Importing the Dependencies

```
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
```

Data Collection and Data Processing

```
#loading the dataset to a pandas Dataframe sonar_data = pd.read_csv('/content/sonar data.csv', header=None)
```

sonar_data.head()

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
0	0.0200	0.0371	0.0428	0.0207	0.0954	0.0986	0.1539	0.1601	0.3109	0.2111	0.1609	0.1582	0.2238	0.0645	0.0660	0.2273	0.3100
1	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
2	0.0262	0.0582	0.1099	0.1083	0.0974	0.2280	0.2431	0.3771	0.5598	0.6194	0.6333	0.7060	0.5544	0.5320	0.6479	0.6931	0.6759
3	0.0100	0.0171	0.0623	0.0205	0.0205	0.0368	0.1098	0.1276	0.0598	0.1264	0.0881	0.1992	0.0184	0.2261	0.1729	0.2131	0.0693
4	0.0762	0.0666	0.0481	0.0394	0.0590	0.0649	0.1209	0.2467	0.3564	0.4459	0.4152	0.3952	0.4256	0.4135	0.4528	0.5326	0.7306
4																	>

number of rows and columns
sonar_data.shape

(208, 61)

sonar_data.describe() #describe --> statistical measures of the data

	0	1	2	3	4	5	6	7	8	9	
count	208.000000	208.000000	208.000000	208.000000	208.000000	208.000000	208.000000	208.000000	208.000000	208.000000	208.00
mean	0.029164	0.038437	0.043832	0.053892	0.075202	0.104570	0.121747	0.134799	0.178003	0.208259	0.23
std	0.022991	0.032960	0.038428	0.046528	0.055552	0.059105	0.061788	0.085152	0.118387	0.134416	0.13
min	0.001500	0.000600	0.001500	0.005800	0.006700	0.010200	0.003300	0.005500	0.007500	0.011300	0.02
25%	0.013350	0.016450	0.018950	0.024375	0.038050	0.067025	0.080900	0.080425	0.097025	0.111275	0.12
50%	0.022800	0.030800	0.034300	0.044050	0.062500	0.092150	0.106950	0.112100	0.152250	0.182400	0.22
75%	0.035550	0.047950	0.057950	0.064500	0.100275	0.134125	0.154000	0.169600	0.233425	0.268700	0.30
max	0.137100	0.233900	0.305900	0.426400	0.401000	0.382300	0.372900	0.459000	0.682800	0.710600	0.73

sonar_data[60].value_counts()

M 111 R 97

Name: 60, dtype: int64

M --> Mine

R --> Rock

sonar_data.groupby(60).mean()

```
1
                                   2
                                            3
                                                               5
                                                                                 7
                                                                                           8
                                                                                                    9
                                                                                                             10
      60
# separating data and Labels
X = sonar data.drop(columns=60, axis=1)
Y = sonar data[60]
print(X)
print(Y)
              0
                      1
                              2
                                                   56
                                                          57
                                                                  58
                                                                          59
                                      3
                                          . . .
                                              0.0180
                                                                      0.0032
          0.0200
                 0.0371
                         0.0428
                                 0.0207
                                                      0.0084
                                                              0.0090
                                                      0.0049
     1
          0.0453 0.0523
                         0.0843
                                 0.0689
                                              0.0140
                                                              0.0052
                                                                      0.0044
                                              0.0316
          0.0262 0.0582
                         0.1099
                                 0.1083
                                                      0.0164
                                                              0.0095
                                                                      0.0078
                 0.0171 0.0623
                                 0.0205
                                              0.0050
                                                      0.0044
                                                              0.0040
          0.0100
                                                                      0.0117
          0.0762 0.0666
                         0.0481
                                 0.0394
                                              0.0072
                                                      0.0048
                                                              0.0107
                                                                      0.0094
                         0.0168
                                              0.0065
                                                      0.0115
          0.0187
                 0.0346
                                  0.0177
                                                              0.0193
                                                                      0.0157
                                                              0.0062
                 0.0101
                         0.0298
                                 0.0564
                                              0.0034
                                                      0.0032
          0.0323
                                                                      0.0067
         0.0522 0.0437
                         0.0180
                                 0.0292
                                              0.0140 0.0138 0.0077
     205
                                                                      0.0031
                                              0.0034 0.0079 0.0036 0.0048
     206 0.0303 0.0353
                          0.0490
                                 0.0608
     207 0.0260 0.0363 0.0136
                                 0.0272
                                              0.0040 0.0036 0.0061 0.0115
     [208 rows x 60 columns]
            R
     1
            R
     2
            R
            R
            R
     203
            Μ
     204
            Μ
     205
            Μ
     206
            Μ
     207
            Μ
```

