Tn [4].	
In [1]:	<pre>import pandas as pd import numpy as np import matplotlib.pyplot as plt import seaborn as sns</pre>
In [2]:	df=pd.read_csv("sonar.csv", header=None)
In [3]:	df.shape
Out[3]: In [4]:	(208, 61) df.head()
	1
In [5]:	df.columns
Out[5]:	<pre>Int64Index([0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16,</pre>
Out[6]:	0 1 2 3 4 5 6 7 8 9 51 52 53 54 55 56 57 58 59 60 203 0.0187 0.0346 0.0168 0.0177 0.0393 0.1630 0.2028 0.1694 0.2328 0.2684 0.0199 0.0033 0.0101 0.0065 0.0115 0.0193 0.0157 M
	204 0.0323 0.0101 0.0298 0.0564 0.0760 0.0958 0.0990 0.1018 0.1030 0.2154 0.0061 0.0093 0.0135 0.0063 0.0063 0.0063 0.0034 0.0032 0.0062 0.0067 M 205 0.0522 0.0437 0.0180 0.0292 0.0351 0.1171 0.1257 0.1178 0.1258 0.2529 0.0160 0.0029 0.0051 0.0062 0.0089 0.0140 0.0138 0.0077 0.0031 M 206 0.0303 0.0353 0.0490 0.0608 0.0167 0.1354 0.1465 0.1123 0.1945 0.2354 0.0086 0.0046 0.0126 0.0036 0.0035 0.0034 0.0079 0.0036 0.0048 M 207 0.0260 0.0363 0.0136 0.0272 0.0214 0.0338 0.0655 0.1400 0.1843 0.2354 0.0146 0.0129 0.0047 0.0039 0.0061 0.0040 0.0036 0.0046 0.0015 M 5 rows × 61 columns
In [7]: In [8]:	<pre>df.rename(columns={60:'Resources'},inplace=True)</pre>
Out[8]:	df.head() 0 1 2 3 4 5 6 7 8 9 51 52 53 54 55 56 57 58 59 Resources
	0 0.0200 0.0371 0.0428 0.0207 0.0954 0.096 0.1539 0.1601 0.3109 0.2111 0.0027 0.0065 0.0159 0.0072 0.0167 0.0160 0.0180 0.0084 0.0090 0.0032 R 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 0.0048 0.0094 0.0191 0.0140 0.0049 0.0052 0.0044 R 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 0.0095 0.0180 0.0244 0.0016 0.0016 0.0164 0.0095 0.0078 R 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 0.0150 0.0016 0.0150 0.0016 0.0016 0.0017 0.0017 0.0017 0.0017 0.0017 0.0018 0.0018 0.0017 0.0018 0.0018 0.0017 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.0018 0.001
In [9]:	<pre>df['Resources'].value_counts()</pre>
Out[9]:	M 111 R 97 Name: Resources, dtype: int64
In [10]: Out[10]:	<pre>df.describe() 0 1 2 3 4 5 6 7 8 9 50 51 52 53 54 55 56</pre>
	count 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.000000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.00000 208.000000 208.00000 208.00000 208.00000
In [11]:	df.info()
In [12]: In [13]:	**Cabas* (paralas cottes* (frant Cabas*) paralas **Bolton** (cottes*) **
In [14]:	<pre>False df.groupby('Resources').mean()</pre>
Out[14]:	0 1 2 3 4 5 6 7 8 9 50 51 52 53 54 55 56 57 58 59 Resources
	M 0.034989 0.045544 0.050720 0.064768 0.086715 0.111864 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 [15]: In [16]:	<pre>from sklearn.model_selection import train_test_split x=df.drop(columns='Resources',axis=1)</pre>
In [17]:	y=df['Resources']
In [17].	<pre>X_train, X_test, y_train, y_test=train_test_split(x, y, test_size=0.3, stratify=y, random_state=10)</pre> <pre>from sklearn.linear_model import LogisticRegression</pre>
In [19]:	lr=LogisticRegression()
In [20]:	<pre>lr.fit(X_train,y_train)</pre>
	LogisticRegression()
In [21]: In [22]:	<pre>pred=lr.predict(X_test)</pre> <pre>from sklears metrics import accuracy score classification report confusion matrix</pre>
In [23]:	<pre>from sklearn.metrics import accuracy_score, classification_report, confusion_matrix print(accuracy_score(y_test, pred))</pre>
	0.77777777778
In [24]:	print(classification_report(y_test, pred)) precision recall f1-score support M 0.86 0.71 0.77 34 R 0.71 0.86 0.78 29
	accuracy 0.78 63 macro avg 0.79 0.78 0.78 63 weighted avg 0.79 0.78 0.78 63
In [25]:	<pre>print(confusion_matrix(y_test,pred)) [[24 10] [4 25]]</pre>