

# Bank of American Stock Price Research

## Abstract:

From 2007 to 2014, the price of stock of Bank of American is relatively stable by checking the RRV (relative realized volatility). Only one day has relatively high RRVs during 2007 to 2014. I also did count on times of no arrival trade by different seconds. Then calculate the probability of no arrival trade with Poisson model. The result is that we can use Poisson model to predict probability of no arrival trade when seconds gap is relatively small. In addition, when I plot the daily 100 seconds accumulated RV(realized volatility) and daily average RV, I found strong linear relationship between these two variables. In the end, Using the Heston model to verify if there exist linear relationship between daily average and mean reversion rate. Then Comparing trend of alpha with weekly VIX from Yahoo Finance. When I use 5 days as a period to calculate the daily average RV and mean reversion rate. The significance of linear relationship is stronger. The overall trend of VIX from Yahoo Finance is similar to the shape of five day period alpha.

## Introduction:

First of all, based on the definition of RV(realized volatility), after we calculate all the 100 seconds RVs, we find all the RVs are scatter in a wide range. In order to normalize RVs and better analyze the jump effect in RV100s, we decide to use RRVs which is divide RVs by median RV. After calculating the RRVs, we divide it into different intervals. For example, in interval RRV greater than 1000, there may be one day. In interval RRV between 500 and 1000, there may be 50 days. Then we check how many days in each interval and check if the days are consecutive. The goal is to analyze the stability of stock price of Bank of American.

Secondly, since the original data have gaps between each trade, I try to use different seconds to calculate the number of gaps daily and sum them up to the whole year. Trying to predict the relationship between seconds unit and gaps, I choose Poisson distribution to calculate the number of gaps.

Thirdly, the daily average RV and daily accumulated RV is another direction to research the stability of stock price. We can check if there are outliers when we plot daily average RV and daily accumulated RV since they have strong correlation.

In the end, building Heston model can help us to discover more about daily average realized volatility. From the formula of Heston Model. I calculate the daily average RV for each day and also 5 days average RV. Preferring 5 days average RV is because 5 day have larger sample size therefore the variance may be smaller. The overall shape of significant alphas are similar to weekly VIX from Yahoo Finance.

## Body of the paper:

For the first question, we get there only 1 day that has RRV greater than 1000. The day is May 6<sup>th</sup>, 2010. From the output we can see that the interval 0 to 2.5 has 95.1% days and it means that most of days do not have large realized volatility. The interval 0 to 2.5 and 2.5 to 5 contains more than 99% of days. It indicates the stock prices of Bank of American are relatively stable and will not change too much in a day.

```
> ALLRRV=RRVreport(20070103,20141231,100)
NO.Observations Min      Max      Median      x1stQ      x3rdQ
1      371499      0 1755.393 0.9999737 0.6474082 1.416034
  Intervals Percentage
1      [0,2.5)      95.1%
2      [2.5,5)      4.12%
3      [5,10)      0.552%
4      [10,100)     0.176%
5      [100,1000)   0.0151%
6      >=1000      0.000808%
> |

> ddply(table7, .(RRVS), summarise, Pool.number = sum(Pool.number))
      RRVS Pool.number
1  RRV<500      1595
2 RRV>=1000         1
```

Secondly, the number of gaps is approximately fit Poisson distribution. The outputs from 2007 to 2014 indicates that if we choose small seconds to count gaps, the result will be much closer to the Poisson distribution probability. However, when we choose larger seconds to count gaps, the difference between empirical probability and Poisson model probability will be relatively large. This trend can be clearly seen in every year.

```
> pro2007
Seconds  average total.gaps empirical.probability poisson.model.probability
1      x5 4402.9402  4680.0000      0.9407992      0.9425175
2     x10 2079.7291  2340.0000      0.8887731      0.8947357
3     x15 1313.6892  1560.0000      0.8421085      0.8539424
4     x20  934.7171  1170.0000      0.7989035      0.8178335
5     x25  710.4741   936.0000      0.7590535      0.7858837
6     x30  562.9442   780.0000      0.7217234      0.7570874
7     x35  459.0598   668.5714      0.6866278      0.7309778
8     x40  383.2908   585.0000      0.6551980      0.7083606
> |

> pro2008
Seconds  average total.gaps empirical.probability poisson.model.probability
1      x5 3458.2806  4680.0000      0.7389489      0.7702415
2     x10 1342.6680  2340.0000      0.5737897      0.6529790
3     x15  723.2846  1560.0000      0.4636440      0.5848756
4     x20  447.2213  1170.0000      0.3822405      0.5391510
5     x25  300.0198   936.0000      0.3205339      0.5068876
6     x30  212.8142   780.0000      0.2728388      0.4832790
7     x35  156.1937   668.5714      0.2336230      0.4646936
8     x40  117.9763   585.0000      0.2016689      0.4500795
> |
```

```

> pro2009
Seconds    average total.gaps empirical.probability poisson.model.probability
1      X5 3329.3373  4680.0000          0.7113969          0.7493095
2     X10 1231.0437  2340.0000          0.5260870          0.6225614
3     X15  636.2063  1560.0000          0.4078246          0.5531227
4     X20  380.1190  1170.0000          0.3248881          0.5090994
5     X25  248.6984   936.0000          0.2657034          0.4798429
6     X30  173.0238   780.0000          0.2218254          0.4592435
7     X35  124.5238   668.5714          0.1862536          0.4431946
8     X40   92.2619   585.0000          0.1577127          0.4307242
> |

> pro2010
Seconds    average total.gaps empirical.probability poisson.model.probability
1      X5 4027.9643  4680.0000          0.8606761          0.8699462
2     X10 1715.1151  2340.0000          0.7329552          0.7656387
3     X15  981.1905  1560.0000          0.6289683          0.6900220
4     X20  638.4444  1170.0000          0.5456790          0.6348789
5     X25  446.4206   936.0000          0.4769451          0.5927071
6     X30  326.3810   780.0000          0.4184371          0.5590240
7     X35  245.9008   668.5714          0.3678003          0.5314216
8     X40  190.2500   585.0000          0.3252137          0.5092652
> |

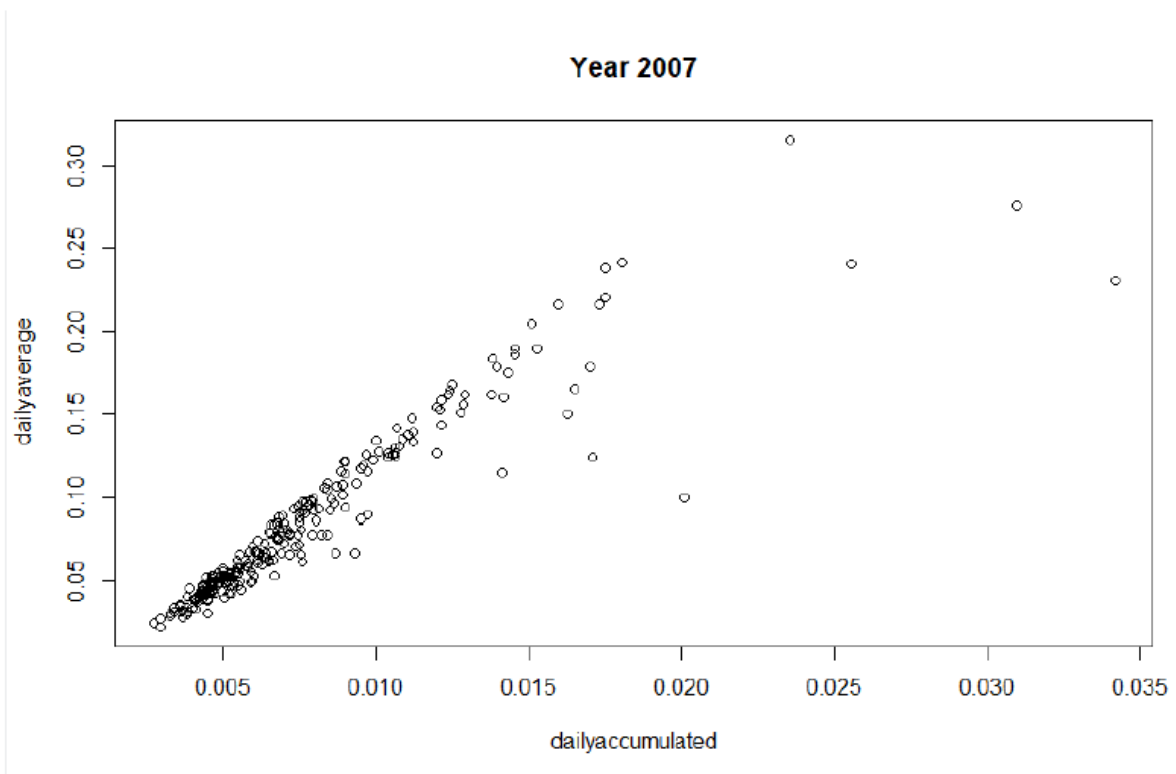
> pro2011
Seconds    average total.gaps empirical.probability poisson.model.probability
1      X5 4034.2421  4680.0000          0.8620175          0.8711140
2     X10 1723.4563  2340.0000          0.7365198          0.7683728
3     X15  992.9524  1560.0000          0.6365079          0.6952442
4     X20  650.7698  1170.0000          0.5562135          0.6416024
5     X25  457.5198   936.0000          0.4888032          0.5997774
6     X30  336.6786   780.0000          0.4316392          0.5664532
7     X35  255.8532   668.5714          0.3826864          0.5393915
8     X40  200.7302   585.0000          0.3431285          0.5184708
> |

> pro2012
Seconds    average total.gaps empirical.probability poisson.model.probability
1      X5 4163.2440  4680.0000          0.8895821          0.8954598
2     X10 1830.1340  2340.0000          0.7821085          0.8042127
3     X15 1079.0766  1560.0000          0.6917157          0.7347064
4     X20  717.4833  1170.0000          0.6132336          0.6792497
5     X25  512.3636   936.0000          0.5473970          0.6359706
6     X30  383.0144   780.0000          0.4910440          0.6011228
7     X35  294.1244   668.5714          0.4399297          0.5711689
8     X40  231.3206   585.0000          0.3954198          0.5463037
> |

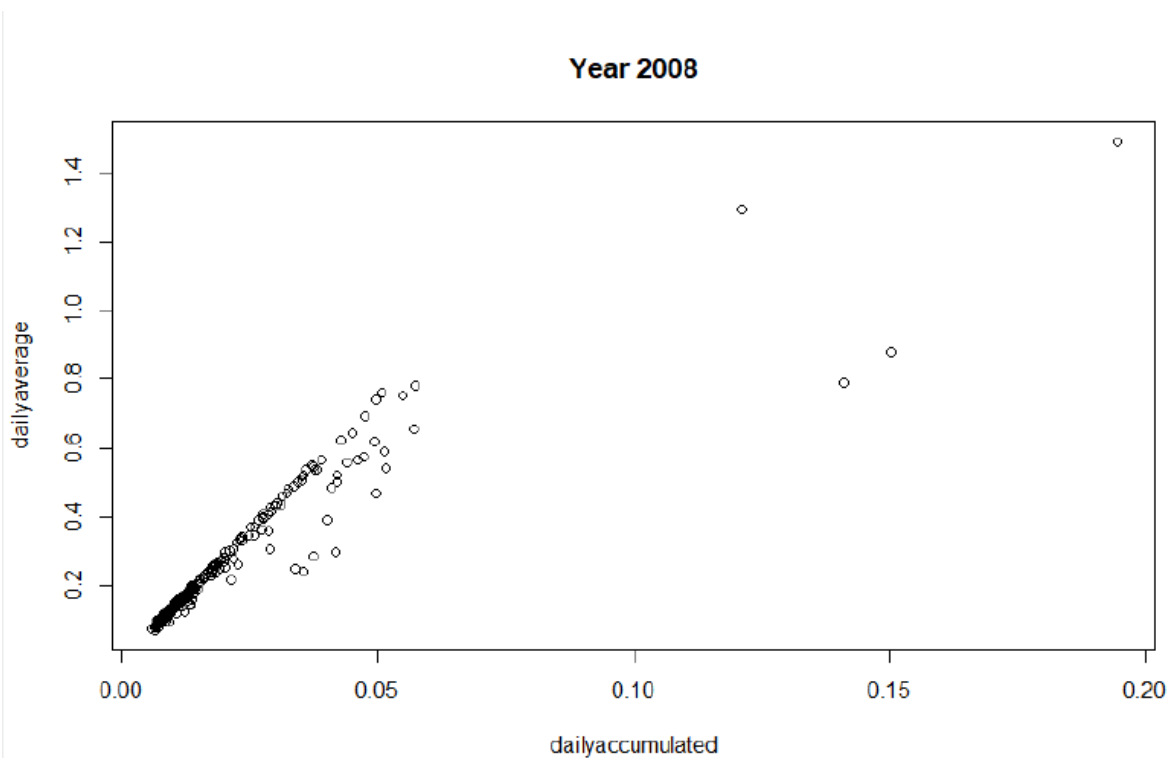
> pro2014
Seconds    average total.gaps empirical.probability poisson.model.probability
1      X5 4109.6535  4680.0000          0.8781311          0.8852644
2     X10 1783.0472  2340.0000          0.7619860          0.7881917
3     X15 1037.8740  1560.0000          0.6653039          0.7155555
4     X20  684.4646  1170.0000          0.5850125          0.6603485
5     X25  482.0079   936.0000          0.5149657          0.6156761
6     X30  354.9921   780.0000          0.4551181          0.5799103
7     X35  269.3228   668.5714          0.4028333          0.5503688
8     X40  209.8346   585.0000          0.3586917          0.5266030
> |

