In [1]: import numpy as np import pandas as pd In [2]: df=pd.read_csv("Titanic.csv") In [3]: df.head() Out[3]: PassengerId Survived Pclass Name Sex Age SibSp Parch **Ticket Fare** Braund, 0 0 1 3 1 0 Mr. Owen male 22.0 7.250021171 Harris Cumings, Mrs. John Bradley 1 2 female 38.0 1 0 PC 17599 71.2833 1 1 (Florence Briggs Th... Heikkinen, STON/O2. 2 3 1 3 0 Miss. female 26.0 7.9250 3101282 Laina Futrelle, Mrs. Jacques 3 4 1 1 female 35.0 1 0 113803 53.1000 Heath (Lily May Peel) Allen, Mr. 5 0 3 William male 35.0 0 373450 8.0500 Henry In [4]: 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					
dtypos, $flort(4/2)$ $int(4/5)$ $object(5)$								

dtypes: float64(2), int64(5), object(5)

memory usage: 83.7+ KB

In [5]: df.describe()

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	Passengerld	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

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In [6]: df.drop(columns=['PassengerId','Name','Ticket','Cabin'],inplace=True)

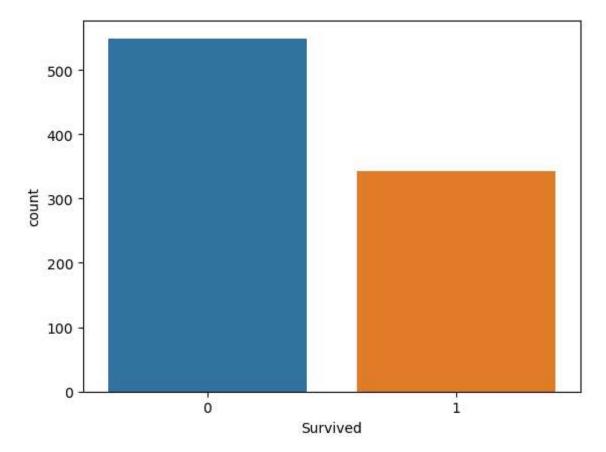
In [7]: df.head()

Out[7]:

	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	0	3	male	22.0	1	0	7.2500	S
1	1	1	female	38.0	1	0	71.2833	С
2	1	3	female	26.0	0	0	7.9250	S
3	1	1	female	35.0	1	0	53.1000	S
4	0	3	male	35.0	0	0	8.0500	S

```
df['Survived'].value_counts()
 In [8]:
 Out[8]: Survived
              549
              342
         Name: count, dtype: int64
In [11]: import seaborn as sns
         sns.countplot(x=df['Sex'])
Out[11]: <Axes: xlabel='Sex', ylabel='count'>
           600
           500
           400
        300
           200
           100
                               male
                                                                female
                                                Sex
In [12]: sns.countplot(x=df['Survived'])
```

Out[12]: <Axes: xlabel='Survived', ylabel='count'>



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

In [14]: df.head()

Out[14]:		Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
	0	0	3	1	22.0	1	0	7.2500	S
	1	1	1	0	38.0	1	0	71.2833	С
	2	1	3	0	26.0	0	0	7.9250	S
	3	1	1	0	35.0	1	0	53.1000	S
	4	0	3	1	35.0	0	0	8.0500	S

```
In [15]: x=df[['Sex', 'Pclass']]
y=df['Survived']
```

In [16]: 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

In [17]: from sklearn.linear_model import LogisticRegression
 log= LogisticRegression(random_state = 42)
 log.fit(X_train, y_train)

```
Out[17]: ▼
                  LogisticRegression
        LogisticRegression(random_state=42)
In [18]: pred = log.predict(X_test)
         pred
1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0,
                1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1,
                0, 0, 0, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1,
               0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0,
               1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0,
               0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 1,
               0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0,
                1, 1, 1], dtype=int64)
In [19]: print(y test)
       709
              1
       439
              0
       840
              0
       720
              1
       39
              1
             . .
       433
              0
       773
              a
       25
              1
       84
              1
       10
              1
       Name: Survived, Length: 179, dtype: int64
In [20]: from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_sco
         accuracy = accuracy_score(y_test, pred)
         print("Accuracy:", accuracy)
         pre = precision_score(y_test , pred , average = "micro")
         print("The precision is:{}".format(pre))
         rec = recall_score(y_test , pred , average = "micro")
         print("The recall is:{}".format(rec))
         f1 = f1 score(y test , pred , average = "micro")
         print("The f1 score is:{}".format(f1))
       Accuracy: 0.7821229050279329
       The precision is:0.7821229050279329
       The recall is:0.7821229050279329
       The f1_score is:0.7821229050279329
In [21]: import warnings
         warnings.filterwarnings("ignore")
         res= log.predict([[0,0]])
         if(res==0):
             print("Not survived")
```

```
else:
    print("survived")

survived

In [22]: import warnings
    warnings.filterwarnings("ignore")
    res= log.predict([[2,1]])

    if(res==0):
        print("Not survived")
    else:
        print("survived")

    Not survived

In []:
```