

# Measuring the Information Content of VIX Volatility

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Context: Humboldt Project

Supervisor: Prof. Franziska Peter

Author: Sophia Gläser (7. Semester BA CME)

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# Introduction

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## Motivation: Why this project? Why does Volatility matter?

- Risk measurement and the stability of the financial system
  - Volatility is closely related to risk
  - Volatility is crucial input to risk measures, such as the Value at Risk
- Volatility is used for..
  - ... the pricing of financial instruments, such as derivatives
  - ... the risk-return trade-off and therefore management decisions
- Forecast potential

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## More closely: What exactly is Volatility?

- In Finance, we are usually interested in the *conditional* standard deviation from the expected value of the underlying asset return (Tsay 2005)
- What causes asset price movement and thus volatility?
  - Assuming Market efficiency (as introduced by Malkiel and Fama), stock prices incorporate all available information from the market
  - But that does not give us information about the volatility of stock prices

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# The Problem: Why is it so hard to measure and forecast volatility?

- Volatility is not directly observable
  - We can estimate it for a given time period
  - The problem is, each period contains different information
- We can however observe stylized facts of volatility
  - Volatility clustering
  - Variation within a fixed range
  - Leverage effect: different reaction to price drop or increase

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## Maybe a solution: How volatility has been calculated so far

- According to this stylized facts, we can use models that best capture the characteristics of volatility
- There are multiple approaches, examples are
  - Econometric models using historic volatility (e.g. ARCH)
  - Black-and-Scholes implied volatility
- Many papers argue, that Black-and Scholes implied volatility contains significant information for realized volatility (e.g. Jiang and Tian (2005))

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## Maybe a solution: Black-and-Scholes implied volatility

- Intuition behind Black-and-Scholes (BS) implied volatility
  - Option pricing model, that uses volatility as an input
  - + Reverse calculation and derive an *option implied volatility*
  - + Options contain the market's expectation of future stock price movement
- Problems with the BS implied volatility
  - The BS model is mainly based on at-the-money options
  - Most importantly the BS model has a pricing assumption
- Joint hypothesis problem

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# Solving the joint hypothesis problem: Model-free implied volatility

- Model-free implied volatility is not based on any particular option pricing model
- One of the first model-free implied volatility indices was the VIX from CBOE (CBOE 2009)
  - Measures the market's expectation of 30-day volatility
  - + Includes information from both at-the-money and out-of-the-money options
  - + No pricing assumption → direct test of the informational efficiency of the options market

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## Solving the joint hypothesis problem: Model-free implied volatility

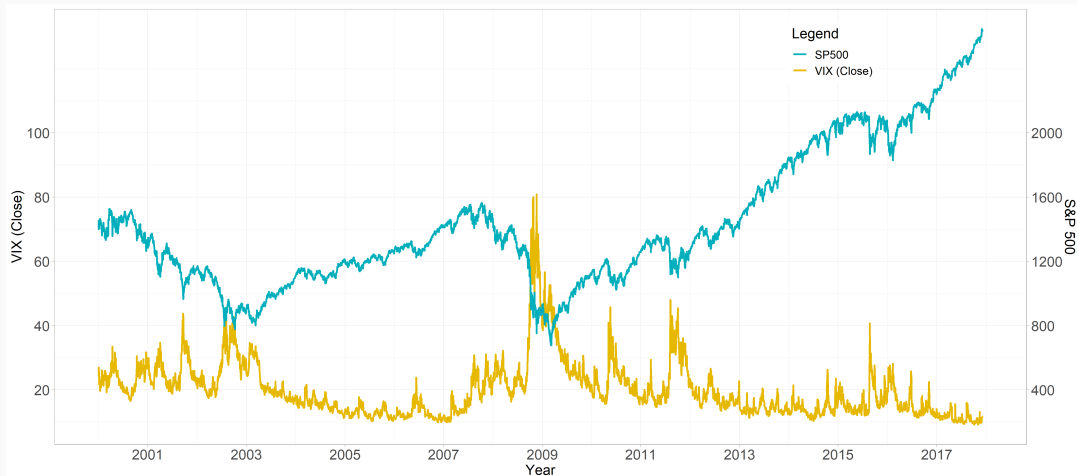
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# Data

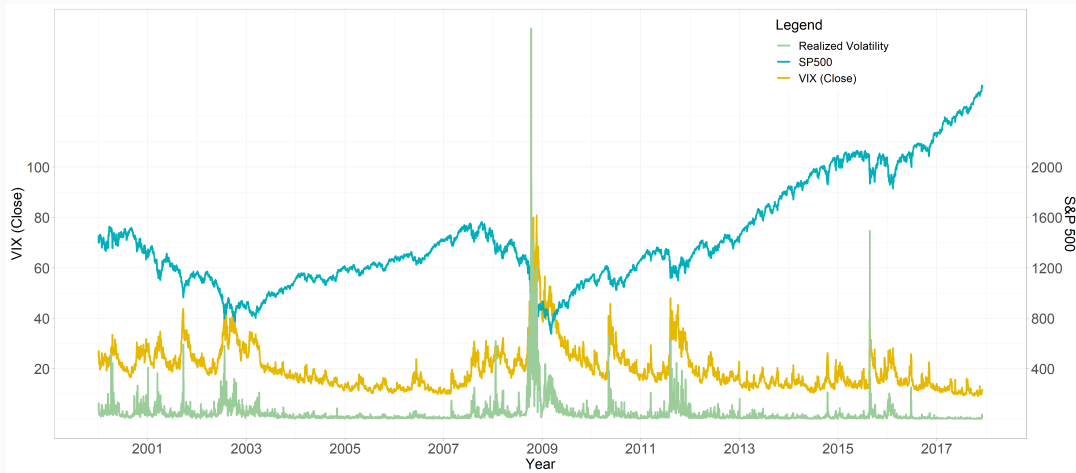
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- Components of Dataset
  - VIX, downloaded from CBOE
  - Realized Volatility, downloaded from Oxford Man Institute
  - S&P 500, downloaded from CBOE
- Sampling period January 2000 - December 2017

# S&P 500 in Comparison to VIX



# S&P 500 in Comparison to VIX and Realized Volatility





# Method

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## My Method: Regression and HAR Model

Stepwise regression to compare information content:

$$\sigma_{t+1} = c + \beta^d RV_{t-1}^d + \beta^w RV_{t-1}^w + \beta^m RV_t^m \quad (1)$$

$$\sigma_{t+1} = c + \beta^d RV_{t-1}^d + \beta^w RV_{t-1}^w + \beta^m RV_t^m + \beta^{VIX} VIX_t^{VIX} \quad (2)$$

+ the same regressions as log-log model

with the weekly aggregation period being (monthly similar with 20 days):

$$\sigma_t^w = \frac{1}{5} (RV_t^d + RV_{t-1d}^d + RV_{t-2d}^d + RV_{t-3d}^d + RV_{t-4d}^d)$$

and Realized Variance (RV) being:  $RV_t^d = \sqrt{\sum_{m=1}^M r_{t,m}^2}$

$\sigma$  = Volatility (sd),  $RV$  = Realized Variance with  $M$  equally spaced intraday returns for a day  $t$ , with  $m = 1, \dots, M$ , here 10min returns

HAR-RV Model (according to Corsi (2009), but not implemented yet), idea:

$$\sigma_{t+1} = c + \beta^d RV_{t-1}^d + \beta^w RV_{t-1}^w + \beta^m RV_t^m + \tilde{w}_{t+1d}^d \quad (3)$$

$\tilde{w}_{t+1d}^d$  = innovation term

## Results so far

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# Regression Results

**Figure 1:** level-level Model

	Historic	Historic with VIX
Intercept	0.00*** (0.00)	-0.01*** (0.00)
$RV_t^d$	0.27*** (0.02)	0.23*** (0.02)
$RV_t^w$	0.39*** (0.03)	0.36*** (0.03)
$RV_t^m$	0.25*** (0.03)	-0.04 (0.03)
VIX		0.00*** (0.00)
$R^2$	0.54	0.57
Adj. $R^2$	0.54	0.57
Num. obs.	4436	4436
RMSE	0.02	0.02

\*\*\*  $p < 0.001$ , \*\*  $p < 0.01$ , \*  $p < 0.05$

**Figure 2:** log-log Model

	Historic	Historic with VIX
Intercept	-0.25*** (0.05)	-7.85*** (0.28)
$RV_t^d$	0.34*** (0.02)	0.24*** (0.02)
$RV_t^w$	0.40*** (0.03)	0.28*** (0.03)
$RV_t^m$	0.21*** (0.02)	-0.13*** (0.03)
VIX		1.60*** (0.06)
$R^2$	0.73	0.77
Adj. $R^2$	0.73	0.77
Num. obs.	4437	4437
RMSE	0.60	0.55

\*\*\*  $p < 0.001$ , \*\*  $p < 0.01$ , \*  $p < 0.05$

## Possible Problems coming up






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## Next steps/Questions to solve

- Having gathered all this information about volatility measurement, what is the most accurate way to set up my regressions and my HAR-RV model?
- Next step: Bring current state of work to paper

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# Appendix

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# The Black and Scholes Equation

price of an option over time:

$$\frac{\partial C}{\partial t} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 C}{\partial C^2} + rS \frac{\partial C}{\partial S} = rC \quad (4)$$

calculate the price of European call and put option:

$$C(S, t) = N(d_1)S - N(d_2)Ke^{-rt} \quad (5)$$

$$d_1 = \frac{1}{\sigma\sqrt{t}} \left[ \ln \left( \frac{S}{K} \right) + t \left( r + \frac{\sigma^2}{2} \right) \right] \quad (6)$$

$$d_2 = \frac{1}{\sigma\sqrt{t}} \left[ \ln \left( \frac{S}{K} \right) + t \left( r - \frac{\sigma^2}{2} \right) \right] \quad (7)$$

$$N(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{1}{2}z^2} dz \quad (8)$$

C = Call option price

S = Current stock price

K = Strike price of the option

r = risk-free interest rate (a number between 0 and 1)

$\sigma$  = volatility of the stocks return (a number between 0 and 1)

t = time to option maturity (in years)

N = normal cumulative distribution function

$$\sigma^2 = \frac{2}{T} \sum_i \frac{\Delta K_i}{K_i^2} e^{RT} Q(K_i) - \frac{1}{T} \left( \frac{F}{K_0} - 1 \right)^2 \quad (9)$$