

# Multi Linear Regresson

```
In [1]: import numpy as np  
import pandas as pd
```

```
In [2]: dataset=pd.read_csv(r"D:\ML_Course\Works_on_python\Multi_Linear_Regression\Fish.csv")
```

```
In [3]: dataset.head()
```

```
Out[3]:
```

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

```
In [4]: dataset.isnull().any()
```

```
Out[4]: Species      False  
Weight      False  
Length1     False  
Length2     False  
Length3     False  
Height      False  
Width       False  
dtype: bool
```

```
In [5]: dataset["Species"].unique()
```

```
Out[5]: array(['Bream', 'Roach', 'Whitefish', 'Parkki', 'Perch', 'Pike', 'Smelt'],  
             dtype=object)
```

```
In [6]: from sklearn.preprocessing import LabelEncoder  
le=LabelEncoder()  
dataset["Species"]=le.fit_transform(dataset["Species"])
```

```
In [7]: dataset.head()
```

```
Out[7]:
```

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

```
In [8]: x=dataset.iloc[:,[0,2,3,4,5,6]].values  
y=dataset.iloc[:,1:2].values
```

```
In [9]: x.shape
```

```
Out[9]: (159, 6)
```

```
In [10]: y.shape
```

```
Out[10]: (159, 1)
```

```
In [11]: from sklearn.preprocessing import OneHotEncoder  
one=OneHotEncoder()  
z=one.fit_transform(x[:,0:1]).toarray()  
x=np.delete(x,0,axis=1)  
x=np.concatenate((z,x),axis=1)
```

C:\Users\anikp\Anaconda3\lib\site-packages\sklearn\preprocessing\\_encoders.py:368: FutureWarning: The handling of integer data will change in version 0.22. Currently, the categories are determined based on the range [0, max(values)], while in the future they will be determined based on the unique values.

If you want the future behaviour and silence this warning, you can specify "categories='auto'".

In case you used a LabelEncoder before this OneHotEncoder to convert the categories to integers, then you can now use the OneHotEncoder directly.

```
warnings.warn(msg, FutureWarning)
```

```
In [12]: x.shape
```

```
Out[12]: (159, 12)
```

```
In [13]: y.shape
```

```
Out[13]: (159, 1)
```

```
In [14]: from sklearn.model_selection import train_test_split  
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=0)
```

```
In [15]: y_test.shape
```

```
Out[15]: (32, 1)
```

```
In [16]: from sklearn.linear_model import LinearRegression  
mlr=LinearRegression()  
mlr.fit(x_train,y_train)
```

```
Out[16]: LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None,  
normalize=False)
```

```
In [17]: y_predict=mlr.predict(x_test)
```

In [18]: x\_test

```
Out[18]: array([[ 1.    ,  0.    ,  0.    ,  0.    ,  0.    ,  0.    ,  0.    ,
 27.6   , 30.    , 35.    , 12.67  , 4.69   ],
 [ 0.    ,  0.    ,  0.    ,  0.    ,  1.    ,  0.    ,  0.    ,
 19.    , 20.5   , 22.8   , 6.4752 , 3.3516 ],
 [ 0.    ,  0.    ,  1.    ,  0.    ,  0.    ,  0.    ,  0.    ,
 21.5   , 23.5   , 25.    , 6.275  , 3.725  ],
 [ 0.    ,  0.    ,  0.    ,  0.    ,  1.    ,  0.    ,  0.    ,
 20.5   , 22.5   , 25.3   , 7.0334 , 3.8203 ],
 [ 0.    ,  0.    ,  1.    ,  0.    ,  0.    ,  0.    ,  0.    ,
 32.    , 34.5   , 36.5   , 10.2565, 6.3875 ],
 [ 0.    ,  0.    ,  1.    ,  0.    ,  0.    ,  0.    ,  0.    ,
 37.    , 40.    , 42.5   , 11.73   , 7.225  ],
 [ 0.    ,  0.    ,  0.    ,  0.    ,  0.    ,  0.    ,  1.    ,
 33.7   , 36.4   , 39.6   , 11.7612, 6.5736 ],
 [ 0.    ,  0.    ,  0.    ,  1.    ,  0.    ,  0.    ,  0.    ,
 34.8   , 37.3   , 39.8   , 6.2884 , 4.0198 ],
 [ 1.    ,  0.    ,  0.    ,  0.    ,  0.    ,  0.    ,  0.    ,
 37.4   , 41.    , 45.9   , 18.6354, 6.7473 ],
 [ 0.    ,  0.    ,  1.    ,  0.    ,  0.    ,  0.    ,  0.    ,
 19.    , 21.    , 22.5   , 5.9175 , 3.3075 ],
 [ 0.    ,  0.    ,  0.    ,  1.    ,  0.    ,  0.    ,  0.    ,
 30.    , 32.3   , 34.8   , 5.568  , 3.3756 ],
 [ 0.    ,  0.    ,  0.    ,  1.    ,  0.    ,  0.    ,  0.    ,
 40.    , 42.5   , 45.5   , 7.28   , 4.3225 ],
 [ 0.    ,  0.    ,  0.    ,  0.    ,  0.    ,  0.    ,  1.    ,
 37.3   , 40.    , 43.5   , 12.354  , 6.525  ],
 [ 0.    ,  0.    ,  1.    ,  0.    ,  0.    ,  0.    ,  0.    ,
 40.2   , 43.5   , 46.    , 12.604  , 8.142  ],
 [ 0.    ,  1.    ,  0.    ,  0.    ,  0.    ,  0.    ,  0.    ,
 14.3   , 15.5   , 17.4   , 6.5772 , 2.3142 ],
 [ 0.    ,  0.    ,  0.    ,  0.    ,  1.    ,  0.    ,  0.    ,
 17.5   , 18.8   , 21.2   , 5.5756 , 2.9044 ],
 [ 0.    ,  0.    ,  0.    ,  0.    ,  1.    ,  0.    ,  0.    ,
 20.5   , 22.    , 24.3   , 6.6339 , 3.5478 ],
 [ 0.    ,  0.    ,  0.    ,  1.    ,  0.    ,  0.    ,  0.    ,
 56.    , 60.    , 64.    , 9.6     , 6.144  ],
 [ 0.    ,  0.    ,  1.    ,  0.    ,  0.    ,  0.    ,  0.    ,
 19.3   , 21.3   , 22.8   , 6.384  , 3.534  ],
 [ 1.    ,  0.    ,  0.    ,  0.    ,  0.    ,  0.    ,  0.    ,
 32.    , 35.    , 40.6   , 16.3618, 6.09   ],
 [ 0.    ,  1.    ,  0.    ,  0.    ,  0.    ,  0.    ,  0.    ,
```

```

13.5 , 14.7 , 16.5 , 6.8475, 2.3265],
[ 0. , 0. , 0. , 0. , 1. , 0. , 0. ,
29.5 , 31.7 , 35. , 9.485 , 5.355 ],
[ 0. , 0. , 1. , 0. , 0. , 0. , 0. ,
20. , 22. , 23.5 , 6.11 , 3.4075],
[ 0. , 0. , 0. , 1. , 0. , 0. , 0. ,
59. , 63.4 , 68. , 10.812 , 7.48 ],
[ 0. , 1. , 0. , 0. , 0. , 0. , 0. ,
16.3 , 17.7 , 19.8 , 7.4052, 2.673 ],
[ 1. , 0. , 0. , 0. , 0. , 0. , 0. ,
27.6 , 30. , 35.1 , 14.0049, 4.8438],
[ 1. , 0. , 0. , 0. , 0. , 0. , 0. ,
31.9 , 35. , 40.5 , 16.2405, 5.589 ],
[ 0. , 0. , 0. , 0. , 0. , 0. , 1. ,
24.1 , 26.5 , 29.3 , 8.1454, 4.2485],
[ 0. , 0. , 1. , 0. , 0. , 0. , 0. ,
36.9 , 40. , 42.3 , 11.9286, 7.1064],
[ 0. , 0. , 0. , 0. , 0. , 1. , 0. ,
10.4 , 11. , 12. , 2.196 , 1.38 ],
[ 1. , 0. , 0. , 0. , 0. , 0. , 0. ,
31. , 33.5 , 38.7 , 14.4738, 5.7276],
[ 0. , 0. , 1. , 0. , 0. , 0. , 0. ,
20. , 22. , 23.5 , 5.5225, 3.995 ]])

```

```
In [19]: y_test
```

```
Out[19]: array([[ 390. ],
 [    0. ],
 [ 170. ],
 [ 160. ],
 [ 556. ],
 [ 900. ],
 [ 800. ],
 [ 300. ],
 [ 975. ],
 [ 115. ],
 [ 200. ],
 [ 456. ],
 [1000. ],
 [1000. ],
 [  60. ],
 [  78. ],
 [ 145. ],
 [1600. ],
 [ 130. ],
 [ 720. ],
 [  55. ],
 [ 390. ],
 [ 120. ],
 [1650. ],
 [  90. ],
 [ 450. ],
 [ 700. ],
 [ 270. ],
 [ 850. ],
 [   9.7],
 [ 650. ],
 [ 110. ]])
```

```
In [20]: y_predict
```

```
Out[20]: array([[ 428.88533577],
 [  98.08363614],
 [ 216.67998922],
 [ 208.66936638],
 [ 657.24094116],
 [ 876.38855413],
 [ 665.97861965],
 [ 407.27203048],
 [ 965.65306863],
 [ 146.62291102],
 [ 255.15532231],
 [ 561.63685124],
 [ 765.67575361],
 [1012.38234027],
 [-118.72798063],
 [  14.47341216],
 [ 137.60789564],
 [1155.53572308],
 [ 170.97092949],
 [ 724.93548455],
 [-128.48675188],
 [ 525.45508599],
 [ 175.3519065 ],
 [1322.74816983],
 [ -23.0873263 ],
 [ 475.44172778],
 [ 719.96841977],
 [ 280.94571114],
 [ 885.12085107],
 [ -16.63412226],
 [ 585.26038657],
 [ 164.22863371]])
```

```
In [21]: from sklearn.metrics import r2_score
accuracy=r2_score(y_test,y_predict)
```

```
In [22]: accuracy
```

```
Out[22]: 0.9102350316202583
```

```
In [23]: y=mlr.predict([[1,0,0,0,0,0,0,28,28,30,15,6]])
```

```
In [24]: y
```

```
Out[24]: array([[257.56434014]])
```

## Decision Tree

```
In [25]: from sklearn.tree import DecisionTreeRegressor  
dtr=DecisionTreeRegressor(random_state=0)  
dtr.fit(x_train,y_train)
```

```
Out[25]: DecisionTreeRegressor(criterion='mse', max_depth=None, max_features=None,  
                                max_leaf_nodes=None, min_impurity_decrease=0.0,  
                                min_impurity_split=None, min_samples_leaf=1,  
                                min_samples_split=2, min_weight_fraction_leaf=0.0,  
                                presort=False, random_state=0, splitter='best')
```

```
In [26]: y_dtr=dtr.predict(x_test)
```

```
In [27]: y_dtr
```

```
Out[27]: array([ 430. ,  110. ,  145. ,  145. ,  514. , 1015. ,  685. ,  345. ,  
                 950. ,  110. ,  345. ,  500. ,  820. , 1100. ,   51.5,   87. ,  
                 150. , 1550. ,  110. ,  725. ,   51.5,  500. ,  130. , 1550. ,  
                 69. ,  450. ,  725. ,  300. , 1015. ,    8.7,  700. ,  120. ])
```



```
In [28]: y_test
```

```
Out[28]: array([[ 390. ],
 [    0. ],
 [ 170. ],
 [ 160. ],
 [ 556. ],
 [ 900. ],
 [ 800. ],
 [ 300. ],
 [ 975. ],
 [ 115. ],
 [ 200. ],
 [ 456. ],
 [1000. ],
 [1000. ],
 [  60. ],
 [  78. ],
 [ 145. ],
 [1600. ],
 [ 130. ],
 [ 720. ],
 [  55. ],
 [ 390. ],
 [ 120. ],
 [1650. ],
 [  90. ],
 [ 450. ],
 [ 700. ],
 [ 270. ],
 [ 850. ],
 [   9.7],
 [ 650. ],
 [ 110. ]])
```

```
In [29]: accuracy_dtr=r2_score(y_test,y_dtr)
```

```
In [30]: accuracy_dtr
```

```
Out[30]: 0.9724345073124595
```

# Random Forest

```
In [55]: from sklearn.ensemble import RandomForestRegressor  
rfr=RandomForestRegressor(n_estimators=16,random_state=0)  
rfr.fit(x_train,y_train)
```

C:\Users\anikp\Anaconda3\lib\site-packages\ipykernel\_launcher.py:3: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n\_samples,), for example using ravel().  
This is separate from the ipykernel package so we can avoid doing imports until

```
Out[55]: RandomForestRegressor(bootstrap=True, criterion='mse', max_depth=None,  
                                max_features='auto', max_leaf_nodes=None,  
                                min_impurity_decrease=0.0, min_impurity_split=None,  
                                min_samples_leaf=1, min_samples_split=2,  
                                min_weight_fraction_leaf=0.0, n_estimators=16, n_jobs=None,  
                                oob_score=False, random_state=0, verbose=0, warm_start=False)
```

```
In [56]: y_rfr=rfr.predict(x_test)
```

```
In [57]: accuracy_rfr=r2_score(y_test,y_rfr)
```

```
In [58]: accuracy_rfr
```

```
Out[58]: 0.9753199702551268
```

```
In [ ]:
```