

L04s02

September 20, 2018

1 Calibration of BSM Volatility

[Ref] Chap 8 of [Hil15]

In simple terms, problem of calibration is to find parameters for the proposed model such that observed market quotes of liquidly traded plain vanilla options are replicated as closely as possible. The general approach is that, one defines an error function that is to be minimized.

In this below, we will use - Sep-30-2014 Market quote on call options underlying Euro Stoxx 50 - BSM model - RMSE as an error function to illustrate the calibration of volatility.

```
In [1]: import numpy as np
import pandas as pd
import scipy.optimize as sop
```

```
In [2]: # Market Data from www.eurexchange.com
# as of 30. September 2014
# European call & put option data (3 maturities)

options = pd.read_csv('SX5E_Options_140930.csv')
S0 = 3225.93 # EURO STOXX 50 level
r = 0.0005 # ECB base rate
options['Date'] = pd.to_datetime(options['Date'])
options['Maturity'] = pd.to_datetime(options['Maturity'])

options
```

```
Out[2]:
```

	Date	Strike	Call	Maturity	Put
0	2014-09-30	3175.0	126.8	2014-12-19	78.8
1	2014-09-30	3200.0	110.9	2014-12-19	87.9
2	2014-09-30	3225.0	96.1	2014-12-19	98.1
3	2014-09-30	3250.0	82.3	2014-12-19	109.3
4	2014-09-30	3275.0	69.6	2014-12-19	121.6
5	2014-09-30	3175.0	171.0	2015-03-20	129.2
6	2014-09-30	3200.0	156.1	2015-03-20	139.4
7	2014-09-30	3225.0	142.0	2015-03-20	150.3
8	2014-09-30	3250.0	128.5	2015-03-20	161.8
9	2014-09-30	3275.0	115.8	2015-03-20	174.0
10	2014-09-30	3175.0	82.3	2014-10-17	24.5

11	2014-09-30	3200.0	64.3	2014-10-17	31.5
12	2014-09-30	3225.0	48.3	2014-10-17	40.5
13	2014-09-30	3250.0	34.6	2014-10-17	51.8
14	2014-09-30	3275.0	23.5	2014-10-17	65.8

The above data is the market quote on Sep-30-2014 on call and put options underlying Euro Stoxx 50. The question of the calibration is, given that the market follows BSM, what is the volatility σ fitting to the above market quote?

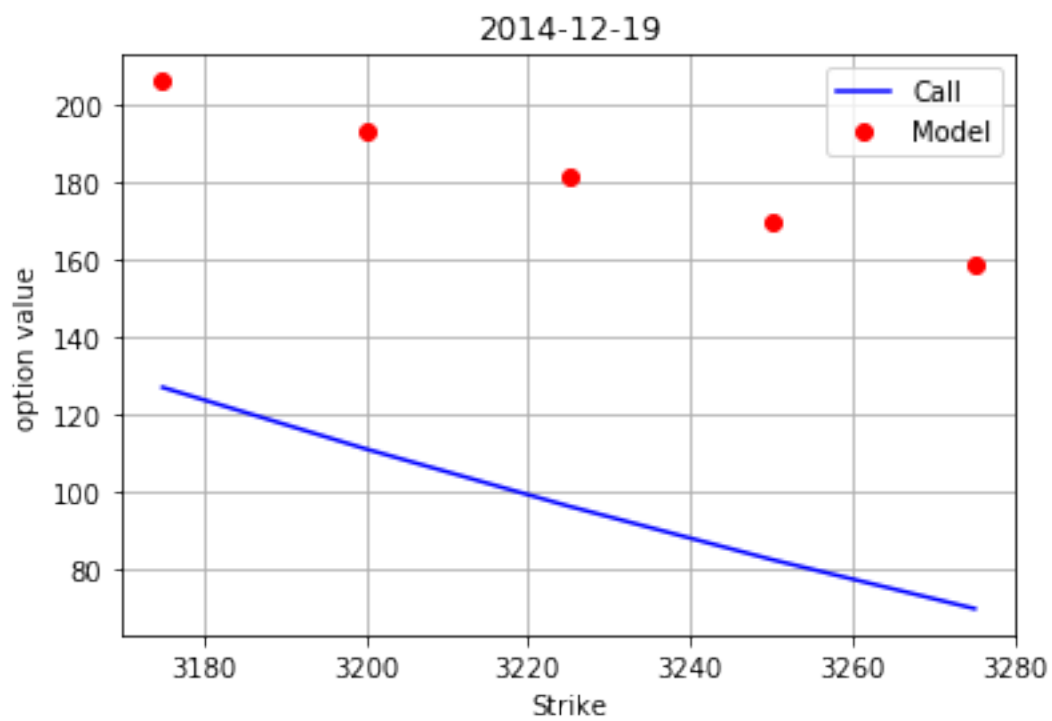
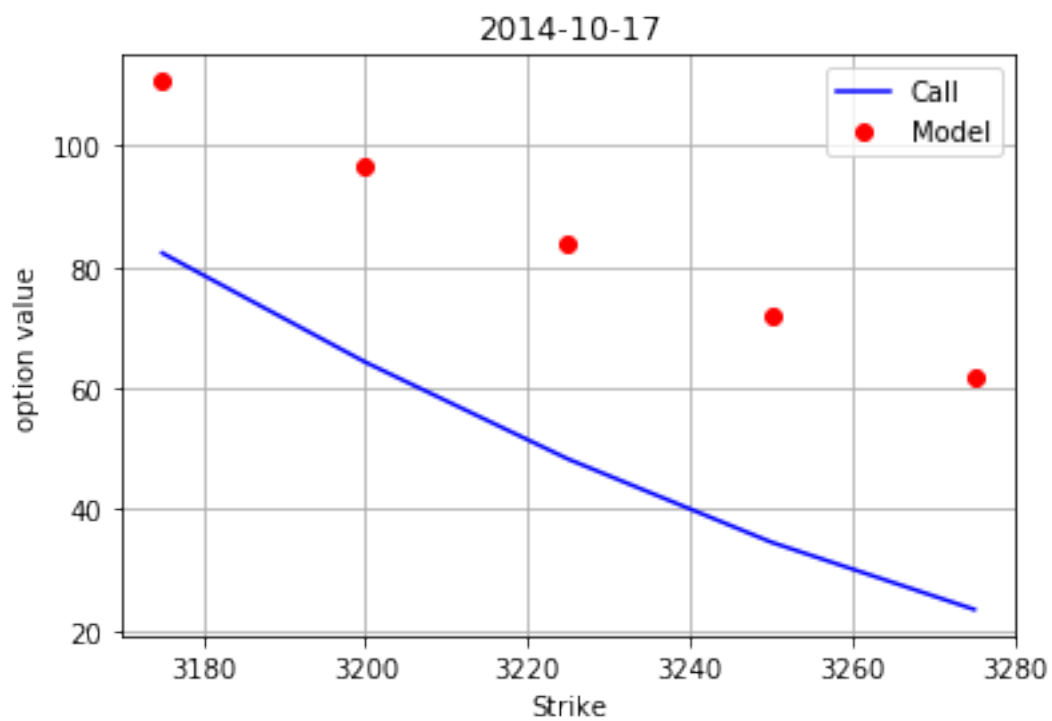
```
In [3]: from BSM_option_valuation import *
```

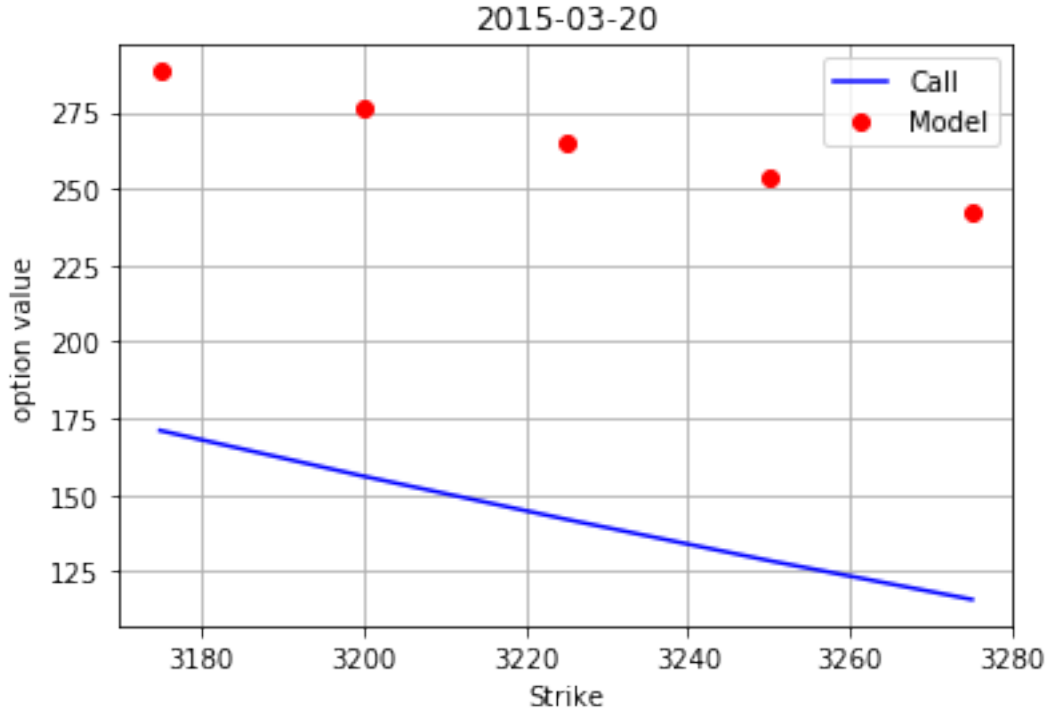
```
In [4]: def generate_plot(opt, options):
    #
    # Calculating Model Prices
    #
    sigma = opt
    options['Model'] = 0.0
    for row, option in options.iterrows():
        T = (option['Maturity'] - option['Date']).days / 365.
        options.loc[row, 'Model'] = BSM_call_value(S0, option['Strike'], 0, T, r, sigma)

    #
    # Plotting
    #
    mats = sorted(set(options['Maturity']))
    options = options.set_index('Strike')
    for i, mat in enumerate(mats):
        options[options['Maturity'] == mat][['Call', 'Model']].\
            plot(style=['b-', 'ro'], title='%s' % str(mat)[:10],
                grid=True)
        plt.ylabel('option value')
        plt.savefig('BSM_calibration_3_%s.pdf' % i)
```

The following graph shows that how the market data differs from the theoretical price by just making a guess of 30% volatility.

```
In [5]: generate_plot(.3, options); #Arbitrary input for the volatility
```





1.1 Calibration by minimizing RMSE

Suppose there are N options available in the market quote with various strike and maturities. We denote the market quotes of N options by

$$\{C_n^* : n = 1, 2, \dots, N.\}.$$

Correspondingly, there are N BSM theoretical prices available as a function of the volatility. We denote it by

$$\{C_n(\sigma) : n = 1, 2, \dots, N.\}$$

The error function RMSE is defined by

$$RMSE(\sigma) = \sqrt{\frac{1}{N} \sum_{n=1}^N |C_n^* - C_n(\sigma)|^2}.$$

Our goal is to find out the calibrated volatility $\hat{\sigma}$, which is given by

$$\hat{\sigma} = \arg \min_{\sigma} RMSE(\sigma).$$

```
In [6]: # Error Function
#
def BSM_error_function(p0):
    ''' Error Function for parameter calibration in BSM Model via
```

```

Parameters
=====
sigma: float
    volatility factor in diffusion term

Returns
=====
RMSE: float
    root mean squared error
'''

global i, min_RMSE
sigma = p0
if sigma < 0.0:
    return 500.0
se = []
for row, option in options.iterrows():
    T = (option['Maturity'] - option['Date']).days / 365.
    model_value = BSM_call_value(S0, option['Strike'], 0, T, r, sigma)
    se.append((model_value - option['Call']) ** 2)
RMSE = math.sqrt(sum(se) / len(se))
min_RMSE = min(min_RMSE, RMSE)
if i % 5 == 0:
    print('%4d |' % i, np.array(p0), '| %7.3f | %7.3f' % (RMSE, min_RMSE))
i += 1
return RMSE

```

```

In [7]: # Calibration initialization
#
i = 0 # counter initialization
min_RMSE = 200 # minimal RMSE initialization
opt = sop.fmin(BSM_error_function, 0.5,
               maxiter=500, maxfun=750,
               xtol=0.000001, ftol=0.000001)

```

```

0 | [0.5] | 214.284 | 200.000
5 | [0.35] | 119.761 | 119.761
10 | [0.2] | 25.162 | 7.525
15 | [0.175] | 9.808 | 3.796
20 | [0.1609375] | 3.594 | 3.594
25 | [0.1609375] | 3.594 | 3.584
30 | [0.16049805] | 3.584 | 3.584
35 | [0.16049805] | 3.584 | 3.584
40 | [0.16051178] | 3.584 | 3.584

```

Optimization terminated successfully.

Current function value: 3.584264

Iterations: 22

Function evaluations: 44

```
In [8]: print(opt)
```

```
[0.16051331]
```

```
In [9]: generate_plot(opt, options);
```

