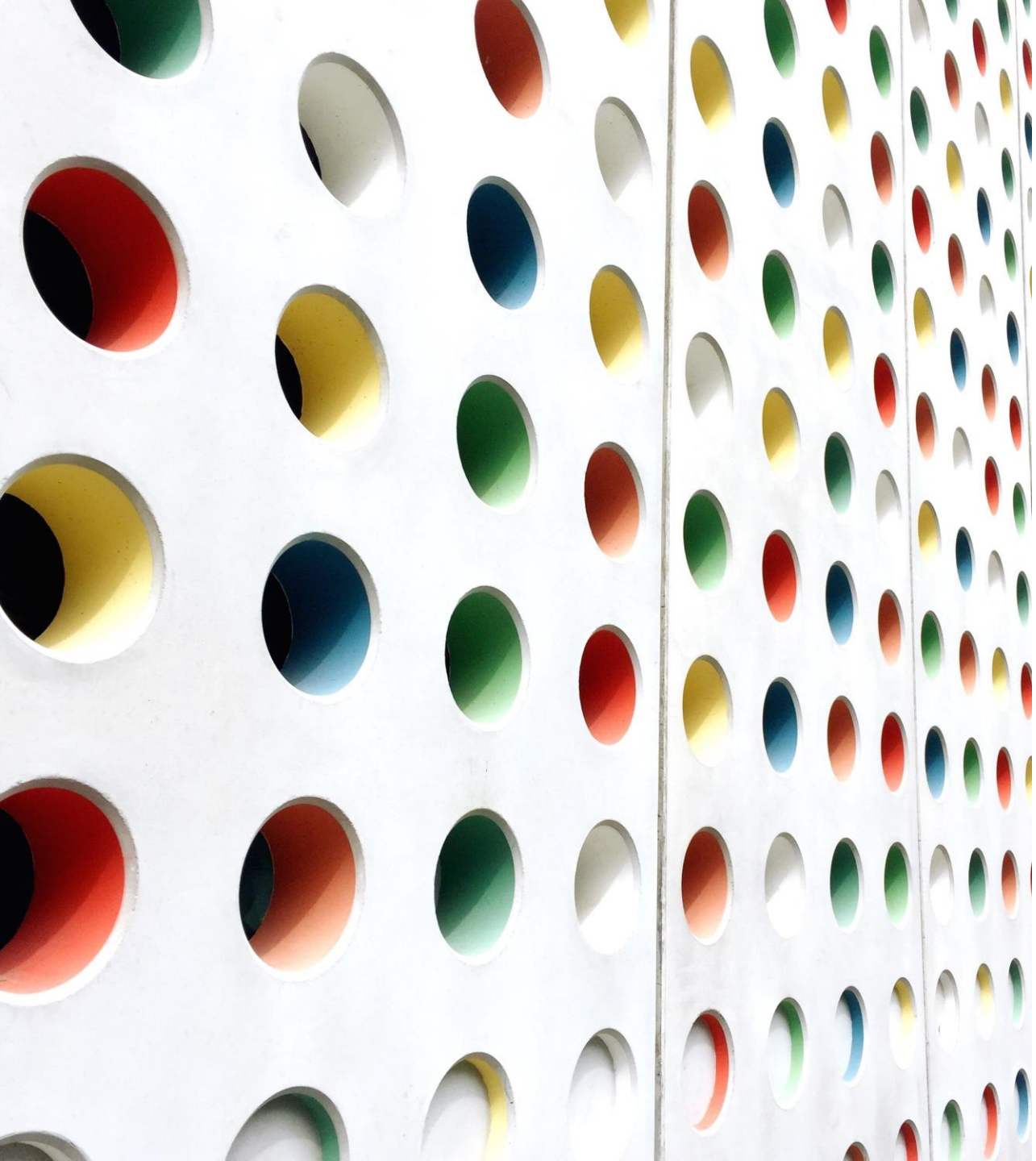


Air Quality - Forecast

SARA SANCHEZ

TIME SERIES ANALYSIS AND MODELING

Objective

To predict the Relative Humidity (RH) based on several features from a data set over the period March 2004 to February 2005.

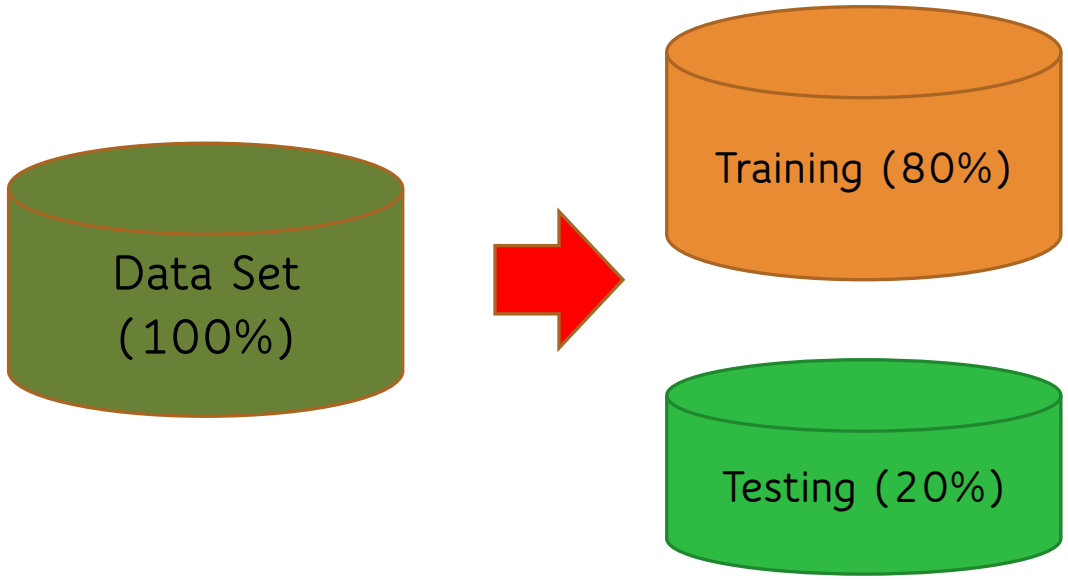
Data Set

- Repository: UCI Machine Learning
- 9,358 Instances
- Dependent variable: Relative Humidity (RH)

