

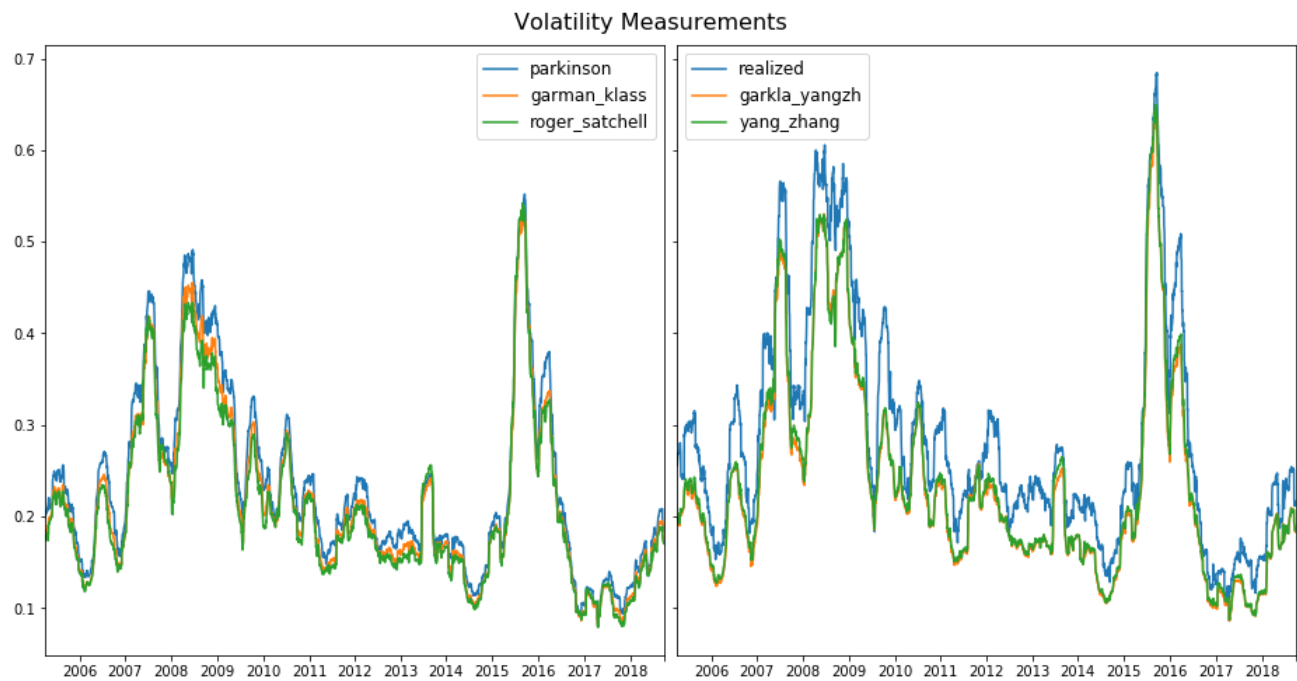
volatility

计算波动率的六种方法（仅供参考）：

Volatility	Price Information
Realized	Close
Parkinson	High, Low
Garman-Klass	Open, High, Low, Close
Roger-Satchell	Open, High, Low, Close
Garman-Klass-Yang-Zhang	Open, High, Low, Close
Yang-Zhang	Open, High, Low, Close

后五种方法均采用了连续收益率，导致波动率被低估

中证500指数的波动率的计算结果如下图所示



Parkinson, Garman-Klass, Roger-Satchell都是只用到当日的价格信息，可以看做是日内波动率；

Realized, Garman-Klass-Yang-Zhang, Yang-Zhang都用到了前一日和当日的价格信息，可以看做是日间波动率，从图上来看，前三者得到的波动率明显小于后三者得到波动率。

1. Realized Volatility: Close-Close

$$\sigma_{realized} = \sqrt{\frac{N}{n-2} \sum_{i=1}^{n-1} (r_t - \bar{r})^2}$$

$r_t = \log \frac{C_t}{C_{t-1}}$: 收益率

$\bar{r} = \frac{1}{n} \sum_{t=1}^{t-1} r_t$: 平均收益率

2. Parkinson Volatility: High-Low Volatility

$$\sigma_{parkinson} = \sqrt{\frac{1}{4 \cdot \ln 2} * \frac{252}{n} * \sum_{t=1}^n \ln \left(\frac{H_t}{L_t} \right)^2}$$

一般的波动率只考虑了收盘价，Parkinson Volatility 将最高价和最低价纳入了考虑范围，underestimate

3. Garman-Klass Volatility: OHLC volatility

Assumes Brown motion with zero drift and no opening jumps.

$$\sigma_{garman-klass} = \sqrt{\frac{N}{n} \sum_{i=1}^n \left[\frac{1}{2} * \left(\log \frac{H_i}{L_i} \right)^2 - (2 * \log 2 - 1) * \left(\log \frac{C_i}{O_i} \right)^2 \right]}$$

相比于Parkinson Volatility进一步考虑了开盘价和收盘价，纳入了更多的价格信息，underestimate

4. Roger-Satchell Volatility: OHLC Volatility

Assumes for non-zero drift, but assumed no opening jump.

$$\sigma_{roger-satchel} = \sqrt{\frac{N}{n} \sum_{i=1}^n \left[\log \frac{H_i}{L_i} * \log \frac{H_i}{O_i} + \log \frac{H L_i}{L_i} * \log \frac{L_i}{O_i} \right]}$$

underestimate

5. Garman-Klass-Yang-Zhang Volatility: OHLC Volatility

A modified version of Garman-Klass estimator that allows for opening jumps.

$$\sigma_{garkla-yangzh} = \sqrt{\frac{N}{n} \sum_{i=1}^n \left[\left(\log \frac{O_i}{C_{i-1}} \right)^2 + \frac{1}{2} * \left(\log \frac{H_i}{L_i} \right)^2 - (2 * \log 2 - 1) * \left(\log \frac{C_i}{O_i} \right)^2 \right]}$$

当资产收益率均不为零时，会高估波动率

6. Yang-Zhang Volatility: OHLC Volatility

$$\sigma_{yang-zhang} = \sqrt{\sigma_o^2 + k * \sigma_c^2 + (1 - k) * \sigma_{rs}^2}$$

$$\mu_o = \frac{1}{n} \sum_{i=1}^n \log \frac{O_i}{C_{i-1}}$$

$$\sigma_o^2 = \frac{N}{n-1} \sum_{i=1}^n \left(\log \frac{O_i}{C_{i-1}} - \mu_o \right)^2, \text{ Open-Close Volatility or Overnight Volatility}$$

$$\mu_c = \frac{1}{n} \sum_{i=1}^n \log \frac{C_i}{O_i}, \text{ Close-Open Volatility}$$

$$\sigma_c^2 = \frac{N}{n-1} \sum_{i=1}^n \left(\log \frac{C_i}{O_i} - \mu_c \right)^2$$

$$\sigma_{rs}^2 = \sigma_{roger-satchel}^2$$

$$k^* = \frac{\alpha}{1 + \alpha + \frac{n+1}{n-1}}, \alpha \text{通常为} 0.34$$

Has minimum estimator error, and is independent of drift and open gaps. It can be interpreted as a weighted average of the Roger-Satchell estimator, the Close-Open Volatility and the Open-Close Volatility.

References

1. [Volatility and its Measurements](#)
2. [Drift Independent Volatility Estimation Based on High, Low, Open and Close Price](#)
3. [volatility function | R Documentation](#)
4. [Parkinson volatility - Breaking Down Finance](#)
5. [MEASURING HISTORICAL VOLATILITY](#)