```
In [1]: import warnings warnings.filterwarnings("ignore")

import numpy as np import pandas as pd import seaborn as sns import matplotlib.pyplot as plt
```

In [2]: df = pd.read_csv("titanic_train.csv")
 df.head(20)

Out[2]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	С
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	s
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	s
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	s
5	6	0	3	Moran, Mr. James	male	NaN	0	0	330877	8.4583	NaN	Q
6	7	0	1	McCarthy, Mr. Timothy J	male	54.0	0	0	17463	51.8625	E46	S
7	8	0	3	Palsson, Master. Gosta Leonard	male	2.0	3	1	349909	21.0750	NaN	s
8	9	1	3	Johnson, Mrs. Oscar W (Elisabeth Vilhelmina Berg)	female	27.0	0	2	347742	11.1333	NaN	S
9	10	1	2	Nasser, Mrs. Nicholas (Adele Achem)	female	14.0	1	0	237736	30.0708	NaN	С
10	11	1	3	Sandstrom, Miss. Marguerite Rut	female	4.0	1	1	PP 9549	16.7000	G6	S
11	12	1	1	Bonnell, Miss. Elizabeth	female	58.0	0	0	113783	26.5500	C103	s
12	13	0	3	Saundercock, Mr. William Henry	male	20.0	0	0	A/5. 2151	8.0500	NaN	s
13	14	0	3	Andersson, Mr. Anders Johan	male	39.0	1	5	347082	31.2750	NaN	S
14	15	0	3	Vestrom, Miss. Hulda Amanda Adolfina	female	14.0	0	0	350406	7.8542	NaN	s
15	16	1	2	Hewlett, Mrs. (Mary D Kingcome)	female	55.0	0	0	248706	16.0000	NaN	S
16	17	0	3	Rice, Master. Eugene	male	2.0	4	1	382652	29.1250	NaN	Q
17	18	1	2	Williams, Mr. Charles Eugene	male	NaN	0	0	244373	13.0000	NaN	S
18	19	0	3	Vander Planke, Mrs. Julius (Emelia Maria Vande	female	31.0	1	0	345763	18.0000	NaN	S
19	20	1	3	Masselmani, Mrs. Fatima	female	NaN	0	0	2649	7.2250	NaN	С

```
In [3]: df["Name"].to_list()
Out[3]: ['Braund, Mr. Owen Harris',
           'Cumings, Mrs. John Bradley (Florence Briggs Thayer)',
          'Heikkinen, Miss. Laina',
          'Futrelle, Mrs. Jacques Heath (Lily May Peel)', 'Allen, Mr. William Henry',
          'Moran, Mr. James',
          'McCarthy, Mr. Timothy J',
          'Palsson, Master. Gosta Leonard',
          'Johnson, Mrs. Oscar W (Elisabeth Vilhelmina Berg)',
'Nasser, Mrs. Nicholas (Adele Achem)',
          'Sandstrom, Miss. Marguerite Rut',
          'Bonnell, Miss. Elizabeth',
          'Saundercock, Mr. William Henry',
          'Andersson, Mr. Anders Johan',
          'Vestrom, Miss. Hulda Amanda Adolfina',
          'Hewlett, Mrs. (Mary D Kingcome) ',
          'Rice, Master. Eugene',
          'Williams, Mr. Charles Eugene',
          'Vander Planke, Mrs. Julius (Emelia Maria Vandemoortele)',
```

```
In [4]: df["Name"].unique()
Out[4]: array(['Braund, Mr. Owen Harris',
                   Cumings, Mrs. John Bradley (Florence Briggs Thayer)',
                   'Heikkinen, Miss. Laina',
                   'Futrelle, Mrs. Jacques Heath (Lily May Peel)',
                  'Allen, Mr. William Henry', 'Moran, Mr. James', 'McCarthy, Mr. Timothy J', 'Palsson, Master. Gosta Leonard',
                   'Johnson, Mrs. Oscar W (Elisabeth Vilhelmina Berg)',
                   'Nasser, Mrs. Nicholas (Adele Achem)',
                  'Sandstrom, Miss. Marguerite Rut', 'Bonnell, Miss. Elizabeth', 'Saundercock, Mr. William Henry', 'Andersson, Mr. Anders Johan',
                   'Vestrom, Miss. Hulda Amanda Adolfina',
                   'Hewlett, Mrs. (Mary D Kingcome) ', 'Rice, Master. Eugene',
                   'Williams, Mr. Charles Eugene',
                   'Vander Planke, Mrs. Julius (Emelia Maria Vandemoortele)',
                  'Masselmani, Mrs. Fatima', 'Fynney, Mr. Joseph J',
'Beesley, Mr. Lawrence', 'McGowan, Miss. Anna "Annie"'
                   'Sloper, Mr. William Thompson', 'Palsson, Miss. Torborg Danira',
                   'Asplund, Mrs. Carl Oscar (Selma Augusta Emilia Johansson)',
                  'Emir, Mr. Farred Chehab', 'Fortune, Mr. Charles Alexander',
```

