

CME_Tick_Changes_JPY

October 9, 2019

1 The Robert and Rosenbaum Uncertainty Zones model

2 An application to EURUSD FX Futures at CME

2.1 Implementation by

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2.3 and

2.4 Florian Huchedé (CME)

2.5 Aug-2019

2.6 Import packages

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import statsmodels.api as sm
import glob
```

```
[2]: pd.set_option('display.max_columns', 50)
```

```
[3]: pd.set_option('display.max_rows', 200)
```

```
[4]: import cme_processing as cme
```

2.7 File paths and initial values

```
[5]: PATHPROJ = '/Users/marcoscscarreira/Documents/X/CME project/CME_data/'
URL_ROOT = 'https://raw.githubusercontent.com/MarcosCarreira/UZStats/master/
↳CME_data/'
```

```
[6]: CURR = 'JPY'
```

```
[7]: PATH_PRIOR = PATHPROJ+CURR+'/prior/'  
PATH_AFTER = PATHPROJ+CURR+'/after/'  
URL_1 = CURR+'/prior/'  
URL_2 = CURR+'/after/'  
#PATH_PRIOR = URL_ROOT+URL_1  
#PATH_AFTER = URL_ROOT+URL_2
```

```
[8]: TRADING_HOURS = 8
```

```
[9]: TICK_PRIOR = 1.0  
TICK_AFTER = 0.5
```

```
[10]: PRIOR_CDATES_LIST = [['6JH5', '20150105'], ['6JH5', '20150106'], ['6JH5', '20150107'],\  
    ↪ '20150108'], ['6JH5', '20150109'], ['6JH5', '20150110'], ['6JH5', '20150111'], ['6JH5', '20150112'], ['6JH5', '20150113'],\  
    ↪ '20150114'], ['6JH5', '20150115'], ['6JH5', '20150116'], ['6JH5', '20150117'], ['6JH5', '20150118'], ['6JH5', '20150119'],\  
    ↪ '20150120'], ['6JH5', '20150121'], ['6JH5', '20150122'], ['6JH5', '20150123'], ['6JH5', '20150124'], ['6JH5', '20150125'],\  
    ↪ '20150126'], ['6JH5', '20150127'], ['6JH5', '20150128'], ['6JH5', '20150129'], ['6JH5', '20150130'], ['6JH5', '20150131'],\  
    ↪ '20150201'], ['6JH5', '20150202'], ['6JH5', '20150203'], ['6JH5', '20150204'], ['6JH5', '20150205'], ['6JH5', '20150206'],\  
    ↪ '20150207'], ['6JH5', '20150208'], ['6JH5', '20150209'], ['6JH5', '20150210'], ['6JH5', '20150211'], ['6JH5', '20150212'],\  
    ↪ '20150213'], ['6JH5', '20150214'], ['6JH5', '20150215'], ['6JH5', '20150216'], ['6JH5', '20150217'], ['6JH5', '20150218'],\  
    ↪ '20150219'], ['6JH5', '20150220'], ['6JH5', '20150221'], ['6JH5', '20150222'], ['6JH5', '20150223'], ['6JH5', '20150224'],\  
    ↪ '20150225'], ['6JH5', '20150226'], ['6JH5', '20150227'], ['6JH5', '20150228'], ['6JH5', '20150229'], ['6JH5', '20150301'],\  
    ↪ '20150302'], ['6JH5', '20150303'], ['6JH5', '20150304'], ['6JH5', '20150305'], ['6JH5', '20150306'], ['6JH5', '20150307'],\  
    ↪ '20150308'], ['6JH5', '20150309'], ['6JH5', '20150310'], ['6JH5', '20150311'], ['6JH5', '20150312'], ['6JH5', '20150313'],\  
    ↪ '20150314'], ['6JM5', '20150315'], ['6JM5', '20150316'], ['6JM5', '20150317'], ['6JM5', '20150318'], ['6JM5', '20150319'],\  
    ↪ '20150320'], ['6JM5', '20150321'], ['6JM5', '20150322'], ['6JM5', '20150323'], ['6JM5', '20150324'], ['6JM5', '20150325'],\  
    ↪ '20150326'], ['6JM5', '20150327'], ['6JM5', '20150328'], ['6JM5', '20150329'], ['6JM5', '20150330'], ['6JM5', '20150331'],\  
    ↪ '20150401'], ['6JM5', '20150402'], ['6JM5', '20150403'], ['6JM5', '20150404'], ['6JM5', '20150405'], ['6JM5', '20150406'],\  
    ↪ '20150407'], ['6JM5', '20150408'], ['6JM5', '20150409'], ['6JM5', '20150410'], ['6JM5', '20150411'], ['6JM5', '20150412'],\  
    ↪ '20150413'], ['6JM5', '20150414'], ['6JM5', '20150415'], ['6JM5', '20150416'], ['6JM5', '20150417'], ['6JM5', '20150418'],\  
    ↪ '20150419'], ['6JM5', '20150420'], ['6JM5', '20150421'], ['6JM5', '20150422'], ['6JM5', '20150423'], ['6JM5', '20150424'],\  
    ↪ '20150425'], ['6JM5', '20150426'], ['6JM5', '20150427'], ['6JM5', '20150428'], ['6JM5', '20150429'], ['6JM5', '20150430']]
```

```

    ['6JM5', '20150409'], ['6JM5', '20150410'], ['6JM5', '20150413'], ['6JM5', '20150414'],\
    ['6JM5', '20150415'], ['6JM5', '20150416'], ['6JM5', '20150417'], ['6JM5', '20150420'],\
    ['6JM5', '20150421'], ['6JM5', '20150422'], ['6JM5', '20150423'], ['6JM5', '20150424'],\
    ['6JM5', '20150427'], ['6JM5', '20150428'], ['6JM5', '20150429'], ['6JM5', '20150430'],\
    ['6JM5', '20150501'], ['6JM5', '20150504'], ['6JM5', '20150505'], ['6JM5', '20150506'],\
    ['6JM5', '20150507'], ['6JM5', '20150508'], ['6JM5', '20150511'], ['6JM5', '20150512'],\
    ['6JM5', '20150513'], ['6JM5', '20150514'], ['6JM5', '20150515'], ['6JM5', '20150518'],\
    ['6JM5', '20150519'], ['6JM5', '20150520'], ['6JM5', '20150521'], ['6JM5', '20150522'],\
    ['6JM5', '20150525'], ['6JM5', '20150526'], ['6JM5', '20150527'], ['6JM5', '20150528'],\
    ['6JM5', '20150529'], ['6JM5', '20150601'], ['6JM5', '20150602'], ['6JM5', '20150603'],\
    ['6JM5', '20150604'], ['6JM5', '20150605'], ['6JM5', '20150608'], ['6JM5', '20150609'],\
    ['6JM5', '20150610'], ['6JM5', '20150611'], ['6JU5', '061215'], ['6JU5', '061515'],\
    ['6JU5', '061615'], ['6JU5', '061715'], ['6JU5', '061815'], ['6JU5', '061915']]

```

```

[11]: AFTER_CDATES_LIST = [['6JU5', '062215'], ['6JU5', '062315'], ['6JU5', '062415'],\
    ['6JU5', '062515'], ['6JU5', '062615'], ['6JU5', '062915'], ['6JU5', '063015'],\
    ['6JU5', '070115'], ['6JU5', '070215'], ['6JU5', '070315'], ['6JU5', '070615'],\
    ['6JU5', '070715'], ['6JU5', '070815'], ['6JU5', '070915'], ['6JU5', '071015'],\
    ['6JU5', '20150713'], ['6JU5', '20150714'], ['6JU5', '20150715'], ['6JU5', '20150716'],\
    ['6JU5', '20150717'], ['6JU5', '20150720'], ['6JU5', '20150721'], ['6JU5', '20150722'],\
    ['6JU5', '20150723'], ['6JU5', '20150724'], ['6JU5', '20150727'], ['6JU5', '20150728'],\
    ['6JU5', '20150729'], ['6JU5', '20150730'], ['6JU5', '20150731'], ['6JU5', '20150803'],\
    ['6JU5', '20150804'], ['6JU5', '20150805'], ['6JU5', '20150806'], ['6JU5', '20150807']]

```

```

    ['6JU5', '20150810'], ['6JU5', '20150811'], ['6JU5', '20150812'], ['6JU5', '20150813'], \
    ['6JU5', '20150814'], ['6JU5', '20150817'], ['6JU5', '20150818'], ['6JU5', '20150819'], \
    ['6JU5', '20150820'], ['6JU5', '20150821'], ['6JU5', '20150824'], ['6JU5', '20150825'], \
    ['6JU5', '20150826'], ['6JU5', '20150827'], ['6JU5', '20150828'], ['6JU5', '20150831'], \
    ['6JU5', '20150901'], ['6JU5', '20150902'], ['6JU5', '20150903'], ['6JU5', '20150904'], \
    ['6JU5', '20150907'], ['6JU5', '20150908'], ['6JU5', '20150909'], ['6JU5', '20150910'], \
    ['6JU5', '20150911'], ['6JZ5', '20150914'], ['6JZ5', '20150915'], ['6JZ5', '20150916'], \
    ['6JZ5', '20150917'], ['6JZ5', '20150918'], ['6JZ5', '20150921'], ['6JZ5', '20150922'], \
    ['6JZ5', '20150923'], ['6JZ5', '20150924'], ['6JZ5', '20150925'], ['6JZ5', '20150928'], \
    ['6JZ5', '20150929'], ['6JZ5', '20150930'], ['6JZ5', '20151001'], ['6JZ5', '20151002'], \
    ['6JZ5', '20151005'], ['6JZ5', '20151006'], ['6JZ5', '20151007'], ['6JZ5', '20151008'], \
    ['6JZ5', '20151009'], ['6JZ5', '20151012'], ['6JZ5', '20151013'], ['6JZ5', '20151014'], \
    ['6JZ5', '20151015'], ['6JZ5', '20151016'], ['6JZ5', '20151019'], ['6JZ5', '20151020'], \
    ['6JZ5', '20151021'], ['6JZ5', '20151022'], ['6JZ5', '20151023'], ['6JZ5', '20151026'], \
    ['6JZ5', '20151027'], ['6JZ5', '20151028'], ['6JZ5', '20151029'], ['6JZ5', '20151030'], \
    ['6JZ5', '20151102'], ['6JZ5', '20151103'], ['6JZ5', '20151104'], ['6JZ5', '20151105'], \
    ['6JZ5', '20151106'], ['6JZ5', '20151109'], ['6JZ5', '20151110'], ['6JZ5', '20151111'], \
    ['6JZ5', '20151112'], ['6JZ5', '20151113'], ['6JZ5', '20151116'], ['6JZ5', '20151117'], \
    ['6JZ5', '20151118'], ['6JZ5', '20151119'], ['6JZ5', '20151120'], ['6JZ5', '20151123'], \
    ['6JZ5', '20151124'], ['6JZ5', '20151125'], ['6JZ5', '20151126'], ['6JZ5', '20151127'], \
    ['6JZ5', '20151130'], ['6JZ5', '20151201'], ['6JZ5', '20151202'], ['6JZ5', '20151203'], \
    ['6JZ5', '20151204'], ['6JZ5', '20151207'], ['6JZ5', '20151208'], ['6JZ5', '20151209'], \
    ['6JZ5', '20151210'], ['6JZ5', '20151211']]

```

2.7.1 Processing files

Prior

```
[12]: #PRIOR_CDATES_LIST = cme.list_files(PATH_PRIOR)

[13]: #PRIOR_CDATES_LIST

[14]: PRIOR_CDATES, FILES_PRIOR_CATicks, FILES_PRIOR_COSTtrades,\
      FILES_PRIOR_OBstats, FILES_PRIOR_OTtrans,\
      FILES_PRIOR_RDFtrans, FILES_PRIOR_UZstats = \
      cme.process_files(PATH_PRIOR, PRIOR_CDATES_LIST, 'prior', TICK_PRIOR)

[15]: PRIOR_OB_UZ_STATS = cme.ob_uz_stats(PRIOR_CDATES, FILES_PRIOR_OBstats,\
      FILES_PRIOR_UZstats, FILES_PRIOR_CATicks, TRADING_HOURS)

[16]: PRIOR_IMBAL_STATS = cme.imbal_stats(PRIOR_CDATES, FILES_PRIOR_OTtrans)

[17]: PRIOR_IMBAL_STATS_TS = cme.time_series_imbal(PRIOR_IMBAL_STATS, pd.
      ↪to_datetime(PRIOR_CDATES['Date']), 'prior')

[18]: PRIOR_IMBAL_STATS_TS['eta1'] = PRIOR_OB_UZ_STATS['eta1'].values

[19]: PRIOR_TRADE_STATS_TS = cme.time_series_imbal_trd(PRIOR_IMBAL_STATS, pd.
      ↪to_datetime(PRIOR_CDATES['Date']), 'prior')

[20]: PRIOR_DEPL_STATS = cme.depl_stats(PRIOR_CDATES, FILES_PRIOR_RDFtrans)

[21]: PRIOR_DEPL_STATS_TS = cme.time_series_depl(PRIOR_DEPL_STATS, pd.
      ↪to_datetime(PRIOR_CDATES['Date']), 'prior')

[22]: PRIOR_DEPL_STATS_TS['eta1'] = PRIOR_OB_UZ_STATS['eta1'].values

[23]: PRIOR_ABSDEPL_STATS_TS = cme.time_series_absdepl(PRIOR_DEPL_STATS, pd.
      ↪to_datetime(PRIOR_CDATES['Date']), 'prior')

[24]: PRIOR_ABSDEPL_STATS_TS['eta1'] = PRIOR_OB_UZ_STATS['eta1'].values
      PRIOR_ABSDEPL_STATS_TS['M'] = PRIOR_OB_UZ_STATS['M'].values

[25]: PRIOR_COST_STATS = cme.cost_stats(PRIOR_CDATES, FILES_PRIOR_COSTtrades)

[26]: PRIOR_COST_STATS['Status'] = 'prior'
```

After

```
[27]: #AFTER_CDATES_LIST = cme.list_files(PATH_AFTER)

[28]: #AFTER_CDATES_LIST
```

```

[29]: AFTER_CDATES, FILES_AFTER_CATicks, FILES_AFTER_COSTtrades,\
      FILES_AFTER_OBstats, FILES_AFTER_OTtrans,\
      FILES_AFTER_RDFtrans, FILES_AFTER_UZstats = \
      cme.process_files(PATH_AFTER, AFTER_CDATES_LIST, 'after', TICK_AFTER)

[30]: AFTER_OB_UZ_STATS = cme.ob_uz_stats(AFTER_CDATES, FILES_AFTER_OBstats,\
      FILES_AFTER_UZstats, FILES_AFTER_CATicks, TRADING_HOURS)

[31]: AFTER_IMBAL_STATS = cme.imbal_stats(AFTER_CDATES, FILES_AFTER_OTtrans)

[32]: AFTER_IMBAL_STATS_TS = cme.time_series_imbal(AFTER_IMBAL_STATS, pd.
      ↪to_datetime(AFTER_CDATES['Date']), 'after')

[33]: AFTER_IMBAL_STATS_TS['eta1'] = AFTER_OB_UZ_STATS['eta1'].values

[34]: AFTER_TRADE_STATS_TS = cme.time_series_imbal_trd(AFTER_IMBAL_STATS, pd.
      ↪to_datetime(AFTER_CDATES['Date']), 'after')

[35]: AFTER_DEPL_STATS = cme.depl_stats(AFTER_CDATES, FILES_AFTER_RDFtrans)

[36]: AFTER_DEPL_STATS_TS = cme.time_series_depl(AFTER_DEPL_STATS, pd.
      ↪to_datetime(AFTER_CDATES['Date']), 'after')

[37]: AFTER_DEPL_STATS_TS['eta1'] = AFTER_OB_UZ_STATS['eta1'].values

[38]: AFTER_ABSDEPL_STATS_TS = cme.time_series_absdepl(AFTER_DEPL_STATS, pd.
      ↪to_datetime(AFTER_CDATES['Date']), 'after')

[39]: AFTER_ABSDEPL_STATS_TS['eta1'] = AFTER_OB_UZ_STATS['eta1'].values
      AFTER_ABSDEPL_STATS_TS['M'] = AFTER_OB_UZ_STATS['M'].values

[40]: AFTER_COST_STATS = cme.cost_stats(AFTER_CDATES, FILES_AFTER_COSTtrades)

[41]: AFTER_COST_STATS['Status'] = 'after'

```

Join prior and after

```

[42]: OB_UZ_STATS = pd.concat([PRIOR_OB_UZ_STATS, AFTER_OB_UZ_STATS], sort=False)

[43]: IMBAL_STATS_TS = pd.concat([PRIOR_IMBAL_STATS_TS, AFTER_IMBAL_STATS_TS],
      ↪sort=False)

[44]: TRADE_STATS_TS = pd.concat([PRIOR_TRADE_STATS_TS, AFTER_TRADE_STATS_TS],
      ↪sort=False)

```

```
[45]: DEPL_STATS_TS = pd.concat([PRIOR_DEPL_STATS_TS, AFTER_DEPL_STATS_TS],  
    ↪sort=False)
```

```
[46]: ABSDEPL_STATS_TS = pd.concat([PRIOR_ABSDEPL_STATS_TS, AFTER_ABSDEPL_STATS_TS],  
    ↪sort=False)
```

2.7.2 Tables

```
[47]: TABLE_MATHIEU = cme.table_mathieu(OB_UZ_STATS)  
TABLE_MATHIEU_ERR = cme.table_mathieu_err(OB_UZ_STATS)
```

```
[48]: TABLE_MATHIEU
```

```
[48]:      Tick  chgavg  ndfpr_pred      ndfpr          M      Volume \  
Status  
prior    1.0  1.02531  2539.57754  1993.7913  13109.66957  75510.02609  
after     0.5  0.52336  6152.96473  3809.2800  17209.84000  68863.32000  
  
      eta1      S1  lambda1  twspr1  duration  dt_avg  rvxe \  
Status  
prior    0.23994  0.98738  0.98544  1.04473  16.91766  17.73267  0.00400  
after     0.33870  0.96178  0.97067  1.16956   9.82449  10.58436  0.00348  
  
      spot_avg  
Status  
prior    8330.35714  
after     8205.53733
```

```
[49]: TABLE_MATHIEU_ERR
```

```
[49]:      Tick  chgavg  ndfpr_pred      ndfpr          M      Volume \  
Status  
prior    0.0  0.04983  2079.37315  1024.36377  6368.99859  32942.32568  
after     0.0  0.03153  10626.03777  3207.30554  9190.30675  34884.79666  
  
      eta1      S1  lambda1  twspr1  duration  dt_avg  rvxe \  
Status  
prior    0.05823  0.01332  0.01780  0.04863  11.28649  8.49775  0.00175  
after     0.06312  0.03516  0.03298  0.12921   9.44989  6.71907  0.00206  
  
      spot_avg  
Status  
prior    135.75089  
after     119.74845
```

```
[50]: cme.avg_perc_mat(PRIOR_IMBAL_STATS, pd.to_datetime(PRIOR_CDATES['Date']))
```

```
[50]:
```

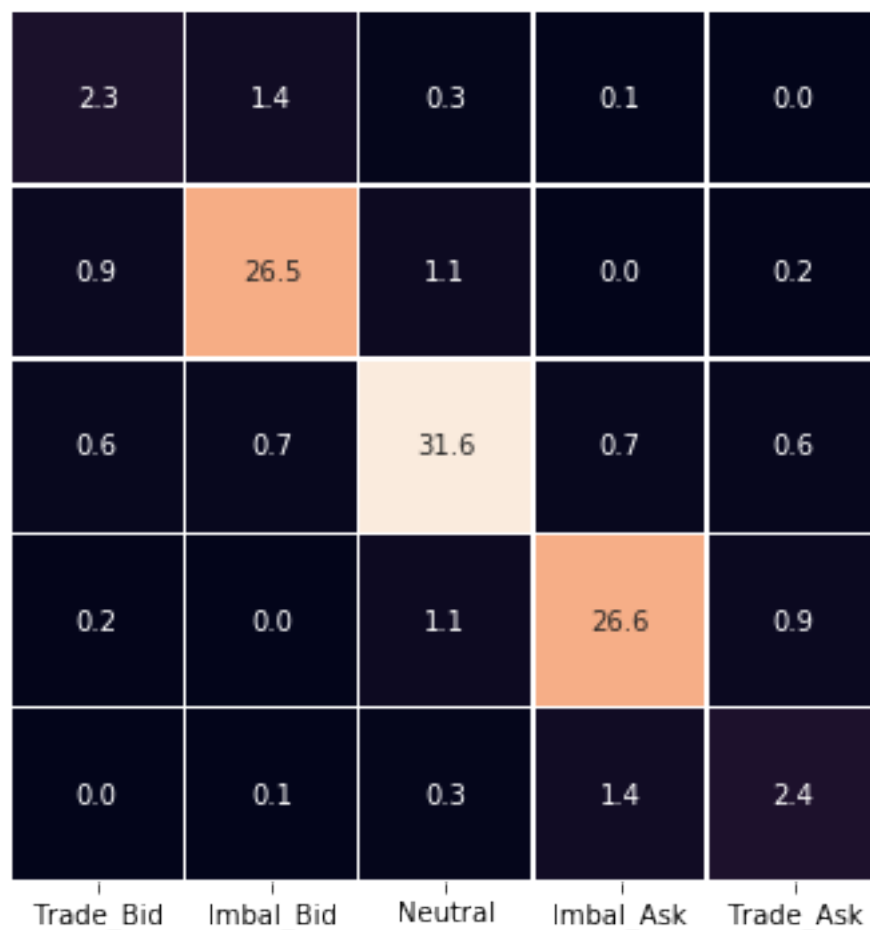
	Trade_Bid	Imbal_Bid	Neutral	Imbal_Ask	Trade_Ask	Total Cols
Trade_Bid	2.33	1.39	0.27	0.08	0.00	4.07
Imbal_Bid	0.88	26.48	1.09	0.04	0.23	28.72
Neutral	0.62	0.74	31.63	0.73	0.60	34.32
Imbal_Ask	0.23	0.04	1.08	26.57	0.88	28.81
Trade_Ask	0.00	0.07	0.26	1.38	2.37	4.08
Total Rows	4.07	28.72	34.32	28.81	4.08	100.00

```
[51]: cme.avg_perc_mat(AFTER_IMBAL_STATS, pd.to_datetime(AFTER_CDATES['Date']))
```

```
[51]:
```

	Trade_Bid	Imbal_Bid	Neutral	Imbal_Ask	Trade_Ask	Total Cols
Trade_Bid	0.91	1.06	0.28	0.09	0.00	2.34
Imbal_Bid	0.61	24.46	1.64	0.16	0.24	27.11
Neutral	0.58	1.35	37.09	1.35	0.57	40.94
Imbal_Ask	0.25	0.16	1.64	24.61	0.60	27.25
Trade_Ask	0.00	0.09	0.28	1.05	0.93	2.35
Total Rows	2.34	27.11	40.94	27.25	2.35	100.00

```
[52]: AVG_IMBAL_PRIOR = cme.avg_perc_mat(PRIOR_IMBAL_STATS, pd.
      ↪to_datetime(PRIOR_CDATES['Date']))
plt.figure(figsize=(9, 6))
sns.heatmap(AVG_IMBAL_PRIOR.iloc[:-1].drop(columns=['Total Cols']),\
            annot=True, fmt=".1f",\
            linewidths=.5, square=True,\
            xticklabels=True,\
            yticklabels=False,\
            cbar=False);
```

```
[53]: cme.avg_perc_mat_2(PRIOR_DEPL_STATS, pd.to_datetime(PRIOR_CDATES['Date']))
```

```
[53]:
```

	same				oppo				Total Cols
	D C	D T	D T+F	F	D C	D T	D T+F	F	
D C	0.02	0.03	0.11	17.36	0.02	0.01	2.32	0.59	20.48
D T	0.01	0.11	0.52	18.07	0.02	0.02	3.81	2.72	25.30
D T+F	0.01	0.02	0.24	5.77	0.01	0.01	1.54	0.95	8.55
F	12.54	8.70	0.00	0.10	7.84	16.40	0.00	0.10	45.67
Total Rows	12.59	8.86	0.88	41.30	7.89	16.44	7.68	4.37	100.00

