finn

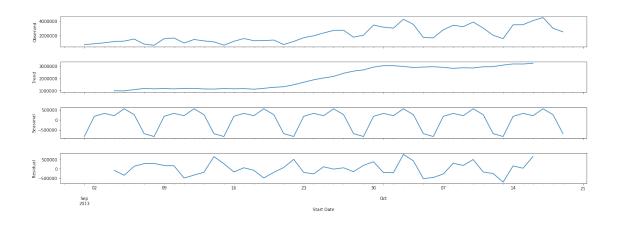
September 4, 2018

```
In [1]: import pandas as pd
                         df_raw = pd.read_excel('Q4_2013_Groupon_North_America_Data_XLSX.xlsx', 'Q4 2013 Raw Data_XLSX.xlsx', 'Q4 2013 Raw Data_XLSX.xl
                         df_Local = df_raw[df_raw['Segment'] == 'Local']
                         df_Goods = df_raw[df_raw['Segment'] == 'Goods']
                         df_Travel = df_raw[df_raw['Segment'] == 'Travel']
In [2]: %matplotlib inline
                          import pandas as pd
                          import numpy as np
                          import matplotlib.pyplot as plt
                          import datetime
                         from dateutil.relativedelta import relativedelta
                          import seaborn as sns
                          import statsmodels.api as sm
                         from statsmodels.tsa.stattools import acf
                         from statsmodels.tsa.stattools import pacf
                         from statsmodels.tsa.seasonal import seasonal_decompose
In [3]: Local_Billings = df_Local.groupby('Start Date')['Start Date', 'Billings'].sum()
In [4]: data = Local_Billings['2013-09-01':'2013-10-19']
In [5]: data.Billings.plot(figsize=(22,8), title= 'Local Billings', fontsize=14)
                         plt.savefig('Local_Billings.png', bbox_inches='tight')
                                                                                                                                 Local Billings
              4500000
              4000000
              3500000
              3000000
              2500000
              2000000
               500000
```

Start Date

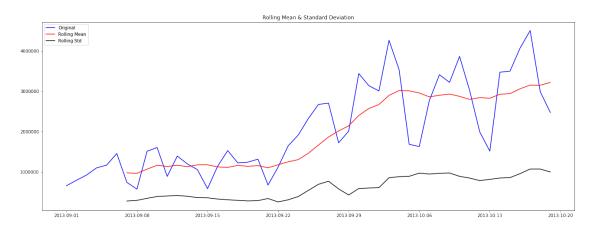
```
In [6]: decomposition = seasonal_decompose(data.Billings, freq=7)
    fig = plt.figure()
    fig = decomposition.plot()
    fig.set_size_inches(22, 8)
```

<Figure size 432x288 with 0 Axes>



```
In [7]: from statsmodels.tsa.stattools import adfuller
        def test_stationarity(timeseries):
            #Determing rolling statistics
            rolmean = timeseries.rolling(window=7).mean()
            rolstd = timeseries.rolling(window=7).std()
            #Plot rolling statistics:
            fig = plt.figure(figsize=(22, 8))
            orig = plt.plot(timeseries, color='blue',label='Original')
           mean = plt.plot(rolmean, color='red', label='Rolling Mean')
            std = plt.plot(rolstd, color='black', label = 'Rolling Std')
           plt.legend(loc='best')
           plt.title('Rolling Mean & Standard Deviation')
           plt.show()
            #Perform Dickey-Fuller test:
            print ('Results of Dickey-Fuller Test:')
            dftest = adfuller(timeseries, autolag='AIC')
            dfoutput = pd.Series(dftest[0:4], index=['Test Statistic','p-value','#Lags Used',']
            for key,value in dftest[4].items():
                dfoutput['Critical Value (%s)'%key] = value
            print (dfoutput)
```

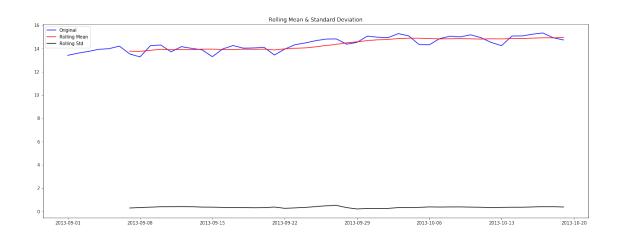
In [8]: test_stationarity(data.Billings)



Test Statistic -0.493765
p-value 0.893227
#Lags Used 8.000000
Number of Observations Used 40.000000
Critical Value (1%) -3.605565
Critical Value (5%) -2.937069
Critical Value (10%) -2.606986

dtype: float64

/Users/vicky/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: UserWarning: Panda """Entry point for launching an IPython kernel.



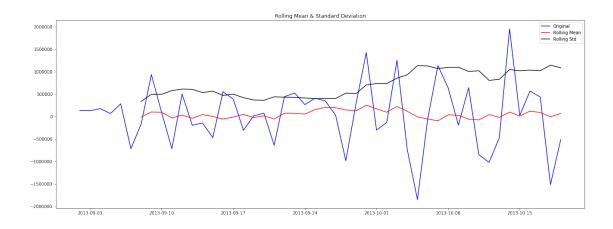
Test Statistic -0.676594
p-value 0.852720
#Lags Used 8.000000
Number of Observations Used 40.000000
Critical Value (1%) -3.605565
Critical Value (5%) -2.937069
Critical Value (10%) -2.606986

dtype: float64

/Users/vicky/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: SettingWithCopyWars
A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.htm """Entry point for launching an IPython kernel.



Results of Dickey-Fuller Test:

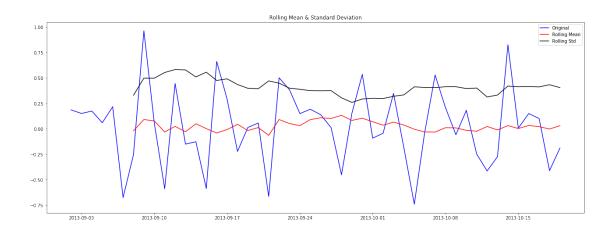
Test Statistic -1.771613
p-value 0.394643
#Lags Used 7.000000
Number of Observations Used 40.000000
Critical Value (1%) -3.605565
Critical Value (5%) -2.937069
Critical Value (10%) -2.606986

dtype: float64

/Users/vicky/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: SettingWithCopyWars
A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.htm """Entry point for launching an IPython kernel.



Results of Dickey-Fuller Test:

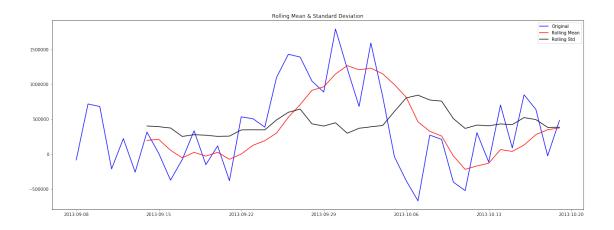
Test Statistic	-1.831369
p-value	0.365003
#Lags Used	7.000000
Number of Observations Used	40.000000
Critical Value (1%)	-3.605565
Critical Value (5%)	-2.937069
Critical Value (10%)	-2.606986

dtype: float64

/Users/vicky/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: SettingWithCopyWars
A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.htm """Entry point for launching an IPython kernel.



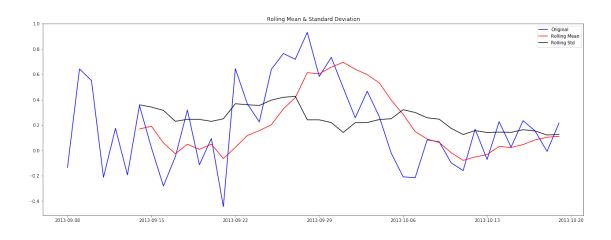
Test Statistic		-3.247368
p-value		0.017401
#Lags Used		8.000000
Number of Obser	vations Used	33.000000
Critical Value	(1%)	-3.646135
Critical Value	(5%)	-2.954127
Critical Value	(10%)	-2.615968

dtype: float64

/Users/vicky/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: SettingWithCopyWars
A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.htm """Entry point for launching an IPython kernel.



