

```
In [1]: import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score
```

```
In [2]: sonar_data = pd.read_csv("Copy of sonar data.csv", header=None)
```

```
In [3]: sonar_data.head(10)
```

```
Out[3]:
```

	0	1	2	3	4	5	6	7	8	9	...	51	52	
0	0.0200	0.0371	0.0428	0.0207	0.0954	0.0986	0.1539	0.1601	0.3109	0.2111	...	0.0027	0.0065	0.
1	0.0453	0.0523	0.0843	0.0689	0.1183	0.2583	0.2156	0.3481	0.3337	0.2872	...	0.0084	0.0089	0.
2	0.0262	0.0582	0.1099	0.1083	0.0974	0.2280	0.2431	0.3771	0.5598	0.6194	...	0.0232	0.0166	0.
3	0.0100	0.0171	0.0623	0.0205	0.0205	0.0368	0.1098	0.1276	0.0598	0.1264	...	0.0121	0.0036	0.
4	0.0762	0.0666	0.0481	0.0394	0.0590	0.0649	0.1209	0.2467	0.3564	0.4459	...	0.0031	0.0054	0.
5	0.0286	0.0453	0.0277	0.0174	0.0384	0.0990	0.1201	0.1833	0.2105	0.3039	...	0.0045	0.0014	0.
6	0.0317	0.0956	0.1321	0.1408	0.1674	0.1710	0.0731	0.1401	0.2083	0.3513	...	0.0201	0.0248	0.
7	0.0519	0.0548	0.0842	0.0319	0.1158	0.0922	0.1027	0.0613	0.1465	0.2838	...	0.0081	0.0120	0.
8	0.0223	0.0375	0.0484	0.0475	0.0647	0.0591	0.0753	0.0098	0.0684	0.1487	...	0.0145	0.0128	0.
9	0.0164	0.0173	0.0347	0.0070	0.0187	0.0671	0.1056	0.0697	0.0962	0.0251	...	0.0090	0.0223	0.

10 rows × 61 columns

```
In [4]: sonar_data.shape
```

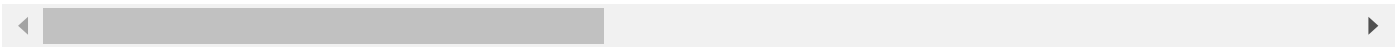
```
Out[4]: (208, 61)
```

```
In [5]: sonar_data.describe() # describe --> statistical measures the data
```

Out[5]:

	0	1	2	3	4	5	6	7
count	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
std	0.022991	0.032960	0.038428	0.046528	0.055552	0.059105	0.061788	0.085152
min	0.001500	0.000600	0.001500	0.005800	0.006700	0.010200	0.003300	0.005500
25%	0.013350	0.016450	0.018950	0.024375	0.038050	0.067025	0.080900	0.080425
50%	0.022800	0.030800	0.034300	0.044050	0.062500	0.092150	0.106950	0.112100
75%	0.035550	0.047950	0.057950	0.064500	0.100275	0.134125	0.154000	0.169600
max	0.137100	0.233900	0.305900	0.426400	0.401000	0.382300	0.372900	0.459000

8 rows × 60 columns



In [6]: `sonar_data.info()`

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 208 entries, 0 to 207
```

```
Data columns (total 61 columns):
```

#	Column	Non-Null Count	Dtype
0	0	208 non-null	float64
1	1	208 non-null	float64
2	2	208 non-null	float64
3	3	208 non-null	float64
4	4	208 non-null	float64
5	5	208 non-null	float64
6	6	208 non-null	float64
7	7	208 non-null	float64
8	8	208 non-null	float64
9	9	208 non-null	float64
10	10	208 non-null	float64
11	11	208 non-null	float64
12	12	208 non-null	float64
13	13	208 non-null	float64
14	14	208 non-null	float64
15	15	208 non-null	float64
16	16	208 non-null	float64
17	17	208 non-null	float64
18	18	208 non-null	float64
19	19	208 non-null	float64
20	20	208 non-null	float64
21	21	208 non-null	float64
22	22	208 non-null	float64
23	23	208 non-null	float64
24	24	208 non-null	float64
25	25	208 non-null	float64
26	26	208 non-null	float64
27	27	208 non-null	float64
28	28	208 non-null	float64
29	29	208 non-null	float64
30	30	208 non-null	float64
31	31	208 non-null	float64
32	32	208 non-null	float64
33	33	208 non-null	float64
34	34	208 non-null	float64
35	35	208 non-null	float64
36	36	208 non-null	float64
37	37	208 non-null	float64
38	38	208 non-null	float64
39	39	208 non-null	float64
40	40	208 non-null	float64
41	41	208 non-null	float64
42	42	208 non-null	float64
43	43	208 non-null	float64
44	44	208 non-null	float64
45	45	208 non-null	float64
46	46	208 non-null	float64
47	47	208 non-null	float64
48	48	208 non-null	float64
49	49	208 non-null	float64
50	50	208 non-null	float64
51	51	208 non-null	float64
52	52	208 non-null	float64
53	53	208 non-null	float64
54	54	208 non-null	float64

```

55 55      208 non-null    float64
56 56      208 non-null    float64
57 57      208 non-null    float64
58 58      208 non-null    float64
59 59      208 non-null    float64
60 60      208 non-null    object

```

dtypes: float64(60), object(1)

memory usage: 99.2+ KB

In [7]: `sonar_data[60].value_counts()` # M --> Mine And R --> Rock

Out[7]:

```

M      111
R       97

```

Name: 60, dtype: int64

In [8]: `sonar_data.groupby(60).mean(60)`

Out[8]:

	0	1	2	3	4	5	6	7	8	9
60										
M	0.034989	0.045544	0.050720	0.064768	0.086715	0.111864	0.128359	0.149832	0.213492	0.251022
R	0.022498	0.030303	0.035951	0.041447	0.062028	0.096224	0.114180	0.117596	0.137392	0.159325

2 rows × 60 columns

In [9]: `# Seprating data and Lables`

In [10]:

```

X = sonar_data.drop(columns=60,axis=1)
Y = sonar_data[60]

```

In [11]: `X`

```
Out[11]:
```

	0	1	2	3	4	5	6	7	8	9	...	50	51
0	0.0200	0.0371	0.0428	0.0207	0.0954	0.0986	0.1539	0.1601	0.3109	0.2111	...	0.0232	0.0027
1	0.0453	0.0523	0.0843	0.0689	0.1183	0.2583	0.2156	0.3481	0.3337	0.2872	...	0.0125	0.0084
2	0.0262	0.0582	0.1099	0.1083	0.0974	0.2280	0.2431	0.3771	0.5598	0.6194	...	0.0033	0.0232
3	0.0100	0.0171	0.0623	0.0205	0.0205	0.0368	0.1098	0.1276	0.0598	0.1264	...	0.0241	0.0121
4	0.0762	0.0666	0.0481	0.0394	0.0590	0.0649	0.1209	0.2467	0.3564	0.4459	...	0.0156	0.0031
...
203	0.0187	0.0346	0.0168	0.0177	0.0393	0.1630	0.2028	0.1694	0.2328	0.2684	...	0.0203	0.0116
204	0.0323	0.0101	0.0298	0.0564	0.0760	0.0958	0.0990	0.1018	0.1030	0.2154	...	0.0051	0.0061
205	0.0522	0.0437	0.0180	0.0292	0.0351	0.1171	0.1257	0.1178	0.1258	0.2529	...	0.0155	0.0160
206	0.0303	0.0353	0.0490	0.0608	0.0167	0.1354	0.1465	0.1123	0.1945	0.2354	...	0.0042	0.0086
207	0.0260	0.0363	0.0136	0.0272	0.0214	0.0338	0.0655	0.1400	0.1843	0.2354	...	0.0181	0.0146

208 rows × 60 columns

```
In [12]: Y
```

```
Out[12]:
```

0	R
1	R
2	R
3	R
4	R
...	...
203	M
204	M
205	M
206	M
207	M

Name: 60, Length: 208, dtype: object

Training and Test data

```
In [13]: X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.1,random_state=1,stra
```

```
In [14]: X.shape,X_train.shape,X_test.shape
```

```
Out[14]: ((208, 60), (187, 60), (21, 60))
```

```
In [15]: print(X_train)
          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	
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	
..	
140	0.0412	0.1135	0.0518	0.0232	0.0646	0.1124	0.1787	0.2407	0.2682	
5	0.0286	0.0453	0.0277	0.0174	0.0384	0.0990	0.1201	0.1833	0.2105	
154	0.0117	0.0069	0.0279	0.0583	0.0915	0.1267	0.1577	0.1927	0.2361	
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	0.0475	...	0.0149	0.0125	0.0134	0.0026	0.0038	0.0018	0.0113	
56	0.1504	...	0.0048	0.0049	0.0041	0.0036	0.0013	0.0046	0.0037	
123	0.2122	...	0.0197	0.0189	0.0204	0.0085	0.0043	0.0092	0.0138	
18	0.1520	...	0.0045	0.0084	0.0010	0.0018	0.0068	0.0039	0.0120	
..	
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
..
140	0.0225	0.0098	0.0085
5	0.0027	0.0051	0.0062
154	0.0062	0.0026	0.0052
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

Model Training Logistic regression

```
In [16]: # Training the Logistic regression model with training data
```

```
In [17]: model = LogisticRegression()
```

```
In [18]: model.fit(X_train,Y_train)
```

```
Out[18]: LogisticRegression()
```

Model Evaluation

```
In [19]: # Accuracy on training data
```

```
In [20]: X_train_prediction = model.predict(X_train)
training_data_accuracy = accuracy_score(X_train_prediction,Y_train)
```

```
In [21]: print(' Accuracy on training data :' ,training_data_accuracy)
```

```
Accuracy on training data : 0.8342245989304813
```

```
In [22]: # Accuracy on training data
```

```
X_test_prediction = model.predict(X_test)
test_data_accuracy = accuracy_score(X_test_prediction,Y_test)
```

```
In [23]: print(' Accuracy on test data :' , accuracy_score(X_test_prediction,Y_test) )
```

```
Accuracy on test data : 0.7619047619047619
```

Making a predictive system

```
In [25]: input_data = (0.0519,0.0548,0.0842,0.0319,0.1158,0.0922,0.1027,0.0613,0.1465,0.2838,0.
0.0121,0.0097,0.0085,0.0047,0.0048,0.0053)
```

```
# 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_reshape = input_data_as_numpy_array.reshape(1,-1)
```

```
prediction = model.predict(input_data_reshape )
```

```
print(prediction)
```

```
if (prediction[0]=='M'):
    print('The object is a Mine')
```

```
else:
    print('The object is a Rock')
```

```
['M']
```

```
The object is a Mine
```