```

Thirdly, the daily accumulated RV and daily average RV are really close to each other and they have approximately linear relationship. Except just a few outliers, many points are approximately in linear relationship. The followings are the graphs for each year.

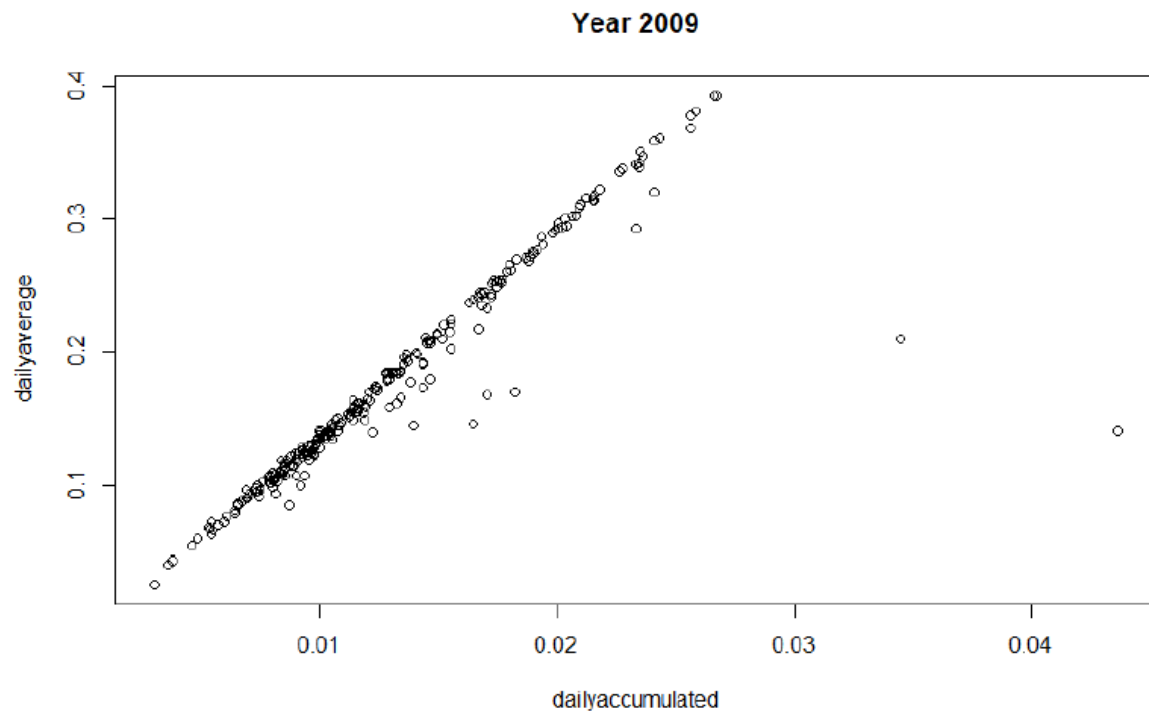


The plot of 2007 is closed to linear relationship, but it does have several obvious outliers. Those outliers are on 20070227, 20070228, 20070301 and 20070816. Overall the shape is approximately linear.

