



```
In [13]: 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

# Loading the dataset to a pandas DataFrame
print("Loading dataset:")
sonar_data = pd.read_csv('/content/sonar_data.csv', header=None)

# Number of rows and columns
print("\nDataset Shape (rows, columns):")
print(sonar_data.shape)

# Statistical measures of the data
print("\nStatistical Description of Dataset:")
print(sonar_data.describe())

# Count of Rock (R) and Mine (M)
print("\nCount of Rock and Mine Labels:")
print(sonar_data[60].value_counts())

# Mean of each feature per label
print("\nMean of each column grouped by label (R/M):")
print(sonar_data.groupby(60).mean())

# Separating data and labels
X = sonar_data.drop(columns=60, axis=1)
Y = sonar_data[60]

print("\nFeatures (X):")
print(X)

print("\nLabels (Y):")
print(Y)

# Splitting into training and testing data
print("\nSplitting data into training and testing sets:")
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.1, strat

print("\nShapes of datasets:")
print(f"X: {X.shape}, X_train: {X_train.shape}, X_test: {X_test.shape}")

print("\nTraining Feature Data (X_train):")
print(X_train)

print("\nTraining Labels (Y_train):")
print(Y_train)

# Model Training
print("\nTraining Logistic Regression model:")
model = LogisticRegression()
model.fit(X_train, Y_train)
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# Accuracy on training data
X_train_prediction = model.predict(X_train)
training_data_accuracy = accuracy_score(X_train_prediction, Y_train)
print("\nAccuracy on training data:", training_data_accuracy)

# Accuracy on test data
X_test_prediction = model.predict(X_test)
test_data_accuracy = accuracy_score(X_test_prediction, Y_test)
print("Accuracy on test data:", test_data_accuracy)

# Making a prediction on new input data
print("\nMaking a prediction on custom input data:")
input_data = (
    0.0307,0.0523,0.0653,0.0521,0.0611,0.0577,0.0665,0.0664,0.1460,0.2792,
    0.3877,0.4992,0.4981,0.4972,0.5607,0.7339,0.8230,0.9173,0.9975,0.9911,
    0.8240,0.6498,0.5980,0.4862,0.3150,0.1543,0.0989,0.0284,0.1008,0.2636,
    0.2694,0.2930,0.2925,0.3998,0.3660,0.3172,0.4609,0.4374,0.1820,0.3376,
    0.6202,0.4448,0.1863,0.1420,0.0589,0.0576,0.0672,0.0269,0.0245,0.0190,
    0.0063,0.0321,0.0189,0.0137,0.0277,0.0152,0.0052,0.0121,0.0124,0.0055
)

# Changing input data to a numpy array
input_data_as_numpy_array = np.asarray(input_data)

# Reshape the numpy array
input_data_reshaped = input_data_as_numpy_array.reshape(1, -1)

# Predict using the trained model
prediction = model.predict(input_data_reshaped)
print("\nPrediction for the input data:", prediction[0])

# Output result
if prediction[0] == 'R':
    print("The object is a Rock")
else:
    print("The object is a Mine")

```

Loading dataset:

Dataset Shape (rows, columns):
(208, 61)

Statistical Description of Dataset:

	0	1	2	3	4	5
\						
count	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
std	0.022991	0.032960	0.038428	0.046528	0.055552	0.059105
min	0.001500	0.000600	0.001500	0.005800	0.006700	0.010200
25%	0.013350	0.016450	0.018950	0.024375	0.038050	0.067025
50%	0.022800	0.030800	0.034300	0.044050	0.062500	0.092150
75%	0.035550	0.047950	0.057950	0.064500	0.100275	0.134125
max	0.137100	0.233900	0.305900	0.426400	0.401000	0.382300

	6	7	8	9	...	50	\
count	208.000000	208.000000	208.000000	208.000000	...	208.000000	
mean	0.121747	0.134799	0.178003	0.208259	...	0.016069	
std	0.061788	0.085152	0.118387	0.134416	...	0.012008	
min	0.003300	0.005500	0.007500	0.011300	...	0.000000	
25%	0.080900	0.080425	0.097025	0.111275	...	0.008425	
50%	0.106950	0.112100	0.152250	0.182400	...	0.013900	
75%	0.154000	0.169600	0.233425	0.268700	...	0.020825	
max	0.372900	0.459000	0.682800	0.710600	...	0.100400	

	51	52	53	54	55	56
\						
count	208.000000	208.000000	208.000000	208.000000	208.000000	208.000000
mean	0.013420	0.010709	0.010941	0.009290	0.008222	0.007820
std	0.009634	0.007060	0.007301	0.007088	0.005736	0.005785
min	0.000800	0.000500	0.001000	0.000600	0.000400	0.000300
25%	0.007275	0.005075	0.005375	0.004150	0.004400	0.003700
50%	0.011400	0.009550	0.009300	0.007500	0.006850	0.005950
75%	0.016725	0.014900	0.014500	0.012100	0.010575	0.010425
max	0.070900	0.039000	0.035200	0.044700	0.039400	0.035500

	57	58	59
count	208.000000	208.000000	208.000000
mean	0.007949	0.007941	0.006507
std	0.006470	0.006181	0.005031
min	0.000300	0.000100	0.000600
25%	0.003600	0.003675	0.003100
50%	0.005800	0.006400	0.005300
75%	0.010350	0.010325	0.008525
max	0.044000	0.036400	0.043900

[8 rows x 60 columns]

Count of Rock and Mine Labels:

60

M 111

R 97

Name: count, dtype: int64

Mean of each column grouped by label (R/M):

	0	1	2	3	4	5	6	\
60								
M	0.034989	0.045544	0.050720	0.064768	0.086715	0.111864	0.128359	
R	0.022498	0.030303	0.035951	0.041447	0.062028	0.096224	0.114180	
	7	8	9	...	50	51	52	53
\								
60				...				
M	0.149832	0.213492	0.251022	...	0.019352	0.016014	0.011643	0.012185
R	0.117596	0.137392	0.159325	...	0.012311	0.010453	0.009640	0.009518
	54	55	56	57	58	59		
60								
M	0.009923	0.008914	0.007825	0.009060	0.008695	0.006930		
R	0.008567	0.007430	0.007814	0.006677	0.007078	0.006024		

[2 rows x 60 columns]

Features (X):

	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	0.2280	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	
..	
203	0.0187	0.0346	0.0168	0.0177	0.0393	0.1630	0.2028	0.1694	0.2328	
204	0.0323	0.0101	0.0298	0.0564	0.0760	0.0958	0.0990	0.1018	0.1030	
205	0.0522	0.0437	0.0180	0.0292	0.0351	0.1171	0.1257	0.1178	0.1258	
206	0.0303	0.0353	0.0490	0.0608	0.0167	0.1354	0.1465	0.1123	0.1945	
207	0.0260	0.0363	0.0136	0.0272	0.0214	0.0338	0.0655	0.1400	0.1843	
	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	0.1264	...	0.0241	0.0121	0.0036	0.0150	0.0085	0.0073	0.0050	
4	0.4459	...	0.0156	0.0031	0.0054	0.0105	0.0110	0.0015	0.0072	
..	
203	0.2684	...	0.0203	0.0116	0.0098	0.0199	0.0033	0.0101	0.0065	
204	0.2154	...	0.0051	0.0061	0.0093	0.0135	0.0063	0.0063	0.0034	
205	0.2529	...	0.0155	0.0160	0.0029	0.0051	0.0062	0.0089	0.0140	
206	0.2354	...	0.0042	0.0086	0.0046	0.0126	0.0036	0.0035	0.0034	
207	0.2354	...	0.0181	0.0146	0.0129	0.0047	0.0039	0.0061	0.0040	
	57	58	59							
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  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]

Labels (Y):

```

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

Splitting data into training and testing sets:

Shapes of datasets:

X: (208, 60), X_train: (187, 60), X_test: (21, 60)

Training Feature Data (X_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]

Training Labels (Y_train):

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

Training Logistic Regression model:

Accuracy on training data: 0.8342245989304813

Accuracy on test data: 0.7619047619047619

Making a prediction on custom input data:

Prediction for the input data: M

The object is a Mine

In []: