

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

In [2]: `df=pd.read_csv(r'C:\Users\mohan\Downloads\archive (2)\Titanic-Dataset.csv')`

In [3]: `df.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 12 columns):
#   Column              Non-Null Count  Dtype  
---  --
0   PassengerId         891 non-null   int64  
1   Survived            891 non-null   int64  
2   Pclass              891 non-null   int64  
3   Name                891 non-null   object  
4   Sex                 891 non-null   object  
5   Age                 714 non-null   float64
6   SibSp              891 non-null   int64  
7   Parch              891 non-null   int64  
8   Ticket              891 non-null   object  
9   Fare                891 non-null   float64
10  Cabin               204 non-null   object  
11  Embarked            889 non-null   object  
dtypes: float64(2), int64(5), object(5)
memory usage: 83.7+ KB
```

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

	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	Cummings, 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

In [5]: `df.tail()`

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.00	NaN	S
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.00	B42	S
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.45	NaN	S
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.00	C148	C
890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.75	NaN	Q

In [6]: `df.describe()`

	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 [7]: `print("The column values are:")
df.columns.values`

The column values are:

Out[7]: `array(['PassengerId', 'Survived', 'Pclass', 'Name', 'Sex', 'Age', 'SibSp', 'Parch', 'Ticket', 'Fare', 'Cabin', 'Embarked'], dtype=object)`

In [8]: `print("Survival:")