```
[54]: cme.avg_perc_mat_2(AFTER_DEPL_STATS, pd.to_datetime(AFTER_CDATES['Date']))
```

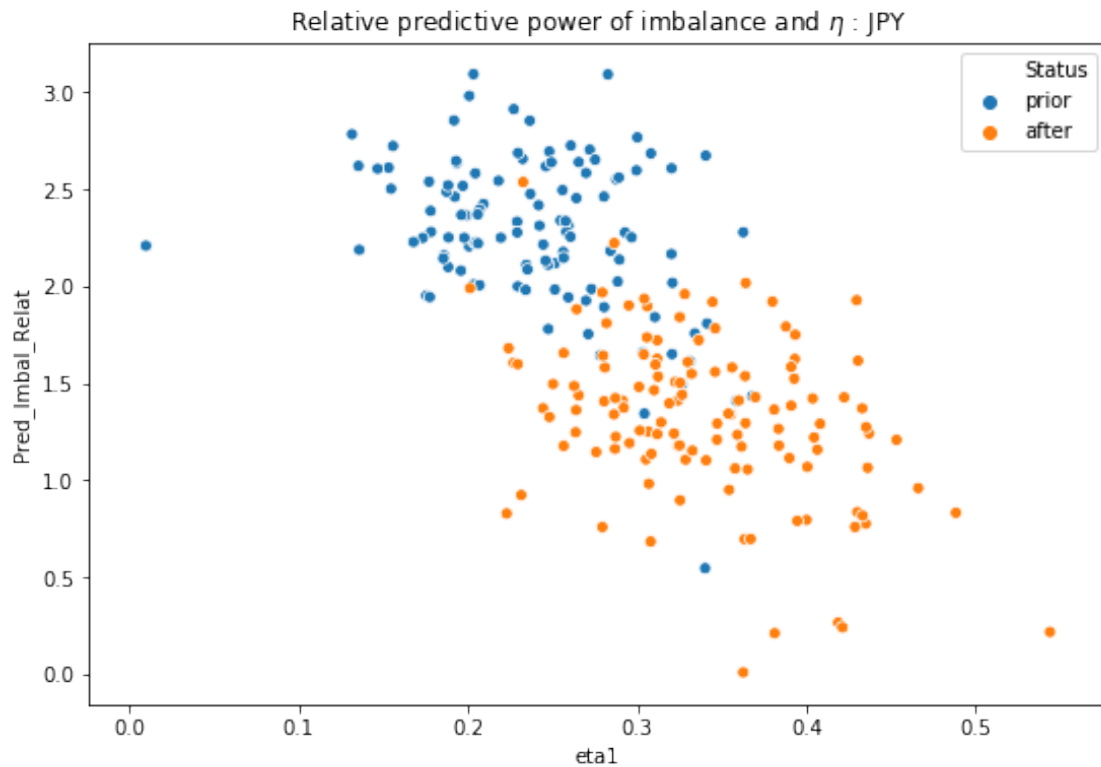
```
[54]:
```

	same				oppo				Total Cols
	D C	D T	D T+F	F	D C	D T	D T+F	F	
D C	0.04	0.05	0.15	23.48	0.03	0.02	1.24	1.49	26.51
D T	0.02	0.14	0.49	13.94	0.05	0.04	2.28	4.14	21.08
D T+F	0.02	0.03	0.19	3.12	0.02	0.02	0.61	0.97	4.96
F	19.04	7.43	0.00	0.16	7.29	13.37	0.00	0.15	47.44

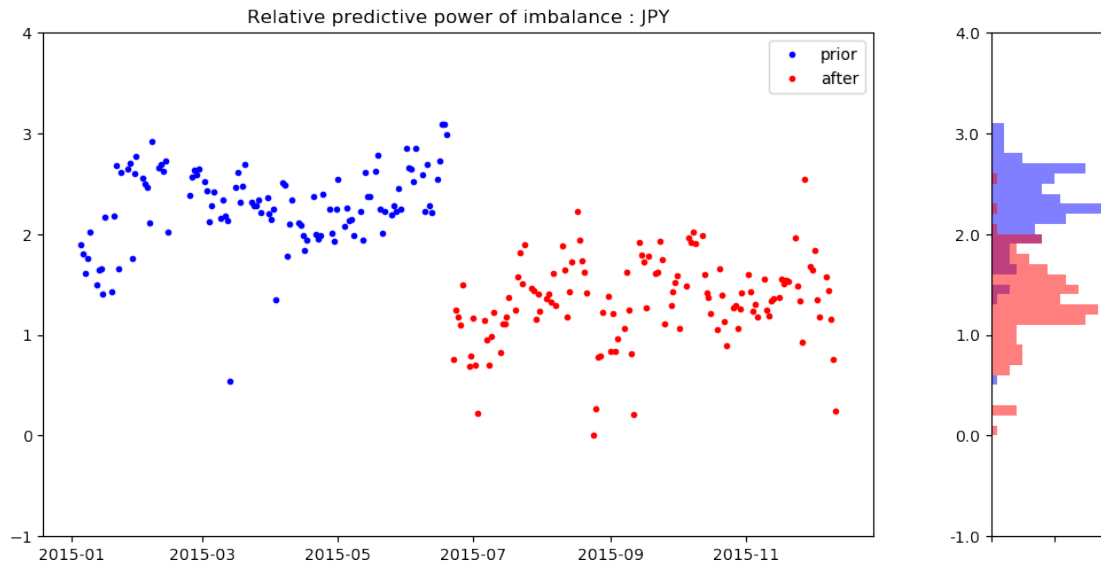
Total Rows 19.13 7.65 0.83 40.69 7.38 13.44 4.13 6.75 100.00

2.8 Charts and Regressions

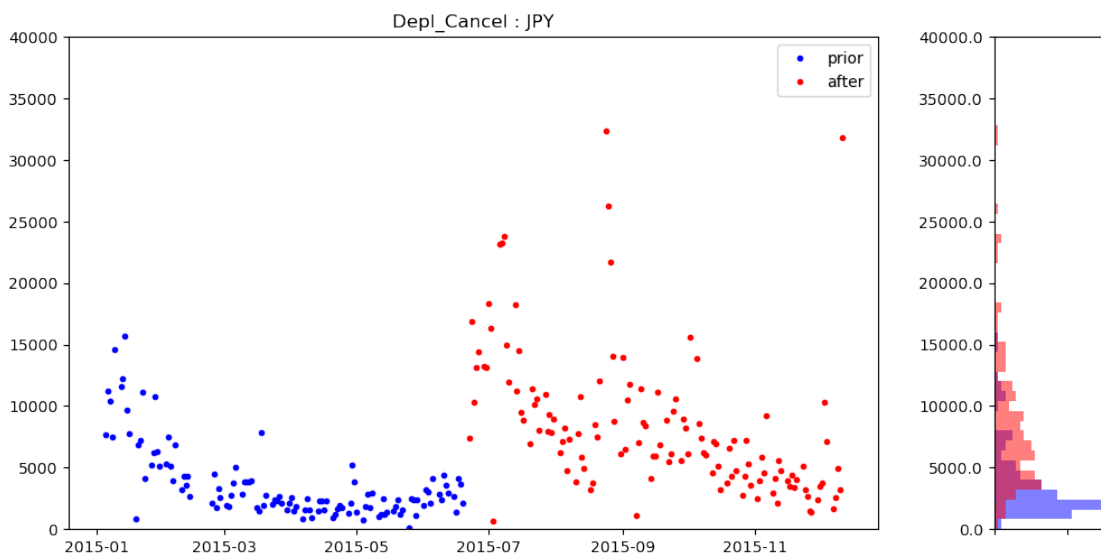
```
[55]: plt.figure(figsize=(9, 6))
sns.scatterplot(x='eta1', y='Pred_Imbal_Relat', hue='Status',\
               data=IMBAL_STATS_TS);
plt.title('Relative predictive power of imbalance and  $\eta$  : '+CURR);
```



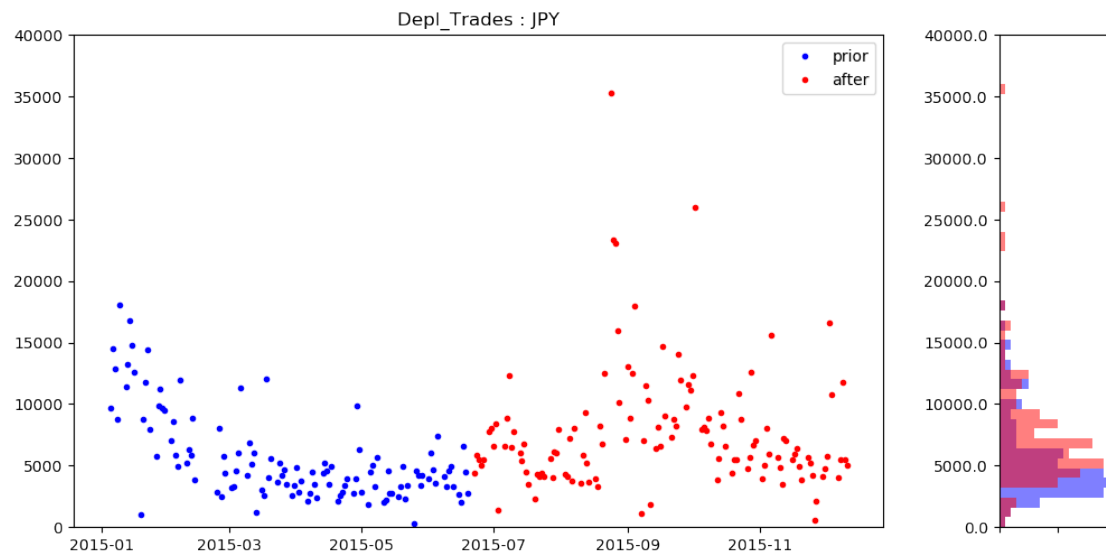
```
[164]: cme.time_series_hist_plot(IMBAL_STATS_TS, 'Pred_Imbal_Relat',\
                                'Relative predictive power of imbalance : '+CURR, -1.0, 4.0, 50)
```



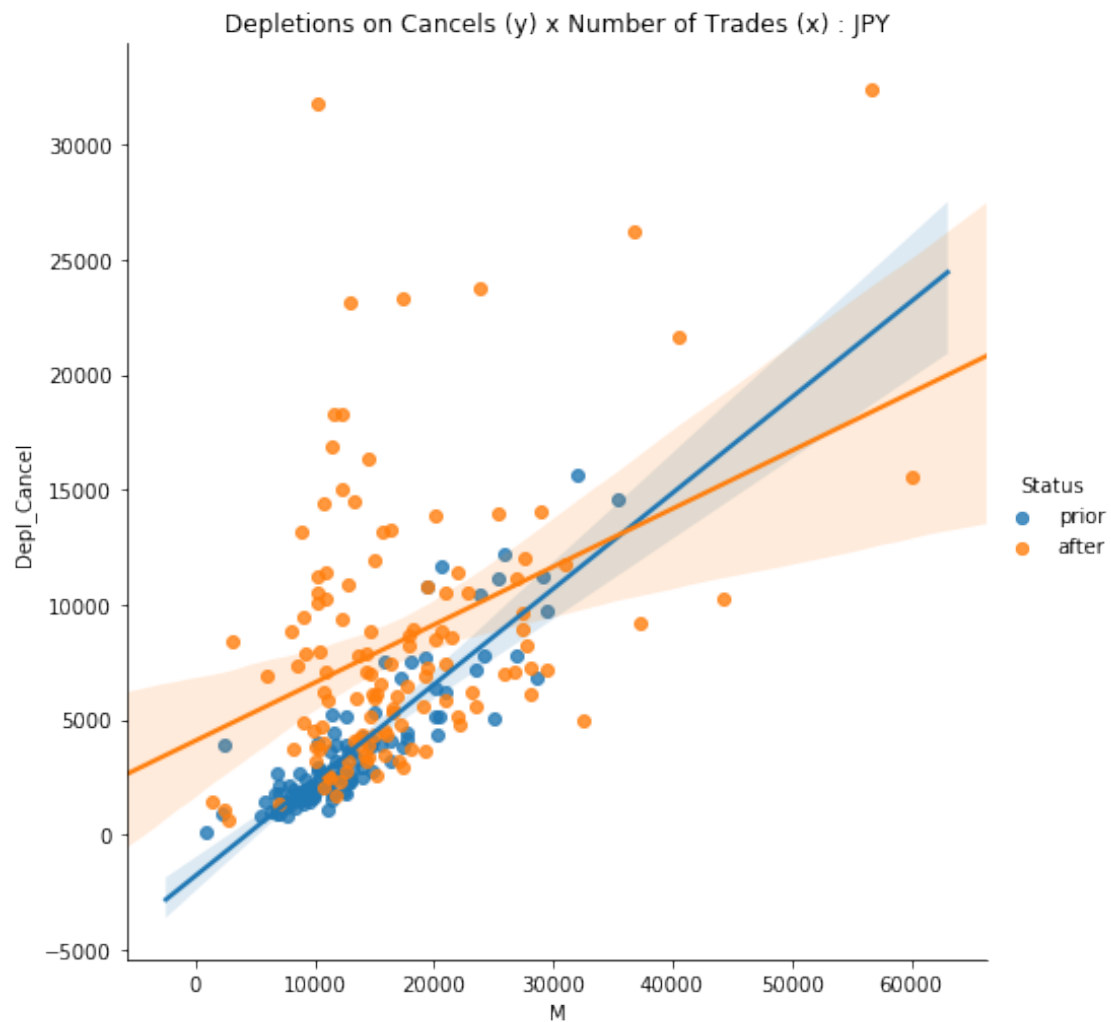
```
[165]: cme.time_series_hist_plot(ABSDEPL_STATS_TS, 'Depl_Cancel',\
    'Depl_Cancel : '+CURR, 0.0, 40000.0, 50)
```



```
[166]: cme.time_series_hist_plot(ABSDEPL_STATS_TS, 'Depl_Trades',\
    'Depl_Trades : '+CURR, 0.0, 40000.0, 50)
```



```
[59]: cme.regr_plot(ABSDEPL_STATS_TS, 'M', 'Depl_Cancel',\
    'Depletions on Cancels (y) x Number of Trades (x) : '+CURR)
```



```
[60]: cme.regr_plot(ABSDEPL_STATS_TS, 'M', 'Depl_Trades',\
    'Depletions on Trades (y) x Number of Trades (x) : '+CURR)
```



```
[61]: cme.lin_reg(ABSDEPL_STATS_TS, ['M'], 'Depl_Cancel')
```

OLS Regression Results

```
=====
Dep. Variable:          Depl_Cancel    R-squared:                0.309
Model:                  OLS            Adj. R-squared:          0.306
Method:                 Least Squares   F-statistic:             106.2
Date:                   Wed, 09 Oct 2019 Prob (F-statistic):       7.88e-21
Time:                   16:10:27        Log-Likelihood:          -2352.2
No. Observations:       240            AIC:                    4708.
Df Residuals:           238            BIC:                    4715.
Df Model:                1
Covariance Type:        nonrobust
=====
```

```
=====
               coef      std err          t      P>|t|      [0.025      0.975]
-----
```

const	714.4925	598.043	1.195	0.233	-463.640	1892.625
M	0.3561	0.035	10.305	0.000	0.288	0.424

```
=====
```

Omnibus:	146.578	Durbin-Watson:	0.608
Prob(Omnibus):	0.000	Jarque-Bera (JB):	944.074
Skew:	2.452	Prob(JB):	9.93e-206
Kurtosis:	11.388	Cond. No.	3.66e+04

```
=====
```

Warnings:

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

[2] The condition number is large, 3.66e+04. This might indicate that there are strong multicollinearity or other numerical problems.

/Users/marcoscscarreira/anaconda3/envs/CondaEnv36/lib/python3.6/site-packages/numpy/core/fromnumeric.py:2495: FutureWarning: Method .ptp is deprecated and will be removed in a future version. Use numpy.ptp instead.

return ptp(axis=axis, out=out, **kwargs)

```
[62]: cme.lin_reg(ABSDEPL_STATS_TS, ['M'], 'Depl_Trades')
```

OLS Regression Results

```
=====
```

Dep. Variable:	Depl_Trades	R-squared:	0.915
Model:	OLS	Adj. R-squared:	0.915
Method:	Least Squares	F-statistic:	2558.
Date:	Wed, 09 Oct 2019	Prob (F-statistic):	2.56e-129
Time:	16:10:27	Log-Likelihood:	-2059.3
No. Observations:	240	AIC:	4123.
Df Residuals:	238	BIC:	4130.
Df Model:	1		
Covariance Type:	nonrobust		

```
=====
```

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	-1183.8240	176.517	-6.707	0.000	-1531.560	-836.088
M	0.5158	0.010	50.576	0.000	0.496	0.536

```
=====
```

Omnibus:	54.536	Durbin-Watson:	0.695
Prob(Omnibus):	0.000	Jarque-Bera (JB):	326.201
Skew:	0.708	Prob(JB):	1.47e-71
Kurtosis:	8.533	Cond. No.	3.66e+04

```
=====
```

Warnings:

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

[2] The condition number is large, 3.66e+04. This might indicate that there are strong multicollinearity or other numerical problems.

```
[63]: cme.lin_reg(PRIOR_ABSDEPL_STATS_TS, ['M'], 'Depl_Trades')
```

```

                                OLS Regression Results
=====
Dep. Variable:                  Depl_Trades    R-squared:                  0.956
Model:                          OLS           Adj. R-squared:            0.956
Method:                        Least Squares   F-statistic:               2462.
Date:                          Wed, 09 Oct 2019 Prob (F-statistic):        1.50e-78
Time:                           16:10:27      Log-Likelihood:           -923.95
No. Observations:              115           AIC:                     1852.
Df Residuals:                  113           BIC:                     1857.
Df Model:                      1
Covariance Type:               nonrobust
=====
               coef      std err          t      P>|t|      [0.025      0.975]
-----
const      -1613.3969    161.273    -10.004      0.000    -1932.908    -1293.885
M              0.5495      0.011     49.617      0.000       0.528       0.571
=====
Omnibus:                 4.271    Durbin-Watson:           1.485
Prob(Omnibus):            0.118    Jarque-Bera (JB):         4.318
Skew:                     0.241    Prob(JB):                 0.115
Kurtosis:                 3.818    Cond. No.                 3.34e+04
=====

```

Warnings:

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

[2] The condition number is large, 3.34e+04. This might indicate that there are strong multicollinearity or other numerical problems.

