

Household Power Consumption

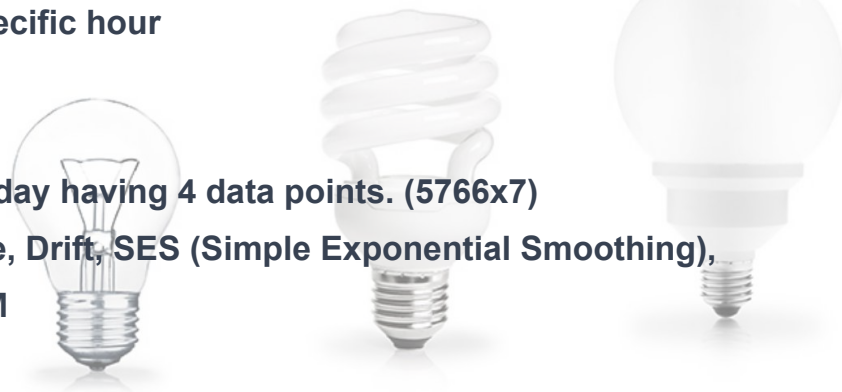
Final Presentation of 2023 Spring Time Series

HaeLee Kim



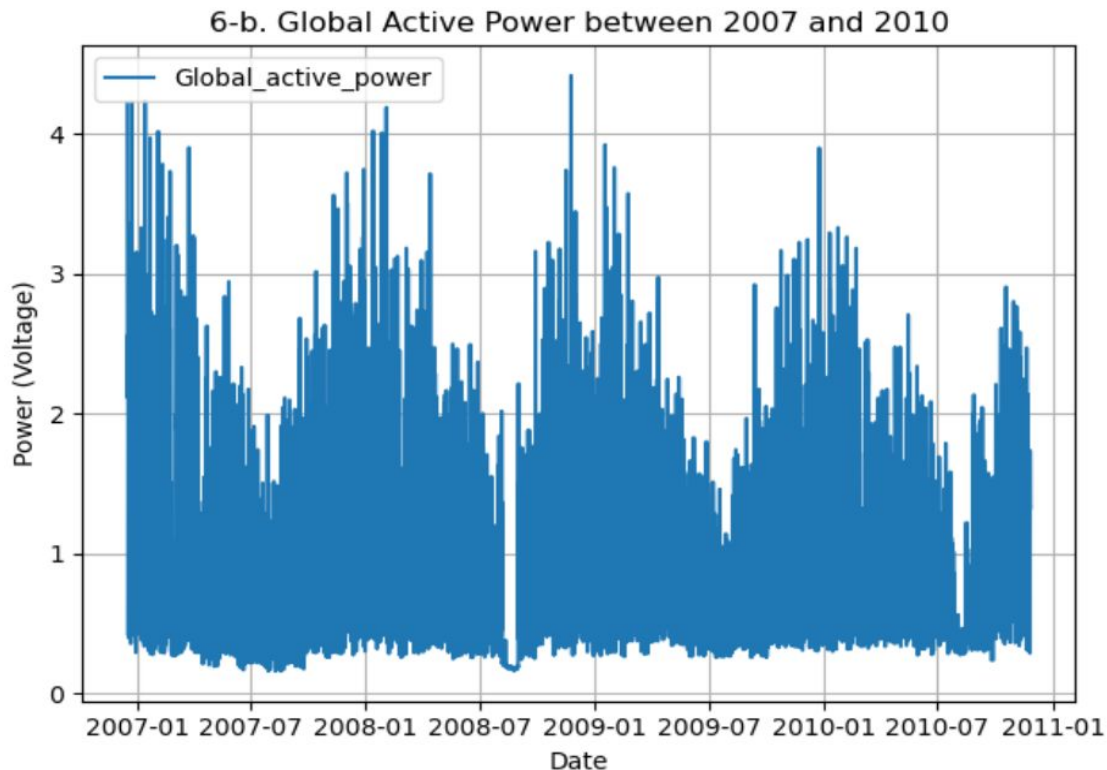
Introduction

- Goal: Accurate predictions of the household power consumption
- Dataset: Individual household electric power consumption Data Set
- Source: <https://archive.ics.uci.edu/ml/datasets/Individual+household+electric+power+consumption>
- Contains 2,075,259 data points with 7 features (type: float)
- Collected in a local area - Sceaux (7km of Paris) - of France
between December 2006 and November 2010 (47 months), with a data point recorded every minute.
- The missing values filled with the mean of that specific hour
- Resampled Dataset :
Step 1. resampled hourly first
Step 2. resampled every 6 hours, resulting in one day having 4 data points. (5766x7)
- Time Series Models: Base Models - Average, Naive, Drift, SES (Simple Exponential Smoothing),
Multiple Linear Regression, ARMA, SARIMA, LSTM

