

ZINC

November 5, 2022

1 Import required libraries

```
[25]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
from matplotlib.pyplot import rcParams
rcParams['figure.figsize'] = 15,6
from pandas.tseries.offsets import DateOffset

import math
from datetime import datetime as dt
from statsmodels.tsa.stattools import adfuller, acf, pacf
from statsmodels.tsa.arima_model import ARIMA

import warnings
warnings.filterwarnings('ignore')
```

2 Visualize the time series

```
[2]: df = pd.read_csv('zinc_prices_IMF.csv')
```

```
[3]: df['Date'].head()
```

```
[3]: 0    1-Jan-80
1    1-Feb-80
2    1-Mar-80
3    1-Apr-80
4    1-May-80
Name: Date, dtype: object
```

```
[4]: df = df.set_index('Date')
df.head
```

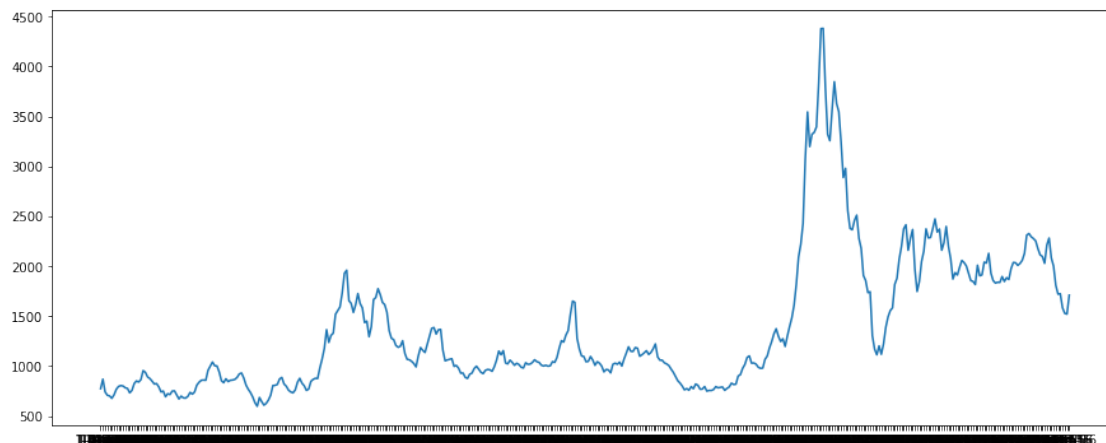
```
[4]: <bound method NDFrame.head of
Date
1-Jan-80    773.82
1-Feb-80    868.62
1-Mar-80    740.75
1-Apr-80    707.68
1-May-80    701.07
...
1-Oct-15   1724.34
1-Nov-15   1583.31
1-Dec-15   1527.79
1-Jan-16   1520.36
1-Feb-16   1709.85

[434 rows x 1 columns]>
```

```
[5]: ts = df['Price']
```

```
[6]: plt.plot(ts)
```

```
[6]: [<matplotlib.lines.Line2D at 0x7f2a07e89710>]
```