```
[64]: cme.lin_reg(AFTER_ABSDEPL_STATS_TS, ['M'], 'Depl_Trades')
```

```

                                OLS Regression Results
=====
Dep. Variable:                  Depl_Trades    R-squared:                  0.889
Model:                          OLS           Adj. R-squared:            0.889
Method:                        Least Squares   F-statistic:               990.1
Date:                          Wed, 09 Oct 2019 Prob (F-statistic):        1.12e-60
Time:                           16:10:27      Log-Likelihood:           -1101.0
No. Observations:              125           AIC:                     2206.
Df Residuals:                  123           BIC:                     2212.
Df Model:                      1
Covariance Type:               nonrobust
=====

```

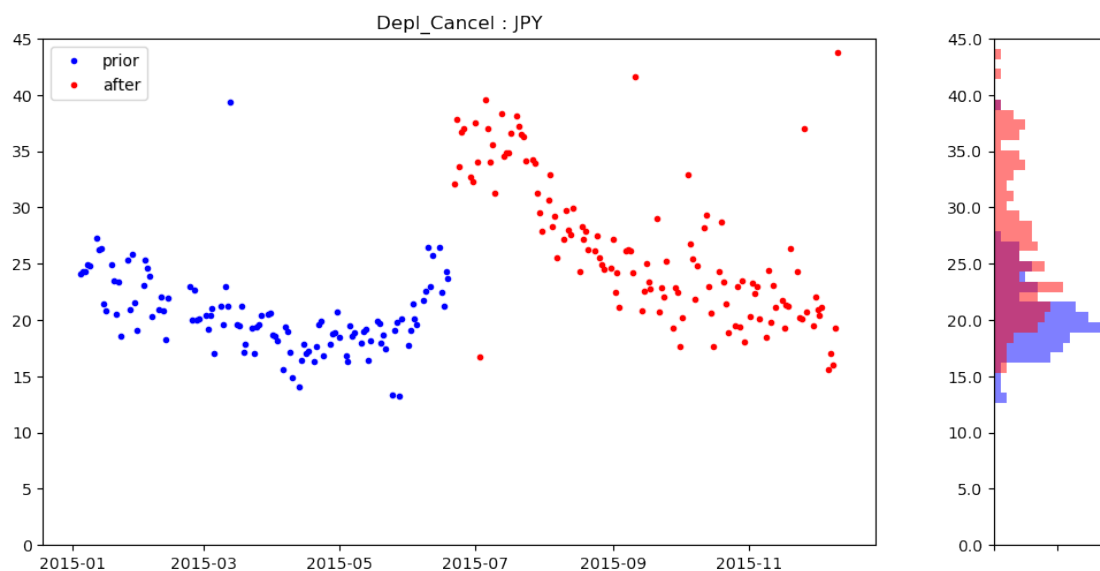

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	-947.3854	310.675	-3.049	0.003	-1562.347	-332.424
M	0.5015	0.016	31.466	0.000	0.470	0.533
=====	=====	=====	=====	=====	=====	=====
Omnibus:		38.024	Durbin-Watson:			0.596
Prob(Omnibus):		0.000	Jarque-Bera (JB):			126.173
Skew:		1.050	Prob(JB):			4.00e-28
Kurtosis:		7.452	Cond. No.			4.15e+04
=====	=====	=====	=====	=====	=====	=====

Warnings:

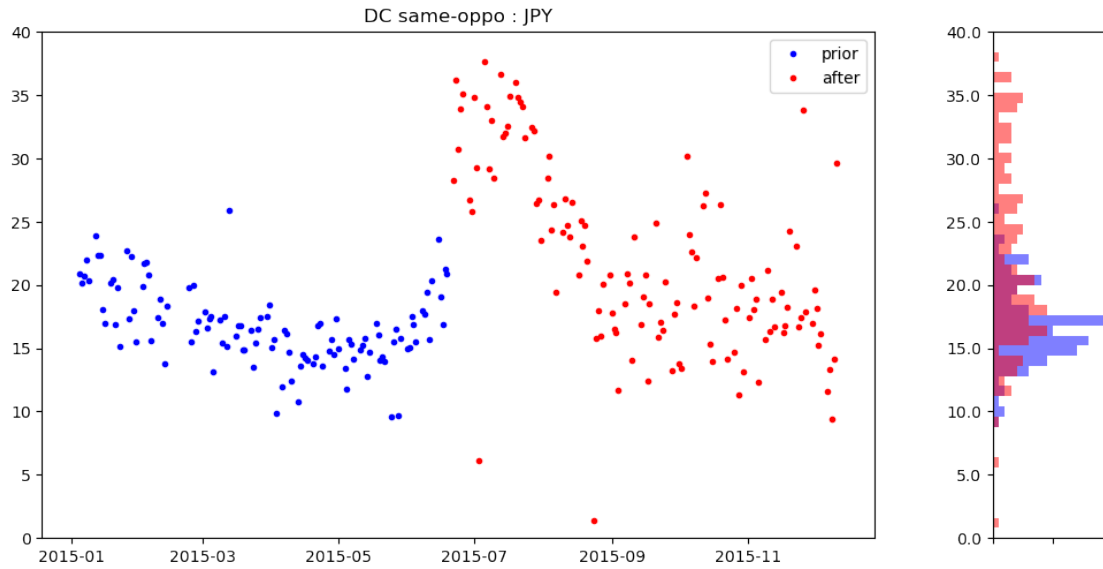
[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

[2] The condition number is large, 4.15e+04. This might indicate that there are strong multicollinearity or other numerical problems.

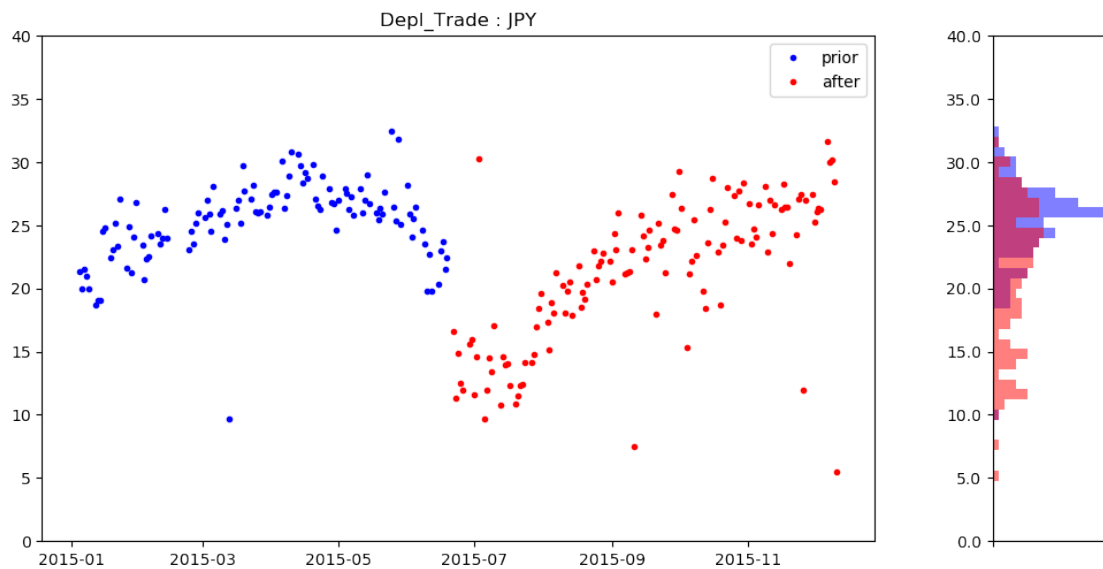
```
[170]: cme.time_series_hist_plot(DEPL_STATS_TS, 'Depl_Cancel',\
    'Depl_Cancel : '+CURR, 0, 45, 50)
```



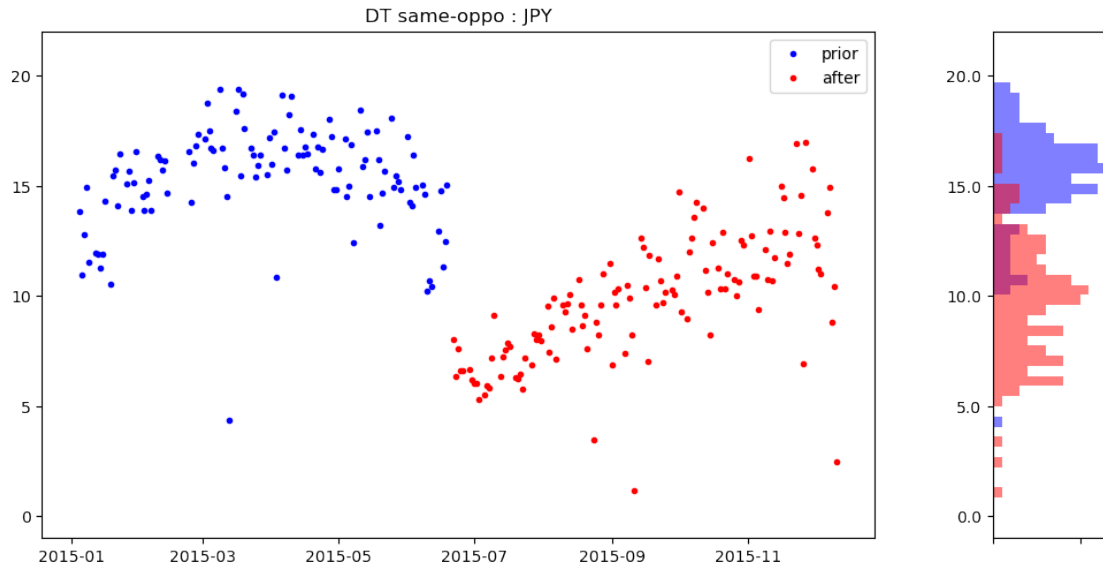
```
[172]: cme.time_series_hist_plot(DEPL_STATS_TS, 'DC same-oppo',\
    'DC same-oppo : '+CURR, 0, 40, 50)
```



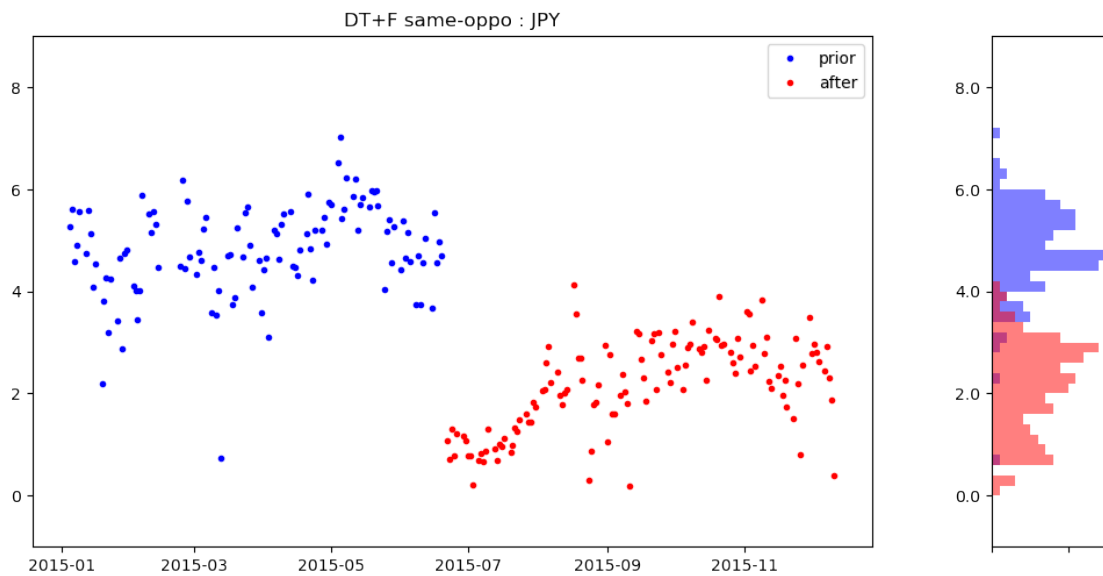
```
[67]: cme.time_series_hist_plot(DEPL_STATS_TS, 'Depl_Trade',\
    'Depl_Trade : '+CURR, 0, 40, 50)
```



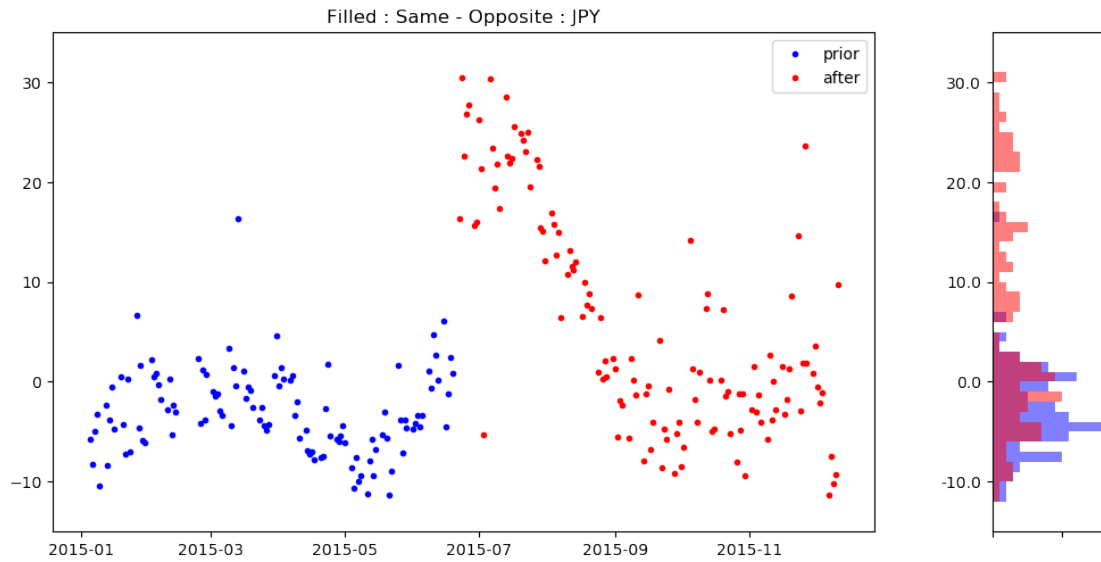
```
[176]: cme.time_series_hist_plot(DEPL_STATS_TS, 'DT same-oppo',\
    'DT same-oppo : '+CURR, -1, 22, 50)
```



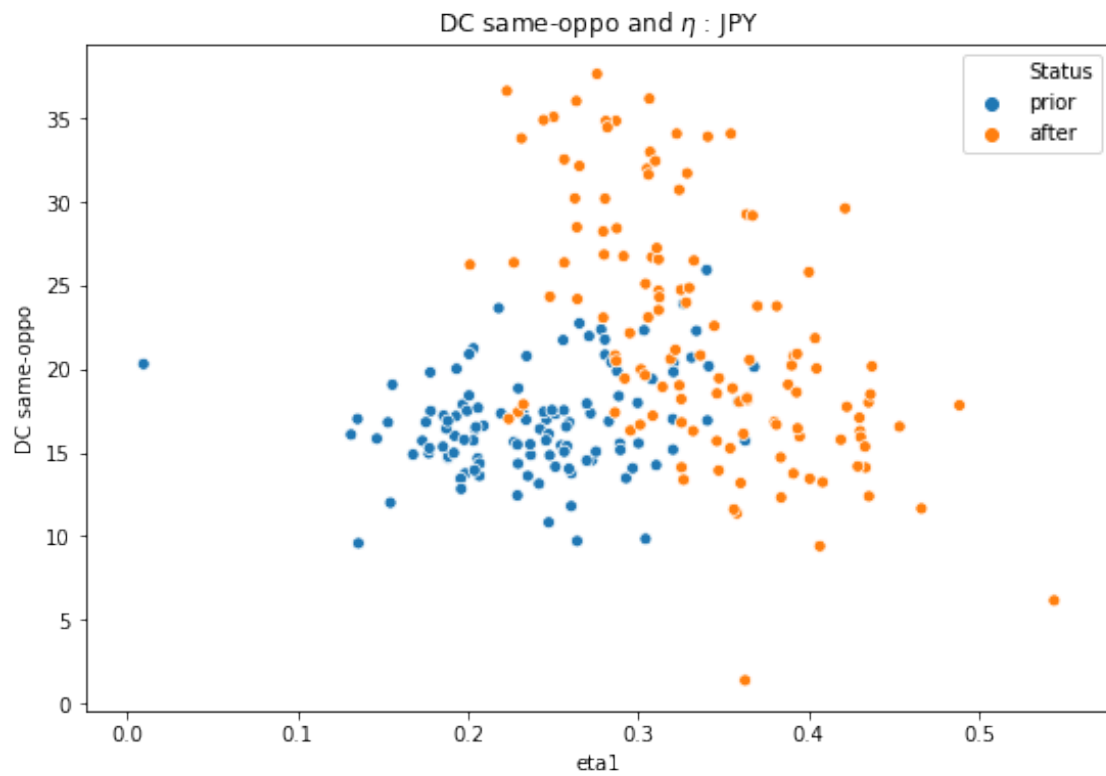
```
[178]: cme.time_series_hist_plot(DEPL_STATS_TS, 'DT+F same-oppo',\
    'DT+F same-oppo : '+CURR, -1, 9, 50)
```



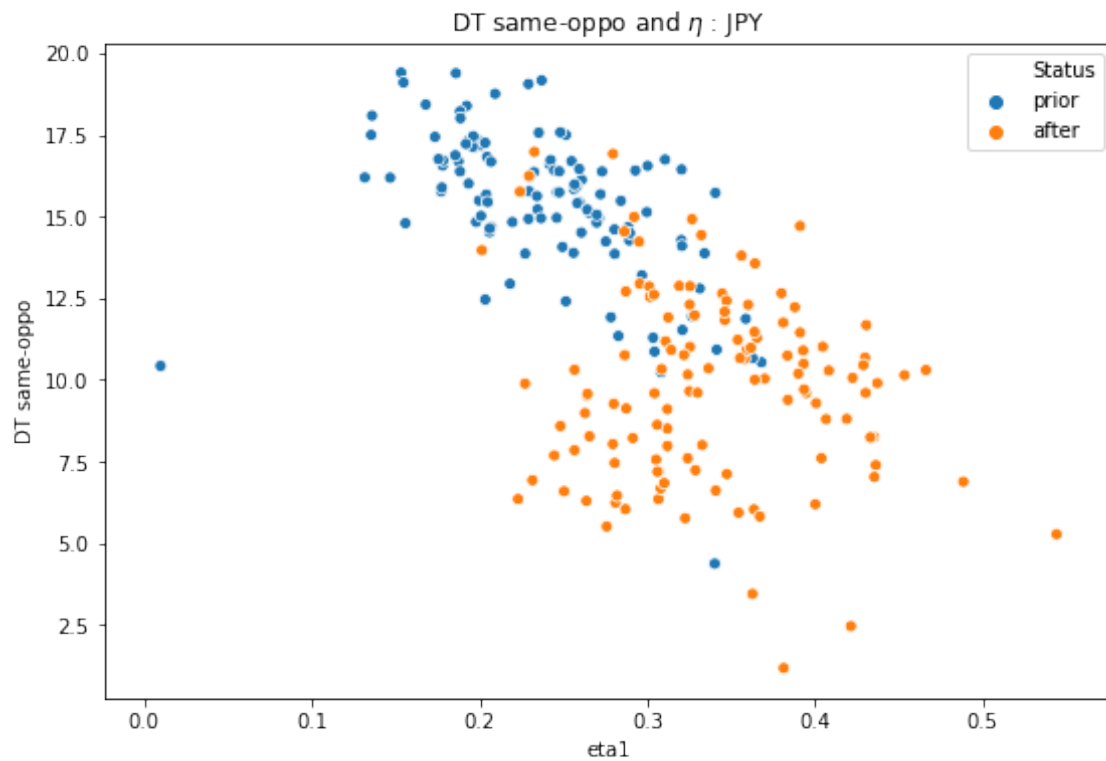
```
[181]: cme.time_series_hist_plot(DEPL_STATS_TS, 'Fill same-oppo',\
    'Filled : Same - Opposite : '+CURR, -15, 35, 50)
```



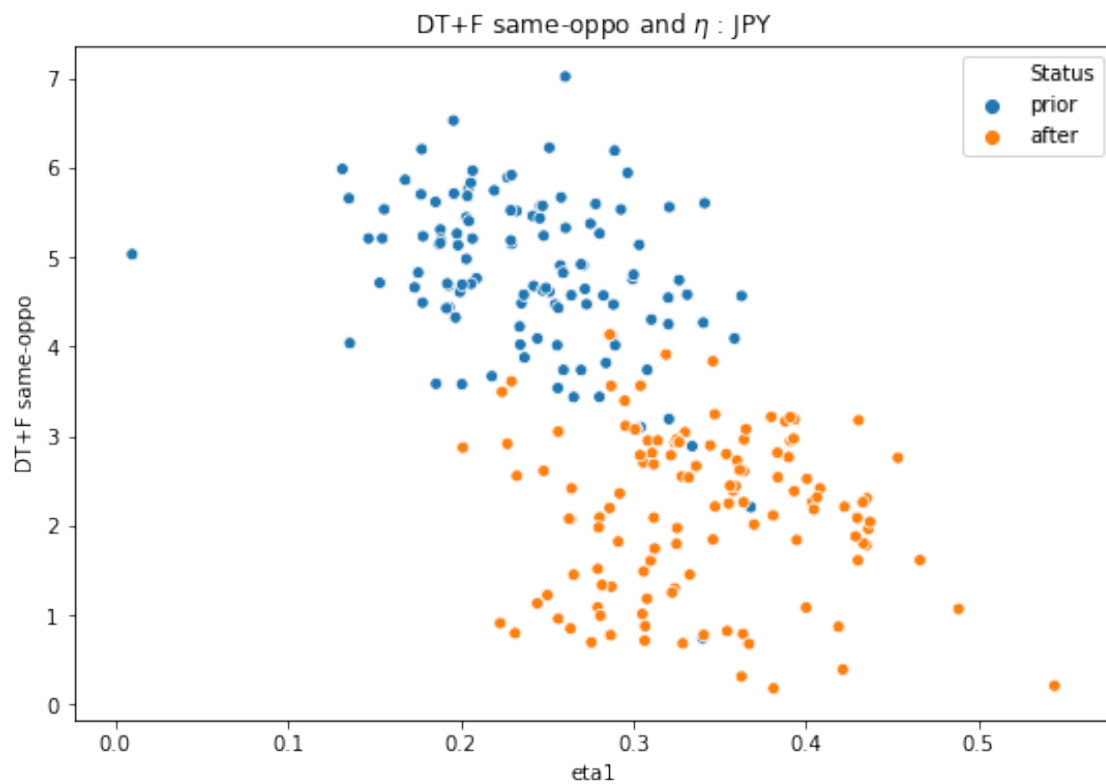
```
[71]: plt.figure(figsize=(9, 6))
sns.scatterplot(x='eta1', y='DC same-oppo', hue='Status',\
               data=DEPL_STATS_TS);
plt.title('DC same-oppo and  $\eta$  : JPY');
```



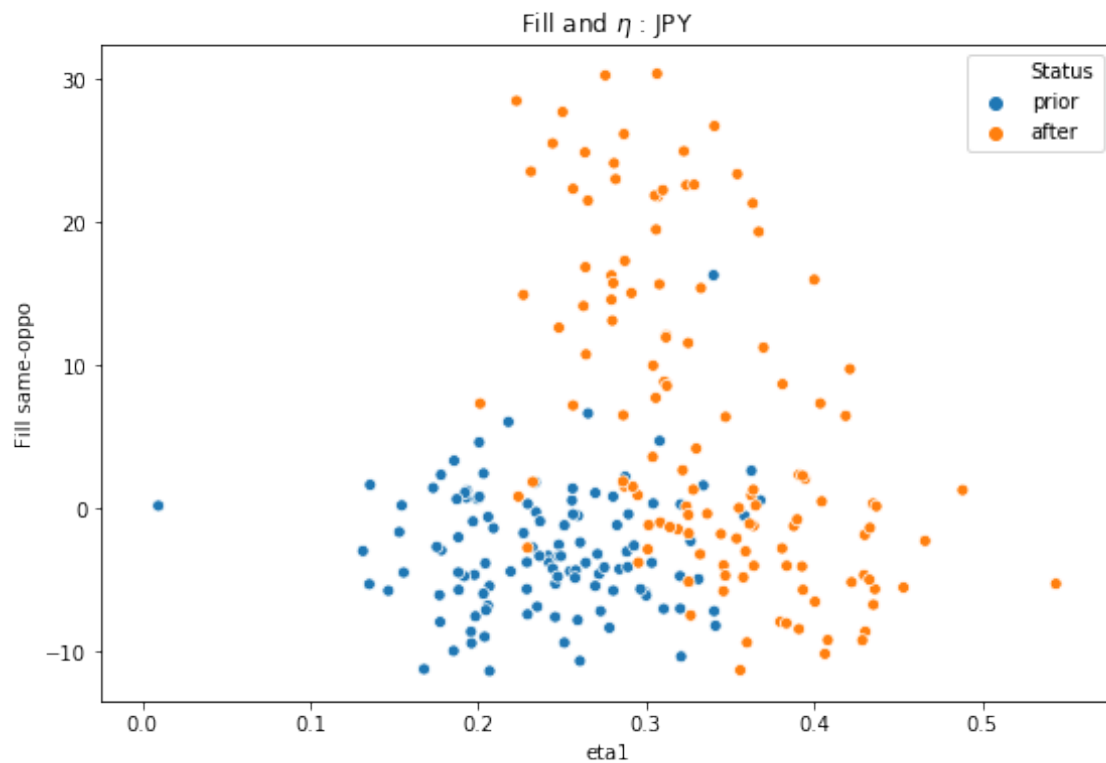
```
[72]: plt.figure(figsize=(9, 6))
sns.scatterplot(x='eta1', y='DT same-oppo', hue='Status',\
                data=DEPL_STATS_TS);
plt.title('DT same-oppo and  $\eta$  : JPY');
```



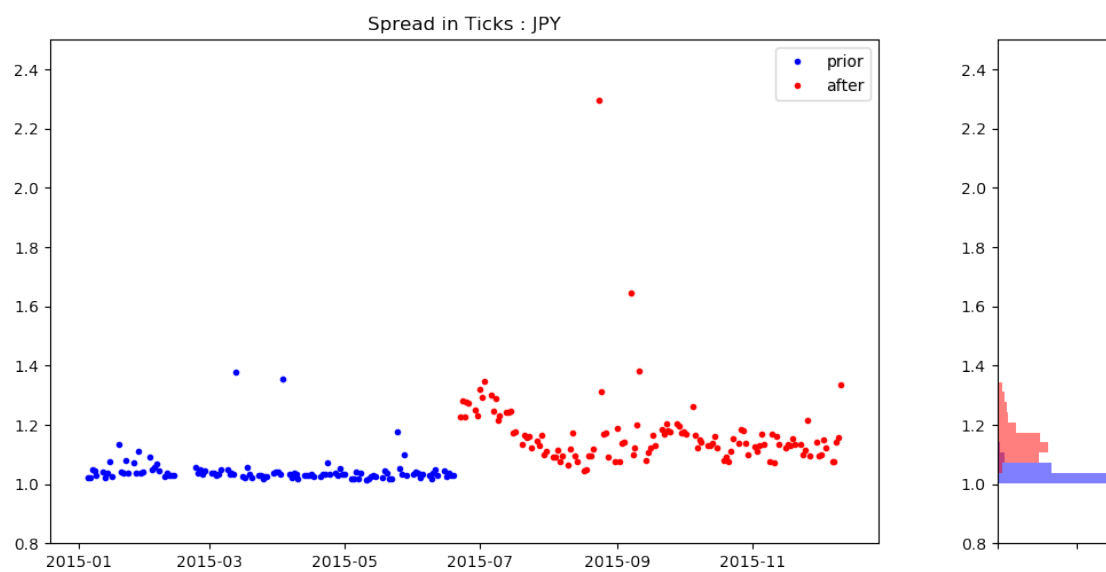
```
[73]: plt.figure(figsize=(9, 6))
sns.scatterplot(x='eta1', y='DT+F same-oppo', hue='Status',\
                data=DEPL_STATS_TS);
plt.title('DT+F same-oppo and  $\eta$  : JPY');
```



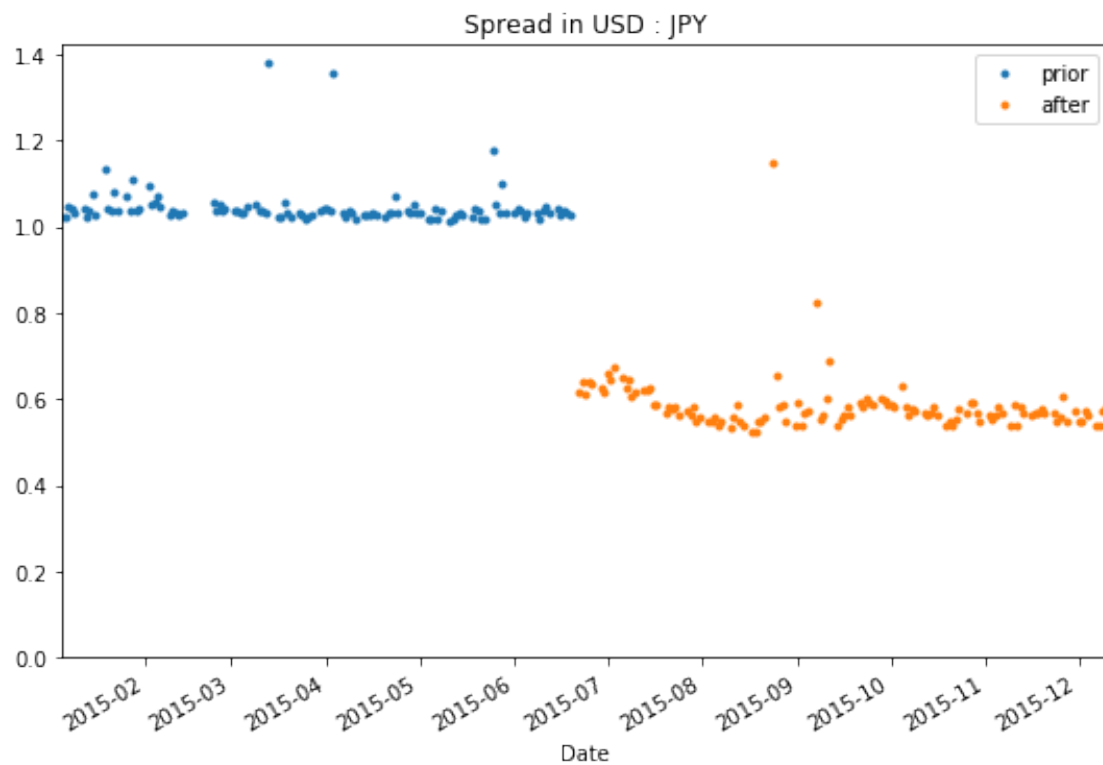
```
[74]: plt.figure(figsize=(9, 6))
sns.scatterplot(x='eta1', y='Fill same-oppo', hue='Status',\
               data=DEPL_STATS_TS);
plt.title('Fill and  $\eta$  : '+CURR);
```



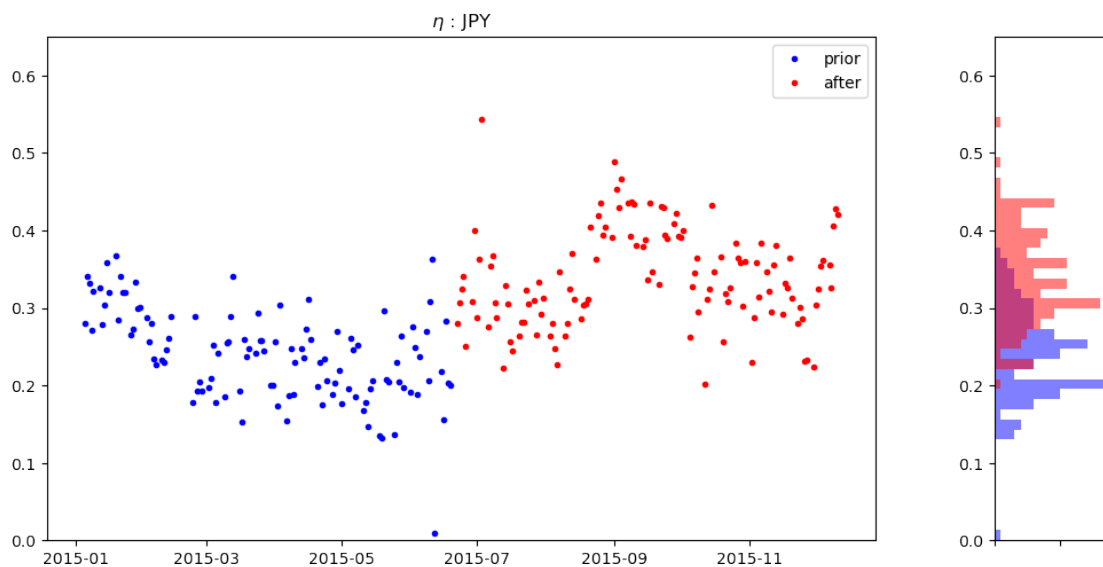
```
[183]: cme.time_series_hist_plot(OB_UZ_STATS, 'twspr1',\
    'Spread in Ticks : '+CURR, 0.8, 2.5, 50)
```



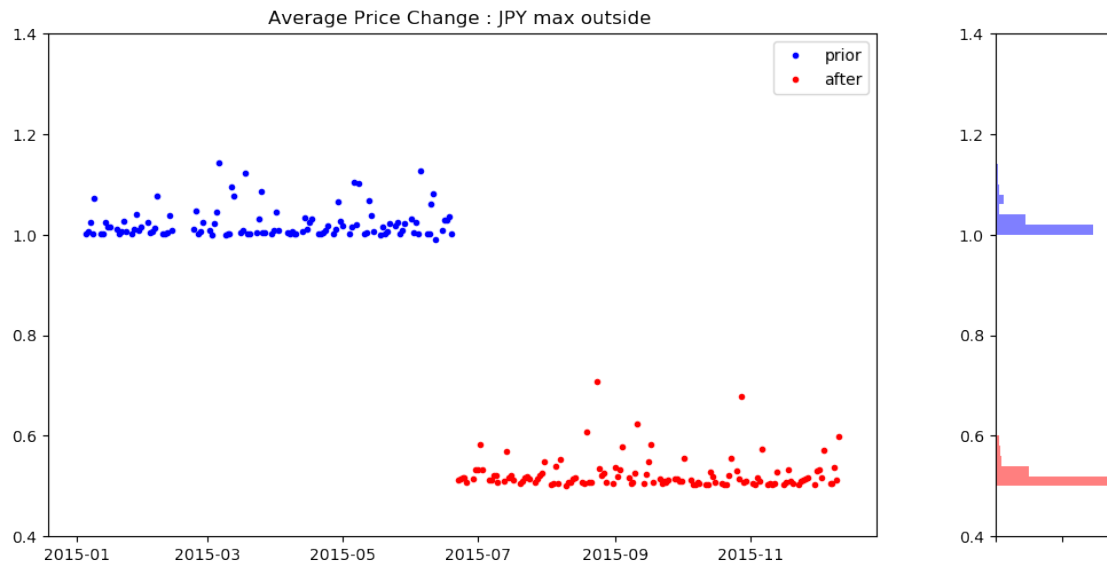
```
[76]: cme.twspr_plot_USD(OB_UZ_STATS, CURR)
```



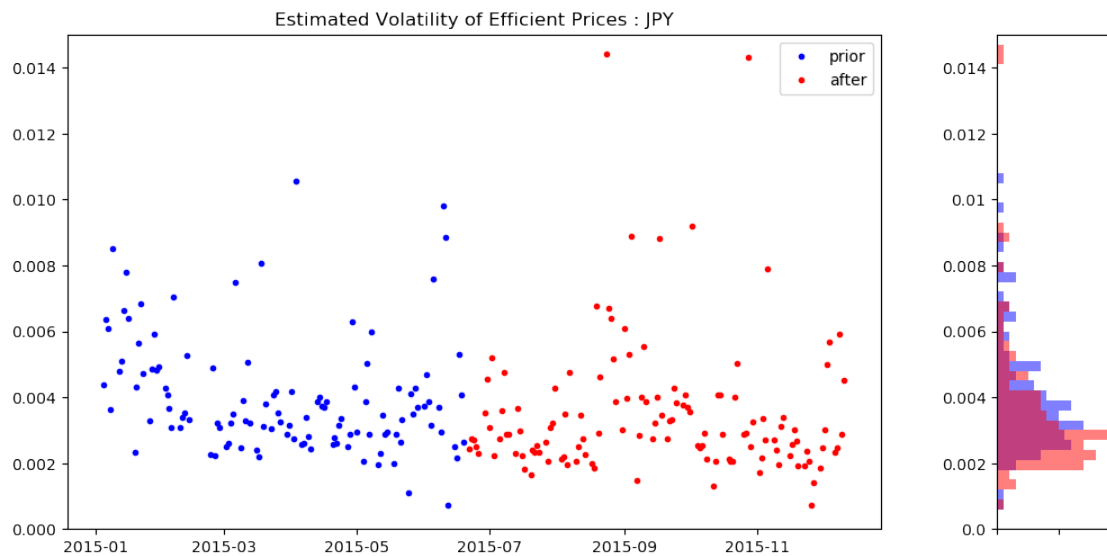
```
[184]: cme.time_series_hist_plot(OB_UZ_STATS, 'eta1',\
    '$\eta$ : '+CURR, 0, 0.65, 50)
```



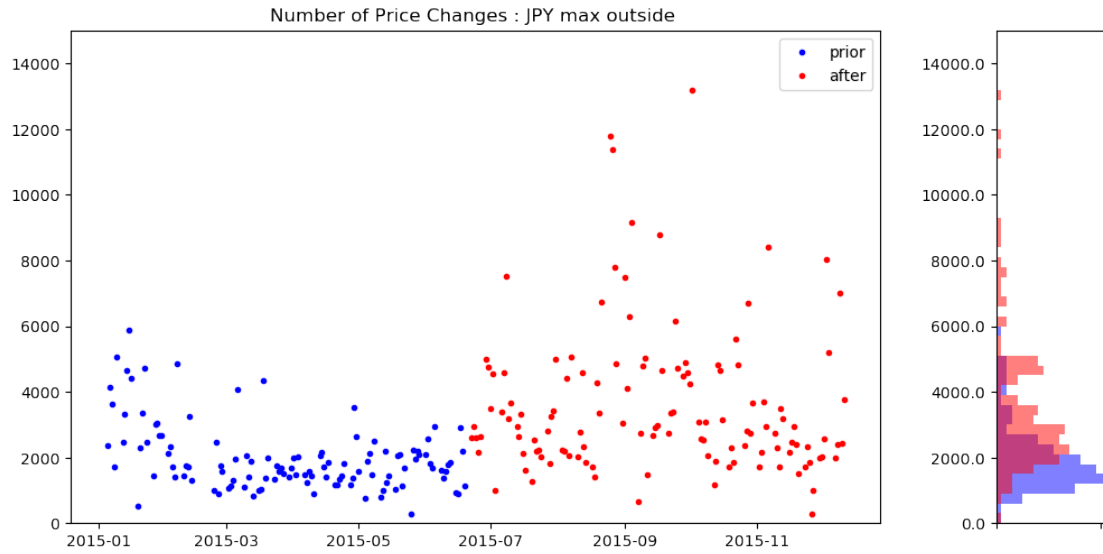

```
[78]: cme.time_series_hist_plot(OB_UZ_STATS, 'chgavg',\
    'Average Price Change : '+CURR, 0.4, 1.4, 50)
```



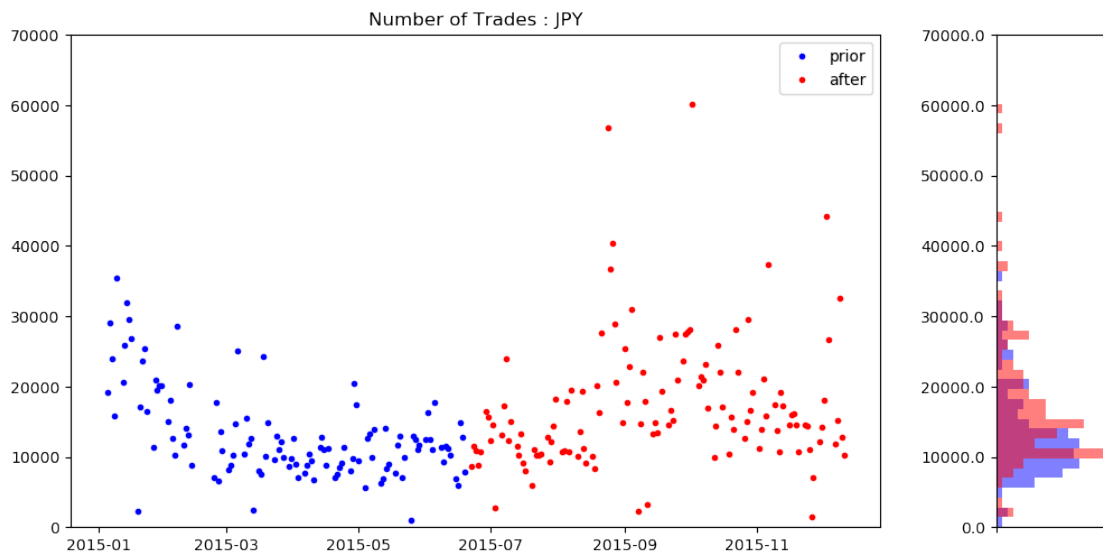
```
[186]: cme.time_series_hist_plot(OB_UZ_STATS, 'rvxe',\
    'Estimated Volatility of Efficient Prices : '+CURR, 0, 0.015, 50)
```



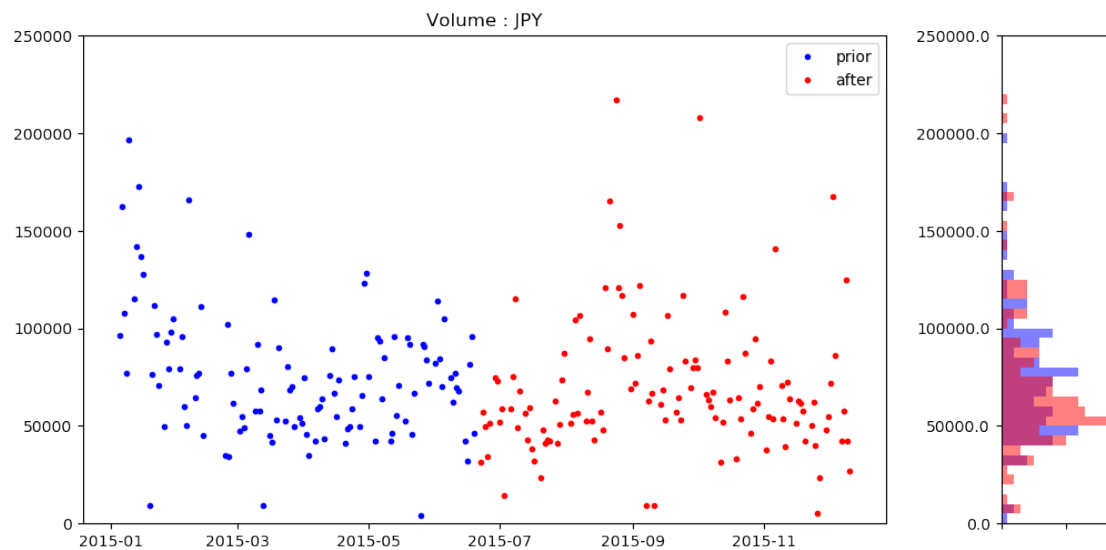
```
[188]: cme.time_series_hist_plot(OB_UZ_STATS, 'ndfpr',\
    'Number of Price Changes : '+CURR, 0, 15000, 50)
```



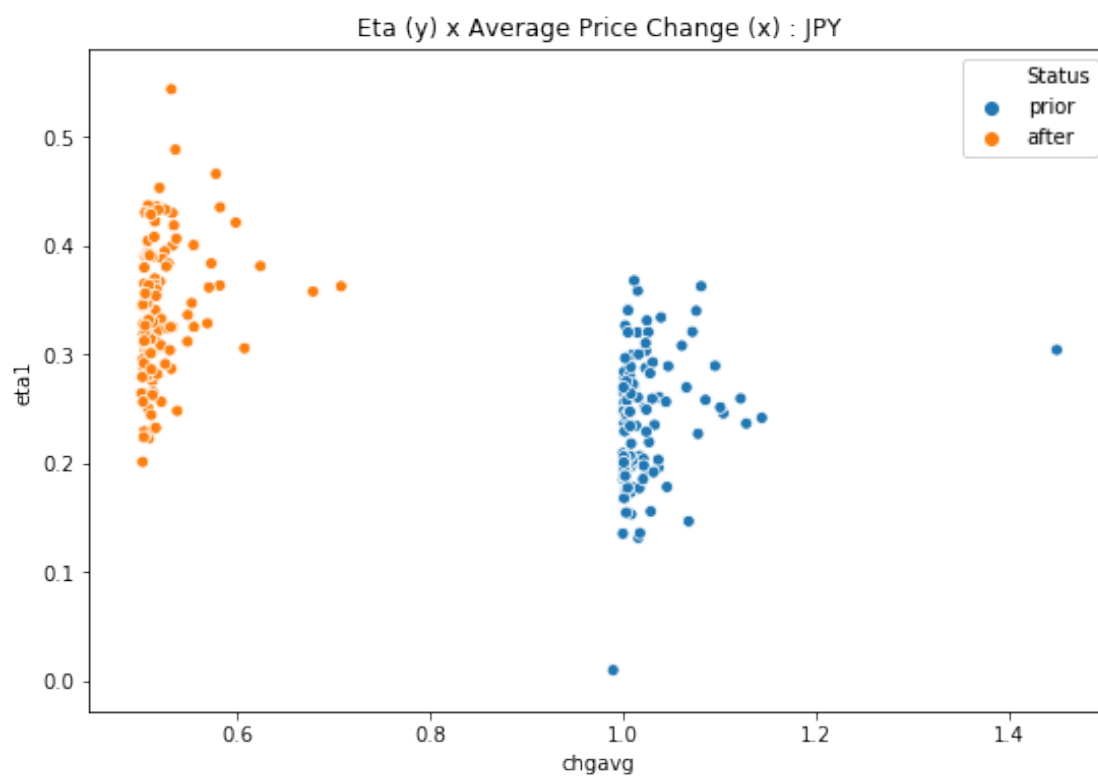
```
[190]: cme.time_series_hist_plot(OB_UZ_STATS, 'M',\
    'Number of Trades : '+CURR, 0, 70000, 50)
```



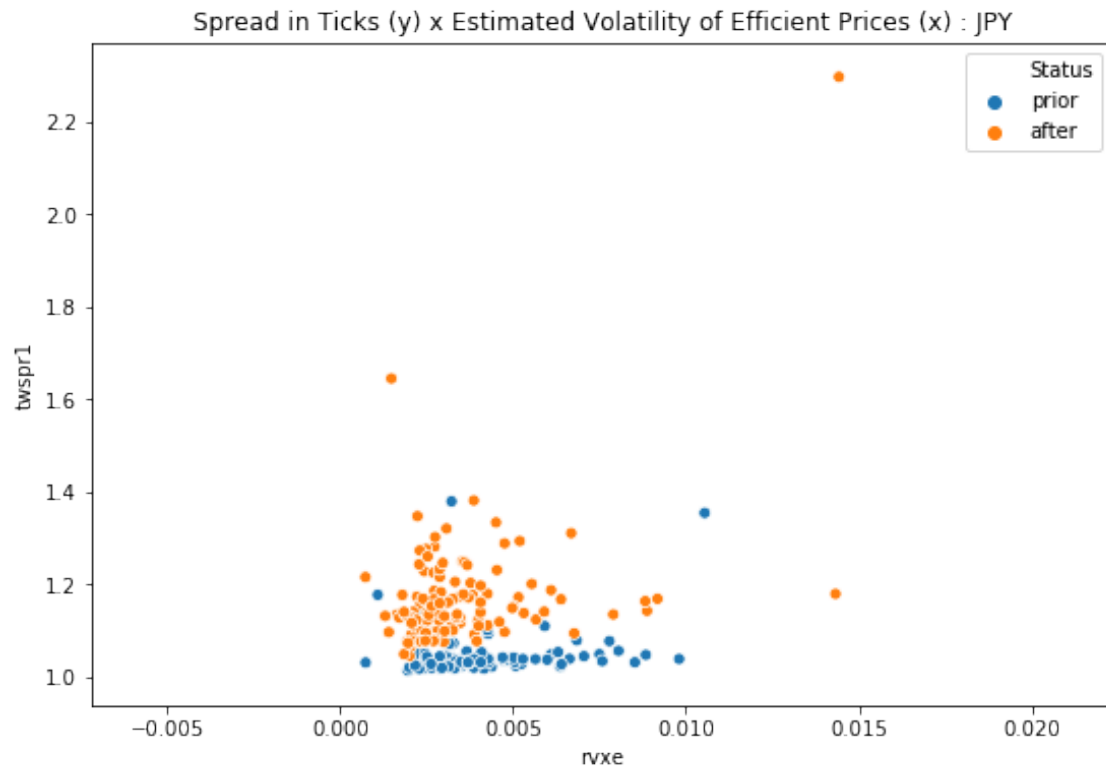
```
[192]: cme.time_series_hist_plot(OB_UZ_STATS, 'Volume',\
    'Volume : '+CURR, 0, 250000, 50)
```



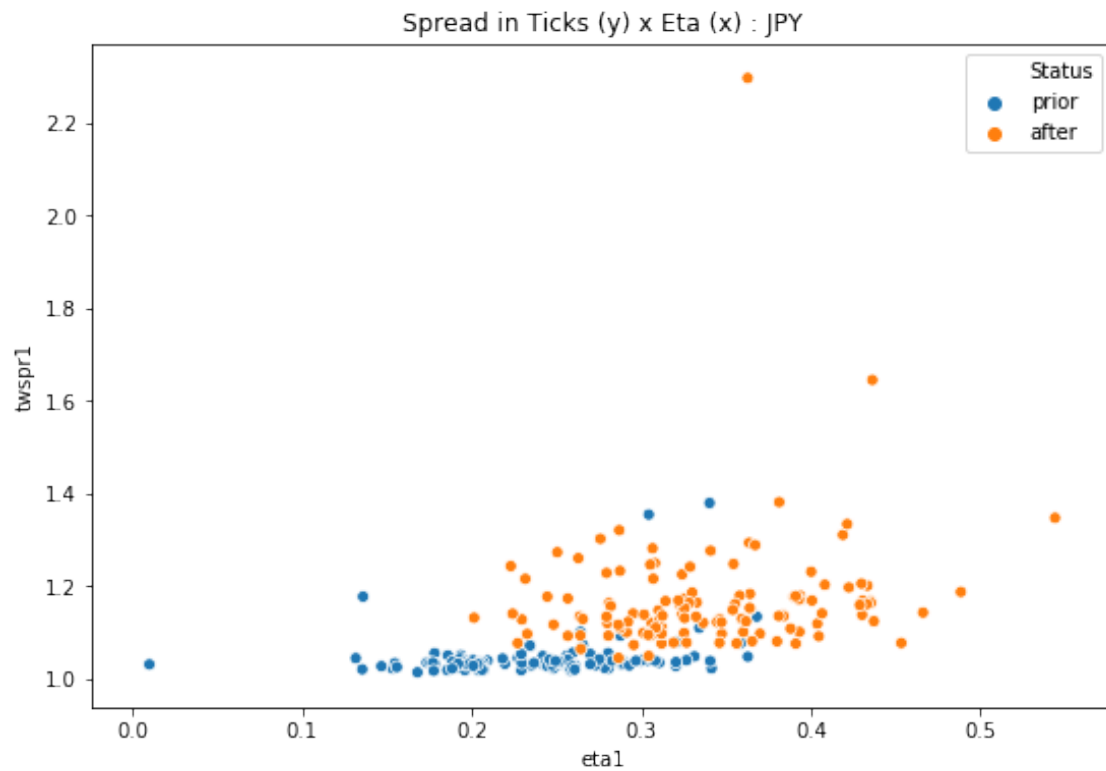
```
[83]: cme.scatter_plot(OB_UZ_STATS, 'chgavg', 'eta1',\
    'Eta (y) x Average Price Change (x) : '+CURR)
```



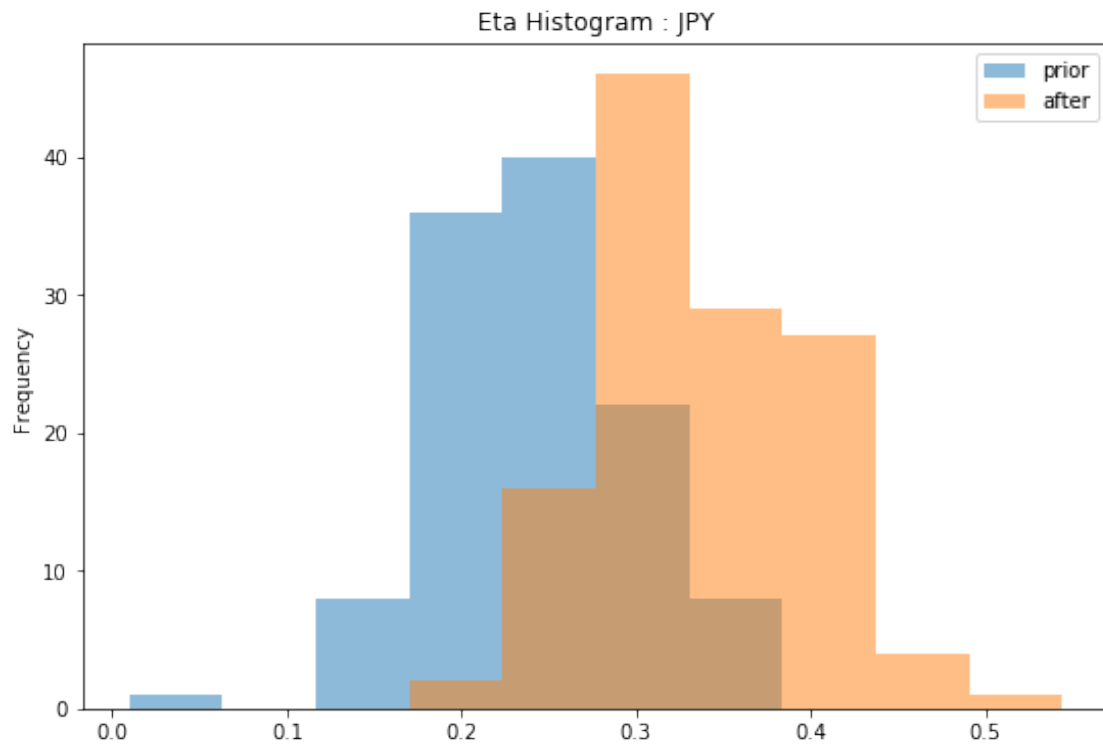
```
[84]: cme.scatter_plot(OB_UZ_STATS, 'rvxe', 'twspr1',\
    'Spread in Ticks (y) x Estimated Volatility of Efficient Prices (x) : JPY',\
    '+CURR')
```



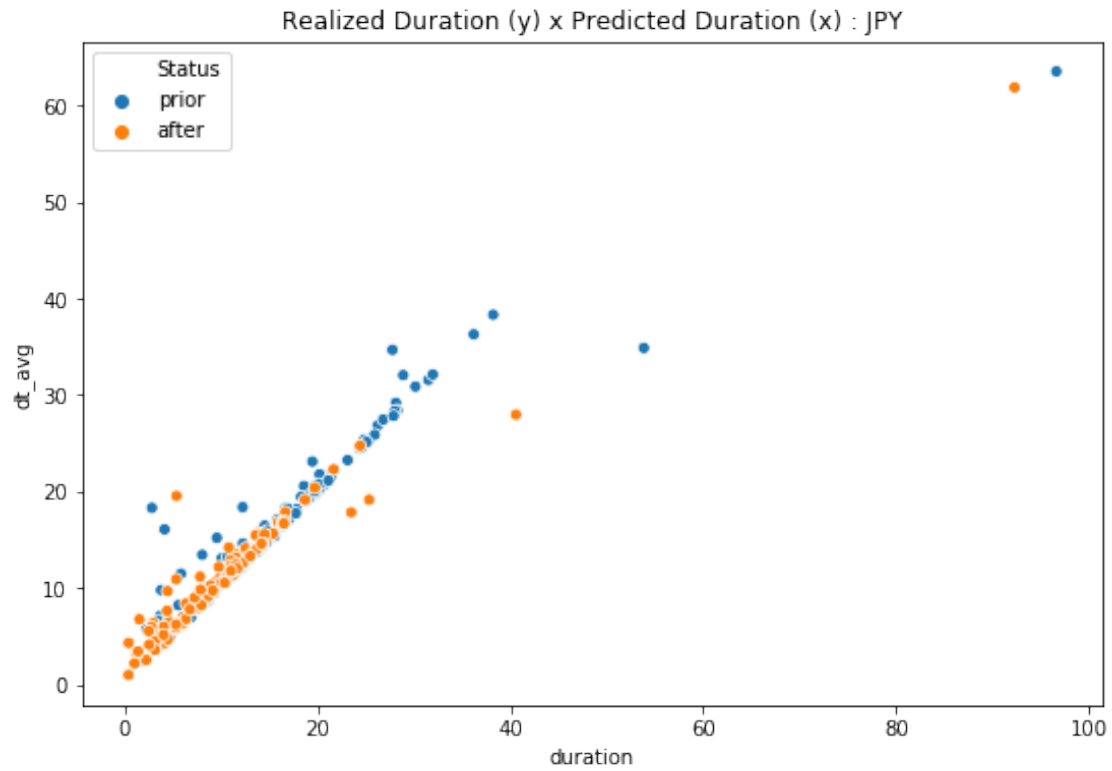
```
[85]: cme.scatter_plot(OB_UZ_STATS, 'eta1', 'twspr1',\
    'Spread in Ticks (y) x Eta (x) : '+CURR)
```



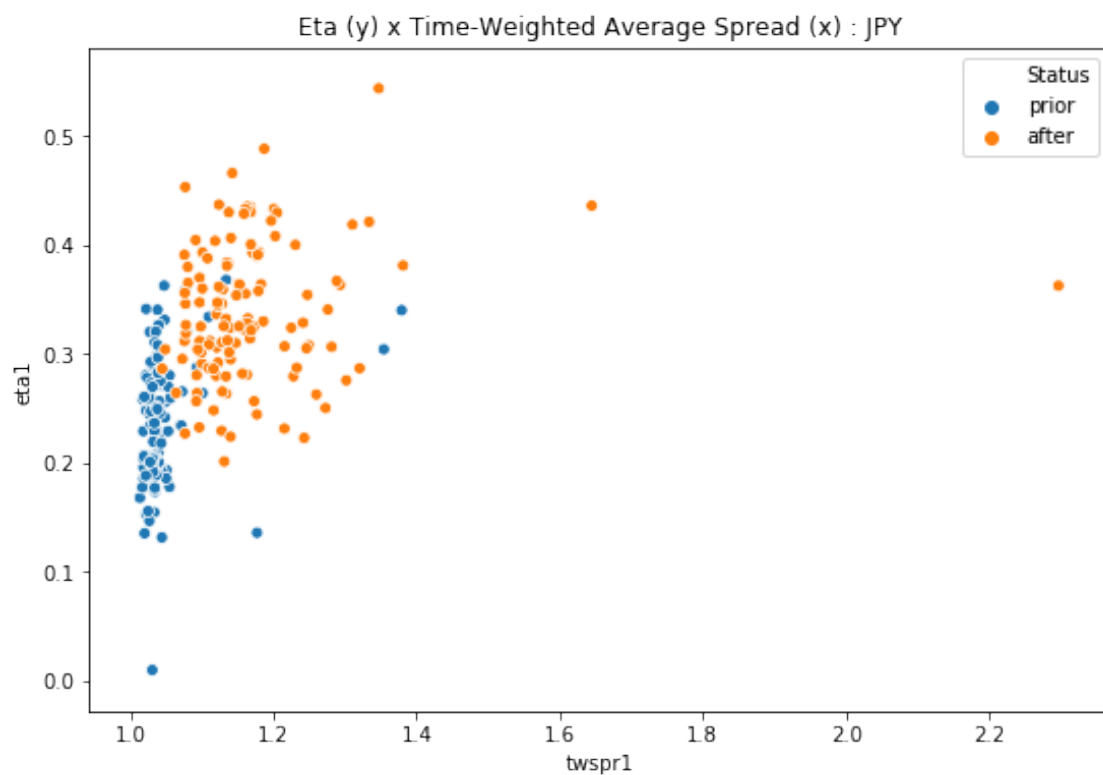
```
[86]: cme.time_series_hist(OB_UZ_STATS, 'eta1',\
      'Eta Histogram : '+CURR)
```



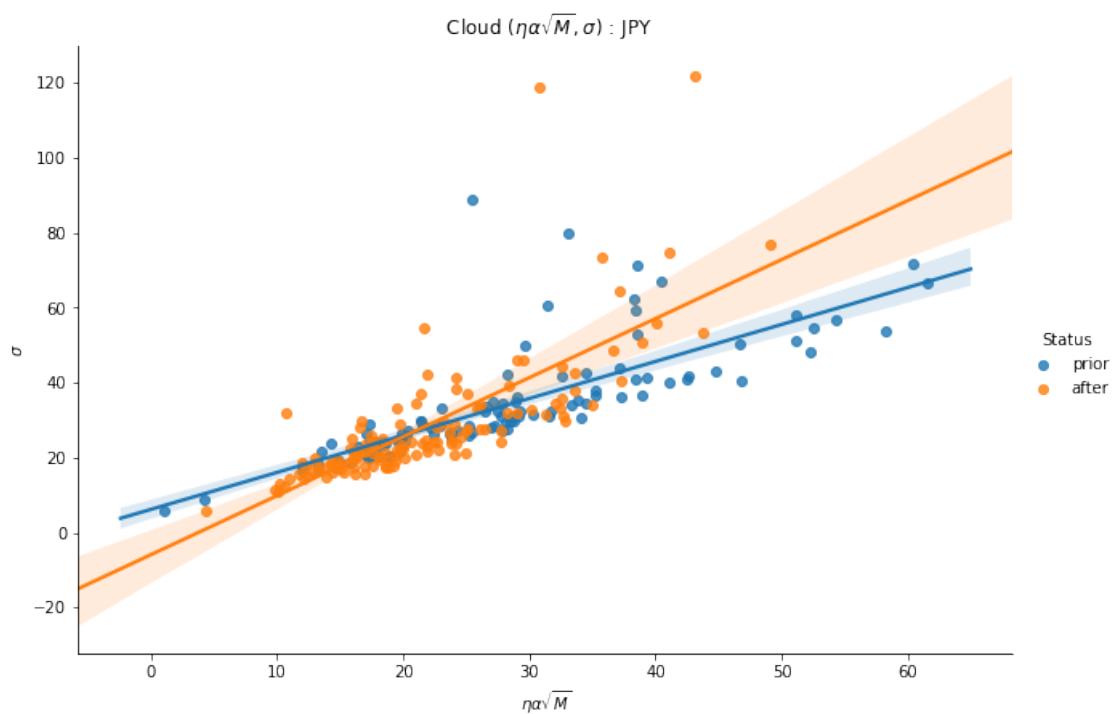
```
[87]: cme.scatter_plot(OB_UZ_STATS, 'duration', 'dt_avg',\  
    'Realized Duration (y) x Predicted Duration (x) : '+CURR)
```



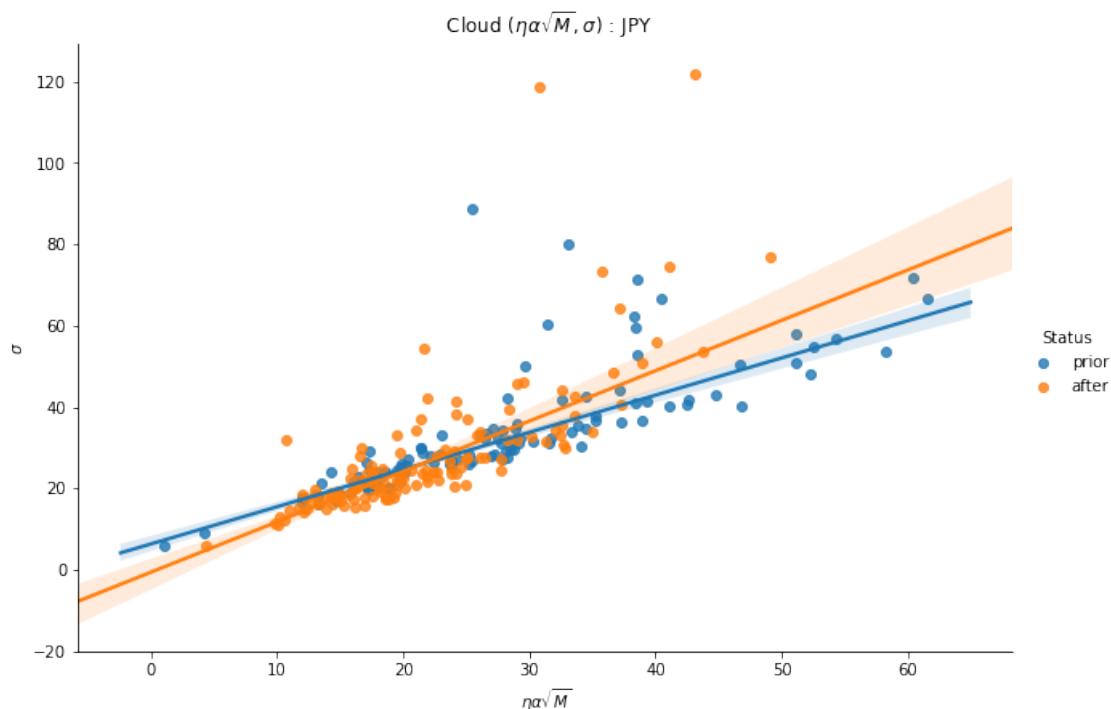
```
[88]: cme.scatter_plot(OB_UZ_STATS, 'twspr1', 'eta1',\
    'Eta (y) x Time-Weighted Average Spread (x) : '+CURR)
```



[89]: `cme.cloud1(OB_UZ_STATS, CURR)`




```
[90]: cme.cloud1(OB_UZ_STATS, CURR, True)
```



```
[91]: cme.lin_reg(PRIOR_OB_UZ_STATS, ['eta*alpha*sqrt(M)', 'S*sqrt(M)'], 'sigma')
```

OLS Regression Results

```
=====
Dep. Variable:          sigma    R-squared:                0.629
Model:                  OLS      Adj. R-squared:            0.623
Method:                 Least Squares    F-statistic:           95.04
Date:                   Wed, 09 Oct 2019    Prob (F-statistic):    7.40e-25
Time:                   16:10:48    Log-Likelihood:       -414.17
No. Observations:       115      AIC:                  834.3
Df Residuals:           112      BIC:                  842.6
Df Model:                2
Covariance Type:        nonrobust
=====
```

```
=====
              coef      std err          t      P>|t|      [0.025
0.975]
-----
const          -0.5655      4.570      -0.124      0.902     -9.621
```

```

8.490
eta*alpha*sqrt(M)    0.7324    0.168    4.371    0.000    0.400
1.064
S*sqrt(M)           0.1185    0.071    1.680    0.096    -0.021
0.258
=====
Omnibus:                125.596    Durbin-Watson:                1.696
Prob(Omnibus):          0.000    Jarque-Bera (JB):            1959.511
Skew:                   3.873    Prob(JB):                    0.00
Kurtosis:               21.680    Cond. No.:                   670.
=====

```

Warnings:

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

```
[92]: cme.lin_reg_rob(PRIOR_OB_UZ_STATS, ['eta*alpha*sqrt(M)', 'S*sqrt(M)'], 'sigma')
```

```

Robust linear Model Regression Results
=====
Dep. Variable:          sigma    No. Observations:          115
Model:                  RLM      Df Residuals:                112
Method:                 IRLS     Df Model:                    2
Norm:                   HuberT
Scale Est.:             mad
Cov Type:               H1
Date:                   Wed, 09 Oct 2019
Time:                   16:10:48
No. Iterations:         50
=====
=====
coef      std err          z      P>|z|      [0.025
0.975]
-----
const      -1.4132      1.786      -0.791      0.429      -4.914
2.087
eta*alpha*sqrt(M)  0.6077      0.065      9.281      0.000      0.479
0.736
S*sqrt(M)    0.1397      0.028      5.063      0.000      0.086
0.194
=====
=====

```

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[93]: cme.lin_reg(AFTER_OB_UZ_STATS, ['eta*alpha*sqrt(M)', 'S*sqrt(M)'], 'sigma')
```

```

                                OLS Regression Results
=====
Dep. Variable:                  sigma    R-squared:                  0.677
Model:                        OLS      Adj. R-squared:            0.672
Method:                      Least Squares  F-statistic:                127.8
Date:                        Wed, 09 Oct 2019  Prob (F-statistic):      1.17e-30
Time:                        16:10:48    Log-Likelihood:             -462.49
No. Observations:              125      AIC:                       931.0
Df Residuals:                  122      BIC:                       939.5
Df Model:                      2
Covariance Type:               nonrobust
=====
=====
              coef      std err          t      P>|t|      [0.025
0.975]
-----
const          -13.5533      2.791      -4.856      0.000     -19.078
-8.029
eta*alpha*sqrt(M)  0.6932      0.175       3.958      0.000       0.347
1.040
S*sqrt(M)        0.3620      0.057       6.368      0.000       0.250
0.475
=====
Omnibus:                 140.664    Durbin-Watson:                1.753
Prob(Omnibus):            0.000    Jarque-Bera (JB):             3653.972
Skew:                     3.913    Prob(JB):                     0.00
Kurtosis:                 28.304    Cond. No.                     259.
=====

```

Warnings:

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

```
[94]: cme.lin_reg_rob(AFTER_OB_UZ_STATS, ['eta*alpha*sqrt(M)', 'S*sqrt(M)'], 'sigma')
```

```

                    Robust linear Model Regression Results
=====
Dep. Variable:                  sigma    No. Observations:            125
Model:                        RLM      Df Residuals:                122
Method:                      IRLS      Df Model:                    2
Norm:                        HuberT
Scale Est.:                  mad
Cov Type:                    H1
Date:                        Wed, 09 Oct 2019
Time:                        16:10:48

```

```

No. Iterations:                    50
=====
=====
coef      std err      z      P>|z|      [0.025
0.975]
-----
-----
const      -11.4321      1.362      -8.391      0.000      -14.102
-8.762
eta*alpha*sqrt(M)      0.5201      0.085      6.084      0.000      0.353
0.688
S*sqrt(M)      0.3630      0.028      13.080      0.000      0.309
0.417
=====
=====

```

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```

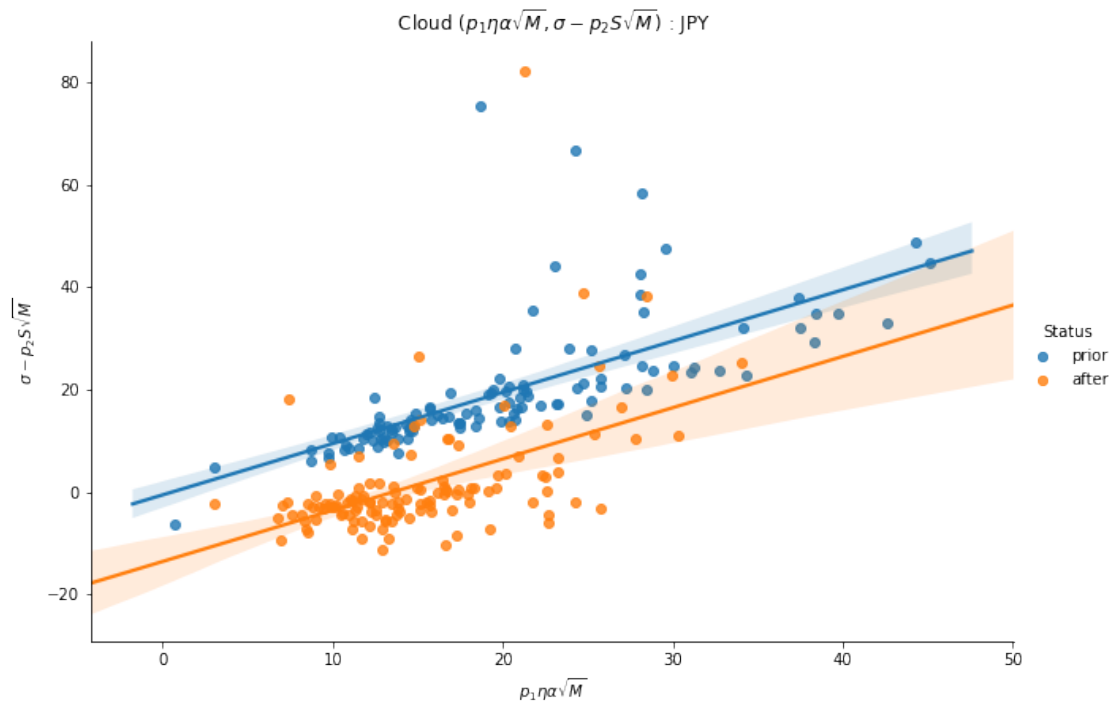
[95]: OB_UZ_STATS['p1*eta*alpha*sqrt(M)'] = np.where(OB_UZ_STATS['Status']=='prior',\
    cme.lin_reg_params(PRIOR_OB_UZ_STATS, ['eta*alpha*sqrt(M)', 'S*sqrt(M)'],\
    ↪ 'sigma')['eta*alpha*sqrt(M)'],\
    cme.lin_reg_params(AFTER_OB_UZ_STATS, ['eta*alpha*sqrt(M)', 'S*sqrt(M)'],\
    ↪ 'sigma')['eta*alpha*sqrt(M)'])\
    *OB_UZ_STATS['eta*alpha*sqrt(M)']
OB_UZ_STATS['sigma-p2*S*sqrt(M)'] = OB_UZ_STATS['sigma']-\
    np.where(OB_UZ_STATS['Status']=='prior',\
    cme.lin_reg_params(PRIOR_OB_UZ_STATS, ['eta*alpha*sqrt(M)', 'S*sqrt(M)'],\
    ↪ 'sigma')['S*sqrt(M)'],\
    cme.lin_reg_params(AFTER_OB_UZ_STATS, ['eta*alpha*sqrt(M)', 'S*sqrt(M)'],\
    ↪ 'sigma')['S*sqrt(M)'])*\
    OB_UZ_STATS['S*sqrt(M)']

```

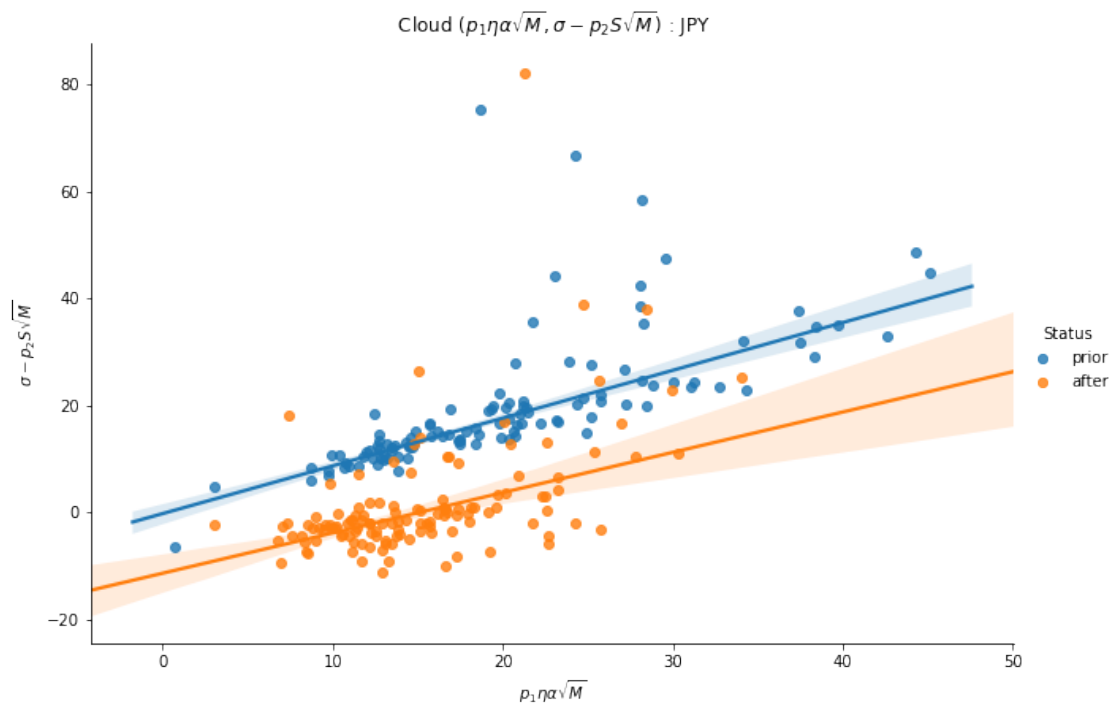
```

[96]: cme.cloud2(OB_UZ_STATS, CURR)

```



```
[97]: cme.cloud2(OB_UZ_STATS, CURR, True)
```



```
[98]: cme.lin_reg(OB_UZ_STATS[OB_UZ_STATS['Status']=='prior'],  
→['p1*eta*alpha*sqrt(M)'], 'sigma-p2*S*sqrt(M)')
```

```

                                OLS Regression Results
=====
Dep. Variable:    sigma-p2*S*sqrt(M)    R-squared:                0.480
Model:                OLS    Adj. R-squared:            0.475
Method:            Least Squares    F-statistic:            104.2
Date:                Wed, 09 Oct 2019    Prob (F-statistic):      1.00e-17
Time:                16:11:02    Log-Likelihood:          -414.17
No. Observations:    115    AIC:                    832.3
Df Residuals:        113    BIC:                    837.8
Df Model:            1
Covariance Type:        nonrobust
=====
=====
                                coef      std err          t      P>|t|      [0.025
0.975]
-----
const                -0.5655        2.141      -0.264      0.792      -4.807
3.676
p1*eta*alpha*sqrt(M)  1.0000        0.098     10.205      0.000        0.806
1.194
=====
Omnibus:                125.596    Durbin-Watson:           1.696
Prob(Omnibus):           0.000    Jarque-Bera (JB):        1959.511
Skew:                    3.873    Prob(JB):                0.00
Kurtosis:                21.680    Cond. No.                56.2
=====

```

Warnings:

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

```
[99]: cme.lin_reg_rob(OB_UZ_STATS[OB_UZ_STATS['Status']=='prior'],  
→['p1*eta*alpha*sqrt(M)'], 'sigma-p2*S*sqrt(M)')
```

```

                        Robust linear Model Regression Results
=====
Dep. Variable:    sigma-p2*S*sqrt(M)    No. Observations:            115
Model:                RLM    Df Residuals:                113
Method:            IRLS    Df Model:                    1
Norm:                HuberT
Scale Est.:                mad
Cov Type:                H1
Date:                Wed, 09 Oct 2019
Time:                16:11:02

```

No. Iterations: 37

=====					
	coef	std err	z	P> z	[0.025
0.975]					

const	-0.2559	0.843	-0.304	0.761	-1.908
1.396					
p1*eta*alpha*sqrt(M)	0.8932	0.039	23.155	0.000	0.818
0.969					
=====					
=====					

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[100]: cme.lin_reg(OB_UZ_STATS[OB_UZ_STATS['Status']=='after'],  
             ↪['p1*eta*alpha*sqrt(M)'], 'sigma-p2*S*sqrt(M)')
```

OLS Regression Results

=====					
Dep. Variable:	sigma-p2*S*sqrt(M)	R-squared:	0.255		
Model:	OLS	Adj. R-squared:	0.249		
Method:	Least Squares	F-statistic:	42.04		
Date:	Wed, 09 Oct 2019	Prob (F-statistic):	1.96e-09		
Time:	16:11:02	Log-Likelihood:	-462.49		
No. Observations:	125	AIC:	929.0		
Df Residuals:	123	BIC:	934.6		
Df Model:	1				
Covariance Type:	nonrobust				
=====					
=====					
	coef	std err	t	P> t	[0.025
0.975]					

const	-13.5533	2.506	-5.408	0.000	-18.514
-8.592					
p1*eta*alpha*sqrt(M)	1.0000	0.154	6.484	0.000	0.695
1.305					
=====					
Omnibus:	140.664	Durbin-Watson:	1.753		
Prob(Omnibus):	0.000	Jarque-Bera (JB):	3653.972		
Skew:	3.913	Prob(JB):	0.00		
Kurtosis:	28.304	Cond. No.	46.3		
=====					

Warnings:

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

```
[101]: cme.lin_reg_rob(OB_UZ_STATS[OB_UZ_STATS['Status']=='after'],  
↳['p1*eta*alpha*sqrt(M)'], 'sigma-p2*S*sqrt(M)')
```

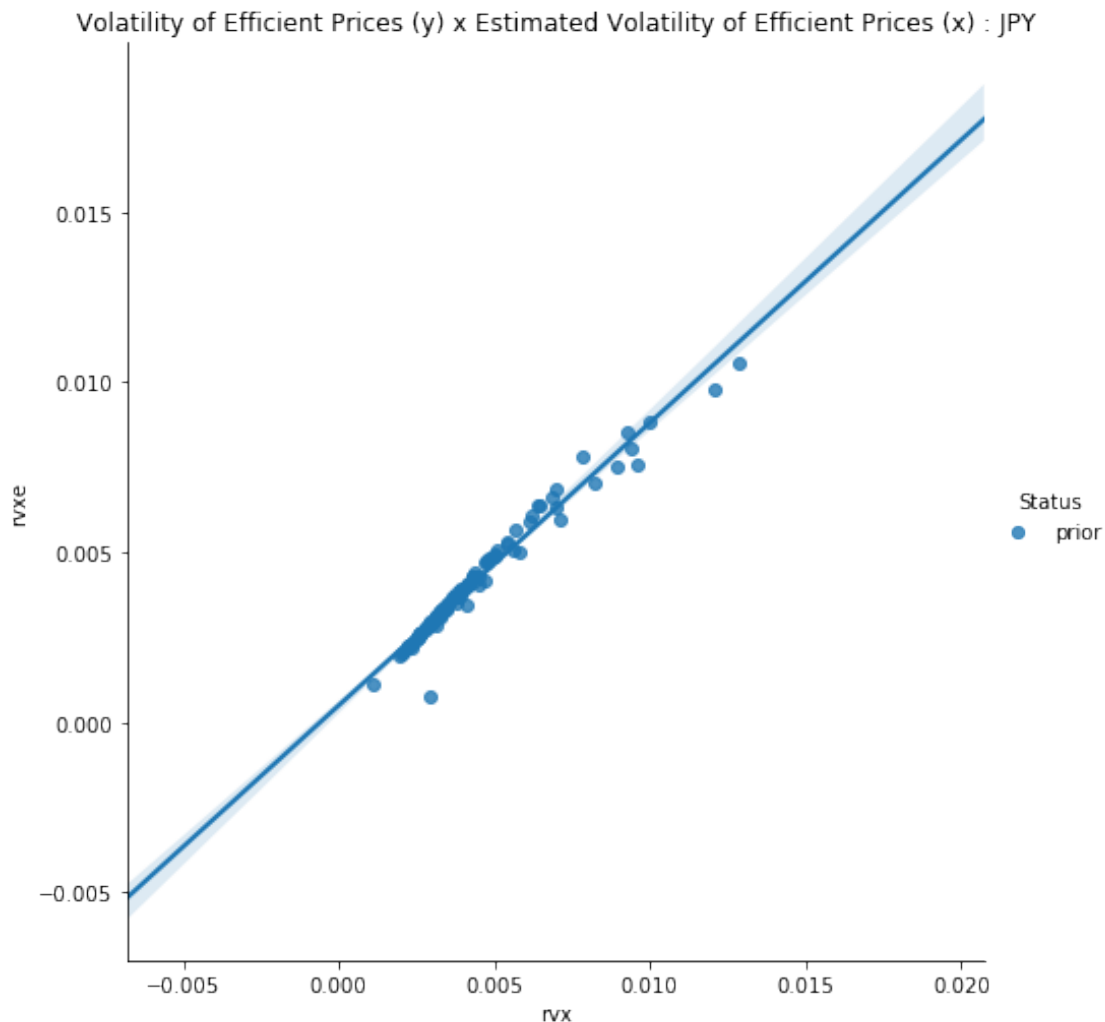
Robust linear Model Regression Results

```
=====
Dep. Variable:      sigma-p2*S*sqrt(M)    No. Observations:      125
Model:              RLM                   Df Residuals:          123
Method:             IRLS                  Df Model:              1
Norm:               HuberT
Scale Est.:         mad
Cov Type:           H1
Date:               Wed, 09 Oct 2019
Time:               16:11:02
No. Iterations:     50
=====
```

```
=====
                                coef      std err          z      P>|z|      [0.025
0.975]
-----
const                -11.4088         1.226      -9.306      0.000      -13.812
-9.006
p1*eta*alpha*sqrt(M)  0.7537         0.075       9.991      0.000         0.606
0.902
=====
```

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[102]: cme.regr_plot(PRIOR_OB_UZ_STATS, 'rvx', 'rvxe',\  
↳'Volatility of Efficient Prices (y) x Estimated Volatility of Efficient_  
↳Prices (x) : '+CURR)
```

```
[103]: cme.lin_reg(PRIOR_OB_UZ_STATS, 'rvx', 'rvxe', True)
```

OLS Regression Results

=====					
Dep. Variable:	rvxe	R-squared:	0.897		
Model:	OLS	Adj. R-squared:	0.896		
Method:	Least Squares	F-statistic:	981.3		
Date:	Wed, 09 Oct 2019	Prob (F-statistic):	1.53e-57		
Time:	16:11:03	Log-Likelihood:	69.473		
No. Observations:	115	AIC:	-134.9		
Df Residuals:	113	BIC:	-129.5		
Df Model:	1				
Covariance Type:	nonrobust				
=====					
	coef	std err	t	P> t	[0.025 0.975]

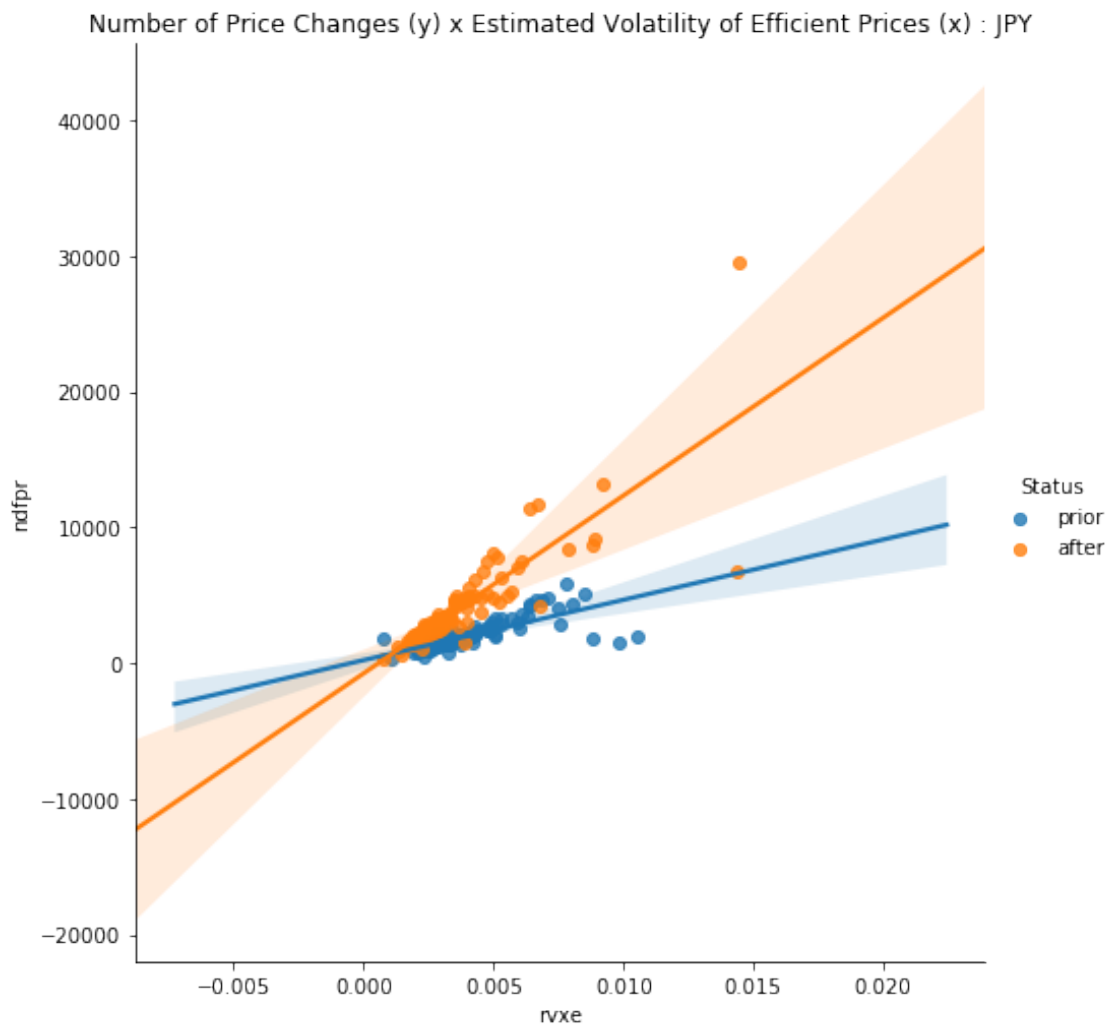
const	-0.3790	0.167	-2.265	0.025	-0.711	-0.047
rvx	0.9396	0.030	31.326	0.000	0.880	0.999

Omnibus:	230.242	Durbin-Watson:	1.934
Prob(Omnibus):	0.000	Jarque-Bera (JB):	40562.325
Skew:	-9.171	Prob(JB):	0.00
Kurtosis:	93.160	Cond. No.	77.5

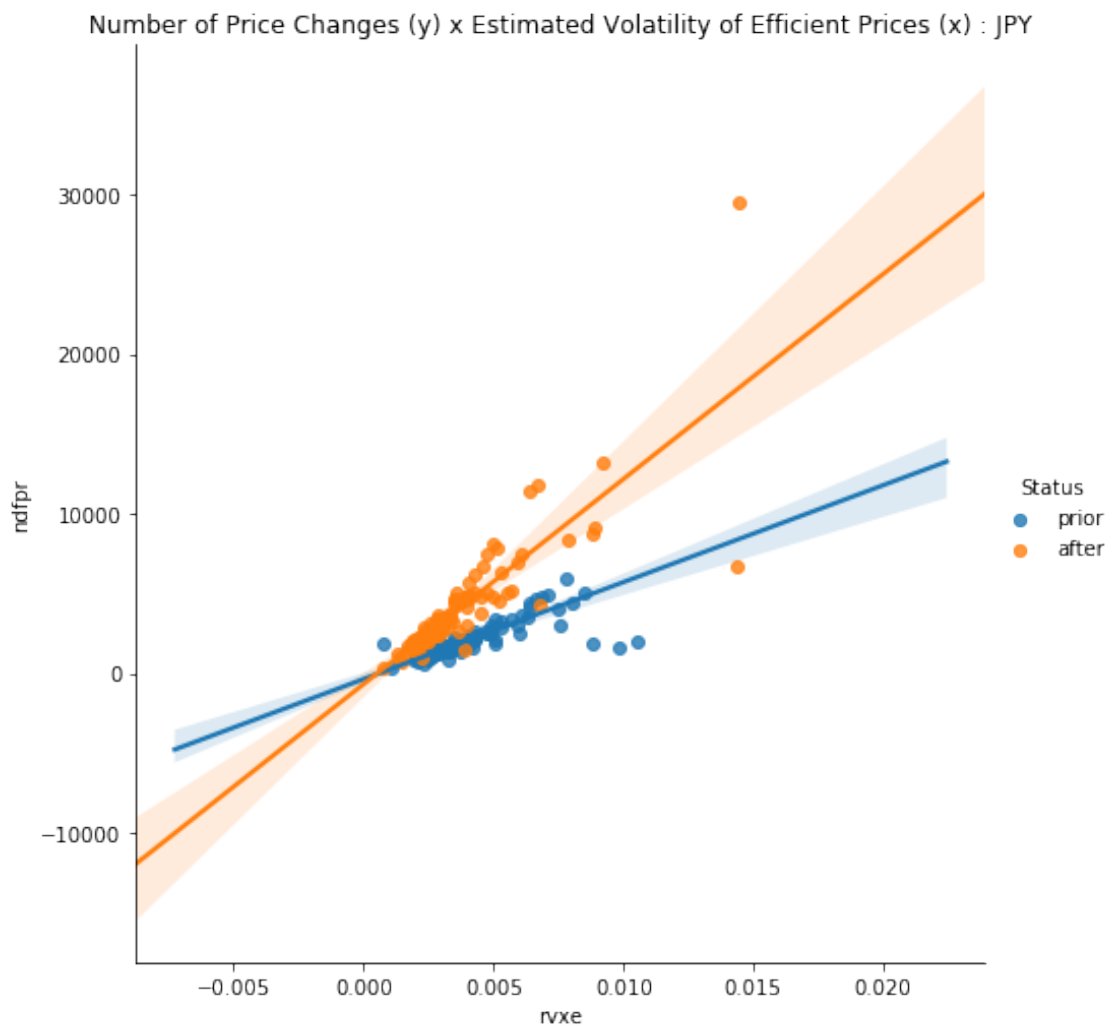
Warnings:

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

```
[104]: cme.regr_plot(OB_UZ_STATS, 'rvxe', 'ndfpr', \
    'Number of Price Changes (y) x Estimated Volatility of Efficient Prices (x) \
    ↪: '+CURR)
```



```
[105]: cme.regr_plot(OB_UZ_STATS, 'rvxe', 'ndfpr',\
    'Number of Price Changes (y) x Estimated Volatility of Efficient Prices (x)'\
    ↪: '+CURR, True)
```



```
[106]: cme.lin_reg(PRIOR_OB_UZ_STATS, 'rvxe', 'ndfpr', True)
```

OLS Regression Results

```
=====
Dep. Variable:          ndfpr    R-squared:                0.640
Model:                  OLS      Adj. R-squared:           0.636
Method:                 Least Squares    F-statistic:         200.5
Date:                   Wed, 09 Oct 2019    Prob (F-statistic):   8.50e-27
Time:                   16:11:16    Log-Likelihood:       -20.584
No. Observations:      115        AIC:                 45.17
=====
```

Df Residuals: 113 BIC: 50.66
Df Model: 1
Covariance Type: nonrobust

```
=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const         12.7349        0.372      34.241      0.000       11.998       13.472
rvxe           0.9367        0.066      14.161      0.000        0.806        1.068
=====
Omnibus:                 30.409   Durbin-Watson:           2.081
Prob(Omnibus):            0.000   Jarque-Bera (JB):        326.946
Skew:                     0.258   Prob(JB):               1.01e-71
Kurtosis:                 11.244   Cond. No.                79.2
=====
```

Warnings:

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

```
[107]: cme.lin_reg_rob(PRIOR_OB_UZ_STATS, 'rvxe', 'ndfpr', True)
```

Robust linear Model Regression Results

```
=====
Dep. Variable:          ndfpr   No. Observations:          115
Model:                RLM      Df Residuals:              113
Method:               IRLS     Df Model:                  1
Norm:                 HuberT
Scale Est.:           mad
Cov Type:             H1
Date:                 Wed, 09 Oct 2019
Time:                 16:11:16
No. Iterations:       26
=====
              coef      std err          z      P>|z|      [0.025      0.975]
-----
const         14.4039        0.178      80.836      0.000       14.055       14.753
rvxe           1.2298        0.032      38.805      0.000        1.168        1.292
=====
```

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[108]: cme.lin_reg(AFTER_OB_UZ_STATS, 'rvxe', 'ndfpr', True)
```

OLS Regression Results

```
=====
Dep. Variable:          ndfpr   R-squared:              0.841
Model:                OLS     Adj. R-squared:         0.840
```

```

Method:          Least Squares    F-statistic:          649.7
Date:            Wed, 09 Oct 2019  Prob (F-statistic):       6.46e-51
Time:            16:11:16          Log-Likelihood:          0.55552
No. Observations:      125          AIC:                      2.889
Df Residuals:          123          BIC:                      8.546
Df Model:              1
Covariance Type:      nonrobust

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const         15.1309      0.279      54.283      0.000      14.579      15.683
rvxe           1.2265      0.048      25.489      0.000       1.131       1.322
=====
Omnibus:                 65.568    Durbin-Watson:                 1.618
Prob(Omnibus):            0.000    Jarque-Bera (JB):            251.141
Skew:                     -1.909    Prob(JB):                     2.92e-55
Kurtosis:                  8.800    Cond. No.                      76.5
=====

```

Warnings:

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

```
[109]: cme.lin_reg_rob(AFTER_OB_UZ_STATS, 'rvxe', 'ndfpr', True)
```

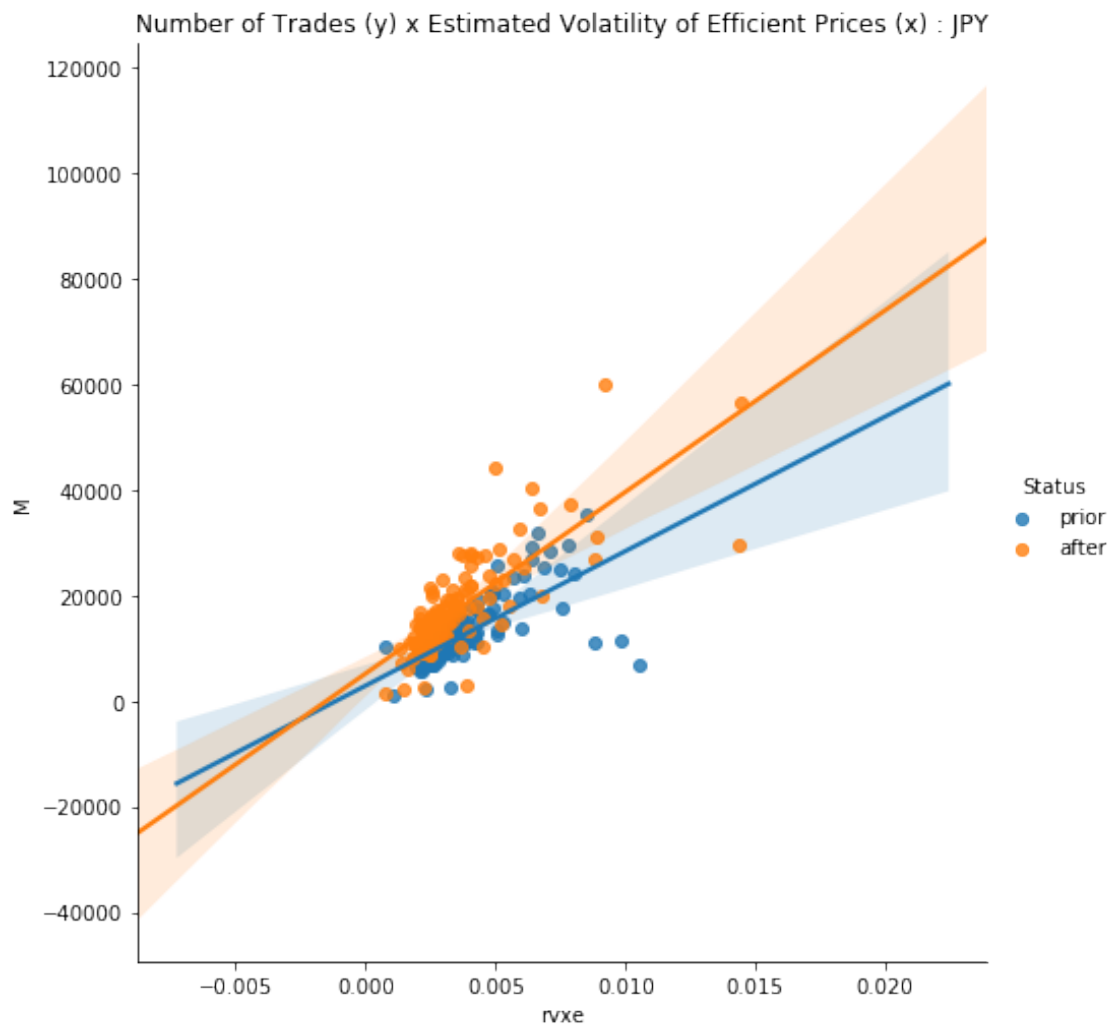
```

              Robust linear Model Regression Results
=====
Dep. Variable:          ndfpr    No. Observations:          125
Model:                  RLM      Df Residuals:              123
Method:                 IRLS     Df Model:                  1
Norm:                   HuberT
Scale Est.:             mad
Cov Type:               H1
Date:                   Wed, 09 Oct 2019
Time:                   16:11:16
No. Iterations:         37
=====
              coef      std err          z      P>|z|      [0.025      0.975]
-----
const         15.4635      0.203      76.033      0.000      15.065      15.862
rvxe           1.2784      0.035      36.411      0.000       1.210       1.347
=====

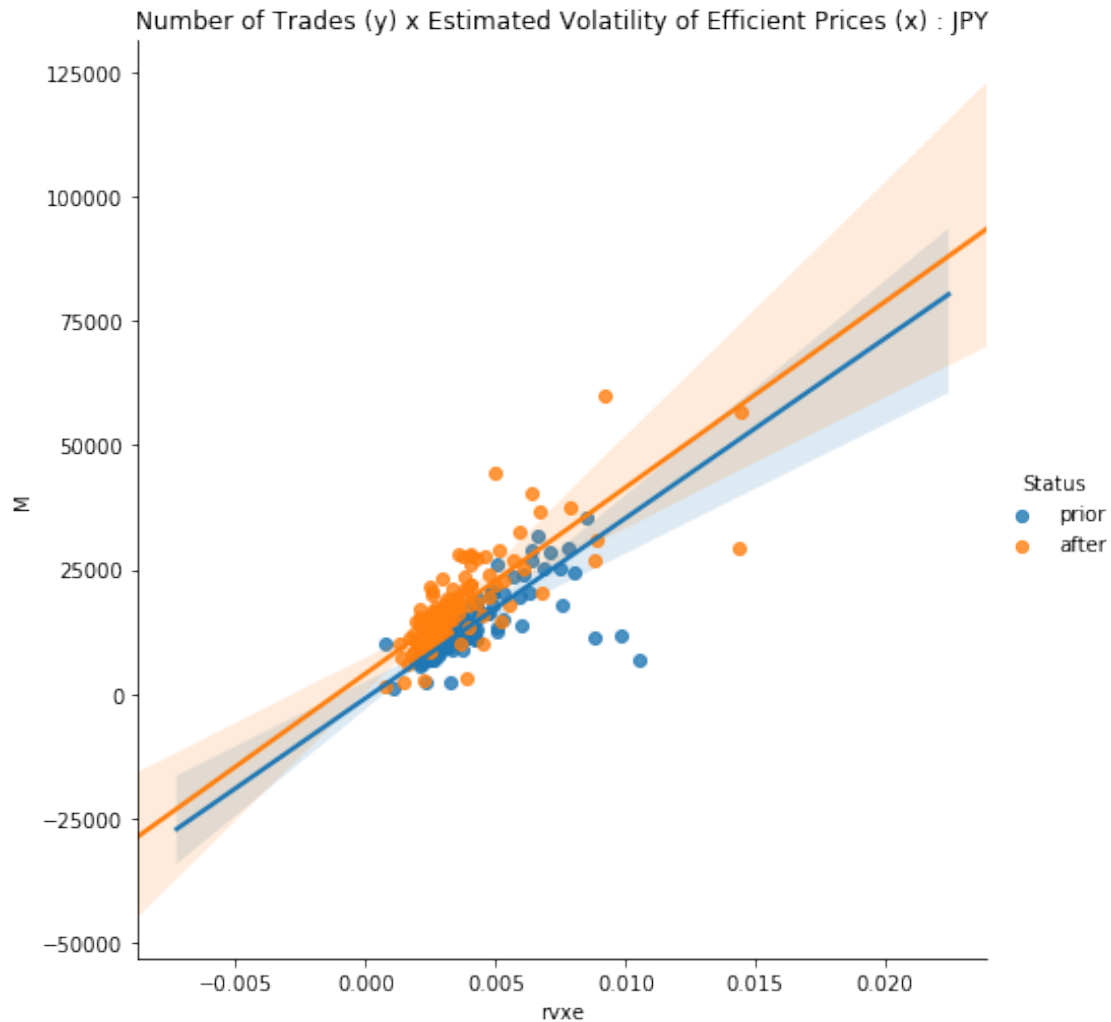
```

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[110]: cme.regr_plot(OB_UZ_STATS, 'rvxe', 'M',\
    'Number of Trades (y) x Estimated Volatility of Efficient Prices (x) : JPY'
    ↪ '+CURR)
```



```
[111]: cme.regr_plot(OB_UZ_STATS, 'rvxe', 'M',\
    'Number of Trades (y) x Estimated Volatility of Efficient Prices (x) : JPY'
    ↪ '+CURR, True)
```



```
[112]: cme.lin_reg(PRIOR_OB_UZ_STATS, 'rvxe', 'M', True)
```

OLS Regression Results

=====					
Dep. Variable:	M	R-squared:	0.522		
Model:	OLS	Adj. R-squared:	0.518		
Method:	Least Squares	F-statistic:	123.4		
Date:	Wed, 09 Oct 2019	Prob (F-statistic):	7.87e-20		
Time:	16:11:24	Log-Likelihood:	-44.537		
No. Observations:	115	AIC:	93.07		
Df Residuals:	113	BIC:	98.56		
Df Model:	1				
Covariance Type:	nonrobust				
=====					
	coef	std err	t	P> t	[0.025 0.975]

const	14.4390	0.458	31.523	0.000	13.532	15.346
rvxe	0.9051	0.081	11.110	0.000	0.744	1.067

```
=====
```

Omnibus:	62.069	Durbin-Watson:	1.838
Prob(Omnibus):	0.000	Jarque-Bera (JB):	336.094
Skew:	-1.739	Prob(JB):	1.04e-73
Kurtosis:	10.619	Cond. No.	79.2

```
=====
```

Warnings:

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

```
[113]: cme.lin_reg_rob(PRIOR_OB_UZ_STATS, 'rvxe', 'M', True)
```

```
Robust linear Model Regression Results
```

```
=====
```

Dep. Variable:	M	No. Observations:	115
Model:	RLM	Df Residuals:	113
Method:	IRLS	Df Model:	1
Norm:	HuberT		
Scale Est.:	mad		
Cov Type:	H1		
Date:	Wed, 09 Oct 2019		
Time:	16:11:24		
No. Iterations:	18		

```
=====
```

	coef	std err	z	P> z	[0.025	0.975]

const	15.5520	0.238	65.373	0.000	15.086	16.018
rvxe	1.0956	0.042	25.893	0.000	1.013	1.179

```
=====
```

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[114]: cme.lin_reg(AFTER_OB_UZ_STATS, 'rvxe', 'M', True)
```

```
OLS Regression Results
```

```
=====
```

Dep. Variable:	M	R-squared:	0.572
Model:	OLS	Adj. R-squared:	0.569
Method:	Least Squares	F-statistic:	164.6
Date:	Wed, 09 Oct 2019	Prob (F-statistic):	1.93e-24
Time:	16:11:24	Log-Likelihood:	-47.153
No. Observations:	125	AIC:	98.31
Df Residuals:	123	BIC:	104.0
Df Model:	1		

Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	14.8451	0.408	36.360	0.000	14.037	15.653
rvxe	0.9043	0.070	12.831	0.000	0.765	1.044
Omnibus:	69.215	Durbin-Watson:	1.222			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	304.490			
Skew:	-1.964	Prob(JB):	7.60e-67			
Kurtosis:	9.560	Cond. No.	76.5			

Warnings:

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

```
[115]: cme.lin_reg_rob(AFTER_OB_UZ_STATS, 'rvxe', 'M', True)
```

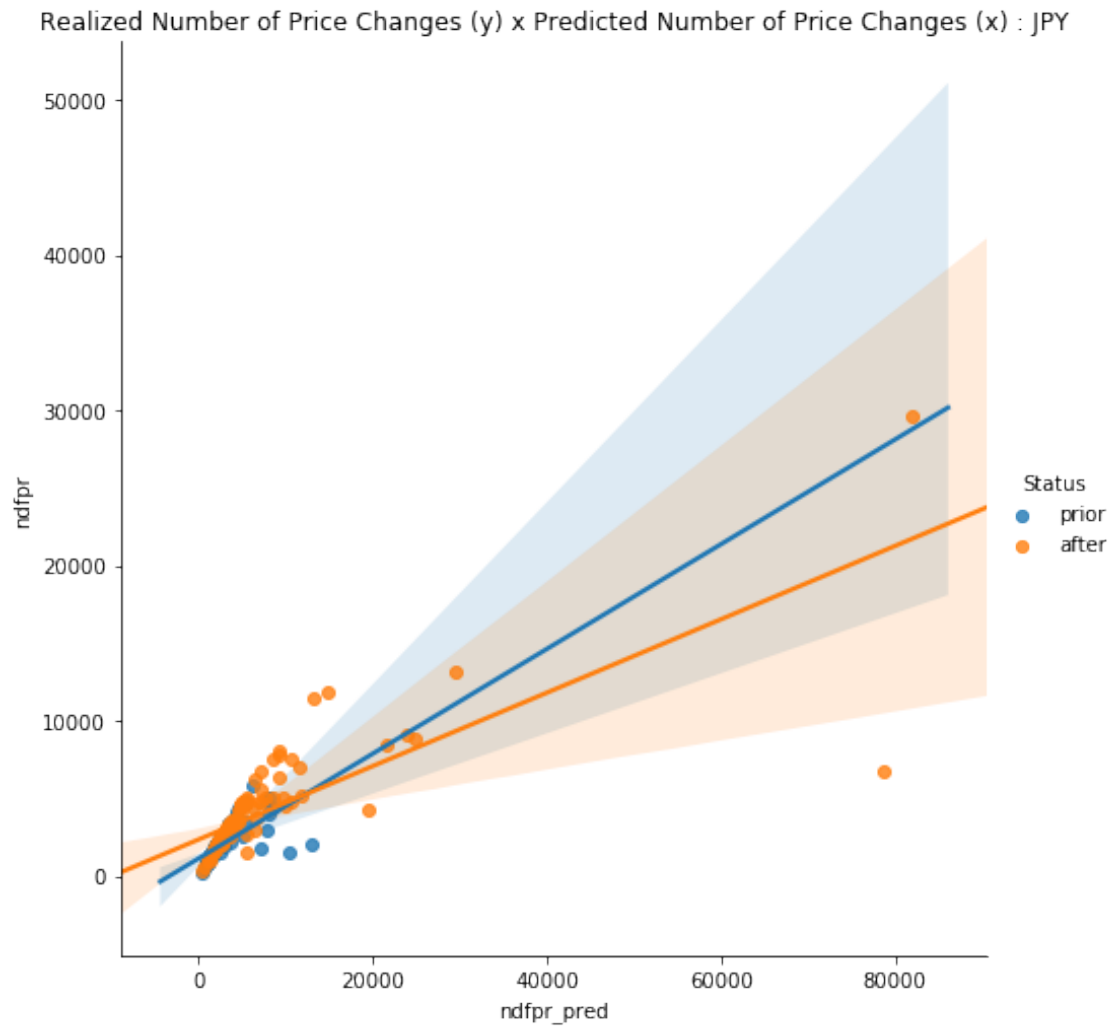
Robust linear Model Regression Results

Dep. Variable:	M	No. Observations:	125			
Model:	RLM	Df Residuals:	123			
Method:	IRLS	Df Model:	1			
Norm:	HuberT					
Scale Est.:	mad					
Cov Type:	H1					
Date:	Wed, 09 Oct 2019					
Time:	16:11:24					
No. Iterations:	19					
=====						
	coef	std err	z	P> z	[0.025	0.975]

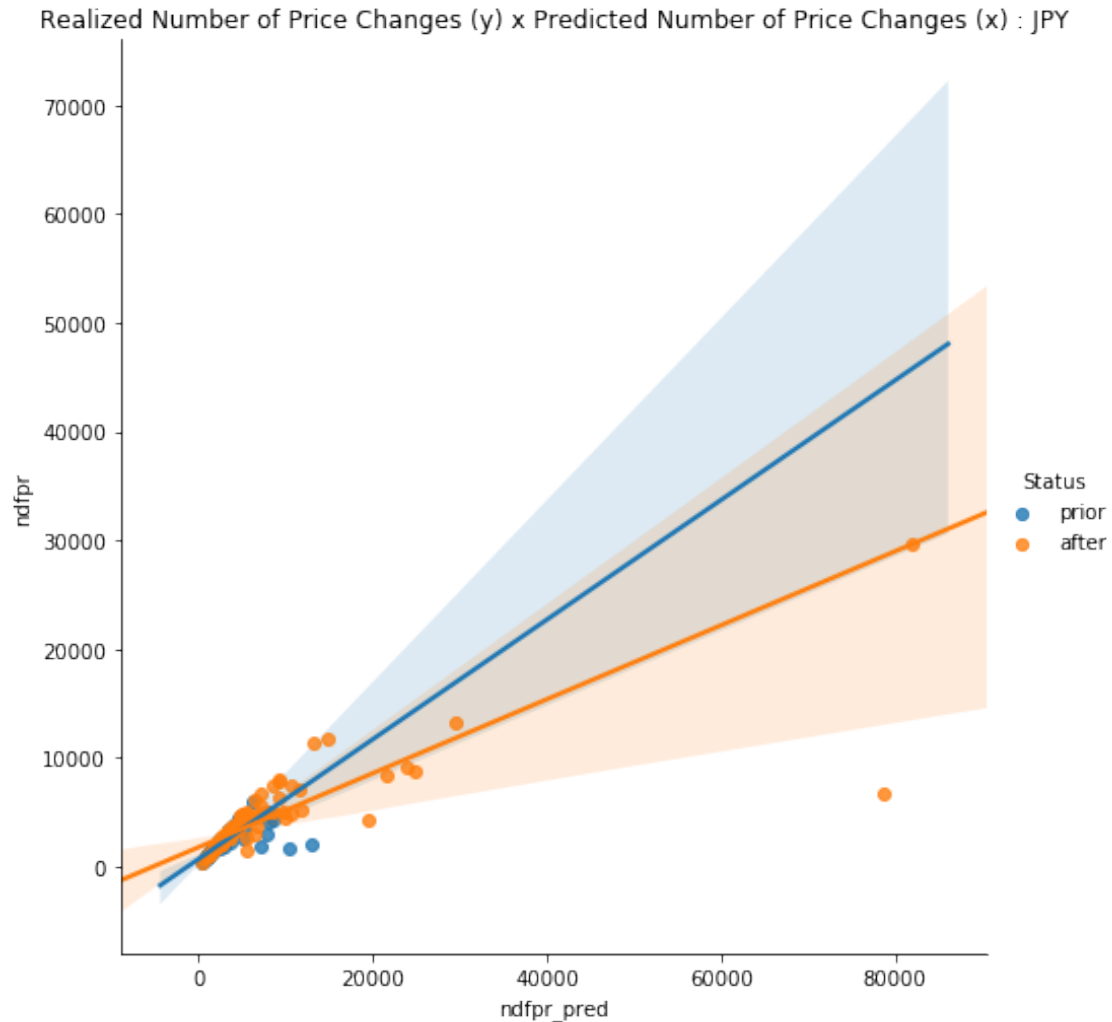
const	14.6011	0.311	46.884	0.000	13.991	15.211
rvxe	0.8556	0.054	15.915	0.000	0.750	0.961
=====						

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[116]: cme.regr_plot(OB_UZ_STATS, 'ndfpr_pred', 'ndfpr', \
    'Realized Number of Price Changes (y) x Predicted Number of Price Changes_\
    ↪(x) : '+CURR)
```



```
[117]: cme.regr_plot(OB_UZ_STATS, 'ndfpr_pred', 'ndfpr',\
    'Realized Number of Price Changes (y) x Predicted Number of Price Changes_\
    ↪(x) : '+CURR, True)
```



```
[118]: cme.lin_reg(PRIOR_OB_UZ_STATS, 'ndfpr_pred', 'ndfpr')
```

OLS Regression Results

```
=====
Dep. Variable:          ndfpr    R-squared:                0.469
Model:                  OLS      Adj. R-squared:           0.464
Method:                 Least Squares    F-statistic:           99.66
Date:                   Wed, 09 Oct 2019    Prob (F-statistic):     3.30e-17
Time:                   16:11:38    Log-Likelihood:         -923.48
No. Observations:       115      AIC:                   1851.
Df Residuals:           113      BIC:                   1856.
Df Model:                1
Covariance Type:        nonrobust
=====
```

```
=====
               coef      std err          t      P>|t|      [0.025      0.975]
-----
```

const	1137.3348	110.685	10.275	0.000	918.048	1356.622
ndfpr_pred	0.3372	0.034	9.983	0.000	0.270	0.404

```
=====
```

Omnibus:	33.430	Durbin-Watson:	1.205
Prob(Omnibus):	0.000	Jarque-Bera (JB):	219.570
Skew:	-0.673	Prob(JB):	2.09e-48
Kurtosis:	9.634	Cond. No.	5.19e+03

```
=====
```

Warnings:

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

[2] The condition number is large, 5.19e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[119]: cme.lin_reg_rob(PRIOR_OB_UZ_STATS, 'ndfpr_pred', 'ndfpr')
```

```

Robust linear Model Regression Results
=====
Dep. Variable:          ndfpr    No. Observations:          115
Model:                  RLM      Df Residuals:              113
Method:                 IRLS     Df Model:                  1
Norm:                   HuberT
Scale Est.:             mad
Cov Type:               H1
Date:                   Wed, 09 Oct 2019
Time:                   16:11:38
No. Iterations:         24
=====

```

	coef	std err	z	P> z	[0.025	0.975]
const	682.0525	47.193	14.452	0.000	589.556	774.549
ndfpr_pred	0.5505	0.014	38.218	0.000	0.522	0.579

```
=====
```

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[120]: cme.lin_reg(AFTER_OB_UZ_STATS, 'ndfpr_pred', 'ndfpr')
```

```

OLS Regression Results
=====
Dep. Variable:          ndfpr    R-squared:              0.613
Model:                  OLS      Adj. R-squared:         0.610
Method:                 Least Squares
F-statistic:            194.7
Date:                   Wed, 09 Oct 2019
Time:                   16:11:38
Log-Likelihood:         -1126.7
AIC:                    2257.
No. Observations:       125
Prob (F-statistic):     4.10e-27

```

```

Df Residuals:          123    BIC:          2263.
Df Model:              1
Covariance Type:      nonrobust

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const          2355.3637      207.299      11.362      0.000      1945.028      2765.699
ndfpr_pred        0.2363       0.017      13.954      0.000         0.203         0.270
=====
Omnibus:                90.852    Durbin-Watson:                1.298
Prob(Omnibus):           0.000    Jarque-Bera (JB):        2561.497
Skew:                   -1.873    Prob(JB):                 0.00
Kurtosis:               24.858    Cond. No.                1.42e+04
=====

```

Warnings:

```

[1] Standard Errors assume that the covariance matrix of the errors is correctly
specified.
[2] The condition number is large, 1.42e+04. This might indicate that there are
strong multicollinearity or other numerical problems.

```

```
[121]: cme.lin_reg_rob(AFTER_OB_UZ_STATS, 'ndfpr_pred', 'ndfpr')
```

Robust linear Model Regression Results

```

=====
Dep. Variable:          ndfpr    No. Observations:          125
Model:                RLM      Df Residuals:              123
Method:              IRLS      Df Model:                1
Norm:                HuberT
Scale Est.:          mad
Cov Type:            H1
Date:                Wed, 09 Oct 2019
Time:                16:11:38
No. Iterations:      15
=====
              coef      std err          z      P>|z|      [0.025      0.975]
-----
const          1757.2247      90.328      19.454      0.000      1580.185      1934.264
ndfpr_pred        0.3403       0.007      46.127      0.000         0.326         0.355
=====

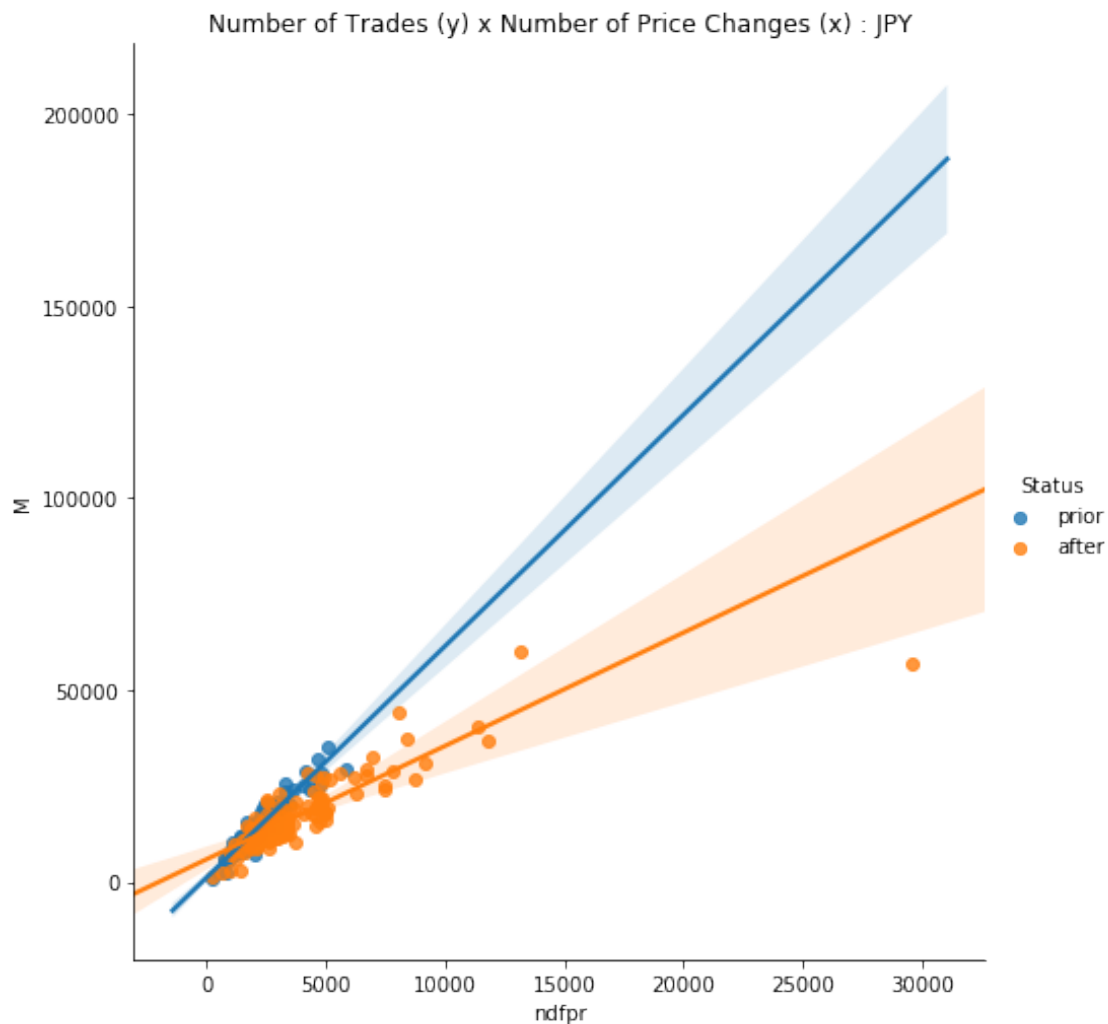
```

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[122]: cme.regr_plot(OB_UZ_STATS, 'ndfpr', 'M',\
                    'Number of Trades (y) x Number of Price Changes (x) : '+CURR)
```



```
[123]: cme.regr_plot(OB_UZ_STATS, 'ndfpr', 'M', \
    'Number of Trades (y) x Number of Price Changes (x) : '+CURR, True)
```



```
[124]: cme.lin_reg(PRIOR_OB_UZ_STATS, 'ndfpr', 'M')
```

OLS Regression Results

```
=====
Dep. Variable:          M      R-squared:          0.909
Model:                  OLS    Adj. R-squared:       0.908
Method:                 Least Squares    F-statistic:      1133.
Date:                   Wed, 09 Oct 2019    Prob (F-statistic): 1.00e-60
Time:                   16:11:44    Log-Likelihood:    -1032.0
No. Observations:      115    AIC:              2068.
Df Residuals:          113    BIC:              2073.
Df Model:              1
Covariance Type:       nonrobust
=====
```

```
=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
```

const	1288.8412	394.509	3.267	0.001	507.247	2070.436
ndfpr	5.9288	0.176	33.656	0.000	5.580	6.278

```
=====
```

Omnibus:	7.519	Durbin-Watson:	1.181
Prob(Omnibus):	0.023	Jarque-Bera (JB):	12.085
Skew:	-0.220	Prob(JB):	0.00238
Kurtosis:	4.526	Cond. No.	4.92e+03

```
=====
```

Warnings:

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

[2] The condition number is large, 4.92e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
[125]: cme.lin_reg_rob(PRIOR_OB_UZ_STATS, 'ndfpr', 'M')
```

```

Robust linear Model Regression Results
=====
Dep. Variable:          M    No. Observations:          115
Model:                RLM    Df Residuals:              113
Method:               IRLS    Df Model:                  1
Norm:                 HuberT
Scale Est.:           mad
Cov Type:             H1
Date:                 Wed, 09 Oct 2019
Time:                 16:11:44
No. Iterations:       7
=====

```

	coef	std err	z	P> z	[0.025	0.975]
const	1087.7832	353.087	3.081	0.002	395.745	1779.821
ndfpr	6.0329	0.158	38.264	0.000	5.724	6.342

```
=====
```

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[126]: cme.lin_reg(AFTER_OB_UZ_STATS, 'ndfpr', 'M')
```

```

OLS Regression Results
=====
Dep. Variable:          M    R-squared:          0.727
Model:                OLS    Adj. R-squared:     0.725
Method:               Least Squares    F-statistic:        328.1
Date:                 Wed, 09 Oct 2019    Prob (F-statistic):  1.65e-36
Time:                 16:11:44    Log-Likelihood:     -1236.4
No. Observations:     125    AIC:                2477.

```



```
Df Residuals:      123    BIC:      2482.
Df Model:          1
Covariance Type:  nonrobust
```

```
=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const      7900.9868      670.723      11.780      0.000      6573.333      9228.641
ndfpr       2.4437       0.135      18.113      0.000       2.177       2.711
=====
Omnibus:            25.963    Durbin-Watson:           1.047
Prob(Omnibus):      0.000    Jarque-Bera (JB):        204.287
Skew:               0.153    Prob(JB):               4.36e-45
Kurtosis:           9.255    Cond. No.                7.74e+03
=====
```

Warnings:

```
[1] Standard Errors assume that the covariance matrix of the errors is correctly
specified.
[2] The condition number is large, 7.74e+03. This might indicate that there are
strong multicollinearity or other numerical problems.
```

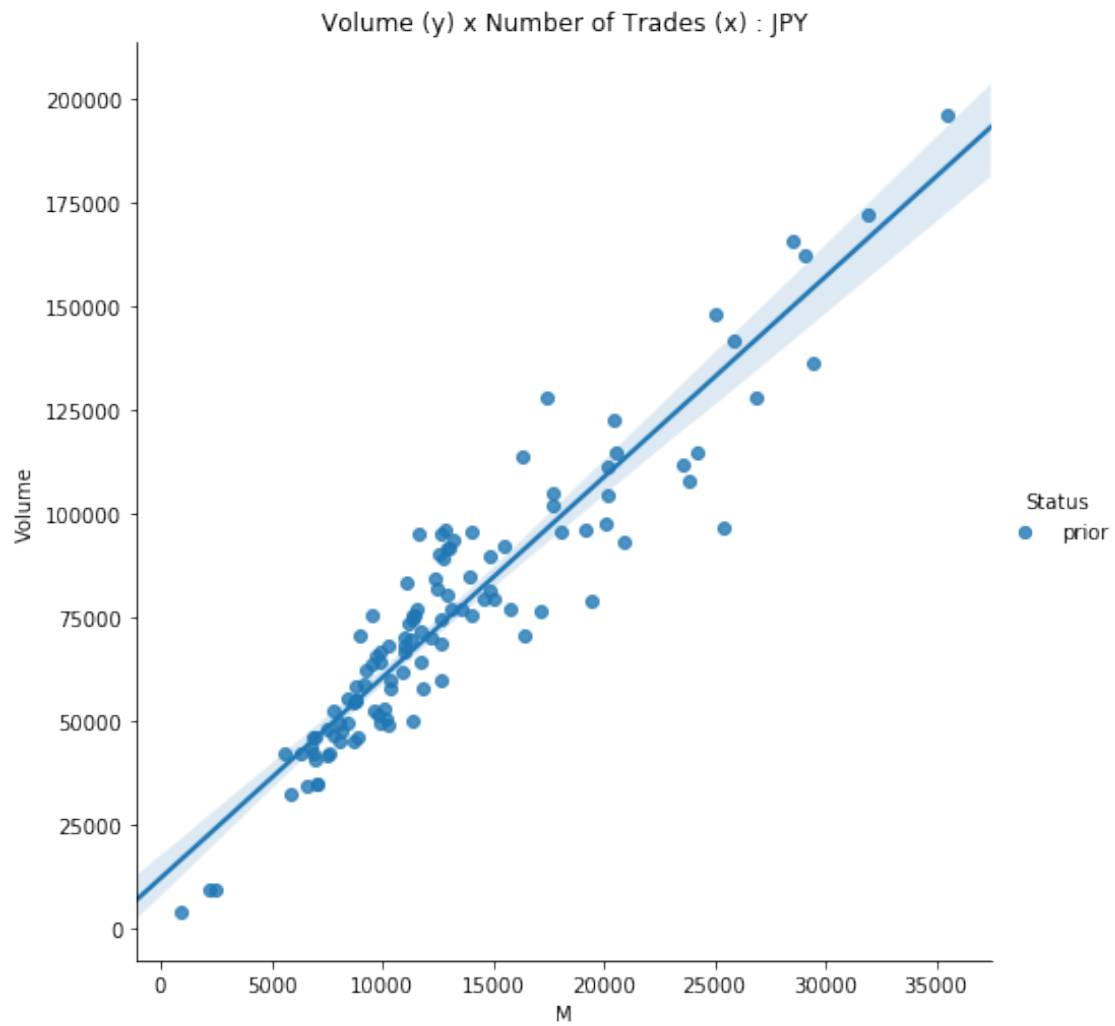
```
[127]: cme.lin_reg_rob(AFTER_OB_UZ_STATS, 'ndfpr', 'M')
```

Robust linear Model Regression Results

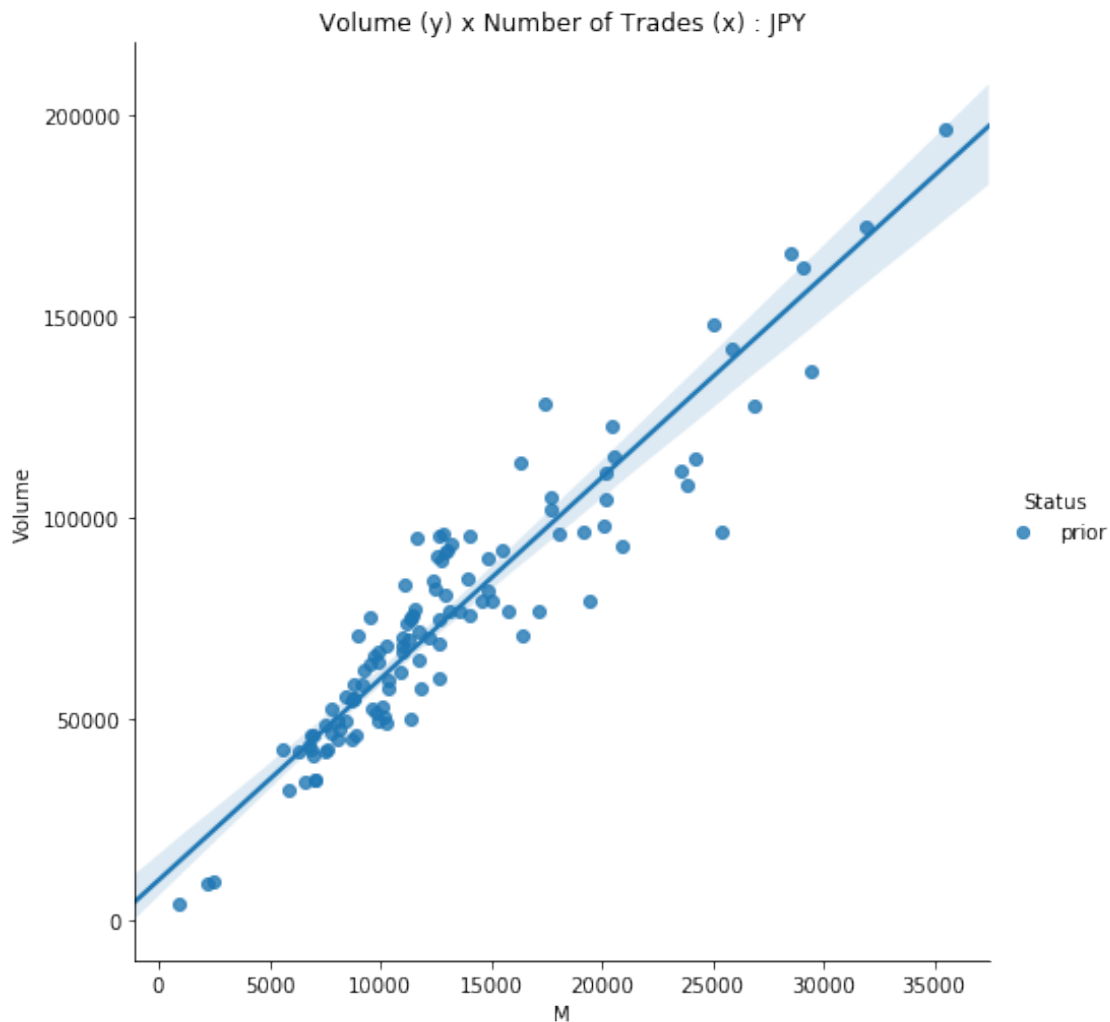
```
=====
Dep. Variable:      M    No. Observations:      125
Model:              RLM    Df Residuals:        123
Method:             IRLS    Df Model:          1
Norm:               HuberT
Scale Est.:         mad
Cov Type:           H1
Date:               Wed, 09 Oct 2019
Time:               16:11:44
No. Iterations:     9
=====
              coef      std err          z      P>|z|      [0.025      0.975]
-----
const      5883.7790      500.390      11.758      0.000      4903.033      6864.525
ndfpr       2.9554       0.101      29.362      0.000       2.758       3.153
=====
```

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[128]: cme.regr_plot(PRIOR_OB_UZ_STATS, 'M', 'Volume',\
                    'Volume (y) x Number of Trades (x) : '+CURR)
```



```
[129]: cme.regr_plot(PRIOR_OB_UZ_STATS, 'M', 'Volume',\
    'Volume (y) x Number of Trades (x) : '+CURR, True)
```



```
[130]: cme.lin_reg(PRIOR_OB_UZ_STATS, 'M', 'Volume')
```

OLS Regression Results

=====					
Dep. Variable:	Volume	R-squared:	0.875		
Model:	OLS	Adj. R-squared:	0.874		
Method:	Least Squares	F-statistic:	789.6		
Date:	Wed, 09 Oct 2019	Prob (F-statistic):	8.24e-53		
Time:	16:11:47	Log-Likelihood:	-1239.5		
No. Observations:	115	AIC:	2483.		
Df Residuals:	113	BIC:	2488.		
Df Model:	1				
Covariance Type:	nonrobust				
=====					
	coef	std err	t	P> t	[0.025 0.975]

const	1.209e+04	2507.128	4.822	0.000	7122.220	1.71e+04
M	4.8377	0.172	28.100	0.000	4.497	5.179

```
=====
```

Omnibus:	1.366	Durbin-Watson:	1.022
Prob(Omnibus):	0.505	Jarque-Bera (JB):	0.897
Skew:	-0.045	Prob(JB):	0.638
Kurtosis:	3.423	Cond. No.	3.34e+04

```
=====
```

Warnings:

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

[2] The condition number is large, 3.34e+04. This might indicate that there are strong multicollinearity or other numerical problems.

```
[131]: cme.lin_reg_rob(PRIOR_OB_UZ_STATS, 'M', 'Volume')
```

```
Robust linear Model Regression Results
```

```
=====
```

Dep. Variable:	Volume	No. Observations:	115
Model:	RLM	Df Residuals:	113
Method:	IRLS	Df Model:	1
Norm:	HuberT		
Scale Est.:	mad		
Cov Type:	H1		
Date:	Wed, 09 Oct 2019		
Time:	16:11:47		
No. Iterations:	4		

```
=====
```

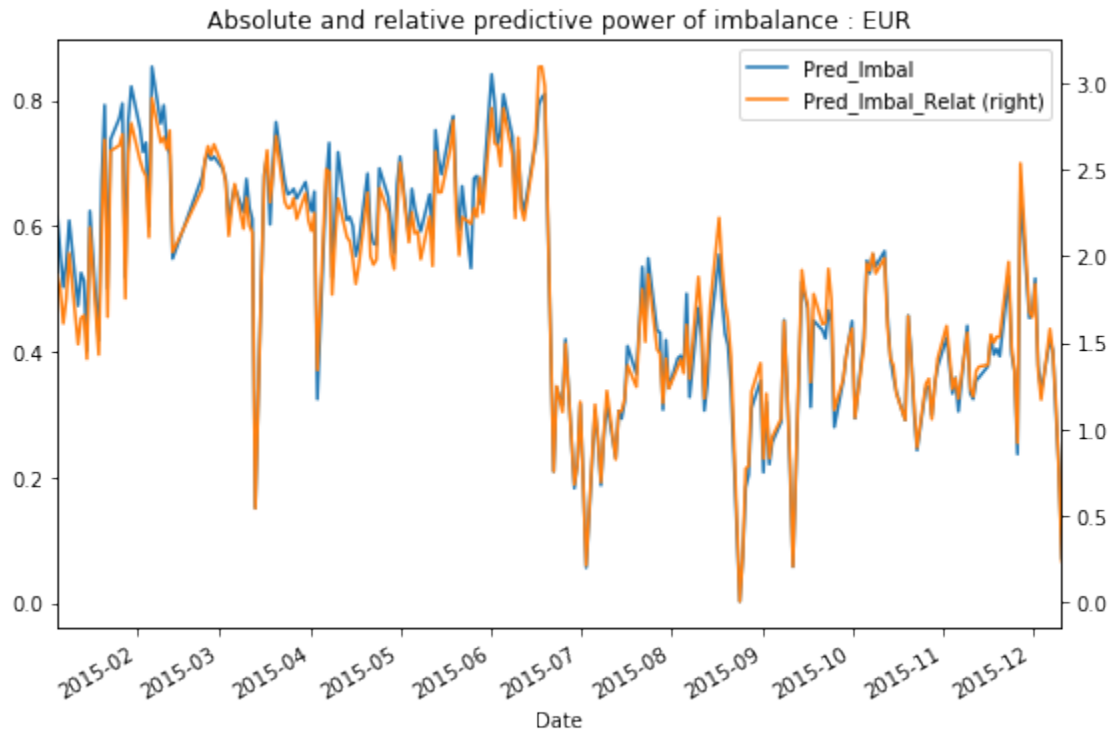
	coef	std err	z	P> z	[0.025	0.975]

const	9923.0892	2430.707	4.082	0.000	5158.992	1.47e+04
M	5.0035	0.167	29.977	0.000	4.676	5.331

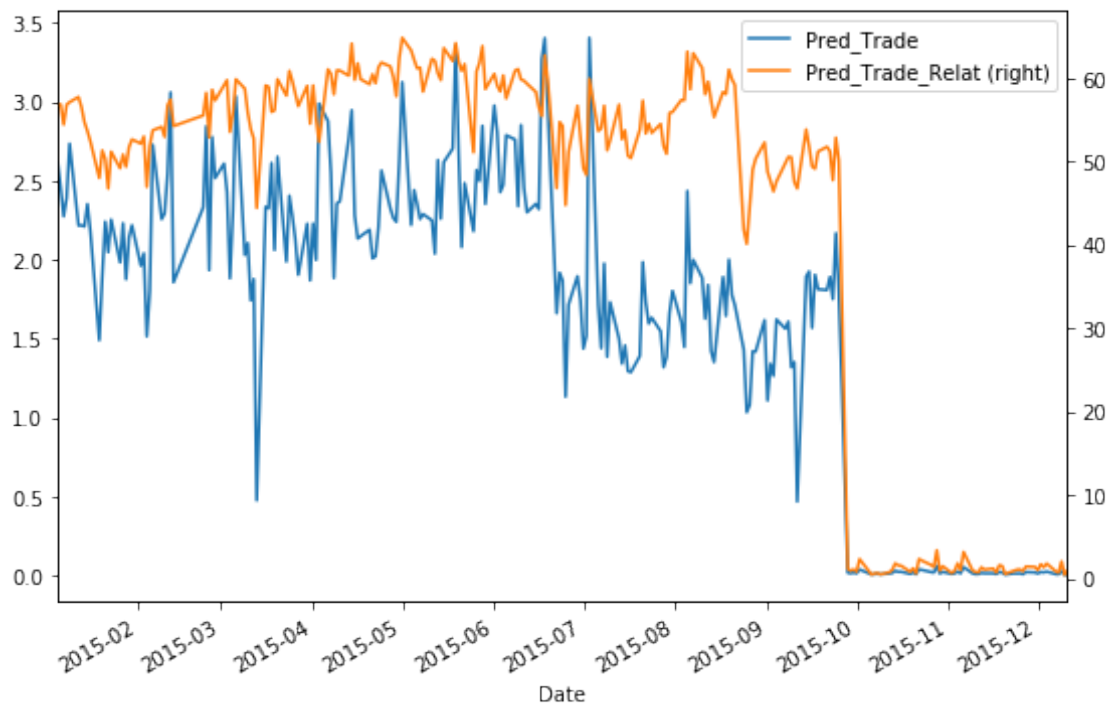
```
=====
```

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[132]: IMBAL_STATS_TS.drop(columns=['eta1']).plot(secondary_y=['Pred_Imbal_Relat'],\
        figsize=(9,6), title='Absolute and relative predictive power of imbalance :_\
        ↳EUR');
```

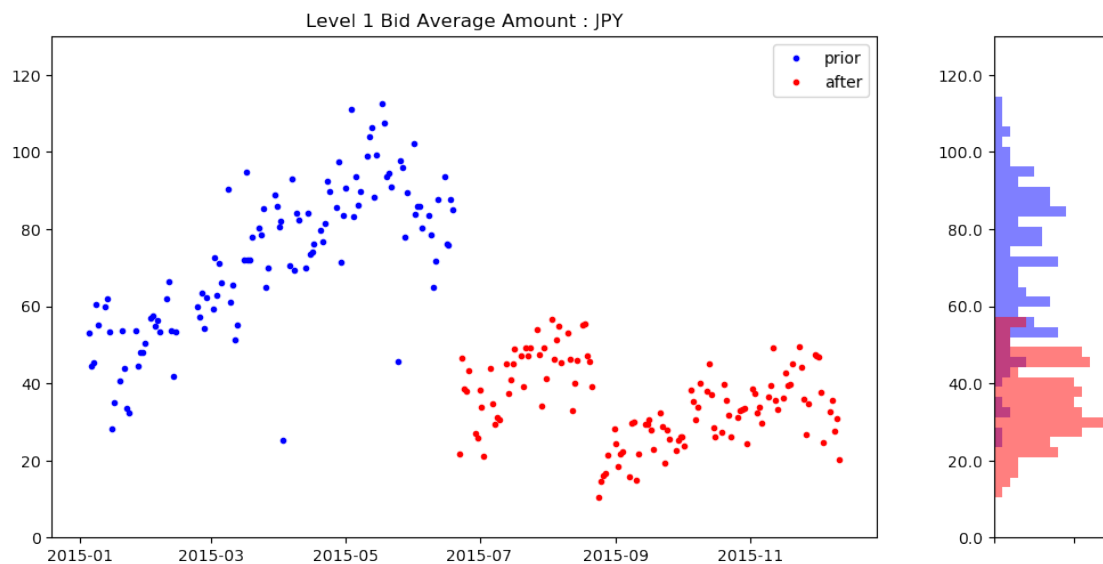


```
[133]: TRADE_STATS_TS.plot(secondary_y=['Pred_Trade_Relat'], figsize=(9,6));
```

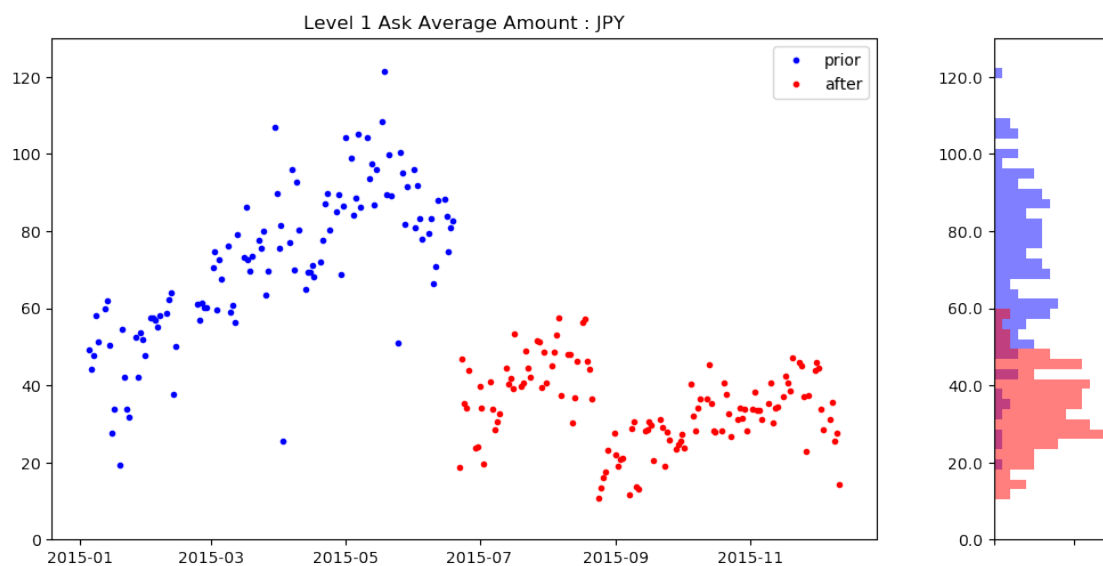


```
[134]: OB_UZ_STATS_SPREADS = cme.spread_stats(OB_UZ_STATS)
```

```
[195]: cme.time_series_hist_plot(OB_UZ_STATS_SPREADS, 'bid1qty',\
    'Level 1 Bid Average Amount : '+CURR, 0, 130, 50)
```



```
[197]: cme.time_series_hist_plot(OB_UZ_STATS_SPREADS, 'ask1qty',\
    'Level 1 Ask Average Amount : '+CURR, 0, 130, 50)
```



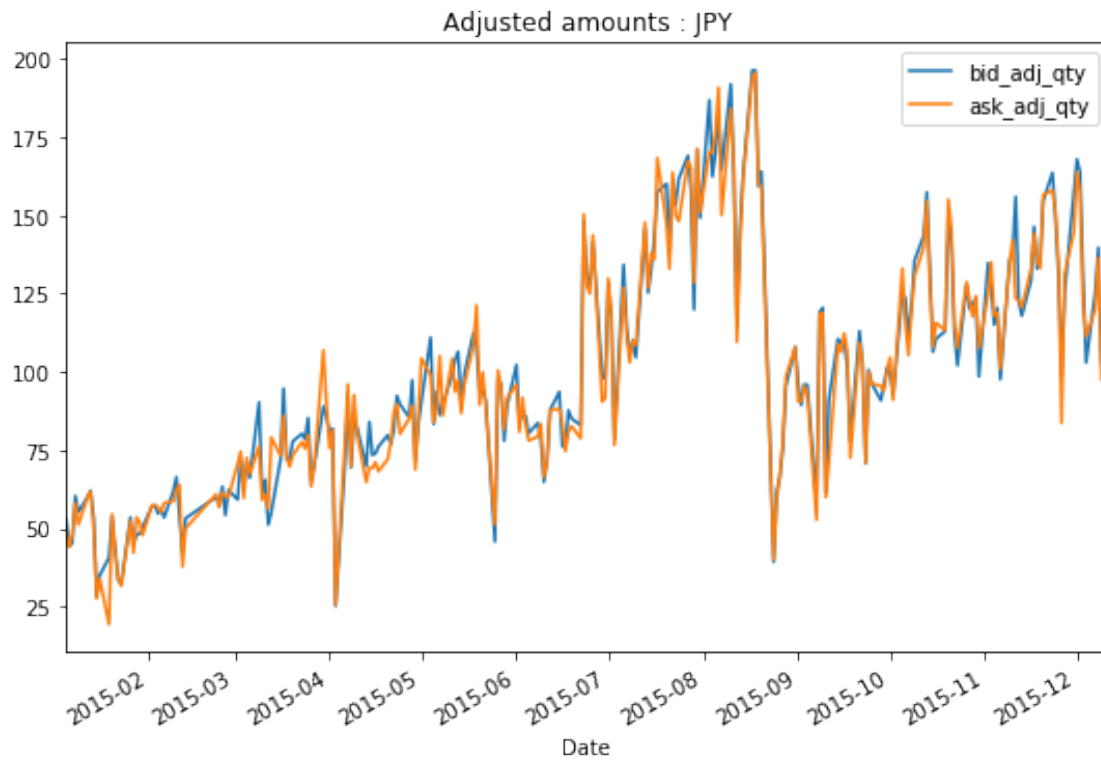
```
[137]: OB_UZ_STATS_SPREADS[['bid1qty', 'ask1qty']].plot(figsize=(9,6));
```



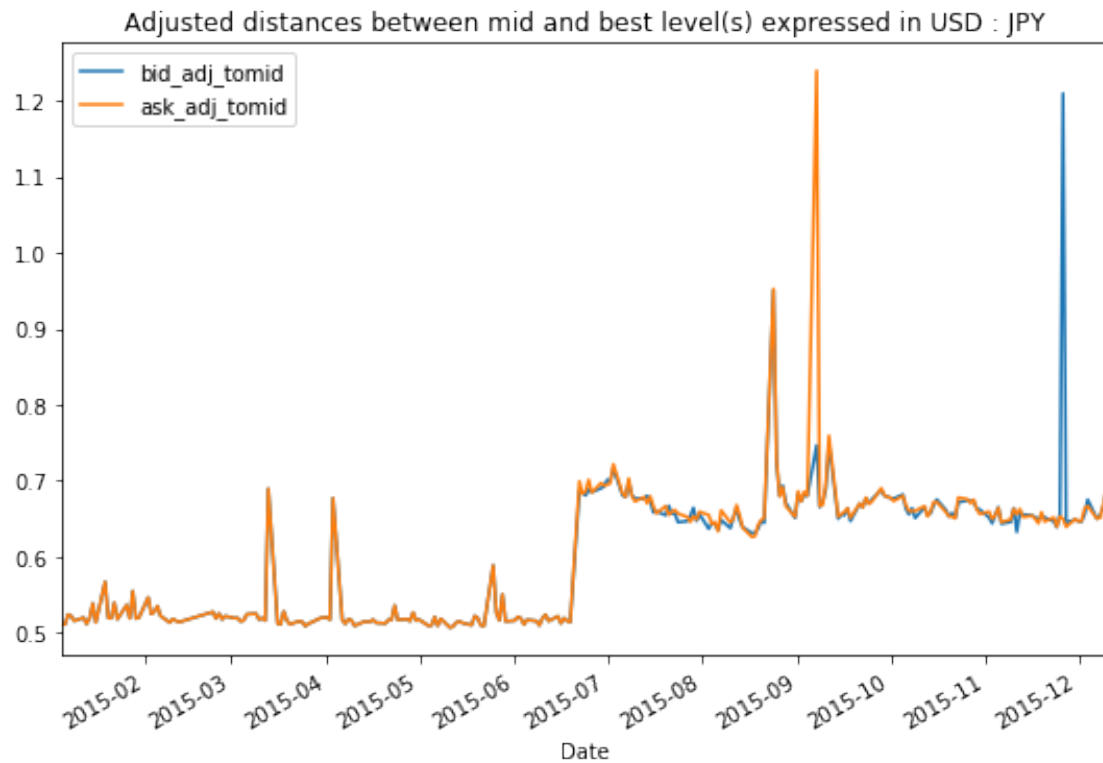
```
[138]: OB_UZ_STATS_SPREADS[OB_UZ_STATS_SPREADS['Status'] == 'prior'][['bid1qty',
↪ 'ask1qty']].mean()/\
      OB_UZ_STATS_SPREADS[OB_UZ_STATS_SPREADS['Status'] == 'after'][['bid1qty',
↪ 'ask1qty']].mean()
```

```
[138]: bid1qty    2.066217
ask1qty    2.104551
dtype: float64
```

```
[139]: OB_UZ_STATS_SPREADS[['bid_adj_qty', 'ask_adj_qty']].plot(figsize=(9,6),\
      title='Adjusted amounts : '+CURR);
```

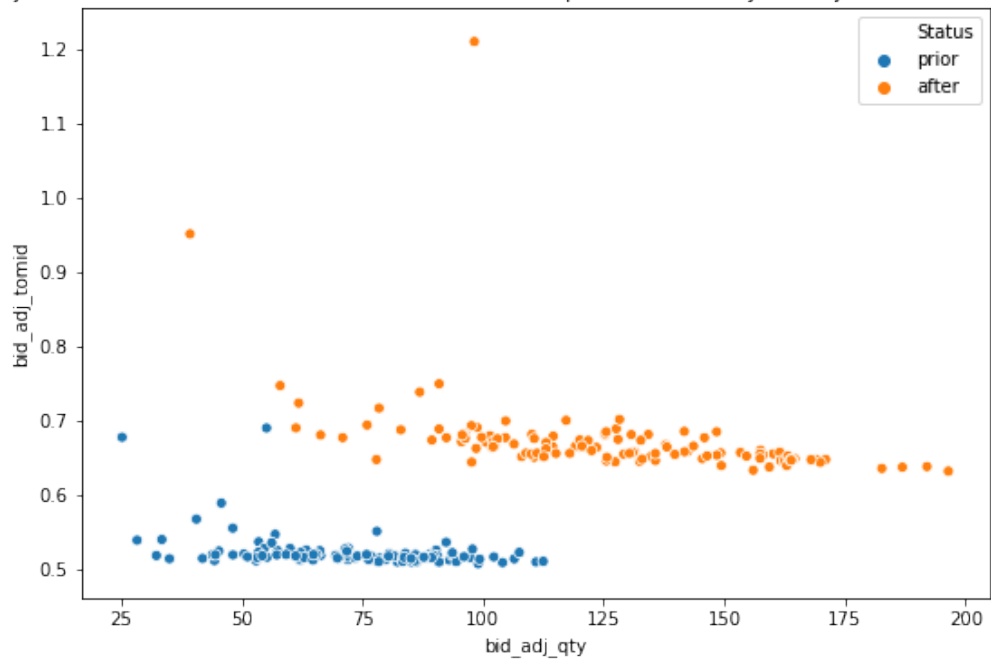


```
[140]: OB_UZ_STATS_SPREADS[['bid_adj_tomid', 'ask_adj_tomid']].plot(figsize=(9,6),\
    title='Adjusted distances between mid and best level(s) expressed in USD :_↵\
    ↵'+CURR);
```

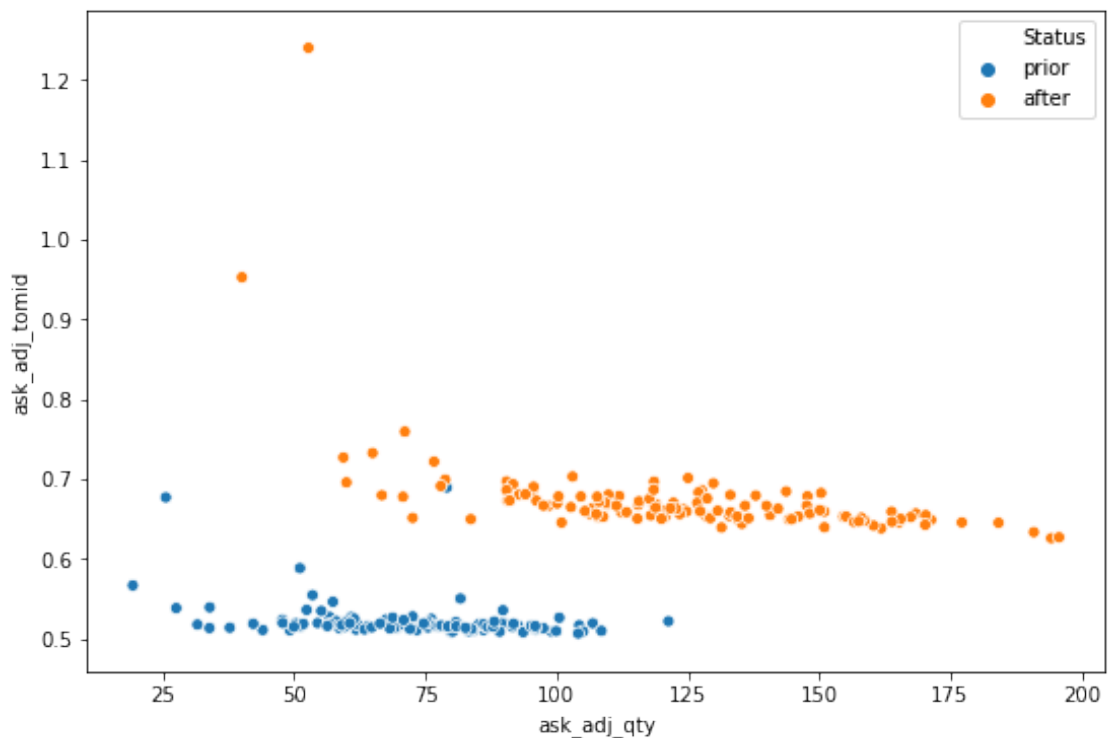



```
[141]: plt.figure(figsize=(9, 6))
sns.scatterplot(x='bid_adj_qty', y='bid_adj_tomid',\
                hue='Status', data=OB_UZ_STATS_SPREADS);
plt.title('Adjusted distances between mid and best level(s) expressed in USD_\
↪(y) vs Adjusted amount (x) : '+CURR);
```

Adjusted distances between mid and best level(s) expressed in USD (y) vs Adjusted amount (x) : JPY



```
[142]: plt.figure(figsize=(9, 6))
sns.scatterplot(x='ask_adj_qty', y='ask_adj_tomid',\
                hue='Status', data=OB_UZ_STATS_SPREADS);
```



2.8.1 Costs

```
[143]: PRIOR_MEAN_COST = cme.cost_mean(PRIOR_COST_STATS, 100)

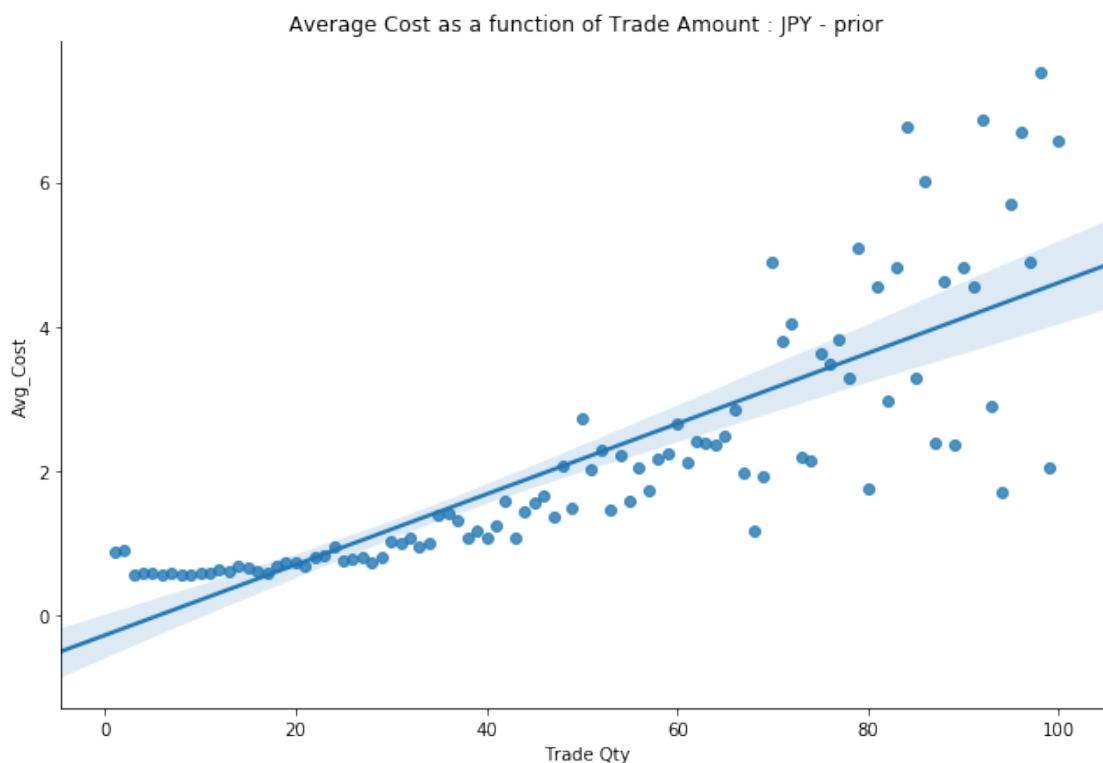
[144]: PRIOR_MEAN_COST['Status'] = 'prior'

[145]: AFTER_MEAN_COST = cme.cost_mean(AFTER_COST_STATS, 100)

[146]: AFTER_MEAN_COST['Status'] = 'after'

[147]: MEAN_COST_STATS = pd.concat([PRIOR_MEAN_COST, AFTER_MEAN_COST], sort=False)

[148]: sns.lmplot(x='Trade Qty', y='Avg_Cost', data=PRIOR_MEAN_COST.reset_index(),\
                height=6, aspect=1.5);
plt.title('Average Cost as a function of Trade Amount : '+CURR+' - prior');
```



```
[149]: sns.lmplot(x='Trade Qty', y='Avg_Cost', data=PRIOR_MEAN_COST.reset_index(),\
                height=6, aspect=1.5, robust=True);
plt.title('Average Cost as a function of Trade Amount : '+CURR+' - prior');
```



```
[150]: cme.lin_reg(cme.cost_mean(PRIOR_COST_STATS, 50).reset_index(), 'Trade Qty',
↪ 'Avg_Cost')
```

OLS Regression Results

```
=====
Dep. Variable:          Avg_Cost    R-squared:                0.669
Model:                  OLS         Adj. R-squared:            0.662
Method:                 Least Squares   F-statistic:              96.89
Date:                  Wed, 09 Oct 2019   Prob (F-statistic):       4.24e-13
Time:                  16:11:58         Log-Likelihood:           -1.4328
No. Observations:      50             AIC:                     6.866
Df Residuals:          48             BIC:                     10.69
Df Model:              1
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.3491	0.073	4.784	0.000	0.202	0.496
Trade Qty	0.0245	0.002	9.843	0.000	0.020	0.030

```
=====
Omnibus:                43.627    Durbin-Watson:           1.118
Prob(Omnibus):          0.000    Jarque-Bera (JB):        161.855
Skew:                   2.326    Prob(JB):                 7.14e-36
=====
```

Kurtosis: 10.487 Cond. No. 59.5

Warnings:

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

```
[151]: cme.lin_reg_rob(cme.cost_mean(PRIOR_COST_STATS, 50).reset_index(), 'Trade Qty',  
↳ 'Avg_Cost')
```

Robust linear Model Regression Results

```
=====
Dep. Variable:          Avg_Cost    No. Observations:          50
Model:                  RLM        Df Residuals:              48
Method:                 IRLS       Df Model:                  1
Norm:                   HuberT
Scale Est.:             mad
Cov Type:               H1
Date:                   Wed, 09 Oct 2019
Time:                   16:11:58
No. Iterations:         24
=====
```

	coef	std err	z	P> z	[0.025	0.975]
const	0.3657	0.051	7.125	0.000	0.265	0.466
Trade Qty	0.0223	0.002	12.750	0.000	0.019	0.026

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[152]: cme.lin_reg(cme.cost_mean(PRIOR_COST_STATS, 100).reset_index(), 'Trade Qty',  
↳ 'Avg_Cost')
```

OLS Regression Results

```
=====
Dep. Variable:          Avg_Cost    R-squared:              0.675
Model:                  OLS        Adj. R-squared:          0.671
Method:                 Least Squares    F-statistic:          203.1
Date:                   Wed, 09 Oct 2019    Prob (F-statistic):    1.26e-25
Time:                   16:11:58    Log-Likelihood:        -139.84
No. Observations:       100        AIC:                  283.7
Df Residuals:           98        BIC:                  288.9
Df Model:                1
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
--	------	---------	---	------	--------	--------

const	-0.2767	0.199	-1.387	0.168	-0.672	0.119
Trade Qty	0.0489	0.003	14.252	0.000	0.042	0.056

Omnibus:	11.623	Durbin-Watson:	2.148
Prob(Omnibus):	0.003	Jarque-Bera (JB):	16.776
Skew:	0.535	Prob(JB):	0.000228
Kurtosis:	4.698	Cond. No.	117.

Warnings:

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

```
[153]: cme.lin_reg_rob(cme.cost_mean(PRIOR_COST_STATS, 100).reset_index(), 'Trade_
↪Qty', 'Avg_Cost')
```

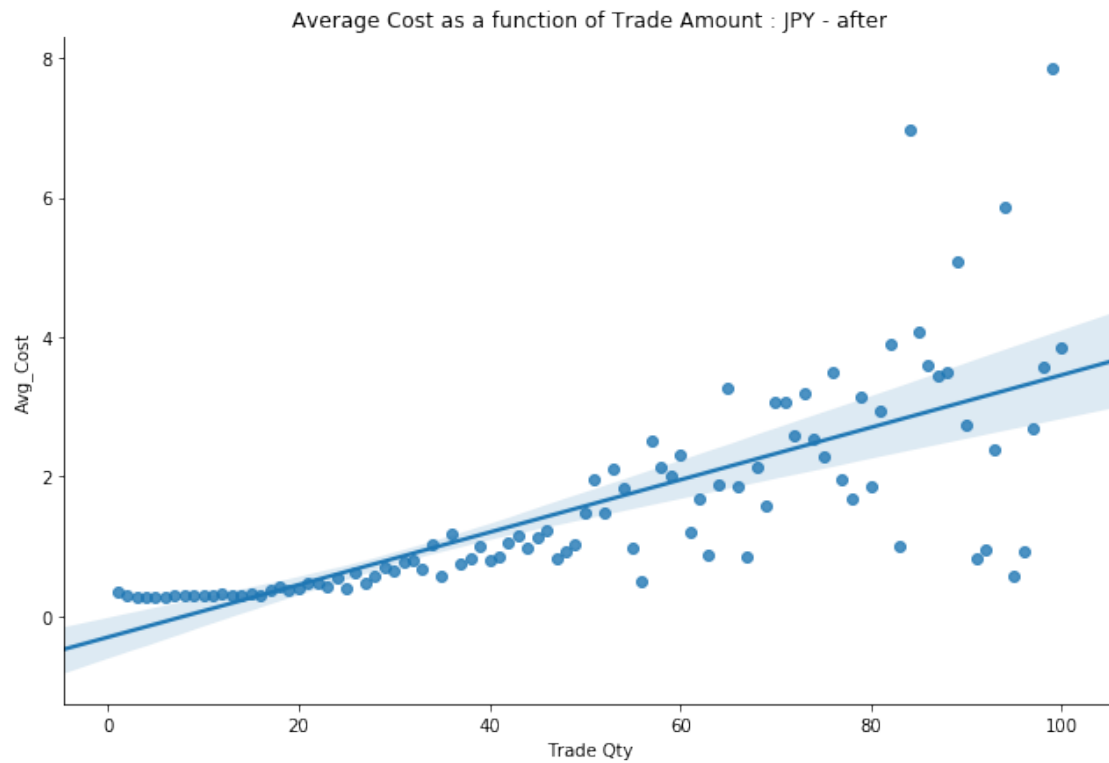
Robust linear Model Regression Results

Dep. Variable:	Avg_Cost	No. Observations:	100
Model:	RLM	Df Residuals:	98
Method:	IRLS	Df Model:	1
Norm:	HuberT		
Scale Est.:	mad		
Cov Type:	H1		
Date:	Wed, 09 Oct 2019		
Time:	16:11:58		
No. Iterations:	37		

	coef	std err	z	P> z	[0.025	0.975]
const	-0.1532	0.146	-1.050	0.294	-0.439	0.133
Trade Qty	0.0445	0.003	17.721	0.000	0.040	0.049

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[154]: sns.lmplot(x='Trade Qty', y='Avg_Cost', data=AFTER_MEAN_COST.reset_index(),\
height=6, aspect=1.5);
plt.title('Average Cost as a function of Trade Amount : '+CURR+ ' - after');
```



```
[155]: sns.lmplot(x='Trade Qty', y='Avg_Cost', data=AFTER_MEAN_COST.reset_index(),\
               height=6, aspect=1.5, robust=True);\
plt.title('Average Cost as a function of Trade Amount : '+CURR+' - after');
```



```
[156]: cme.lin_reg(cme.cost_mean(AFTER_COST_STATS, 50).reset_index(), 'Trade Qty',
↪ 'Avg_Cost')
```

OLS Regression Results

```
=====
Dep. Variable:          Avg_Cost    R-squared:                0.831
Model:                  OLS         Adj. R-squared:           0.828
Method:                 Least Squares   F-statistic:             236.4
Date:                   Wed, 09 Oct 2019   Prob (F-statistic):      3.58e-20
Time:                   16:12:05         Log-Likelihood:          30.399
No. Observations:       50             AIC:                    -56.80
Df Residuals:           48             BIC:                    -52.97
Df Model:               1
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0992	0.039	2.570	0.013	0.022	0.177
Trade Qty	0.0203	0.001	15.374	0.000	0.018	0.023

```
=====
Omnibus:                6.257    Durbin-Watson:           1.786
Prob(Omnibus):           0.044    Jarque-Bera (JB):         5.337
Skew:                    0.775    Prob(JB):                 0.0694
=====
```


Kurtosis: 3.398 Cond. No. 59.5

Warnings:

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

```
[157]: cme.lin_reg_rob(cme.cost_mean(AFTER_COST_STATS, 50).reset_index(), 'Trade Qty',  
↳ 'Avg_Cost')
```

Robust linear Model Regression Results

```
=====
Dep. Variable:          Avg_Cost    No. Observations:          50
Model:                  RLM        Df Residuals:              48
Method:                 IRLS       Df Model:                  1
Norm:                   HuberT
Scale Est.:             mad
Cov Type:               H1
Date:                   Wed, 09 Oct 2019
Time:                   16:12:05
No. Iterations:         21
=====
```

	coef	std err	z	P> z	[0.025	0.975]
const	0.1086	0.036	2.976	0.003	0.037	0.180
Trade Qty	0.0194	0.001	15.583	0.000	0.017	0.022

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[158]: cme.lin_reg(cme.cost_mean(AFTER_COST_STATS, 100).reset_index(), 'Trade Qty',  
↳ 'Avg_Cost')
```

OLS Regression Results

```
=====
Dep. Variable:          Avg_Cost    R-squared:              0.548
Model:                  OLS        Adj. R-squared:          0.543
Method:                 Least Squares    F-statistic:          118.6
Date:                   Wed, 09 Oct 2019    Prob (F-statistic):    1.41e-18
Time:                   16:12:05    Log-Likelihood:        -140.55
No. Observations:       100    AIC:                   285.1
Df Residuals:           98    BIC:                   290.3
Df Model:                1
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
--	------	---------	---	------	--------	--------

const	-0.3018	0.201	-1.503	0.136	-0.700	0.097
Trade Qty	0.0376	0.003	10.893	0.000	0.031	0.044

```
=====
```

Omnibus:	42.575	Durbin-Watson:	1.804
Prob(Omnibus):	0.000	Jarque-Bera (JB):	207.085
Skew:	1.245	Prob(JB):	1.08e-45
Kurtosis:	9.595	Cond. No.	117.

```
=====
```

Warnings:

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

```
[159]: cme.lin_reg_rob(cme.cost_mean(AFTER_COST_STATS, 100).reset_index(), 'Trade_Qty', 'Avg_Cost')
```

```
Robust linear Model Regression Results
```

```
=====
```

Dep. Variable:	Avg_Cost	No. Observations:	100
Model:	RLM	Df Residuals:	98
Method:	IRLS	Df Model:	1
Norm:	HuberT		
Scale Est.:	mad		
Cov Type:	H1		
Date:	Wed, 09 Oct 2019		
Time:	16:12:05		
No. Iterations:	24		

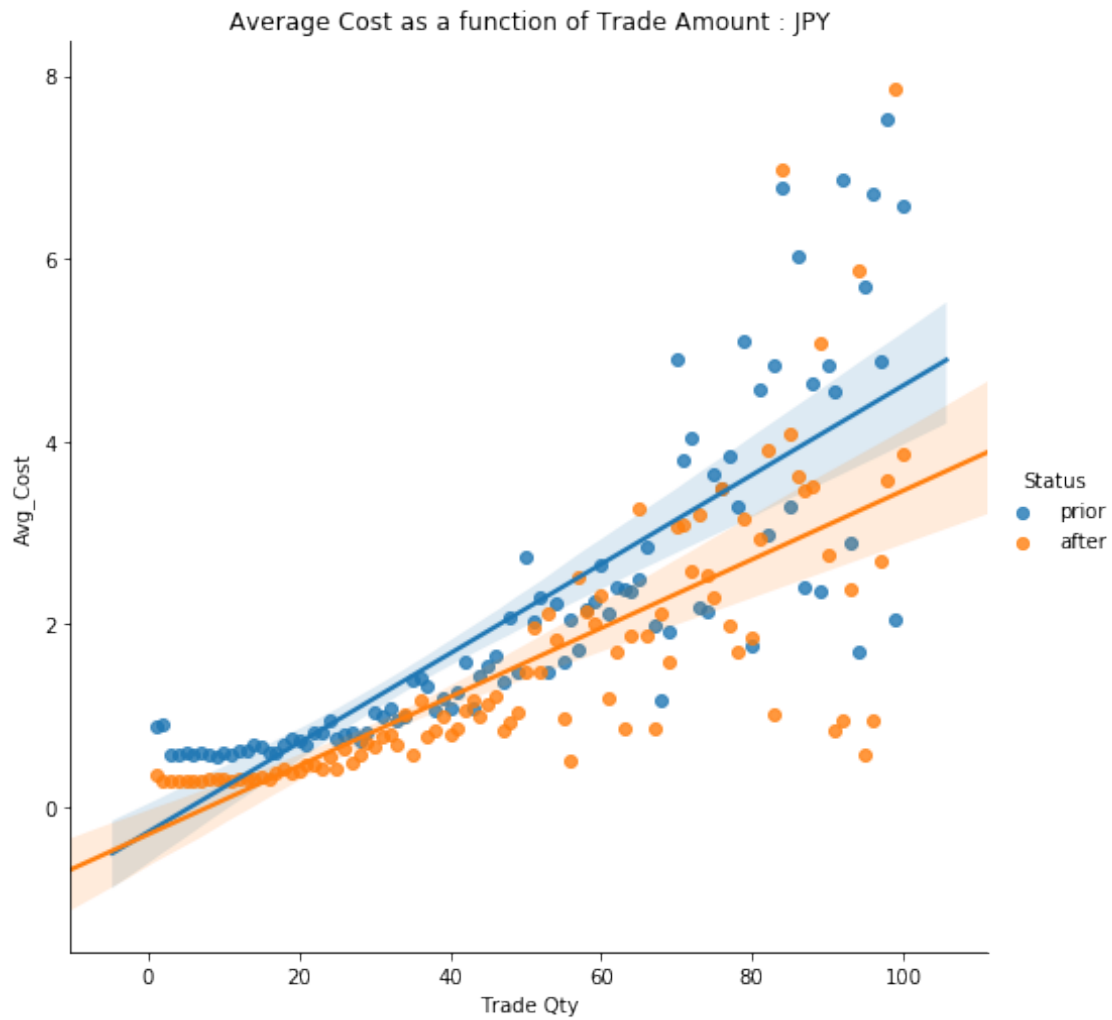
```
=====
```

	coef	std err	z	P> z	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	-0.2237	0.117	-1.917	0.055	-0.452	0.005
Trade Qty	0.0351	0.002	17.504	0.000	0.031	0.039

```
=====
```

If the model instance has been used for another fit with different fit parameters, then the fit options might not be the correct ones anymore .

```
[160]: cme.regr_plot(MEAN_COST_STATS.reset_index(), 'Trade Qty', 'Avg_Cost',\
'Average Cost as a function of Trade Amount : '+CURR)
```

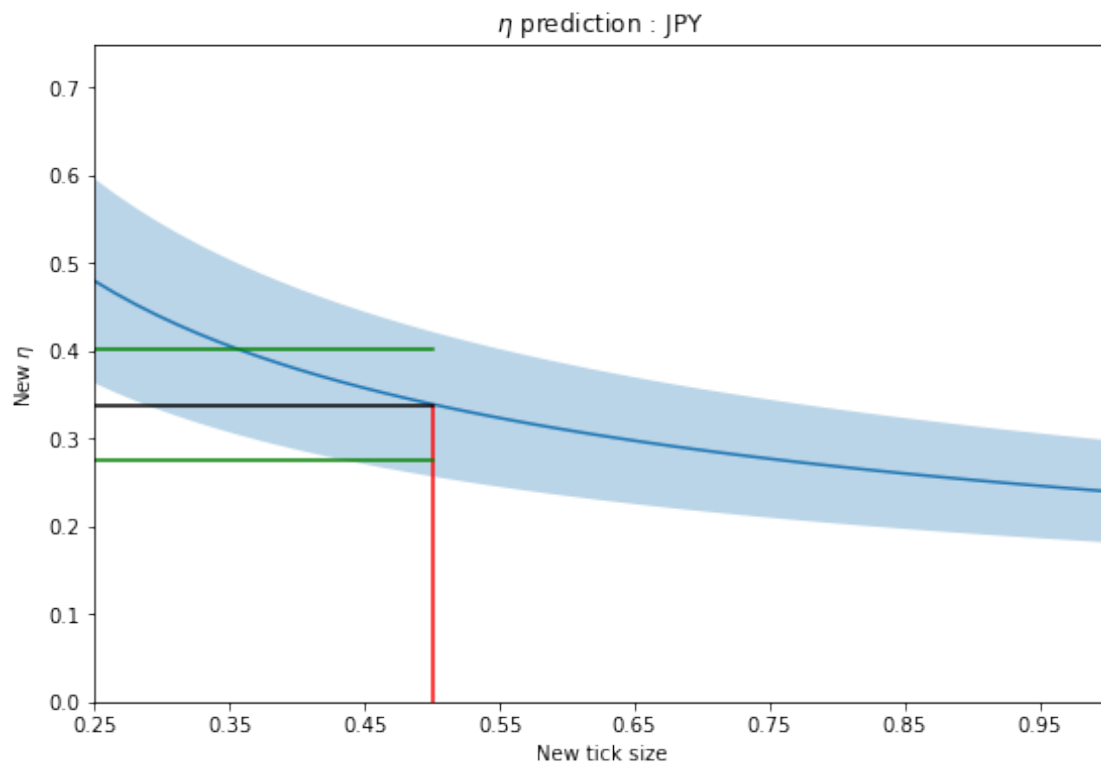


```
[161]: cme.regr_plot(MEAN_COST_STATS.reset_index(), 'Trade Qty', 'Avg_Cost',\
    'Average Cost as a function of Trade Amount : '+CURR, True)
```



2.9 Eta prediction

```
[162]: cme.plot_eta(TICK_PRIOR, TICK_AFTER,\
    TABLE_MATHIEU.loc['prior']['eta1'], TABLE_MATHIEU.loc['after']['eta1'],\
    TABLE_MATHIEU_ERR.loc['prior']['eta1'], TABLE_MATHIEU_ERR.\
    ↳loc['after']['eta1'],\
    CURR)
```



[]: