

titanic-survival-prediction

March 17, 2024

Import necessary libs

```
[5]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score
```

Data Collection and Processing

```
[9]: df=pd.read_csv('train.csv')
df.head()
```

```
[9]:
```

	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

```
[11]: df.shape
```

```
[11]: (891, 12)
```

```
[12]: 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
```

```
[13]: df.isnull().sum()
```

```
[13]: 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
```

```
[14]: #remove missing/null values
df = df.drop(columns='Cabin', axis=1)
```

```
[15]: #replacing missing values with mean number
df['Age'].fillna(df['Age'].mean(), inplace=True)
```

```
[16]: df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 11 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              891 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  Embarked         889 non-null    object
dtypes: float64(2), int64(5), object(4)
memory usage: 76.7+ KB

```

```
[17]: df.isnull().sum()
```

```

[17]: PassengerId      0
      Survived        0
      Pclass          0
      Name            0
      Sex             0
      Age             0
      SibSp           0
      Parch           0
      Ticket          0
      Fare            0
      Embarked        2
      dtype: int64

```

```

[18]: #lets fix Embarked
      print(df['Embarked'].mode())

```

```

0    S
Name: Embarked, dtype: object

```

```
[19]: print(df['Embarked'].mode()[0])
```

```
S
```

```

[21]: #replace the mode value with the missing value
      df['Embarked'].fillna(df['Embarked'].mode()[0], inplace=True)

```

```
[24]: df.isnull().sum()
```

```
[24]: PassengerId    0
      Survived      0
      Pclass       0
      Name         0
      Sex          0
      Age         0
      SibSp        0
      Parch        0
      Ticket       0
      Fare         0
      Embarked     0
      dtype: int64
```

Analysing the data

```
[25]: df.describe()
```

```
[25]:
```

	PassengerId	Survived	Pclass	Age	SibSp \
count	891.000000	891.000000	891.000000	891.000000	891.000000
mean	446.000000	0.383838	2.308642	29.699118	0.523008
std	257.353842	0.486592	0.836071	13.002015	1.102743
min	1.000000	0.000000	1.000000	0.420000	0.000000
25%	223.500000	0.000000	2.000000	22.000000	0.000000
50%	446.000000	0.000000	3.000000	29.699118	0.000000
75%	668.500000	1.000000	3.000000	35.000000	1.000000
max	891.000000	1.000000	3.000000	80.000000	8.000000

	Parch	Fare
count	891.000000	891.000000
mean	0.381594	32.204208
std	0.806057	49.693429
min	0.000000	0.000000
25%	0.000000	7.910400
50%	0.000000	14.454200
75%	0.000000	31.000000
max	6.000000	512.329200

```
[26]: #how many survived?
      df['Survived'].value_counts()
```

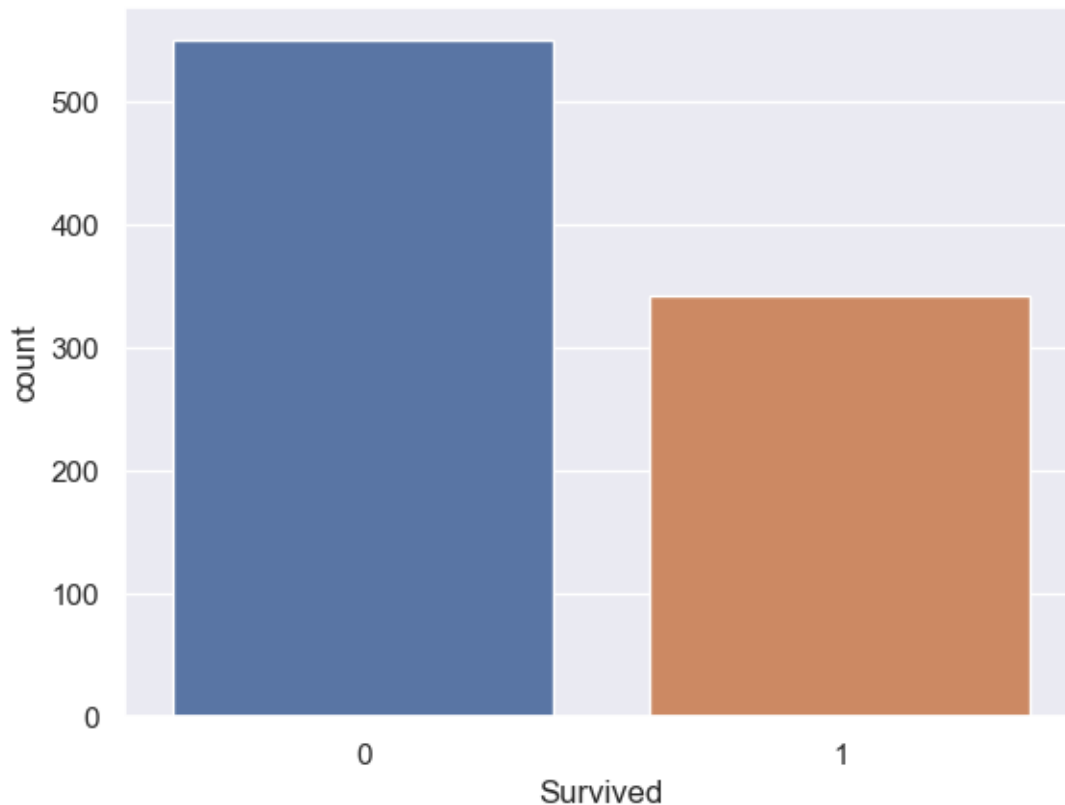
```
[26]: 0    549
      1    342
      Name: Survived, dtype: int64
```

```
[27]: #visualizing data
      sns.set()
```

```
[28]: sns.countplot(df['Survived'])
```

```
C:\Users\CHANDRA ADITYA\anaconda3\lib\site-packages\seaborn\_decorators.py:36:
FutureWarning: Pass the following variable as a keyword arg: x. From version
0.12, the only valid positional argument will be `data`, and passing other
arguments without an explicit keyword will result in an error or
misinterpretation.
  warnings.warn(
```

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



```
[29]: df['Sex'].value_counts()
```

```
[29]: male      577
female    314
Name: Sex, dtype: int64
```

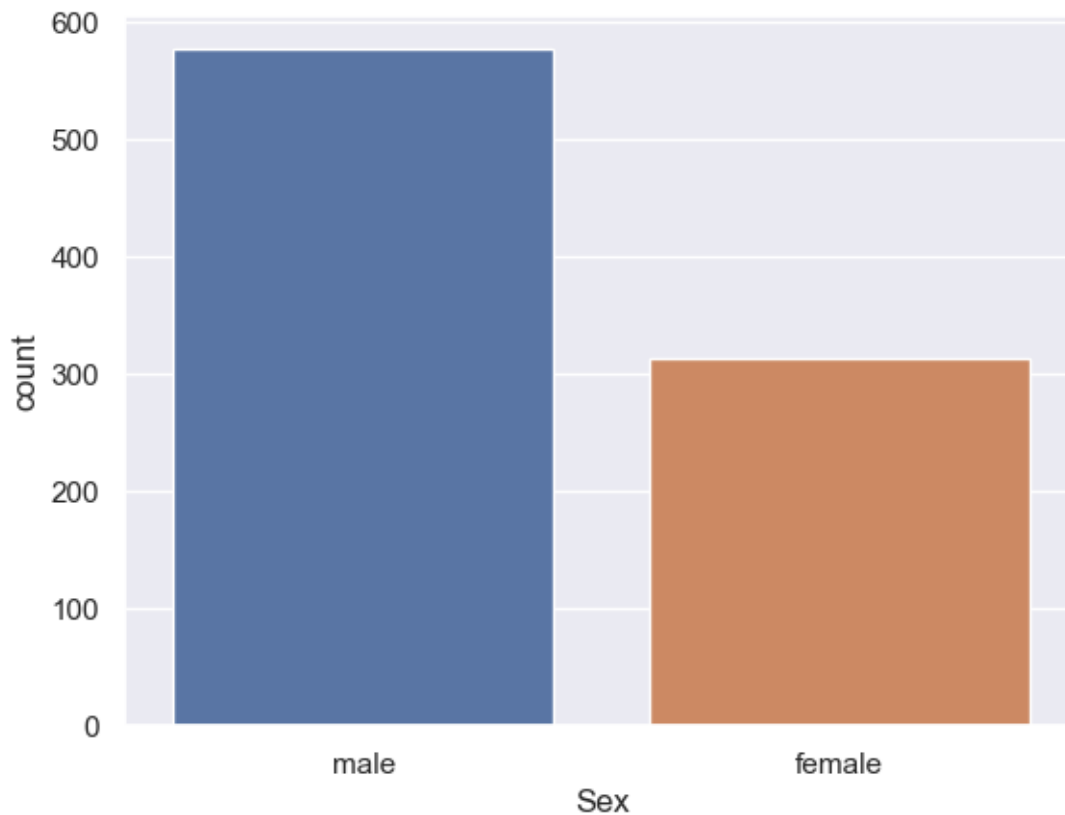
```
[30]: #count plot for sex column
sns.countplot(df['Sex'])
```

```
C:\Users\CHANDRA ADITYA\anaconda3\lib\site-packages\seaborn\_decorators.py:36:
```

FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

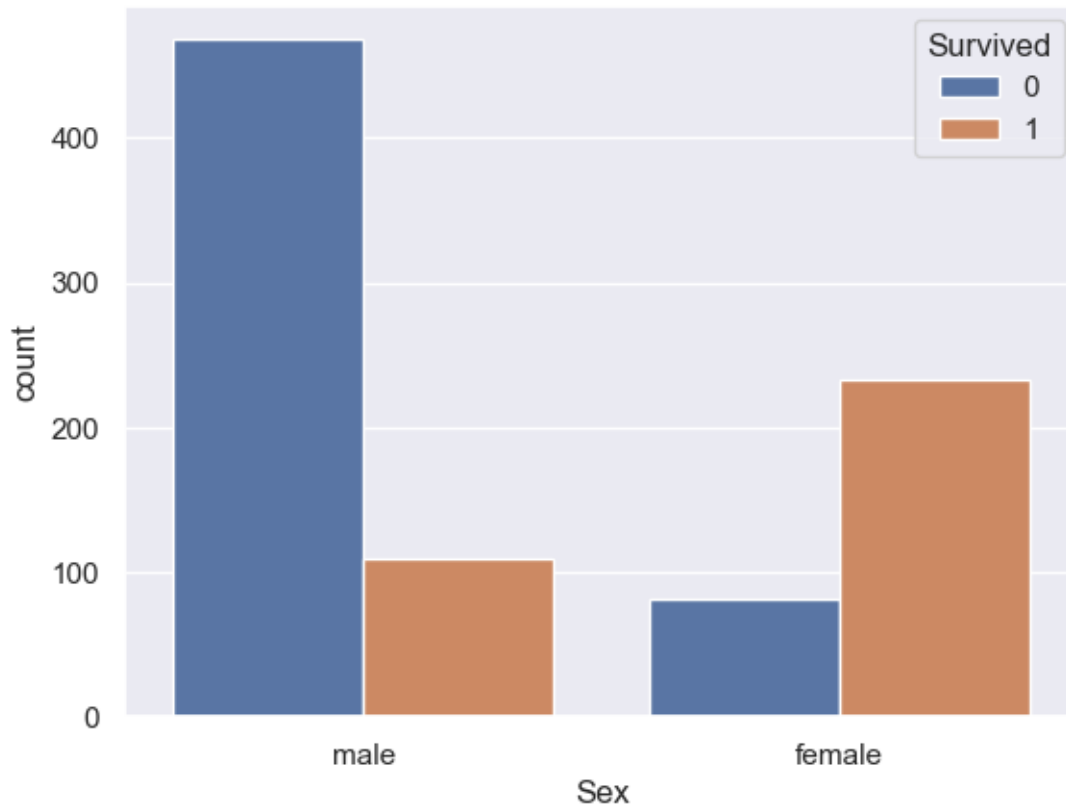
```
warnings.warn(
```

```
[30]: <AxesSubplot:xlabel='Sex', ylabel='count'>
```



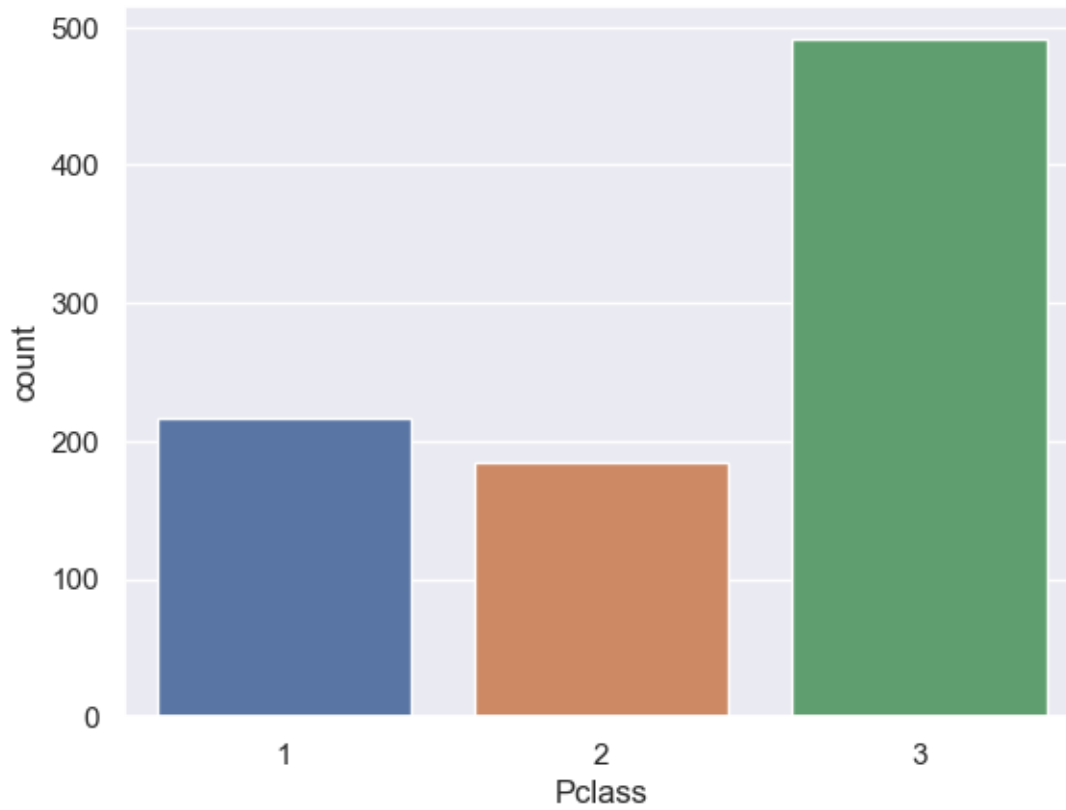
```
[31]: #Analysing gender wise survivals  
sns.countplot(x='Sex', hue='Survived', data=df)
```

```
[31]: <AxesSubplot:xlabel='Sex', ylabel='count'>
```



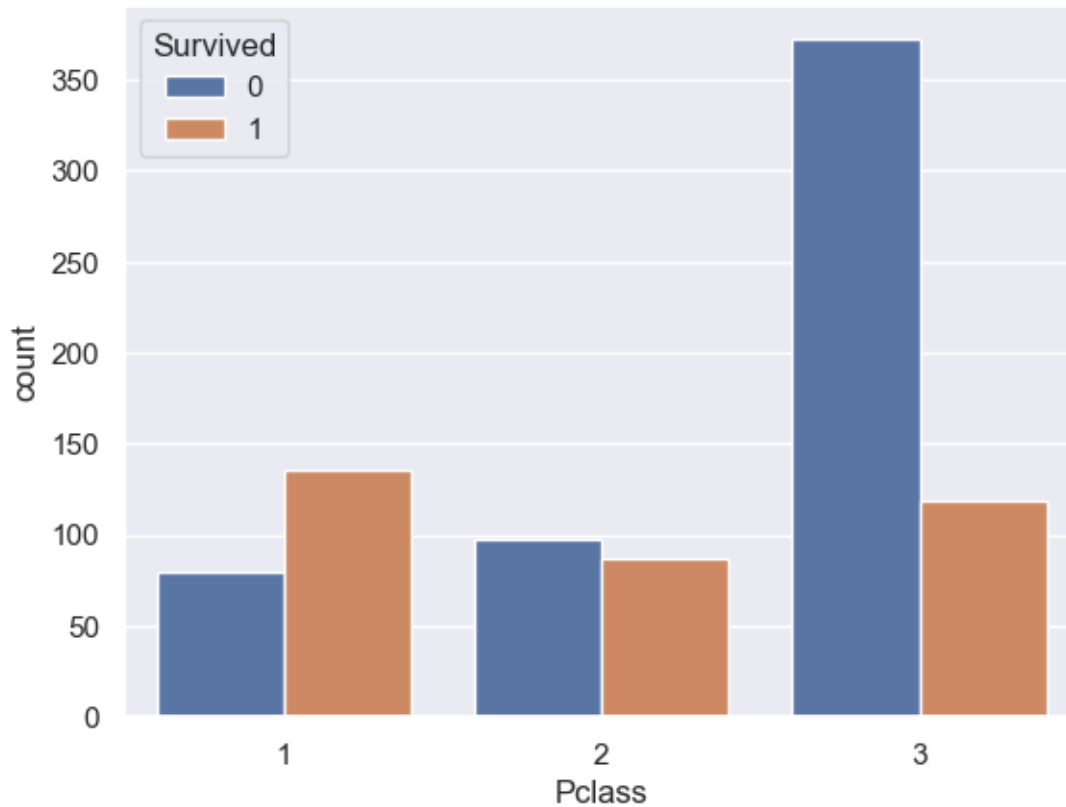
```
[32]: #count plot for "Pclass" column  
sns.countplot(x='Pclass', data=df)
```

```
[32]: <AxesSubplot:xlabel='Pclass', ylabel='count'>
```



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

```
[33]: <AxesSubplot:xlabel='Pclass', ylabel='count'>
```

Encode categorical columns/data

```
[34]: df['Sex'].value_counts()
```

```
[34]: male      577
      female    314
      Name: Sex, dtype: int64
```

```
[35]: df['Embarked'].value_counts()
```

```
[35]: S      646
      C      168
      Q       77
      Name: Embarked, dtype: int64
```

```
[37]: df.replace({'Sex':{'male':0, 'female':1}, 'Embarked':{'S':0, 'C':1, 'Q':2}})
```

```
[37]:   PassengerId  Survived  Pclass  \
0             1         0        3
1             2         1        1
2             3         1        3
```

3	4	1	1
4	5	0	3
..
886	887	0	2
887	888	1	1
888	889	0	3
889	890	1	1
890	891	0	3

	Name	Sex	Age	SibSp	\
0	Braund, Mr. Owen Harris	0	22.000000	1	
1	Cummings, Mrs. John Bradley (Florence Briggs Th...	1	38.000000	1	
2	Heikkinen, Miss. Laina	1	26.000000	0	
3	Futrelle, Mrs. Jacques Heath (Lily May Peel)	1	35.000000	1	
4	Allen, Mr. William Henry	0	35.000000	0	
..	
886	Montvila, Rev. Juozas	0	27.000000	0	
887	Graham, Miss. Margaret Edith	1	19.000000	0	
888	Johnston, Miss. Catherine Helen "Carrie"	1	29.699118	1	
889	Behr, Mr. Karl Howell	0	26.000000	0	
890	Dooley, Mr. Patrick	0	32.000000	0	

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

[891 rows x 11 columns]

```
[40]: X = df.drop(columns = ['PassengerId', 'Name', 'Ticket', 'Survived'], axis=1)
      Y = df['Survived']
```

```
[41]: print(X)
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	male	22.000000	1	0	7.2500	S
1	1	female	38.000000	1	0	71.2833	C
2	3	female	26.000000	0	0	7.9250	S
3	1	female	35.000000	1	0	53.1000	S

4	3	male	35.000000	0	0	8.0500	S
..	
886	2	male	27.000000	0	0	13.0000	S
887	1	female	19.000000	0	0	30.0000	S
888	3	female	29.699118	1	2	23.4500	S
889	1	male	26.000000	0	0	30.0000	C
890	3	male	32.000000	0	0	7.7500	Q

[891 rows x 7 columns]

[42]: `print(Y)`

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