In [*]:
 import pandas as pd
 import numpy as np
 import matplotlib.pyplot as plt
 %matplotlib inline

Load the dataset

In [*]: df=pd.read_csv('abalone.csv')

In [3]: df

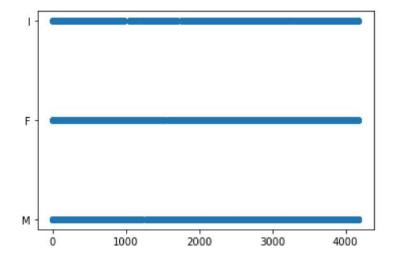
Out[3]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	М	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.1500	15
1	М	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700	7
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100	9
3	М	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550	10
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550	7
		•••		•••					
4172	F	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	11
4173	М	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	10
4174	М	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080	9
4175	F	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960	10
4176	М	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12

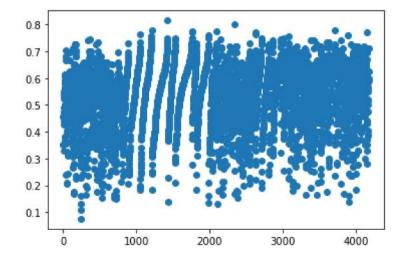
4177 rows × 9 columns

Univariate analysis

```
In [4]: import seaborn as sns
plt.scatter(df.index,df['Sex'])
plt.show()
```

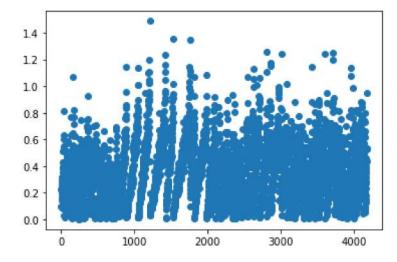


In [5]: plt.scatter(df.index,df['Length'])
plt.show()



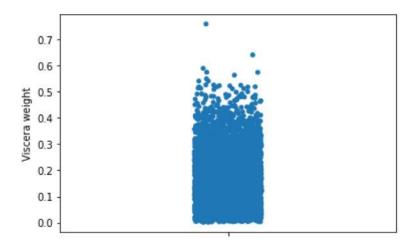
```
In [6]: sns.stripplot(y=df['Diameter'])
Out[6]: <AxesSubplot:ylabel='Diameter'>
             0.6
             0.5
          Diameter
0.0
0.3
             0.2
             0.1
In [7]: sns.stripplot(y=df['Height'])
Out[7]: <AxesSubplot:ylabel='Height'>
             1.0
             0.8
          Height
9.0
             0.4
             0.2
             0.0
In [8]: sns.stripplot(y=df['Whole weight'])
Out[8]: <AxesSubplot:ylabel='Whole weight'>
             2.5
             2.0
          Whole weight 1.0
             0.5
             0.0
```

```
In [9]: plt.scatter(df.index,df['Shucked weight'])
plt.show()
```



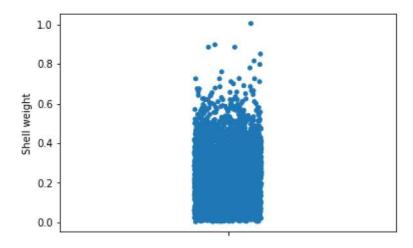
```
In [10]: sns.stripplot(y=df['Viscera weight'])
```

Out[10]: <AxesSubplot:ylabel='Viscera weight'>

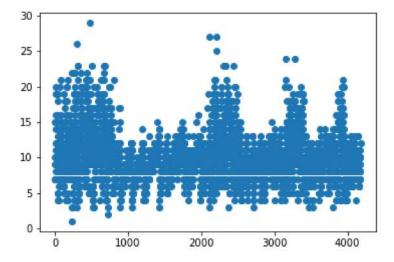


```
In [11]: sns.stripplot(y=df['Shell weight'])
```

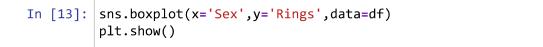
Out[11]: <AxesSubplot:ylabel='Shell weight'>

