	Imorting the Dependencies
In [1]:	<pre>import numpy as np import pandas as pd from sklearn.model_selection import train_test_split</pre>
	<pre>from sklearn.linear_model import LogisticRegression from sklearn.metrics import accuracy_score</pre>
In [2]:	Data Collection and Data Processing #loading the dataset to a panda Dataframe
In [3]:	<pre>sonar_data = pd.read_csv(r"C:\Users\hp\Downloads\Copy of sonar data.csv", header=None) sonar_data.head()</pre>
Out[3]:	0 1 2 3 4 5 6 7 8 9 51 52 53 54 55 56 57 58 59 60 0 0.0200 0.0371 0.0428 0.0207 0.0984 0.0539 0.1611 0.3109 0.2111 0.0027 0.0052 0.0160 0.0180 0.0180 0.0084 0.0090 0.0032 R 1 0.0453 0.0523 0.0883 0.0180 0.0183 0.0180 0.0180 0.0180 0.0092 0.0044 R 2 0.0262 0.0583 0.1083 0.2431 0.3711 0.5598 0.6194 0.0242 0.0160 0.0140 0.0140 0.0044 0.0045 0.0045 0.0044 0.0045 0.0044 0.0046 0.0140 0.0044 0.0045 0.0045 0.0047 0.0048 0.0040 0.0044 0.0040 0.0044 0.0040 0.0040 0.0044 0.0040 0
In [4]:	<pre># number of rows and columns sonar_data.shape (208, 61)</pre>
Out[4]: In [5]:	#descibe the statistical measures of the data sonar_data.describe()
Out[5]:	count 1 2 3 4 5 6 7 8 9 50 51 52 53 54 55 56 count 28.000000 28.00000 28.00000 28.00000 28.00000 28.00000 28.00000 28.00000 28.00000 28.000000 28.000000 28.000000 28.000000<
In [6]: Out[6]:	<pre>sonar_data[60].value_counts() M 111 R 97</pre>
In [7]:	Name: 60, dtype: int64 sonar_data.groupby(60).mean()
Out[7]:	60 M 0.034989 0.045544 0.050720 0.064768 0.086715 0.11864 0.128359 0.149832 0.213492 0.251022 0.019352 0.016014 0.011643 0.012185 0.009923 0.008914 0.007825 0.009060 0.008695 0.006930
	R 0.022498 0.030303 0.035951 0.041447 0.062028 0.096224 0.114180 0.117596 0.137392 0.159325 0.012311 0.010453 0.009640 0.009518 0.008567 0.007430 0.007814 0.006677 0.007078 0.006024 2 rows × 60 columns
In [8]:	<pre>#separating data and Labels X = sonar_data.drop(columns=60, axis=1)</pre>
In [9]:	<pre>Y = sonar_data[60] print(X)</pre>
	print(Y) 0 1 2 3 4 5 6 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 0.2260 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 4 0.0762 0.0666 0.0481 0.0394 0.0590 0.0649 0.1209 0.2467 0.3564
	9 50 51 52 53 54 55 56 \ 0 0.2111 0.0232 0.0027 0.0065 0.0159 0.0072 0.0167 0.0180 1 0.2872 0.0125 0.0084 0.0089 0.0048 0.0094 0.0191 0.0140 2 0.6194 0.0033 0.0232 0.0166 0.0095 0.0180 0.0244 0.0316
	3
	0 0.0084 0.0090 0.0032 1 0.0049 0.0052 0.0044 2 0.0164 0.0095 0.0078 3 0.0044 0.0040 0.0117
	4 0.0048 0.0107 0.0094 203 0.0115 0.0193 0.0157 204 0.0032 0.0062 0.0067
	205
	[208 rows x 60 columns] 0 R 1 R 2 R
	3 R 4 R 203 M
	204 M 205 M 206 M 207 M
In [10]:	<pre>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)</pre>
In [11]:	print(X.shape, X_train.shape, X_test.shape) (208, 60) (187, 60) (21, 60)
In [12]: In [13]:	<pre>#Model Training - Logistic Regression model=LogisticRegression() print(X_train)</pre>
111 [10].	<pre>print(Y_train) 0 1 2 3 4 5 6 7 8 \ 115 0.0414 0.0436 0.0447 0.0844 0.0419 0.1215 0.2002 0.1516 0.0818</pre>
	38 0.0123 0.0022 0.0196 0.0206 0.0180 0.0492 0.0033 0.0398 0.0791 56 0.0152 0.0102 0.0113 0.0263 0.0097 0.0391 0.0857 0.0915 0.0949 123 0.0270 0.0163 0.0341 0.0247 0.0822 0.1256 0.1323 0.1584 0.2017 18 0.0270 0.0092 0.0145 0.0278 0.0412 0.0757 0.1026 0.1138 0.0794
	131 0.1150 0.1163 0.0866 0.0358 0.0232 0.1267 0.2417 0.2661 0.4346 203 0.0187 0.0346 0.0168 0.0177 0.0393 0.1630 0.2028 0.1694 0.2328 9 50 51 52 53 54 55 56 \ 115 0.1975 0.0222 0.0045 0.0136 0.0113 0.0053 0.0165 0.0141
	38
	140 0.2058 0.0798 0.0376 0.0143 0.0272 0.0127 0.0166 0.0095 5 0.3039 0.0104 0.0045 0.0014 0.0038 0.0013 0.0089 0.0057 154 0.2169 0.0039 0.0053 0.0029 0.0020 0.0013 0.0029 0.0020
	131 0.5378 0.0228 0.0099 0.0065 0.0085 0.0166 0.0110 0.0190 203 0.2684 0.0203 0.0116 0.0098 0.0199 0.0033 0.0101 0.0065 57 58 59
	115 0.0077 0.0246 0.0198 38 0.0058 0.0047 0.0071 56 0.0011 0.0034 0.0033 123 0.0094 0.0105 0.0093 18 0.0132 0.0070 0.0088
	18
	131 0.0141 0.0068 0.0086 203 0.0115 0.0193 0.0157 [187 rows x 60 columns]
	115 M 38 R 56 R 123 M
	18 R 140 M 5 R 154 M
	131 M 203 M Name: 60, Length: 187, dtype: object
In [14]: Out[14]:	# Training the Logistic Regression model with training data model.fit(X_train,Y_train) * LogisticRegression
	LogisticRegression()
	Model Evaluation
In [15]:	<pre>#accuracy on the training data X_train_prediction=model.predict(X_train) training_data_accuracy=accuracy_score(X_train_prediction,Y_train)</pre>
In [16]:	print('Accuracy on training data :',training_data_accuracy) Accuracy on training data : 0.8342245989304813 #accuracy on the training data
In [17]:	<pre>#accuracy on the training data X_test_prediction=model.predict(X_test) test_data_accuracy=accuracy_score(X_test_prediction,Y_test)</pre>
In [18]:	print('Accuracy on test data :',test_data_accuracy) Accuracy on test data : 0.7619047619047619
Tn Fact	Making a Predictive System input data=(0.0039.0.0063.0.0152.0.0336.0.0310.0.0284.0.0396.0.0272.0.0323.0.0452.0.0492.0.0996.0.1424.0.1194.0.0628.0.0907.0.1177.0.1429.0.1223.0.1104.0.1847.0.3715.0.4382.0.5707.0
In [19]:	<pre>input_data=(0.0039,0.0063,0.0152,0.0336,0.0310,0.0284,0.0396,0.0272,0.0323,0.0452,0.0492,0.0996,0.1424,0.1194,0.0628,0.0907,0.1177,0.1429,0.1223,0.1104,0.1847,0.3715,0.4382,0.5707,0.000000000000000000000000000000</pre>
	<pre>#Reshape the numpy array as we are predicting for one insrances input_data_reshaped = input_data_as_numpy_array.reshape(1,-1) prediction = model.predict(input_data_reshaped) print(prediction)</pre>
	<pre>if(prediction[0]=='R'): print("The object is a Rock") else:</pre>
	<pre>else: print('The object is mine') ['R'] The object is a Rock</pre>
In [20]:	<pre>input_data=(0.0294,0.0123,0.0117,0.0113,0.0497,0.0998,0.1326,0.1117,0.2984,0.3473,0.4231,0.5044,0.5237,0.4398,0.3236,0.2956,0.3286,0.3231,0.4528,0.6339,0.7044,0.8314,0.8449,0.8512,0.4231,0.4231,0.4231,0.4231,0.5044,0.5237,0.4398,0.3236,0.2956,0.3286,0.3231,0.4528,0.6339,0.7044,0.8314,0.8449,0.8512,0.4231,0.4231,0.4231,0.4231,0.4231,0.5044,0.5237,0.4398,0.3236,0.2956,0.3286,0.3231,0.4528,0.6339,0.7044,0.8314,0.8449,0.8512,0.4231,0.4231,0.4231,0.4231,0.4231,0.5044,0.5237,0.4398,0.3236,0.2956,0.3286,0.3231,0.4528,0.6339,0.7044,0.8314,0.8449,0.8512,0.4231</pre>
	<pre>#Reshape the numpy array as we are predicting for one insrances input_data_reshaped = input_data_as_numpy_array.reshape(1,-1)</pre>
	<pre>prediction = model.predict(input_data_reshaped) print(prediction) if(prediction[0]=='R'): print("The object is a Rock")</pre>
	<pre>print("The object is a Rock") else: print('The object is mine') ['M']</pre>
	The object is mine