MultipleLinearRegression

Program to implement multiple linear regression technique using any standard dataset available in the public domain and evaluate its performance.

The description for all the columns containing data for air pollutants, temperature, relative humidity and absolute humidity is provided below.

Columns	Description
PT08.S1(CO)	PT08.S1 (tin oxide) hourly averaged sensor response (nominally CO targeted)
C6H6(GT)	True hourly averaged Benzene concentration in $\frac{\mu g}{m^3}$
PT08.S2(NMHC)	PT08.S2 (titania) hourly averaged sensor response (nominally $\overline{\mathrm{NMHC}}$ targeted)
PT08.S3(NOx)	PT08.S3 (tungsten oxide) hourly averaged sensor response (nominally NO_x targeted)
PT08.S4(N02)	PT08.S4 (tungsten oxide) hourly averaged sensor response (nominally NO_2 targeted)
PT08.S5(03)	PT08.S5 (indium oxide) hourly averaged sensor response (nominally O_3 targeted)
Т	Temperature in °C
RH	Relative Humidity (%)
АН	AH Absolute Humidity

▼ Multiple Linear Regression Model Using sklearn Module

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1 #Load Dataset & display 1st 5 rows. Github link is as follows:
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⁵ ds.head()

	DateTime	PT08.S1(CO)	C6H6(GT)	PT08.S2(NMHC)	PT08.S3(NOx)	PT08.S4(NO2)	PT08.:
0	2004-03- 10 18:00:00	1360.0	11.9	1046.0	1056.0	1692.0	
1	2004-03- 10 19:00:00	1292.0	9.4	955.0	1174.0	1559.0	
2	2004-03- 10	14N2 N	9.0	939 N	1140 0	1555 0	

^{1 #}Display the columns in dataframe

☐→ Index(['DateTime', 'PT08.S1(C0)', 'C6H6(GT)', 'PT08.S2(NMHC)', 'PT08.S3(NOx)',

^{2 #} https://raw.githubusercontent.com/jiss-sngce/air/main/airquality.csv.csv

³ import pandas as pd

⁴ ds=pd.read_csv('https://raw.githubusercontent.com/jiss-sngce/air/main/airquality.csv.cs

² ds.columns

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'PT08.S4(NO2)', 'PT08.S5(O3)', 'T', 'RH', 'AH', 'Year', 'Month', 'Day',
                'Day Name'],
               dtype='object')
    1 # Build a linear regression model using the sklearn module by including all the feature
    2 from sklearn.model_selection import train_test_split
    3 from sklearn.linear_model import LinearRegression
    4 features=list(ds.columns.values[1:-1])
    5 features.remove('RH')
    6 f=ds[features]
    7 t=ds['RH']
    8 # Splitting the DataFrame into the train and test sets.
    9 # Test set will have 33% of the values.
   10 f_train,f_test,t_train,t_test=train_test_split(f,t,test_size=0.33,random_state=42)
   11 t_train_reshaped=t_train.values.reshape(-1,1)
   12 t_test_reshaped=t_test.values.reshape(-1,1)
   14 # Build a linear regression model using the 'sklearn.linear_model' module.
   15 sklearn_lin_reg=LinearRegression()
   16 sklearn_lin_reg.fit(f_train,t_train_reshaped)
   17 # Print the value of the intercept .
   18
   19 print("intercept = ",sklearn_lin_reg.intercept_[0])
   20
   21 # Print the names of the features along with the values of their corresponding coeffici
   22 print("\ncoeff = ",sklearn_lin_reg.coef_)
   23 for item in list(zip(f.columns.values,sklearn_lin_reg.coef_[0])):
        print(item[0],item[1])
        intercept = -15028.451823247718
        coeff = [[ 1.48327948e-02 -9.03464156e-01 -5.88095941e-03 1.50325488e-03
            2.64965020e-02 -1.06574176e-03 -2.35491907e+00 2.95517421e+01
            7.50515310e+00 1.16786097e+00 3.52321248e-02]]
        PT08.S1(CO) 0.014832794792690625
        C6H6(GT) -0.9034641560183382
        PT08.S2(NMHC) -0.005880959405385411
        PT08.S3(NOx) 0.0015032548783276978
        PT08.S4(NO2) 0.026496502045666503
        PT08.S5(03) -0.001065741763271788
        T -2.354919067592639
        AH 29.551742104329783
        Year 7.505153097892558
        Month 1.1678609682998067
        Day 0.03523212478929974
    1 # Evaluate the linear regression model using the 'r2 score', 'mean squared error' & 'me
    2 from sklearn.metrics import r2 score, mean squared error, mean absolute error
    3 import numpy as np
    4 t_train_pred=sklearn_lin_reg.predict(f_train)
    5 t_test_pred=sklearn_lin_reg.predict(f_test)
    7 print("**** TRAIN SET ****")
    8 print("R-squared = ",r2_score(t_train_reshaped,t_train_pred))
    9 print("mean squared error = ",mean_squared_error(t_train_reshaped,t_train_pred))
    10 nrint("root mean squared error = " nn sart(mean squared error(t train reshaned t train
https://colab.research.google.com/drive/1yPINOh6ZrEw7v7sZCFofiHxYk3H0FLT-#scrollTo=-BxonYJweOIM&printMode=true
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11 print("mean absolute error = ",mean_absolute_error(t_train_reshaped,t_train_pred))
12 print("\n**** TEST SET ****")
13 print("R-squared = ",r2_score(t_test_reshaped,t_test_pred))
14 print("mean squared error = ",mean_squared_error(t_test_reshaped,t_test_pred))
15 print("root mean squared error = ",np.sqrt(mean_squared_error(t_test_reshaped,t_test_pr
16 print("mean absolute error = ",mean_absolute_error(t_test_reshaped,t_test_pred))
    **** TRAIN SET ****
    R-squared = 0.8785638240066055
    mean squared error = 35.11591834141915
    root mean squared error = 5.925868572742662
    mean absolute error = 4.571994849644625
    **** TEST SET ****
    R-squared = 0.8787020691681189
    mean squared error = 34.702124455429534
    root mean squared error = 5.8908509109830245
    mean absolute error = 4.5644604329243466
```