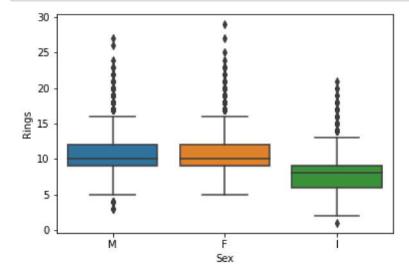


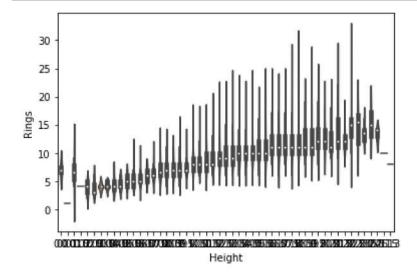
```
In [12]: plt.scatter(df.index,df['Rings'])
    plt.show()
```



Bi-variate analysis

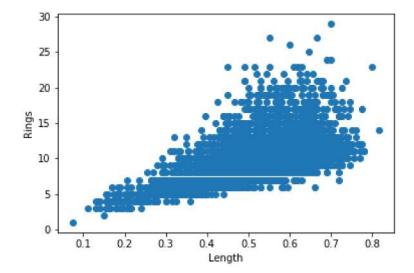






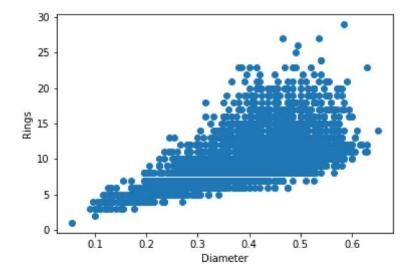
```
In [15]: plt.plot(df["Length"],df["Rings"],'o')
    plt.ylabel("Rings")
    plt.xlabel("Length")
```

Out[15]: Text(0.5, 0, 'Length')



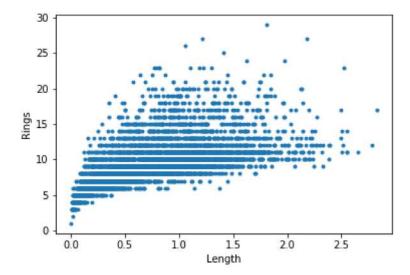
```
In [16]: plt.plot(df["Diameter"],df["Rings"],'o')
    plt.ylabel("Rings")
    plt.xlabel("Diameter")
```

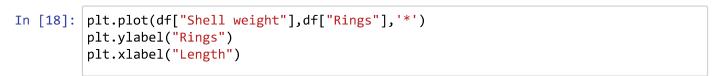
Out[16]: Text(0.5, 0, 'Diameter')



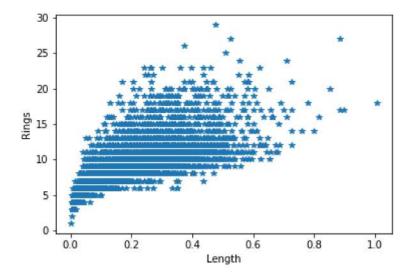
```
In [17]: plt.plot(df["Whole weight"],df["Rings"],'.')
    plt.ylabel("Rings")
    plt.xlabel("Length")
```

Out[17]: Text(0.5, 0, 'Length')



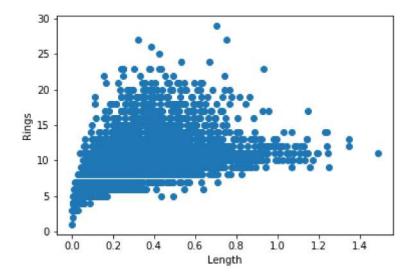


Out[18]: Text(0.5, 0, 'Length')



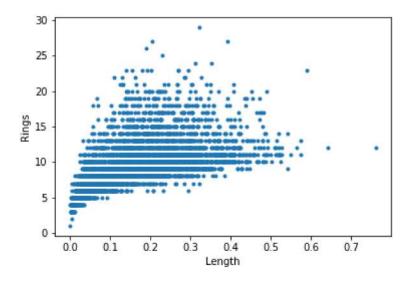
```
In [19]: plt.plot(df["Shucked weight"],df["Rings"],'o')
    plt.ylabel("Rings")
    plt.xlabel("Length")
```

```
Out[19]: Text(0.5, 0, 'Length')
```



