In [7]: import pandas as pd import numpy as np import seaborn as sns

import matplotlib.pyplot as plt

titanic_data = pd.read_csv("tested.csv") In [8]:

In [9]: titanic_data

Out[9]:		Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cak
	0	892	0	3	Kelly, Mr. James	male	34.5	0	0	330911	7.8292	Ni
	1	893	1	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	7.0000	Ni
	2	894	0	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875	Ni
	3	895	0	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625	Ni
	4	896	1	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875	Ni
	•••	•••	•••	•••	•••	•••	•••	•••		•••	•••	
	413	1305	0	3	Spector, Mr. Woolf	male	NaN	0	0	A.5. 3236	8.0500	Ni
	414	1306	1	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	108.9000	C1
	415	1307	0	3	Saether, Mr. Simon Sivertsen	male	38.5	0	0	SOTON/O.Q. 3101262	7.2500	Ni
	416	1308	0	3	Ware, Mr. Frederick	male	NaN	0	0	359309	8.0500	Ni
	417	1309	0	3	Peter, Master. Michael J	male	NaN	1	1	2668	22.3583	Ni

418 rows × 12 columns

```
titanic_data.head()
In [10]:
             PassengerId Survived Pclass
                                            Name
                                                                               Ticket
                                                                                         Fare Cabin Em
Out[10]:
                                                      Sex Age SibSp Parch
                                          Kelly, Mr.
          0
                    892
                                0
                                       3
                                                     male 34.5
                                                                    0
                                                                          0
                                                                              330911
                                                                                       7.8292
                                                                                               NaN
                                             James
                                            Wilkes,
                                              Mrs.
          1
                    893
                                1
                                       3
                                            James
                                                   female 47.0
                                                                    1
                                                                              363272
                                                                                       7.0000
                                                                                               NaN
                                             (Ellen
                                            Needs)
                                            Myles,
                                               Mr.
          2
                                0
                                       2
                    894
                                                     male 62.0
                                                                    0
                                                                              240276
                                                                                       9.6875
                                                                                               NaN
                                           Thomas
                                            Francis
                                          Wirz, Mr.
          3
                    895
                                0
                                       3
                                                     male 27.0
                                                                    0
                                                                              315154
                                                                                       8.6625
                                                                                               NaN
                                             Albert
                                          Hirvonen,
                                              Mrs.
          4
                    896
                                1
                                       3 Alexander
                                                   female 22.0
                                                                    1
                                                                          1 3101298 12.2875
                                                                                               NaN
                                           (Helga E
                                          Lindqvist)
          titanic_data.shape
In [12]:
          (418, 12)
Out[12]:
In [13]:
          titanic_data.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 418 entries, 0 to 417
          Data columns (total 12 columns):
                              Non-Null Count Dtype
               Column
           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
               Ticket
           8
                              418 non-null
                                               object
           9
                                               float64
               Fare
                              417 non-null
               Cabin
                              91 non-null
                                               object
           10
               Embarked
                             418 non-null
                                               object
          dtypes: float64(2), int64(5), object(5)
          memory usage: 39.3+ KB
          titanic_data.isnull().sum()
In [14]:
```

```
PassengerId
                           0
Out[14]:
         Survived
                           0
         Pclass
                           0
         Name
                           0
         Sex
                           0
         Age
                          86
         SibSp
                           0
         Parch
                           0
         Ticket
                           0
         Fare
                           1
         Cabin
                         327
         Embarked
                           0
         dtype: int64
```

In [15]: titanic_data.drop(columns='Cabin',axis=1)

Out[15]:		Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Em
	0	892	0	3	Kelly, Mr. James	male	34.5	0	0	330911	7.8292	
	1	893	1	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	7.0000	
	2	894	0	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875	
	3	895	0	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625	
	4	896	1	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875	
	•••	•••				•••						
	413	1305	0	3	Spector, Mr. Woolf	male	NaN	0	0	A.5. 3236	8.0500	
	414	1306	1	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	108.9000	
	415	1307	0	3	Saether, Mr. Simon Sivertsen	male	38.5	0	0	SOTON/O.Q. 3101262	7.2500	
	416	1308	0	3	Ware, Mr. Frederick	male	NaN	0	0	359309	8.0500	
	417	1309	0	3	Peter, Master. Michael J	male	NaN	1	1	2668	22.3583	

418 rows × 11 columns

```
titanic_data['Fare'].fillna(titanic_data['Fare'].mode()[0],inplace=True)
In [23]:
           titanic_data.isnull().sum()
In [24]:
          PassengerId
                              0
Out[24]:
          Survived
                              0
          Pclass
                              0
          Name
                              0
          Sex
                              0
          Age
                              0
          SibSp
                              0
          Parch
                              0
          Ticket
                              0
          Fare
                              0
          Cabin
                           327
          Embarked
                              0
          dtype: int64
           titanic_data=titanic_data.drop(columns='Cabin',axis=1)
In [43]:
           titanic data.isnull().sum()
In [44]:
          PassengerId
                           0
Out[44]:
          Survived
                           0
          Pclass
                           0
          Name
                           0
          Sex
                           0
          Age
                           0
          SibSp
                           0
                           0
          Parch
          Ticket
                           0
                           0
          Fare
                           0
          Embarked
          dtype: int64
           titanic_data.describe()
In [45]:
                                              Pclass
                                                                                            Parch
Out[45]:
                  Passengerld
                                Survived
                                                            Sex
                                                                       Age
                                                                                 SibSp
                                                                                                         Fa
           count
                  418.000000 418.000000 418.000000 418.000000 418.000000 418.000000 418.000000 418.000000
           mean 1100.500000
                                0.363636
                                            2.265550
                                                       0.363636
                                                                  30.272590
                                                                              0.447368
                                                                                          0.392344
                                                                                                    35.56049
                  120.810458
                                0.481622
                                            0.841838
                                                       0.481622
                                                                  12.634534
                                                                              0.896760
                                                                                          0.981429
                                                                                                    55.85714
             std
                   892.000000
                                0.000000
                                            1.000000
                                                       0.000000
                                                                   0.170000
                                                                              0.000000
                                                                                          0.000000
                                                                                                     0.00000
            min
            25%
                   996.250000
                                0.000000
                                            1.000000
                                                       0.000000
                                                                  23.000000
                                                                              0.000000
                                                                                          0.000000
                                                                                                     7.89580
            50%
                1100.500000
                                0.000000
                                            3.000000
                                                       0.000000
                                                                  30.272590
                                                                              0.000000
                                                                                          0.000000
                                                                                                    14.4542(
            75%
                 1204.750000
                                1.000000
                                            3.000000
                                                       1.000000
                                                                  35.750000
                                                                              1.000000
                                                                                          0.000000
                                                                                                    31.47187
            max 1309.000000
                                1.000000
                                            3.000000
                                                       1.000000
                                                                  76.000000
                                                                              8.000000
                                                                                          9.000000 512.32920
           titanic_data["Survived"].value_counts()
```

Out[46]: 0 266 1 152

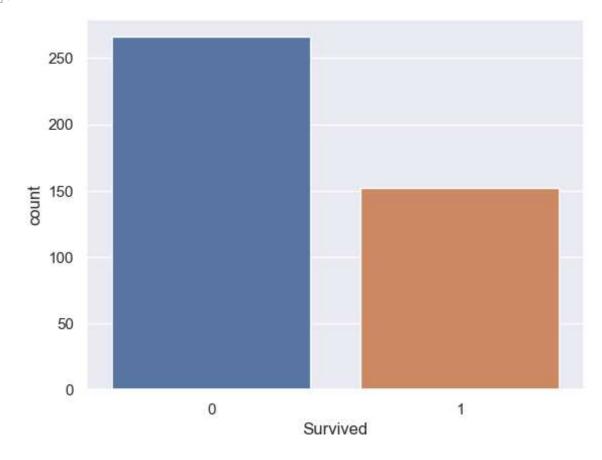
Name: Survived, dtype: int64

In [47]: sns.set()
 sns.countplot("Survived",data=titanic_data)

C:\Users\SAHITHI\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarnin g: 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 k eyword will result in an error or misinterpretation.

warnings.warn(

Out[47]: <AxesSubplot:xlabel='Survived', ylabel='count'>

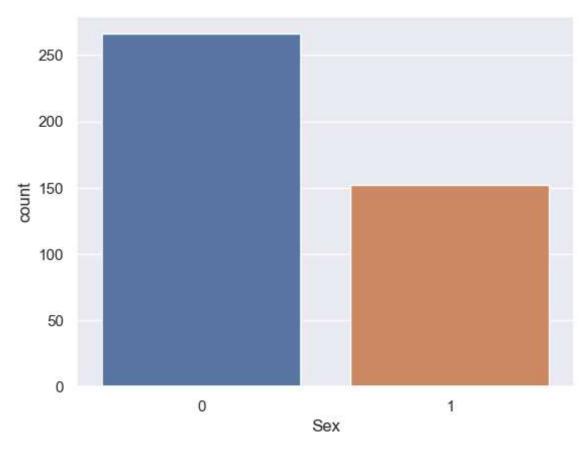


In [48]: sns.set()
sns.countplot("Sex",data=titanic_data)

C:\Users\SAHITHI\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarnin g: 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 k eyword will result in an error or misinterpretation.

warnings.warn(

Out[48]: <AxesSubplot:xlabel='Sex', ylabel='count'>

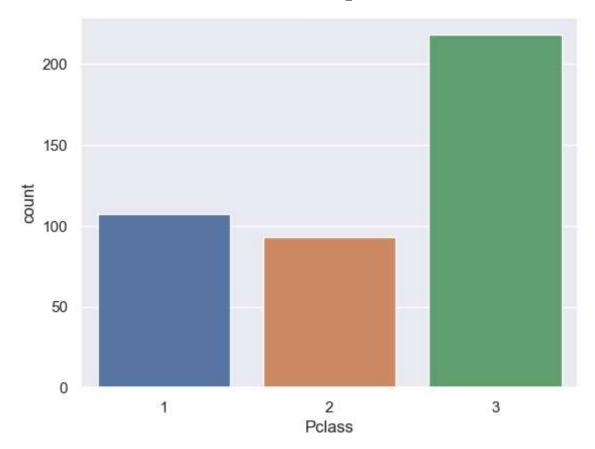


In [49]: sns.countplot("Pclass",data=titanic_data)

C:\Users\SAHITHI\anaconda3\lib\site-packages\seaborn\ decorators.py:36: FutureWarnin g: 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 \boldsymbol{k} eyword will result in an error or misinterpretation.

warnings.warn(

<AxesSubplot:xlabel='Pclass', ylabel='count'> Out[49]:



titanic_data.replace({"Sex":{"male":0,"female":1}, "Embarked":{"S":0,"C":1,"Q":2}},ing In [50]: titanic_data.head() In [51]: Name Sex Age SibSp Parch Out[51]: PassengerId Survived Pclass **Ticket** Fare Embarked Kelly, Mr. 3 0 892 0 0 34.5 330911 7.8292 2 James Wilkes, Mrs. 1 893 1 3 James 1 47.0 363272 7.0000 0 1 (Ellen Needs) Myles, 2 2 0 2 894 0 62.0 240276 9.6875 Thomas Francis Wirz, Mr. 3 895 3 0 27.0 315154 8.6625 0 Albert Hirvonen, Mrs. 4 896 1 3 Alexander 1 22.0 1 3101298 12.2875 0 (Helga E Lindqvist)

```
X = titanic_data.drop(columns=["Name","Ticket","PassengerId","Survived"],axis=1)
In [52]:
          y = titanic_data["Survived"]
In [53]:
                              Age SibSp Parch
                                                    Fare Embarked
Out[53]:
               Pclass Sex
            0
                   3
                        0 34.50000
                                       0
                                              0
                                                   7.8292
                                                                 2
                   3
                        1 47.00000
                                              0
                                                   7.0000
                                                                 0
            2
                   2
                        0 62.00000
                                       0
                                              0
                                                   9.6875
                                                                 2
            3
                   3
                        0 27.00000
                                              0
                                                   8.6625
                                                                 0
            4
                   3
                        1 22.00000
                                       1
                                              1
                                                  12.2875
                                                                 0
                   •••
          413
                   3
                        0 30.27259
                                       0
                                              0
                                                   8.0500
                                                                 0
          414
                        1 39.00000
                                              0 108.9000
                                                                 1
          415
                   3
                        0 38.50000
                                                                 0
                                       0
                                              0
                                                   7.2500
          416
                   3
                        0 30.27259
                                       0
                                              0
                                                   8.0500
                                                                 0
          417
                   3
                        0 30.27259
                                       1
                                              1
                                                  22.3583
                                                                 1
         418 rows × 7 columns
In [54]:
                 0
Out[54]:
                 1
          2
                 0
          3
                 0
          4
                 1
          413
                 0
          414
                 1
          415
                 0
          416
                 0
          417
          Name: Survived, Length: 418, dtype: int64
In [55]:
          from sklearn.model_selection import train_test_split
          from sklearn.linear model import LogisticRegression
          from sklearn.metrics import accuracy_score
In [56]:
          X_train,X_test, y_train,y_test = train_test_split(X,y,test_size=0.2,random_state=2)
In [57]:
          print(X.shape,X_train.shape,X_test.shape)
          (418, 7) (334, 7) (84, 7)
          model=LogisticRegression()
In [59]:
```

In [60]:

model.fit(X_train,y_train)

```
C:\Users\SAHITHI\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:814: C
      onvergenceWarning: lbfgs failed to converge (status=1):
      STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
      Increase the number of iterations (max iter) or scale the data as shown in:
         https://scikit-learn.org/stable/modules/preprocessing.html
      Please also refer to the documentation for alternative solver options:
         https://scikit-learn.org/stable/modules/linear model.html#logistic-regression
       n_iter_i = _check_optimize_result(
      LogisticRegression()
Out[60]:
      X_train_prediction=model.predict(X_train)
In [61]:
      print(X train prediction)
In [62]:
      1\ 1\ 1\ 0\ 0\ 0\ 1\ 0\ 0\ 0\ 1\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 0\ 0\ 1\ 0\ 1\ 0\ 1\ 1\ 1\ 0\ 1
       0\;1\;0\;0\;0\;0\;0\;0\;0\;0\;0\;0\;1\;1\;0\;1\;0\;1\;0\;0\;0\;0\;0\;0\;1\;0\;1\;1\;1\;0\;1\;0\;1\;0
       1]
In [63]:
      training data accuracy=accuracy score(y train,X train prediction)
      print("accuracy score:",training_data_accuracy)
      accuracy score: 1.0
      X_test_predict=model.predict(X_test)
In [64]:
In [66]:
      print(X test predict)
      0 1 1 0 1 0 0 0 0 0]
      test_data_accuracy=accuracy_score(y_test,X_test_predict)
In [68]:
      print("accuracy score:",test data accuracy)
      accuracy score: 1.0
In [ ]:
```