3 Check for the stationarity of your data using Rolling Statistics and Dickey-Fuller test

```
[7]: ts_log = np.log(ts)
```

```
[8]: def test_stationarity (timeseries):
    rolmean = timeseries.rolling(window=52, center = False).mean()
```

```

rolstd = timeseries.rolling(window=52, center = False).std()

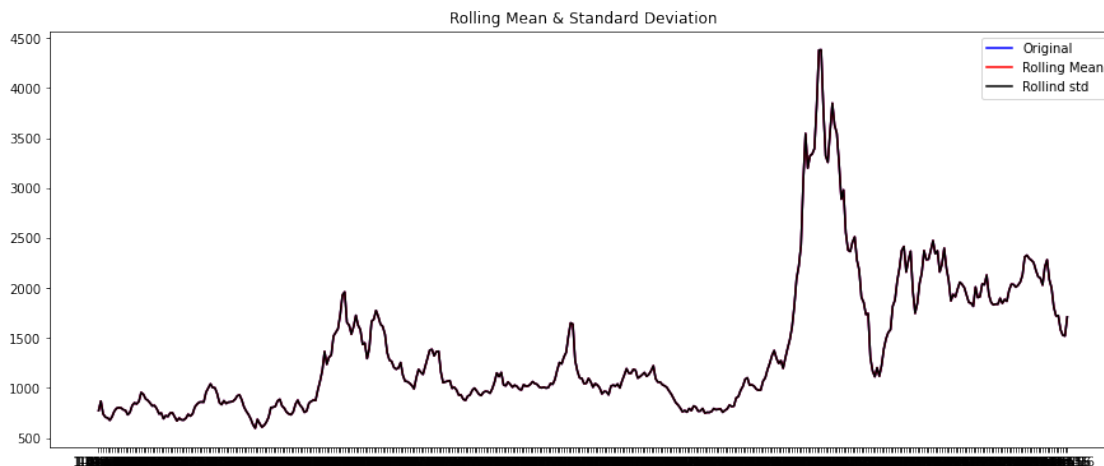
orig = plt.plot(timeseries,color='blue', label='Original')
mean = plt.plot(timeseries, color='red', label='Rolling Mean')
std = plt.plot(timeseries, color='black', label='Rollind std')
plt.legend(loc='best')
plt.title('Rolling Mean & Standard Deviation')
plt.show(block= False)

print('Results of Dickey-Fuller Test:')
dftest = adfuller(timeseries, autolag='AIC')
dfoutput = pd.Series(dftest[0:4],index=['Test Statistic', 'p-values', '#Lags_
→Used',
                                     'Number of Observations Used'])

for key,value in dftest[4].items():
    dfoutput['Critical Value (%s)'%key] = value
print(dfoutput)

```

```
[9]: test_stationarity(df['Price'])
```



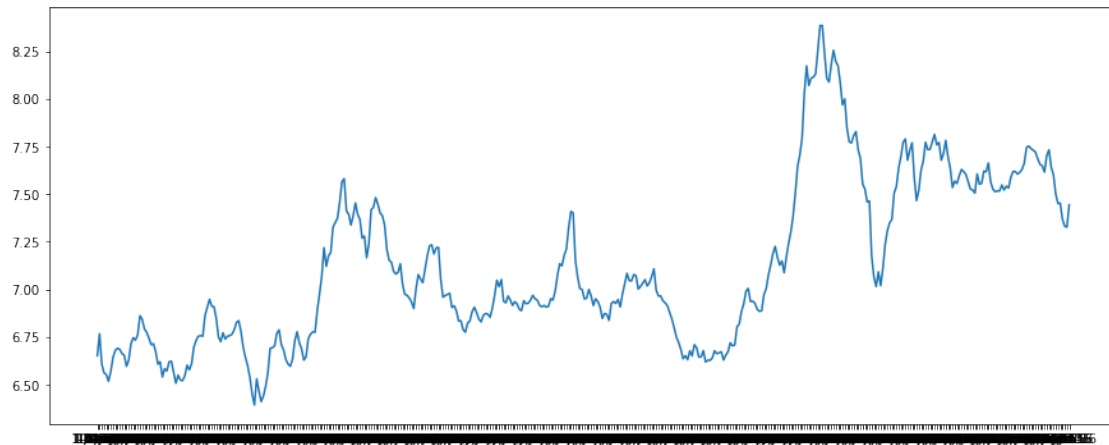
```

Results of Dickey-Fuller Test:
Test Statistic          -3.139601
p-values                0.023758
#Lags Used              7.000000
Number of Observations Used  426.000000
Critical Value (1%)     -3.445794
Critical Value (5%)     -2.868349
Critical Value (10%)    -2.570397
dtype: float64