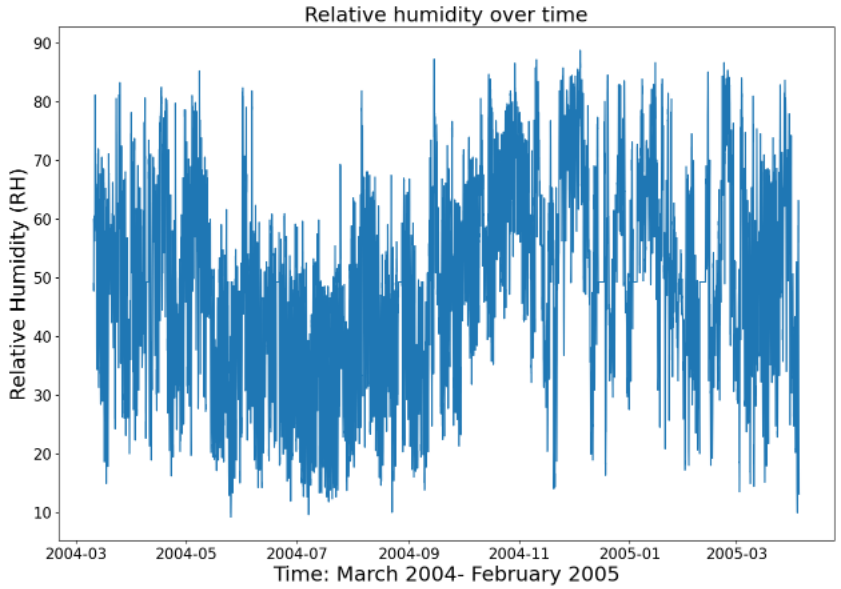
Data Preprocesing

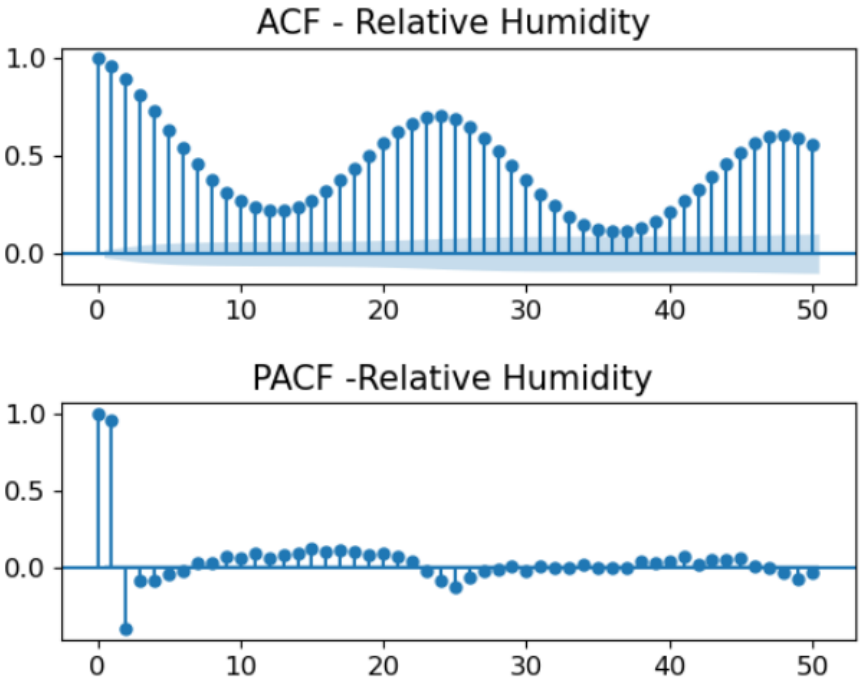
As parte of the data preprocesing:

- Removing Unamed columns
- Manage the dates
- Changing some datatypes
- Handling Null values
- Handling NaN values

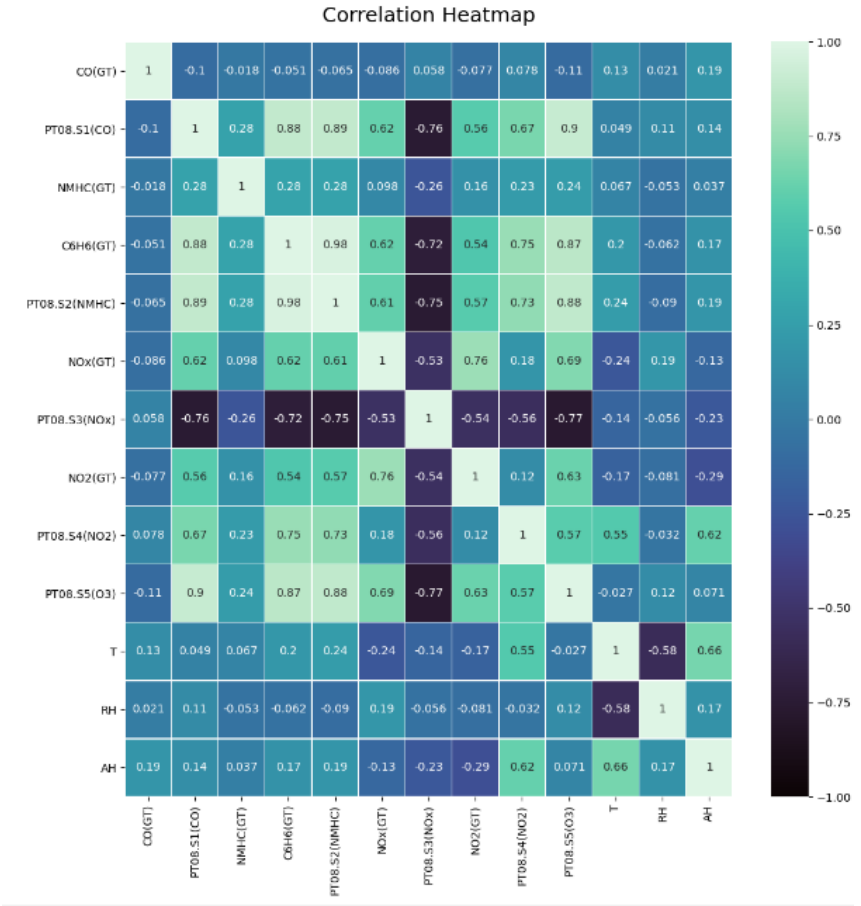


EDA


Relative humidity over time
Relative Humidity (RH)
90
80
70
60
50
40
30
20
10
2004-03
2004-05
2004-07
2004-09
2004-11
2005-01
2005-03
Time: March 2004- February 2005


ACF - Relative Humidity
1.0
0.5
0.0
0
10
20
30
40
50
PACF -Relative Humidity
1.0
0.5
0.0
0
10
20
30
40
50

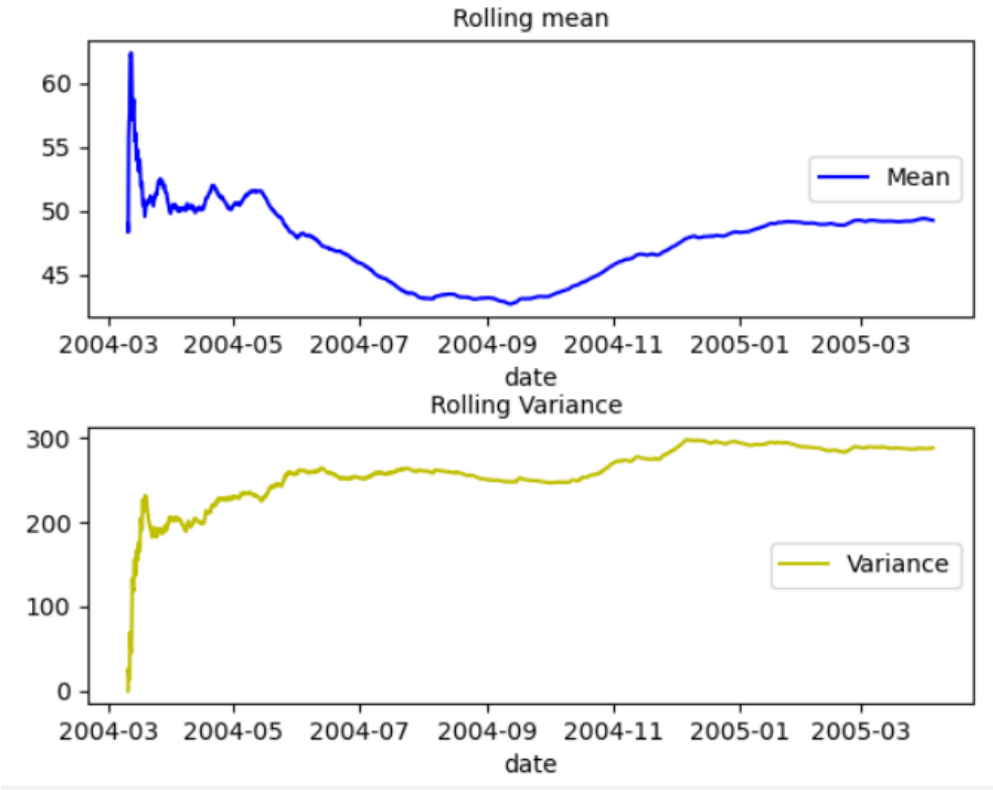
Correlation Heatmap



Checking - Stationality

```
ADF Statistic: -7.391164
p-value: 0.000000
Critical Values:
    1%: -3.431
    5%: -2.862
    10%: -2.567
```

```
Results of KPSS Test:
Test Statistic            2.963095
p-value                   0.010000
LagsUsed                 52.000000
Critical Value (10%)      0.347000
Critical Value (5%)       0.463000
Critical Value (2.5%)     0.574000
Critical Value (1%)       0.739000
dtype: float64
```

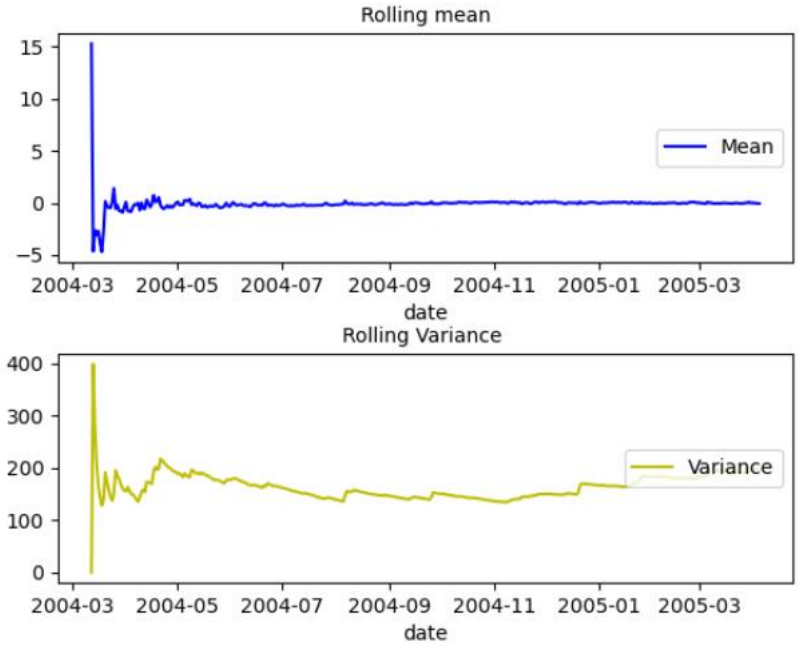


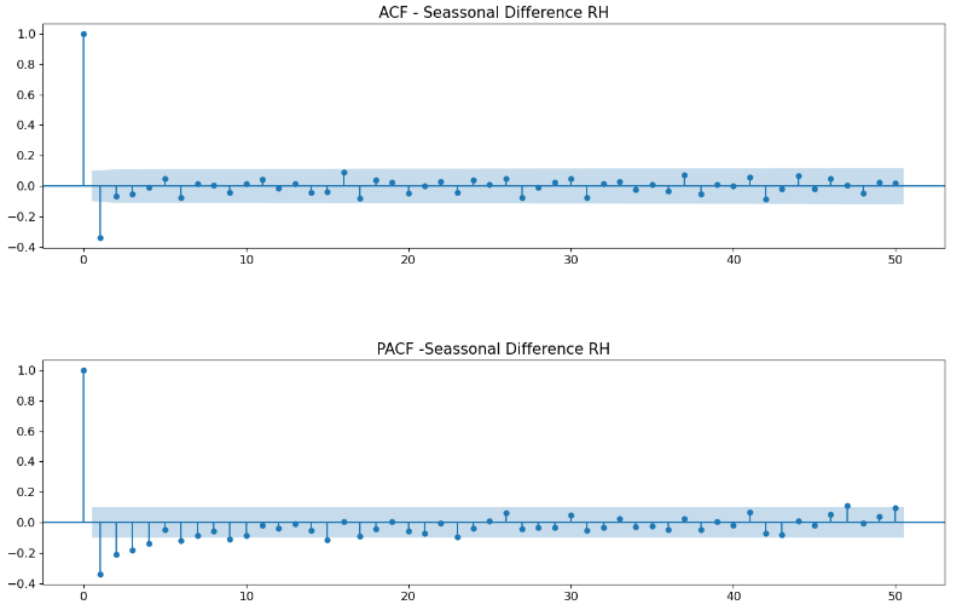
Seasonal difference – 24 periods

ADF Statistic: -9.538226
p-value: 0.000000
Critical Values:
    1%: -3.448
    5%: -2.869
    10%: -2.571

