IBM ASSIGNMENT 4

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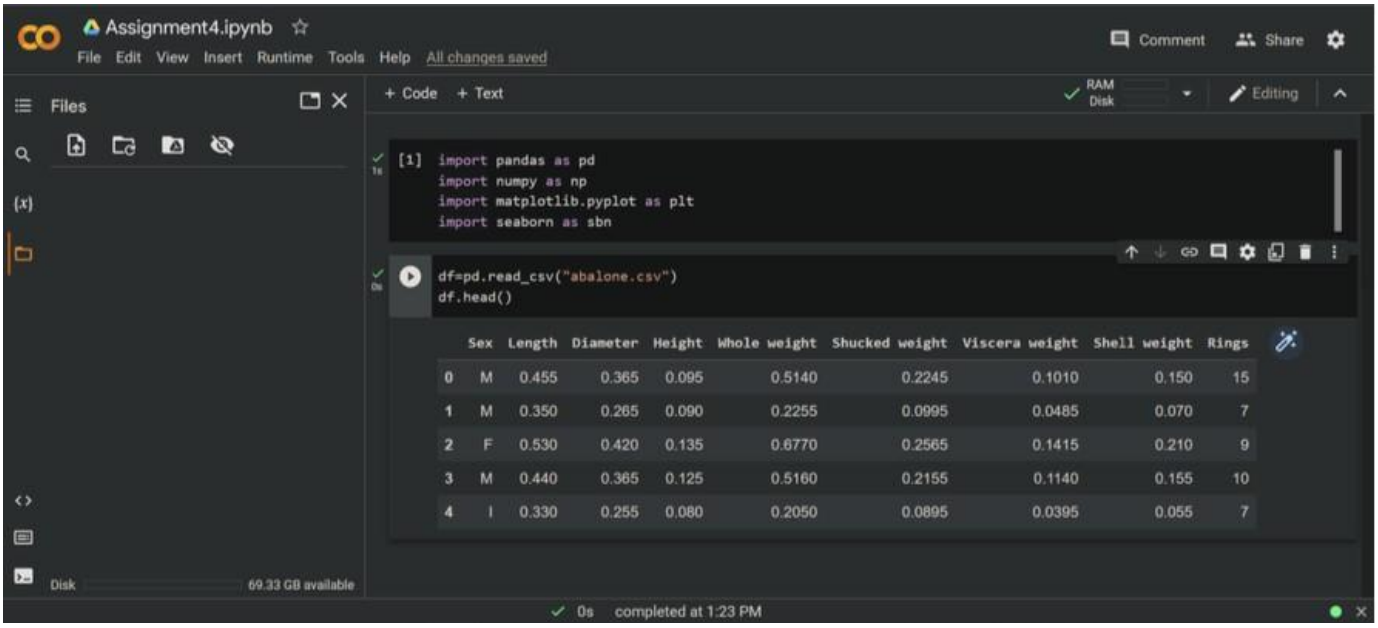
CHALLENGE:

Abalone Age Prediction

LOADING THE DATASET:



