

# Metro Traffic Volume Analysis and Prediction

Time Series Analysis and Modelling  
Final Term Project

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A dark blue diagonal gradient bar that starts from the bottom left and extends towards the top right, covering the lower half of the slide.

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# Data Description

```
Int64Index: 40575 entries, 0 to 48203
Data columns (total 9 columns):
#   Column              Non-Null Count  Dtype
---  -
0   holiday             40575 non-null  object
1   temp                40575 non-null  float64
2   rain_1h             40575 non-null  float64
3   snow_1h             40575 non-null  float64
4   clouds_all          40575 non-null  int64
5   weather_main        40575 non-null  object
6   weather_description  40575 non-null  object
7   date_time           40575 non-null  object
8   traffic_volume      40575 non-null  int64
dtypes: float64(3), int64(2), object(4)
memory usage: 3.1+ MB
None
```

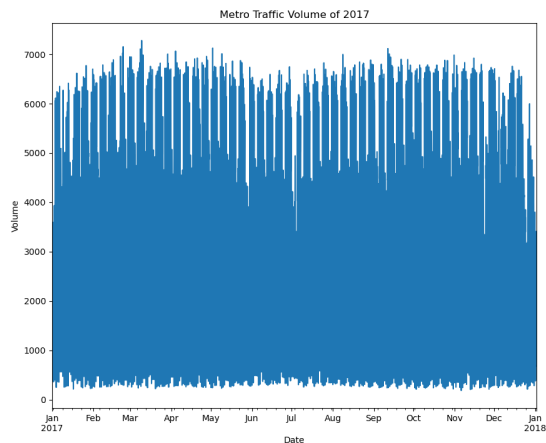
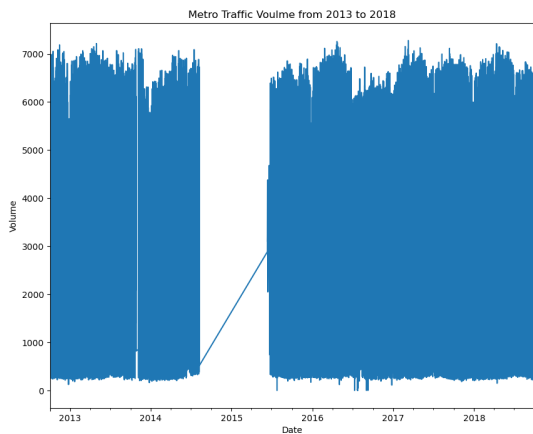
```
In [31]: print(data.head().to_string())
```

	holiday	temp	rain_1h	snow_1h	clouds_all	weather_main	weather_description	date_time	traffic_volume
0	None	288.28	0.0	0.0	40	Clouds	scattered clouds	2012-10-02 09:00:00	5545
1	None	289.36	0.0	0.0	75	Clouds	broken clouds	2012-10-02 10:00:00	4516
2	None	289.58	0.0	0.0	90	Clouds	overcast clouds	2012-10-02 11:00:00	4767
3	None	290.13	0.0	0.0	90	Clouds	overcast clouds	2012-10-02 12:00:00	5026
4	None	291.14	0.0	0.0	75	Clouds	broken clouds	2012-10-02 13:00:00	4918

- Number of Observation: 48203
- Target Variable: traffic\_volume
- Numerical features: temp, rain\_1h, snow\_1h, clouds\_all
- Categorical features: holiday, weather\_main, weather\_description

# Data Description – Missing records

- There are some missing records: Interpolation is used
- Categorical features about Weather : `df.fillna(method='ffill')` method is used
- There are many duplicated or missing observations from 2014-2015 → Decided to take year 2017 only



# Data Description – Final Dataset(2017)

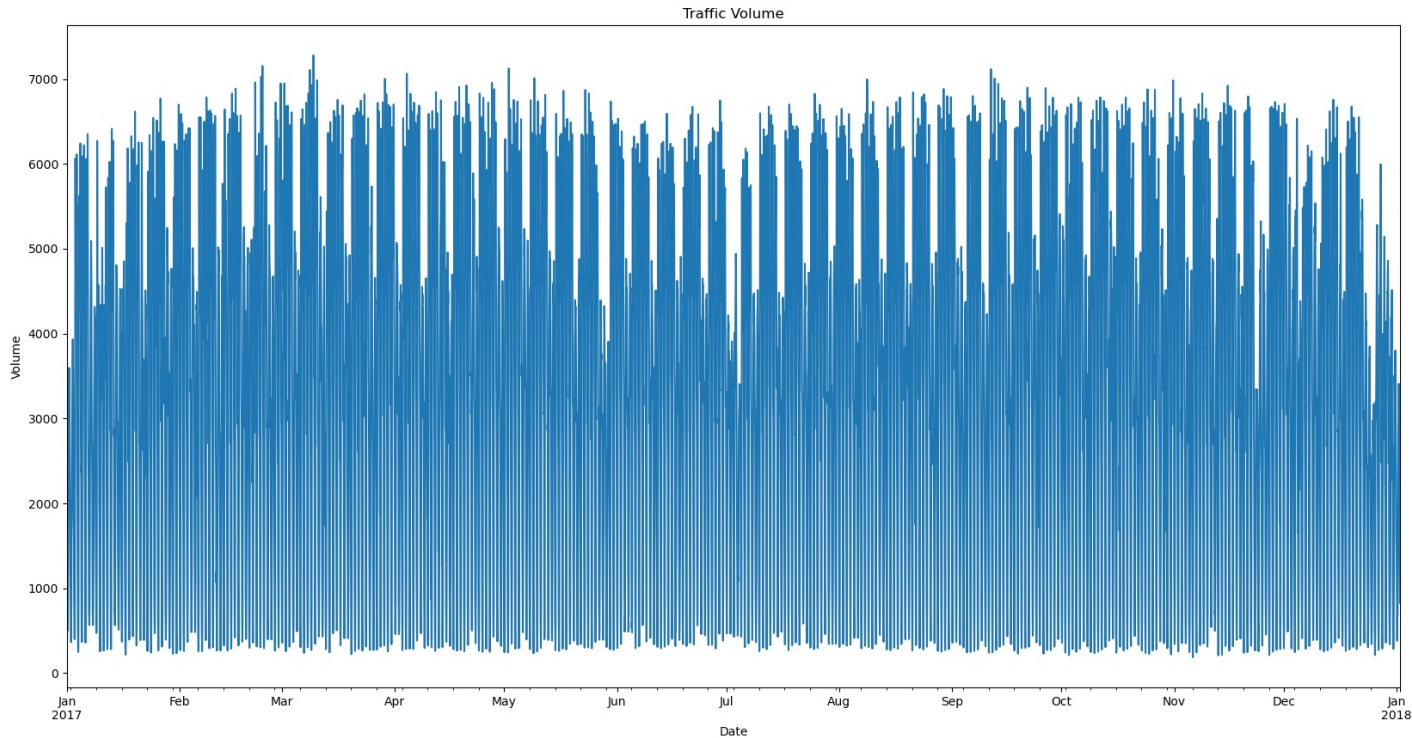
- Number of Observation: 8784
- Statistical description for Numerical features
- Every value in rain\_1h, snow\_1h categories at the year of 2017 is 0

```
In [29]: df.describe()
```