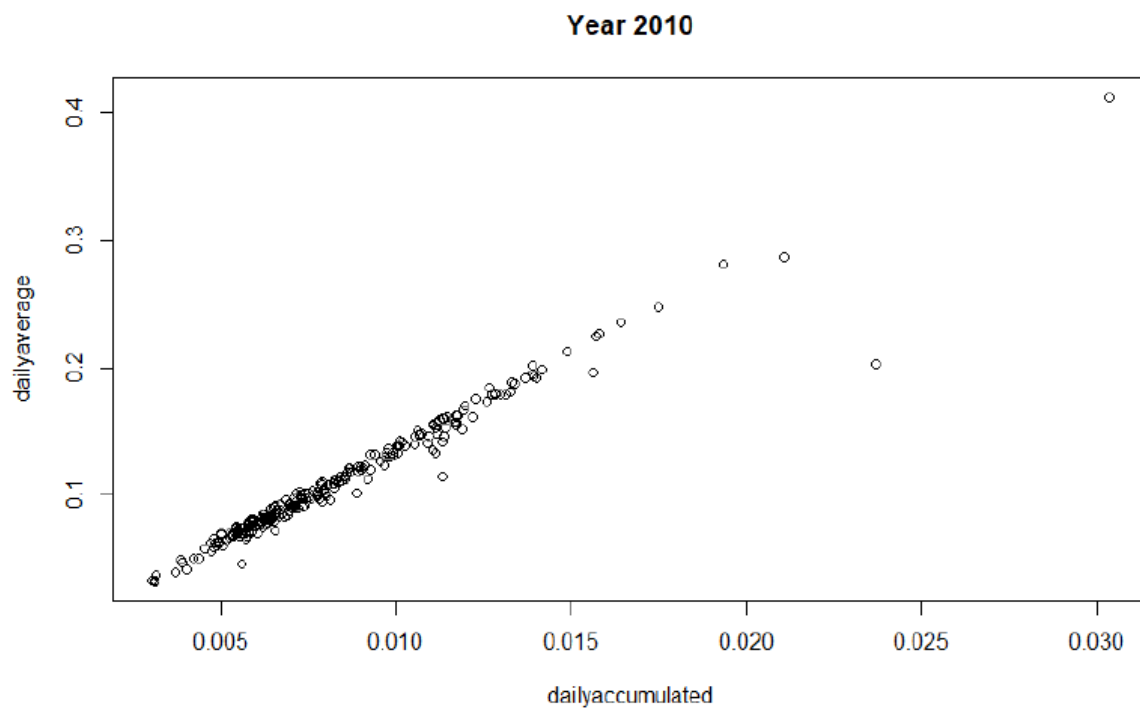


The plot of 2008 looks like that it has two separate parts. The first part which looks closed to

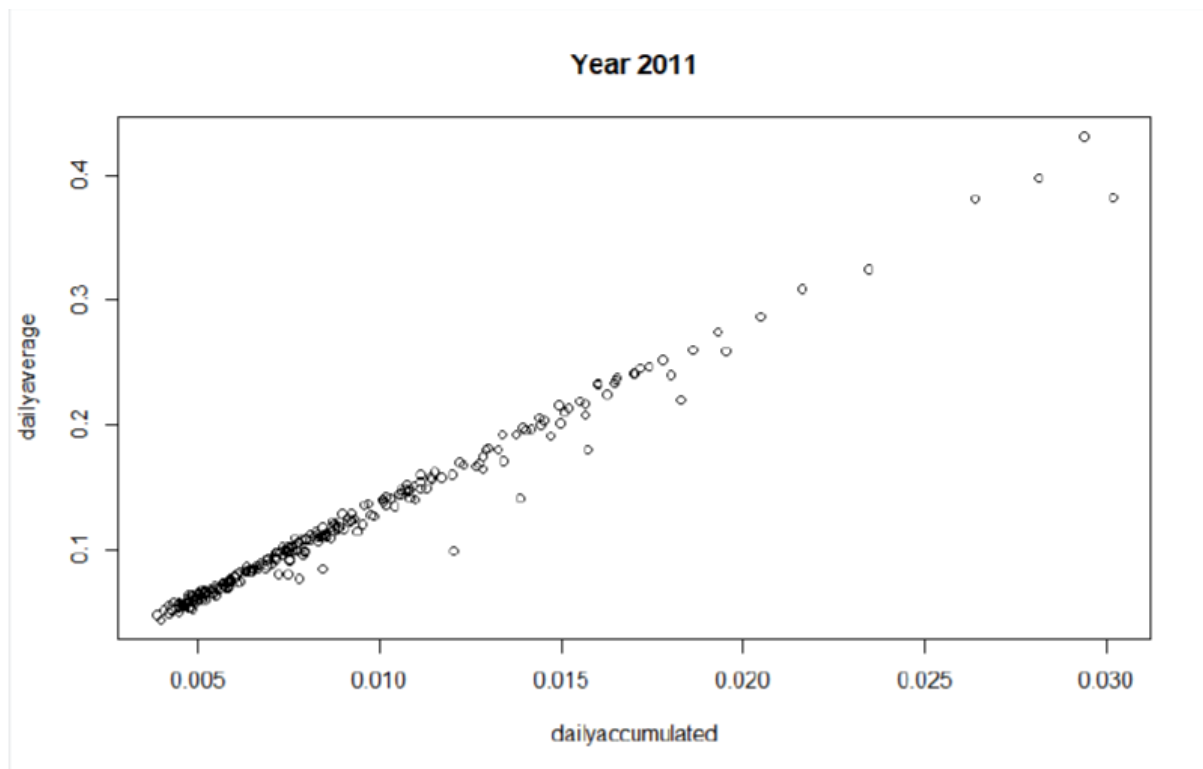
linear and the second part is 4 distinct outlier points. These outlier points are 20081010, 20081016, 20080919 and 20080122.



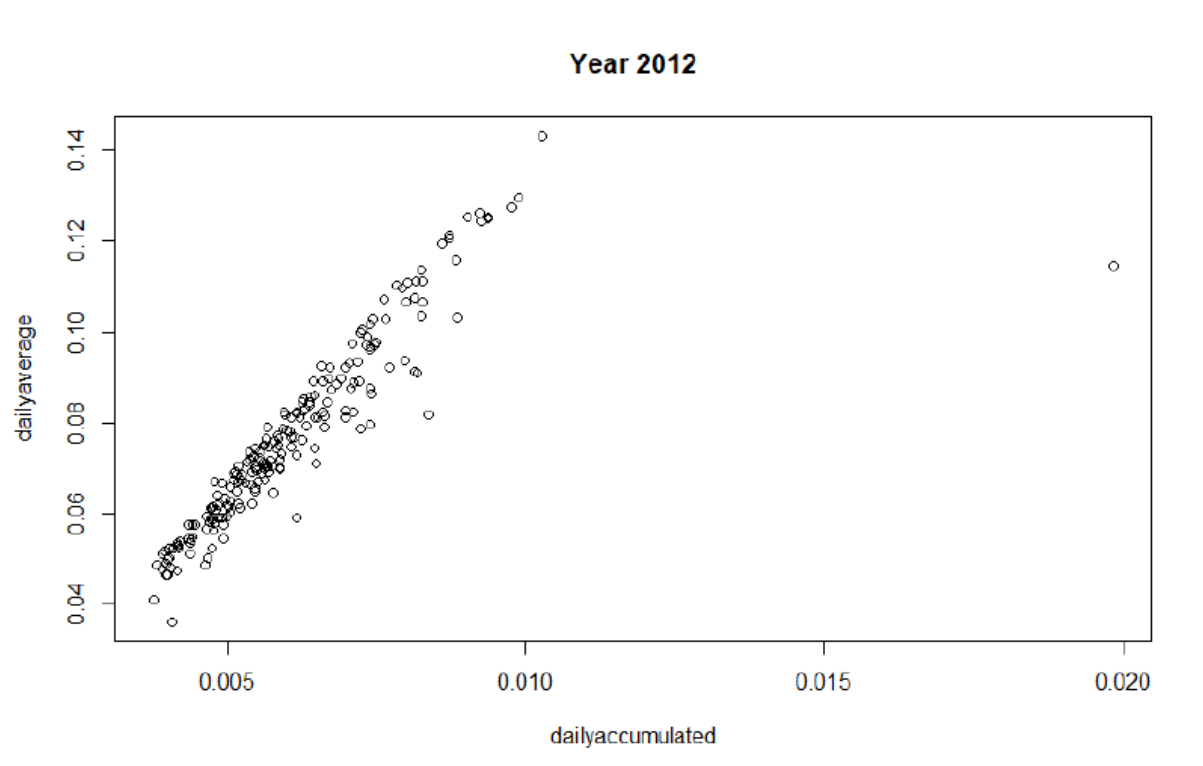
The plot of 2009 has major points remaining on linear relationship and two outlier points on the right side. These outlier points' dates are 20091207 and 20090916.



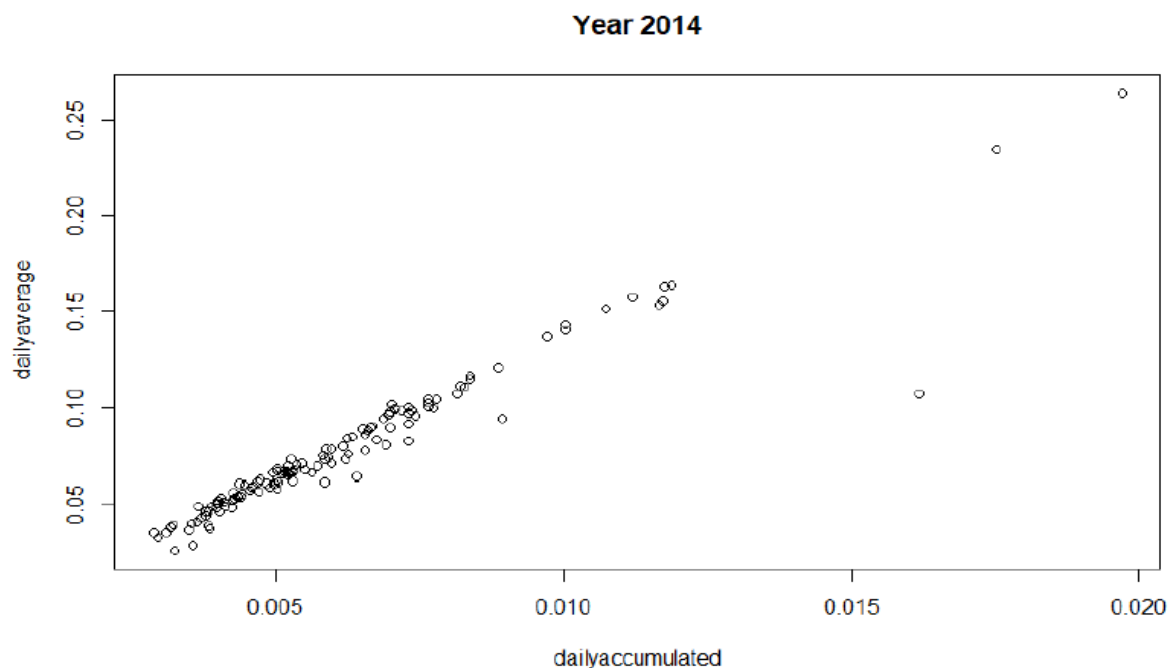
The plot of 2010 does not have obvious outlier point and is closed to linear relationship.



The plot of year 2011 is interesting since it is closed to linear relationship and it does not have obvious outliers.



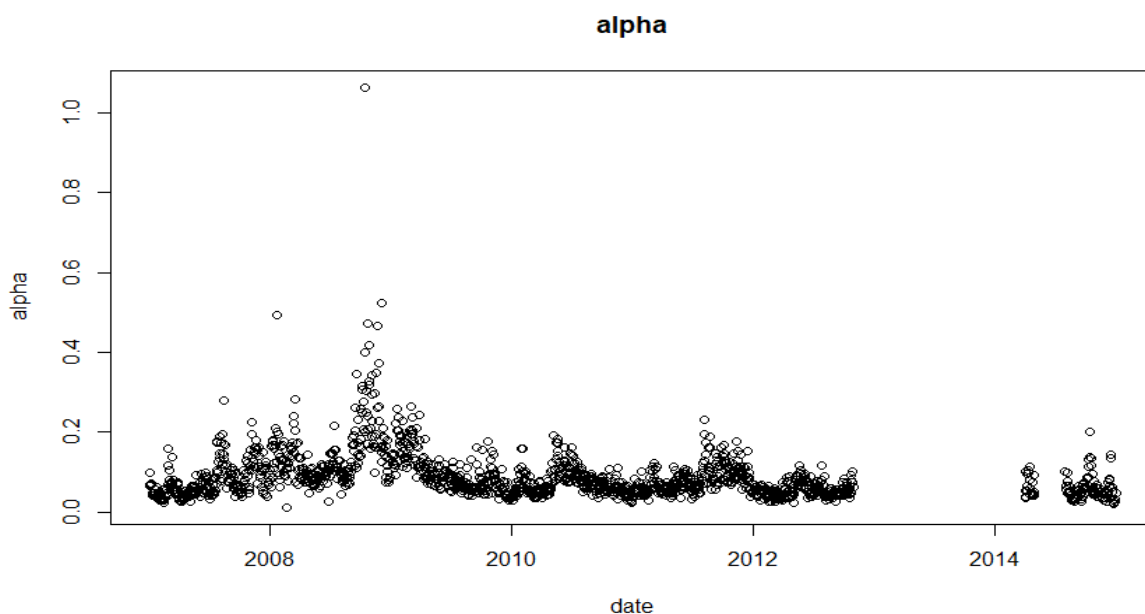
There is one obvious outlier point which is on the date 20120724



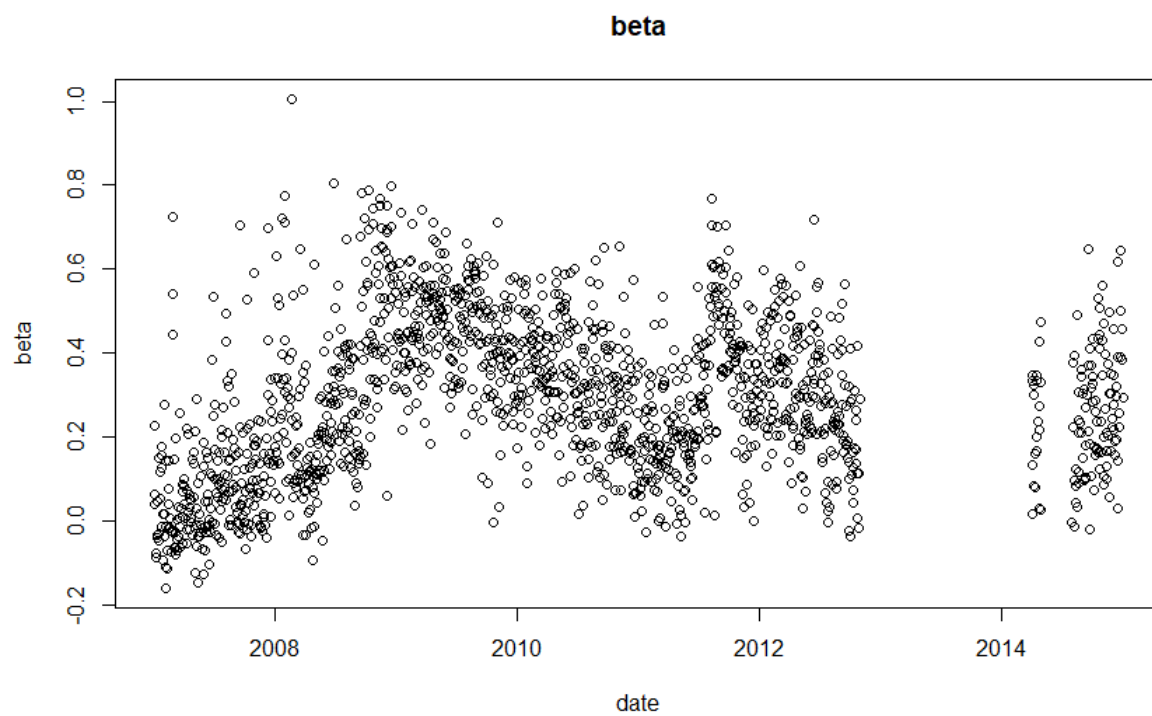
The plot of 2014 is almost on the one line except for one point on the data 20141027

In the end, after we built the Heston model and calculate the one day period and five day period for daily average and mean reversion rate. We find that since we use different number of days as period, the 5 day period obviously have smaller variance. The most obvious is the outlier point date 20100506. In one day period, the alpha of that day is more than 4 but in five day period, the alpha is less than 4. After computing the alpha, we will use 5-day period to compare with the VIX from yahoo finance from 2007 to 2014.

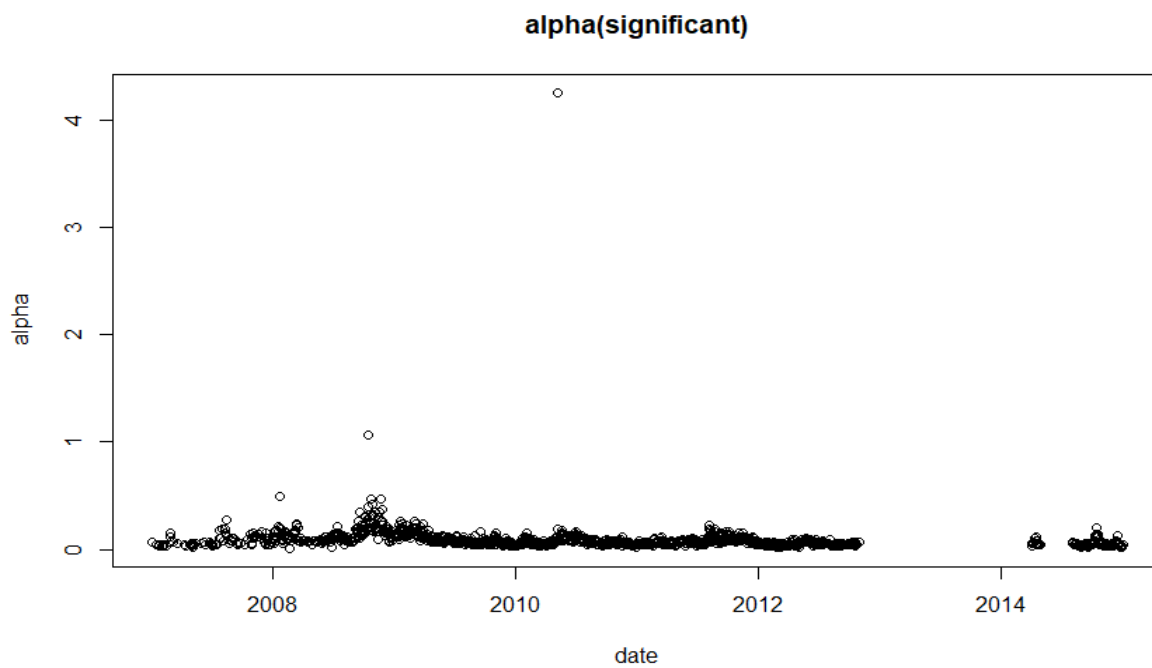
The following is one day period alpha and beta plots.



(alpha without outlier 20100506)

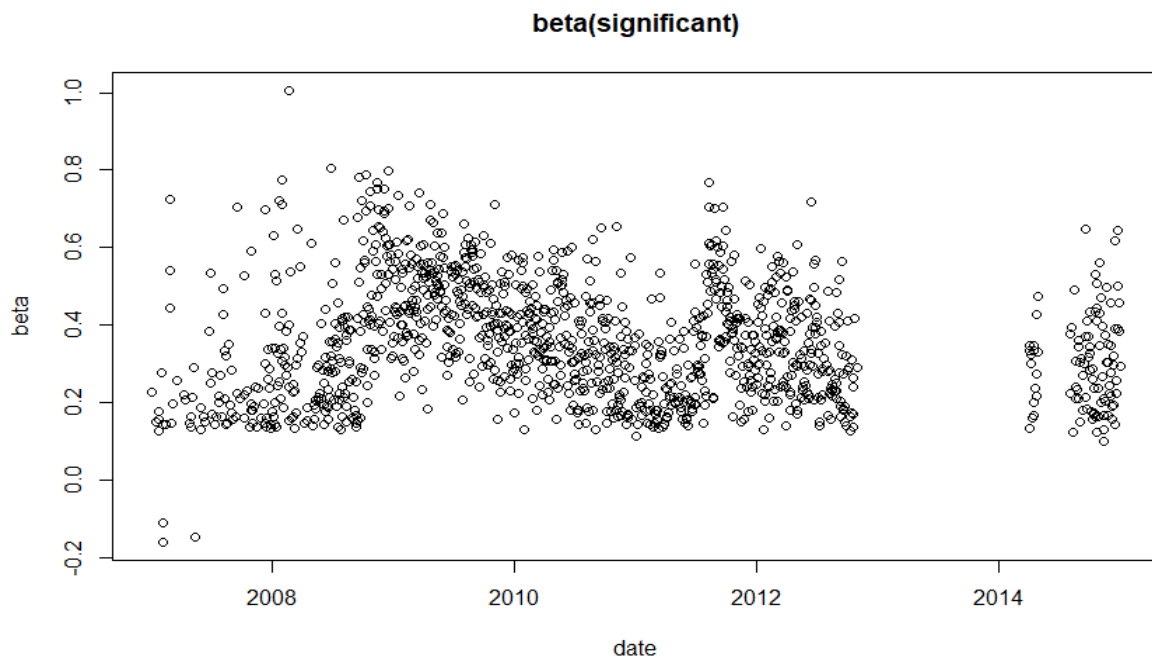


The beta plot does not have obvious or significant outliers. The highest beta is a little bit larger than 1.

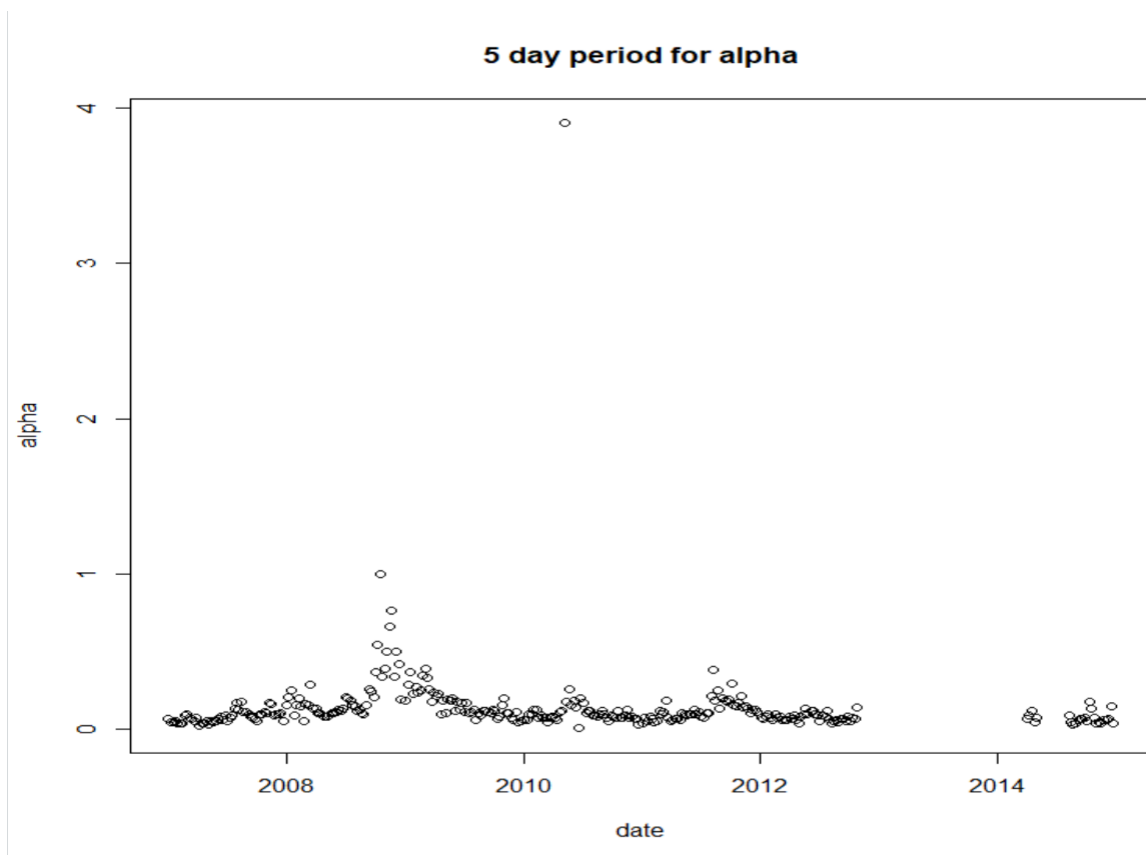


(The plot with outlier 20100506)

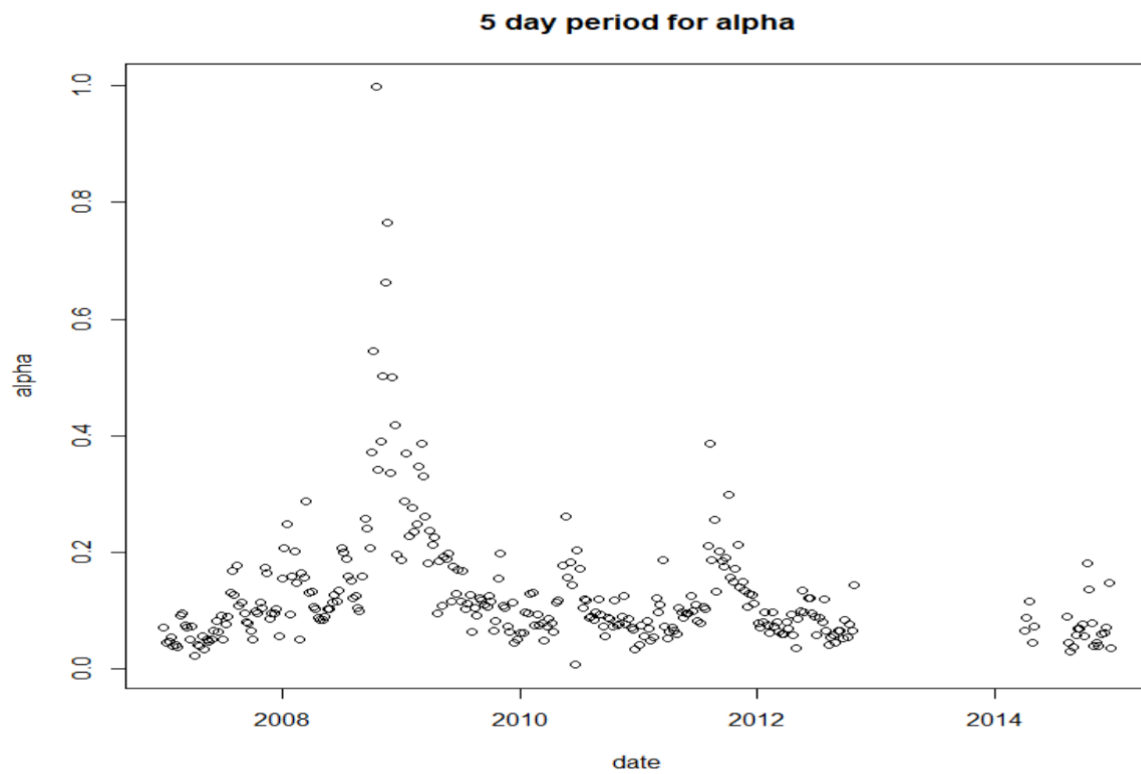




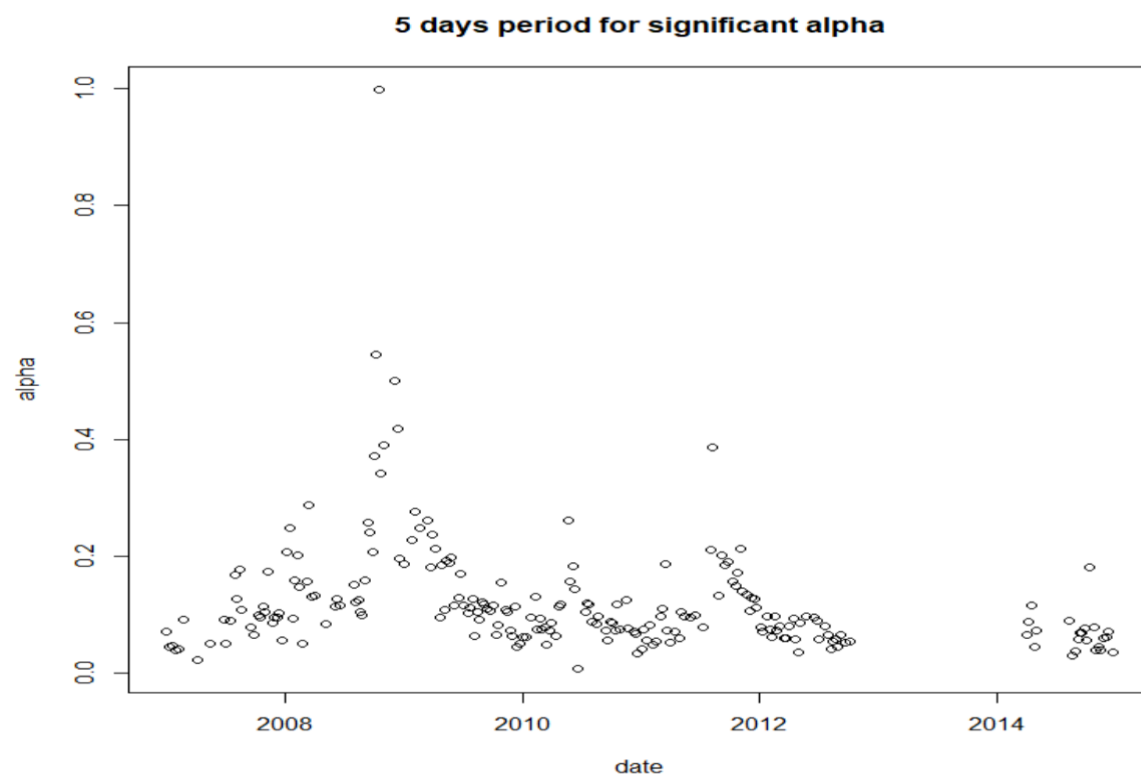
Beta graph with only significant days does not have significant or obvious outlier.  
The following is 5 day period:



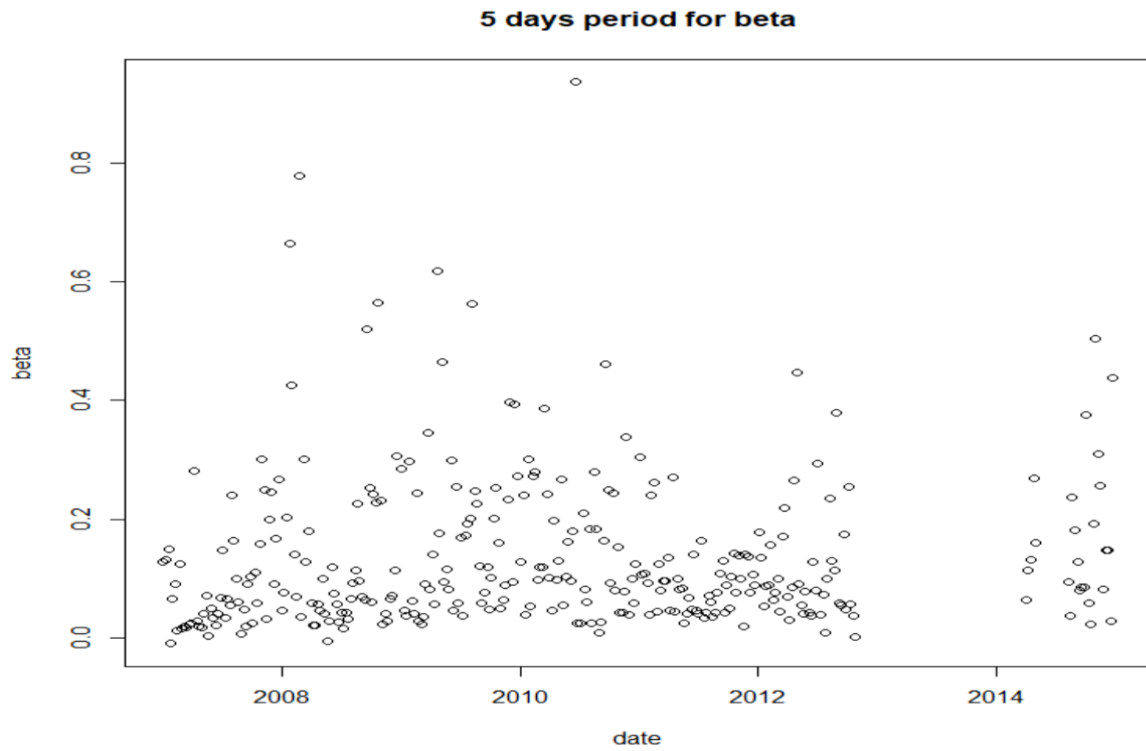
Here we use five day as a period, then it is clearly that the variance of the alpha becomes smaller. Previously in one day period, the outlier 20100506 is slightly greater than 4 but in five day period, it is below 4 now.



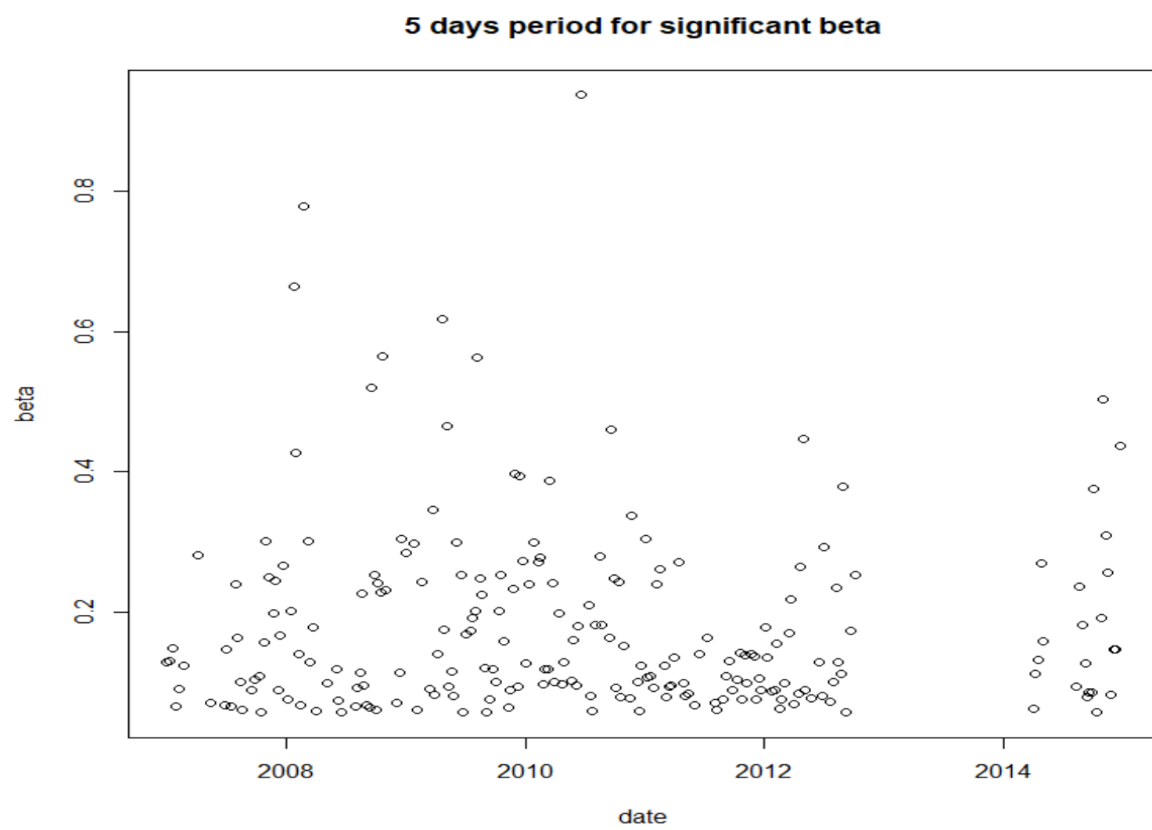
(Five day period without outlier)



The overall shape of five day period is similar to one day period but the variance is smaller due to larger sample size.

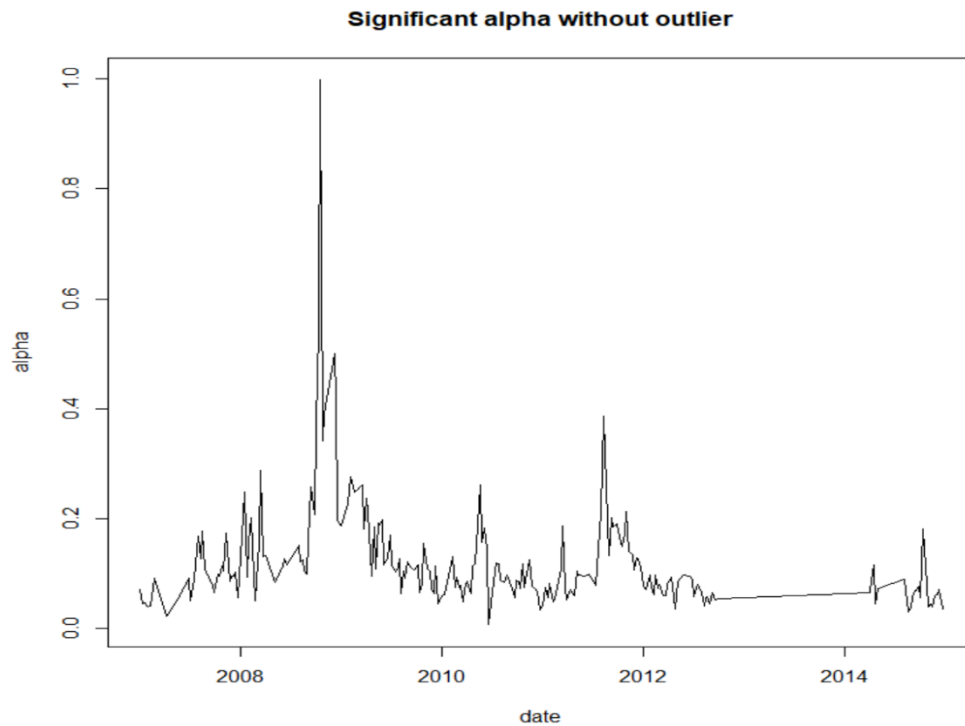


There is no significant outlier in five day period beta.

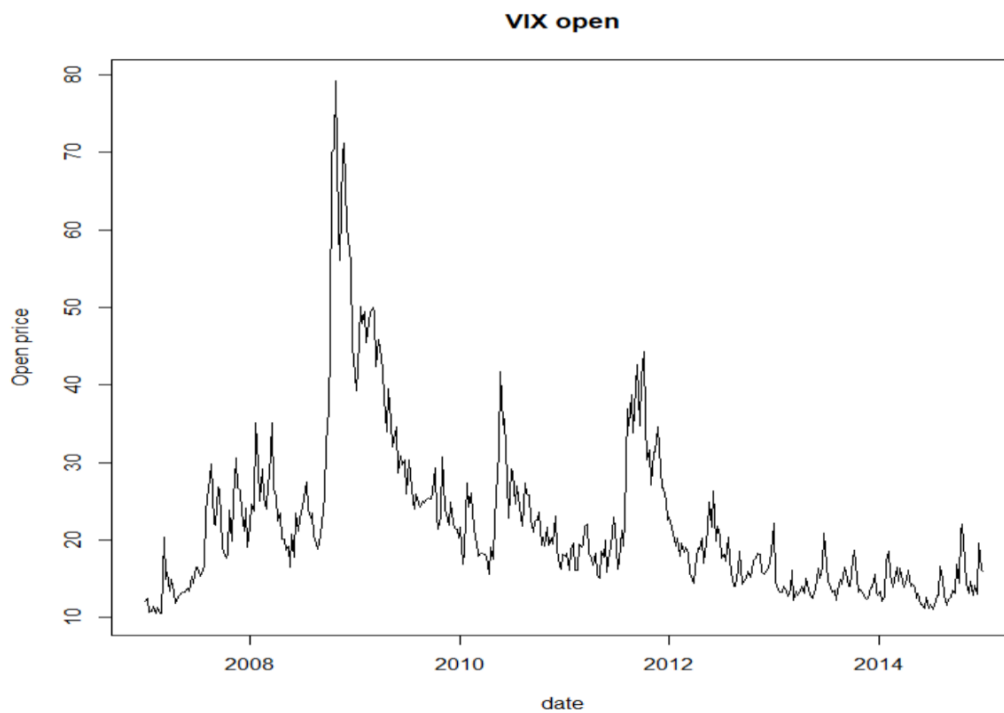


On the one hand, since here we have all alphas and betas, in order to compare with the

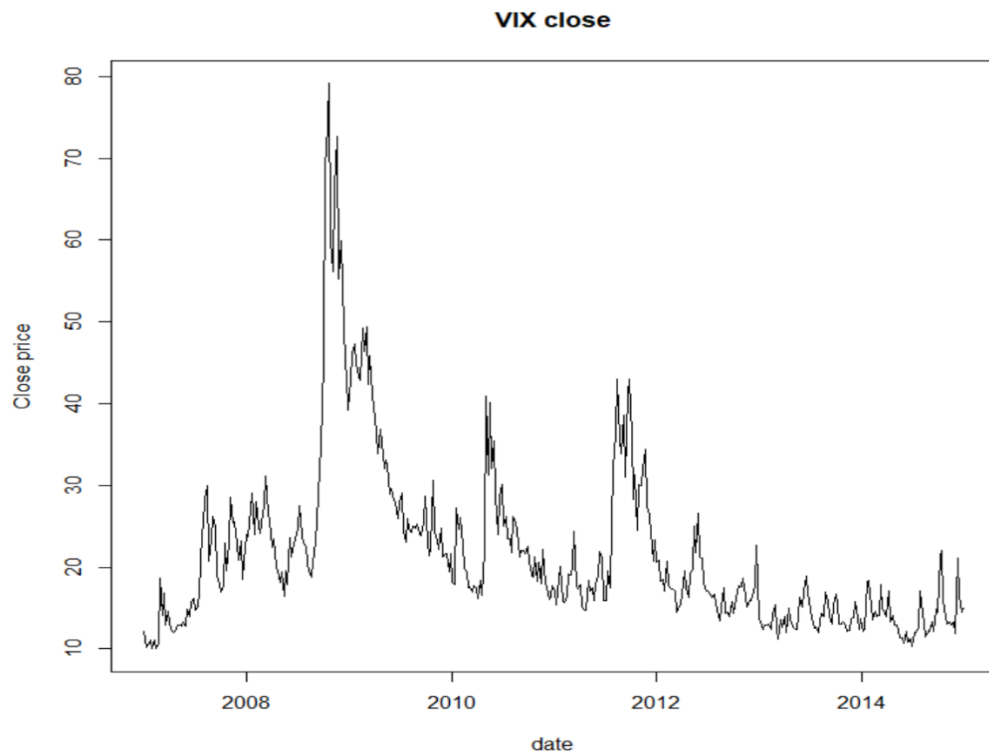
movement of the market volatility, I choose to download VIX from yahoo finance and plot with time from 2007 to 2014. On the other hand, I fill out all the insignificant alphas. Then I decide to compare with these two plots. The following are the plot about VIX and plot about significant alpha.



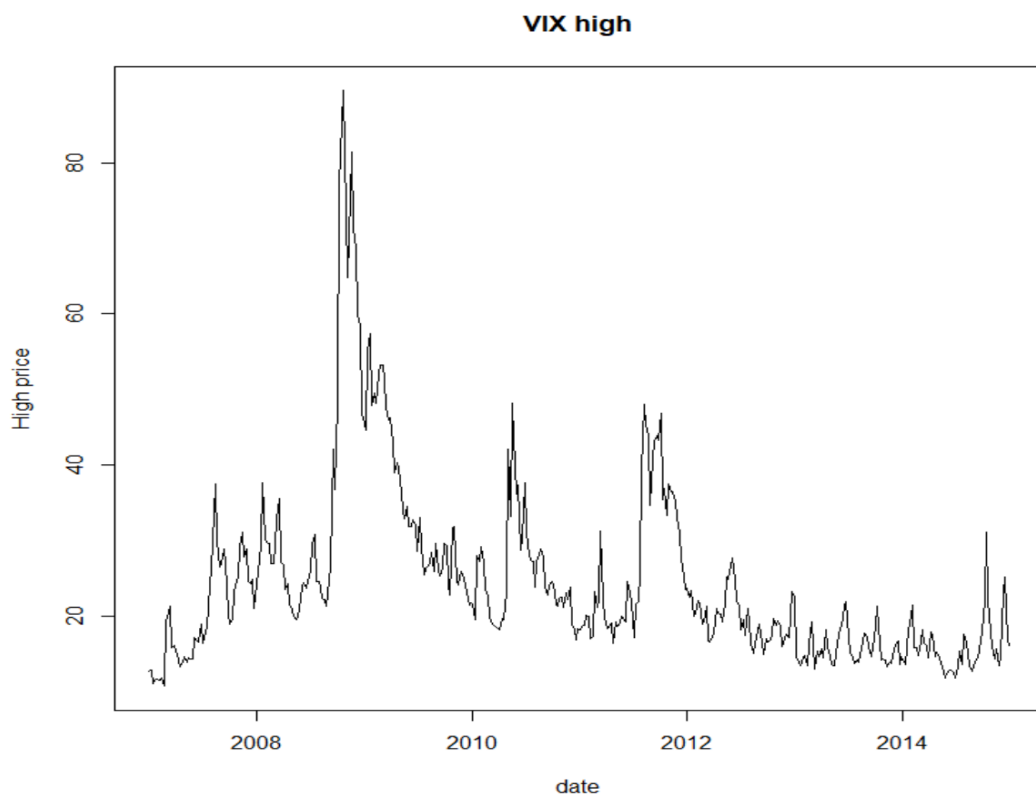
(note: the data is missing year 2013)



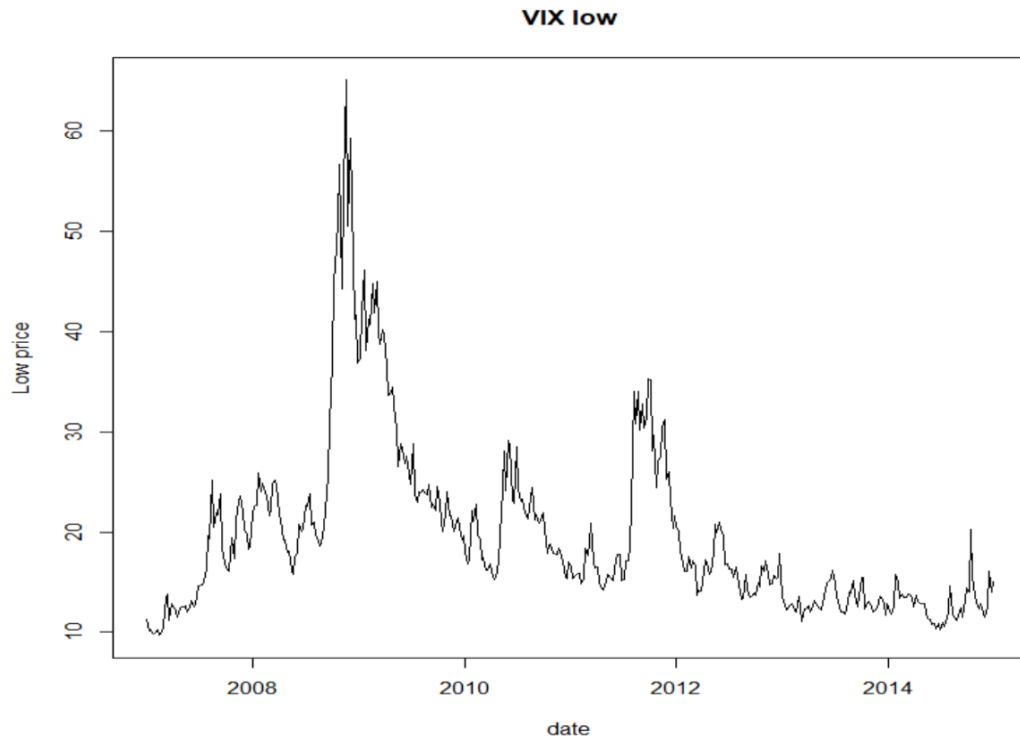
(plot with open price)



(plot with closed price)



(plot with high price)



(plot with low price)

The VIX values are calculated from the implied volatility of stock options. However, our RVs are from actual stock prices. Comparing the four plots from VIX, their shapes are approximately the same, but surprisingly, our significant alphas with 5 day period have close shape to the plots from VIX by ignoring the missing values in year 2013.

## Theory:

Log return: the formula of log return is following:

$$\text{Log return} = \log(t) - \log(t-1)$$

Realized Volatility: the realized volatility is calculated by taking the sum over the past squared return

$$RV_t = \sum_{i=1}^N r_t^2$$

The definition and use of RRV: in order to observed pronounced volatility jumps in high-frequency data instead of in frequent data. to better analyze the jump effects in RV100s, it is helpful to calculate the relative realized volatility to normalize RV100s. RRV has the formula as follows:

$$RRV = \frac{RV}{\text{Daily Median RV}}$$

From the mean-reversion feature of Heston model, we have the following:

$$RV_t - RV_{t-1} = \beta \cdot (RV_{t-1} - R_{bar}) + error$$

By removing terms, we got the following equation:

$$RV_t = \beta \cdot RV_{t-1} + \alpha + \epsilon$$

We have the alpha which is daily average and beta as mean-reversion rate. Since the equation is similar to linear equation formula, we decide to test if there is strong linear relationship between mean reversion rate and daily average.

### Additional section:

From the analysis of data of stock price, the most interesting values are RV in 2010/05/06. There was a significant flash crash happened on that day. It started at 14:32 pm and lasted about 36 minutes. Some research indicated that the most obvious reason is the debt crisis from Greece. The equity market began to fall rapidly and followed one 300-point drop and two 600-point drop. It was the most significant crash on year 2010. The following are the RV versus time, log return versus time and price versus time.

