19/10/2020 task1

```
In [2]:
         #LINEAR REGRESSION
         #Basic numpy, pandas and matplotlib imports
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         import os
         from sklearn import metrics
         from sklearn.linear model import LinearRegression
         #getting the directory
         THIS FOLDER = os.path.abspath('')
         #setting the training and testing dataset path
         class Dataset:
           train=os.path.join(THIS_FOLDER, 'wineQualityRed_train.csv')
           test=os.path.join(THIS FOLDER, 'wineQualityRed test.csv')
         fields=["fixed acidity","volatile acidity","citric acid","residual sugar","ch
         fields1=["fixed acidity","volatile acidity","citric acid","residual sugar","c
         #X TRAINING DATASET
         train = pd.read csv(Dataset.train, delimiter=';', header=None, skiprows=1,
         #X TESTING DATASET
         test = pd.read csv(Dataset.test, delimiter=';', header=None, skiprows=1,
         train.head()
                                                   free
                                                          total
Out[2]:
             fixed
                  volatile
                         citric
                               residual
                                       chlorides
                                                 sulfur
                                                         sulfur
                                                               density
                                                                       pH sulphates alcohol
           acidity
                  acidity
                          acid
                                 sugar
                                                dioxide dioxide
        0
             10.6
                    0.28
                          0.39
                                  15.5
                                          0.069
                                                   6.0
                                                          23.0
                                                                1.0026 3.12
                                                                                0.66
                                                                                        9.2
                    0.30
                          0.56
        1
              9.4
                                   2.8
                                          0.080
                                                   6.0
                                                          17.0
                                                                0.9964
                                                                      3.15
                                                                                0.92
                                                                                       11.7
        2
             10.6
                    0.36
                          0.59
                                   2.2
                                          0.152
                                                   6.0
                                                          18.0
                                                                0.9986
                                                                      3.04
                                                                                1.05
                                                                                        9.4
        3
             10.6
                    0.36
                          0.60
                                   2.2
                                          0.152
                                                   7.0
                                                          18.0
                                                                0.9986
                                                                                1.06
                                                                      3.04
                                                                                        9.4
        4
             10.6
                    0.44
                          0.68
                                                   6.0
                                                          24.0
                                                                0.9970 3.06
                                                                                0.66
                                   4.1
                                          0.114
                                                                                       13.4
         test.head()
In [3]:
Out[3]:
                                                   free
                                                          total
             fixed volatile
                         citric residual
                                       chlorides
                                                 sulfur
                                                         sulfur
                                                               density
                                                                       pH sulphates alcohol
           acidity
                  acidity
                          acid
                                 sugar
                                                dioxide
                                                       dioxide
        0
              7.4
                    0.70
                          0.00
                                          0.076
                                                           34
                                                                      3.51
                                                                                0.56
                                   1.9
                                                   11.0
                                                                0.9978
                                                                                        94
        1
              7.8
                    0.88
                          0.00
                                   2.6
                                          0.098
                                                  25.0
                                                           67
                                                                0.9968
                                                                      3.20
                                                                                0.68
                                                                                        9.8
        2
              7.8
                    0.76
                          0.04
                                   2.3
                                          0.092
                                                  15.0
                                                           54
                                                                0.9970
                                                                      3.26
                                                                                0.65
                                                                                        9.8
        3
             11.2
                    0.28
                          0.56
                                   1.9
                                          0.075
                                                  17.0
                                                           60
                                                                0.9980
                                                                      3.16
                                                                                0.58
                                                                                        9.8
              7.4
        4
                     0.70
                          0.00
                                   1.9
                                          0.076
                                                   11.0
                                                                0.9978 3.51
                                                                                0.56
                                                                                        9.4
In [4]:
         ##LINEAR REGRESSION FOR SINGLE ATTRIBUTE
          #considering only 'fixed acidity' attribute
         x train=train.iloc[:,0:1].values
```

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```
y_train=train.iloc[:,-1].values
regressor = LinearRegression()
regressor.fit(x_train, y_train)

#taking all test data for regression
x_test=test.iloc[:,0:1].values
y_test=test.iloc[:,-1].values

#predict y for test data
y_pred = regressor.predict(x_test)
df = pd.DataFrame({'Actual': y_test, 'Predicted': y_pred})
df.head()
```

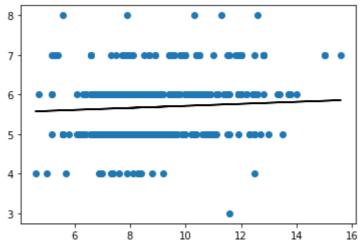
```
Actual Predicted
Out[4]:
                     5.646777
                  5
          1
                  5
                     5.657007
          2
                     5.657007
                  5
                    5.743954
          3
                  6
                     5.646777
          4
                  5
```

Mean Squared Error: 0.5839186420366159

Intercept: 5.457538106303752

Slope: [0.02557288]

fixed acidity x_test vs y_pred



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```
task1
          y_train_all = np.array(train['quality'])
          x test all = np.array(test.drop(['quality'], axis=1))
          y test all = np.array(test['quality'])
          regressor all = LinearRegression()
          regressor all.fit(x train all,y train all)
          #intercept
          print("Intercept:", regressor all.intercept )
          #slope
          print("Slope:", regressor all.coef )
          test pred = regressor all.predict(x test all)
          print(test pred[:10])
          predicted data = np.round (test pred)
          print(predicted_data[:10])
         Intercept: 31.741831140740295
         Slope: [-5.73100144e-03 -1.07148926e+00 -6.30475021e-02 1.02337212e-02
          \hbox{-1.94044757e+00} \quad \hbox{3.51582257e-03} \quad \hbox{-2.15501378e-03} \quad \hbox{-2.68912590e+01}
          -6.34244261e-01 1.24194845e+00 2.67754540e-01]
         [4.94084106 5.16790302 5.214698 5.64933519 4.94084106 4.97871927
          5.05868681 5.26228712 5.26138255 5.75318328]
         [5. 5. 5. 6. 5. 5. 5. 5. 6.]
In [4]: #mean absolute, mean squared, and root mean squared error calculation
         print('Mean Absolute Error:', metrics.mean_absolute_error(y_test_all, test_pr
print('Mean Squared Error:', metrics.mean_squared_error(y_test_all, test_prec
          print('Root Mean Squared Error:', np.sqrt(metrics.mean_squared_error(y_test_&
         Mean Absolute Error: 0.5224781243375761
         Mean Squared Error: 0.4463021333444595
```

Root Mean Squared Error: 0.6680584804824047