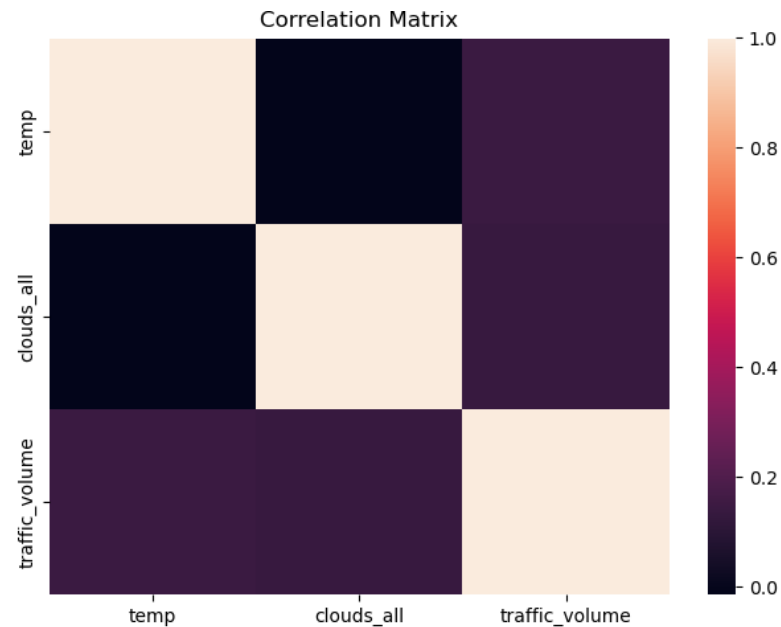
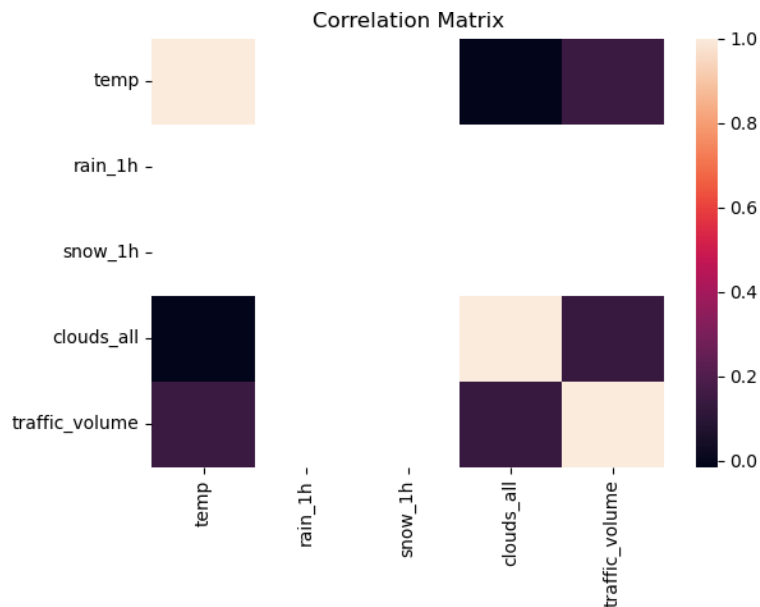
```
Out[29]:
```

	temp	rain_1h	snow_1h	clouds_all	traffic_volume
count	8784.000000	8784.0	8784.0	8784.000000	8784.000000
mean	281.420064	0.0	0.0	43.781933	3369.585781
std	12.085219	0.0	0.0	39.161671	1981.338474
min	246.150000	0.0	0.0	0.000000	186.000000
25%	272.580000	0.0	0.0	1.000000	1351.500000
50%	282.600000	0.0	0.0	40.000000	3563.500000
75%	291.160000	0.0	0.0	90.000000	4990.000000
max	307.020000	0.0	0.0	92.000000	7280.000000

# Data Description – Target Variable



# Data Description – Correlation Matrix



- rain\_1h and snow\_1h is dropped
- High correlation between numerical features and the target variable is not clearly identified
- High correlation within numerical features is not identified

# Data Description – Rolling Mean & Variance



- From the plots of Rolling Mean and Variance, the dataset seems stationary



# Data Description – ADF & KPSS test

```
ADF Statistic: -11.254188
```

```
p-value: 0.000000
```

```
Critical Values:
```

```
1%: -3.431
```

```
5%: -2.862
```

```
10%: -2.567
```

```
Results of KPSS Test:
```

```
Test Statistic      0.243493
```

```
p-value             0.100000
```

```
Lags Used            29.000000
```

```
Critical Value (10%) 0.347000
```

```
Critical Value (5%)  0.463000
```

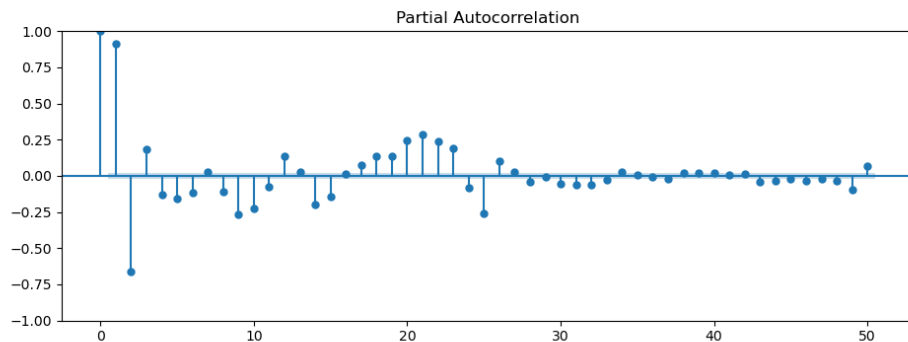
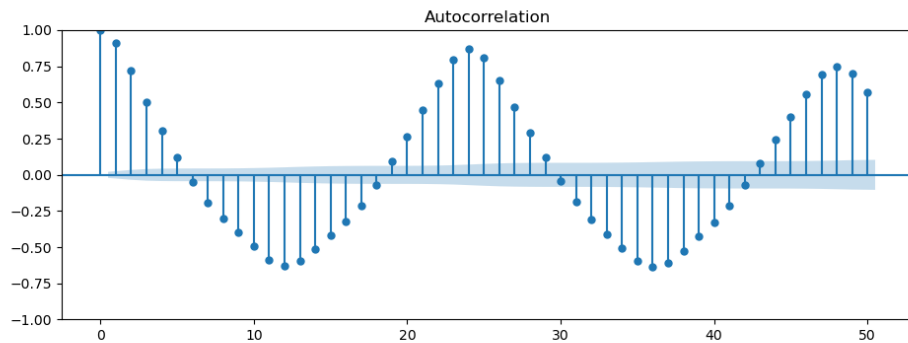
```
Critical Value (2.5%) 0.574000
```

```
Critical Value (1%)  0.739000
```

```
dtype: float64
```

- P-value from ADF is lower than threshold(0.05), and P-value from KPSS is higher
- From the ADF and KPSS test, the dataset is stationary

# Data Description – ACF / PACF



- ACF has oscillating shape
- The peak is 24, 48, ...
- The dataset has a severe seasonality

# Data Description – ACF / PACF from Differencing

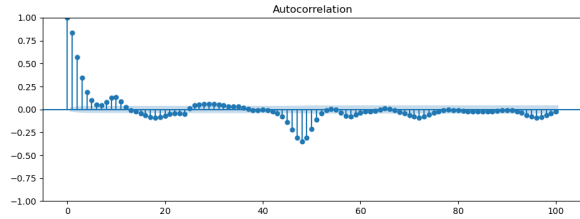


figure 1

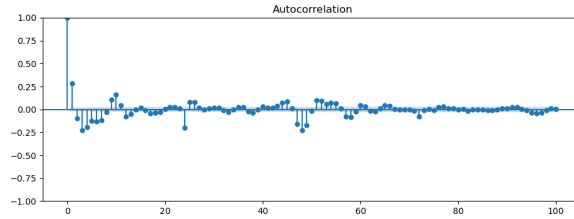


figure 2

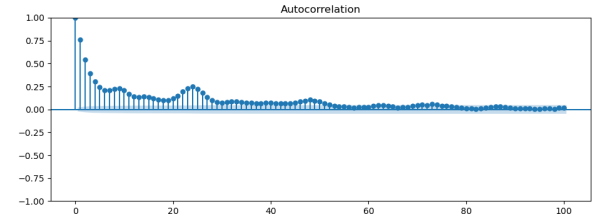
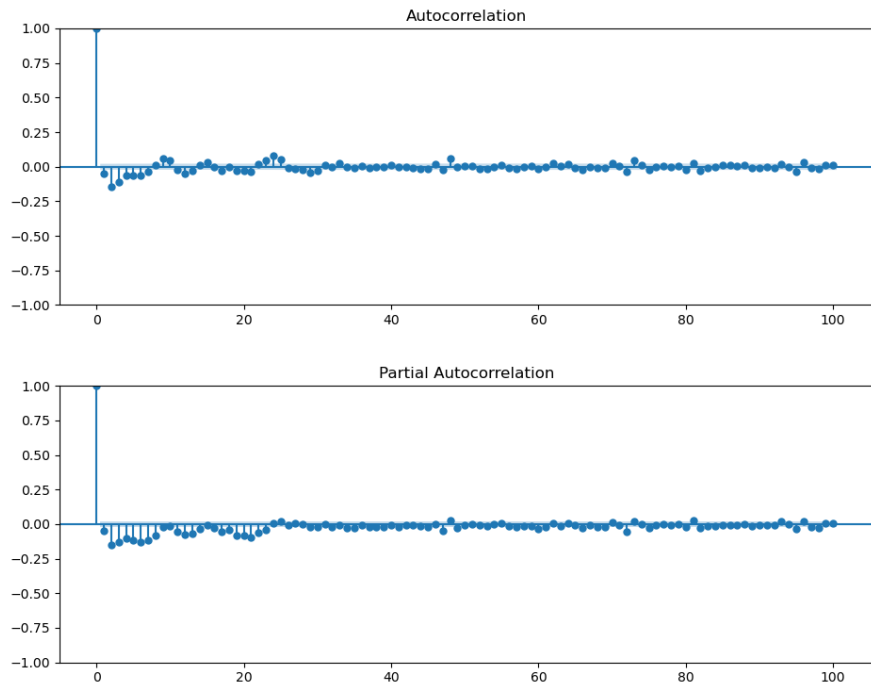