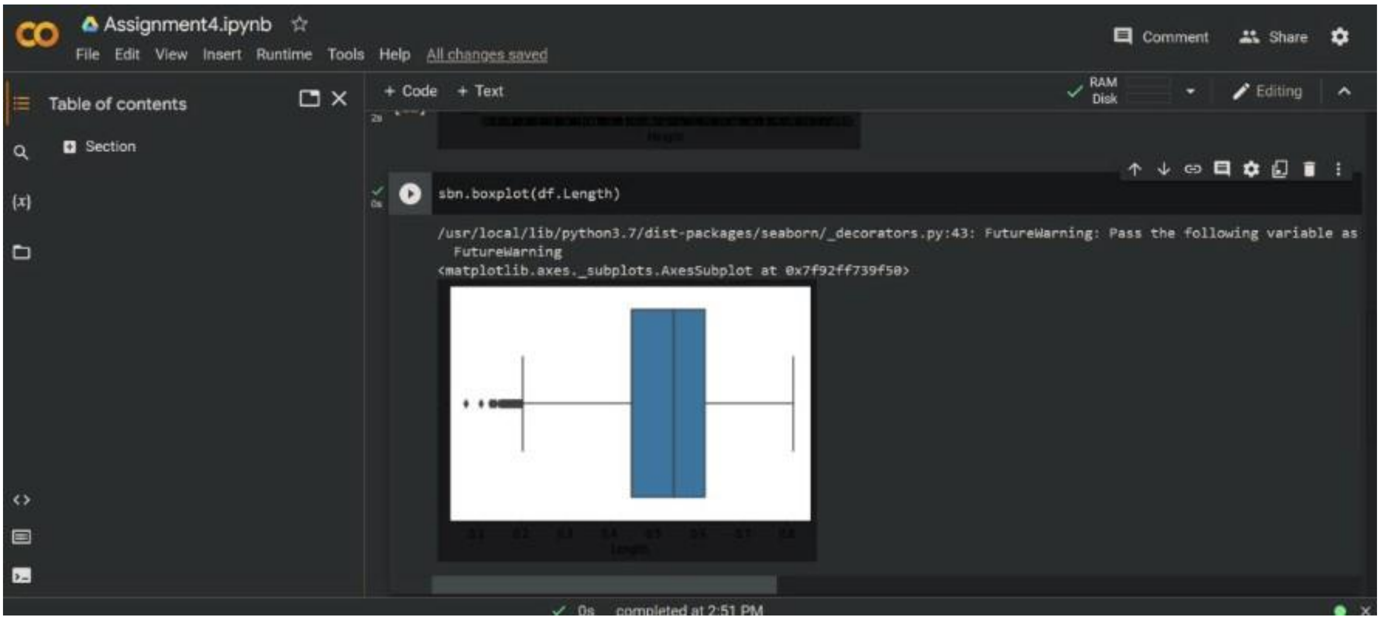

```
[1] import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
    import seaborn as sbn
```

```
df=pd.read_csv("abalone.csv")
df.head()
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	M	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	15
1	M	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	7
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	9
3	M	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	10
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	7

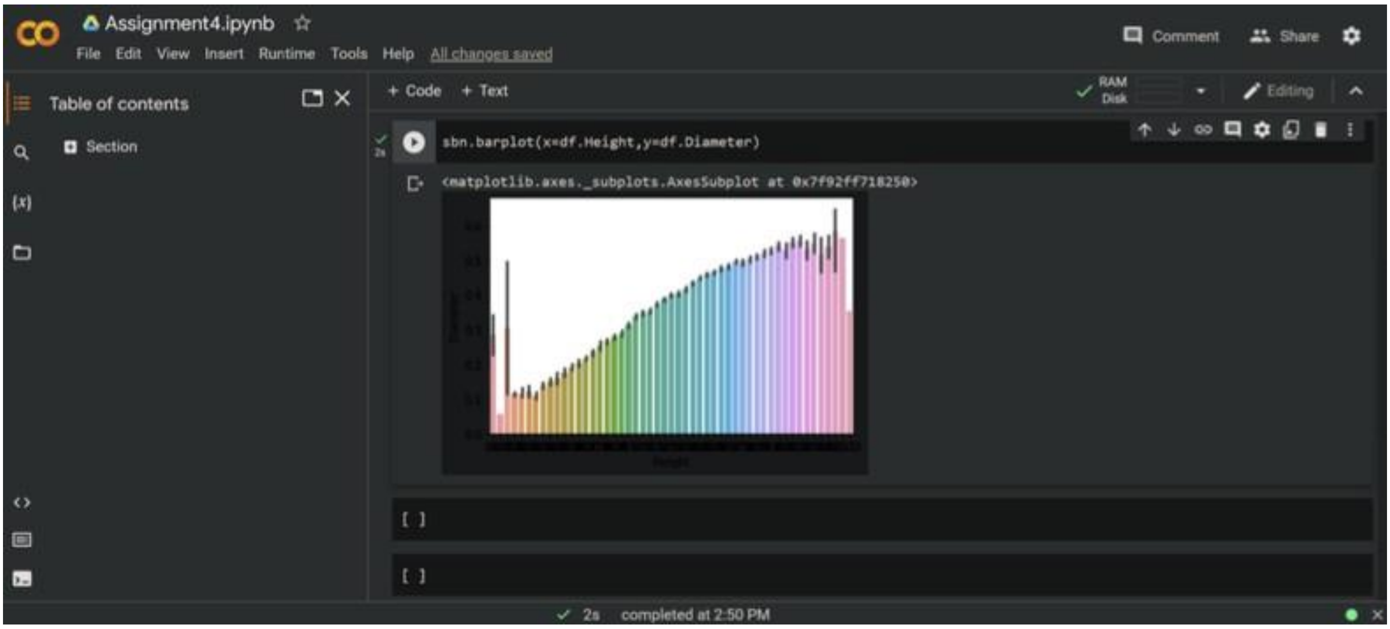
VISUALIZATIONS:

Univariate Analysis
Bi-Variate Analysis

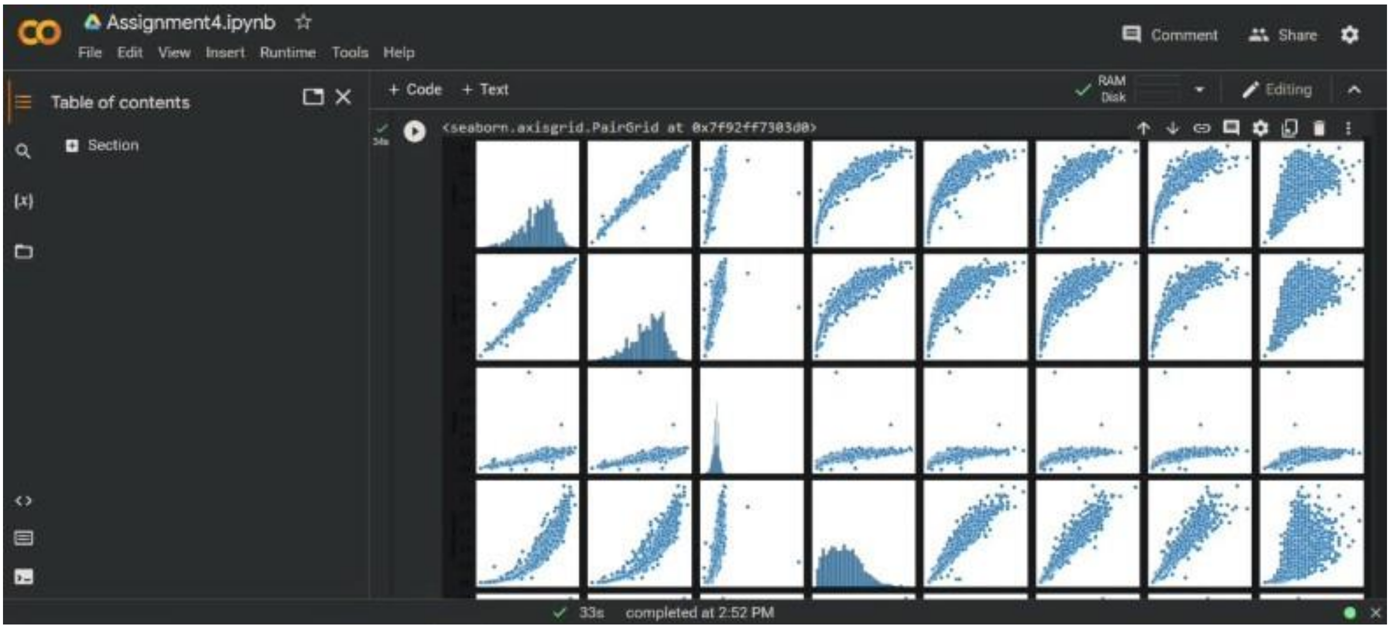



```
sbn.boxplot(df.Length)
```

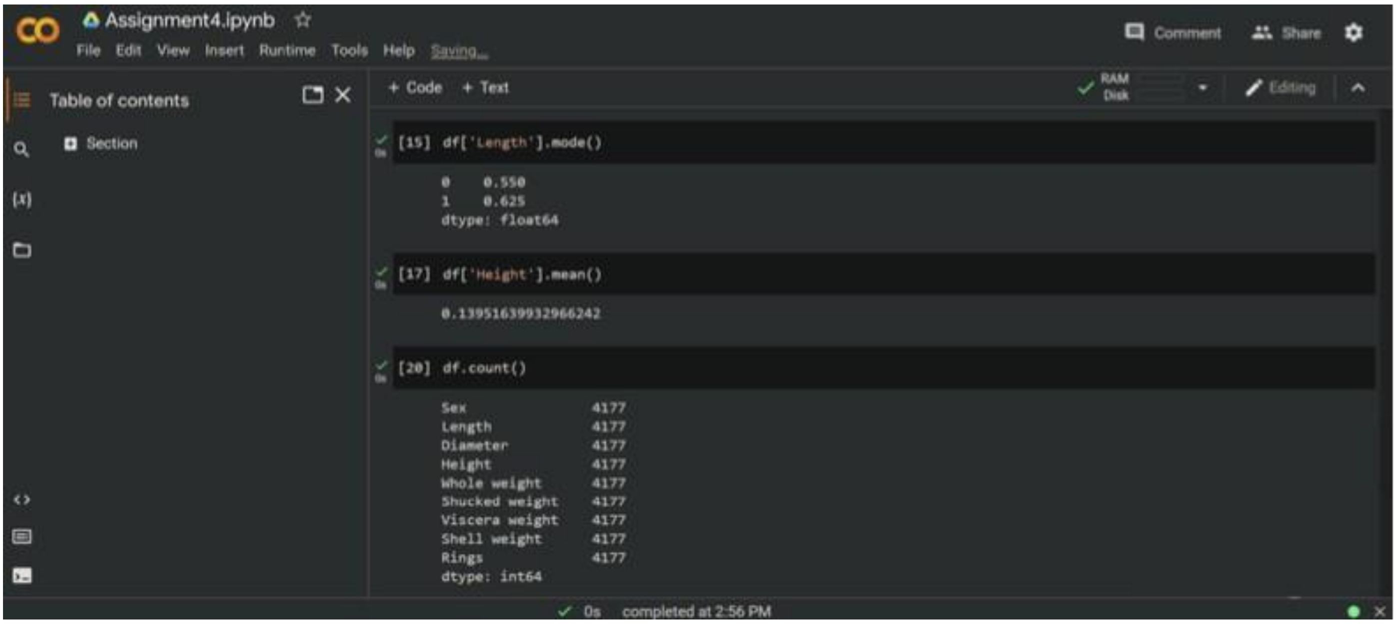
```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as
  FutureWarning
<matplotlib.axes._subplots.AxesSubplot at 0x7f92ff739f50>
```

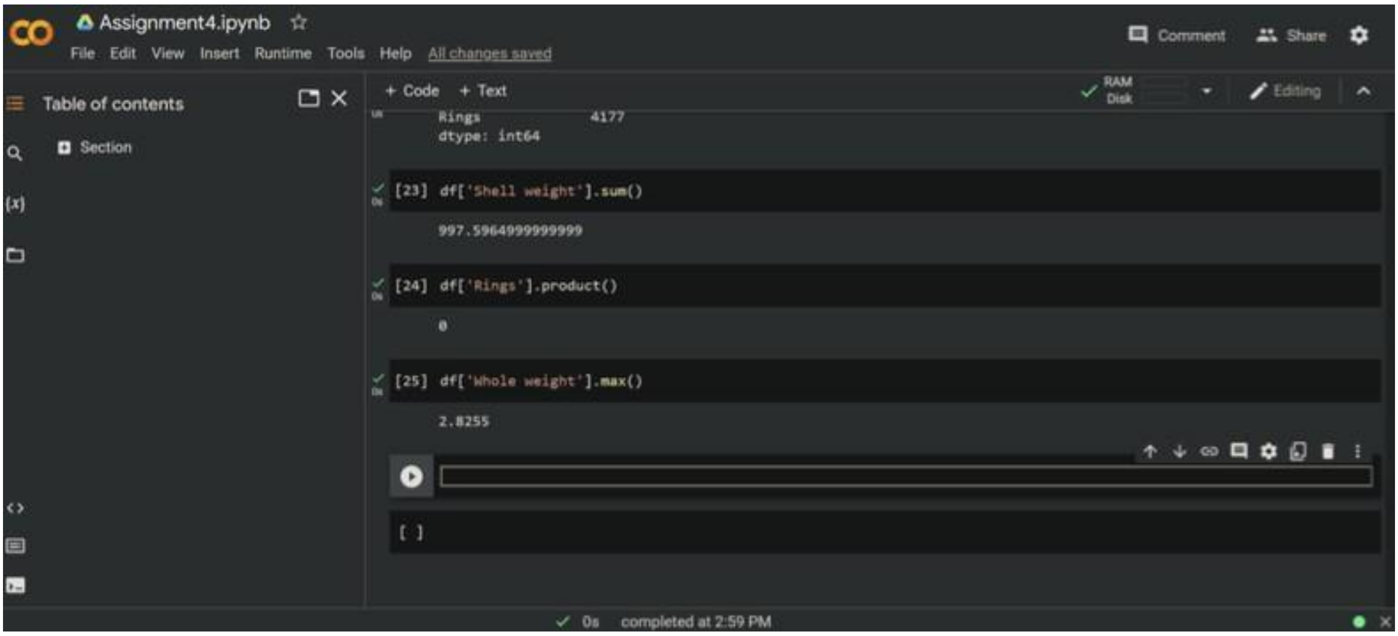


Multi-Variate Analysis