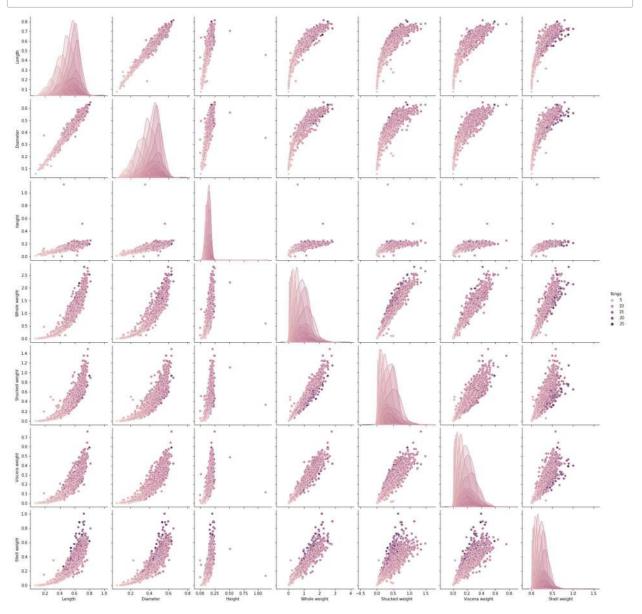
```
In [20]: plt.plot(df["Viscera weight"],df["Rings"],'.')
    plt.ylabel("Rings")
    plt.xlabel("Length")
```

Out[20]: Text(0.5, 0, 'Length')



Multivariate Analysis

In [21]: sns.pairplot(df,hue="Rings",height=3)
plt.show()



Descriptive analysis

In [22]: df.describe().T

Out[22]:

	count	mean	std	min	25%	50%	75%	max
Length	4177.0	0.523992	0.120093	0.0750	0.4500	0.5450	0.615	0.8150
Diameter	4177.0	0.407881	0.099240	0.0550	0.3500	0.4250	0.480	0.6500
Height	4177.0	0.139516	0.041827	0.0000	0.1150	0.1400	0.165	1.1300
Whole weight	4177.0	0.828742	0.490389	0.0020	0.4415	0.7995	1.153	2.8255
Shucked weight	4177.0	0.359367	0.221963	0.0010	0.1860	0.3360	0.502	1.4880
Viscera weight	4177.0	0.180594	0.109614	0.0005	0.0935	0.1710	0.253	0.7600
Shell weight	4177.0	0.238831	0.139203	0.0015	0.1300	0.2340	0.329	1.0050
Rings	4177.0	9.933684	3.224169	1.0000	8.0000	9.0000	11.000	29.0000

Handling with missing data

In [23]: df = pd.DataFrame(df)
 df.isnull()

Out[23]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False
4172	False	False	False	False	False	False	False	False	False
4173	False	False	False	False	False	False	False	False	False
4174	False	False	False	False	False	False	False	False	False
4175	False	False	False	False	False	False	False	False	False
4176	False	False	False	False	False	False	False	False	False

4177 rows × 9 columns

In [24]: df.fillna(0)

Out[24]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	М	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.1500	15
1	М	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700	7
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100	9
3	М	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550	10
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550	7
4172	F	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	11
4173	М	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	10
4174	М	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080	9
4175	F	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960	10
4176	М	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12

4177 rows × 9 columns

```
In [25]: df.isnull().sum()
```

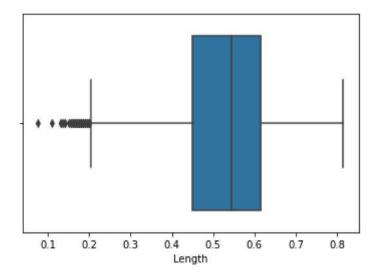
Out[25]: Sex

0 0 Length Diameter 0 Height 0 Whole weight 0 Shucked weight 0 0 Viscera weight Shell weight 0 Rings 0 dtype: int64

Outliers in each attribute

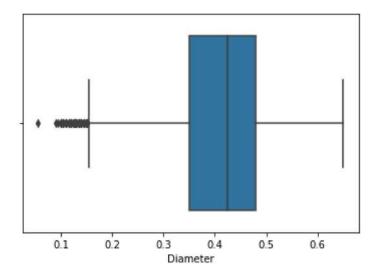
In [26]: sns.boxplot(df['Length'],data=df)

Out[26]: <AxesSubplot:xlabel='Length'>



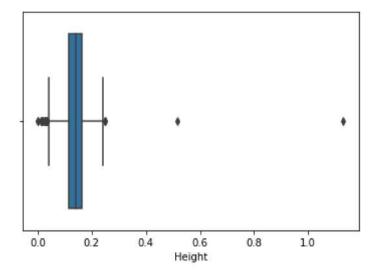
In [27]: sns.boxplot(df['Diameter'],data=df)

