```
1 import numpy as np
2 import pandas as pd
3 import matplotlib.pyplot as plt
4 from sklearn.model_selection import train_test_split
5 from sklearn.linear_model import LogisticRegression
6 from sklearn.metrics import accuracy_score
```

→ Data Processing

```
1 sonar_data = pd.read_csv('/content/Sonar_data.csv', header=None)
1 sonar_data.head()
                   1
                                                               7
                                                                              9
    0 0.0200 0.0371 0.0428 0.0207 0.0954 0.0986 0.1539 0.1601 0.3109 0.2111
    1 0.0453 0.0523 0.0843 0.0689 0.1183 0.2583 0.2156 0.3481 0.3337 0.2872
                                                                                     0.0
    2 0.0262 0.0582 0.1099 0.1083 0.0974 0.2280 0.2431 0.3771 0.5598 0.6194
    3 0.0100 0.0171 0.0623 0.0205 0.0205 0.0368 0.1098 0.1276 0.0598
                                                                         0.1264
                                                                                  0.0
    4 0.0762 0.0666 0.0481 0.0394 0.0590 0.0649 0.1209 0.2467 0.3564 0.4459
                                                                                  ... 0.1
    5 rows × 61 columns
1 sonar_data.shape
    (208, 61)
1 sonar_data[60].value_counts()
3 # here we got almost similar numbers for both the categories so we don't have to stratify the data compulsarly
4 # M
        111
5 # R
         97
6 # Name: 60, dtype: int64
   Μ
        111
         97
    Name: 60, dtype: int64
```

Separating label and features

```
1 X = sonar_data.drop(60, axis=1)
2 # (shape= 208, 60)
3 # (if we didn't specified axis above, shape would have (207, 61) by dropping 60th row)
4 X
```

```
7
          0.0200 0.0371 0.0428 0.0207 0.0954 0.0986
                                                    0.1539
                                                           0.1601 0.3109
                                                                          0.2111
      1
          0.2872
          0.0262 0.0582 0.1099 0.1083 0.0974 0.2280 0.2431 0.3771 0.5598
                                                                         0.6194
          0.0100 0.0171 0.0623 0.0205 0.0205 0.0368 0.1098 0.1276 0.0598
 1 y = sonar_data[60]
 2 y.head()
 3 # (m is mine and r is rock)
    1
         R
         R
    2
    3
         R
    4
         R
    Name: 60, dtype: object
     207 0.0260 0.0363 0.0136 0.0272 0.0214 0.0338 0.0655 0.1400 0.1843 0.2354
1 # a simple explanation of how groupby works:
 3 # Splitting: The DataFrame is split into groups based on one or more criteria.
4 # Applying: A function is applied to each group independently.
5 # Combining: The results of the function applications are combined back into a new DataFrame.
7 # data = {'Category': ['A', 'B', 'A', 'B', 'A', 'B'],
            'Value': [10, 20, 30, 40, 50, 60]}
10 # df = pd.DataFrame(data)
11
12 # # Grouping by 'Category'
13 # grouped = df.groupby('Category')
15 # # Calculating the mean for each group
16 # mean_values = grouped.mean()
17
18 # print(mean_values)
19
1 sonar_data.groupby(60).mean()
 2 # taking mean of each grp based on grouping(categrising) clmn 60
     60
     M 0.034989 0.045544 0.050720 0.064768 0.086715 0.111864 0.128359 0.149832 0.213
      R 0.022498 0.030303 0.035951 0.041447 0.062028 0.096224 0.114180 0.117596 0.137
    2 rows × 60 columns
 1 sonar_data.groupby(60).sum()
             0
                    1
                            2
                                   3
                                                  5
                                                                  7
                                                                          8
     M 3.8838 5.0554 5.6299 7.1892 9.6254 12.4169 14.2478 16.6314 23.6976 27.8634
      R 2.1823 2.9394 3.4872 4.0204 6.0167 9.3337 11.0755 11.4068 13.3270 15.4545
    2 rows × 60 columns
```

Splitting training and testing data

```
1 x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.2, stratify=y, random_state=42)

1 len(x_test)

42

1 len(x_train)

166
```

Model training

```
1 model = LogisticRegression()
```

Traing our loogistic regression model

```
1 model.fit(x_train, y_train)
2
3 # The .fit() method takes the feature matrix X_train and target variable y_train as arguments and fits
4 # the logistic regression model to the training data. During this process, the model learns the coefficients
5 # and intercept that define the decision boundary.
6
7 # After the model is trained using .fit(), we can use the trained model to make predictions on new data using
8 # the .predict() method.
```

Model Evaluation

```
1 # let's see the accuracy on training data first
2 x_train_prediction = model.predict(x_train)

1 x_train_accuracy = accuracy_score(x_train_prediction, y_train)

1 print('Accuracy on training data :', x_train_accuracy)
2 # Accuracy on training data : 0.8373493975903614

Accuracy on training data : 0.8373493975903614

1 # Accuracy on test data
2 x_test_prediction = model.predict(x_test)
3 x_test_accuracy = accuracy_score(x_test_prediction, y_test)
4 print('Accuracy on testing data :', x_test_accuracy)
5 # Accuracy on testing data : 0.8571428571428571
6 # Don't know how i'm getting better prediction in my test data than train data Accuracy on testing data : 0.8571428571428571
```

Maling a predictive system

```
1 input_data = ()
2 input_data_as_nparray = np.asarray(input_data)
```