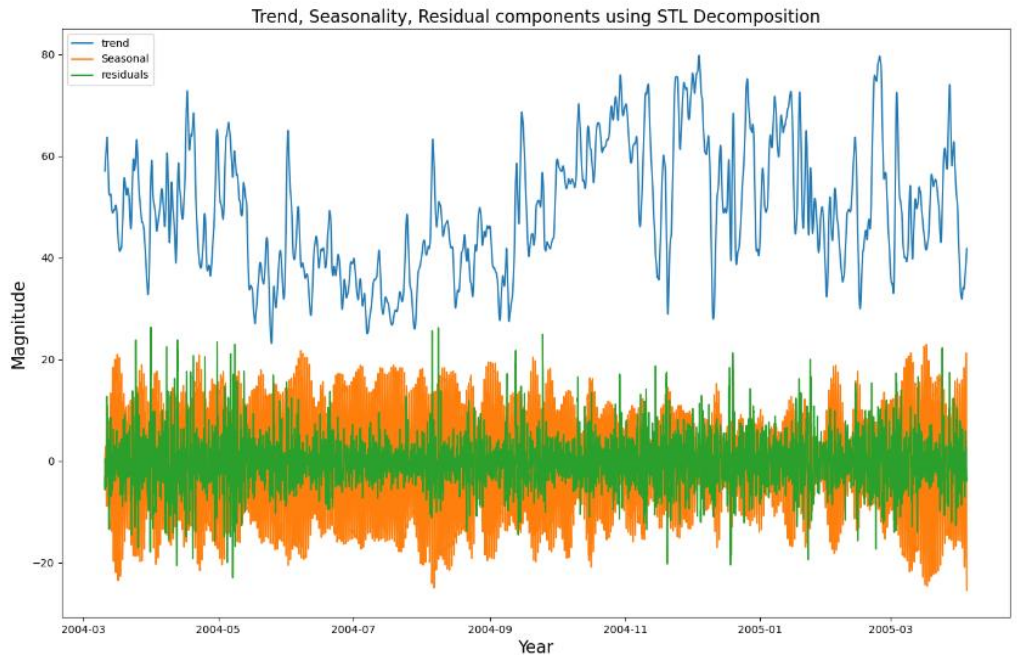
Results of KPSS Test:

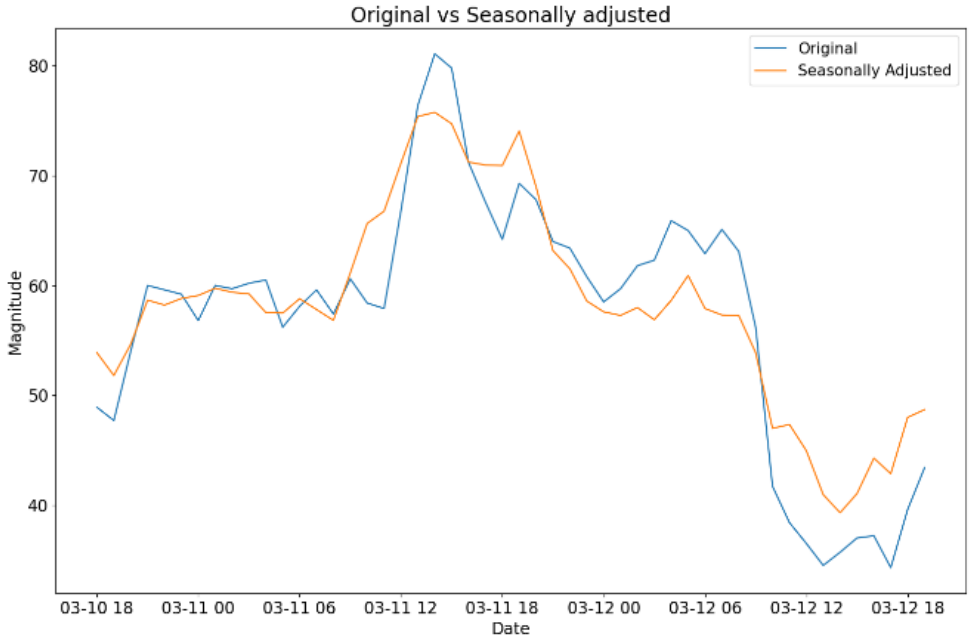
Test Statistic	0.117078
p-value	0.100000
LagsUsed	53.000000
Critical Value (10%)	0.347000
Critical Value (5%)	0.463000
Critical Value (2.5%)	0.574000
Critical Value (1%)	0.739000





Time Series Decomposition

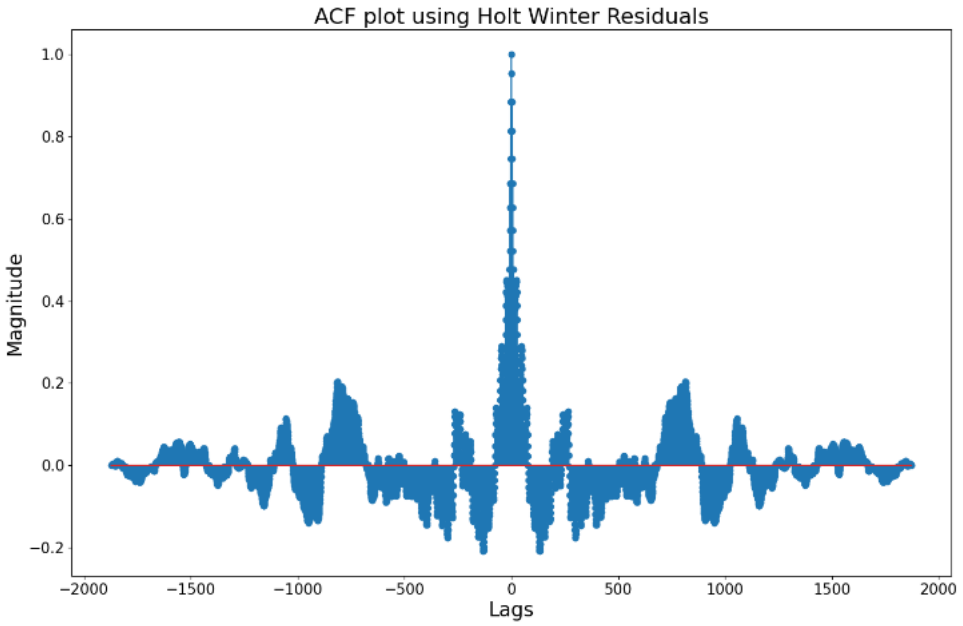


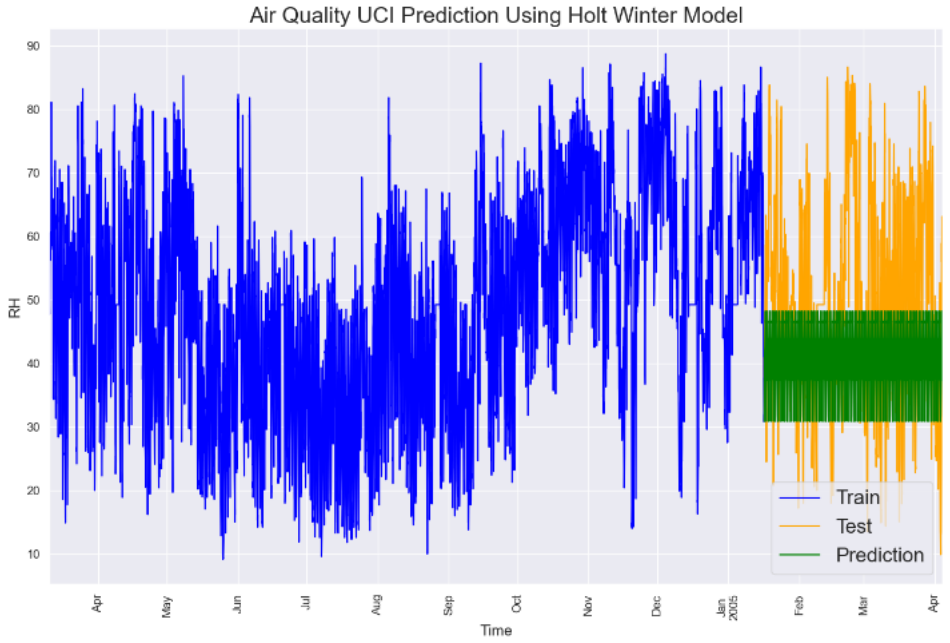


Strength of trend for Air quality dataset is 0.879

Strength of seasonality for Air quality dataset is 0.807

Holt-Winter


ACF plot using Holt Winter Residuals
Magnitude
Lags
1.0
0.8
0.6
0.4
0.2
0.0
−0.2
−2000
−1500
−1000
−500
0
500
1000
1500
2000


Air Quality UCI Prediction Using Holt Winter Model
RH
Time
Train
Test
Prediction
90
80
70
60
50
40
30
20
10
Apr
May
Jun
Jul
Aug
Sep
Oct
Nov
Dec
Jan
2005
Feb
Mar
Apr

Feature Selection

OLS Regression Results

Dep. Variable:	RH	R-squared (uncentered):	0.987
Model:	OLS	Adj. R-squared (uncentered):	0.987
Method:	Least Squares	F-statistic:	4.919e+04
Date:	Sat, 30 Apr 2022	Prob (F-statistic):	0.00
Time:	16:41:34	Log-Likelihood:	-23784.
No. Observations:	7485	AIC:	4.759e+04
Df Residuals:	7473	BIC:	4.767e+04
Df Model:	12		
Covariance Type:	nonrobust		

	coef	std err	t	P>|t|	[0.025	0.975]
CO(GT)	0.0006	0.000	3.610	0.000	0.000	0.001
PT08.S1(CO)	0.0135	0.001	15.513	0.000	0.012	0.015
NMHC(GT)	0.0033	0.001	3.339	0.001	0.001	0.005
C6H6(GT)	-1.6980	0.037	-45.856	0.000	-1.771	-1.625
PT08.S2(NMHC)	0.0339	0.001	29.384	0.000	0.032	0.036
NOx(GT)	0.0157	0.001	21.564	0.000	0.014	0.017
PT08.S3(NOx)	0.0104	0.000	30.786	0.000	0.010	0.011
NO2(GT)	-0.0241	0.003	-7.061	0.000	-0.031	-0.017
PT08.S4(NO2)	0.0122	0.001	22.974	0.000	0.011	0.013
PT08.S5(O3)	-0.0017	0.001	-3.173	0.002	-0.003	-0.001
T	-2.3287	0.014	-169.999	0.000	-2.356	-2.302
AH	34.5833	0.321	107.714	0.000	33.954	35.213

Omnibus:	479.490	Durbin-Watson:	0.163
Prob(Omnibus):	0.000	Jarque-Bera (JB):	607.479
Skew:	0.607	Prob(JB):	1.22e-132
Kurtosis:	3.688	Cond. No.	1.23e+04

OLS Regression Results

Dep. Variable:	RH	R-squared (uncentered):	0.987
Model:	OLS	Adj. R-squared (uncentered):	0.987
Method:	Least Squares	F-statistic:	5.887e+04
Date:	Sat, 30 Apr 2022	Prob (F-statistic):	0.00
Time:	17:05:23	Log-Likelihood:	-23795.
No. Observations:	7485	AIC:	4.761e+04
Df Residuals:	7475	BIC:	4.768e+04
Df Model:	10		
Covariance Type:	nonrobust		

	coef	std err	t	P>|t|	[0.025	0.975]
CO(GT)	0.0006	0.000	3.876	0.000	0.000	0.001
PT08.S1(CO)	0.0127	0.001	15.830	0.000	0.011	0.014
C6H6(GT)	-1.7151	0.037	-46.484	0.000	-1.787	-1.643
PT08.S2(NMHC)	0.0333	0.001	30.100	0.000	0.031	0.035
NOx(GT)	0.0156	0.001	21.357	0.000	0.014	0.017
PT08.S3(NOx)	0.0108	0.000	33.278	0.000	0.010	0.011
NO2(GT)	-0.0249	0.003	-7.416	0.000	-0.032	-0.018
PT08.S4(NO2)	0.0124	0.001	23.597	0.000	0.011	0.013
T	-2.3144	0.013	-178.661	0.000	-2.340	-2.289
AH	34.4295	0.318	108.315	0.000	33.806	35.053

Omnibus:	497.707	Durbin-Watson:	0.163
Prob(Omnibus):	0.000	Jarque-Bera (JB):	631.717
Skew:	0.624	Prob(JB):	6.67e-138
Kurtosis:	3.686	Cond. No.	1.10e+04

Multiple Linear Regression

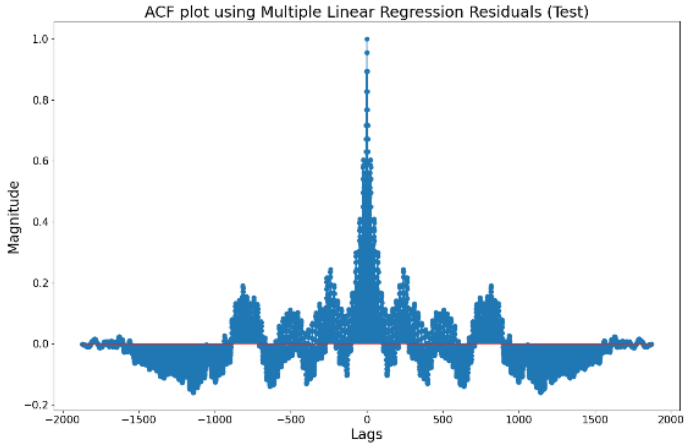
OLS Regression Results

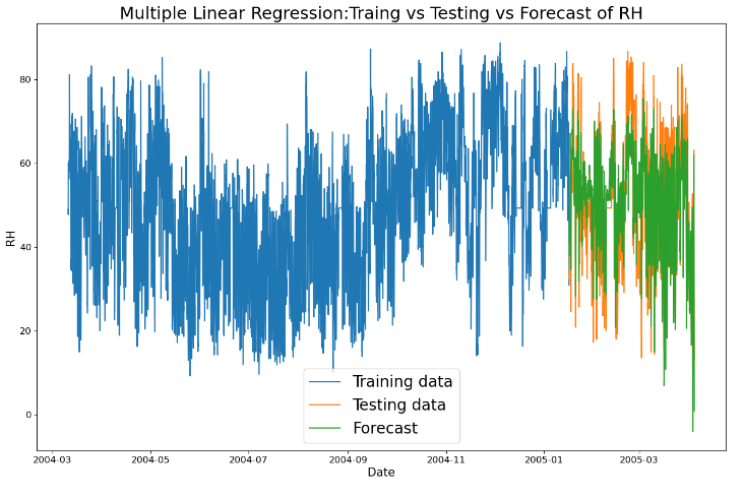
Dep. Variable:	RH	R-squared (uncentered):	0.987
Model:	OLS	Adj. R-squared (uncentered):	0.987
Method:	Least Squares	F-statistic:	4.919e+04
Date:	Mon, 02 May 2022	Prob (F-statistic):	0.00
Time:	11:32:36	Log-Likelihood:	-23784.
No. Observations:	7485	AIC:	4.759e+04
Df Residuals:	7473	BIC:	4.767e+04
Df Model:	12		
Covariance Type:	nonrobust		

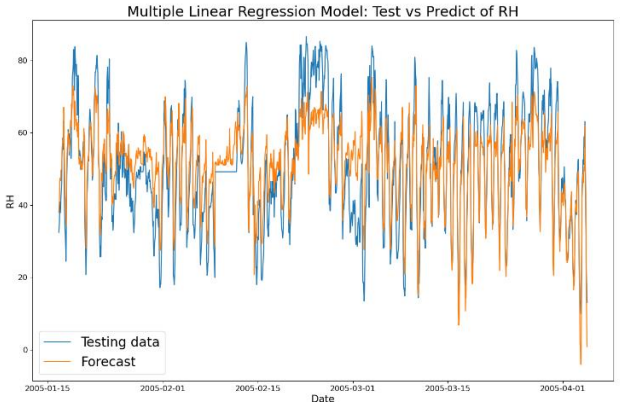
	coef	std err	t	P>|t|	[0.025	0.975]
CO(GT)	0.0006	0.000	3.610	0.000	0.000	0.001
PT08.S1(CO)	0.0135	0.001	15.513	0.000	0.012	0.015
NMHC(GT)	0.0033	0.001	3.339	0.001	0.001	0.005
C6H6(GT)	-1.6980	0.037	-45.856	0.000	-1.771	-1.625
PT08.S2(NMHC)	0.0339	0.001	29.384	0.000	0.032	0.036
NOx(GT)	0.0157	0.001	21.564	0.000	0.014	0.017
PT08.S3(NOx)	0.0104	0.000	30.786	0.000	0.010	0.011
NO2(GT)	-0.0241	0.003	-7.061	0.000	-0.031	-0.017
PT08.S4(NO2)	0.0122	0.001	22.974	0.000	0.011	0.013
PT08.S5(O3)	-0.0017	0.001	-3.173	0.002	-0.003	-0.001
T	-2.3287	0.014	-169.999	0.000	-2.356	-2.302
AH	34.5833	0.321	107.714	0.000	33.954	35.213

Omnibus:	479.490	Durbin-Watson:	0.163
Prob(Omnibus):	0.000	Jarque-Bera (JB):	607.479
Skew:	0.607	Prob(JB):	1.22e-132
Kurtosis:	3.688	Cond. No.	1.23e+04

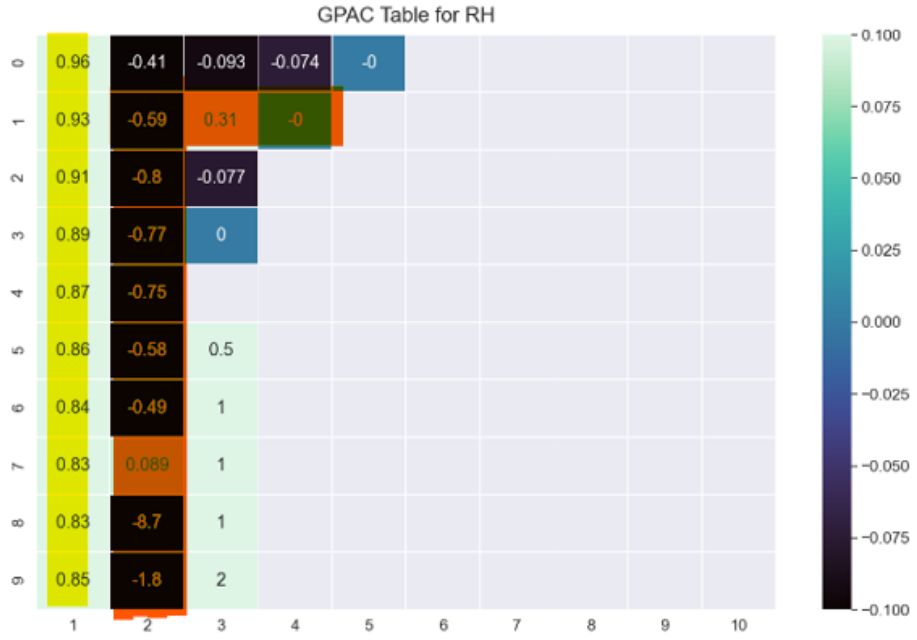
Notes:
[1] R^2 is computed without centering (uncentered) since the model does not contain a constant.
[2] Standard Errors assume that the covariance matrix of the errors is correctly specified.
[3] The condition number is large, 1.23e+04. This might indicate that there are
strong multicollinearity or other numerical problems.







ARMA - Models



ARMA (1,0)

ARMA (2,1)

ARMA (1,0)

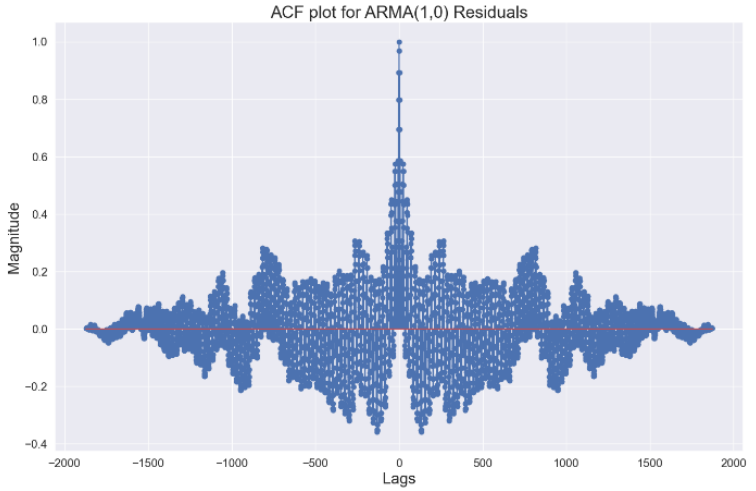
ARMA Model Results

Dep. Variable:	RH	No. Observations:	9357
Model:	ARMA(1, 0)	Log Likelihood	-27529.364
Method:	css-mle	S.D. of innovations	4.586
Date:	Wed, 04 May 2022	AIC	55062.729
Time:	00:04:44	BIC	55077.016
Sample:	03-10-2004 - 04-04-2005	HQIC	55067.581

	coef	std err	z	P>|z|	[0.025	0.975]
ar.L1.RH	0.9629	0.003	345.193	0.000	0.957	0.968

Roots

	Real	Imaginary	Modulus	Frequency
AR.1	1.0385	+0.0000j	1.0385	0.0000

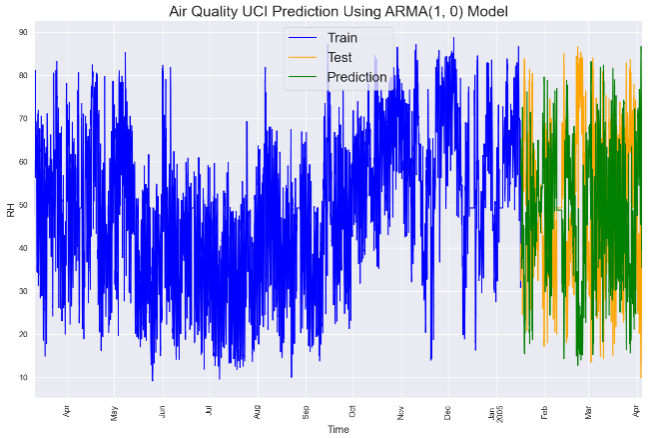


LM – Parameter Estimation

PARAMETER ESTIMATED

LM - The AR coefficient a0 is: 0.9960090800559419

The AR coefficient a0 is: 0.995995649538989



ARMA (2,1)

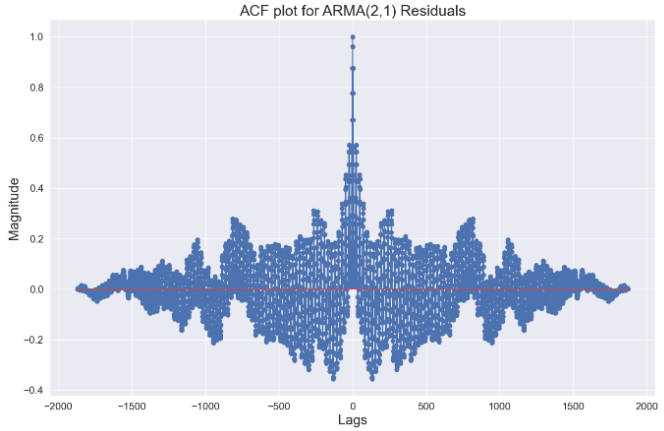
ARMA Model Results

Dep. Variable:	RH	No. Observations:	9357
Model:	ARMA(2, 1)	Log Likelihood	-26603.418
Method:	css-mle	S.D. of innovations	4.154
Date:	Wed, 04 May 2022	AIC	53214.837
Time:	00:12:15	BIC	53243.412
Sample:	03-10-2004 - 04-04-2005	HQIC	53224.542

	coef	std err	z	P>|z|	[0.025	0.975]
ar.L1.RH	1.6266	0.019	85.550	0.000	1.589	1.664
ar.L2.RH	-0.6685	0.018	-36.681	0.000	-0.704	-0.633
ma.L1.RH	-0.3351	0.025	-13.540	0.000	-0.384	-0.287

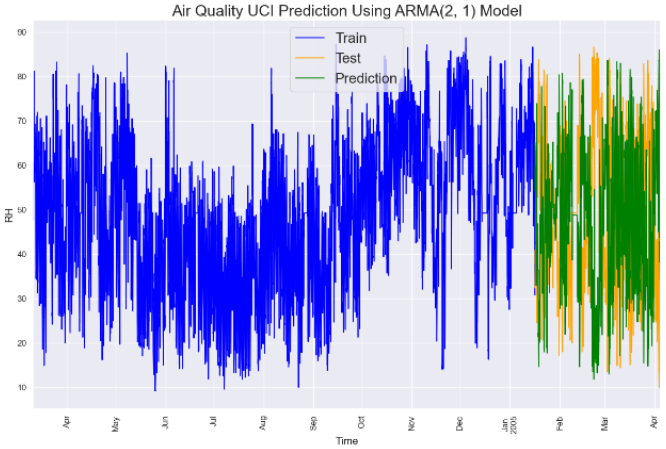
Roots

	Real	Imaginary	Modulus	Frequency
AR.1	1.2166	-0.1258j	1.2230	-0.0164
AR.2	1.2166	+0.1258j	1.2230	0.0164
MA.1	2.9844	+0.0000j	2.9844	0.0000



LM – Parameter Estimation

```
LM - The AR coefficient a0 is: 1.5249896053962666
LM - The AR coefficient a1 is: -0.5298550819910275
LM - The MA coefficient b0 is: -0.1843455766129109
The AR coefficient a0 is: 1.5190248961591442
The AR coefficient a1 is: -0.5239233881532351
The MA coefficient b0 is: -0.17136892003716545
```



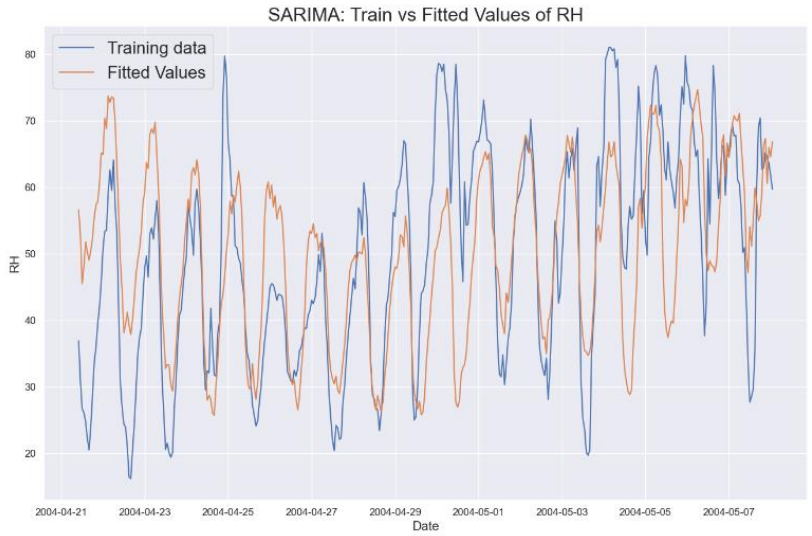
SARIMA (0,0,0) x(0,1,1,24)

SARIMAX Results

Dep. Variable:	RH	No. Observations:	7485
Model:	SARIMAX(0, 1, [1], 24)	Log Likelihood	-28612.806
Date:	Wed, 04 May 2022	AIC	57229.612
Time:	00:42:17	BIC	57243.447
Sample:	03-10-2004	HQIC	57234.365
	- 01-16-2005		
Covariance Type:	opg		

	coef	std err	z	P>|z|	[0.025	0.975]
ma.S.L24	-0.6081	0.008	-73.520	0.000	-0.624	-0.592
sigma2	125.2868	1.544	81.151	0.000	122.261	128.313

Ljung-Box (L1) (Q):	6459.24	Jarque-Bera (JB):	892.57
Prob(Q):	0.00	Prob(JB):	0.00
Heteroskedasticity (H):	1.12	Skew:	0.33
Prob(H) (two-sided):	0.00	Kurtosis:	4.56

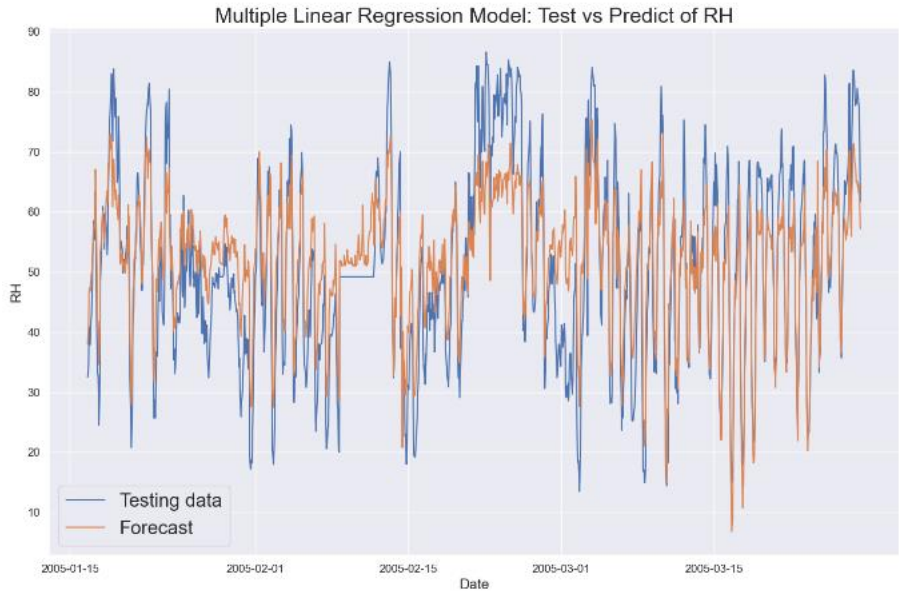


Model's Comparison

BASE MODEL COMPARISON

Model	MSE	RMSE	Residual Mean	Residual Variance	Train Residual Mean	Train Residual Variance	Q Value
Average Model	261.847603	16.181706	1.231013	260.332210	-2.269931e-09	294.757911	310975.648032
Naive Model	592.261704	24.336428	18.218932	260.332210	1.698792e+01	294.757911	310975.648032
Drift Model	675.720932	25.994633	20.333690	262.261970	2.544018e+01	367.999923	305120.133891
Simple Exponential Smoothing Model	582.260222	24.130069	17.942352	260.332210	-3.701496e-03	34.940240	310975.648032
Holt Winter Model	298.016539	17.263156	9.938193	199.248865	-5.502696e-02	19.478492	174779.944494
Multiple Linear Regression Model	60.433018	7.773868	-0.640819	60.022369	1.103621e-01	33.677514	45559.267884
ARMA(1, 0) Model	986.902672	31.415007	2.426130	981.016567	2.185490e-03	1115.137897	316119.924675
ARMA(2, 1) Model	1003.068885	31.671263	2.389274	997.360256	9.273044e-04	1125.326443	315440.233711
SARIMA (0, 0, 0) (0, 1, 1, 24) Model	418.858124	20.466024	-14.524584	207.894582	2.365511e-01	137.638637	178853.964200

Final Model – Multiple Linear Regression



Final Equation:

$$Y= 0.0006 * CD(GT) + 0.0135 * PT08.S1(CO) + 0.0033 * NMHC(GT) - 1.6980 * C6H6(G) + 0.0339 * PT08.S2(NMHC) + 0.0157 * NOx(GT) + 0.0104 * PT08.S3(N0x) - 0.0241 * NO2(GT) + 0.0122 * PT08.S4(NO2) - 0.0017 * PT08.S4(NO2) - 2.3287 * T + 34.5833 * AH$$

Thanks!!!!!!