

▼ Import Library

```
import pandas as pd
```

```
import numpy as np
```


```
import matplotlib.pyplot as plt
```

▼ Import Data Set

```
df = pd.read_csv(r'https://github.com/YBI-Foundation/Dataset/raw/main/Fish.csv')
```

▼ Get the First Five Rows of Dataframe

```
df.head()
```

	Category	Species	Weight	Height	Width	Length1	Length2	Length3	
0	1	Bream	242.0	11.5200	4.0200	23.2	25.4	30.0	
1	1	Bream	290.0	12.4800	4.3056	24.0	26.3	31.2	
2	1	Bream	340.0	12.3778	4.6961	23.9	26.5	31.1	
3	1	Bream	363.0	12.7300	4.4555	26.3	29.0	33.5	
4	1	Bream	430.0	12.4440	5.1340	26.5	29.0	34.0	

▼ Get information about the DataSet

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 159 entries, 0 to 158
Data columns (total 8 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Category    159 non-null    int64
1   Species     159 non-null    object
2   Weight      159 non-null    float64
3   Height      159 non-null    float64
```

```

4   Width      159 non-null    float64
5   Length1    159 non-null    float64
6   Length2    159 non-null    float64
7   Length3    159 non-null    float64
dtypes: float64(6), int64(1), object(1)
memory usage: 10.1+ KB

```

▼ Get the Summary Statistics

```
df.describe()
```

	Category	Weight	Height	Width	Length1	Length2	Length3
count	159.000000	159.000000	159.000000	159.000000	159.000000	159.000000	159.000000
mean	3.264151	398.326415	8.970994	4.417486	26.247170	28.415723	31.227000
std	1.704249	357.978317	4.286208	1.685804	9.996441	10.716328	11.610000
min	1.000000	0.000000	1.728400	1.047600	7.500000	8.400000	8.800000
25%	2.000000	120.000000	5.944800	3.385650	19.050000	21.000000	23.150000
50%	3.000000	273.000000	7.786000	4.248500	25.200000	27.300000	29.400000
75%	4.500000	650.000000	12.365900	5.584500	32.700000	35.500000	39.650000
max	7.000000	1650.000000	18.957000	8.142000	59.000000	63.400000	68.000000

▼ Get the Shape Of the Data Frame

```
df.shape
```

```
(159, 8)
```

▼ Get the Column names

```
df.columns
```

```

Index(['Category', 'Species', 'Weight', 'Height', 'Width', 'Length1',
       'Length2', 'Length3'],
      dtype='object')

```

Define y(dependent or label or target variable) and X
(independent or features or attribute Variable)

```
y = df['Weight']
```

```
X = df.drop(['Category','Species','Weight'],axis =1)
```

```
y.shape
```

```
(159,)
```

```
X.shape
```

```
(159, 5)
```

```
y
```

```
0      242.0  
1      290.0  
2      340.0  
3      363.0  
4      430.0
```

```
...  
154     12.2  
155     13.4  
156     12.2  
157     19.7  
158     19.9
```

```
Name: Weight, Length: 159, dtype: float64
```

```
X
```

▼ Add Constant to Features (X) for intercept Estimation

```
import statsmodels.api as sm
```

```
/usr/local/lib/python3.7/dist-packages/statsmodels/tools/_testing.py:19: FutureWarning
import pandas.util.testing as tm
```

```
X = sm.add_constant(X)
```

```
/usr/local/lib/python3.7/dist-packages/statsmodels/tsa/tsatools.py:117: FutureWarning
x = pd.concat(x[::order], 1)
```

```
X.head()
```

	const	Height	Width	Length1	Length2	Length3
0	1.0	11.5200	4.0200	23.2	25.4	30.0
1	1.0	12.4800	4.3056	24.0	26.3	31.2
2	1.0	12.3778	4.6961	23.9	26.5	31.1
3	1.0	12.7300	4.4555	26.3	29.0	33.5
4	1.0	12.4440	5.1340	26.5	29.0	34.0

▼ Get Train Test split

```
from sklearn.model_selection import train_test_split
```

```
X_train,X_test,y_train,y_test = train_test_split(X,y,test_size = 0.3,random_state = 2529)
```

```
X_train.shape,X_test.shape,y_train.shape,y_test.shape
```

```
((111, 6), (48, 6), (111,), (48,))
```

▼ Get the model train

```
import statsmodels.api as sm
```

```
model = sm.OLS(y_train,X_train).fit()
```

▼ Get model Prediction

```
y_pred = model.predict(X_test)
```

```
y_pred
```

```
6      485.768263
54     502.247209
80      94.723820
138    876.571171
91     184.078918
48     219.301305
52     322.325322
103    376.223260
57     372.357305
149   -182.675371
153   -160.604868
108    454.335862
90     159.597558
118    843.485252
131    587.216806
100    299.535214
15     597.729508
46     197.146054
132    639.890467
79      91.200679
64     150.954248
35    -103.083206
133    627.197128
116    795.691769
31     814.687330
146   -204.149651
53     329.987469
28     715.892880
1      359.756344
117    792.324392
9      532.703671
12     552.008323
129    433.484727
111    687.617503
147   -204.763625
125    932.536683
120    810.742342
158    -80.062172
51     284.362879
34     907.080360
23     642.582834
127    959.338482
21     675.287923
113    718.863055
109    623.898492
101    376.483470
10     530.838281
157    -86.235707
dtype: float64
```

```
y_pred.shape
(48,)
```

▼ Get Model Evaluation

```
from sklearn.metrics import mean_squared_error, mean_absolute_error, mean_absolute_percentage_error

mean_squared_error(y_test, y_pred)

16397.34452441141

mean_absolute_error(y_test, y_pred)

103.02952922678567

mean_absolute_percentage_error(y_test, y_pred)

2.508285347160016

r2_score

<function sklearn.metrics._regression.r2_score>
```

▼ Get model Summary

```
print(model.summary())
```

```

                                OLS Regression Results
=====
Dep. Variable:                  Weight    R-squared:                0.896
Model:                          OLS      Adj. R-squared:           0.891
Method:                        Least Squares  F-statistic:              181.2
Date:                          Sat, 23 Apr 2022  Prob (F-statistic):    5.84e-50
Time:                          02:08:58    Log-Likelihood:          -689.20
No. Observations:              111        AIC:                     1390.
Df Residuals:                  105        BIC:                     1407.
Df Model:                      5
Covariance Type:               nonrobust
=====
               coef    std err          t      P>|t|      [0.025      0.975]
-----
const        -519.2834    34.659   -14.983    0.000   -588.005   -450.562
Height         29.8643    10.826     2.759    0.007     8.398    51.330
Width          2.2594    26.105     0.087    0.931   -49.502    54.020
Length1       58.3379    52.151     1.119    0.266   -45.068   161.743
Length2        8.5339    51.806     0.165    0.869   -94.189   111.256

```

```

Length3      -36.1521    21.444    -1.686     0.095    -78.671     6.367
=====
Omnibus:                5.384    Durbin-Watson:      2.008
Prob(Omnibus):          0.068    Jarque-Bera (JB):  4.993
Skew:                   0.391    Prob(JB):         0.0824
Kurtosis:               3.684    Cond. No.         331.
=====

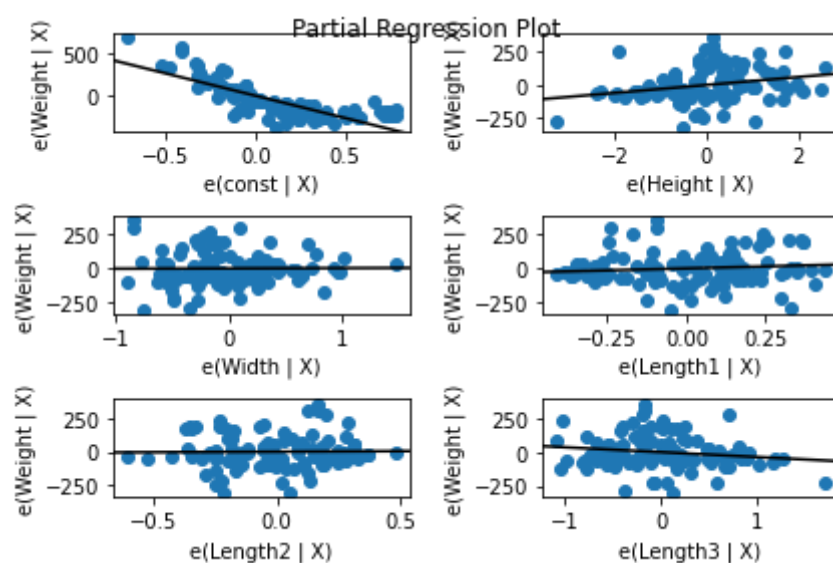
```

Warnings:

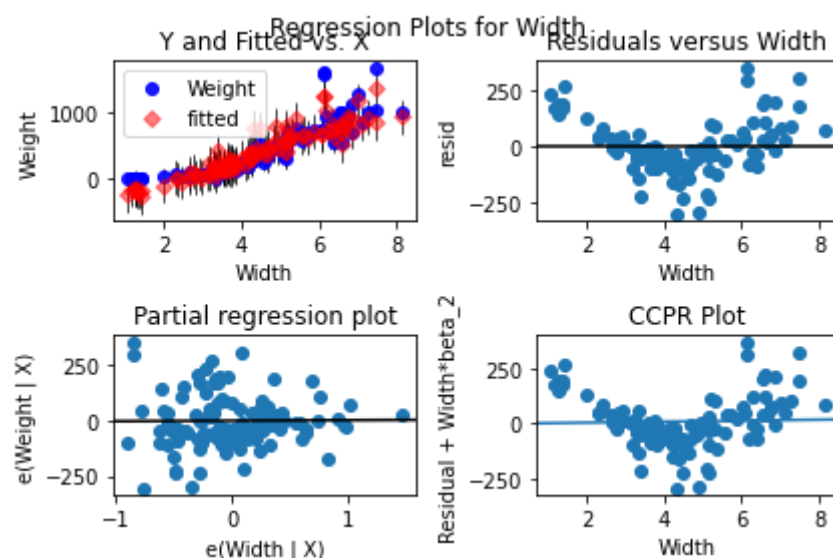
[1] Standard Errors assume that the covariance matrix of the errors is correctly spec

*Model has insignificant variables hence need remodelling

```
fig = sm.graphics.plot_partregress_grid(model)
```



```
fig = sm.graphics.plot_regress_exog(model, 'Width')
```



► Get future Predictions

```
df.new = df.sample(1)
```

```
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1: UserWarning: Pandas c
"""Entry point for launching an IPython kernel.
```

```
df.new
```

	Category	Species	Weight	Height	Width	Length1	Length2	Length3
125	3	Perch	1100.0	12.5125	7.4165	40.1	43.0	45.5

```
df.columns
```

```
Index(['Category', 'Species', 'Weight', 'Height', 'Width', 'Length1',
      'Length2', 'Length3'],
      dtype='object')
```

```
X_new = df.new[['Height', 'Width', 'Length1',
               'Length2', 'Length3']]
```

```
X_new = sm.add_constant(X_new, has_constant = 'add')
```

```
/usr/local/lib/python3.7/dist-packages/statsmodels/tsa/tsatools.py:117: FutureWarning
x = pd.concat(x[::order], 1)
```

```
X_new
```

	const	Height	Width	Length1	Length2	Length3
125	1.0	12.5125	7.4165	40.1	43.0	45.5

```
X_new.shape
```

```
(1, 6)
```

```
y_pred_new = model.predict(X_new)
```

```
y_pred_new
```

```
125    932.536683
dtype: float64
```

✓

0s

completed at 8:08 AM

×