```
In [46]: import numpy as np
        import pandas as pd
        from sklearn.model_selection import train test split
        from sklearn.linear model import LogisticRegression
        from sklearn.metrics import accuracy score
In [47]: sonar_data=pd.read_csv(r'C:\my edu\Machine Learning\udemy_ml\ml_projects\Rock vs Mine Prediction\sonar data.csv', header=None)
In [48]: print(sonar_data.head())
                                    3
                                                  5
                                                                 7
                                                                         8 \
       0 0.0200 0.0371 0.0428 0.0207 0.0954 0.0986 0.1539 0.1601 0.3109
       1 0.0453 0.0523 0.0843 0.0689 0.1183 0.2583 0.2156 0.3481 0.3337
       2 0.0262 0.0582 0.1099 0.1083 0.0974 0.2280 0.2431 0.3771 0.5598
        3 0.0100 0.0171 0.0623 0.0205 0.0205 0.0368 0.1098 0.1276 0.0598
                                       0.0590 0.0649 0.1209 0.2467 0.3564
       4 0.0762 0.0666 0.0481 0.0394
                                 52
                                        53
                          51
                                                54
                                                       55
              9
                 . . .
                                                               56
                                                                      57 \
       0 0.2111 ... 0.0027 0.0065 0.0159 0.0072 0.0167 0.0180 0.0084
       1 0.2872 ... 0.0084 0.0089 0.0048
                                            0.0094 0.0191 0.0140 0.0049
                 ... 0.0232 0.0166 0.0095 0.0180 0.0244 0.0316 0.0164
       2 0.6194
       3 0.1264 ... 0.0121 0.0036 0.0150 0.0085 0.0073 0.0050 0.0044
       4 0.4459 ... 0.0031 0.0054 0.0105 0.0110 0.0015 0.0072 0.0048
              58
                     59 60
        0 0.0090 0.0032
       1 0.0052 0.0044
       2 0.0095 0.0078
        3 0.0040 0.0117
       4 0.0107 0.0094
       [5 rows x 61 columns]
In [49]: sonar_data.shape
Out[49]: (208, 61)
In [50]: sonar_data.describe()
```

Out[50]:		0	1	2	3	4	5	6	7	8	9	•••	50	51	52	
	count	208.000000	208.000000	208.000000	208.000000	208.000000	208.000000	208.000000	208.000000	208.000000	208.000000		208.000000	208.000000	208.000000	208.
	mean	0.029164	0.038437	0.043832	0.053892	0.075202	0.104570	0.121747	0.134799	0.178003	0.208259		0.016069	0.013420	0.010709	0.
	std	0.022991	0.032960	0.038428	0.046528	0.055552	0.059105	0.061788	0.085152	0.118387	0.134416		0.012008	0.009634	0.007060	0.
	min	0.001500	0.000600	0.001500	0.005800	0.006700	0.010200	0.003300	0.005500	0.007500	0.011300		0.000000	0.000800	0.000500	0.
	25%	0.013350	0.016450	0.018950	0.024375	0.038050	0.067025	0.080900	0.080425	0.097025	0.111275		0.008425	0.007275	0.005075	0.
	50%	0.022800	0.030800	0.034300	0.044050	0.062500	0.092150	0.106950	0.112100	0.152250	0.182400		0.013900	0.011400	0.009550	0.
	75%	0.035550	0.047950	0.057950	0.064500	0.100275	0.134125	0.154000	0.169600	0.233425	0.268700		0.020825	0.016725	0.014900	0.
	max	0.137100	0.233900	0.305900	0.426400	0.401000	0.382300	0.372900	0.459000	0.682800	0.710600		0.100400	0.070900	0.039000	0.