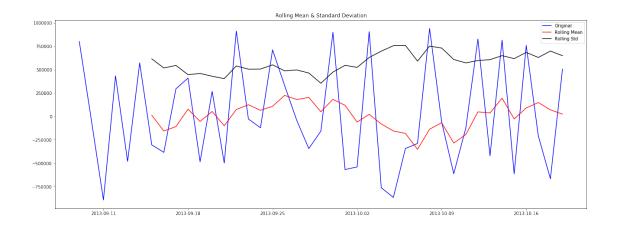
Test Statistic	-2.713955
p-value	0.071654
#Lags Used	8.000000
Number of Observat	ions Used 33.000000
Critical Value (1%) -3.646135
Critical Value (5%) -2.954127
Critical Value (10	%) -2.615968

dtype: float64

/Users/vicky/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: SettingWithCopyWars
A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.htm """Entry point for launching an IPython kernel.



Results of Dickey-Fuller Test:

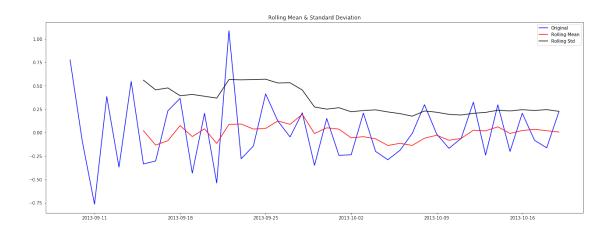
Test Statistic -8.999948e+00
p-value 6.538471e-15
#Lags Used 0.000000e+00
Number of Observations Used 4.000000e+01
Critical Value (1%) -3.605565e+00
Critical Value (5%) -2.937069e+00
Critical Value (10%) -2.606986e+00

dtype: float64

/Users/vicky/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: SettingWithCopyWars A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.htm """Entry point for launching an IPython kernel.



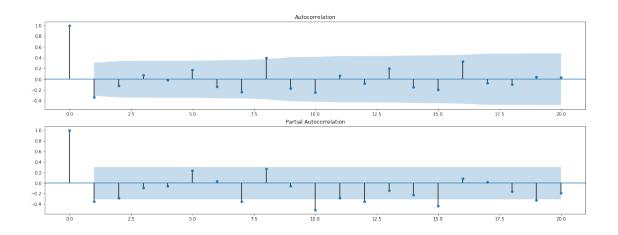
Results of Dickey-Fuller Test:

Test Statistic -2.755552
p-value 0.064909
#Lags Used 9.000000
Number of Observations Used 31.000000
Critical Value (1%) -3.661429
Critical Value (5%) -2.960525
Critical Value (10%) -2.619319

In [16]: fig = plt.figure(figsize=(22,8))

dtype: float64

```
ax1 = fig.add_subplot(211)
fig = sm.graphics.tsa.plot_acf(data.seasonal_first_difference.iloc[8:], lags=20, ax=a:
ax2 = fig.add_subplot(212)
fig = sm.graphics.tsa.plot_pacf(data.seasonal_first_difference.iloc[8:], lags=20, ax=a:
```



Statespace Model Results

Dep. Variable: Model: Date: Time: Sample:	SARIM	MAX(0, 1, 0) Tue		, 7) Log L 2018 AIC 9:39 BIC 2013 HQIC	bservations: ikelihood		49 -599.473 1202.946 1206.373 1204.194
Covariance Type:				opg			
	coef	std err	z	P> z	[0.025	0.975]	
			-1.581 2e+24		-0.938 3.79e+11		
Ljung-Box (Q): Prob(Q): Heteroskedasticit Prob(H) (two-side			69.66 0.00 1.32 0.62	Jarque-Bera Prob(JB): Skew: Kurtosis:	(JB):		1.29 0.52 0.11 2.16