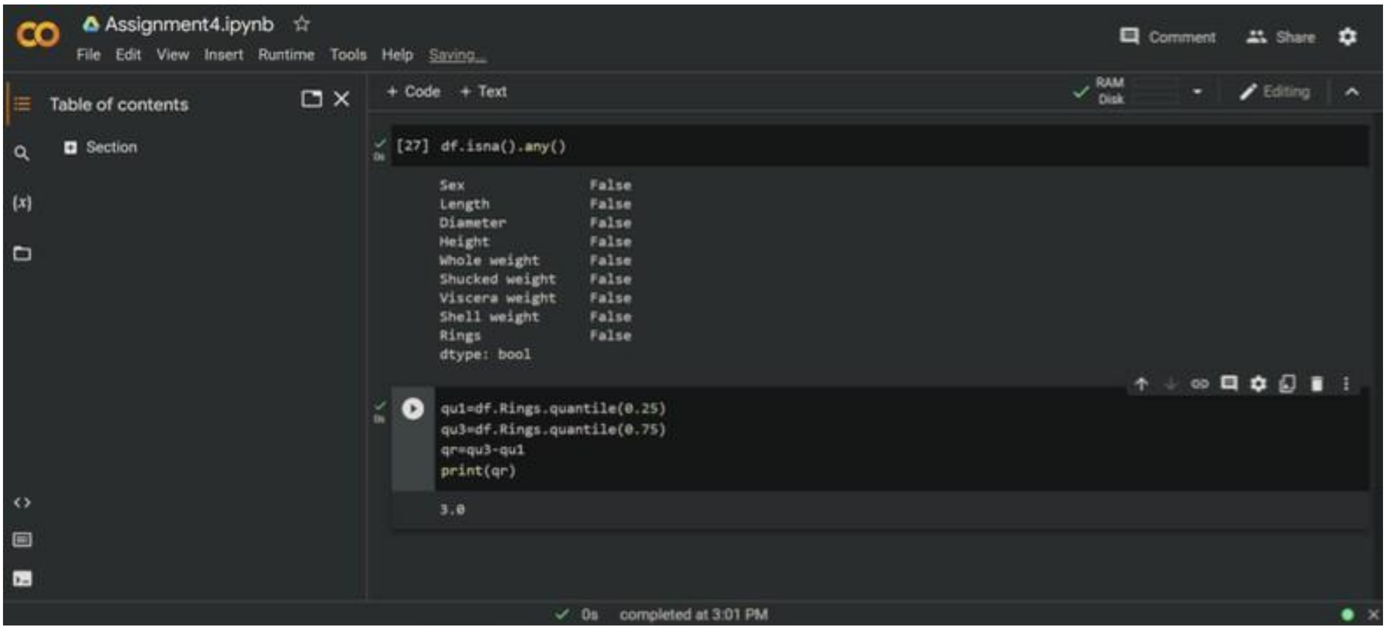


Perform descriptive Analysis on datasets





Checking for missing values and deal with them , Finding the outliers and replace them outliers



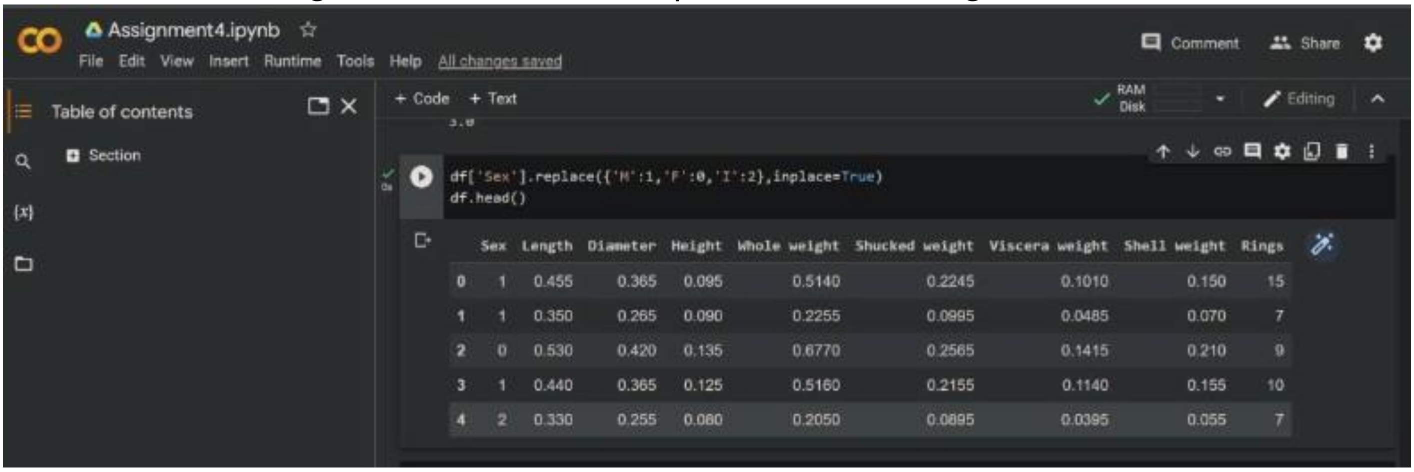
```
[27] df.isna().any()
```

