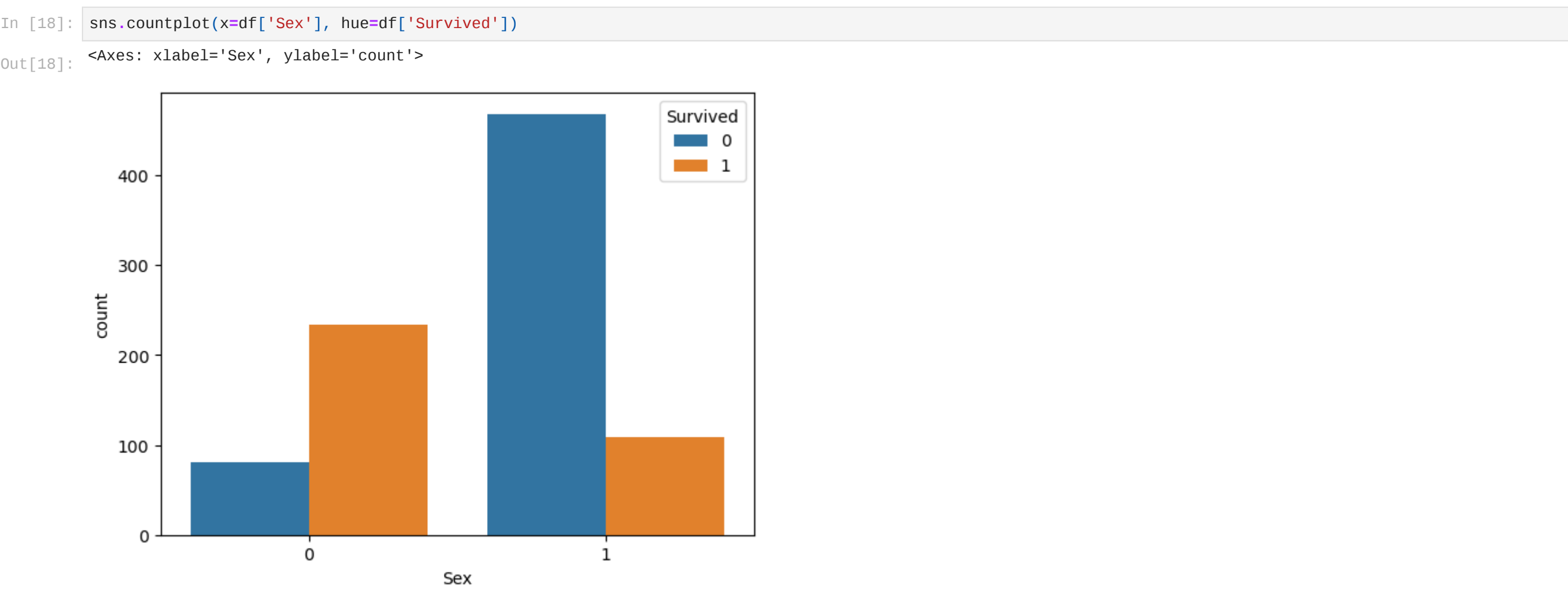
df['Sex']= labelencoder.fit_transform(df['Sex'])
df.head()

Out[16]:
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	1	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	0	38.0	1	0	PC 17599	71.2833	C85	C
2	3	1	3	Heikkinen, Miss. Laina	0	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	0	35.0	1	0	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	1	35.0	0	0	373450	8.0500	NaN	S

```
In [17]: df['Sex'], df['Survived']

Out[17]: (0    1
1    0
2    0
3    0
4    1
...
886    1
887    0
888    0
889    1
890    1
Name: Sex, Length: 891, dtype: int32,
0    1
1    1
2    1
3    1
4    0
...
886    0
887    1
888    0
889    1
890    0
Name: Survived, Length: 891, dtype: int64)
```



```
In [20]: df.isna().sum()

Out[20]:
PassengerId    0
Survived        0
Pclass          0
Name            0
Sex             0
Age            177
SibSp           0
Parch           0
Ticket          0
Fare            0
Cabin          687
Embarked        2
dtype: int64
```

After dropping non required column

```
In [21]: df=df.drop(['Age'], axis=1)

In [22]: df_final =df
df_final.head(10)
```

	PassengerId	Survived	Pclass	Name	Sex	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	1	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	0	1	0	PC 17599	71.2833	C85	C
2	3	1	3	Heikkinen, Miss. Laina	0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	0	1	0	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	1	0	0	373450	8.0500	NaN	S
5	6	0	3	Moran, Mr. James	1	0	0	330877	8.4583	NaN	Q
6	7	0	1	McCarthy, Mr. Timothy J	1	0	0	17463	51.8625	E46	S
7	8	0	3	Palsson, Master. Gosta Leonard	1	3	1	349909	21.0750	NaN	S
8	9	1	3	Johnson, Mrs. Oscar W (Elisabeth Vilhelmina Berg)	0	0	2	347742	11.1333	NaN	S
9	10	1	2	Nasser, Mrs. Nicholas (Adele Achem)	0	1	0	237736	30.0708	NaN	C

Model Training

```
In [23]: X=df[['Pclass','Sex']]
Y=df['Survived']

In [28]: 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)

In [29]: from sklearn.linear_model import LogisticRegression

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

Out[29]: LogisticRegression
```

Model Prediction

```
In [30]: pred = print(log.predict(X_test))

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

In [31]: print(Y_test)

495    0
648    0
278    0
31     1
255    1
...
780    1
837    0
215    1
833    0
372    0
Name: Survived, Length: 179, dtype: int64

In [33]: import warnings
warnings.filterwarnings("ignore")

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

if(res==0):
    print("so sorry! Not Suvised")
    print("Survived")

so sorry! Not Suvised
Survived

In [ ]: *****END OF THE CODE*****

In [ ]: *****THANK YOU *****
```