df['Survived']`

Survival:

Out[8]: `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`

In [9]: `df['Survived'].value_counts()`

Survived

Out[9]: `0 549
1 342
Name: count, dtype: int64`

In [10]: `print('Survival grouped by gender')
gender=df.groupby('Sex')['Survived'].count()
print(gender)`

Survival grouped by gender

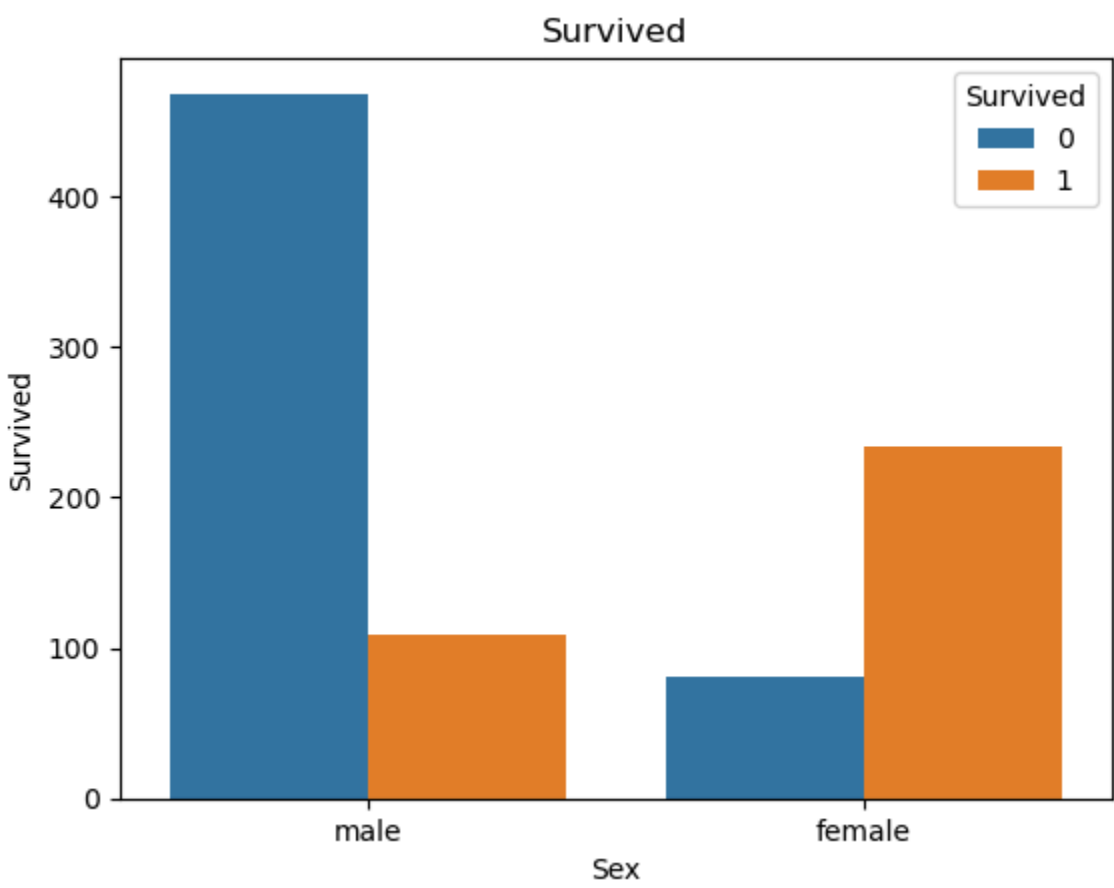
Sex

female 314

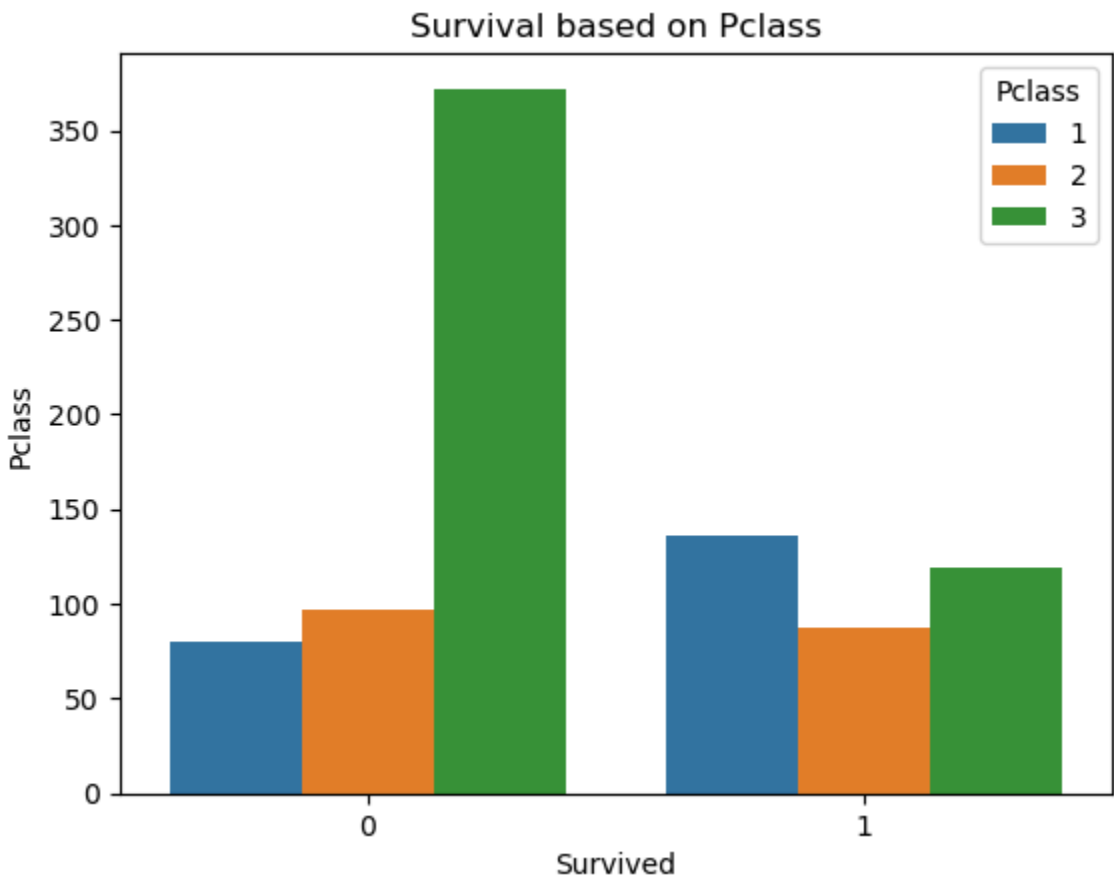
male 577

Name: Survived, dtype: int64

In [11]: `#Survival based on Sex
sns.countplot(x=df['Sex'],hue=df['Survived'])
plt.title('Survived')
plt.xlabel('Sex')
plt.ylabel('Survived')
plt.show()`



In [12]: `#Survival based on Pclass
sns.countplot(x=df['Survived'],hue=df['Pclass'])
plt.title('Survival based on Pclass')
plt.xlabel('Survived')
plt.ylabel('Pclass')
plt.show()`



In [13]: `from sklearn.preprocessing import LabelEncoder
labelencoder=LabelEncoder()
df['Sex']=labelencoder.fit_transform(df['Sex'])
df.head()`

Out[13]:

	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	Cummings, 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 [14]: `#model training
x=df[['Sex','Pclass']]
y=df['Survived']`

In [15]: `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 [16]: `from sklearn.linear_model import LogisticRegression
log=LogisticRegression(random_state=0)
log.fit(x_train,y_train)`

Out[16]: `LogisticRegression`
`LogisticRegression(random_state=0)`

In [17]: `print(log.predict(x_test))`

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

In [18]: `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 [19]: `import warnings
warnings.filterwarnings("ignore")
res=log.predict([[0,3]])
if(res!=0):
 print("Passenger survive")
else:
 print("Passenger not survived")`

Passenger survive

In []: