

Bharat Intern

Task - 2 (Titanic Classification)

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Importing the Libraries

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
import warnings
warnings.filterwarnings('ignore')
```

Loading the dataset

```
In [4]: df = pd.read_csv("tested.csv")
df.head()
```

```
Out[4]:
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin
0	892	0	3	Kelly, Mr. James	male	34.5	0	0	330911	7.8292	NaN
1	893	1	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	7.0000	NaN
2	894	0	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875	NaN
3	895	0	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625	NaN
4	896	1	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875	NaN

```
In [5]: df.shape
```

```
Out[5]: (418, 12)
```

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

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 418 entries, 0 to 417
Data columns (total 12 columns):
 #   Column        Non-Null Count  Dtype
---  ---
 0   PassengerId   418 non-null    int64
 1   Survived      418 non-null    int64
 2   Pclass        418 non-null    int64
 3   Name          418 non-null    object
 4   Sex           418 non-null    object
 5   Age           332 non-null    float64
 6   SibSp         418 non-null    int64
 7   Parch         418 non-null    int64
 8   Ticket        418 non-null    object
 9   Fare          417 non-null    float64
10   Cabin         91 non-null     object
11   Embarked      418 non-null    object
dtypes: float64(2), int64(5), object(5)
memory usage: 39.3+ KB
```

```
In [10]: df.describe()
```

```
Out[10]:
```

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare
count	418.000000	418.000000	418.000000	332.000000	418.000000	418.000000	417.000000
mean	1100.500000	0.363636	2.265550	30.272590	0.447368	0.392344	35.627188
std	120.810458	0.481622	0.841838	14.181209	0.896760	0.981429	55.907576
min	892.000000	0.000000	1.000000	0.170000	0.000000	0.000000	0.000000
25%	996.250000	0.000000	1.000000	21.000000	0.000000	0.000000	7.895800
50%	1100.500000	0.000000	3.000000	27.000000	0.000000	0.000000	14.454200
75%	1204.750000	1.000000	3.000000	39.000000	1.000000	0.000000	31.500000
max	1309.000000	1.000000	3.000000	76.000000	8.000000	9.000000	512.329200

```
In [7]: df.isnull().any()
```

```
Out[7]: PassengerId    False
Survived      False
Pclass        False
Name          False
Sex           False
Age           True
SibSp         False
Parch         False
Ticket        False
Fare          True
Cabin         True
Embarked      False
dtype: bool
```

```
In [8]: df['Survived'].value_counts()
# 0 means Not Survived
# 1 means Survived
```

```
Out[8]: 0    266
1     152
Name: Survived, dtype: int64
```

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

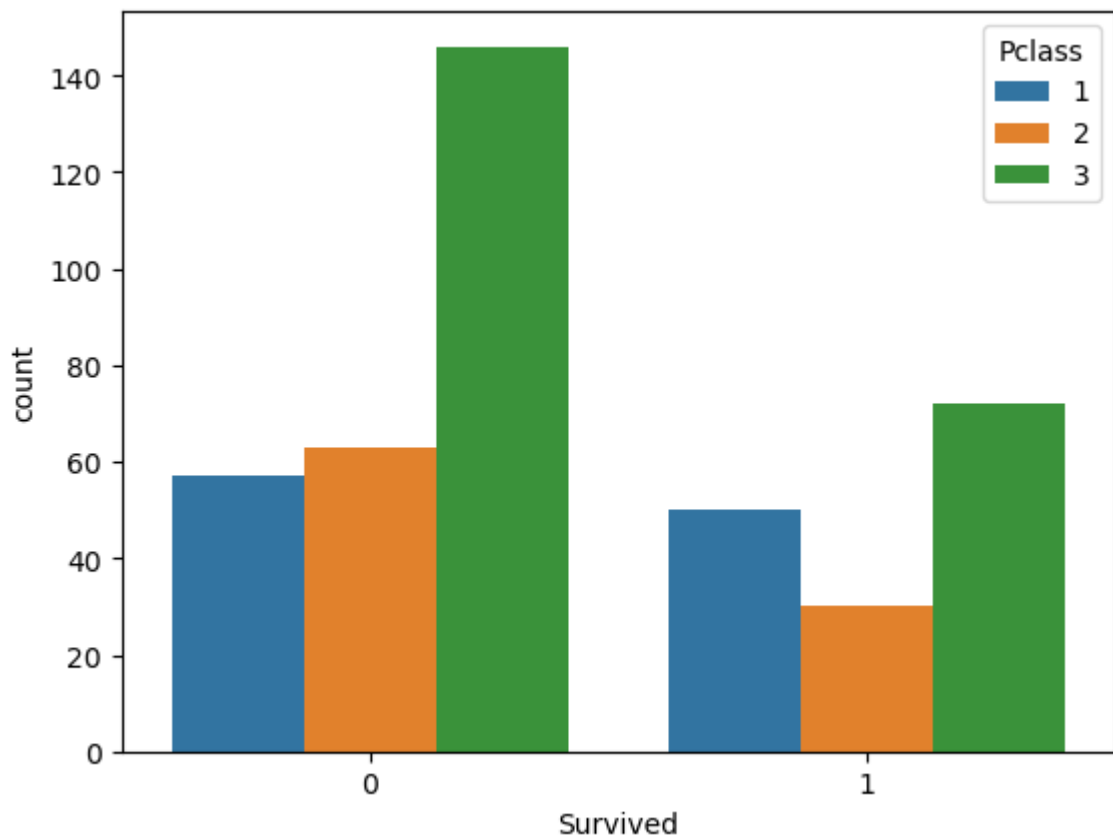
```
Out[9]: male      266  
female    152  
Name: Sex, dtype: int64
```

Data Visualisation

Visualising the count of Survivals with respect to Pclass parameter

```
In [13]: sns.countplot(x=df['Survived'], hue=df['Pclass'])
```

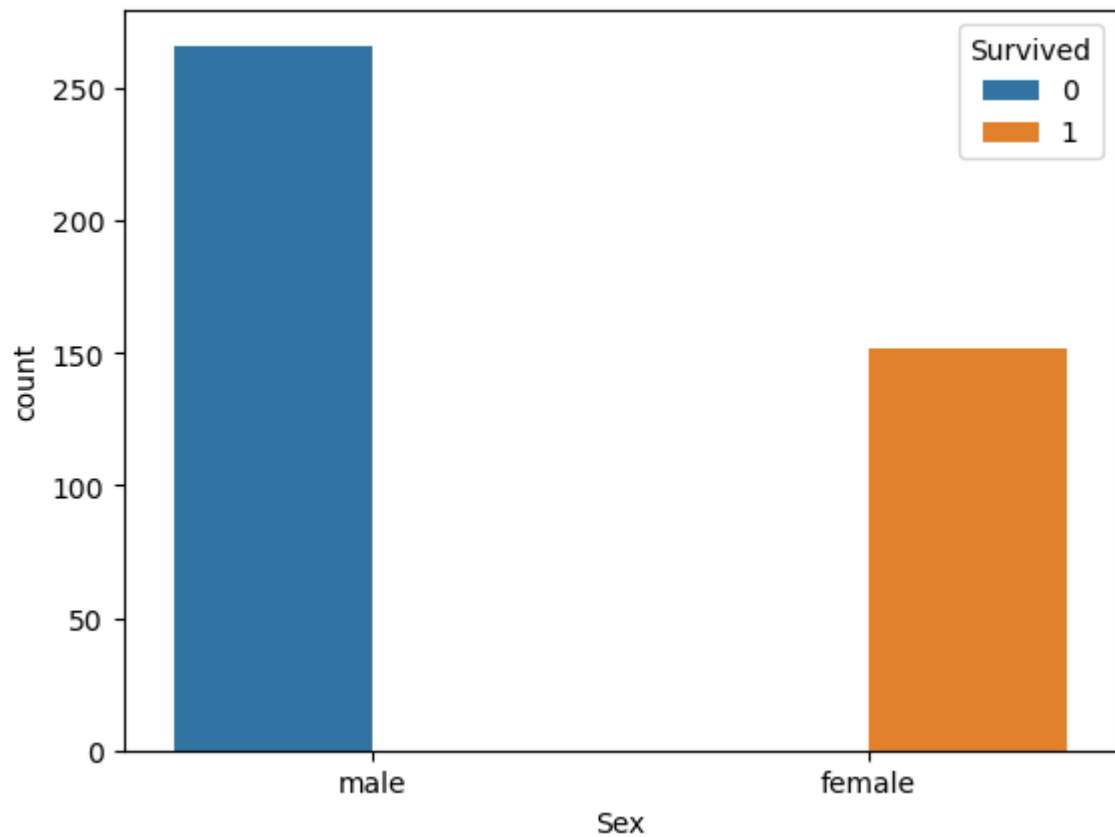
```
Out[13]: <Axes: xlabel='Survived', ylabel='count'>
```



Visualising the count of Survivals with respect to Gender parameter

```
In [13]: sns.countplot(x=df['Sex'], hue=df['Survived'])
```

```
Out[13]: <Axes: xlabel='Sex', ylabel='count'>
```



```
In [14]: df['Sex']
```

```
Out[14]: 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
```

```
In [15]: from sklearn.preprocessing import LabelEncoder
labelencoder = LabelEncoder()

df['Sex'] = labelencoder.fit_transform(df['Sex'])
df.head()
# Male = 1
# Female = 0
```

Out[15]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	En
0	892	0	3	Kelly, Mr. James	1	34.5	0	0	330911	7.8292	NaN	
1	893	1	3	Wilkes, Mrs. James (Ellen Needs)	0	47.0	1	0	363272	7.0000	NaN	
2	894	0	2	Myles, Mr. Thomas Francis	1	62.0	0	0	240276	9.6875	NaN	
3	895	0	3	Wirz, Mr. Albert	1	27.0	0	0	315154	8.6625	NaN	
4	896	1	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	0	22.0	1	1	3101298	12.2875	NaN	

In [16]:

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

Out[16]:

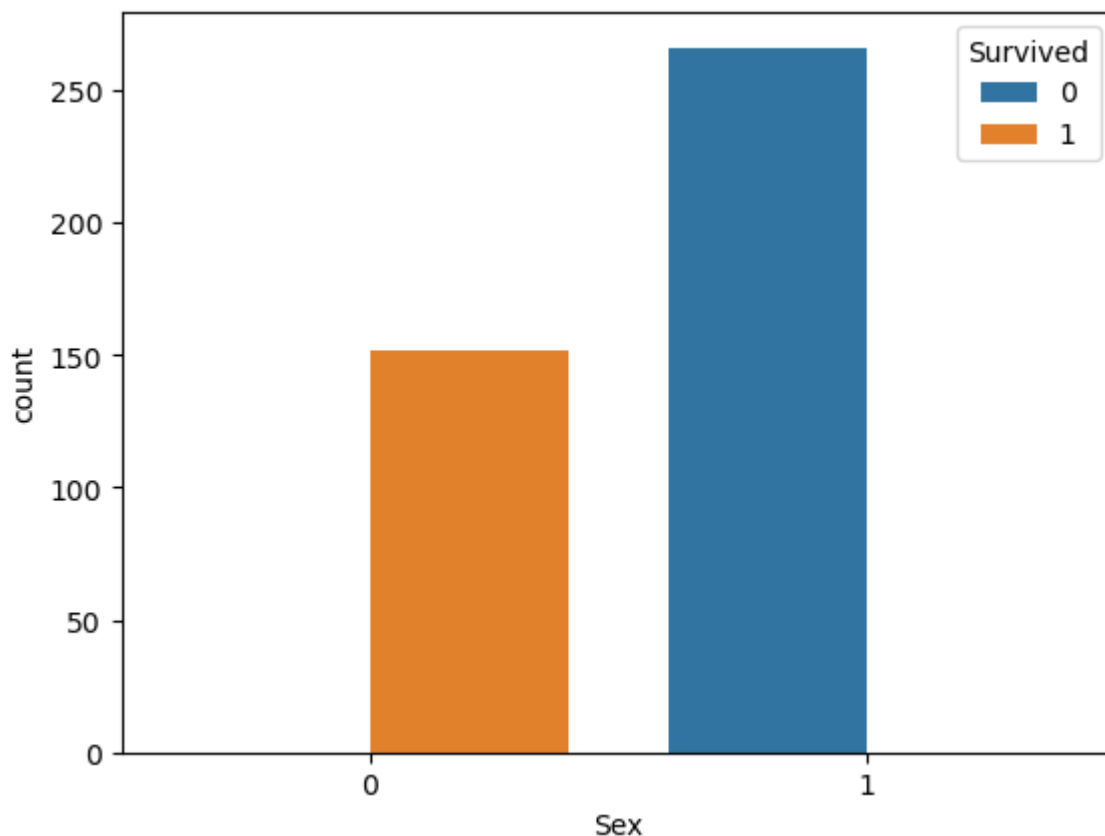
(0 1
1 0
2 1
3 1
4 0
..
413 1
414 0
415 1
416 1
417 1
Name: Sex, Length: 418, dtype: int32,
0 0
1 1
2 0
3 0
4 1
..
413 0
414 1
415 0
416 0
417 0
Name: Survived, Length: 418, dtype: int64)

In [17]:

sns.countplot(x=df['Sex'], hue=df['Survived'])

Out[17]:

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



```
In [18]: df.isna().sum()
```

```
Out[18]: 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
```

Dropping the Age column

```
In [19]: if "Age" in df.columns:
df.drop("Age", axis=1, inplace=True)
```

```
In [20]: new_df = df
new_df.head()
```

Out[20]:

	PassengerId	Survived	Pclass	Name	Sex	SibSp	Parch	Ticket	Fare	Cabin	Embarke
0	892	0	3	Kelly, Mr. James	1	0	0	330911	7.8292	NaN	
1	893	1	3	Wilkes, Mrs. James (Ellen Needs)	0	1	0	363272	7.0000	NaN	
2	894	0	2	Myles, Mr. Thomas Francis	1	0	0	240276	9.6875	NaN	
3	895	0	3	Wirz, Mr. Albert	1	0	0	315154	8.6625	NaN	
4	896	1	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	0	1	1	3101298	12.2875	NaN	

Training the Model

In [21]:

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

In [22]:

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_s

In [23]:

from sklearn.linear_model import LogisticRegression
log = LogisticRegression(random_state=0)
log.fit(X_train, Y_train)

Out[23]:

▼

LogisticRegression

LogisticRegression(random_state=0)

Model Prediction

In [26]:

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]

In [27]:

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
```

```
In [42]: import warnings
warnings.filterwarnings("ignore")

res = log.predict([[2,0]])

if(res==0):
    print("Not Survived")
else:
    print("Survived")
```

Survived