Out[27]: <AxesSubplot:xlabel='Diameter'>



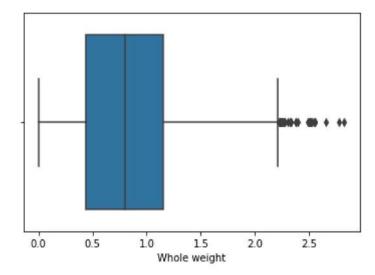
In [28]: sns.boxplot(df['Height'],data=df)

Out[28]: <AxesSubplot:xlabel='Height'>



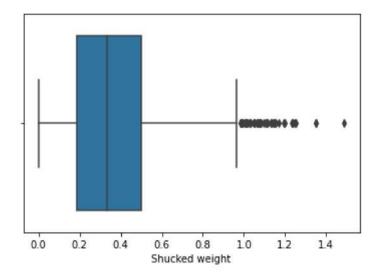
In [29]: sns.boxplot(df['Whole weight'],data=df)

Out[29]: <AxesSubplot:xlabel='Whole weight'>



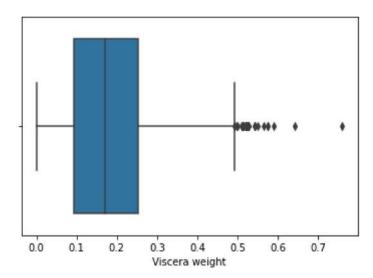
In [30]: sns.boxplot(df['Shucked weight'],data=df)

Out[30]: <AxesSubplot:xlabel='Shucked weight'>



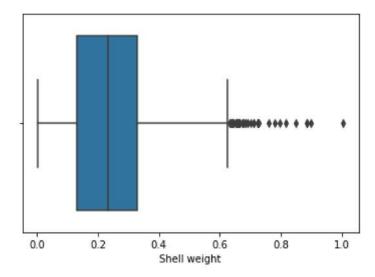
In [31]: sns.boxplot(df['Viscera weight'],data=df)

Out[31]: <AxesSubplot:xlabel='Viscera weight'>



In [32]: sns.boxplot(df['Shell weight'],data=df)

Out[32]: <AxesSubplot:xlabel='Shell weight'>

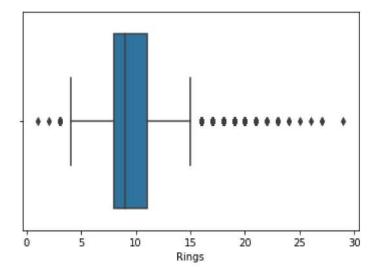


```
In [33]: sns.boxplot(df['Rings'],data=df)
```

C:\Users\ABI\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(

Out[33]: <AxesSubplot:xlabel='Rings'>



```
In [38]: Q1 = df.quantile(0.25)
         Q3 = df.quantile(0.75)
         IQR = Q3-Q1
         print(IQR)
         Length
                            0.1650
         Diameter
                            0.1300
         Height
                            0.0500
         Whole weight
                            0.7115
         Shucked weight
                            0.3160
         Viscera weight
                            0.1595
         Shell weight
                            0.1990
         Rings
                            3.0000
         dtype: float64
In [39]: df.shape
```

Out[39]: (4177, 9)

```
In [43]: upper = np.where(df >= (Q3+1.5*IQR))
    lower = np.where(df <= (Q1-1.5*IQR))
    df.drop(upper[0], inplace = True)
    df.drop(lower[0], inplace = True)
    df.shape</pre>
```

C:\Users\ABI\AppData\Local\Temp/ipykernel_15784/1776641308.py:1: FutureWarning:
Automatic reindexing on DataFrame vs Series comparisons is deprecated and will
raise ValueError in a future version. Do `left, right = left.align(right, axis
=1, copy=False)` before e.g. `left == right`

```
upper = np.where(df >= (Q3+1.5*IQR))
```

C:\Users\ABI\AppData\Local\Temp/ipykernel_15784/1776641308.py:2: FutureWarning:
Automatic reindexing on DataFrame vs Series comparisons is deprecated and will
raise ValueError in a future version. Do `left, right = left.align(right, axis
=1, copy=False)` before e.g. `left == right`

```
lower = np.where(df <= (Q1-1.5*IQR))
```

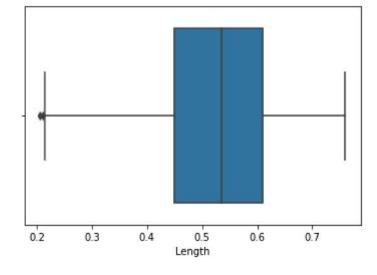
```
Out[43]: (3781, 9)
```

In [44]: | sns.boxplot(df['Length'],data=df)