Warnings:

- [1] Covariance matrix calculated using the outer product of gradients (complex-step).
- [2] Covariance matrix is singular or near-singular, with condition number 2.76e+40. Standard ex

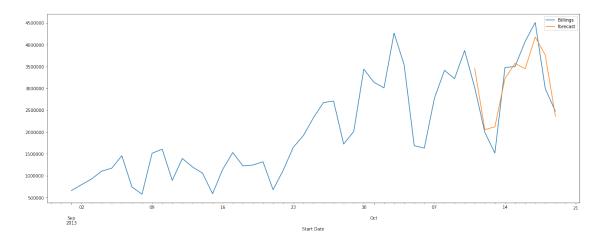
/Users/vicky/anaconda3/lib/python3.6/site-packages/statsmodels/tsa/base/tsa_model.py:171: Value % freq, ValueWarning)

/Users/vicky/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: SettingWithCopyWars
A value is trying to be set on a copy of a slice from a DataFrame.

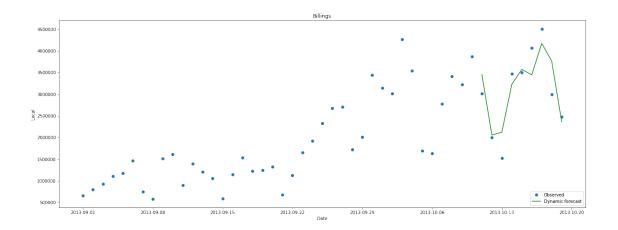
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.htm """Entry point for launching an IPython kernel.

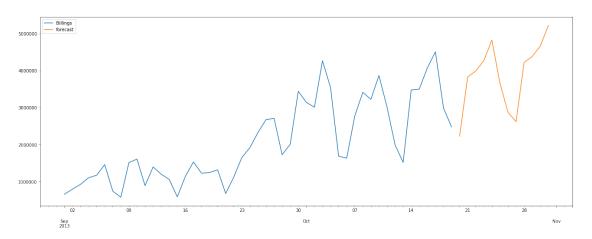
Out[18]: <matplotlib.axes._subplots.AxesSubplot at 0x10e279a58>



```
In [19]: npredict = data.Billings['2013-09-01':].shape[0]
    fig, ax = plt.subplots(figsize=(22,8))
    npre = 7
    ax.set(title='Billings', xlabel='Date', ylabel='Local')
    ax.plot(data.index[-npredict-npre+1:], data.iloc[-npredict-npre+1:]['Billings'], 'o',
    ax.plot(data.index[-npredict-npre+1:], data.iloc[-npredict-npre+1:]['forecast'], 'g',
    legend = ax.legend(loc='lower right')
    legend.get_frame().set_facecolor('w')
    #plt.savefig('ts_predict_compare.png', bbox_inches='tight')
```



Out[21]: <matplotlib.axes._subplots.AxesSubplot at 0x10f722a20>



```
In [22]: data
```

2013-09-04	1.104587e+06	1.777438e+05	0.175442
2013-09-05	1.173147e+06	6.855923e+04	0.060218
2013-09-06	1.459212e+06	2.860653e+05	0.218207
2013-09-07	7.433991e+05	-7.158128e+05	-0.674419
2013-09-08	5.773694e+05	-1.660297e+05	-0.252751
2013-09-09	1.515168e+06	9.377988e+05	0.964799
2013-09-10	1.607966e+06	9.279752e+04	0.059443
2013-09-11	8.929169e+05	-7.150488e+05	-0.588232
2013-09-12	1.395849e+06	5.029323e+05	0.446765
2013-09-13	1.201757e+06	-1.940918e+05	-0.149718
2013-09-14	1.059236e+06	-1.425211e+05	-0.126237
2013-09-15	5.890799e+05	-4.701564e+05	-0.586742
2013-09-16	1.143040e+06	5.539596e+05	0.662884
2013-09-17	1.531653e+06	3.886131e+05	0.292656
2013-09-18	1.227678e+06	-3.039750e+05	-0.221223
2013-09-19	1.245592e+06	1.791426e+04	0.014487
2013-09-20	1.319770e+06	7.417796e+04	0.057847
2013-09-21	6.793485e+05	-6.404214e+05	-0.664078
2013-09-22	1.123031e+06	4.436824e+05	0.502652
2013-09-23	1.650251e+06	5.272196e+05	0.384896
2013-09-24	1.918048e+06	2.677974e+05	0.150381
2013-09-25	2.327159e+06	4.091115e+05	0.193340
2013-09-26	2.675751e+06	3.485919e+05	0.139582
2013-09-27	2.709123e+06	3.337194e+04	0.012395
2013-09-28	1.725309e+06	-9.838143e+05	-0.451219
2013-09-29	2.012469e+06	2.871604e+05	0.153956
2013-09-30	3.440883e+06	1.428414e+06	0.536366
2013-10-10	3.866735e+06	6.439296e+05	0.182158
2013-10-11	3.015991e+06	-8.507440e+05	-0.248482
2013-10-12	1.994842e+06	-1.021149e+06	-0.413364
2013-10-13	1.519141e+06	-4.757016e+05	-0.272420
2013-10-14	3.475041e+06	1.955900e+06	0.827461
2013-10-15	3.501416e+06	2.637561e+04	0.007561
2013-10-16	4.072212e+06	5.707955e+05	0.151019
2013-10-17	4.507996e+06	4.357840e+05	0.101666
2013-10-18	2.991175e+06	-1.516821e+06	-0.410186
2013-10-19	2.477089e+06	-5.140854e+05	-0.188582
2013-10-20	NaN	NaN	NaN
2013-10-21	NaN	NaN	NaN
2013-10-22	NaN	NaN	NaN
2013-10-23	NaN	NaN	NaN
2013-10-24	NaN	NaN	NaN
2013-10-25	NaN	NaN	NaN
2013-10-26	NaN	NaN	NaN
2013-10-27	NaN	NaN	NaN
2013-10-28	NaN	NaN	NaN
2013-10-29	NaN	NaN	NaN

2013-10-30	NaN	NaN	NaN
2013-10-31	NaN	NaN	NaN
2013-11-01	NaN	NaN	NaN
2013-11-02	NaN	NaN	NaN
2013-11-03	NaN	NaN	NaN
2013-11-04	NaN	NaN	NaN
2013-11-05	NaN	NaN	NaN
2013-11-06	NaN	NaN	NaN
2013-11-07	NaN	NaN	NaN
2013-11-08	NaN	NaN	NaN
	seasonal_difference	log_seasonal_difference	\
2013-09-01	NaN	NaN	
2013-09-02	NaN	NaN	
2013-09-03	NaN	NaN	
2013-09-04	NaN	NaN	
2013-09-05	NaN	NaN	
2013-09-06	NaN	NaN	
2013-09-07	NaN	NaN	
2013-09-08	-8.342101e+04	-0.134954	
2013-09-09	7.188332e+05	0.643262	
2013-09-10	6.811222e+05	0.550940	
2013-09-11	-2.116704e+05	-0.212734	
2013 09 11	2.227027e+05	0.173813	
2013 09 12	-2.574545e+05	-0.194111	
2013-09-13	3.158372e+05	0.354070	
2013-09-14	1.171051e+04	0.020080	
2013-09-16	-3.721286e+05 -7.631301e+04	-0.281835	
2013-09-17		-0.048622	
2013-09-18	3.347608e+05	0.318386	
2013-09-19	-1.502573e+05	-0.113892	
2013-09-20	1.180125e+05	0.093672	
2013-09-21	-3.798877e+05	-0.444169	
2013-09-22	5.339511e+05	0.645225	
2013-09-23	5.072110e+05	0.367236	
2013-09-24	3.863953e+05	0.224961	
2013-09-25	1.099482e+06	0.639524	
2013-09-26	1.430159e+06	0.764619	
2013-09-27	1.389353e+06	0.719168	
2013-09-28	1.045960e+06	0.932027	
2013-09-29	8.894385e+05	0.583331	
2013-09-30	1.790632e+06	0.734801	
		• • •	
2013-10-10	-4.001325e+05	-0.098469	
2013-10-11	-5.220921e+05	-0.159657	
2013-10-12	3.069616e+05	0.167091	
2013-10-13	-1.134224e+05	-0.072006	
2013-10-14	7.030876e+05	0.226054	