Goal:- Given the features we need to predict if passenger will survived or not

EDA

```
In [5]: df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 891 entries, 0 to 890
        Data columns (total 12 columns):
             Column
                          Non-Null Count
                                          Dtvpe
         0
             PassengerId 891 non-null
                                          int64
             Survived
                          891 non-null
             Pclass
                          891 non-null
                                          int64
                          891 non-null
         3
             Name
                                          object
         4
             Sex
                          891 non-null
                                          object
                          714 non-null
                                          float64
             Age
         6
             SibSp
                          891 non-null
                                          int64
                          891 non-null
                                          int64
             Parch
         8
                          891 non-null
             Ticket
                                          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
In [6]: df.isna().sum()
Out[6]: PassengerId
        Survived
                         0
        Pclass
                         0
        Name
        Sex
                         0
        Age
        SibSp
        Parch
        Ticket
                         0
        Fare
                         а
        Cabin
                       687
        Embarked
        dtype: int64
In [7]: df["Survived"].value_counts()
Out[7]: 0
             549
             342
        Name: Survived, dtype: int64
```

```
Titanic Logistic - Jupyter Notebook
 In [8]: sns.countplot(data=df, x="Survived")
 Out[8]: <AxesSubplot: xlabel='Survived', ylabel='count'>
             500
             400
           300
Tig
             200
            100
                           ó
                                                  i
                                    Survived
 In [9]: sns.countplot(data=df, x="Survived", palette="RdBu_r")
 Out[9]: <AxesSubplot: xlabel='Survived', ylabel='count'>
             500
             400
           300
1
             200
             100
                           ò
                                                  i
                                    Survived
In [10]: plt.pie(df["Survived"].value_counts(), labels=["Die", "Survived"], colors=["red", "green"], autopct="%1.1f%%", radius=1.5)
          plt.show()
          4
                   Die
                                    Survived
In [11]: df["Sex"].value_counts()
Out[11]: male
```

female

314 Name: Sex, dtype: int64

```
In [12]: sns.countplot(data=df,x="Sex")
Out[12]: <AxesSubplot: xlabel='Sex', ylabel='count'>
             500
             400
           300
mt
             200
             100
              0
                          male
                                                 female
                                       Sex
In [13]: plt.pie(df["Sex"].value_counts(), labels=["Male","Female"], colors=["red","green"], autopct="%1.1f%%",radius=1.5)
plt.show()
                 Male
                                      Female
In [14]: sns.countplot(data=df,x="Survived",hue="Sex")
Out[14]: <AxesSubplot: xlabel='Survived', ylabel='count'>
                                                        Sex
                                                         male
             400
             300
             200
             100
              0
                                     Survived
In [15]: df["Pclass"].value_counts()
Out[15]: 3
               491
               216
               184
```

Name: Pclass, dtype: int64

```
In [16]: sns.countplot(data=df, x="Survived",hue="Pclass",palette="rainbow")
Out[16]: <AxesSubplot: xlabel='Survived', ylabel='count'>
             350
                                                          1
2
             300
                                                          ____3
             250
           15 200
8
             150
             100
              50
                                     Survived
In [17]: plt.pie(df["Pclass"].value_counts(),explode=(0,0.1,0),shadow=True,labels=["Lower Class","First Class","Second Class"], colors=["rollower Class"].
          plt.show()
         4
                  Lower Class
                                              Second Class
            First Class
In [18]: df["SibSp"].value_counts()
Out[18]: 0
                608
                209
          2
4
                28
                18
          3
                16
          8
          Name: SibSp, dtype: int64
In [19]: sns.countplot(data=df,x="SibSp")
Out[19]: <AxesSubplot: xlabel='SibSp', ylabel='count'>
             600
             500
             400
           300
Mit
             200
             100
```

```
In [20]: sns.countplot(data=df,x="Survived",hue="SibSp")
Out[20]: <AxesSubplot: xlabel='Survived', ylabel='count'>
                                                           SibSp
0
             350
             300
             250
           ting 200
             150
             100
              50
                                      Survived
In [21]: sns.countplot(data=df,x="Sex",hue="SibSp")
Out[21]: <AxesSubplot: xlabel='Sex', ylabel='count'>
                                                           SibSp
0
             400
             300
           8 <sub>200</sub>
             100
                           male
                                                  female
In [22]: df["Parch"].value_counts()
Out[22]: 0
                678
                118
          1
2
5
                80
          3
          4
                  4
          Name: Parch, dtype: int64
In [23]: sns.countplot(data=df,x="Parch")
Out[23]: <AxesSubplot: xlabel='Parch', ylabel='count'>
              700
             600
             500
             400
             300
             200
             100
                                       Parch
```

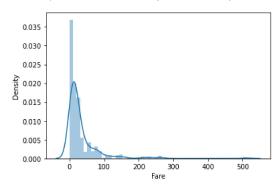
In [24]: df.describe()

Out[24]:

	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

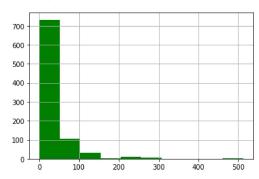
In [25]: sns.distplot(df["Fare"])

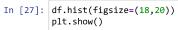
Out[25]: <AxesSubplot: xlabel='Fare', ylabel='Density'>

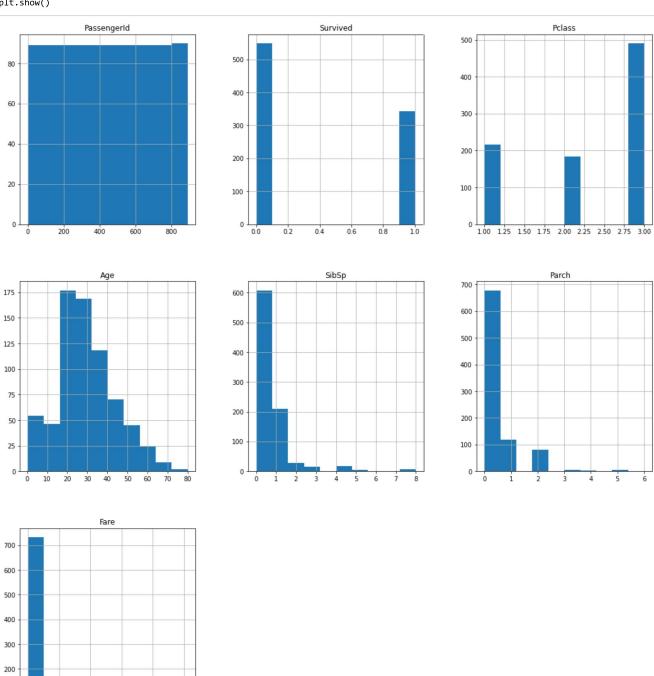


In [26]: df["Fare"].hist(color="green")

Out[26]: <AxesSubplot: >







0 -

100

200

300

400

500

In [28]: sns.pairplot(df,hue ="Survived")

Out[28]: <seaborn.axisgrid.PairGrid at 0x207ff4056f0>



In [29]: sns.heatmap(df.corr(),annot=True)

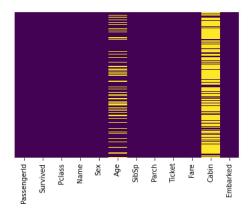
Out[29]: <AxesSubplot: >



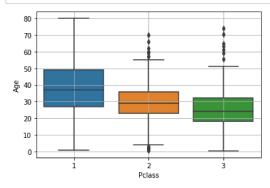
Data Cleaning And Preprocessing

```
In [31]: df.isna().sum()
Out[31]: PassengerId
         Survived
                          0
         Pclass
                          0
         Name
                          0
         Sex
                          0
         Age
                        177
         SibSp
                          0
         Parch
         Ticket
                          0
         Fare
                          0
         Cabin
                        687
         Embarked
                          2
         dtype: int64
In [32]: sns.heatmap(df.isnull(),yticklabels=False, cbar=False, cmap="viridis")
```