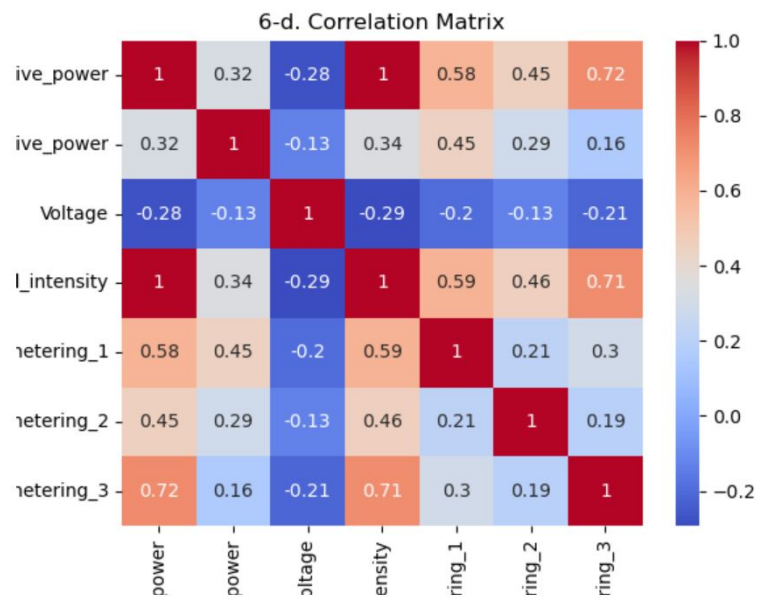
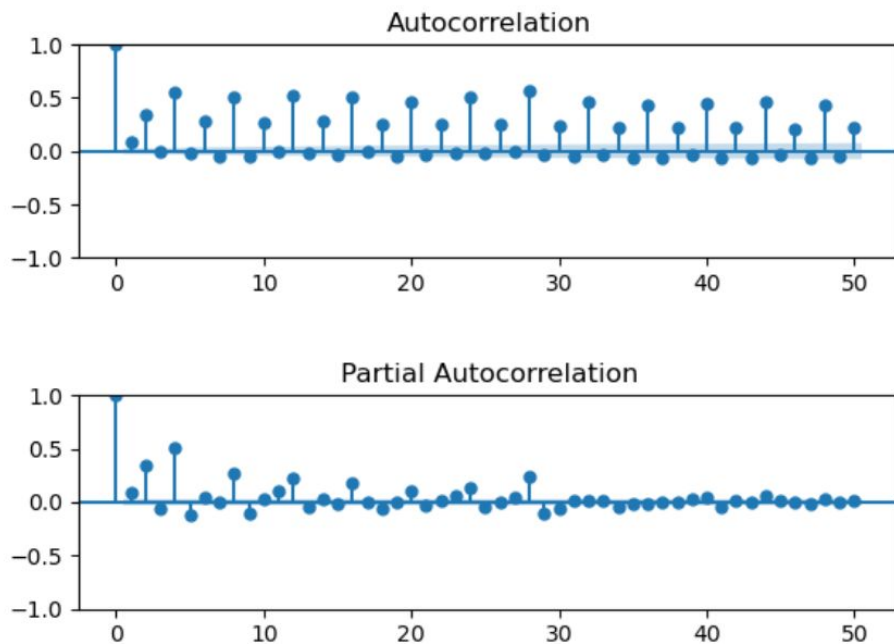


1. Description of the dataset (a) (b)

#	Column	Non-Null Count	Dtype
---	-----	-----	-----
0	Global_active_power	5766 non-null	float64
1	Global_reactive_power	5766 non-null	float64
2	Voltage	5766 non-null	float64
3	Global_intensity	5766 non-null	float64
4	Sub_metering_1	5766 non-null	float64
5	Sub_metering_2	5766 non-null	float64
6	Sub_metering_3	5766 non-null	float64



1. Description of the dataset (c) (d) (e)



e. split the dataset into train set (80%) and test set (20%)

2. Stationarity - ADF / KPSS / Rolling Mean & Variance

ADF Statistic: -5.844490

p-value: 0.000000

Critical Values:

1%: -3.431

5%: -2.862

10%: -2.567

Results of KPSS Test:

Test Statistic 0.443840

p-value 0.058259

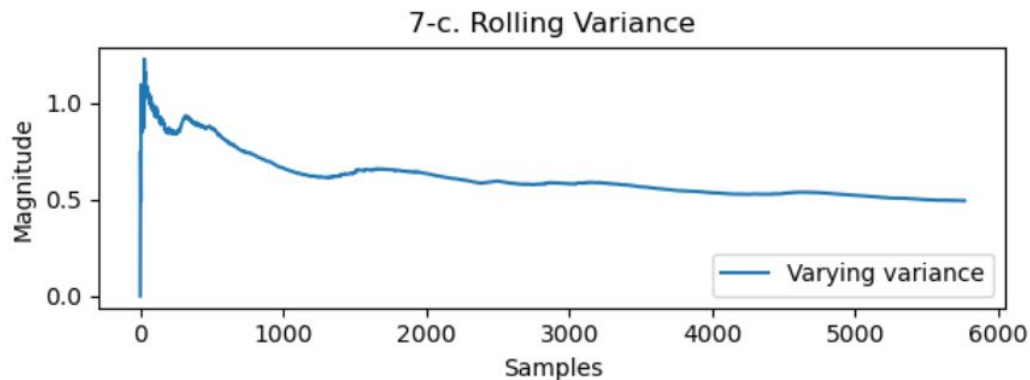
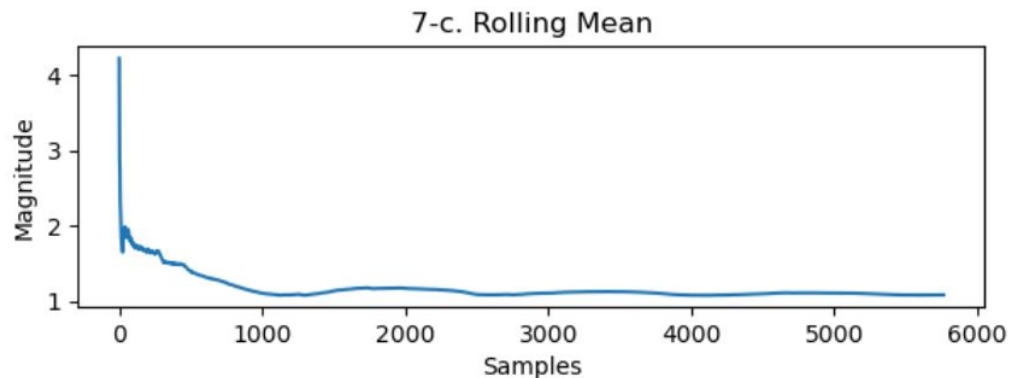
Lags Used 39.000000

Critical Value (10%) 0.347000

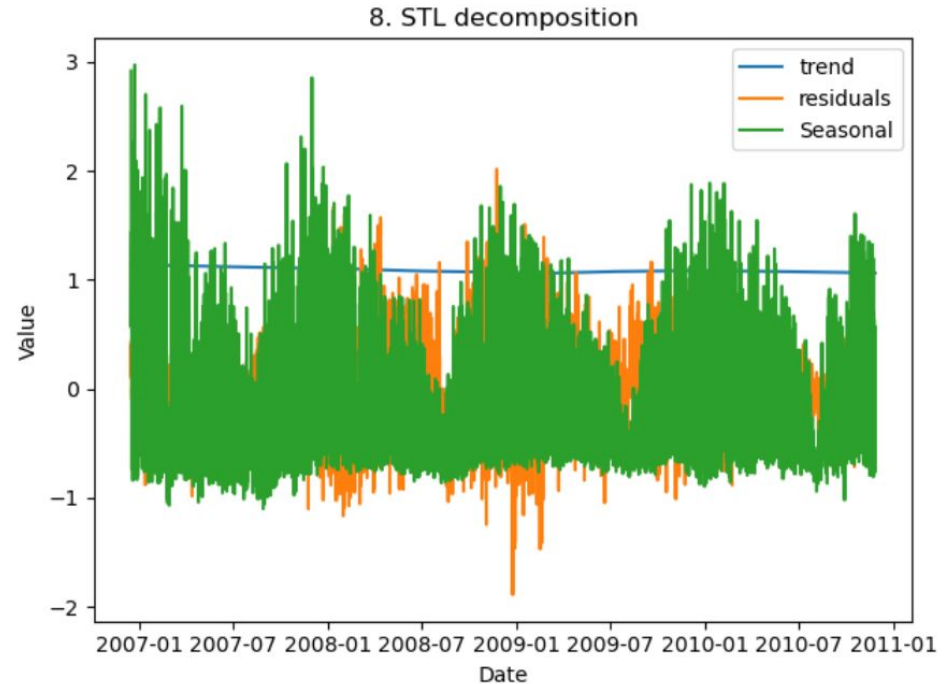
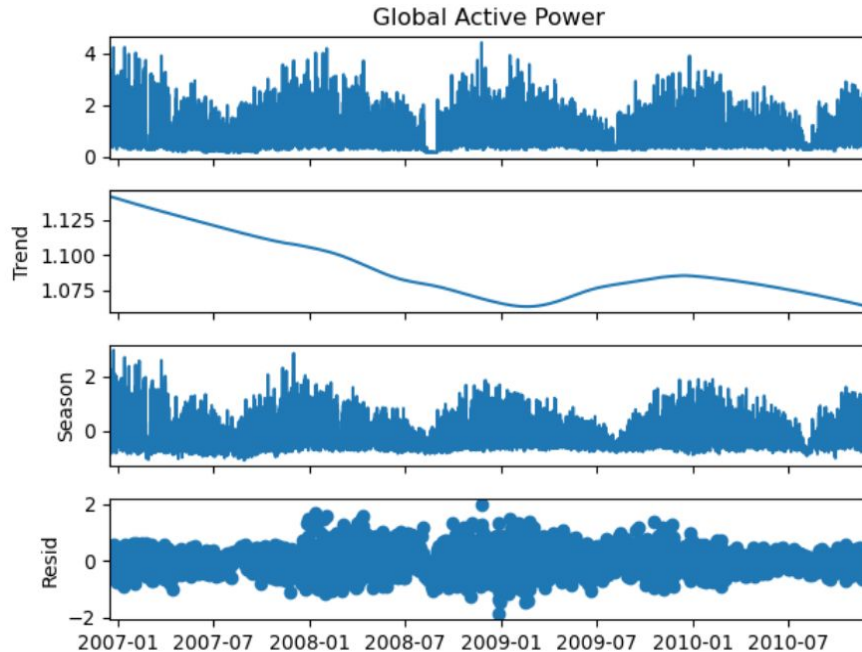
Critical Value (5%) 0.463000

Critical Value (2.5%) 0.574000

Critical Value (1%) 0.739000



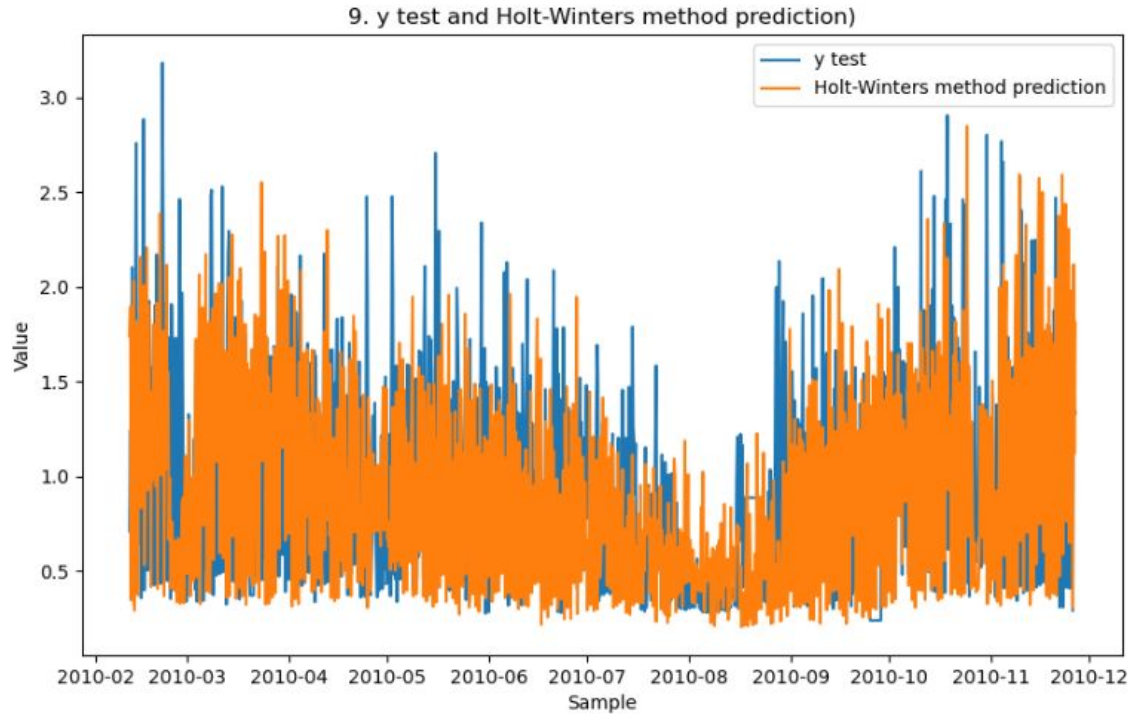
3. Time Series Decomposition



8. The strength of trend for this data set is 0.005443627734045942

8. The strength of Seasonality for this data set is 0.799827216323149

4. Holt-Winters method



Mean Squared Error: 0.2565

5. Feature Selection / Elimination

o 1st step:

VIF = [('Global_reactive_power', 10.17), ('Voltage', 10.43), ('Global_intensity', 13.16), ('Sub_metering_1', 2.36), ('Sub_metering_2', 1.75), ('Sub_metering_3', 5.60)]

Singular Values = [3.34852823e+08 2.01072139e+05 4.46070182e+04 2.44847237e+04 8.30242560e+03 9.90714491e+00]

Condition Number = 5813.70

o 2nd step:

VIF = [('Global_reactive_power', 10.16), ('Voltage', 9.31), ('Sub_metering_1', 1.71), ('Sub_metering_2', 1.40), ('Sub_metering_3', 2.73)]

Condition Number = 5811.26

o 3rd step:

VIF = [('Voltage', 2.53), ('Sub_metering_1', 1.44), ('Sub_metering_2', 1.33), ('Sub_metering_3', 2.73)]

Condition Number = 5811.26

o 4th step:

Backward Stepwise Regression



5. Feature Selection / Elimination

OLS Regression Results

```
=====
Dep. Variable:          y    R-squared:                0.734
Model:                  OLS    Adj. R-squared:           0.734
Method:                 Least Squares    F-statistic:       3178.
Date:                   Tue, 09 May 2023    Prob (F-statistic):    0.00
Time:                   16:34:02    Log-Likelihood:       -3688.4
No. Observations:       4612    AIC:                  7387.
Df Residuals:           4607    BIC:                  7419.
Df Model:                4
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0237	0.008	2.979	0.003	0.008	0.039
Voltage	-0.0457	0.008	-5.818	0.000	-0.061	-0.030
Sub_metering_1	0.3568	0.008	42.999	0.000	0.341	0.373
Sub_metering_2	0.2596	0.008	32.915	0.000	0.244	0.275
Sub_metering_3	0.5756	0.008	67.832	0.000	0.559	0.592

```
=====
Omnibus:                1789.002    Durbin-Watson:           1.393
Prob(Omnibus):           0.000    Jarque-Bera (JB):        7697.182
Skew:                    1.876    Prob(JB):                 0.00
Kurtosis:                8.097    Cond. No.                 1.56
=====
```

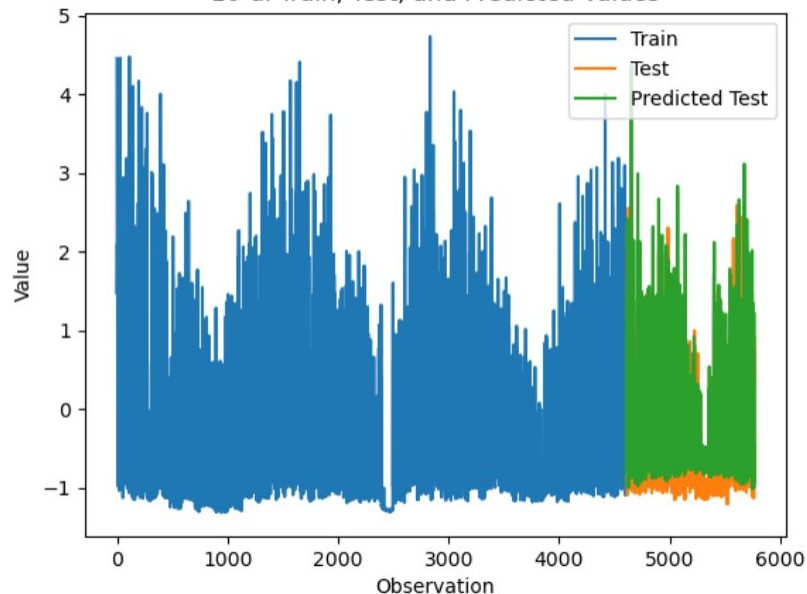
Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

10-d. Features to keep: ['Voltage', 'Sub_metering_1', 'Sub_metering_2', 'Sub_metering_3']

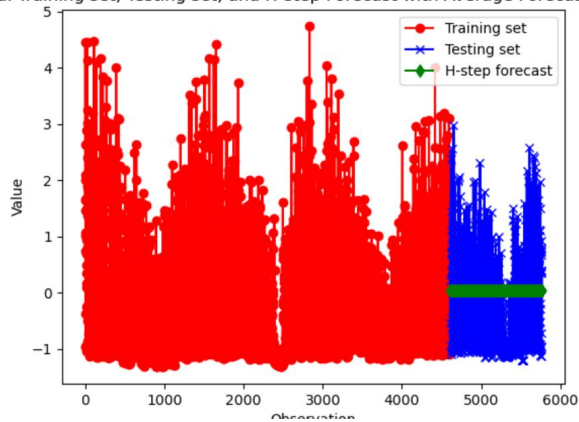
10-d. Features to eliminate: []

10-d. Train, Test, and Predicted Values

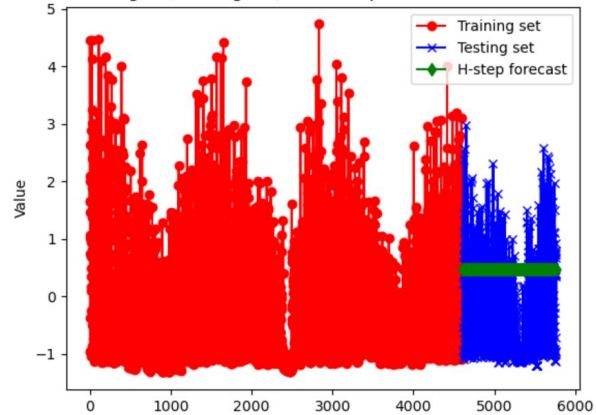


6. Base Models - average, naïve, drift, simple and exponential smoothing

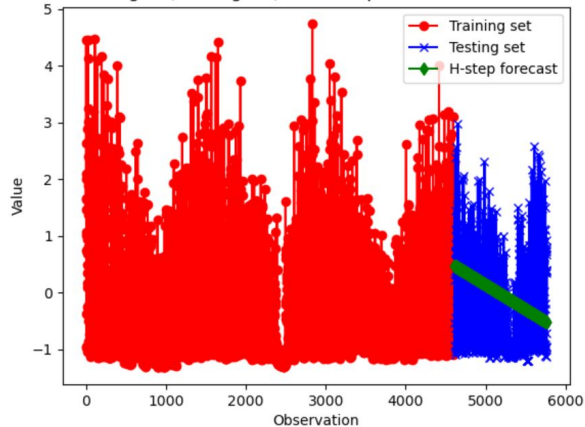
1-a. Training set, Testing set, and H-step Forecast with Average Forecast Method



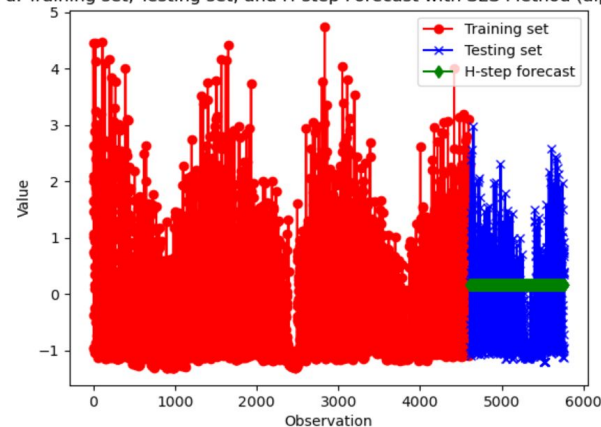
11-b. Training set, Testing set, and H-step Forecast with Naïve Method



11-c. Training set, Testing set, and H-step Forecast with Drift Method



1-d. Training set, Testing set, and H-step Forecast with SES Method ($\alpha=0$)



7. Multiple Linear Regression Model

a. MSE - One-step Ahead Prediction : 0.289

MSE - Forecasting: 1.650

b. AIC: 7386.812

c. BIC: 7418.994

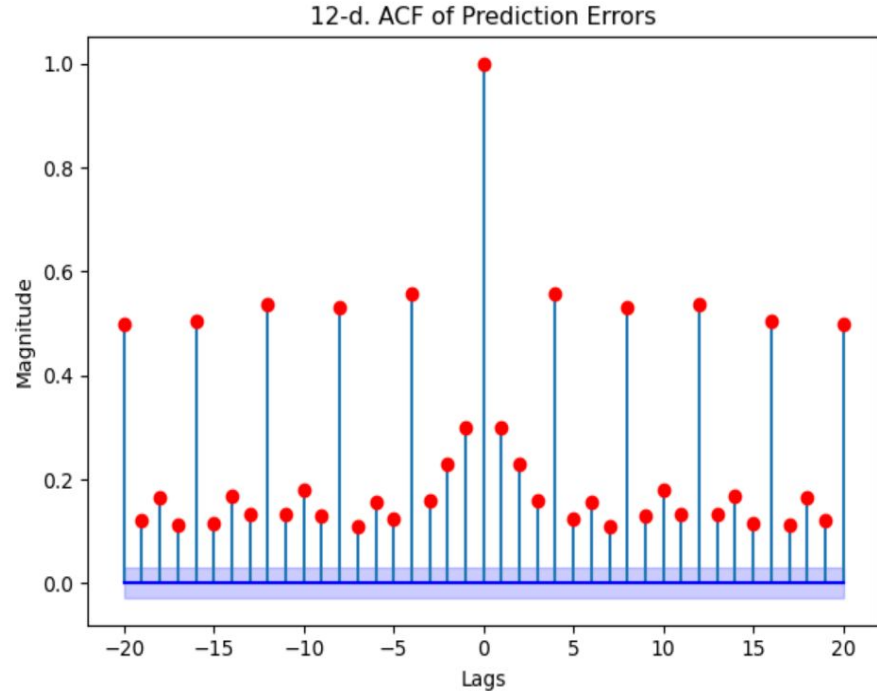
d. RMSE: 0.538

e. Adjusted R-squared: 0.733

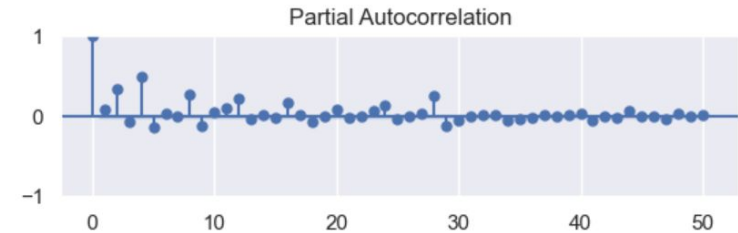
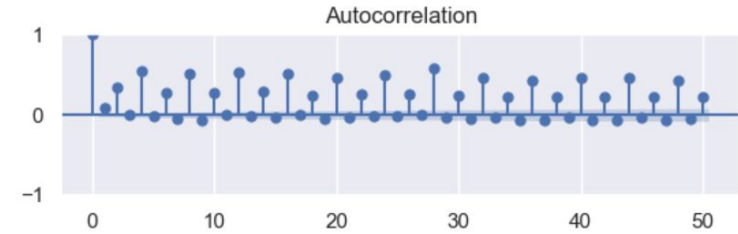
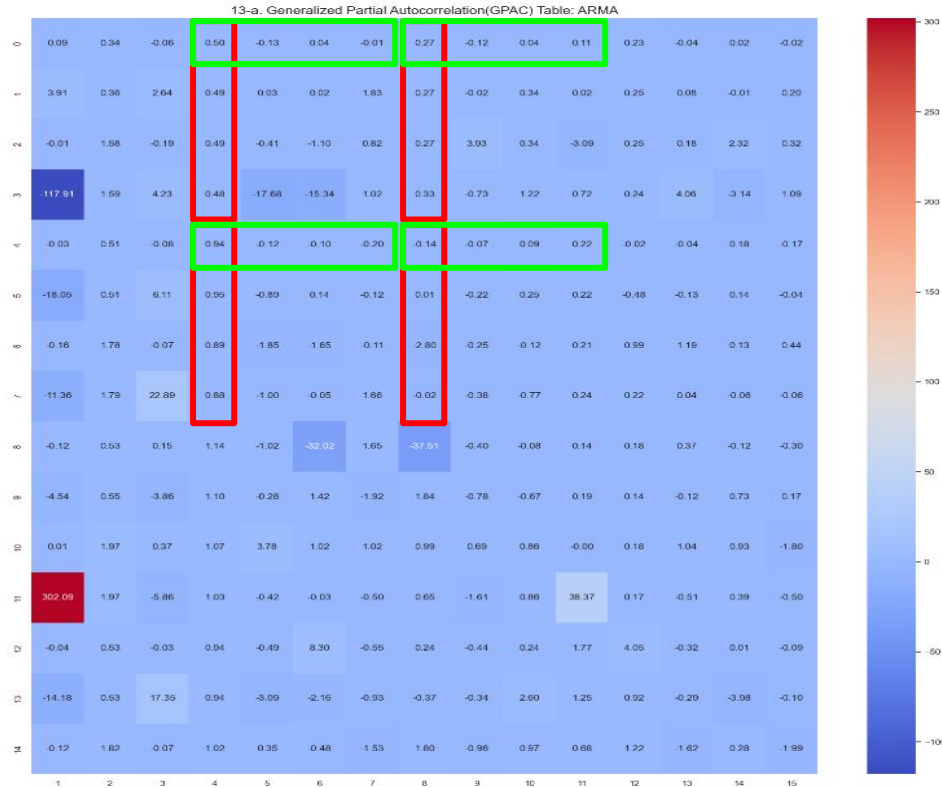
f. Q value is 8325.847

g. Mean of the residuals is 5.145

h. Variance of the residuals is 0.289



8. ARMA / ARIMA / SARIMA model order determination (a) (b) (c)



9. Levenberg Marquardt Algorithm

```
SARIMAX Results
=====
Dep. Variable:          0    No. Observations:          4612
Model:                 ARIMA(4, 0, 4)    Log Likelihood:    -5186.447
Date:                 Tue, 09 May 2023    AIC:              10392.894
Time:                 18:23:51            BIC:              10457.258
Sample:               0    HQIC:             10415.545
                             - 4612
Covariance Type:      opg
=====
              coef    std err          z      P>|z|      [0.025    0.975]
-----
const          0.0662     0.074      0.894     0.371     -0.079     0.211
ar.L1         -0.0087     0.002     -3.827     0.000     -0.013    -0.004
ar.L2         -0.0064     0.002     -2.856     0.004     -0.011    -0.002
ar.L3         -0.0084     0.002     -3.588     0.000     -0.013    -0.004
ar.L4          0.9890     0.002    412.210     0.000     0.984     0.994
ma.L1          0.1149     0.008     13.982     0.000     0.099     0.131
ma.L2          0.1291     0.009     14.925     0.000     0.112     0.146
ma.L3          0.0947     0.008     11.752     0.000     0.079     0.110
ma.L4         -0.8180     0.008    -97.656     0.000    -0.834    -0.802
sigma2         0.5545     0.009     64.535     0.000     0.538     0.571
=====
Ljung-Box (L1) (Q):          74.83    Jarque-Bera (JB):          1321.23
Prob(Q):                   0.00    Prob(JB):                   0.00
Heteroskedasticity (H):      0.61    Skew:                       0.79
Prob(H) (two-sided):         0.00    Kurtosis:                   5.09
=====
```

ARMA(4, 0):

Train MSE (Prediction) = 0.71, Test MSE (Forecasting) = 0.64

ARMA(4, 4):

Train MSE (Prediction) = 0.56, Test MSE (Forecasting) = 0.60

ARMA(8, 0):

Train MSE (Prediction) = 0.64, Test MSE (Forecasting) = 0.64

ARMA(8, 4):

Train MSE (Prediction) = 0.54, Test MSE (Forecasting) = 0.67

10. Diagnostic Analysis

SARIMAX Results

```

=====
Dep. Variable:      0    No. Observations:      4612
Model:              ARIMA(4, 0, 4)    Log Likelihood      -5186.447
Date:              Tue, 09 May 2023    AIC              10392.894
Time:              18:23:51    BIC              10457.258
Sample:            0    HQIC              10415.545
                  - 4612
Covariance Type:    opg
=====

```

	coef	std err	z	P> z	[0.025	0.975]
const	0.0662	0.074	0.894	0.371	-0.079	0.211
ar.L1	-0.0087	0.002	-3.827	0.000	-0.013	-0.004
ar.L2	-0.0064	0.002	-2.856	0.004	-0.011	-0.002
ar.L3	-0.0084	0.002	-3.588	0.000	-0.013	-0.004
ar.L4	0.9890	0.002	412.210	0.000	0.984	0.994
ma.L1	0.1149	0.008	13.982	0.000	0.099	0.131
ma.L2	0.1291	0.009	14.925	0.000	0.112	0.146
ma.L3	0.0947	0.008	11.752	0.000	0.079	0.110
ma.L4	-0.8180	0.008	-97.656	0.000	-0.834	-0.802
sigma2	0.5545	0.009	64.535	0.000	0.538	0.571

```

=====
Ljung-Box (L1) (Q):      74.83    Jarque-Bera (JB):      1321.23
Prob(Q):                 0.00    Prob(JB):              0.00
Heteroskedasticity (H):   0.61    Skew:                  0.79
Prob(H) (two-sided):      0.00    Kurtosis:              5.09
=====

```

MSE of the SARIMA model (Training: Prediction): 0.5599

MSE of the SARIMA model (Test: Forecasting): 0.5730

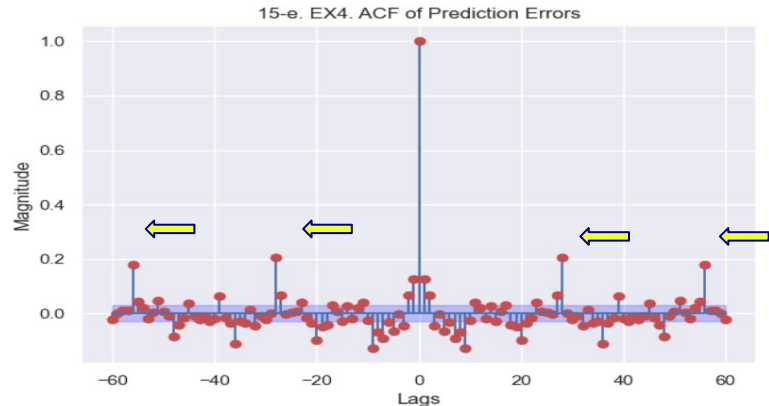
a. Poles - AR roots:

$[-9.99935888e-01+0.j \ -6.77291587e-05+0.99885101j \ -6.77291587e-05-0.99885101j \ 9.91381616e-01+0.j]$

b. Zeros - MA roots:

$[-0.97227237+0.j \ -0.00465345+0.98527024j \ -0.00465345-0.98527024j \ 0.86666145+0.j]$

c. Estimated variance of the error for ARMA(4,4): 1.0



10. Diagnostic Analysis - Finding Better Models

SARIMAX Results

```
=====
Dep. Variable:          0    No. Observations:      4612
Model:                SARIMAX(4, 0, 4)x(1, 0, [1], 28)    Log Likelihood    -4829.440
Date:                 Tue, 09 May 2023    AIC    9680.879
Time:                 18:50:46    BIC    9751.680
Sample:               0    HQIC    9705.796
                        - 4612
Covariance Type:      opg
=====

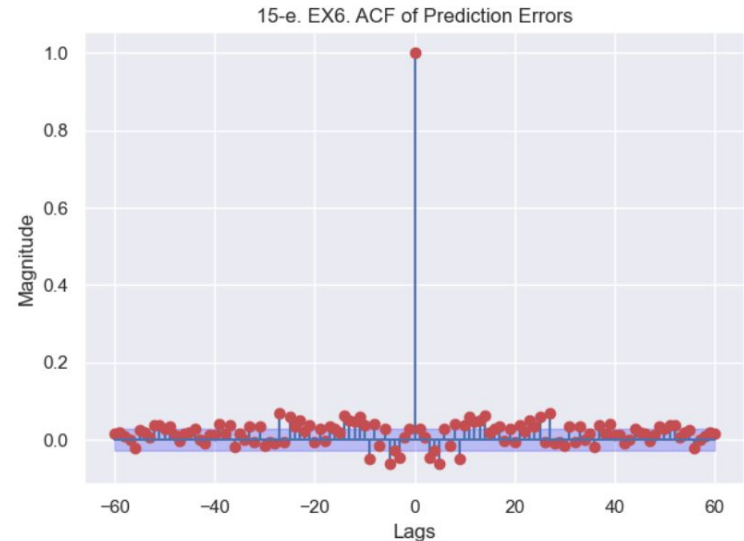
```

	coef	std err	z	P> z	[0.025	0.975]
ar.L1	-0.1402	0.017	-8.129	0.000	-0.174	-0.106
ar.L2	-0.1004	0.019	-5.354	0.000	-0.137	-0.064
ar.L3	-0.1242	0.019	-6.444	0.000	-0.162	-0.086
ar.L4	0.8012	0.017	47.262	0.000	0.768	0.834
ma.L1	0.3580	0.023	15.292	0.000	0.312	0.404
ma.L2	0.3382	0.027	12.671	0.000	0.286	0.391
ma.L3	0.2913	0.026	11.310	0.000	0.241	0.342
ma.L4	-0.5031	0.024	-21.105	0.000	-0.550	-0.456
ar.S.L28	0.9697	0.004	218.533	0.000	0.961	0.978
ma.S.L28	-0.8283	0.011	-73.737	0.000	-0.850	-0.806
sigma2	0.4744	0.007	68.398	0.000	0.461	0.488