```

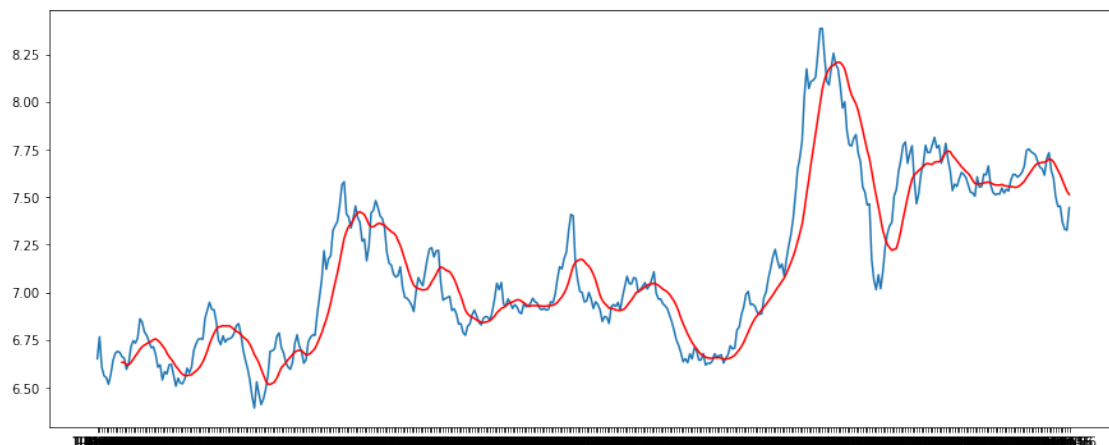
```
[10]: plt.plot(ts_log)
```

```
[10]: [<matplotlib.lines.Line2D at 0x7f2a0552e110>]
```



```
[11]: MovingAverage = ts_log.rolling(window=12).mean()  
MovingSTD = ts_log.rolling(window=12).std()  
plt.plot(ts_log)  
plt.plot(MovingAverage, color='red')
```

```
[11]: [<matplotlib.lines.Line2D at 0x7f2a04a840d0>]
```



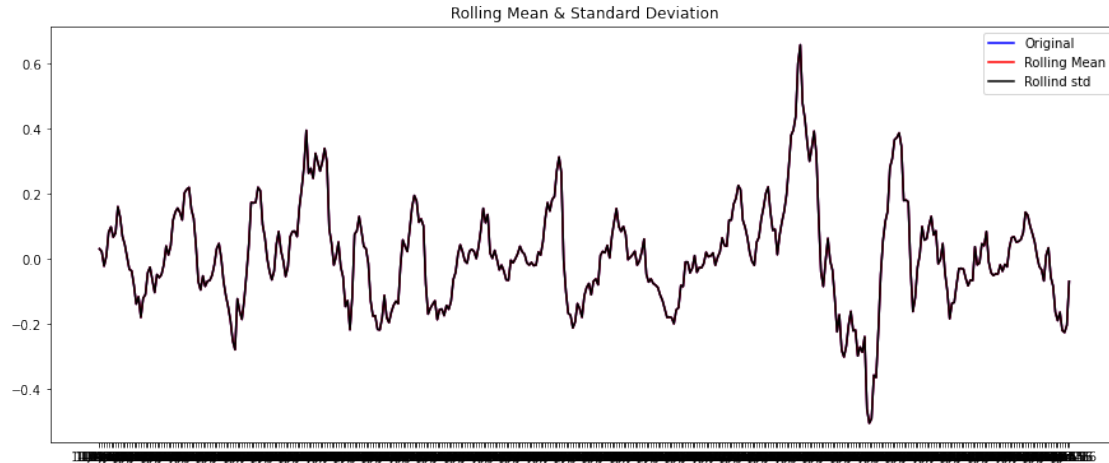
```
[12]: ts_log_mv_diff = ts_log - MovingAverage  
ts_log_mv_diff.head(12)
```

```
[12]: Date
      1-Jan-80      NaN
      1-Feb-80      NaN
      1-Mar-80      NaN
      1-Apr-80      NaN
      1-May-80      NaN
      1-Jun-80      NaN
      1-Jul-80      NaN
      1-Aug-80      NaN
      1-Sep-80      NaN
      1-Oct-80      NaN
      1-Nov-80      NaN
      1-Dec-80      0.030472
      Name: Price, dtype: float64
```

```
[13]: ts_log_mv_diff.dropna(inplace=True)
      ts_log_mv_diff.head(12)
```

```
[13]: Date
      1-Dec-80      0.030472
      1-Jan-81      0.021753
      1-Feb-81     -0.022485
      1-Mar-81      0.008392
      1-Apr-81      0.082191
      1-May-81      0.097617
      1-Jun-81      0.066587
      1-Jul-81      0.078914
      1-Aug-81      0.160180
      1-Sep-81      0.127928
      1-Oct-81      0.068802
      1-Nov-81      0.043854
      Name: Price, dtype: float64
```

```
[14]: test_stationarity(ts_log_mv_diff)
```



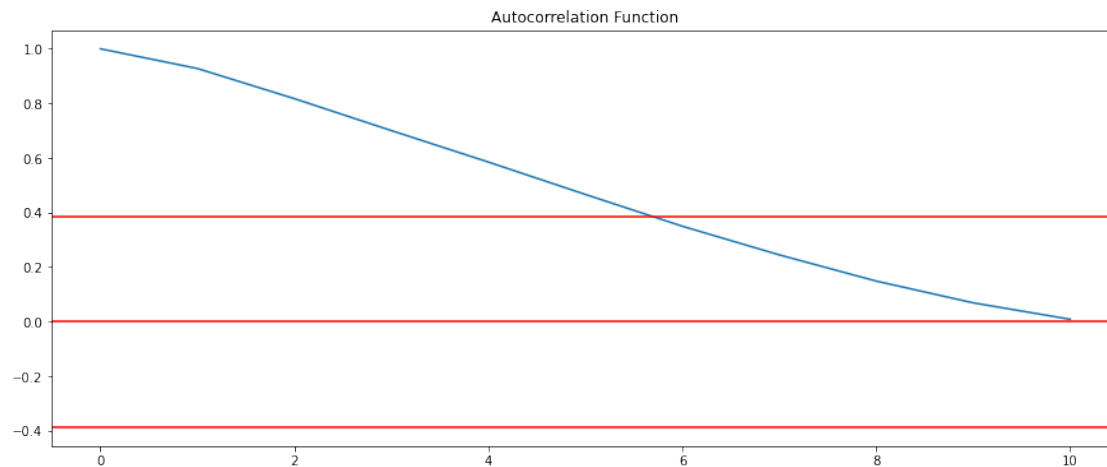
Results of Dickey-Fuller Test:

Test Statistic	-5.898484e+00
p-values	2.814411e-07
#Lags Used	4.000000e+00
Number of Observations Used	4.180000e+02
Critical Value (1%)	-3.446091e+00
Critical Value (5%)	-2.868479e+00
Critical Value (10%)	-2.570466e+00

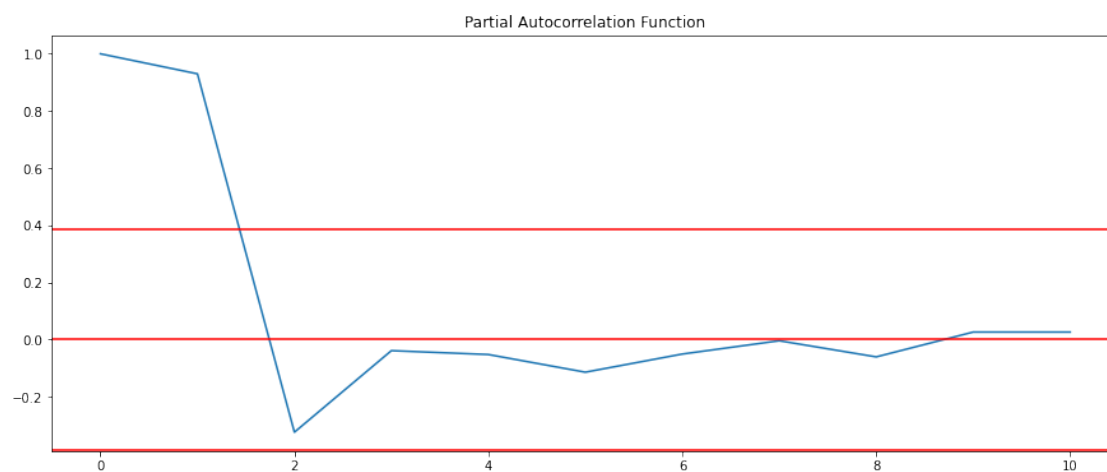
dtype: float64

4 Plot ACF and PACF plots

```
[15]: plt.plot(np.arange(0,11), acf(ts_log_mv_diff,nlags=10))
plt.axhline(y=0, linestyle='-',color='red')
plt.axhline(y=-7.96/np.sqrt(len(ts_log_mv_diff)),linestyle='-', color='red')
plt.axhline(y=7.96/np.sqrt(len(ts_log_mv_diff)),linestyle='-', color='red')
plt.title('Autocorrelation Function')
plt.show()
```



