Predicting Metro Interstate Traffic Volume: A Time Series Forecasting Approach with ARIMA

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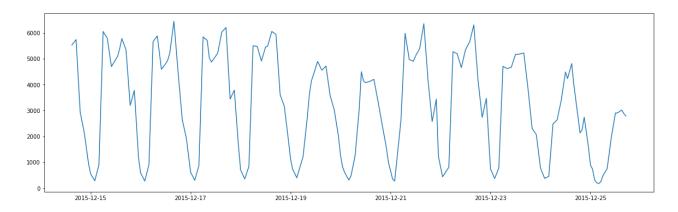
Department of Data Science, Bellevue University DSC680: Applied Data Science Dr. Brett Werner Spring 2023

Dataset: https://archive.ics.uci.edu/ml/datasets/Metro+Interstate+Traffic+Volume

Import and View Data

```
In [1]:
          # Libraries
          import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          from statsmodels.graphics.tsaplots import plot acf, plot pacf
In [2]:
          # import dataset and preview
          df = pd.read_csv("Potential_Datasets/Metro_Interstate_Traffic_Volume.csv.gz", compressi
          df.head()
                     temp rain_1h snow_1h clouds_all weather_main weather_description date_time traffic_v
Out[2]:
            holiday
                                                                                           2012-10-
              None 288.28
                                0.0
                                         0.0
                                                    40
                                                               Clouds
                                                                           scattered clouds
                                                                                                02
                                                                                            09:00:00
                                                                                           2012-10-
              None 289.36
                                0.0
                                         0.0
                                                    75
                                                               Clouds
                                                                            broken clouds
                                                                                                02
                                                                                            10:00:00
                                                                                           2012-10-
              None 289.58
                                         0.0
                                                    90
                                                               Clouds
         2
                                0.0
                                                                           overcast clouds
                                                                                                02
                                                                                            11:00:00
                                                                                           2012-10-
         3
              None 290.13
                                0.0
                                         0.0
                                                    90
                                                               Clouds
                                                                           overcast clouds
                                                                                                02
                                                                                            12:00:00
                                                                                           2012-10-
              None 291.14
                                0.0
                                         0.0
                                                    75
                                                               Clouds
                                                                            broken clouds
                                                                                                02
                                                                                            13:00:00
In [3]:
          # Let's make sure 'date' is actually a date in pandas
          df["date_time"] = pd.to_datetime(df["date_time"])
```

```
In [4]:
           # plot the dataset
           fig, ax = plt.subplots(figsize=(20, 6))
           ax.plot(df["date_time"], df["traffic_volume"]);
          7000
          6000
          5000
          2000
          1000
                    2013
                                    2014
                                                    2015
                                                                                   2017
In [5]:
           # plot the dataset
           fig, ax = plt.subplots(figsize=(20, 6))
           ax.plot(df["date_time"][20000:40000], df["traffic_volume"][20000:40000]);
          6000
          5000
          3000
          2000
          1000
                                                                                                    2017-10
                  2016-01
                             2016-04
                                         2016-07
                                                     2016-10
                                                                                        2017-07
                                                                                                               2018-01
In [6]:
           # plot the dataset
           fig, ax = plt.subplots(figsize=(20, 6))
           ax.plot(df["date_time"][20000:22000], df["traffic_volume"][20000:22000]);
          7000
          6000
          4000
          3000
          2000
          1000
               2015-12-15
                               2016-01-01
                                            2016-01-15
                                                            2016-02-01
                                                                          2016-02-15
                                                                                        2016-03-01
                                                                                                     2016-03-15
In [7]:
           # plot the dataset
           fig, ax = plt.subplots(figsize=(20, 6))
           ax.plot(df["date_time"][20000:20200], df["traffic_volume"][20000:20200]);
```



Clean Dataset

```
In [8]:
          # drop data prior to 2015 - 07
          df_complete = df[df["date_time"] > '2015-07-01 09:00:00']
 In [9]:
          df_complete.shape
          (32038, 9)
 Out[9]:
In [10]:
          # set index
          df_complete2 = df_complete.set_index('date_time')
          df complete2.index = pd.DatetimeIndex(df complete2.index).to period('H')
In [11]:
          # drop variables
          target_df = df_complete['traffic_volume']
          target_df.head()
                   4273
         16166
Out[11]:
         16167
                   4469
         16168
                   4625
         16169
                   4462
                   4996
         16170
         Name: traffic_volume, dtype: int64
```

Checking Stationarity with Augmented Dicky-Fuller Test

https://analyticsindiamag.com/complete-guide-to-dickey-fuller-test-in-time-series-analysis/

```
In [12]: from statsmodels.tsa.stattools import adfuller
In [13]: series = target_df.values
In [14]: # ADF Test
    result = adfuller(series, autolag='AIC')
```

```
In [15]:
          # cite source here: https://analyticsindiamag.com/complete-guide-to-dickey-fuller-test-
          print('ADF Statistic: %f' % result[0])
          print('p-value: %f' % result[1])
          print('Critical Values:')
          for key, value in result[4].items():
              print('\t%s: %.3f' % (key, value))
          if result[0] < result[4]["5%"]:</pre>
              print ("Reject Ho - Time Series is Stationary")
          else:
              print ("Failed to Reject Ho - Time Series is Non-Stationary")
         ADF Statistic: -21.926192
         p-value: 0.000000
         Critical Values:
                  1%: -3.431
                  5%: -2.862
                  10%: -2.567
```

Determining Parameters (p,d,q)

We know d=0 because the time series is stationary.

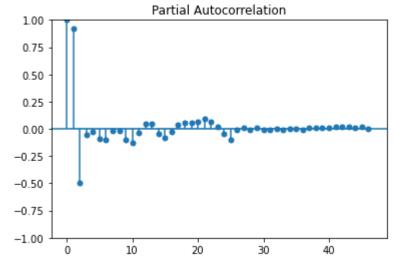
Reject Ho - Time Series is Stationary

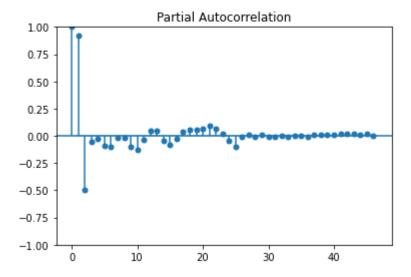
```
In [16]: # Finding p (AR Term)
plot_pacf(series)
```

C:\Users\karli\AppData\Roaming\Python\Python39\site-packages\statsmodels\graphics\tsaplo ts.py:348: FutureWarning: The default method 'yw' can produce PACF values outside of the [-1,1] interval. After 0.13, the default will change tounadjusted Yule-Walker ('ywm'). Y ou can use this method now by setting method='ywm'.

warnings.warn(







Auto ARIMA

In [17]:

! pip install pmdarima --user

Requirement already satisfied: pmdarima in c:\users\karli\appdata\roaming\python\python3 9\site-packages (2.0.3)

Requirement already satisfied: Cython!=0.29.18,!=0.29.31,>=0.29 in c:\programdata\anacon da3\lib\site-packages (from pmdarima) (0.29.24)

Requirement already satisfied: joblib>=0.11 in c:\users\karli\appdata\roaming\python\python\python39\site-packages (from pmdarima) (1.2.0)

Requirement already satisfied: numpy>=1.21.2 in c:\users\karli\appdata\roaming\python\py thon39\site-packages (from pmdarima) (1.22.4)

Requirement already satisfied: statsmodels>=0.13.2 in c:\users\karli\appdata\roaming\python\python39\site-packages (from pmdarima) (0.13.5)

Requirement already satisfied: setuptools!=50.0.0,>=38.6.0 in c:\programdata\anaconda3\l ib\site-packages (from pmdarima) (58.0.4)

Requirement already satisfied: pandas>=0.19 in c:\programdata\anaconda3\lib\site-package s (from pmdarima) (1.3.4)

Requirement already satisfied: urllib3 in c:\programdata\anaconda3\lib\site-packages (fr om pmdarima) (1.26.7)

Requirement already satisfied: scipy>=1.3.2 in c:\programdata\anaconda3\lib\site-package s (from pmdarima) (1.7.1)

Requirement already satisfied: scikit-learn>=0.22 in c:\users\karli\appdata\roaming\pyth on\python39\site-packages (from pmdarima) (1.2.1)

Requirement already satisfied: pytz>=2017.3 in c:\programdata\anaconda3\lib\site-package s (from pandas>=0.19->pmdarima) (2021.3)

Requirement already satisfied: python-dateutil>=2.7.3 in c:\programdata\anaconda3\lib\si te-packages (from pandas>=0.19->pmdarima) (2.8.2)

Requirement already satisfied: six>=1.5 in c:\programdata\anaconda3\lib\site-packages (f rom python-dateutil>=2.7.3->pandas>=0.19->pmdarima) (1.16.0)

Requirement already satisfied: threadpoolctl>=2.0.0 in c:\programdata\anaconda3\lib\site -packages (from scikit-learn>=0.22->pmdarima) (2.2.0)

Requirement already satisfied: packaging>=21.3 in c:\users\karli\appdata\roaming\python \python39\site-packages (from statsmodels>=0.13.2->pmdarima) (23.1)

Requirement already satisfied: patsy>=0.5.2 in c:\programdata\anaconda3\lib\site-package s (from statsmodels>=0.13.2->pmdarima) (0.5.2)

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WARNING: Ignoring invalid distribution -equests (c:\programdata\anaconda3\lib\site-packa ges)
```

```
In [18]:
          from pmdarima.arima import auto_arima
In [19]:
          # train and test split
          train_size = int(len(target_df) * 0.8)
          train = target_df[:train_size]
          test = target df[train size:]
In [20]:
          train.plot()
          test.plot()
          <AxesSubplot:>
Out[20]:
          7000
          6000
          5000
          4000
          3000
          2000
          1000
             0
             15000
                    20000
                          25000
                                 30000
                                        35000
                                               40000
                                                      45000
In [21]:
          arima_model = auto_arima(train, trace=True, information_criterion='bic', max_order = 5)
         Performing stepwise search to minimize bic
          ARIMA(2,0,2)(0,0,0)[0] intercept
                                              : BIC=409019.857, Time=3.69 sec
          ARIMA(0,0,0)(0,0,0)[0] intercept
                                              : BIC=461602.504, Time=0.29 sec
                                              : BIC=415718.150, Time=0.64 sec
          ARIMA(1,0,0)(0,0,0)[0] intercept
                                              : BIC=435734.986, Time=4.36 sec
          ARIMA(0,0,1)(0,0,0)[0] intercept
          ARIMA(0,0,0)(0,0,0)[0]
                                              : BIC=495255.962, Time=0.17 sec
          ARIMA(1,0,2)(0,0,0)[0] intercept
                                              : BIC=409756.291, Time=2.91 sec
                                              : BIC=409026.840, Time=3.11 sec
          ARIMA(2,0,1)(0,0,0)[0] intercept
          ARIMA(3,0,2)(0,0,0)[0] intercept
                                              : BIC=408968.620, Time=13.62 sec
          ARIMA(3,0,1)(0,0,0)[0] intercept
                                              : BIC=409009.407, Time=6.50 sec
          ARIMA(4,0,2)(0,0,0)[0] intercept
                                              : BIC=408904.955, Time=24.20 sec
                                              : BIC=408990.616, Time=8.81 sec
          ARIMA(4,0,1)(0,0,0)[0] intercept
          ARIMA(5,0,2)(0,0,0)[0] intercept
                                              : BIC=408307.740, Time=39.89 sec
          ARIMA(5,0,1)(0,0,0)[0] intercept
                                              : BIC=408794.236, Time=12.31 sec
                                              : BIC=408604.551, Time=40.86 sec
          ARIMA(5,0,3)(0,0,0)[0] intercept
          ARIMA(4,0,3)(0,0,0)[0] intercept
                                              : BIC=408766.504, Time=30.91 sec
          ARIMA(5,0,2)(0,0,0)[0]
                                              : BIC=410777.842, Time=18.99 sec
         Best model: ARIMA(5,0,2)(0,0,0)[0] intercept
         Total fit time: 211.328 seconds
In [22]:
          arima model.summary()
```

Out[22]:

SARIMAX Results

25630 Dep. Variable: y No. Observations: **Log Likelihood** -204108.188 Model: SARIMAX(5, 0, 2) **Date:** Sun, 07 May 2023 AIC 408234.376 Time: 11:52:31 BIC 408307.740 0 Sample: HQIC 408258.093

- 25630

Covariance Type: opg

	coef	std err	z	P> z	[0.025	0.975]
intercept	548.2602	16.349	33.534	0.000	516.216	580.304
ar.L1	1.8325	0.010	187.205	0.000	1.813	1.852
ar.L2	-1.9355	0.019	-101.544	0.000	-1.973	-1.898
ar.L3	1.2753	0.019	67.209	0.000	1.238	1.313
ar.L4	-0.2299	0.012	-19.206	0.000	-0.253	-0.206
ar.L5	-0.1175	0.006	-20.842	0.000	-0.129	-0.106
ma.L1	-0.5065	0.009	-56.699	0.000	-0.524	-0.489
ma.L2	0.8779	0.009	100.832	0.000	0.861	0.895
sigma2	4.883e+05	3010.923	162.182	0.000	4.82e+05	4.94e+05

Ljung-Box (L1) (Q): 5.85 Jarque-Bera (JB): 39939.19

Prob(Q): 0.02 Prob(JB): 0.00

Heteroskedasticity (H): 0.51 Skew: 0.73

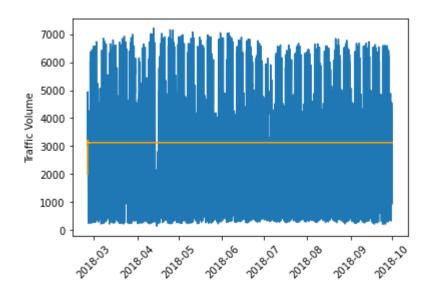
Prob(H) (two-sided): 0.00 **Kurtosis:** 8.94

Warnings:

[1] Covariance matrix calculated using the outer product of gradients (complex-step).

```
In [28]:
          test_plot = df_complete[train_size:]
          plt.plot(test_plot['date_time'], test)
          plt.plot(test_plot['date_time'], arima_model.predict(n_periods=test.shape[0]), color='o
          plt.xticks(rotation=45)
          plt.ylabel('Traffic Volume')
          plt.show()
```

C:\Users\karli\AppData\Roaming\Python\Python39\site-packages\statsmodels\tsa\base\tsa_mo del.py:834: ValueWarning: No supported index is available. Prediction results will be gi ven with an integer index beginning at `start`. return get prediction index(



In	[]:	
In	[]:	
In	[]:	