```
=====
Ljung-Box (L1) (Q):      4.11    Jarque-Bera (JB):      1508.06
Prob(Q):                 0.04    Prob(JB):           0.00
Heteroskedasticity (H):  0.63    Skew:              0.74
Prob(H) (two-sided):     0.00    Kurtosis:          5.38
=====
```

MSE of the SARIMA model (Training: Prediction): 0.4771

MSE of the SARIMA model (Test: Forecasting): 0.6063



11. Deep Learning Model

Mean squared error (MSE) of LSTM is 1.5568

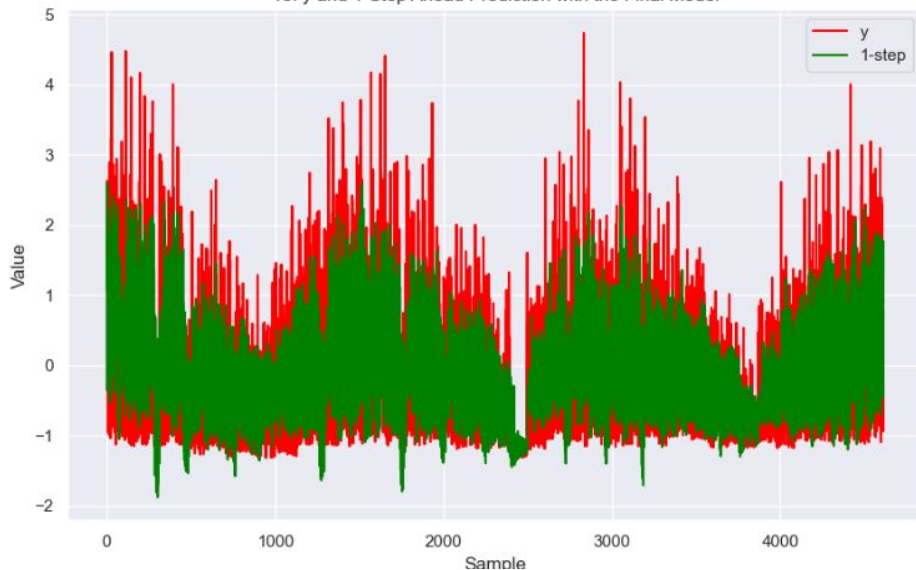


12. Final Model Selection

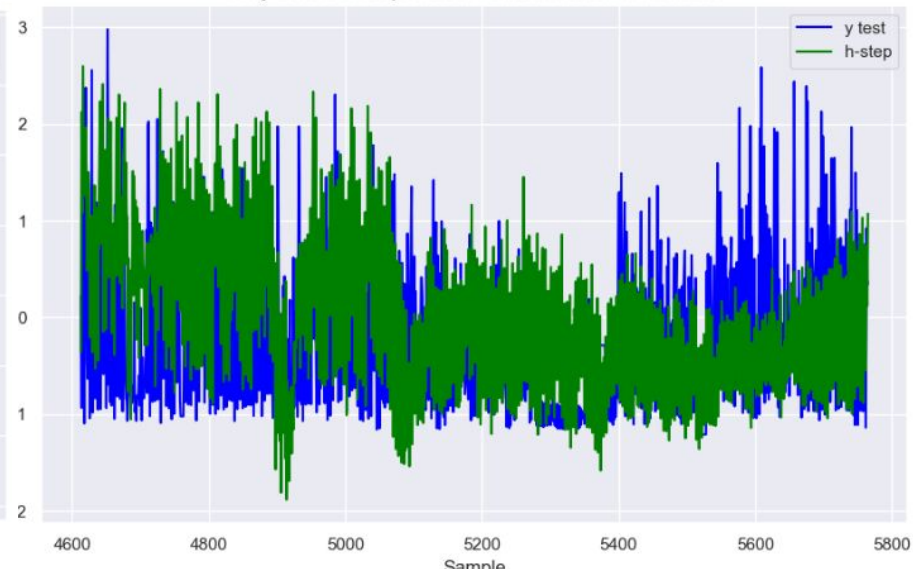
Models	MSE	AIC	BIC
Average	0.6480		
Naive	0.9933		
Drift	0.6838		
SES	0.2008		
Multiple Linear Regression	1.6502		
ARIMA (4,0,4)	0.6032	10392.894	10457.258
SARIMA (4,0,4) x (1,0,1,28)	0.6063	9680.879	9751.680
LSTM	1.5568		

13. Forecast Function & H-step Ahead Prediction

18. y and 1-Step Ahead Prediction with the Final Model



18. y test and h-Step Ahead Prediction with the Final Model



18. h-Step Ahead Prediction with the Final Model



Summary / Conclusion

- aimed to provide accurate predictions of household power consumption using Time Series models
- implemented Time Series models, including Average, Naive, Drift, SES, Multiple Linear Regression, ARMA, SARIMA, and LSTM.
- evaluated the performance of these models using the MSE, AIC, and BIC.
- found that the SARIMA (4,0,4) x (1,0,1,28) model provided the best fit to the data with an MSE of 0.6063.



Thank you!

