In [1]:	<pre>import numpy as np import pandas as pd import seaborn as sns import matplotlib.pyplot as plt from sklearn.model_selection import train_test_split from sklearn.linear_model import LogisticRegression from sklearn.metrics import accuracy_score</pre>
In [3]: In [4]: Out[4]:	titanic_data=pd.read_csv('train.csv')    PassengerId   Survived   Pclass   Name   Sex   Age   SibSp   Parch   Ticket   Fare   Cabin   Embarked
In [6]: Out[6]: In [7]:	4 5 0 3 Allen, Mr. William Henry male 35.0 0 0 373450 8.0500 NaN S  titanic_data.shape  (891, 12)  titanic_data.info()
	<pre>cclass 'pandas.core.frame.DataFrame'&gt; RangeIndex: 891 entries, 0 to 890  Data columns (total 12 columns):     # Column</pre>
In [8]: Out[8]:	memory usage: 83.7+ KB  titanic_data.isnull().sum()  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
In [9]: In [11]: In [12]:	<pre>titanic_data= titanic_data.drop(columns='Cabin', axis=1)  titanic_data['Age'].fillna(titanic_data['Age'].mean(),inplace=True)  print(titanic_data['Embarked'].mode()) 0    S  Total add to the result is at the control of the con</pre>
In [13]: In [15]:	<pre>Name: Embarked, dtype: object  print(titanic_data['Embarked'].mode()[0])  S  titanic_data['Embarked'].fillna(titanic_data['Embarked'].mode()[0],inplace=True)  titanic_data.isnull().sum()</pre>
In [16]: Out[16]:	PassengerId 0 Survived 0 Pclass 0 Name 0 Sex 0 Age 0 SibSp 0 Parch 0 Ticket 0 Fare 0 Embarked 0 dtype: int64
In [17]: Out[17]:	PassengerId         Survived         Pclass         Age         SibSp         Parch         Fare           count         891.000000         891.000000         891.000000         891.000000         891.000000         891.00000         891.000000           mean         446.000000         0.383838         2.308642         29.699118         0.523008         0.381594         32.204208           std         257.353842         0.486592         0.836071         13.002015         1.102743         0.806057         49.693429           min         1.000000         0.000000         1.000000         0.420000         0.000000         0.000000         7.910400           25%         223.500000         0.000000         3.000000         29.699118         0.000000         0.000000         1.4454200           75%         668.500000         1.000000         3.000000         8.000000         6.000000         512.329200
In [18]: Out[18]: In [19]:	titanic_data['Survived'].value_counts()  0    549 1    342 Name: Survived, dtype: int64  sns.countplot('Survived', data=titanic_data)
Out[19]:	C:\Users\Lenovo\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 misinte rpretation.  warnings.warn( <axessubplot:xlabel='survived', ylabel="count"></axessubplot:xlabel='survived',>
	500 - 400 - 200 - 100 - 100 - 1 Survived
<pre>In [20]: Out[20]: In [22]:</pre>	titanic_data['Sex'].value_counts()  male 577 female 314 Name: Sex, dtype: int64  sns.countplot('Survived', data=titanic_data)
Out[22]:	C:\Users\Lenovo\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 misinte rpretation.  warnings.warn( <axessubplot:xlabel='survived', ylabel="count"></axessubplot:xlabel='survived',>
	500 - 400 - 100 -
In [23]:	<pre>sns.countplot('Sex',hue='Survived', data=titanic_data)  C:\Users\Lenovo\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 misinte rpretation.    warnings.warn( <axessubplot:xlabel='sex', ylabel="count"></axessubplot:xlabel='sex',></pre>
Out[23]:	400 - Survived 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
In [24]:	Sex  sns.countplot('Pclass',data=titanic_data)  C:\Users\Lenovo\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 misinte rpretation.
Out[24]:	warnings.warn( <axessubplot:xlabel='pclass', ylabel="count">  500  400  200  100  Pclass  Pclass</axessubplot:xlabel='pclass',>
In [25]: Out[25]:	sns.countplot('Pclass', hue='Survived', data=titanic_data)  C:\Users\Lenonvo\anaconda3\lib\site-packages\seaborn\_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version on the only valid positional argument will be 'data', and passing other arguments without an explicit keyword will result in an error or misinte repretation.  warnings.warn( <pre></pre>
In [26]: Out[26]:	1 2 3 Pclass  titanic_data['Sex'].value_counts()  male 577 formale 214
Out[26]:  In [27]:  Out[27]:	female 314 Name: Sex, dtype: int64  titanic_data['Embarked'].value_counts()  S 646 C 168 Q 77 Name: Embarked, dtype: int64
<pre>In [29]: In [30]: Out[30]:</pre>	titanic_data.replace({'Sex':{'male':0,'female':1}, 'Embarked':{'S':0,'C':1,'Q':2}},inplace=True)  titanic_data.head()  Passengerld Survived Pclass Name Sex Age SibSp Parch Ticket Fare Embarked  1 0 3 Braund, Mr. Owen Harris 0 22.0 1 0 A/5 21171 7.2500 0
-	1       2       1       1       Cumings, Mrs. John Bradley (Florence Briggs Th       1       38.0       1       0       PC 17599       71.2833       1         2       3       1       3       Heikkinen, Miss. Laina       1       26.0       0       0       STON/O2. 3101282       7.9250       0         3       4       1       1       Futrelle, Mrs. Jacques Heath (Lily May Peel)       1       35.0       1       0       113803       53.1000       0         4       5       0       3       Allen, Mr. William Henry       0       35.0       0       0       373450       8.0500       0
In [32]:	<pre>x=titanic_data.drop(columns=['PassengerId', 'Name', 'Ticket', 'Survived'], axis=1) y=titanic_data['Survived']  print(x)  Pclass Sex</pre>
In [34]:	889
	1 1 2 1 3 1 3 1 4 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
In [36]: In [37]:	889 1 890 0 Name: Survived, Length: 891, dtype: int64  x_train,x_test, y_train,y_test= train_test_split(x,y, test_size =0.2, random_state=2)  print(x.shape,x_train.shape,x_test.shape)  (891, 7) (712, 7) (179, 7)
In [39]: In [41]:	<pre>model=LogisticRegression()  model.fit(x_train,y_train)  C:\Users\Lenovo\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:814: ConvergenceWarning: lbfgs failed to converge (status=1): STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.</pre>
Out[41]: In [42]:	<pre>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()  x_train_prediction= model.predict(x_train)</pre>
In [42]: In [43]:	<pre>print(x_train_prediction)  [0 1 0 0 0 0 0 1 0 0 0 1 0 0 1 0 1 0 1 0</pre>
	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$
In [44]:	0 0 0 0 1 0 0 1 0 1 1 0 0 1 0 0 1 0 0 1 0 1 1 1 0 0 1 1 1 0 0 1 1 1 0 0 1 0 1 0
In [44]: In [45]: In [46]: In [48]:	<pre>print('Accuracy score of training data:', training_data_accuracy)  Accuracy score of training data: 0.8075842696629213  x_test_prediction= model.predict(x_test)  test_data_accuracy = accuracy_score(y_test,x_test_prediction)</pre>
In [49]:	print('Accuracy score of test data:', test_data_accuracy)  Accuracy score of test data: 0.7821229050279329