In [43]: import pandas as pd

In [44]: data=pd.read_csv("/home/placement/Downloads/Titanic Dataset.csv")

In [45]: data.describe()

Out[45]:

	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 [46]: data.head(5)

Out[46]:

:		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

In [47]:	data.isna().s	um()
Out[47]:	PassengerId	0
	Survived	0
	Pclass	0
	Name	0
	Sex	0
	Age	177
	SibSp	0
	Parch	0
	Ticket	0

Embarked dtype: int64

0

2

687

In [48]: data.head(10)

Fare

Cabin

Out[48]:

		PassengerId	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

```
In [49]: data['Survived'].unique()
Out[49]: array([0, 1])
In [50]: data['SibSp'].unique()
Out[50]: array([1, 0, 3, 4, 2, 5, 8])
In [51]: data['Age'].unique()
Out[51]: array([22. , 38. , 26. , 35. , nan, 54. , 2. , 27. , 14. ,
                4. , 58. , 20. , 39. , 55. , 31.
                                                   , 34.
                                                         , 15. , 28. ,
                8. , 19. , 40. , 66. , 42. , 21.
                                                    , 18.
                                                          , 3.
                         , 65. , 28.5 , 5. , 11. , 45.
               16. , 25.
                         , 0.83, 30.
                                      , 33. , 23.
                                                    , 24.
               71. , 37. , 47. , 14.5 , 70.5 , 32.5 , 12.
               51. , 55.5 , 40.5 , 44. , 1. , 61.
                                                   , 56.
                                                         , 50.
               45.5 , 20.5 , 62. , 41. , 52. , 63. , 23.5 , 0.92, 43. ,
               60. , 10. , 64. , 13. , 48. , 0.75, 53. , 57. , 80. ,
               70. , 24.5 , 6. , 0.67, 30.5 , 0.42, 34.5 , 74. ])
In [52]: datal=data.drop(['PassengerId', 'Ticket', 'Cabin', 'Name', 'SibSp', 'Parch'],axis=1)
```

In [53]: data1

Out[53]:

	Survived	Pclass	Sex	Age	Fare	Embarked
0	0	3	male	22.0	7.2500	S
1	1	1	female	38.0	71.2833	С
2	1	3	female	26.0	7.9250	S
3	1	1	female	35.0	53.1000	S
4	0	3	male	35.0	8.0500	S
886	0	2	male	27.0	13.0000	S
887	1	1	female	19.0	30.0000	S
888	0	3	female	NaN	23.4500	S
889	1	1	male	26.0	30.0000	С
890	0	3	male	32.0	7.7500	Q

891 rows × 6 columns

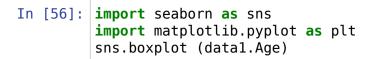
In [54]: data1.fillna(35,inplace=True)

In [55]: data1

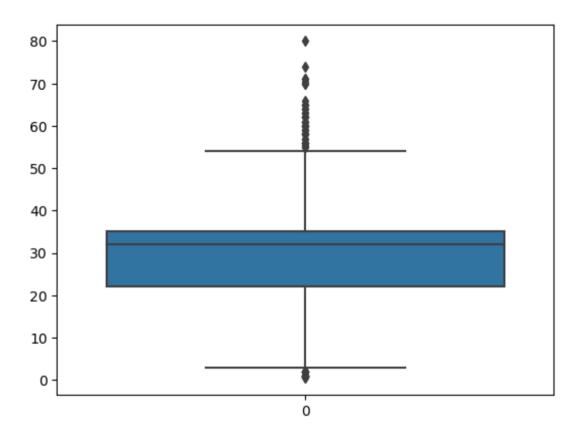
Out[55]:

	Survived	Pclass	Sex	Age	Fare	Embarked
0	0	3	male	22.0	7.2500	S
1	1	1	female	38.0	71.2833	С
2	1	3	female	26.0	7.9250	S
3	1	1	female	35.0	53.1000	S
4	0	3	male	35.0	8.0500	S
886	0	2	male	27.0	13.0000	S
887	1	1	female	19.0	30.0000	S
888	0	3	female	35.0	23.4500	S
889	1	1	male	26.0	30.0000	С
890	0	3	male	32.0	7.7500	Q

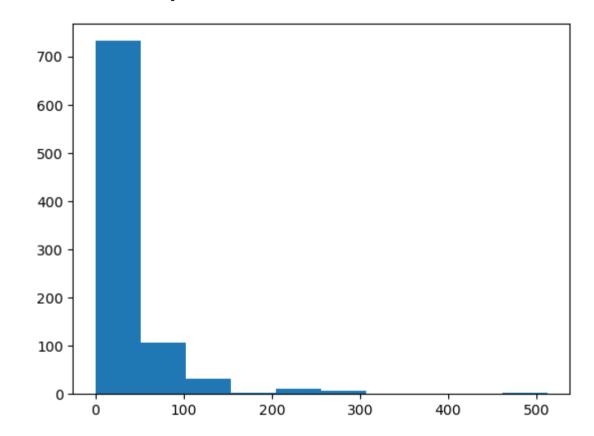
891 rows × 6 columns



Out[56]: <Axes: >



```
In [57]: plt.hist(data1['Age'])
Out[57]: (array([ 54., 46., 177., 169., 295., 70., 45., 24.,
                                                                        2.]),
          array([ 0.42 , 8.378, 16.336, 24.294, 32.252, 40.21 , 48.168, 56.126,
                 64.084, 72.042, 80. ]),
          <BarContainer object of 10 artists>)
          300
          250
          200
          150
          100
           50
                             20
                      10
                                                              70
                                    30
                                          40
                                                 50
                                                        60
                                                                     80
```



```
In [59]: data1.describe()
```

Out[59]:

	Survived	Pclass	Age	Fare
count	891.000000	891.000000	891.000000	891.000000
mean	0.383838	2.308642	30.752155	32.204208
std	0.486592	0.836071	13.173100	49.693429
min	0.000000	1.000000	0.420000	0.000000
25%	0.000000	2.000000	22.000000	7.910400
50%	0.000000	3.000000	32.000000	14.454200
75%	1.000000	3.000000	35.000000	31.000000
max	1.000000	3.000000	80.000000	512.329200

```
In [60]: data1['Age'].unique()
Out[60]: array([22. , 38. , 26. , 35. , 54. , 2. , 27. , 14. , 4. ,
              58. , 20. , 39. , 55. , 31. , 34.
                                                 , 15. , 28.
              19. , 40. , 66. , 42. , 21. , 18.
                                                 , 3.
              29. , 65. , 28.5 , 5. , 11. , 45.
                                                 , 17.
                                                        , 32.
                                                 , 46.
              25. , 0.83, 30. , 33. , 23. , 24.
              37. , 47. , 14.5 , 70.5 , 32.5 , 12.
                                                 , 9. , 36.5 , 51. ,
              55.5 , 40.5 , 44. , 1. , 61. , 56. , 50. , 36.
              20.5 , 62. , 41. , 52. , 63. , 23.5 , 0.92, 43.
              10. , 64. , 13. , 48. , 0.75, 53. , 57. , 80. , 70. ,
              24.5 , 6. , 0.67, 30.5 , 0.42, 34.5 , 74. ])
```

```
In [61]: data1.groupby(['Age']).count()
Out[61]: Survived Pclass Sex Fare Embarked
```

	Surviveu	rciass	Jex	rait	Lilibaikeu
Age					
0.42	1	1	1	1	1
0.67	1	1	1	1	1
0.75	2	2	2	2	2
0.83	2	2	2	2	2
0.92	1	1	1	1	1
70.00	2	2	2	2	2
70.50	1	1	1	1	1
71.00	2	2	2	2	2
74.00	1	1	1	1	1
80.00	1	1	1	1	1

88 rows × 5 columns

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In [64]: datal=pd.get_dummies(datal)

In [65]: data1.shape

Out[65]: (891, 12)

In [66]: data1.head(500)

Out[66]:

	Survived	Age	Fare	Pclass_F	Pclass_S	Pclass_Third	Sex_female	Sex_male	Embarked_35	Embarked_C	Embarked_Q	Embarked_S
0	0	22.0	7.2500	0	0	1	0	1	0	0	0	1
1	1	38.0	71.2833	1	0	0	1	0	0	1	0	0
2	1	26.0	7.9250	0	0	1	1	0	0	0	0	1
3	1	35.0	53.1000	1	0	0	1	0	0	0	0	1
4	0	35.0	8.0500	0	0	1	0	1	0	0	0	1
495	0	35.0	14.4583	0	0	1	0	1	0	1	0	0
496	1	54.0	78.2667	1	0	0	1	0	0	1	0	0
497	0	35.0	15.1000	0	0	1	0	1	0	0	0	1
498	0	25.0	151.5500	1	0	0	1	0	0	0	0	1
499	0	24.0	7.7958	0	0	1	0	1	0	0	0	1

500 rows × 12 columns

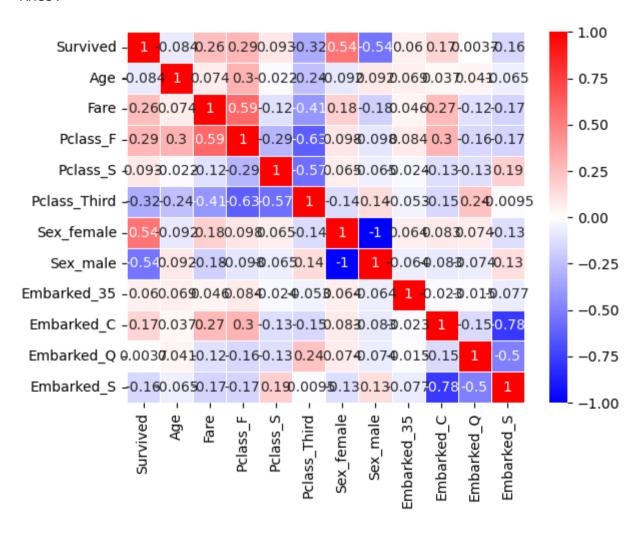
In [67]: cor_mat=data1.corr()
cor_mat

Out[67]:

	Survived	Age	Fare	Pclass_F	Pclass_S	Pclass_Third	Sex_female	Sex_male	Embarked_35	Embarked_C	Embarked_
Survived	1.000000	-0.083713	0.257307	0.285904	0.093349	-0.322308	0.543351	-0.543351	0.060095	0.168240	0.00365
Age	-0.083713	1.000000	0.074199	0.302149	-0.022021	-0.242412	-0.091930	0.091930	0.069343	0.036953	0.04052
Fare	0.257307	0.074199	1.000000	0.591711	-0.118557	-0.413333	0.182333	-0.182333	0.045646	0.269335	-0.11721
Pclass_F	0.285904	0.302149	0.591711	1.000000	-0.288585	-0.626738	0.098013	-0.098013	0.083847	0.296423	-0.15534
Pclass_S	0.093349	-0.022021	-0.118557	-0.288585	1.000000	-0.565210	0.064746	-0.064746	-0.024197	-0.125416	-0.12730
Pclass_Third	-0.322308	-0.242412	-0.413333	-0.626738	-0.565210	1.000000	-0.137143	0.137143	-0.052550	-0.153329	0.23744
Sex_female	0.543351	-0.091930	0.182333	0.098013	0.064746	-0.137143	1.000000	-1.000000	0.064296	0.082853	0.07411
Sex_male	-0.543351	0.091930	-0.182333	-0.098013	-0.064746	0.137143	-1.000000	1.000000	-0.064296	-0.082853	-0.07411
Embarked_35	0.060095	0.069343	0.045646	0.083847	-0.024197	-0.052550	0.064296	-0.064296	1.000000	-0.022864	-0.01458
Embarked_C	0.168240	0.036953	0.269335	0.296423	-0.125416	-0.153329	0.082853	-0.082853	-0.022864	1.000000	-0.14825
Embarked_Q	0.003650	0.040528	-0.117216	-0.155342	-0.127301	0.237449	0.074115	-0.074115	-0.014588	-0.148258	1.00000
Embarked_S	-0.155660	-0.065062	-0.166603	-0.170379	0.192061	-0.009511	-0.125722	0.125722	-0.076588	-0.778359	-0.49662



Out[68]: <Axes: >



```
In [69]: data.groupby('Survived').count()
Out[69]:
                   Passengerld Pclass Name Sex Age SibSp Parch Ticket Fare Cabin Embarked
           Survived
                0
                         549
                                549
                                     549
                                          549 424
                                                    549
                                                          549
                                                                549
                                                                     549
                                                                            68
                                                                                    549
                         342
                                                                                    340
                1
                                342
                                     342 342
                                              290
                                                    342
                                                          342
                                                                342
                                                                     342
                                                                           136
In [70]: y=data1['Survived']
          x=data1.drop('Survived',axis=1)
In [71]: from sklearn.model_selection import train_test_split
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.33,random_state=42)
```

```
In [73]: from sklearn.linear model import LogisticRegression
         classifier=LogisticRegression()
         classifier.fit(x train,y train)
         /home/placement/anaconda3/lib/python3.10/site-packages/sklearn/linear model/ logistic.py:458: ConvergenceWa
         rning: lbfqs 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 (https://scikit-learn.org/stable/modules/pre
         processing.html)
         Please also refer to the documentation for alternative solver options:
             https://scikit-learn.org/stable/modules/linear model.html#logistic-regression (https://scikit-learn.or
         g/stable/modules/linear model.html#logistic-regression)
           n iter i = check optimize result(
Out[73]: LogisticRegression()
         In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
         On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
In [74]: y pred=classifier.predict(x test)
In [75]: y_pred
Out[75]: array([0, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0,
                1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 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, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 0, 0, 0, 1, 1,
                0, 0, 0, 1, 0, 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, 0, 0, 1, 1, 0, 0, 0, 1, 1, 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, 1, 1, 0, 1, 0, 0,
                0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0,
                1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, 0,
                0, 1, 0, 0, 0, 1, 0, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0, 1,
                0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0,
                0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0,
                1, 0, 0, 0, 0, 0, 1, 1, 0])
```

```
In [76]: from sklearn.metrics import confusion matrix
         confusion matrix(y test,y pred)
Out[76]: array([[155, 20],
                [ 36, 84]])
In [78]: from sklearn.metrics import accuracy_score
         accuracy_score(y_test,y_pred)
Out[78]: 0.8101694915254237
In [79]: y
Out[79]: 0
                0
                1
                0
         886
                0
         887
         888
         889
                1
         890
         Name: Survived, Length: 891, dtype: int64
In [ ]:
```