# In [1]:

```
import pandas as pd
import numpy as np
from sklearn.linear_model import LogisticRegression
from sklearn.preprocessing import StandardScaler
```

# In [2]:

```
df=pd.read_csv(r"C:\Users\HP\OneDrive\Documents\Ionospear.csv")
df
```

# Out[2]:

1	0	0.99539	-0.05889	0.85243	0.02306	0.83398	-0.37708	1.1	0.0376		
1	0	1.00000	-0.18829	0.93035	-0.36156	-0.10868	-0.93597	1.00000	-0.04549		-
1	0	1.00000	-0.03365	1.00000	0.00485	1.00000	-0.12062	0.88965	0.01198		-
1	0	1.00000	-0.45161	1.00000	1.00000	0.71216	-1.00000	0.00000	0.00000		
1	0	1.00000	-0.02401	0.94140	0.06531	0.92106	-0.23255	0.77152	-0.16399		-
1	0	0.02337	-0.00592	-0.09924	-0.11949	-0.00763	-0.11824	0.14706	0.06637		-
1	0	0.83508	0.08298	0.73739	-0.14706	0.84349	-0.05567	0.90441	-0.04622		•
1	0	0.95113	0.00419	0.95183	-0.02723	0.93438	-0.01920	0.94590	0.01606		
1	0	0.94701	-0.00034	0.93207	-0.03227	0.95177	-0.03431	0.95584	0.02446		
1	0	0.90608	-0.01657	0.98122	-0.01989	0.95691	-0.03646	0.85746	0.00110		
1	0	0.84710	0.13533	0.73638	-0.06151	0.87873	0.08260	0.88928	-0.09139		
	1 1 1 1 1  1 1 1	1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0	1 0 1.00000 1 0 1.00000 1 0 1.00000 1 0 1.00000 1 0 0.02337 1 0 0.83508 1 0 0.95113 1 0 0.94701 1 0 0.90608	1       0       1.00000       -0.18829         1       0       1.00000       -0.03365         1       0       1.00000       -0.45161         1       0       1.00000       -0.02401         1       0       0.02337       -0.00592               1       0       0.83508       0.08298         1       0       0.95113       0.00419         1       0       0.94701       -0.00034         1       0       0.90608       -0.01657	1       0       1.00000       -0.18829       0.93035         1       0       1.00000       -0.03365       1.00000         1       0       1.00000       -0.45161       1.00000         1       0       1.00000       -0.02401       0.94140         1       0       0.02337       -0.00592       -0.09924               1       0       0.83508       0.08298       0.73739         1       0       0.95113       0.00419       0.95183         1       0       0.94701       -0.00034       0.93207         1       0       0.90608       -0.01657       0.98122	1       0       1.000000       -0.18829       0.93035       -0.36156         1       0       1.000000       -0.03365       1.000000       0.00485         1       0       1.000000       -0.45161       1.000000       1.00000         1       0       1.000000       -0.02401       0.94140       0.06531         1       0       0.02337       -0.00592       -0.09924       -0.11949                1       0       0.83508       0.08298       0.73739       -0.14706         1       0       0.95113       0.00419       0.95183       -0.02723         1       0       0.94701       -0.00034       0.93207       -0.03227         1       0       0.90608       -0.01657       0.98122       -0.01989	1       0       1.00000       -0.18829       0.93035       -0.36156       -0.10868         1       0       1.00000       -0.03365       1.00000       0.00485       1.00000         1       0       1.00000       -0.45161       1.00000       1.00000       0.71216         1       0       1.00000       -0.02401       0.94140       0.06531       0.92106         1       0       0.02337       -0.00592       -0.09924       -0.11949       -0.00763                  1       0       0.83508       0.08298       0.73739       -0.14706       0.84349         1       0       0.95113       0.00419       0.95183       -0.02723       0.93438         1       0       0.94701       -0.00034       0.93207       -0.03227       0.95177         1       0       0.90608       -0.01657       0.98122       -0.01989       0.95691	1       0       1.000000       -0.18829       0.93035       -0.36156       -0.10868       -0.93597         1       0       1.00000       -0.03365       1.00000       0.00485       1.00000       -0.12062         1       0       1.00000       -0.45161       1.00000       1.00000       0.71216       -1.00000         1       0       1.00000       -0.02401       0.94140       0.06531       0.92106       -0.23255         1       0       0.02337       -0.00592       -0.09924       -0.11949       -0.00763       -0.11824                   1       0       0.83508       0.08298       0.73739       -0.14706       0.84349       -0.05567         1       0       0.95113       0.00419       0.95183       -0.02723       0.93438       -0.01920         1       0       0.94701       -0.00034       0.93207       -0.03227       0.95177       -0.03431         1       0       0.90608       -0.01657       0.98122       -0.01989       0.95691       -0.03646	1       0       1.000000       -0.18829       0.93035       -0.36156       -0.10868       -0.93597       1.00000         1       0       1.00000       -0.03365       1.00000       0.00485       1.00000       -0.12062       0.88965         1       0       1.00000       -0.45161       1.00000       1.00000       0.71216       -1.00000       0.00000         1       0       1.00000       -0.02401       0.94140       0.06531       0.92106       -0.23255       0.77152         1       0       0.02337       -0.00592       -0.09924       -0.11949       -0.00763       -0.11824       0.14706                   1       0       0.83508       0.08298       0.73739       -0.14706       0.84349       -0.05567       0.90441         1       0       0.95113       0.00419       0.95183       -0.02723       0.93438       -0.01920       0.94590         1       0       0.94701       -0.00034       0.93207       -0.03227       0.95177       -0.03431       0.95584         1       0       0.90608       -0.01657       0.98122       -0.01	1       0       1.000000       -0.18829       0.93035       -0.36156       -0.10868       -0.93597       1.00000       -0.04549         1       0       1.00000       -0.03365       1.00000       0.00485       1.00000       -0.12062       0.88965       0.01198         1       0       1.00000       -0.45161       1.00000       1.00000       0.71216       -1.00000       0.00000       0.00000         1       0       1.00000       -0.02401       0.94140       0.06531       0.92106       -0.23255       0.77152       -0.16399         1       0       0.02337       -0.00592       -0.09924       -0.11949       -0.00763       -0.11824       0.14706       0.06637	1       0       1.00000       -0.18829       0.93035       -0.36156       -0.10868       -0.93597       1.00000       -0.04549          1       0       1.00000       -0.03365       1.00000       0.00485       1.00000       -0.12062       0.88965       0.01198          1       0       1.00000       -0.45161       1.00000       1.00000       0.71216       -1.00000       0.00000       0.00000          1       0       1.00000       -0.02401       0.94140       0.06531       0.92106       -0.23255       0.77152       -0.16399          1       0       0.02337       -0.00592       -0.09924       -0.11949       -0.00763       -0.11824       0.14706       0.06637