8 rows × 60 columns

```
2
                                            5
                             3
    0.0200 0.0371 0.0428 0.0207 0.0954 0.0986 0.1539 0.1601 0.3109
    0.0453 0.0523 0.0843 0.0689 0.1183 0.2583
                                               0.2156 0.3481 0.3337
    0.0262 0.0582 0.1099 0.1083 0.0974 0.2280
                                               0.2431 0.3771 0.5598
    0.0100 0.0171 0.0623 0.0205 0.0205 0.0368
                                               0.1098 0.1276 0.0598
    0.0762 0.0666 0.0481 0.0394 0.0590 0.0649
                                               0.1209 0.2467 0.3564
    0.0187 0.0346 0.0168 0.0177 0.0393 0.1630
                                               0.2028 0.1694 0.2328
   0.0323 0.0101 0.0298 0.0564 0.0760 0.0958
                                               0.0990 0.1018 0.1030
205 0.0522 0.0437 0.0180 0.0292 0.0351 0.1171 0.1257 0.1178 0.1258
206 0.0303 0.0353 0.0490 0.0608 0.0167 0.1354 0.1465 0.1123 0.1945
207 0.0260 0.0363 0.0136 0.0272 0.0214 0.0338 0.0655 0.1400 0.1843
        9
                   50
                           51
                                  52
                                                        55
                                         53
                                                54
                                                               56 \
    0.2111 ... 0.0232 0.0027 0.0065 0.0159 0.0072 0.0167 0.0180
    0.2872 ... 0.0125 0.0084 0.0089 0.0048 0.0094
                                                    0.0191
                                                           0.0140
    0.6194 ... 0.0033 0.0232 0.0166 0.0095 0.0180
                                                    0.0244
    0.1264 ... 0.0241 0.0121 0.0036 0.0150 0.0085
                                                    0.0073
                                                           0.0050
    0.4459 ... 0.0156 0.0031 0.0054 0.0105 0.0110
                                                    0.0015 0.0072
                                                           0.0065
    0.2684 ... 0.0203 0.0116
                              0.0098 0.0199
                                            0.0033
                                                    0.0101
204 0.2154 ... 0.0051 0.0061 0.0093 0.0135 0.0063
                                                    0.0063
                                                          0.0034
205 0.2529 ... 0.0155 0.0160 0.0029 0.0051 0.0062
                                                    0.0089
                                                           0.0140
206 0.2354 ... 0.0042 0.0086 0.0046 0.0126 0.0036
                                                    0.0035 0.0034
207 0.2354 ... 0.0181 0.0146 0.0129 0.0047 0.0039
        57
               58
                      59
    0.0084 0.0090 0.0032
    0.0049 0.0052 0.0044
    0.0164 0.0095 0.0078
    0.0044 0.0040 0.0117
    0.0048 0.0107 0.0094
              . . .
    0.0115 0.0193 0.0157
204 0.0032 0.0062 0.0067
205 0.0138 0.0077 0.0031
206 0.0079 0.0036 0.0048
207 0.0036 0.0061 0.0115
[208 rows x 60 columns]
```

In [54]: print(y)

```
203
        204
        205
        206
        207
        Name: 60, Length: 208, dtype: object
In [57]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.1,stratify=y,random_state=1)
In [58]: print(x.shape,x_train.shape,x_test.shape)
        (208, 60) (187, 60) (21, 60)
In [59]: model=LogisticRegression()
In [60]: model.fit(x_train,y_train)
Out[60]:
          ▼ LogisticRegression
          ► Parameters
         Model Evaluation
In [61]: x_train_prediction=model.predict(x_train)
         training_data_accuracy=accuracy_score(x_train_prediction,y_train)
In [62]: print("Accuracy on training data is :",training_data_accuracy)
        Accuracy on training data is : 0.8342245989304813
In [63]: x_test_prediction=model.predict(x_test)
         x_test_accuracy=accuracy_score(x_test_prediction,y_test)
```

```
In [64]: print("Accuracy on test data is",x_test_accuracy)

Accuracy on test data is 0.7619047619047619

Making a Predictive System

In [67]: input=(0.0453,0.0523,0.0843,0.0689,0.1183,0.2583,0.2156,0.3481,0.3337,0.2872,0.4918,0.6552,0.6919,0.7797,0.7464,0.9444,1.0000,0.8874,0.8024,0.7818,0.5212,0.4
input_data_ass_numpy_array=np.asarray(input)
input_data_reshaped=input_data_ass_numpy_array.reshape(1,-1)
prediction=model.predict(input_data_reshaped)
print(prediction)
if(prediction=='R'):
    print("The object is a Rock")
else:
    print("The object is Mine")

['R']
The object is a Rock

In []:
```