2013-10-15	8.868560e+04	0.025655	
2013-10-16	8.494065e+05	0.233934	
2013-10-17	6.412609e+05	0.153442	
2013-10-18	-2.481629e+04	-0.008262	
2013-10-19	4.822471e+05	0.216519	
2013-10-20	NaN	NaN	
2013-10-21	NaN	NaN	
2013-10-22	NaN	NaN	
2013-10-23	NaN	NaN	
2013-10-24	NaN	NaN	
2013-10-25	NaN	NaN	
2013-10-26	NaN	NaN	
2013-10-27	NaN	NaN	
2013-10-28	NaN	NaN	
2013-10-29	NaN	NaN	
2013-10-30	NaN	NaN	
2013-10-31	NaN	NaN	
2013-11-01	NaN	NaN	
2013-11-02	NaN	NaN	
2013-11-03	NaN	NaN	
2013-11-04	NaN	NaN	
2013-11-05	NaN	NaN	
2013-11-06	NaN	NaN	
2013-11-07	NaN	NaN	
2013-11-08	NaN	NaN	
2013-11-08			,
	seasonal_first_difference	log_seasonal_first_difference	\
2013-09-01	seasonal_first_difference	log_seasonal_first_difference	\
2013-09-01 2013-09-02	seasonal_first_difference NaN NaN	log_seasonal_first_difference NaN NaN	\
2013-09-01 2013-09-02 2013-09-03	seasonal_first_difference NaN NaN NaN	log_seasonal_first_difference NaN NaN NaN	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04	seasonal_first_difference NaN NaN NaN NaN	log_seasonal_first_difference NaN NaN NaN NaN	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-05	seasonal_first_difference NaN NaN NaN NaN NaN	log_seasonal_first_difference NaN NaN NaN NaN NaN	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-05 2013-09-06	seasonal_first_difference NaN NaN NaN NaN NaN NaN	log_seasonal_first_difference	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-05 2013-09-06 2013-09-07	seasonal_first_difference NaN NaN NaN NaN NaN NaN NaN NaN NaN	log_seasonal_first_difference	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-05 2013-09-06 2013-09-07 2013-09-08	seasonal_first_difference NaN NaN NaN NaN NaN NaN NaN NaN NaN Na	log_seasonal_first_difference NaN NaN NaN NaN NaN NaN NaN NaN NaN Na	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-05 2013-09-06 2013-09-07 2013-09-08 2013-09-09	seasonal_first_difference NaN NaN NaN NaN NaN NaN NaN NaN NaN Na	log_seasonal_first_difference NaN NaN NaN NaN NaN NaN NaN NaN NaN Na	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-06 2013-09-07 2013-09-08 2013-09-09 2013-09-10	seasonal_first_difference NaN NaN NaN NaN NaN NaN NaN NaN NaN Na	log_seasonal_first_difference NaN NaN NaN NaN NaN NaN NaN NaN NaN Na	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-05 2013-09-06 2013-09-07 2013-09-08 2013-09-10 2013-09-11	seasonal_first_difference	log_seasonal_first_difference NaN NaN NaN NaN NaN NaN NaN NaN NaN Na	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-05 2013-09-06 2013-09-07 2013-09-08 2013-09-10 2013-09-11 2013-09-12	seasonal_first_difference	log_seasonal_first_difference	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-06 2013-09-07 2013-09-08 2013-09-09 2013-09-10 2013-09-11 2013-09-12 2013-09-13	seasonal_first_difference	log_seasonal_first_difference	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-05 2013-09-06 2013-09-08 2013-09-09 2013-09-10 2013-09-11 2013-09-12 2013-09-13 2013-09-14	seasonal_first_difference	log_seasonal_first_difference	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-05 2013-09-06 2013-09-08 2013-09-09 2013-09-10 2013-09-11 2013-09-12 2013-09-13 2013-09-14 2013-09-15	seasonal_first_difference	log_seasonal_first_difference	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-06 2013-09-07 2013-09-08 2013-09-09 2013-09-10 2013-09-11 2013-09-12 2013-09-13 2013-09-14 2013-09-15 2013-09-16	seasonal_first_difference	log_seasonal_first_difference	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-05 2013-09-06 2013-09-07 2013-09-09 2013-09-10 2013-09-11 2013-09-11 2013-09-13 2013-09-14 2013-09-15 2013-09-16 2013-09-17	seasonal_first_difference	log_seasonal_first_difference	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-05 2013-09-06 2013-09-08 2013-09-10 2013-09-11 2013-09-11 2013-09-13 2013-09-14 2013-09-15 2013-09-16 2013-09-17 2013-09-18	seasonal_first_difference	log_seasonal_first_difference	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-05 2013-09-06 2013-09-07 2013-09-09 2013-09-10 2013-09-11 2013-09-12 2013-09-13 2013-09-15 2013-09-15 2013-09-16 2013-09-18 2013-09-18 2013-09-19	seasonal_first_difference	log_seasonal_first_difference	\
2013-09-01 2013-09-02 2013-09-03 2013-09-04 2013-09-05 2013-09-06 2013-09-08 2013-09-10 2013-09-11 2013-09-11 2013-09-13 2013-09-14 2013-09-15 2013-09-16 2013-09-17 2013-09-18	seasonal_first_difference	log_seasonal_first_difference	

