

Pollen MSE

May 31, 2019

```
In [31]: import warnings
         warnings.filterwarnings('ignore')
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         sns.set(font='IPAGothic')
         import numpy as np
         import statsmodels.api as sm
         from sklearn.metrics import mean_squared_error
```

```
In [7]: data = pd.read_excel('cleaned_data/Pollen_full_data_Eve.xlsx', parse_dates=['date_time'])
```

```
In [9]: data = data['pollenGlobalIndex']
```

```
In [23]: print(data.index.min(), data.index.max())
```

```
2017-09-06 14:00:00 2019-05-20 11:00:00
```

```
In [89]: a = data['2017-09-06 15:00:00:']
```

```
In [99]: print(data.index.min(), data.index.max())
```

```
2017-09-06 15:00:00 2019-05-20 11:00:00
```

```
In [90]: a.head()
```

```
Out[90]: date_time
         2017-09-06 15:00:00    9
         2017-09-06 16:00:00    9
         2017-09-06 17:00:00    9
         2017-09-06 18:00:00    9
         2017-09-06 19:00:00    9
         Name: pollenGlobalIndex, dtype: int64
```

```
In [91]: a.count()
```

```
Out[91]: 33234
```

```
In [85]: data.count()
```

```
Out[85]: 33235
```

```
In [92]: data = a
```

```
In [93]: data.count()
```

```
Out[93]: 33234
```

```
In [98]: data[16617:]
```

```
Out[98]: date_time
2018-03-05 17:00:00    9
2018-03-05 18:00:00    9
2018-03-05 19:00:00   10
2018-03-05 20:00:00   10
2018-03-05 21:00:00   10
2018-03-05 22:00:00   10
2018-03-05 23:00:00   10
2018-03-06 00:00:00    9
2018-03-06 01:00:00    9
2018-03-06 02:00:00    9
2018-03-06 03:00:00   10
2018-03-06 04:00:00    9
2018-03-06 05:00:00    9
2018-03-06 06:00:00    9
2018-03-06 07:00:00   10
2018-03-06 08:00:00    9
2018-03-06 09:00:00    9
2018-03-06 10:00:00    9
2018-03-06 11:00:00    9
2018-03-06 12:00:00    9
2018-03-06 13:00:00    9
2018-03-06 14:00:00    9
2018-03-06 15:00:00   10
2018-03-06 16:00:00    9
2018-03-06 17:00:00   10
2018-03-06 18:00:00    9
2018-03-06 19:00:00    9
2018-03-06 20:00:00    9
2018-03-06 21:00:00    9
2018-03-06 22:00:00    9
...
2017-10-26 12:00:00    1
2017-11-06 16:00:00    1
2017-11-07 12:00:00    1
2017-11-07 13:00:00    1
2017-11-07 14:00:00    1
```

```

2017-11-16 15:00:00    1
2017-11-16 16:00:00    1
2017-11-16 17:00:00    1
2017-11-16 18:00:00    3
2017-11-16 19:00:00    3
2017-11-16 20:00:00    3
2017-11-16 21:00:00    3
2017-11-16 22:00:00    3
2017-11-16 23:00:00    3
2017-11-17 00:00:00    1
2017-11-17 01:00:00    1
2017-11-17 02:00:00    1
2017-11-17 03:00:00    1
2017-11-17 04:00:00    1
2017-11-17 05:00:00    1
2017-11-17 06:00:00    1
2017-11-17 07:00:00    1
2017-11-17 08:00:00    1
2017-11-17 09:00:00    1
2017-11-17 10:00:00    1
2017-11-17 11:00:00    1
2017-11-17 13:00:00    1
2017-11-17 14:00:00    1
2017-11-17 15:00:00    1
2017-11-17 16:00:00    1
Name: pollenGlobalIndex, Length: 16617, dtype: int64

```

```
In [94]: 33234/2
```

```
Out[94]: 16617.0
```

```
In [70]: data['2019-05-18 15:00:00']
```

```
Out[70]: date_time
2019-05-18 15:00:00    9
Name: pollenGlobalIndex, dtype: int64
```

```
In [100]: tr_start, tr_end = '2017-09-06 15:00:00', '2018-03-05 17:00:00'
          te_start, te_end = '2018-03-05 18:00:00', '2019-05-20 11:00:00'
```

```
In [101]: tra = data[tr_start:tr_end].dropna()
          tes = data[te_start:te_end].dropna()
```

```
In [103]: tra.count()
```

```
Out[103]: 22411
```

```
In [104]: tes.count()
```

```
Out[104]: 22411
```


[illegible]


```

' ignored when e.g. forecasting.', ValueWarning)
/home/omar/.local/lib/python3.5/site-packages/statsmodels/tsa/base/tsa_model.py:225: ValueWarning
' ignored when e.g. forecasting.', ValueWarning)
/home/omar/.local/lib/python3.5/site-packages/statsmodels/tsa/base/tsa_model.py:225: ValueWarning
' ignored when e.g. forecasting.', ValueWarning)
/home/omar/.local/lib/python3.5/site-packages/statsmodels/tsa/base/tsa_model.py:225: ValueWarning
' ignored when e.g. forecasting.', ValueWarning)

```

ARMA(p,q) = (7, 7) is the best.

Parameters

```

In [30]: arima = sm.tsa.statespace.SARIMAX(tra,order=(7,1,7),seasonal_order=(0,0,0,0),
                                             enforce_stationarity=False, enforce_invertibility=False)

arima.summary()

/home/omar/.local/lib/python3.5/site-packages/statsmodels/tsa/base/tsa_model.py:225: ValueWarning
' ignored when e.g. forecasting.', ValueWarning)
/home/omar/.local/lib/python3.5/site-packages/statsmodels/base/model.py:508: ConvergenceWarning:
"Check mle_retvals", ConvergenceWarning)

```

```

Out[30]: <class 'statsmodels.iolib.summary.Summary'>
        """

```

```

                                Statespace Model Results
=====
Dep. Variable:      pollenGlobalIndex      No. Observations:      30344
Model:              SARIMAX(7, 1, 7)      Log Likelihood      -45555.644
Date:              Thu, 30 May 2019      AIC      91141.288
Time:              22:38:07      BIC      91266.089
Sample:            0      HQIC      91181.311
                  - 30344
Covariance Type:      opg
=====

```

	coef	std err	z	P> z	[0.025	0.975]
ar.L1	-0.3179	0.011	-29.973	0.000	-0.339	-0.297
ar.L2	-0.1826	0.009	-19.276	0.000	-0.201	-0.164
ar.L3	0.1550	0.010	14.772	0.000	0.134	0.176
ar.L4	-0.2773	0.011	-25.041	0.000	-0.299	-0.256
ar.L5	-0.1478	0.008	-17.755	0.000	-0.164	-0.131
ar.L6	-0.2519	0.007	-34.589	0.000	-0.266	-0.238
ar.L7	0.6020	0.008	73.819	0.000	0.586	0.618
ma.L1	0.1330	0.010	13.464	0.000	0.114	0.152
ma.L2	0.0894	0.009	10.286	0.000	0.072	0.106
ma.L3	-0.3877	0.009	-42.449	0.000	-0.406	-0.370
ma.L4	0.1502	0.010	14.974	0.000	0.131	0.170

ma.L5	0.0361	0.008	4.311	0.000	0.020	0.052
ma.L6	0.0083	0.009	0.944	0.345	-0.009	0.026
ma.L7	-0.9067	0.008	-107.486	0.000	-0.923	-0.890
sigma2	1.0554	0.011	92.209	0.000	1.033	1.078

=====

Ljung-Box (Q):	5985.72	Jarque-Bera (JB):	375630.19
Prob(Q):	0.00	Prob(JB):	0.00
Heteroskedasticity (H):	0.93	Skew:	0.40
Prob(H) (two-sided):	0.00	Kurtosis:	20.22

=====

Warnings:

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

```
pred = arima.predict(tr_end,te_end, dynamic=True)[1:]      print('ARIMA  model
MSE:{}'.format(mean_squared_error(tes,pred)))
```

```
In [115]: pred = arima.predict(10823,33234, dynamic=True)[1:]
          print('ARIMA model MSE:{}'.format(mean_squared_error(tes,pred)))
```

ARIMA model MSE:29.15488319030185