```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score
```

DATA COLLECTION

```
#importing data
sonar_data=pd.read_csv('/content/Sonar data.csv.csv',header=None)
```

```
sonar_data.head()
```

sonar_data.shape

(208, 61)

sonar_data.describe() # To measure the statistical mea

	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
mean	0.029164	0.038437	0.043832	0.053892	0.075202	0.104570	0.121747	0.134799	0.178003	0.208259
std	0.022991	0.032960	0.038428	0.046528	0.055552	0.059105	0.061788	0.085152	0.118387	0.134416
min	0.001500	0.000600	0.001500	0.005800	0.006700	0.010200	0.003300	0.005500	0.007500	0.011300
25%	0.013350	0.016450	0.018950	0.024375	0.038050	0.067025	0.080900	0.080425	0.097025	0.111275
50%	0.022800	0.030800	0.034300	0.044050	0.062500	0.092150	0.106950	0.112100	0.152250	0.182400
75%	0.035550	0.047950	0.057950	0.064500	0.100275	0.134125	0.154000	0.169600	0.233425	0.268700
max	0.137100	0.233900	0.305900	0.426400	0.401000	0.382300	0.372900	0.459000	0.682800	0.710600

8 rows × 60 columns



sonar_data[60].value_counts()

M 111

R 97

Name: 60, dtype: int64

1

0.0049

```
#separating datas and labels
x=sonar_data.drop(columns=60,axis=1)
y=sonar data[60]
print(x)
print(y)
               0
                       1
                               2
                                        3
                                                         5
                                                                 6
                                                                         7
                                                                                  8
                                                4
     0
          0.0200
                   0.0371
                           0.0428
                                    0.0207
                                            0.0954
                                                    0.0986
                                                             0.1539
                                                                     0.1601
                                                                             0.3109
                                    0.0689
     1
          0.0453
                   0.0523
                           0.0843
                                            0.1183
                                                    0.2583
                                                             0.2156
                                                                     0.3481
                                                                             0.3337
     2
                   0.0582
                           0.1099
                                    0.1083
                                            0.0974
                                                    0.2280
                                                             0.2431
                                                                     0.3771
          0.0262
                                                                             0.5598
     3
                   0.0171
                           0.0623
                                   0.0205
                                            0.0205
                                                    0.0368
                                                             0.1098
                                                                     0.1276
                                                                             0.0598
          0.0100
                           0.0481
          0.0762
                   0.0666
                                    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
     203
          0.0323
                   0.0101
                           0.0298
                                    0.0564
                                            0.0760
                                                    0.0958
                                                             0.0990
                                                                     0.1018
                                                                             0.1030
     204
     205
          0.0522
                  0.0437
                           0.0180
                                   0.0292
                                            0.0351
                                                    0.1171
                                                             0.1257
                                                                     0.1178
                                                                             0.1258
          0.0303
                  0.0353
                           0.0490
                                    0.0608
                                            0.0167
                                                    0.1354
                                                             0.1465
                                                                     0.1123
                                                                             0.1945
     206
                                                    0.0338
     207
          0.0260
                  0.0363
                           0.0136
                                   0.0272
                                            0.0214
                                                             0.0655
                                                                     0.1400
                                                                             0.1843
               9
                            50
                                     51
                                             52
                                                     53
                                                              54
                                                                      55
                                                                               56 \
                        0.0232
                                0.0027
                                         0.0065
                                                 0.0159
                                                         0.0072
                                                                  0.0167
     0
          0.2111
                                                                          0.0180
                                0.0084
          0.2872
                        0.0125
                                         0.0089
                                                 0.0048
                                                          0.0094
                                                                  0.0191
                                                                          0.0140
                        0.0033
                                0.0232
                                         0.0166
          0.6194
                                                 0.0095
                                                         0.0180
                                                                  0.0244
                                                                          0.0316
                                         0.0036
                                                 0.0150
     3
          0.1264
                        0.0241
                                0.0121
                                                         0.0085
                                                                  0.0073
                                                                          0.0050
          0.4459
                        0.0156
                                0.0031
                                         0.0054
                                                 0.0105
                                                                  0.0015
     4
                                                          0.0110
                                                                          0.0072
          0.2684
                        0.0203
                                0.0116
                                         0.0098
                                                 0.0199
                                                          0.0033
                                                                  0.0101
     203
                                                                          0.0065
     204
          0.2154
                        0.0051
                                0.0061
                                         0.0093
                                                 0.0135
                                                          0.0063
                                                                  0.0063
                                                                          0.0034
          0.2529
                        0.0155
                                0.0160
                                         0.0029
                                                 0.0051
                                                         0.0062
                                                                  0.0089
     205
                                                                          0.0140
                                0.0086
          0.2354
                        0.0042
                                         0.0046
                                                 0.0126
                                                         0.0036
                                                                  0.0035
     206
                                                                          0.0034
          0.2354
     207
                        0.0181
                                0.0146
                                         0.0129
                                                 0.0047
                                                         0.0039
                                                                  0.0061
                                                                          0.0040
               57
                       58
                               59
                           0.0032
          0.0084
                   0.0090
     0
```

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
     203
                         0.0157
         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]
           R
     1
           R
     2
           R
           R
           R
     203
           Μ
     204
           Μ
     205
           Μ
     206
           Μ
     207
           Μ
     Name: 60, Length: 208, dtype: object
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)
```

MODEL TRAINING USING LOGISTIC REGRESSION

```
model=LogisticRegression()
model.fit(x_train,y_train)
     LogisticRegression()
#accuracy on training data
x_train_prediction=model.predict(x_train)
training data accuracy = accuracy score(x train prediction, y train)
#accuracy of testing data
x test prediction = model.predict(x test)
testing_data_accuracy = accuracy_score(x_test_prediction,y_test)
print("The training data accuracy: ",training_data_accuracy)
print("The testing data accuracy: ",testing_data_accuracy)
     The training data accuracy:
                                    0.8342245989304813
     The testing data accuracy: 0.7619047619047619
```

Making predictive system

```
input data = (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, 1.08181, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.10818, 0.1
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.
# 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
```

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