Name: 60, Length: 208, dtype: object

12

11

Training and Test data

```
X train, X test, Y train, Y test = train test split(X, Y, test size = 0.1, stratify=Y, random state=1)
print(X.shape, X train.shape, X test.shape)
     (208, 60) (187, 60) (21, 60)
print(X train)
print(Y train)
              0
                      1
                              2
                                      3
                                                   56
                                                          57
                                                                   58
                                                                          59
                                          . . .
     115
          0.0414 0.0436
                          0.0447
                                 0.0844
                                              0.0141 0.0077 0.0246
                                                                      0.0198
     38
          0.0123 0.0022
                          0.0196
                                 0.0206
                                               0.0113
                                                      0.0058
                                                              0.0047
                                                                      0.0071
                                          . . .
                                              0.0037
          0.0152 0.0102
                          0.0113
                                 0.0263
                                                      0.0011 0.0034
                                                                      0.0033
     123
         0.0270 0.0163
                          0.0341
                                 0.0247
                                               0.0138 0.0094
                                                              0.0105
                                                                      0.0093
     18
          0.0270
                 0.0092
                          0.0145
                                 0.0278
                                               0.0120
                                                      0.0132
                                                              0.0070
                                                                      0.0088
                                                      0.0225
     140
          0.0412
                 0.1135
                          0.0518
                                 0.0232
                                               0.0095
                                                              0.0098
                                                                      0.0085
          0.0286
                 0.0453
                          0.0277
                                 0.0174
                                              0.0057
                                                      0.0027
                                                              0.0051 0.0062
         0.0117 0.0069
                         0.0279
                                              0.0020 0.0062 0.0026
                                 0.0583
                                                                      0.0052
                         0.0866
                                              0.0190 0.0141 0.0068 0.0086
     131
         0.1150 0.1163
                                 0.0358
     203
         0.0187 0.0346
                         0.0168 0.0177
                                         . . .
                                              0.0065 0.0115 0.0193 0.0157
     [187 rows x 60 columns]
     115
            Μ
     38
            R
     56
            R
     123
            Μ
     18
            R
     140
            Μ
            R
     154
            Μ
     131
            Μ
     203
            Μ
     Name: 60, Length: 187, dtype: object
```

Model Training --> Logistic Regression

model = LogisticRegression()

```
#training the Logistic Regression model with training data
model.fit(X train, Y train)
     LogisticRegression(C=1.0, class weight=None, dual=False, fit intercept=True,
                        intercept scaling=1, l1 ratio=None, max iter=100,
                        multi class='auto', n jobs=None, penalty='12',
                        random state=None, solver='lbfgs', tol=0.0001, verbose=0,
                        warm start=False)
Model Evaluation
#accuracy on training data
X train prediction = model.predict(X train)
training data accuracy = accuracy score(X train prediction, Y train)
print('Accuracy on training data : ', training data accuracy)
     Accuracy on training data: 0.8342245989304813
#accuracy on test data
X test prediction = model.predict(X test)
```

Accuracy on test data : 0.7619047619047619

print('Accuracy on test data : ', test_data_accuracy)

test data accuracy = accuracy score(X test prediction, Y test)

Making a Predictive System

```
input data = (0.0307, 0.0523, 0.0653, 0.0521, 0.0611, 0.0577, 0.0665, 0.0664, 0.1460, 0.2792, 0.3877, 0.4992, 0.4981, 0.4972, 0.5607, 0.7339, 0.8230, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611, 0.0611
# changing the input data to a numpy array
input data as numpy array = np.asarray(input data)
# reshape the np array as we are predicting for one instance
input data reshaped = input data as numpy array.reshape(1,-1)
prediction = model.predict(input data reshaped)
print(prediction)
if (prediction[0]=='R'):
        print('The object is a Rock')
else:
        print('The object is a mine')
                        ['M']
                       The object is a mine
```