```
Sex               False
Length            False
Diameter          False
Height            False
Whole weight      False
Shucked weight    False
Viscera weight    False
Shell weight      False
Rings             False
dtype: bool
```

```
qu1=df.Rings.quantile(0.25)
qu3=df.Rings.quantile(0.75)
qr=qu3-qu1
print(qr)
```

```
3.0
```

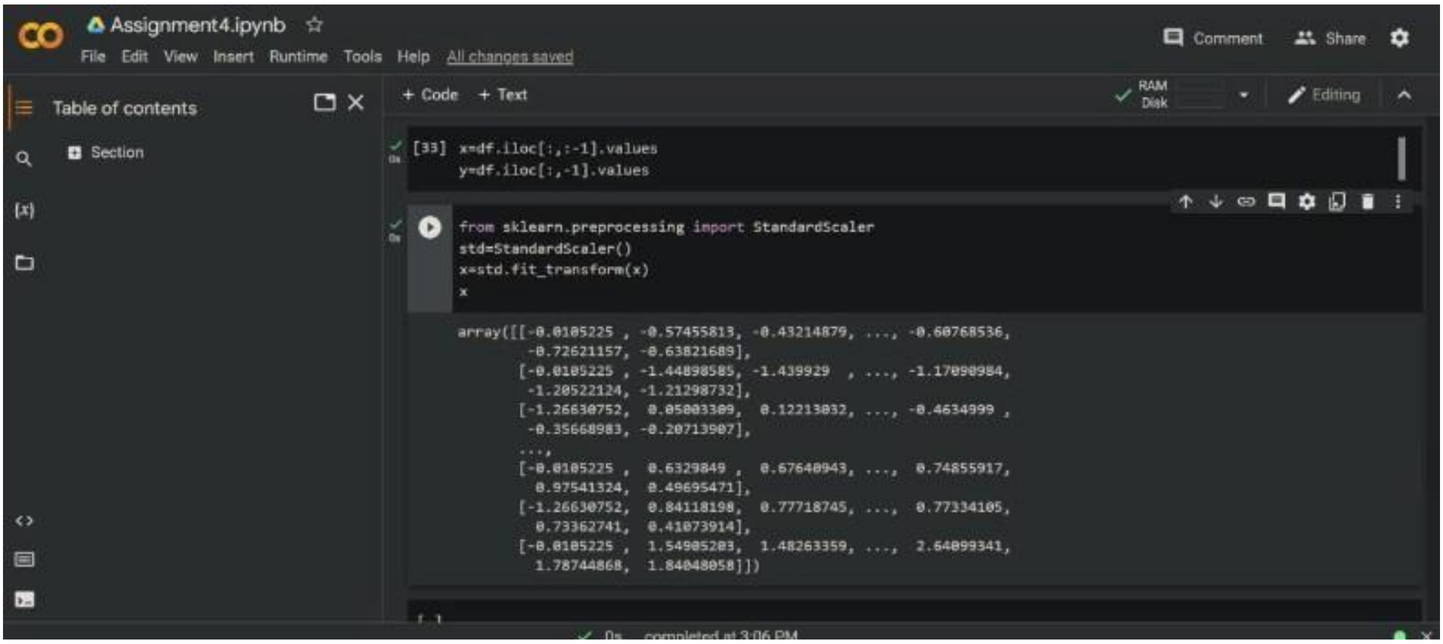
Check for categorical columns and perform encoding



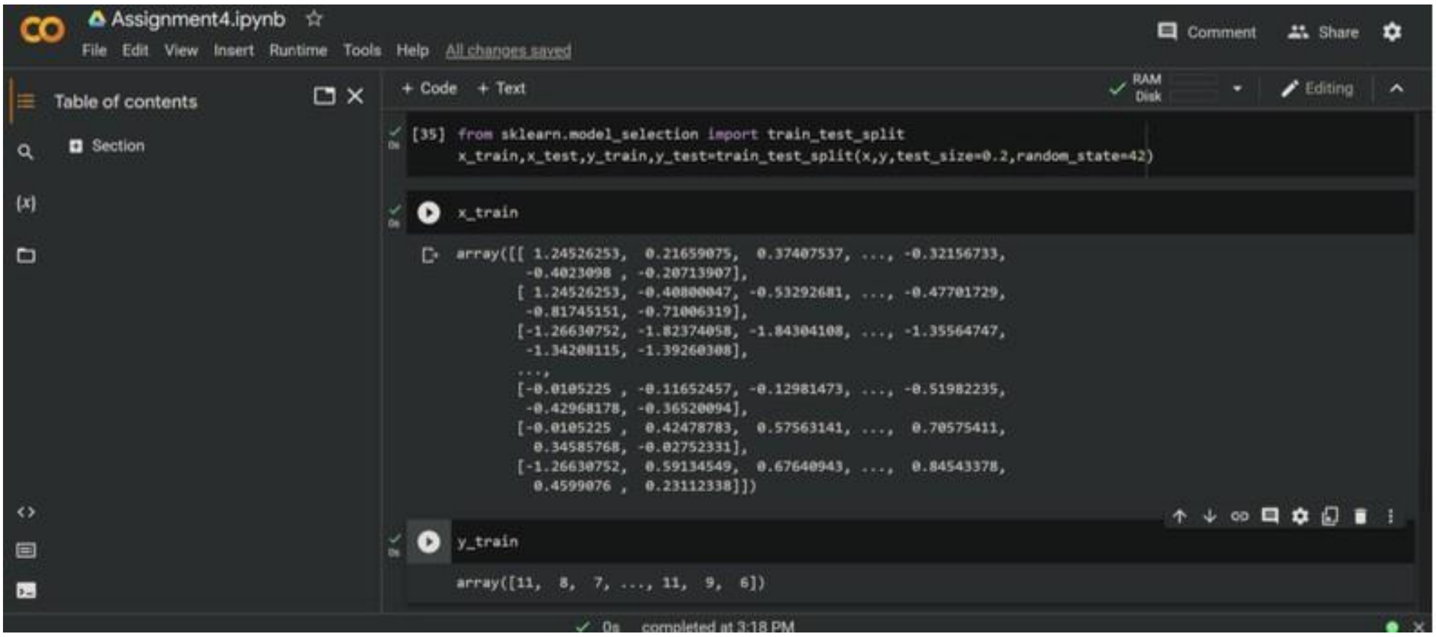
```
df['Sex'].replace({'M':1,'F':0,'I':2},inplace=True)
df.head()
```

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	1	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	15
1	1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	7
2	0	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	9
3	1	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	10
4	2	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	7

Split the data into dependent and independent variables, Scale the independent variable



Split the data into training and testing

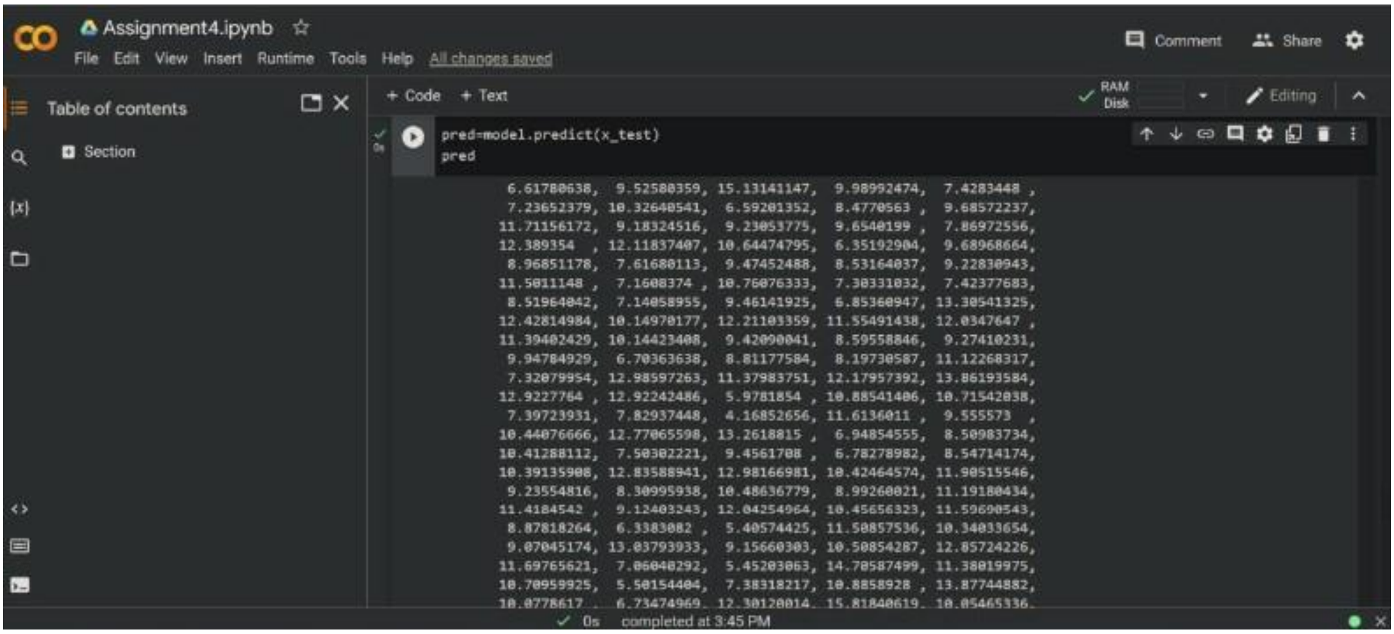


```
[38] x_test

array([[-0.0105225 ,  0.67462432,  0.47485339, ...,  0.27770351,
         1.10314916,  0.61909342],
       [-0.0105225 ,  0.54970607,  0.32368636, ...,  0.12450645,
         0.3139237 ,  0.04432299],
       [-1.26630752,  0.29986958,  0.37407537, ..., -0.2449688 ,
         0.40060164,  0.69093973],
       ...,
       [ 1.24526253,  0.17495134,  0.22290834, ..., -0.0309435 ,
        -0.20614393, -0.22150833],
       [ 1.24526253, -0.4912793 , -0.53292681, ..., -0.47025859,
        -0.81288951, -0.39393946],
       [ 1.24526253, -1.3240676 , -1.33915098, ..., -1.17766853,
        -1.30558517, -1.17706417]])
```

```
y_test

array([ 9,  8, 16,  9, 14, 11,  7,  6,  7, 10, 22,  7, 15,  9,  8, 18, 11,
       14, 13,  9, 20, 12, 12, 11, 10,  7, 11,  8,  9, 10,  9, 10,  6, 10,
        8,  9,  5,  3,  6,  6, 12, 12, 18,  8, 12, 13, 10, 10, 18,  4,  6,
       22,  8,  5,  7, 10, 15, 21, 10,  9, 10, 13, 11,  7,  9, 11,  4,  5,
        7,  9, 10, 11, 10,  7,  9, 12, 23, 14, 15,  9, 15, 13, 10,  6,  7,
       13,  9, 10, 19, 10, 10,  9, 11, 11, 10, 10,  6, 15,  7,  7, 15, 11,
       11, 13,  7,  9, 10,  8,  9, 14, 18,  8, 13,  9, 12,  5,  9, 12, 11,
       13, 11, 10,  8, 14,  9, 20,  9,  9,  9, 10,  9,  9, 10,  5,  8,  8,
       10, 10,  5, 12,  8, 11,  7,  8, 10, 15, 10, 14, 10, 10, 10,  8, 11,
       11,  8, 11, 12,  7,  8,  6,  9,  6, 10, 12,  7, 10, 17, 11,  8,  8,
       10, 12,  9,  8,  8,  7,  9, 11,  9, 10, 13,  7,  8,  8,  7, 10,  8,
       11,  9,  5,  9,  8, 16, 13, 11, 17, 10, 11, 12,  9,  8, 17, 11, 12,
        9, 12, 11,  9,  8, 10,  5,  9, 12,  6,  8, 11, 11,  7,  9, 12, 13,
        9, 12, 11,  9,  8,  7, 13,  9, 12,  5, 10, 10, 12,  7, 10, 10,  7,
        4, 10,  8, 11, 10,  9, 10,  8,  9,  7,  7,  6,  7,  9,  9,  7, 15,
       11,  9,  5, 12, 14, 19, 16,  9,  9,  7,  6,  7, 14, 12,  6,  9,  8,
        6, 12,  8, 18, 10, 16,  9,  6, 15,  9, 13,  8,  5,  9, 10,  5, 10,
       10, 11,  4, 15,  9, 15,  8,  5, 14,  7, 11, 10, 10,  7, 10,  9, 10,
       18,  8,  6,  5,  8,  6,  7, 14, 12, 10,  5, 23,  9,  9, 12,  7,  8,
        8, 13,  6, 13, 17,  7,  8,  8,  7,  7,  9, 14, 10,  9, 13,  8, 10,
       10,  9, 10,  9,  8,  8, 10, 13, 10,  9,  8, 10,  8, 11, 10,  3,  7,
        6,  3,  8,  8, 13, 15,  6, 14,  8,  9, 12,  8,  8, 15, 11,  9,  6,
       10, 13, 13,  7,  7,  9,  9,  8,  7, 10, 11,  5, 10, 12,  8,  7,  9,
        6,  8, 13,  7,  7, 10, 11, 23,  9, 11, 10,  8,  8,  7,  7, 10,  9,
```

Build the model, Train the Model Test the Model

```
[40] from sklearn.ensemble import RandomForestRegressor
     model = RandomForestRegressor(n_estimators = 1000, oob_score = True,n_jobs=-
     1,min_samples_split = 6, min_samples_leaf= 4, max_features = 'sqrt', max_depth= 120,
     bootstrap=True)

[41] model.fit(x_train,y_train)

RandomForestRegressor(max_depth=120, max_features='sqrt', min_samples_leaf=4,
                      min_samples_split=6, n_estimators=1000, n_jobs=-1,
                      oob_score=True)
```



Measure the performance using metrics