350 rows × 35 columns

**→** 

# In [3]:

```
pd.set_option('display.max_rows',10000000000)
pd.set_option('display.max.columns',10000000000)
pd.set_option('display.width',95)
```

## In [4]:

```
1 print('The DataFrame has %d Rows and %d columns'%(df.shape))
```

The DataFrame has 350 Rows and 35 columns

```
In [5]:
```

```
1 df.head()
```

## Out[5]:

	1	0	0.99539	-0.05889	0.85243	0.02306	0.83398	-0.37708	1.1	0.0376	0.85243
0	1	0	1.00000	-0.18829	0.93035	-0.36156	-0.10868	-0.93597	1.00000	-0.04549	0.5087
1	1	0	1.00000	-0.03365	1.00000	0.00485	1.00000	-0.12062	0.88965	0.01198	0.7308
2	1	0	1.00000	-0.45161	1.00000	1.00000	0.71216	-1.00000	0.00000	0.00000	0.0000
3	1	0	1.00000	-0.02401	0.94140	0.06531	0.92106	-0.23255	0.77152	-0.16399	0.5279
4	1	0	0.02337	-0.00592	-0.09924	-0.11949	-0.00763	-0.11824	0.14706	0.06637	0.0378
4											•

### In [6]:

```
feature_matrix=df.iloc[:,0:34]
target_vector=df.iloc[:,-1]
```

### In [7]:

```
print('The Features matrix has %d Rows and %d column(s)'%(feature_matrix.shape))
print('The target matrix has %d Rows and %d column(s)'%(np.array(target_vector).res
```

The Features matrix has 350 Rows and 34 column(s)
The target matrix has 350 Rows and 1 column(s)

#### In [8]:

```
feature_matrix_standardized = StandardScaler().fit_transform(feature_matrix)
```

### In [9]:

```
algorithm = LogisticRegression(penalty='12',dual=False,tol=1e-4,C=1.0,fit_intercept
```

#### In [10]:

1 Logistic\_Regression\_Model = algorithm.fit(feature\_matrix\_standardized,target\_vector

## In [11]:

### In [12]:

```
predictions = Logistic_Regression_Model.predict(observation)
print('The Model predicted The observation to belong to class %s'%(predictions))
```

The Model predicted The observation to belong to class ['g']

# In [13]:

```
1 print('Algorithm was Trained To predict one of the two classes : %s'%(algorithm.cla
```

Algorithm was Trained To predict one of the two classes : ['b' 'g']

### In [14]:

```
print("""The Model ssays the probability of the observation we passed Belonging To
print()
print("""The Model ssays the probability of the observation we passed Belonging To
```

The Model ssays the probability of the observation we passed Belonging To class['b'] Is 0.008044378633976335

The Model ssays the probability of the observation we passed Belonging To class['g'] Is 0.9919556213660237

## In [19]:

```
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
feature=df.columns[0:3]
target=df.columns[-1]

x=df[feature].values
y=df[target].values
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.25)
lr = LinearRegression()
lr.fit(x_train,y_train)
print(lr.score(x-test,y_test))
print(lr.score(x_train,y_train))
```

```
Traceback (most recent call las
ValueError
t)
Cell In[19], line 9
      7 x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.
25)
      8 lr = LinearRegression()
----> 9 lr.fit(x_train,y_train)
     10 print(lr.score(x-test,y_test))
     11 print(lr.score(x_train,y_train))
File ~\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn
\linear model\_base.py:648, in LinearRegression.fit(self, X, y, sample_we
ight)
    644 n_jobs_ = self.n_jobs
    646 accept_sparse = False if self.positive else ["csr", "csc", "coo"]
--> 648 X, y = self._validate_data(
    649
           X, y, accept_sparse=accept_sparse, y_numeric=True, multi_outp
ut=True
    650 )
    652 sample_weight = _check_sample_weight(
    653
            sample_weight, X, dtype=X.dtype, only_non_negative=True
    656 X, y, X_offset, y_offset, X_scale = _preprocess_data(
    657
            Χ,
    658
            у,
   (\ldots)
    661
            sample_weight=sample_weight,
    662 )
File ~\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn
\base.py:584, in BaseEstimator._validate_data(self, X, y, reset, validate
_separately, **check_params)
    582
                y = check_array(y, input_name="y", **check_y_params)
    583
            else:
                X, y = check_X_y(X, y, **check_params)
--> 584
    585
            out = X, y
    587 if not no_val_X and check_params.get("ensure_2d", True):
File ~\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn
\utils\validation.py:1122, in check X y(X, y, accept sparse, accept large
_sparse, dtype, order, copy, force_all_finite, ensure_2d, allow_nd, multi
_output, ensure_min_samples, ensure_min_features, y_numeric, estimator)
            raise ValueError(
   1102
   1103
                f"{estimator_name} requires y to be passed, but the targe
t y is None"
   1104
   1106 X = check array(
   1107
            Χ,
   1108
            accept sparse=accept sparse,
   (\ldots)
   1119
            input name="X",
   1120 )
-> 1122 y = _check_y(y, multi_output=multi_output, y_numeric=y_numeric, e
stimator=estimator)
   1124 check_consistent_length(X, y)
   1126 return X, y
File ~\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn
```

\utils\validation.py:1147, in \_check\_y(y, multi\_output, y\_numeric, estima

localhost:8888/notebooks/python notebook/Inospear.ipynb