2013-09-22		913838.8050	1.089394
2013-09-23		-26740.0170	-0.277989
2013-09-24		-120815.7665	-0.142276
2013-09-25		713086.4460	0.414563
2013-09-26		330677.6475	0.125095
2013-09-27		-40806.0180	-0.045452
2013-09-28		-343392.8820	0.212860
2013-09-29		-156521.9950	-0.348696
2013-09-30		901194.0185	0.151470
2013-10-10		-612539.6850	-0.166649
2013-10-11		-121959.6600	-0.061187
2013-10-12		829053.7765	0.326748
2013-10-13		-420384.0220	-0.239098
2013-10-14		816509.9795	0.298060
2013-10-15		-614401.9965	-0.200399
2013-10-16		760720.9340	0.208279
2013-10-17		-208145.6090	-0.080492
2013-10-18		-666077.2180	-0.161704
2013-10-19		507063.3720	0.224782
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2013-10-21		NaN	NaN
2013-10-22		NaN	NaN
2013-10-23		NaN	NaN
2013-10-24		NaN	NaN
2013-10-25		NaN	NaN
2013-10-26		NaN	NaN
2013-10-27		NaN	NaN
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2013-10-31		NaN	NaN
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2013-11-06		NaN	NaN
2013-11-07		NaN	NaN
2013-11-08		NaN	NaN
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2013-09-02	NaN		
2013-09-03	NaN		
2013-09-04	NaN		
2013-09-05	NaN		
2013-09-06	NaN		

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3.987888e+06 4.267343e+06 4.824704e+06 3.663981e+06
3.987888e+06 4.267343e+06 4.824704e+06 3.663981e+06 2.871038e+06
3.987888e+06 4.267343e+06 4.824704e+06 3.663981e+06 2.871038e+06 2.621852e+06
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      2013-11-03
      NaN

      2013-11-04
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      2013-11-05
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      2013-11-06
      NaN

      2013-11-07
      NaN

      2013-11-08
      NaN
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[69 rows x 8 columns]

```
In [23]: data['forecast']['2013-10-20':'2013-10-30'].sum()
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Out[23]: 41557108.592442185