C:\Users\ABI\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

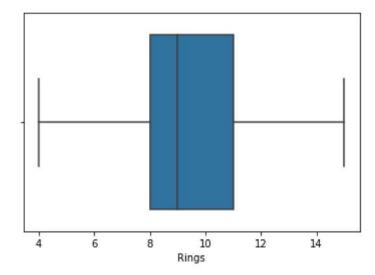
warnings.warn(

Out[44]: <AxesSubplot:xlabel='Length'>



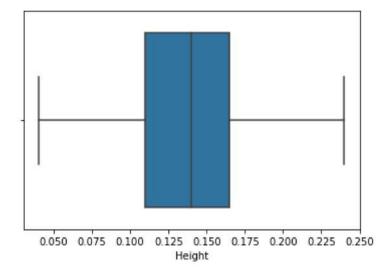
In [45]: sns.boxplot(df['Rings'],data=df)

Out[45]: <AxesSubplot:xlabel='Rings'>



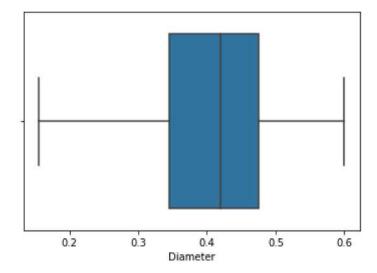
In [46]: sns.boxplot(df['Height'],data=df)

Out[46]: <AxesSubplot:xlabel='Height'>



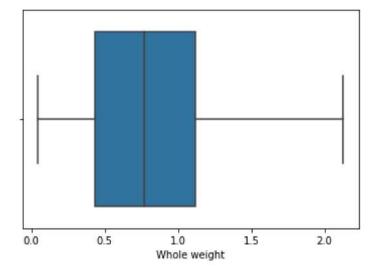
In [47]: | sns.boxplot(df['Diameter'],data=df)

Out[47]: <AxesSubplot:xlabel='Diameter'>



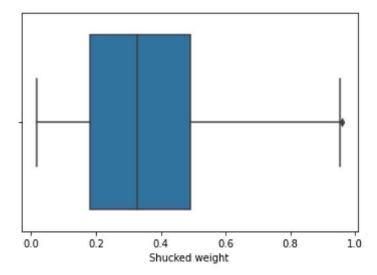
In [48]: sns.boxplot(df['Whole weight'],data=df)

Out[48]: <AxesSubplot:xlabel='Whole weight'>



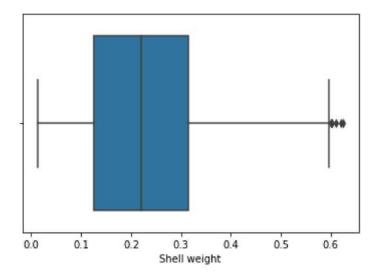
In [49]: sns.boxplot(df['Shucked weight'],data=df)

Out[49]: <AxesSubplot:xlabel='Shucked weight'>



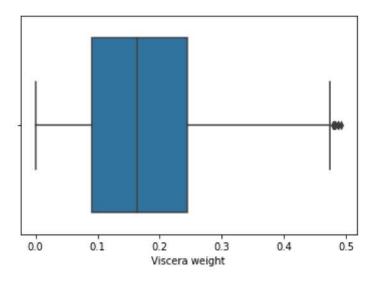
In [50]: sns.boxplot(df['Shell weight'],data=df)

Out[50]: <AxesSubplot:xlabel='Shell weight'>



```
In [51]: sns.boxplot(df['Viscera weight'],data=df)
```

Out[51]: <AxesSubplot:xlabel='Viscera weight'>



```
Label encoding for categorical data
```

```
In [53]: from sklearn.model_selection import train_test_split
    from sklearn.preprocessing import LabelEncoder
    from sklearn.preprocessing import StandardScaler

from sklearn.linear_model import LinearRegression
    from sklearn.svm import SVR
    from sklearn.tree import DecisionTreeRegressor
    from sklearn import metrics
```

```
In [55]: le=LabelEncoder()
df['Sex']=le.fit_transform(df['Sex'])
```

```
In [56]: df
```