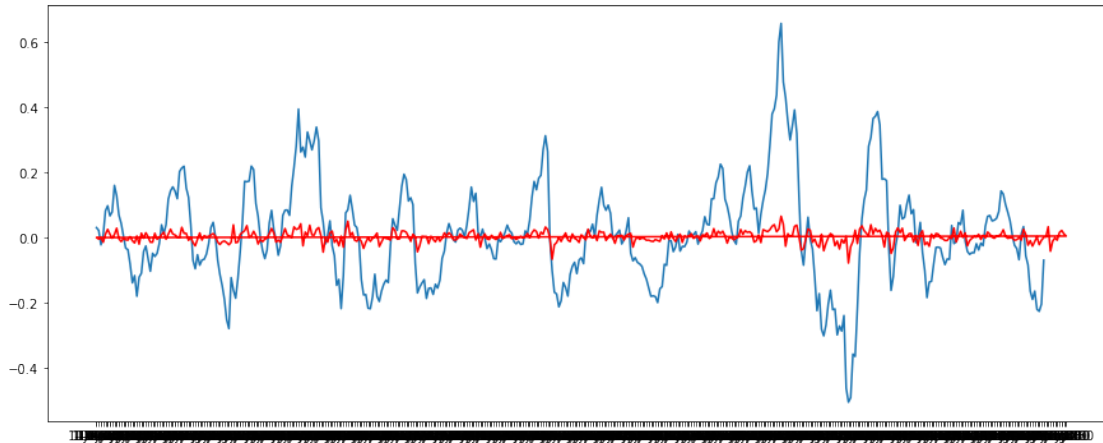
```
[16]: plt.plot(np.arange(0,11), pacf(ts_log_mv_diff,nlags=10))
plt.axhline(y=0, linestyle='-',color='red')
plt.axhline(y=-7.96/np.sqrt(len(ts_log_mv_diff)),linestyle='-', color='red')
plt.axhline(y=7.96/np.sqrt(len(ts_log_mv_diff)),linestyle='-', color='red')
plt.title('Partial Autocorrelation Function')
plt.show()
```



5 Perform ARIMA modeling

```
[17]: model = ARIMA (ts_log, order=(1,1,0))
results_ARIMA = model.fit(dispatch=-1)
plt.plot(ts_log_mv_diff)
plt.plot(results_ARIMA.fittedvalues, color='red')
```

```
[17]: [<matplotlib.lines.Line2D at 0x7f29f87195d0>]
```



```
[18]: predictions_ARIMA_diff = pd.Series(results_ARIMA.fittedvalues, copy=True)
predictions_ARIMA_diff.head()
```

```
[18]: Date
1-Feb-80    0.002030
1-Mar-80    0.033049
1-Apr-80   -0.042031
1-May-80   -0.011002
1-Jun-80   -0.001089
dtype: float64
```

```
[19]: predictions_ARIMA_diff_cumsum = predictions_ARIMA_diff.cumsum()
predictions_ARIMA_diff_cumsum.head()
```

```
[19]: Date
1-Feb-80    0.002030
1-Mar-80    0.035079
1-Apr-80   -0.006952
1-May-80   -0.017955
1-Jun-80   -0.019043
dtype: float64
```

```
[20]: predictions_ARIMA_log = pd.Series(ts_log[0], index=ts_log.index)
```



```

predictions_ARIMA_log = predictions_ARIMA_log.
↪add(predictions_ARIMA_diff_cumsum, fill_value=0)
predictions_ARIMA_log.head()

```

```

[20]: Date
1-Apr-00    7.109676
1-Apr-01    7.099660
1-Apr-02    7.061613
1-Apr-03    7.069247
1-Apr-04    7.177723
dtype: float64

```

6 Forecast the prices using the new model

```

[31]: predictions_ARIMA = np.exp(predictions_ARIMA_log)
plt.plot(ts)
plt.plot(predictions_ARIMA)

```

```

[31]: [<matplotlib.lines.Line2D at 0x7f2a050b62d0>]

```