Out[32]: <AxesSubplot: >



```
In [33]: sns.boxplot(x="Pclass",y="Age",data=df)
plt.grid()
```



```
In [34]: def fillage(col):
    age=col[0]
    pclass=col[1]

if(pd.isnull(age)):
    if(pclass=1):
        return 38
    elif pclass=2:
        return 29
    else:
        return 25
else:
    return age
```

```
In [35]: df["Age"]=df[["Age","Pclass"]].apply(fillage,axis=1)
```

```
In [36]: sns.heatmap(df.isnull(),yticklabels=False, cbar=False, cmap="viridis")
Out[36]: <AxesSubplot: >
In [37]: df.drop("Cabin",axis=1,inplace=True)
In [38]: sns.heatmap(df.isnull(),yticklabels=False, cbar=False, cmap="viridis")
Out[38]: <AxesSubplot: >
In [39]: df.isnull().sum()
Out[39]: PassengerId
                        0
         Survived
                        0
         Pclass
                        0
         Name
                        0
         Sex
                        0
                        0
         Age
                        0
         SibSp
         Parch
         Ticket
                        0
         Fare
                        0
         Embarked
         dtype: int64
In [40]: df.dropna(inplace=True)
In [41]: df.isnull().sum()
Out[41]: PassengerId
         Survived
                        0
         Pclass
                        0
                        0
         Name
         Sex
                        0
         Age
                        0
         SibSp
                        0
                        0
         Parch
         Ticket
                        0
         Fare
         Embarked
         dtype: int64
```

```
In [42]: df.info()
          <class 'pandas.core.frame.DataFrame'>
          Int64Index: 889 entries, 0 to 890
          Data columns (total 11 columns):
               Column
                              Non-Null Count
                                               Dtype
               PassengerId
                              889 non-null
                                                int64
               Survived
                              889 non-null
                                                int64
                              889 non-null
           2
               Pclass
                                                int64
               Name
                              889 non-null
                                                object
               Sex
                              889 non-null
                                                object
                              889 non-null
                                                float64
               Age
                              889 non-null
               SibSp
                                                int64
               Parch
                              889 non-null
                                                int64
               Ticket
                              889 non-null
                                                object
                              889 non-null
                                                float64
               Fare
           10 Embarked
                              889 non-null
                                                object
          dtypes: float64(2), int64(5), object(4)
          memory usage: 83.3+ KB
In [43]: df.head()
Out[43]:
              Passengerld Survived Pclass
                                                                                    Sex Age SibSp Parch
                                                                            Name
                                                                                                                      Ticket
                                                                                                                               Fare Embarked
                                                                                                                                            s
           0
                                                                                         22 0
                                                                                                                   A/5 21171
                                                                                                                             7 2500
                                                              Braund, Mr. Owen Harris
                                                                                    male
                                                                                                                                            С
                       2
                                                                                                         0
                                                                                                                   PC 17599 71.2833
                                       1 Cumings, Mrs. John Bradley (Florence Briggs Th...
                                                                                  female
                                                                                         38.0
           2
                                                                                                           STON/O2. 3101282
                       3
                                       3
                                                                                                         0
                                                                                                                                            S
                                                               Heikkinen, Miss, Laina female
                                                                                         26.0
                                                                                                  0
                                                                                                                             7.9250
                                                                                                         0
                                                                                                                     113803 53.1000
                                                                                                                                            s
                                1
                                       1
                                              Futrelle, Mrs. Jacques Heath (Lily May Peel) female
                                                                                         35.0
                                                                                                  1
                                       3
                                                              Allen, Mr. William Henry
                                                                                                                     373450
                                                                                                                             8.0500
                                                                                                                                            S
                                                                                         35.0
                                                                                                  0
                                                                                   male
In [44]: | df.drop(["PassengerId","Name","Ticket"],axis=1,inplace=True)
In [45]: df.head()
Out[45]:
              Survived Pclass
                                Sex Age SibSp Parch
                                                         Fare Embarked
           0
                                    22.0
                                                        7.2500
                                                                      s
                               male
                                                       71.2833
                                                                      С
                              female
                                    38.0
                                    26.0
                                                        7.9250
                                                                      s
                             female
                                    35.0
                                                       53.1000
                                                                      s
                               male 35.0
                                                        8.0500
                                                                      S
```

Data Preprocessing

```
In [52]: from sklearn.preprocessing import OrdinalEncoder
         oe = OrdinalEncoder()
         x[["Sex","Embarked"]] = oe.fit\_transform(x[["Sex","Embarked"]])
In [53]: x
Out[53]:
              Pclass Sex Age SibSp Parch
                                             Fare Embarked
            0
                      1.0 22.0
                                           7.2500
                                                        2.0
                     0.0 38.0
                                        0 71.2833
                                                        0.0
                                  0
                                          7.9250
                                                        2.0
                   3 0.0 26.0
                                        0
                     0.0 35.0
                                        0 53.1000
                                                        2.0
                                           8.0500
                     1.0 35.0
                                                        2.0
          886
                   2 1.0 27.0
                                  0
                                        0 13.0000
                                                        2.0
          887
                     0.0 19.0
                                        0 30.0000
                                                        2.0
          888
                     0.0 25.0
                                        2 23.4500
                                                        2.0
          889
                      1.0 26.0
                                        0 30.0000
                                                        0.0
                   3 1.0 32.0
                                                        1.0
         889 rows × 7 columns
In [54]: x.Sex.value_counts()
Out[54]: 1.0
         0.0
                312
         Name: Sex, dtype: int64
In [55]: x.Embarked.value_counts()
Out[55]: 2.0
                644
         0.0
                168
         1.0
                 77
         Name: Embarked, dtype: int64
         Model Building
In [56]: from sklearn.model_selection import train_test_split
         xtrain,xtest,ytrain,ytest=train_test_split(x,y,test_size=0.3,random_state=1)
In [57]: from sklearn.linear_model import LogisticRegression
         logreg = LogisticRegression()
         logreg.fit(xtrain,ytrain)
         ypred = logreg.predict(xtest)
In [60]: from sklearn.metrics import accuracy_score,confusion_matrix,classification_report
         ac = accuracy_score(ytest,ypred)
         cm = confusion_matrix(ytest,ypred)
         cr = classification_report(ytest,ypred)
         print(f"Accuracy:- {ac}\n{cm}\n{cr}")
         Accuracy:- 0.8426966292134831
         [[144 22]
          [ 20 81]]
                        precision
                                     recall f1-score
                                                         support
                             0.88
                                       0.87
                                                  0.87
                                                             166
                             0.79
                                       0.80
                                                  0.79
                                                             101
                                                  0.84
                                                             267
             accuracy
             macro avg
                             0.83
                                       0.83
                                                  0.83
                                                             267
         weighted avg
                             0.84
                                       0.84
                                                  0.84
                                                             267
```

```
In [62]: train = logreg.score(xtrain,ytrain)
         test = logreg.score(xtest,ytest)
         print(f"Training Score :- {train}\n Testing Score:- {test}")
         Training Score :- 0.7909967845659164
          Testing Score: - 0.8426966292134831
In [63]: #High Bias + Low Variance ===>underfitting
In [64]: df.Survived.value_counts()
Out[64]: 0
              549
              340
         Name: Survived, dtype: int64
In [65]: xtrain,xtest,ytrain,ytest=train_test_split(x,y,test_size=0.3,random_state=1,stratify=y)
In [66]: from sklearn.linear_model import LogisticRegression
         logreg = LogisticRegression()
         logreg.fit(xtrain,ytrain)
         ypred = logreg.predict(xtest)
In [67]: from sklearn.metrics import accuracy_score,confusion_matrix,classification_report
         ac = accuracy_score(ytest,ypred)
         cm = confusion_matrix(ytest,ypred)
         cr = classification_report(ytest,ypred)
         print(f"Accuracy:- {ac}\n{cm}\n{cr}")
         Accuracy: - 0.8014981273408239
         [[140 25]
[ 28 74]]
                       precision
                                     recall f1-score
                                                        support
                            0.83
                                      0.85
                                                            165
                            0.75
                                      0.73
                                                            102
                                                 0.74
             accuracy
                                                 0.80
                                                            267
            macro avg
                            0.79
                                       0.79
                                                 0.79
                                                            267
                            0.80
                                                            267
         weighted avg
                                      0.80
                                                 0.80
In [68]: | train = logreg.score(xtrain,ytrain)
         test = logreg.score(xtest,ytest)
         print(f"Training Score :- {train}\n Testing Score:- {test}")
         Training Score :- 0.8038585209003215
```

Forecast New Observation

Testing Score:- 0.8014981273408239

```
In [69]: x
Out[69]:
               Pclass Sex Age SibSp Parch
                                              Fare Embarked
                   3 1.0 22.0
                                        0
                                            7.2500
                                                        2.0
                   1 0.0 38.0
                                        0 71.2833
                                                        0.0
            2
                      0.0 26.0
                                  0
                                            7.9250
                                                         2.0
                                        0 53.1000
                      0.0 35.0
                                                        2.0
                   3
                      1.0 35.0
                                  0
                                            8.0500
                                                        2.0
                   2 1.0 27.0
                                  0
                                        0 13.0000
          886
                                                        2.0
                                        0 30.0000
                                                        20
          887
                      0.0 19.0
                   3 0.0 25.0
                                        2 23,4500
                                                        2.0
          888
          889
                   1 1.0 26.0
                                  0
                                        0 30.0000
                                                        0.0
          890
                   3 1.0 32.0
                                        0 7.7500
                                                         1.0
         889 rows × 7 columns
In [70]: oe
Out[70]:
          - OrdinalEncoder
          OrdinalEncoder()
In [71]: | oe.categories_
Out[71]: [array(['female', 'male'], dtype=object), array(['C', 'Q', 'S'], dtype=object)]
In [72]: | s="male"
         oe.transform([[s,e]])
Out[72]: array([[1., 1.]])
In [74]: pclass=1
         sex="male"
         age=24
          sibsp=0
         parch=1
         fare=34
         embarked="S"
          newob=[pclass,sex,age,sibsp,parch,fare,embarked]
         newob[1]
         newob[-1]
Out[74]: 'S'
In [81]: | oe.transform([[newob[1],newob[-1]]])[0]
Out[81]: array([1., 2.])
In [83]: newob[1],newob[-1]=oe.transform([[newob[1],newob[-1]]])[0]
In [84]: newob
Out[84]: [1, 1.0, 24, 0, 1, 34, 2.0]
In [85]: logreg.predict([newob])
Out[85]: array([1], dtype=int64)
```

```
In [100]: df[(df["Fare"]>500)]
Out[100]:
                Survived Pclass
                                  Sex Age SibSp Parch
                                                            Fare Embarked
            258
                              1 female
                                       35.0
                                                      0 512.3292
                                                                         С
            679
                                  male
                                       36.0
                                                0
                                                      1 512.3292
                                                                         С
                                                                         С
            737
                                  male 35.0
                                                0
                                                      0 512.3292
 In [86]: def predictsurvived():
               pclass=int(input("Enter Passenger Class :- "))
               sex=input("Enter Gender of the Passenger :- ")
               age=float(input("Enter the age of the passenger :- "))
sibsp=int(input("Enter no of SibSp of the Passenger:- "))
               parch=int(input("Enter no of parch of the Passenger:- "))
               fare=int(input("Enter Ticket Price:- "))
               embarked=input("Enter Embarked/Port of the Passenger:- ")
               newob=[pclass,sex,age,sibsp,parch,fare,embarked]
               \verb|newob[1]|, \verb|newob[-1]| = oe.transform([[newob[1]], newob[-1]]])[0]|
               v=logreg.predict([newob])[0]
               if v==1:
                   print(f"\n With the Given Feature the Passenger Will Survived")
                   return v
                   print(f"\n With the Given Feature the Passenger Will Not Survived")
In [110]: predictsurvived()
           Enter Passenger Class :- 1
           Enter Gender of the Passenger :- male
           Enter the age of the passenger :- 20
           Enter no of SibSp of the Passenger:- 0
           Enter no of parch of the Passenger: - 0
           Enter Ticket Price:- 512
           Enter Embarked/Port of the Passenger:- C
            With the Given Feature the Passenger Will Survived
Out[110]: 1
  In [ ]:
  In [ ]:
```