figure 3

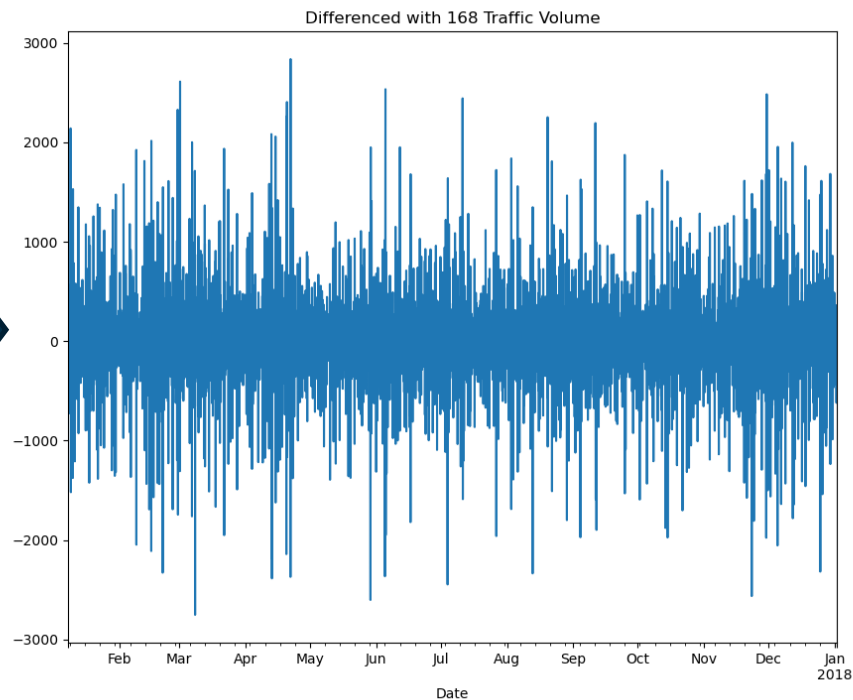
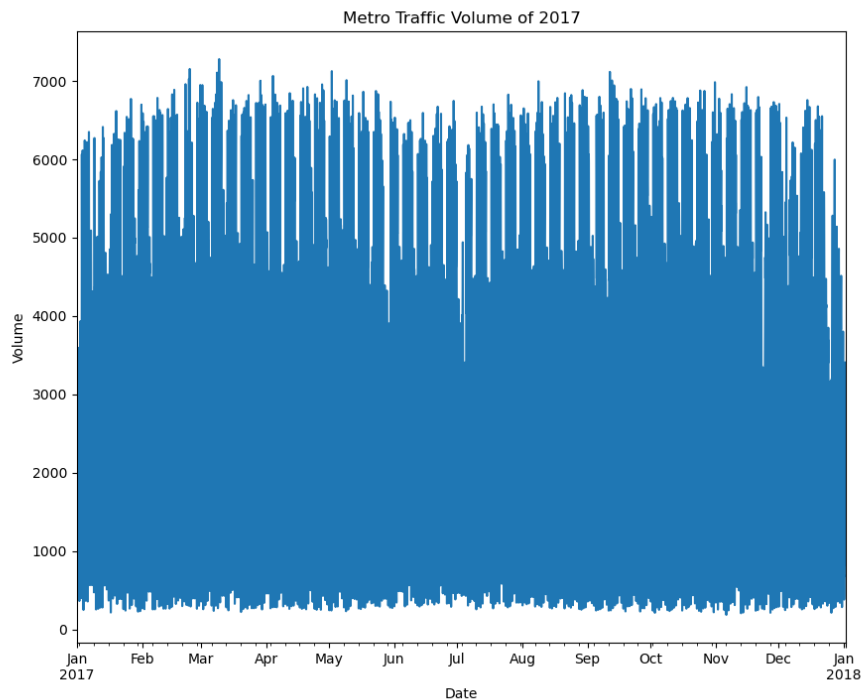
- Figure1 : Seasonal differencing with periods = 24
- Figure2 : Figure1 + First-order non-seasonal differencing
- Figure3: Seasonal differencing with periods = 168 (weekly seasonality)

# Data Description – ACF / PACF from Differencing

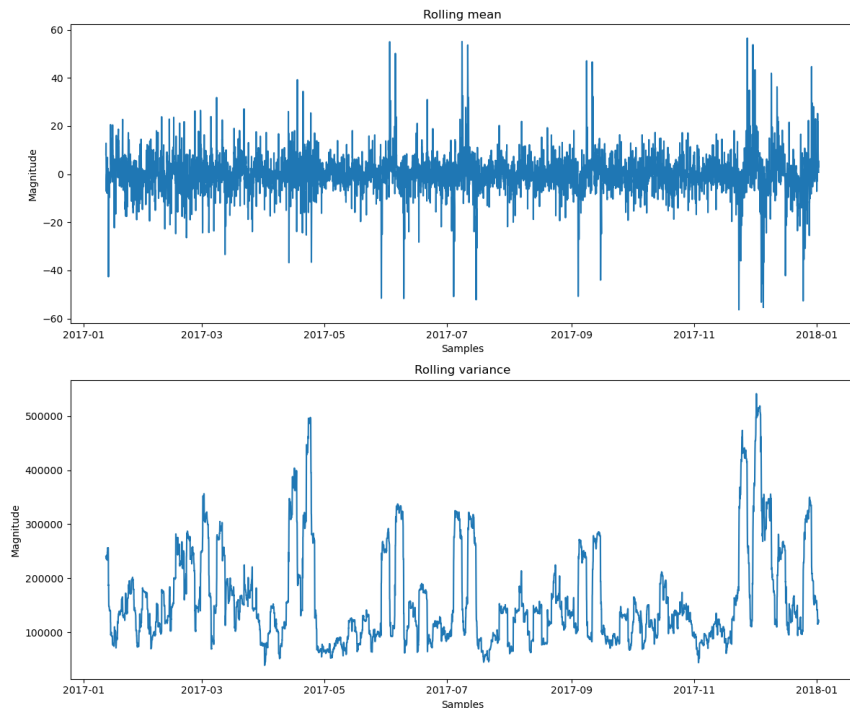


- Seasonal differencing with period = 168
- Non-seasonal first order differencing is followed
- Based on ACF, seasonal differencing with period of 168 followed by first order non-seasonal differencing is used to make dataset stationary

# Data Description – Data after Differencing



# Data Description – Data after Differencing



ADF Statistic: -22.576452

p-value: 0.000000

Critical Values:

1%: -3.431

5%: -2.862

10%: -2.567

ADF for the transformed dataset: None

Results of KPSS Test:

Test Statistic 0.04072

p-value 0.10000

Lags Used 380.00000

Critical Value (10%) 0.34700

Critical Value (5%) 0.46300

Critical Value (2.5%) 0.57400

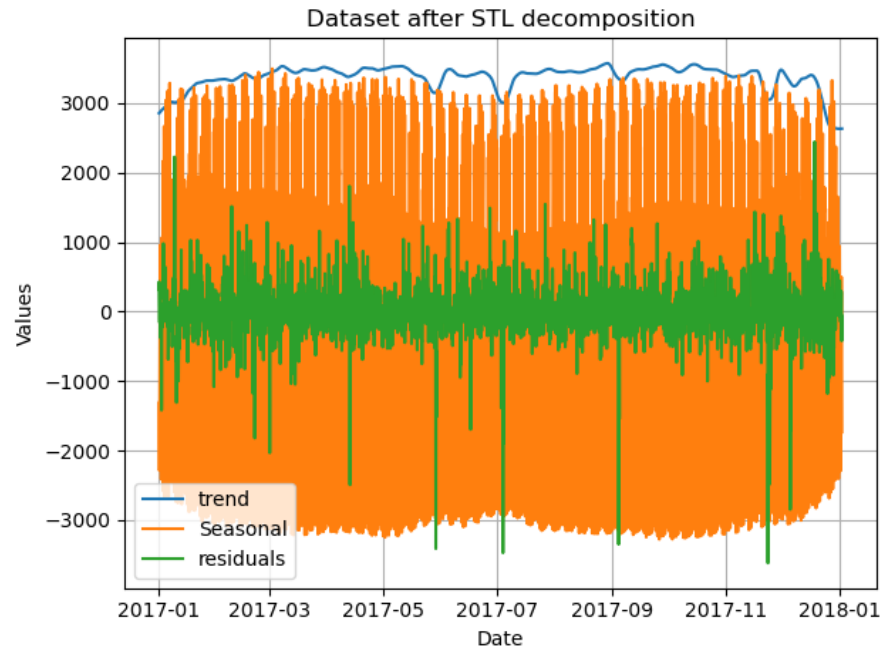
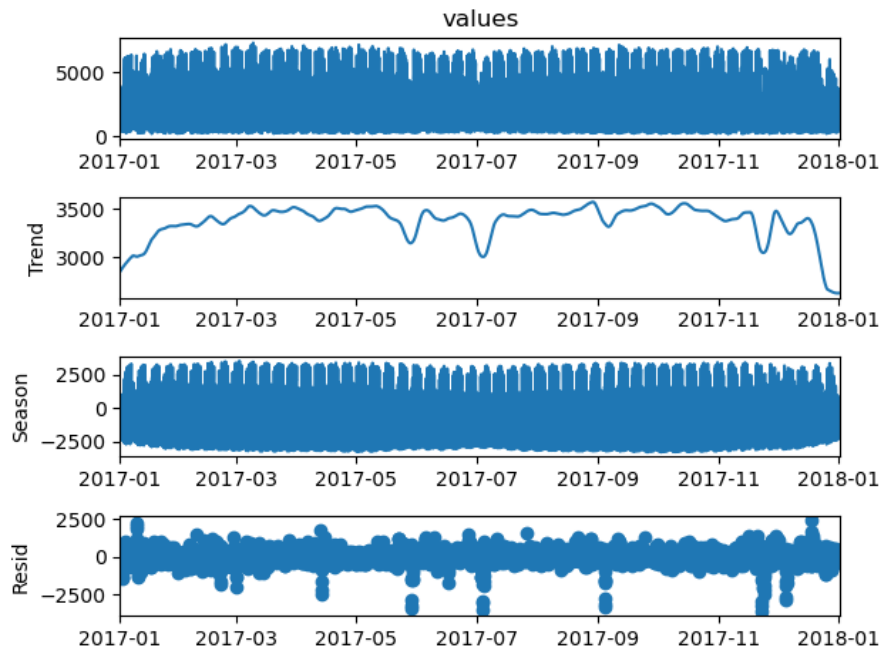
Critical Value (1%) 0.73900

dtype: float64

KPSS for the transformed dataset: None

- Data after Differencing is also stationary

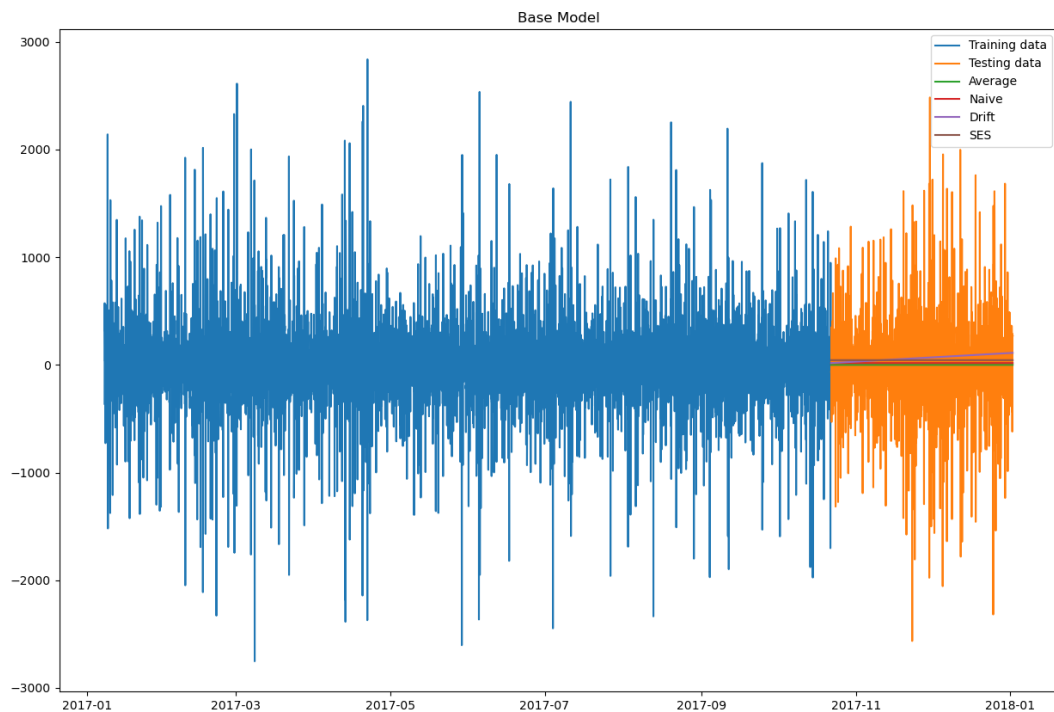
# Data Description – STL Decomposition



The strength of trend for this data set is 0.29159030625136384

The strength of seasonality for this data set is 0.9762908411024094

# Model – Base Models

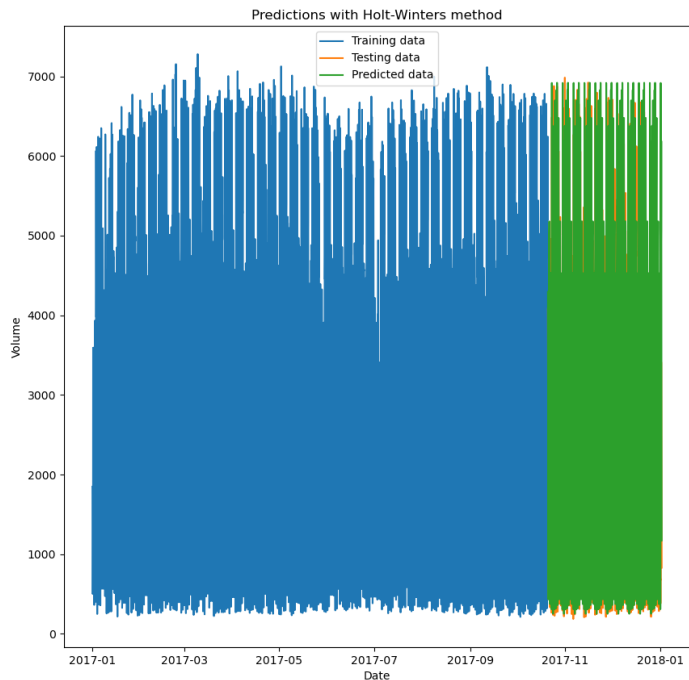


## Best Base Model based on RMSE – **Average Model**

- RMSE for Average Model: 434.09354258940544
- RMSE for Naive Model: 434.43174536549657
- RMSE for Drift Model: 439.732105352747
- RMSE for SES Model: 436.33900696177295



# Model – Holt-Winters Method

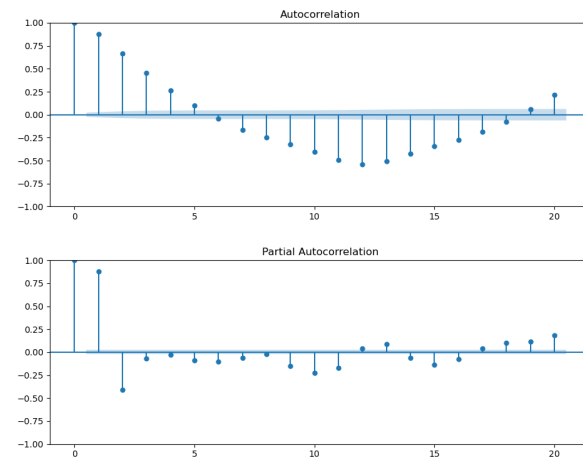


- seasonal='mul'
- seasonal\_periods=168
- RMSE for Holt-Winters Model: 756.0131117149048

# Model – OLS : Full Model

OLS Regression Results							x19	-3.76e+14	5.7e+14	-0.659	0.510	-1.49e+15	7.42e+14
=====							x20	3.388e+14	5.14e+14	0.659	0.510	-6.69e+14	1.35e+15
Dep. Variable:	traffic_volume	R-squared:	0.072				x21	1.145e+15	1.74e+15	0.659	0.510	-2.26e+15	4.55e+15
Model:	OLS	Adj. R-squared:	0.067				x22	-3.887e+14	5.9e+14	-0.659	0.510	-1.54e+15	7.67e+14
Method:	Least Squares	F-statistic:	13.88				x23	-1.26e+14	1.91e+14	-0.659	0.510	-5.01e+14	2.49e+14
Date:	Tue, 09 May 2023	Prob (F-statistic):	6.32e-86				x24	-2.137e+14	3.24e+14	-0.659	0.510	-8.49e+14	4.22e+14
Time:	18:22:23	Log-Likelihood:	-63069.				x25	1.502e+14	2.28e+14	0.659	0.510	-2.96e+14	5.97e+14
No. Observations:	7027	AIC:	1.262e+05				x26	2.913e+14	4.42e+14	0.659	0.510	-5.75e+14	1.16e+15
Df Residuals:	6987	BIC:	1.265e+05				x27	-4.47e+13	6.78e+13	-0.659	0.510	-1.78e+14	8.82e+13
Df Model:	39						x28	1.212e+14	1.84e+14	0.659	0.510	-2.39e+14	4.82e+14
Covariance Type:	nonrobust						x29	-5.198e+13	7.89e+13	-0.659	0.510	-2.07e+14	1.03e+14
=====							x30	-1.767e+14	2.68e+14	-0.659	0.510	-7.02e+14	3.49e+14
	coef	std err	t	P> t	[0.025	0.975]	x31	2.436e+13	3.7e+13	0.659	0.510	-4.81e+13	9.68e+13
-----							x32	2.748e+14	4.17e+14	0.659	0.510	-5.43e+14	1.09e+15
const	3414.2349	29.159	117.092	0.000	3357.075	3471.395	x33	-3.288e+13	4.99e+13	-0.659	0.510	-1.31e+14	6.49e+13
x1	225.8202	25.094	8.999	0.000	176.628	275.013	x34	-0.0549 <td>0.083</td> <td>-0.659</td>	0.083	-0.659	0.510	-0.218 <td>0.108</td>	0.108
x2	421.6712	48.431	8.707	0.000	326.733	516.610	x35	-3.162e+14	4.8e+14	-0.659	0.510	-1.26e+15	6.24e+14
x3	-1.231e+15	1.87e+15	-0.659	0.510	-4.89e+15	2.43e+15	x36 <td>3.396e+14</td> <td>5.15e+14</td> <td>0.659</td> <th>0.510</th> <th>-6.71e+14</th> <td>1.35e+15</td>	3.396e+14	5.15e+14	0.659	0.510	-6.71e+14	1.35e+15
x4	-1.231e+15	1.87e+15	-0.659	0.510	-4.89e+15	2.43e+15	x37 <td>2.267e+14</td> <td>3.44e+14</td> <td>0.659</td> <th>0.510</th> <th>-4.47e+14</th> <td>9.01e+14</td>	2.267e+14	3.44e+14	0.659	0.510	-4.47e+14	9.01e+14
x5	-1.231e+15	1.87e+15	-0.659	0.510	-4.89e+15	2.43e+15	x38 <td>-3.936e+14</td> <td>5.97e+14</td> <td>-0.659</td> <th>0.510</th> <th>-1.56e+15</th> <td>7.77e+14</td>	-3.936e+14	5.97e+14	-0.659	0.510	-1.56e+15	7.77e+14
x6	-1.231e+15	1.87e+15	-0.659	0.510	-4.89e+15	2.43e+15	x39 <td>1.03e+14</td> <td>1.56e+14</td> <td>0.659</td> <th>0.510</th> <th>-2.03e+14</th> <td>4.09e+14</td>	1.03e+14	1.56e+14	0.659	0.510	-2.03e+14	4.09e+14
x7	-1.231e+15	1.87e+15	-0.659	0.510	-4.89e+15	2.43e+15	x40 <td>-1.005e+15</td> <td>1.53e+15</td> <td>-0.659</td> <th>0.510</th> <th>-3.99e+15</th> <td>1.98e+15</td>	-1.005e+15	1.53e+15	-0.659	0.510	-3.99e+15	1.98e+15
x8	-1.231e+15	1.87e+15	-0.659	0.510	-4.89e+15	2.43e+15	x41 <td>-9.071e+13</td> <td>1.38e+14</td> <td>-0.659</td> <th>0.510</th> <th>-3.61e+14</th> <td>1.79e+14</td>	-9.071e+13	1.38e+14	-0.659	0.510	-3.61e+14	1.79e+14
x9	-3.481e+15	5.28e+15	-0.659	0.510	-1.38e+16	6.87e+15	x42 <td>-2.028e+14</td> <td>3.08e+14</td> <td>-0.659</td> <th>0.510</th> <th>-8.06e+14</th> <td>4e+14</td>	-2.028e+14	3.08e+14	-0.659	0.510	-8.06e+14	4e+14
x10	-1.231e+15	1.87e+15	-0.659	0.510	-4.89e+15	2.43e+15	x43 <td>-3.899e+14</td> <td>5.92e+14</td> <td>-0.659</td> <th>0.510</th> <th>-1.55e+15</th> <td>7.7e+14</td>	-3.899e+14	5.92e+14	-0.659	0.510	-1.55e+15	7.7e+14
x11	-1.294e+14	1.96e+14	-0.659	0.510	-5.14e+14	2.55e+14	x44 <td>-113.2027</td> <td>99.089</td> <td>-1.142</td> <td>0.253</td> <td>-307.447</td> <td>81.041</td>	-113.2027	99.089	-1.142	0.253	-307.447	81.041
x12	1.399e+13	2.12e+13	0.659	0.510	-2.76e+13	5.56e+13	x45 <td>-2.325e+13</td> <td>3.53e+13</td> <td>-0.659</td> <th>0.510</th> <td>-9.24e+13</td> <td>4.59e+13</td>	-2.325e+13	3.53e+13	-0.659	0.510	-9.24e+13	4.59e+13
x13	-1.231e+15	1.87e+15	-0.659	0.510	-4.89e+15	2.43e+15	x46 <td>-1.064e+14</td> <td>1.61e+14</td> <td>-0.659</td> <th>0.510</th> <td>-4.23e+14</td> <td>2.1e+14</td>	-1.064e+14	1.61e+14	-0.659	0.510	-4.23e+14	2.1e+14
x14	6.064e+14	9.2e+14	0.659	0.510	-1.2e+15	2.41e+15	x47 <td>-4.151e+14</td> <td>6.3e+14</td> <td>-0.659</td> <th>0.510</th> <td>-1.65e+15</td> <td>8.2e+14</td>	-4.151e+14	6.3e+14	-0.659	0.510	-1.65e+15	8.2e+14
x15	2.205e+14	3.35e+14	0.659	0.510	-4.35e+14	8.76e+14	x48 <td>-1.814e+14</td> <td>2.75e+14</td> <td>-0.659</td> <th>0.510</th> <td>-7.21e+14</td> <td>3.58e+14</td>	-1.814e+14	2.75e+14	-0.659	0.510	-7.21e+14	3.58e+14
x16	-1.502e+14	2.28e+14	-0.659	0.510	-5.97e+14	2.96e+14	x49 <td>-1.283e+14</td> <td>1.95e+14</td> <td>-0.659</td> <th>0.510</th> <td>-5.1e+14</td> <td>2.53e+14</td>	-1.283e+14	1.95e+14	-0.659	0.510	-5.1e+14	2.53e+14
x17	-2.913e+14	4.42e+14	-0.659	0.510	-1.16e+15	5.75e+14	x50 <td>-1.814e+14</td> <td>2.75e+14</td> <td>-0.659</td> <th>0.510</th> <td>-7.21e+14</td> <td>3.58e+14</td>	-1.814e+14	2.75e+14	-0.659	0.510	-7.21e+14	3.58e+14
x18	-3.396e+14	5.15e+14	-0.659	0.510	-1.35e+15	6.71e+14	x51 <td>-9.071e+13</td> <td>1.38e+14</td> <td>-0.659</td> <th>0.510</th> <td>-3.61e+14</td> <td>1.79e+14</td>	-9.071e+13	1.38e+14	-0.659	0.510	-3.61e+14	1.79e+14

- P-values for features except for first 2 are higher than 0.510



# Model – OLS : Feature Selection

- Backward stepwise feature selection is used for feature selection

	AIC	BIC	Adj R <sup>2</sup>	Prev. BIC	Prev. Adj R <sup>2</sup>
24	126202.382899	126387.535808	0.067068	126394.602114	0.067173
25	126202.392573	126380.687967	0.066935	126387.535808	0.067068
23	126202.591689	126394.602114	0.067173	126401.832697	0.067255
26	126202.732032	126374.169911	0.066757	126380.687967	0.066935
22	126202.964758	126401.832697	0.067255	126409.187781	0.067321
21	126203.462326	126409.187781	0.067321	126416.597833	0.067380
27	126203.663992	126368.244356	0.066501	126374.169911	0.066757
20	126204.014863	126416.597833	0.067380	126424.211091	0.067412
19	126204.770606	126424.211091	0.067412	126431.907112	0.067433
28	126205.344846	126363.067695	0.066145	126368.244356	0.066501
18	126205.609112	126431.907112	0.067433	126431.907112	0.067433

- Features to keep: ['temp', 'clouds\_all', 'weather\_main\_Clouds', 'weather\_description\_broken clouds', 'weather\_description\_drizzle', 'weather\_description\_few clouds', 'weather\_description\_heavy intensity drizzle', 'weather\_description\_heavy intensity rain', 'weather\_description\_light intensity drizzle', 'weather\_description\_light rain', 'weather\_description\_light rain and snow', 'weather\_description\_light snow', 'weather\_description\_mist', 'weather\_description\_moderate rain', 'weather\_description\_proximity thunderstorm', 'weather\_description\_proximity thunderstorm with rain', 'weather\_description\_scattered clouds', 'weather\_description\_sky is clear', 'weather\_description\_snow', 'weather\_description\_thunderstorm', 'weather\_description\_thunderstorm with light drizzle']
- Features to eliminate: ['weather\_description\_sleet', 'weather\_main\_Mist', 'weather\_description\_heavy snow', 'weather\_main\_Snow', 'holiday\_Veterans Day', 'weather\_main\_Fog', 'weather\_description\_fog', 'weather\_description\_overcast clouds', 'weather\_main\_Drizzle', 'weather\_description\_haze', 'weather\_main\_Haze', 'weather\_description\_proximity thunderstorm with drizzle', 'weather\_description\_proximity shower rain', 'weather\_main\_Rain', 'holiday\_None', 'weather\_description\_light shower snow', 'holiday\_Martin Luther King Jr Day', 'weather\_description\_light intensity shower rain', 'holiday\_Thanksgiving Day', 'holiday\_Independence Day', 'weather\_main\_Thunderstorm', 'holiday\_Labor Day', 'holiday\_New Years Day', 'holiday\_Memorial Day', 'holiday\_Columbus Day', 'weather\_description\_thunderstorm with rain', 'holiday\_State Fair', 'holiday\_Washingtons Birthday', 'weather\_description\_thunderstorm with light rain', 'weather\_description\_thunderstorm with heavy rain']

# Model – OLS: Multi-collinearity

	feature	VIF
0	temp	1.172962
1	clouds_all	4.008464
2	weather_main_Clouds	8.132498
3	weather_description_broken clouds	1.878463
4	weather_description_drizzle	1.172341
5	weather_description_few clouds	1.610834
6	weather_description_heavy intensity drizzle	1.020216
7	weather_description_heavy intensity rain	1.252323
8	weather_description_light intensity drizzle	1.297529
9	weather_description_light rain	2.185851
10	weather_description_light rain and snow	1.006923
11	weather_description_light snow	1.628223
12	weather_description_mist	2.539087
13	weather_description_moderate rain	1.883428
14	weather_description_proximity thunderstorm	1.394541
15	weather_description_proximity thunderstorm wit...	1.014823
16	weather_description_scattered clouds	2.448045
17	weather_description_sky is clear	6.054076
18	weather_description_snow	1.078326
19	weather_description_thunderstorm	1.065502
20	weather_description_thunderstorm with light dr...	1.007505

```
SingularValues = [2.36529393e+04 2.06674058e+04 2.00305844e+04 1.89750342e+04
1.80837385e+04 1.79407737e+04 1.78695095e+04 1.76470886e+04
1.71115157e+04 1.09917415e+04 9.95726749e+03 9.30202604e+03
9.18585138e+03 8.93128105e+03 8.92387914e+03 8.87402868e+03
8.85318205e+03 8.82558169e+03 8.80142135e+03 8.79675880e+03
8.79407464e+03 8.79381916e+03 8.78933437e+03 8.78870821e+03
8.78806275e+03 8.78787631e+03 8.78643407e+03 8.78626963e+03
8.78523799e+03 8.78516166e+03 8.78501698e+03 8.78500011e+03
8.78500011e+03 8.78500011e+03 8.76670150e+03 8.74388789e+03
8.71415017e+03 8.69908225e+03 8.60062439e+03 6.22119122e+03
1.25070454e+03 3.74014989e+02 1.67038152e+02 3.30725782e-11
2.92202838e-11 1.54935228e-11 5.45708318e-12 1.81533793e-12
1.81533793e-12 1.81533793e-12 1.62549521e-12]
Condition Number= 3614978705674538.5
```

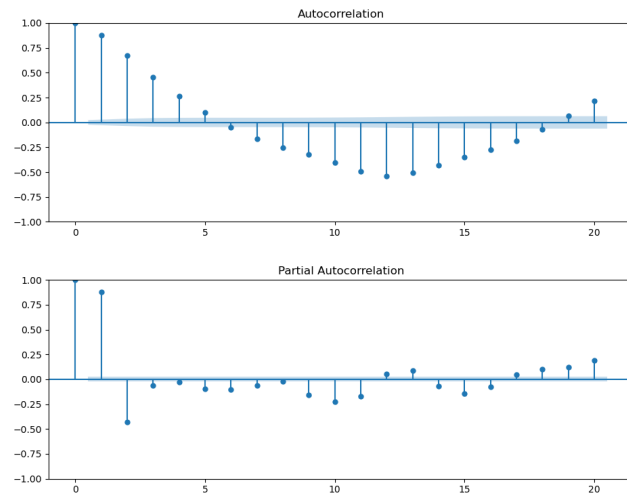
- There is no multi-collinearity in features

# Model – OLS: Final Model

```
=====
                        OLS Regression Results
=====
Dep. Variable:          traffic_volume      R-squared:                0.069
Model:                  OLS                Adj. R-squared:         0.066
Method:                 Least Squares      F-statistic:              24.57
Date:                   Tue, 09 May 2023    Prob (F-statistic):       3.11e-92
Time:                   18:43:18           Log-Likelihood:          -63081.
No. Observations:       7027              AIC:                    1.262e+05
Df Residuals:           7005              BIC:                    1.264e+05
Df Model:               21
Covariance Type:        nonrobust
=====
                        coef    std err          t      Pr>|t|    [0.025    0.975]
-----
const                3402.3239     22.894    148.411     0.000    3357.444    3447.203
temp                 222.2303     24.795     8.963     0.000    173.624    270.836
clouds_all            419.5406     45.837     9.153     0.000    329.686    509.395
weather_main_Clouds   -299.1868     65.289    -4.583     0.000   -427.172   -171.201
weather_description_broken clouds  137.5994     31.378     4.385     0.000     76.089    199.110
weather_description_drizzle -103.0641     24.789    -4.158     0.000   -151.657   -54.471
weather_description_few clouds   202.3416     29.057     6.964     0.000    145.381    259.302
weather_description_heavy intensity drizzle -45.3816     23.124    -1.962     0.050   -90.713    -0.051
weather_description_heavy intensity rain -114.8374     25.620    -4.482     0.000   -165.061   -64.614
weather_description_light intensity drizzle -109.2862     26.079    -4.191     0.000   -160.408   -58.164
weather_description_light rain   -195.1105     33.848    -5.764     0.000   -261.463   -128.758
weather_description_light rain and snow -55.5572     22.973    -2.418     0.016   -100.592   -10.523
weather_description_light snow   -147.5406     29.213    -5.050     0.000   -204.808   -90.273
weather_description_mist        -255.0847     36.481    -6.992     0.000   -326.598   -183.571
weather_description_moderate rain -164.2489     31.420    -5.228     0.000   -225.841   -102.657
weather_description_proximity thunderstorm -160.4352     27.036    -5.934     0.000   -213.434   -107.437
weather_description_proximity thunderstorm with rain -46.5996     23.063    -2.021     0.043   -91.811   -1.389
weather_description_scattered clouds  246.9102     35.821     6.893     0.000    176.691    317.130
weather_description_sky is clear  -115.5526     56.331    -2.051     0.040   -225.979   -5.126
weather_description_snow         -64.7941     23.774    -2.725     0.006   -111.398   -18.190
weather_description_thunderstorm   -82.7141     23.632    -3.500     0.000   -129.040   -36.388
weather_description_thunderstorm with light drizzle -50.5480     22.980    -2.200     0.028   -95.596   -5.500
=====
Omnibus:                2494.971    Durbin-Watson:           0.243
Prob(Omnibus):           0.000      Jarque-Bera (JB):        344.360
Skew:                    -0.060      Prob(JB):                1.67e-75
Kurtosis:                1.922      Cond. No.                6.41
=====

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

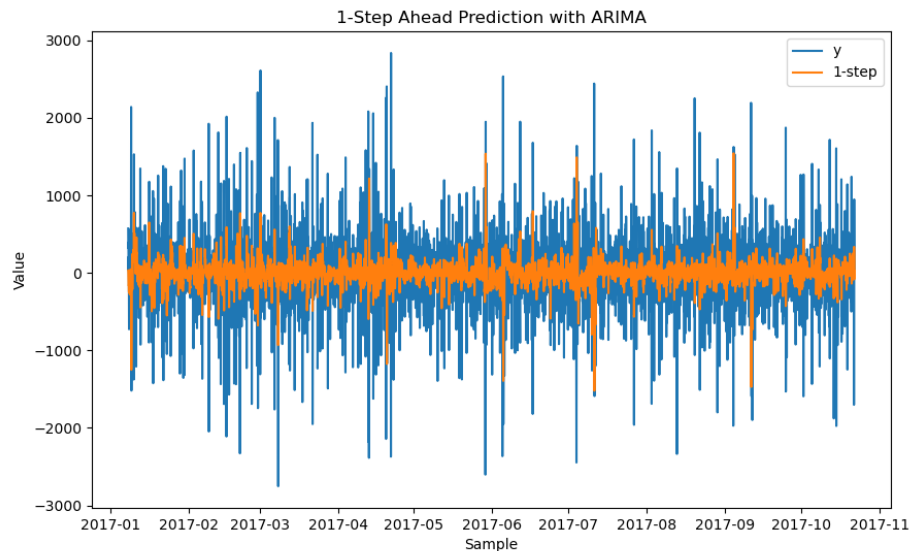
- Mean of residual\_errors: -2.1288414479523444e-12
- Mean of forecast\_errors: 70.6160590398032
- RMSE: 1883.637898288714
- The residual is NOT white
- $R^2$ : 0.069



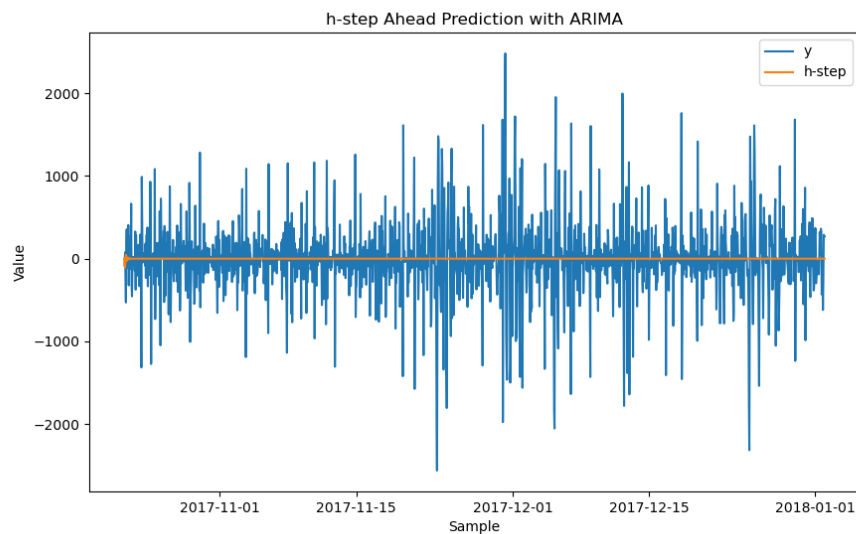
# Model – ARIMA(2, 0, 6)

```
SARIMAX Results
=====
Dep. Variable:    traffic_volume    No. Observations:   6892
Model:            ARIMA(2, 0, 6)    Log Likelihood      -50323.442
Date:             Tue, 09 May 2023  AIC                          100666.884
Time:             22:54:50          BIC                          100735.265
Sample:           01-08-2017        HQIC                         100690.465
- 10-22-2017
Covariance Type:  opg
=====
              coef    std err          z      P>|z|    [0.025    0.975]
-----
const         -0.1640     0.365     -0.449    0.653     -0.880     0.552
ar.L1          0.8375     0.097      8.606    0.000     0.647     1.028
ar.L2         -0.5364     0.057     -9.437    0.000     -0.648     -0.425
ma.L1         -1.0759     0.098    -11.018    0.000     -1.267     -0.884
ma.L2          0.4731     0.078      6.072    0.000     0.320     0.626
ma.L3         -0.0726     0.028     -2.635    0.008     -0.127     -0.019
ma.L4         -0.0918     0.026     -3.566    0.000     -0.142     -0.041
ma.L5         -0.0940     0.020     -4.770    0.000     -0.133     -0.055
ma.L6         -0.0823     0.018     -4.564    0.000     -0.118     -0.047
sigma2        1.321e+05   999.809   132.170    0.000    1.3e+05    1.34e+05
=====
Ljung-Box (L1) (Q):      0.01    Jarque-Bera (JB):      35741.09
Prob(Q):                 0.92    Prob(JB):              0.00
Heteroskedasticity (H):   0.78    Skew:                  0.01
Prob(H) (two-sided):      0.00    Kurtosis:              14.16
=====
```

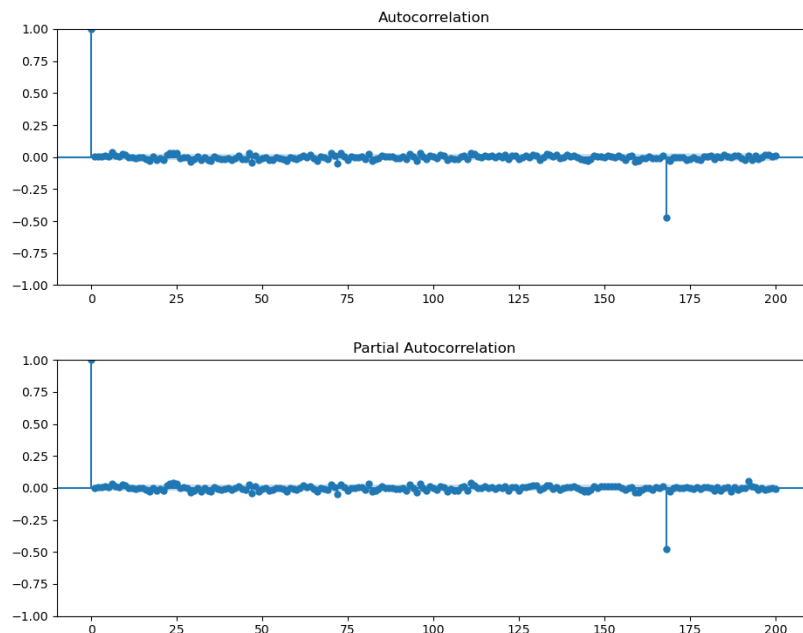
- P-values from the estimated coefficients are less than 0.05
- RMSE for ARIMA: 434.1142080430148



# Model – ACF of Residual from ARIMA(2, 0, 6)



- ACF/PACF of residual has peak at lag=168  
→ SARIMA is needed



# Model – SARIMA(0, 0, 1, 168)

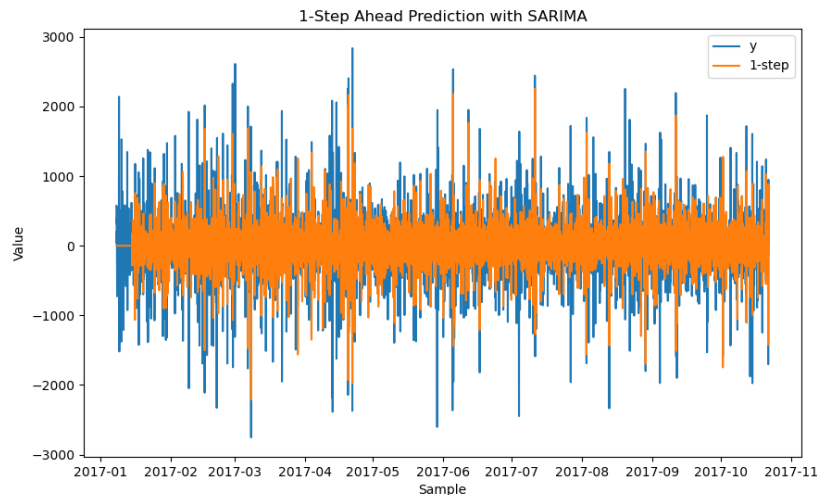
```
SARIMAX Results
=====
Dep. Variable:          traffic_volume    No. Observations:             6892
Model:                 SARIMAX(0, 0, [1], 168)    Log Likelihood              -49162.934
Date:                 Tue, 09 May 2023    AIC                        98329.867
Time:                 15:12:25    BIC                        98343.543
Sample:              01-08-2017    HQIC                       98334.583
                  - 10-22-2017

Covariance Type:          opg
=====
              coef    std err          z      P>|z|      [0.025    0.975]
-----
ma.S.L168      -0.9007      0.006   -154.273     0.000     -0.912     -0.889
sigma2         8.827e+04    631.707    139.732     0.000     8.7e+04    8.95e+04
=====

Ljung-Box (L1) (Q):           47.08    Jarque-Bera (JB):           31726.57
Prob(Q):                     0.00    Prob(JB):                   0.00
Heteroskedasticity (H):       0.71    Skew:                      -0.17
Prob(H) (two-sided):         0.00    Kurtosis:                   13.51
=====

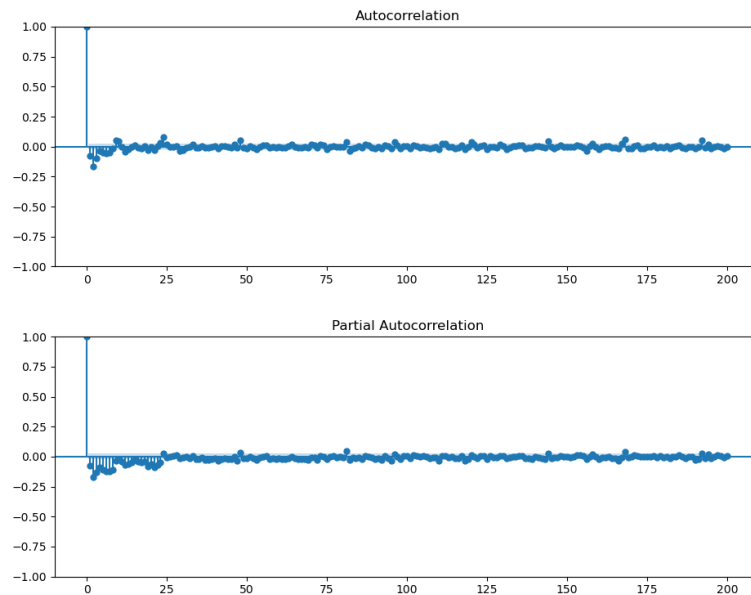
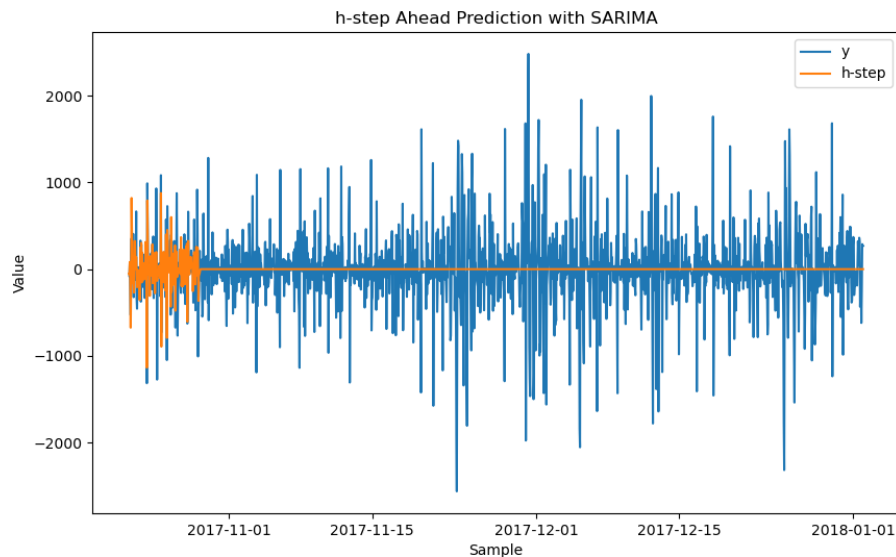
Warnings:
[1] Covariance matrix calculated using the outer product of gradients (complex-step).
RUNNING THE L-BFGS-B CODE
```

- P-value from the estimated coefficient is less than 0.05
- RMSE for SARIMA: 427.19046452962493





# Model – SARIMA(0, 0, 1, 168)



- There is no peak at 168 in ACF/PACF of residuals
- Mean of Residual: 0.36812733693893085
- Variance of Residual: 92899.74806144746

# Final Model Selection – ACF/PACF of Residuals

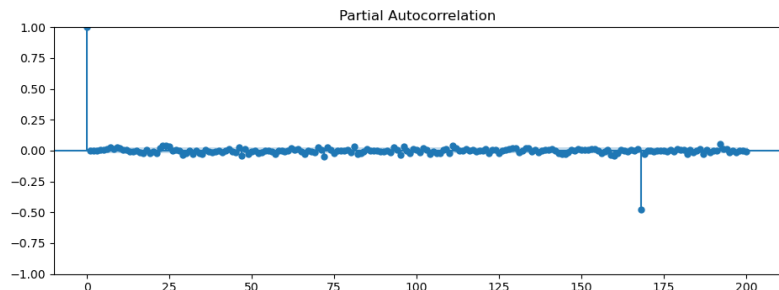
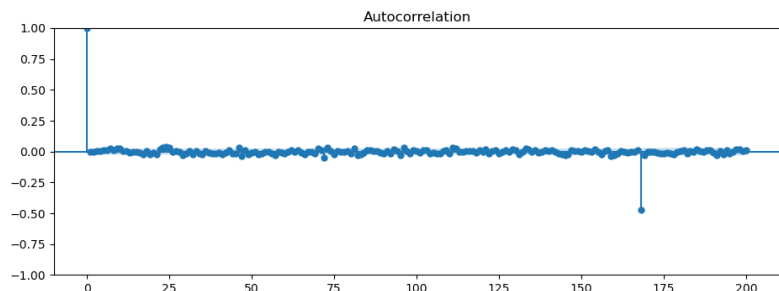


figure 1

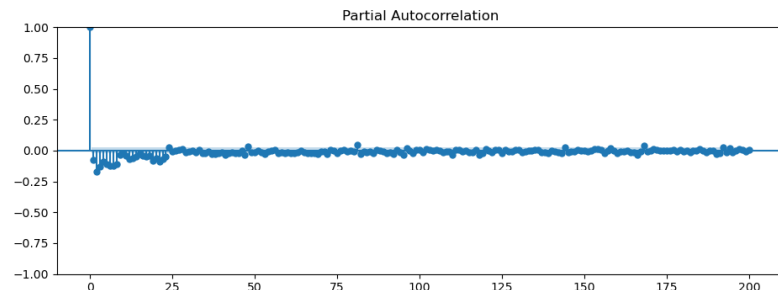
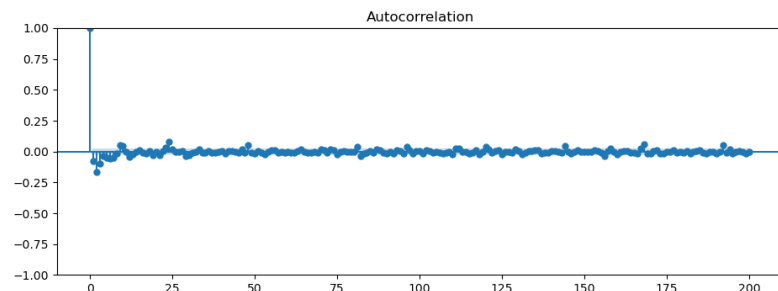


figure 2

- Figure1 : ACF/PACF of residual from ARIMA(2,0,7) model
- Figure2 : ACF/PACF of residual from SARIMA(0, 0, 1, 168) model

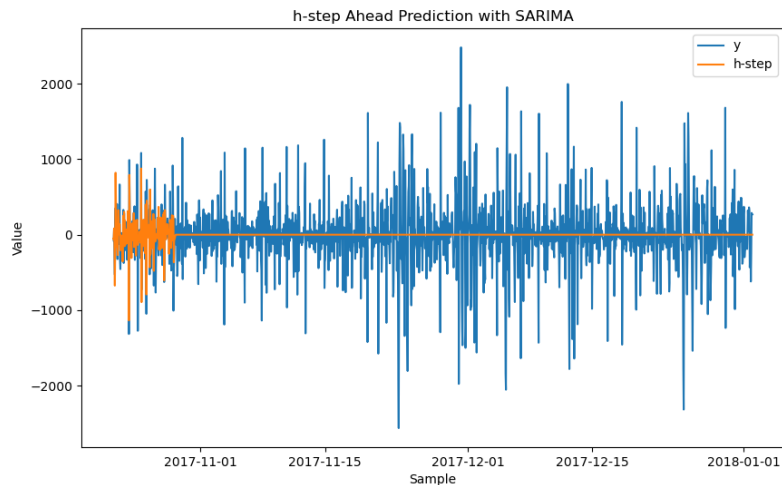
# Final Model Selection

- RMSE for Average Model: 434.09354258940544
- RMSE for ARIMA: 434.13764539656626
- RMSE for SARIMA: 427.19046452962493

→ **SARIMA(0, 0, 1, 168)** is selected as a final model

# Conclusion

- Based on RMSE and ACF/PACF of residual, **SARIMA(0, 0, 1, 168)** is selected as a final model
- RMSE: 427
- However, The residual is NOT white (by Ljung-Box test)
- Therefore, there is a possibility of better model



Thank You