Out[56]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	2	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.1500	15
1	2	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700	7
2	0	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100	9
3	2	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550	10
4	1	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550	7
4172	0	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	11
4173	2	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	10
4174	2	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080	9
4175	0	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960	10
4176	2	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12

3781 rows × 9 columns

Spliting the Data into dependent and Independent Variables

```
In [57]: X = df.iloc[:, :-1].values
y = df.iloc[:, -1].values
```

Scaling independent variables

```
In [58]: scaler = StandardScaler()
scaler.fit(df)
```

Out[58]: StandardScaler()

Spliting training and test data

```
In [59]: train_X,val_X,train_y,val_y = train_test_split(X, y, test_size = 0.2, random_stat
```

```
In [60]: print("Shape of Training X :",train_X.shape)
print("Shape of Validation X :",val_X.shape)
```

```
Shape of Training X: (3024, 8) Shape of Validation X: (757, 8)
```

```
In [61]:
         print("Shape of Training y :",train_y.shape)
         print("Shape of Validation y :",val_y.shape)
         Shape of Training y: (3024,)
         Shape of Validation y: (757,)
         Linear regression
In [62]: | lr = LinearRegression()
         lr.fit(train_X,train_y)
Out[62]: LinearRegression()
In [63]: | %%time
         y_pred_val_lr = lr.predict(val_X)
         print('MAE on Validation set :',metrics.mean_absolute_error(val_y, y_pred_val_lr)
         print("\n")
         print('MSE on Validation set :',metrics.mean_squared_error(val_y, y_pred_val_lr))
         print("\n")
         print('RMSE on Validation set :',np.sqrt(metrics.mean absolute error(val y, y pre
         print("\n")
         print('R2 Score on Validation set :',metrics.r2 score(val y, y pred val lr))
         print("\n")
         MAE on Validation set : 1.2719689486359296
         MSE on Validation set : 2.7606215450501024
         RMSE on Validation set: 1.1278160083257949
         R2 Score on Validation set: 0.5119499107890585
         Wall time: 4.82 ms
         Support vector machine
In [64]:
         svm = SVR()
         svm.fit(train_X,train_y)
Out[64]: SVR()
```

```
In [65]:
         %%time
         y_pred_val_svm = svm.predict(val_X)
         print('MAE on Validation set :',metrics.mean_absolute_error(val_y, y_pred_val_svm
         print("\n")
         print('MSE on Validation set :',metrics.mean_squared_error(val_y, y_pred_val_svm)
         print("\n")
         print('RMSE on Validation set :',np.sqrt(metrics.mean_absolute_error(val_y, y_pre
         print("\n")
         print('R2 Score on Validation set :',metrics.r2_score(val_y, y_pred_val_svm))
         print("\n")
         MAE on Validation set : 1.220895278727089
         MSE on Validation set : 2.7012620714060267
         RMSE on Validation set : 1.104941301032362
         R2 Score on Validation set : 0.5224440679687887
         Wall time: 496 ms
         Decision tree regressor
```

```
In [66]: dc = DecisionTreeRegressor(random_state = 0)
dc.fit(train_X,train_y)
```

Out[66]: DecisionTreeRegressor(random state=0)

```
In [67]:
         %%time
         y_pred_val_dc = dc.predict(val_X)
         print('MAE on Validation set :',metrics.mean_absolute_error(val_y, y_pred_val_dc)
         print("\n")
         print('MSE on Validation set :',metrics.mean_squared_error(val_y, y_pred_val_dc))
         print("\n")
         print('RMSE on Validation set :',np.sqrt(metrics.mean_absolute_error(val_y, y_pre
         print("\n")
         print('R2 Score on Validation set :',metrics.r2_score(val_y, y_pred_val_dc))
         print("\n")
         MAE on Validation set: 1.6393659180977542
         MSE on Validation set : 4.88110964332893
         RMSE on Validation set: 1.2803772561623212
         R2 Score on Validation set : 0.13706896870869845
         Wall time: 6.78 ms
 In [ ]: Overview of R2 scores of all models
In [68]:
         print('Logistic Regression R2 Score on Validation set :',metrics.r2 score(val y,
         print('SVR R2 Score on Validation set :',metrics.r2_score(val_y, y_pred_val_svm))
         print('Decision Tree Regressor R2 Score on Validation set :',metrics.r2 score(val
         Logistic Regression R2 Score on Validation set: 0.5119499107890585
         SVR R2 Score on Validation set : 0.5224440679687887
         Decision Tree Regressor R2 Score on Validation set: 0.13706896870869845
```

In []: