

# A Practical Look at Volatility in Financial Time Series

MATH 287C - Advanced Time Series Analysis  
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# Outline

1. What is Volatility?
2. Normalizing and Variance Stabilizing (NoVaS) Transformation
3. Forecasting Volatility
4. A Simple Volatility Trading Strategy

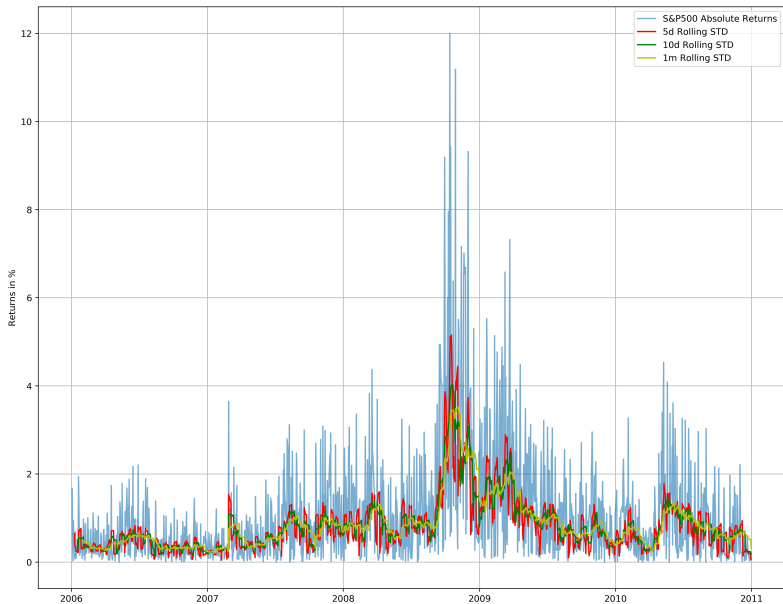
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# What is Volatility?

- ▶ Volatility is a measure of price variability over some period of time
- ▶ Typically described by the standard deviation  $\sigma$  of the return series  $\{X_t\}$
- ▶ Volatility is peculiar in that we know it exists, but in some sense we can't really measure it
- ▶ Bachelier (1900) showed that  $\{X_t\} \sim \text{iid. } N(0, 1)$ , but this is only good for a first order approximation

# Naive Measure - Realized Volatility



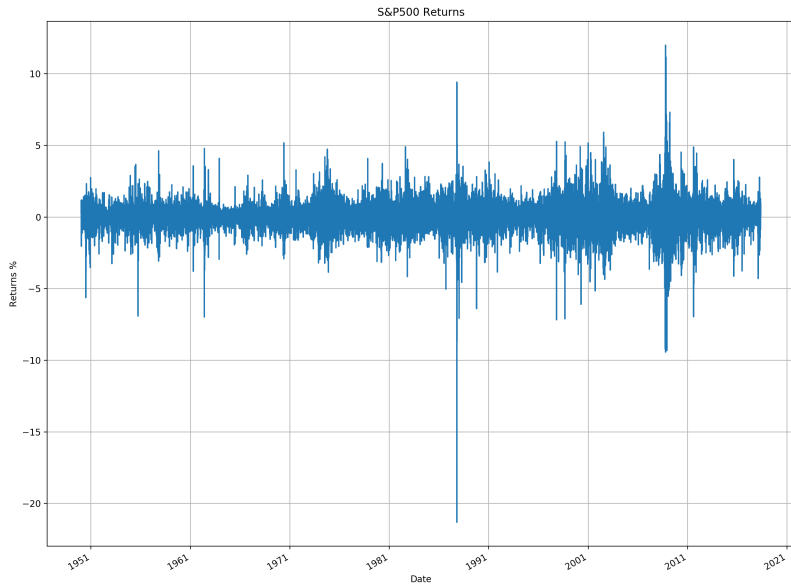
# Stylized Facts

Further analysis of  $\{X_t\}$  reveals other kinds of structure that cannot be explained by the gaussian assumption.

In particular, the return series displays the following distinctive behavior:

1.  $\{X_t\}$  is heavy-tailed, much more so than the Gaussian white noise
2. Although  $\{X_t\}$  is uncorrelated, the series  $\{X_t^2\}$  is highly correlated
3. The changes in  $\{X_t\}$  tend to be clustered, large changes tend to be followed by large changes and vice v
4. Effects are asymmetric, bad news results in larger downward price moves than positive news does to upward price moves

# SP500 Daily Returns (1950-2018)



# GARCH

The Generalized ARCH (GARCH) model of Bollerslev (1986) and its variants are extremely popular (albeit imperfect) methods to model volatility.

GARCH(p,q) model can be expressed as:

$$X_t = \sigma_t \varepsilon_t, \quad \varepsilon_t \sim N(0, 1)$$

where

$$\sigma_t^2 = \alpha_0 + \sum_{i=1}^p \beta_i \sigma_{t-i}^2 + \sum_{j=1}^q \alpha_j X_{t-j}^2$$

For the purposes of this talk, we'll focus on GARCH(1,1) models where  $\sigma_t^2 = \alpha_0 + \beta_1 \sigma_{t-1}^2 + \alpha_1 X_{t-1}^2$



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# NoVaS Transformation (Politis 2007)

The NoVaS Transformation is defined as

$$W_{t,a} = \frac{X_t}{\sqrt{\alpha s_{t-1}^2 + a_0 X_t^2 + \sum_{i=1}^p a_i X_{t-i}^2}}$$

for  $t = p + 1, p + 2, \dots, n$

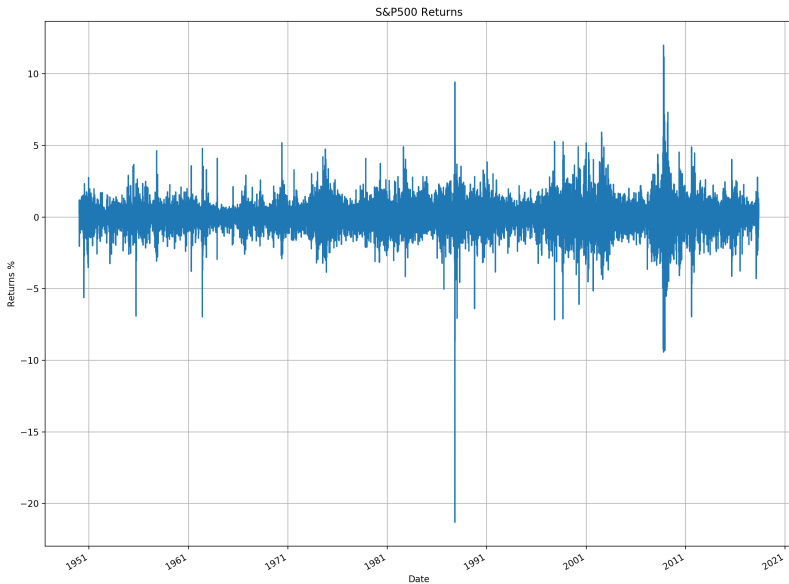
It is a clever extension of the ARCH model where we include the value  $X_t$  in order to “studentize” the returns.

The order  $p$  and the vector of nonnegative parameters  $(\alpha, a_0, \dots, a_p)$  are chosen by the practitioner with the twin goals of normalization and variance-stabilization.

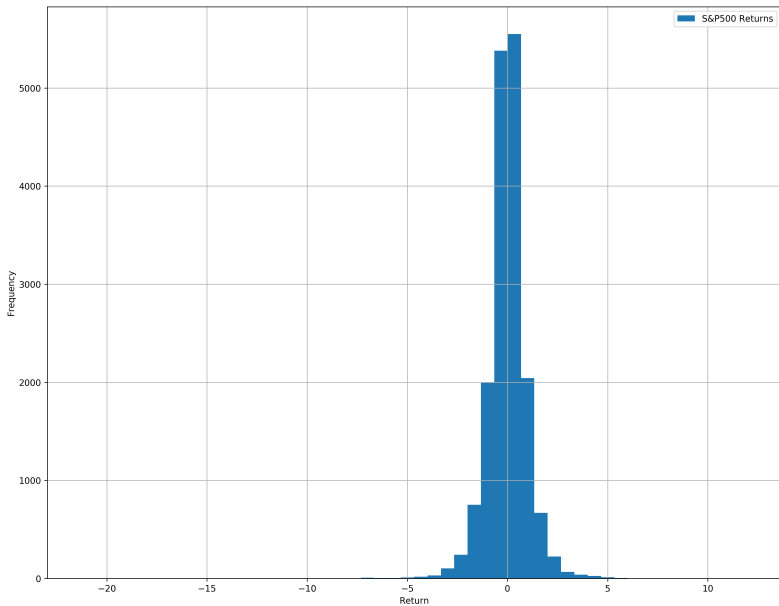
Algorithm for Simple NoVaS:

- ▶ Let  $\alpha = 0$  and  $a_i = \frac{1}{p+1}$  for all  $0 \leq i \leq p$
- ▶ Pick  $p$  such that  $|KURT_n(W_{t,p}^S)| \approx 3$

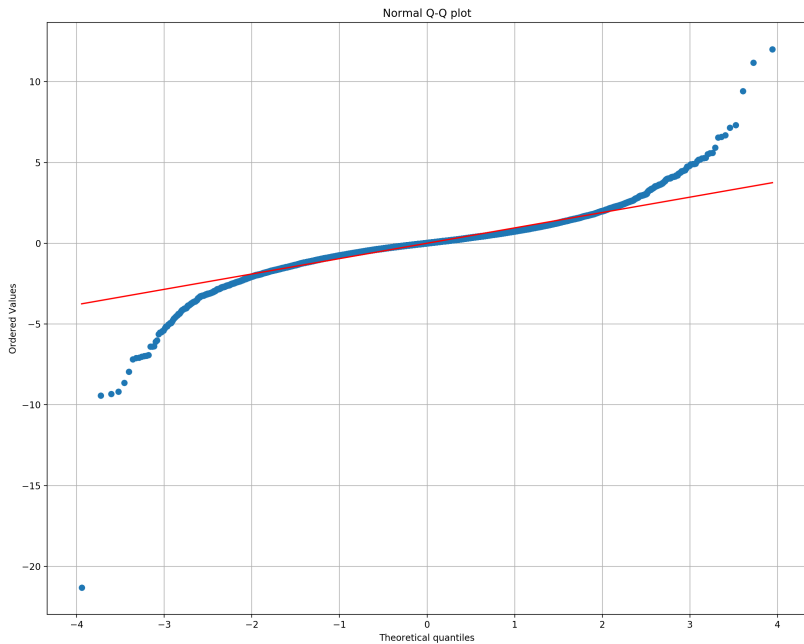
# SP500 Daily Returns (1950-2018)



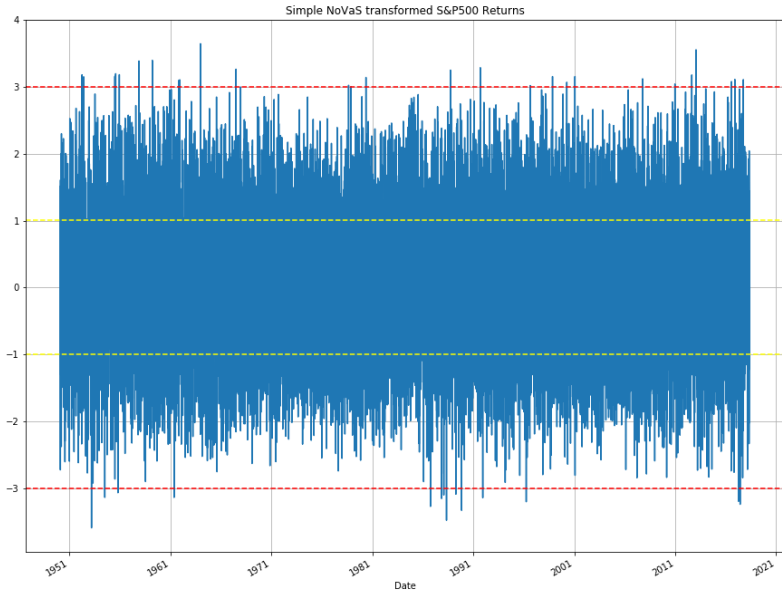
# SP500 Daily Returns Histogram



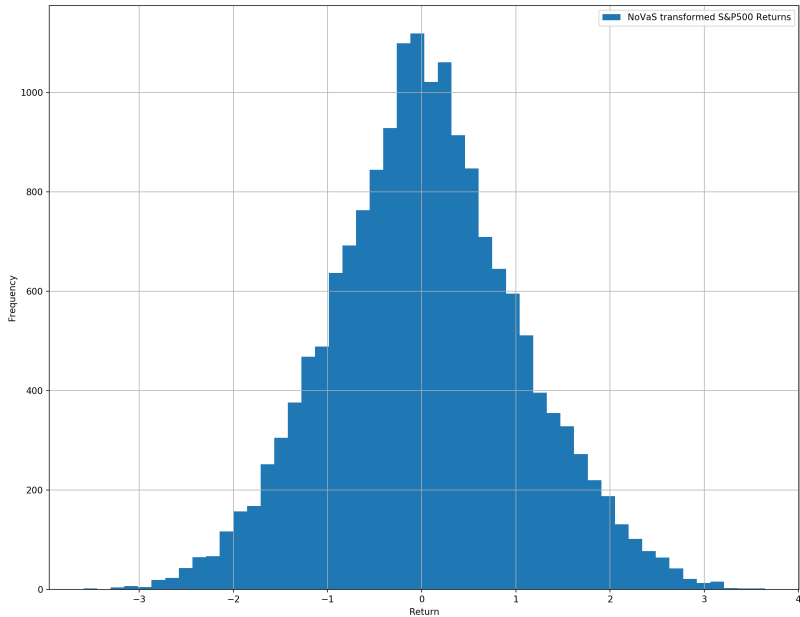
# SP500 Daily Returns Q-Q Plot



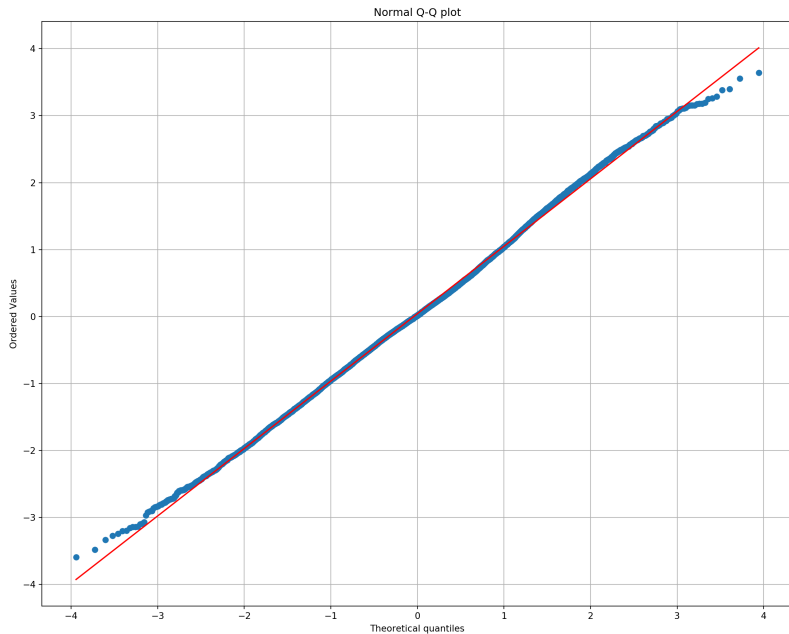
# NoVaS Transformed SP500 Daily Returns ( $p=16$ )



# NoVaS Transformed SP500 Histogram (p=16)

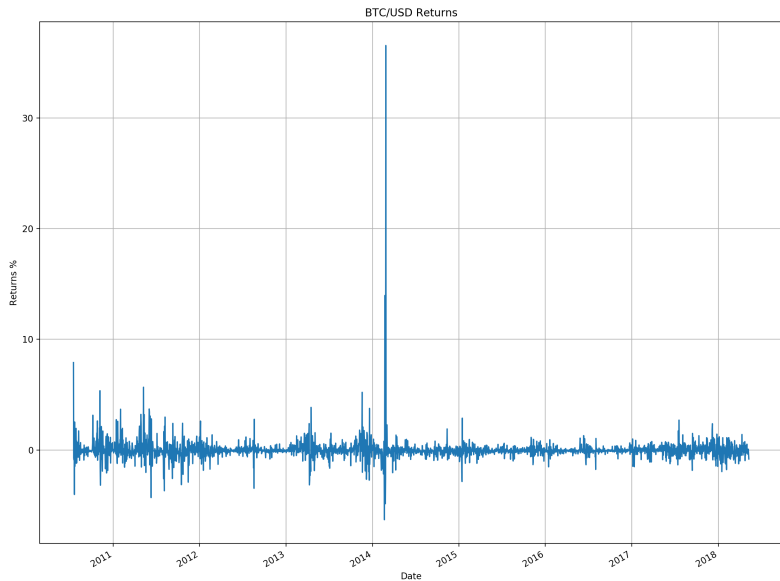


# NoVaS Transformed SP500 QQ-Plot ( $p=16$ )

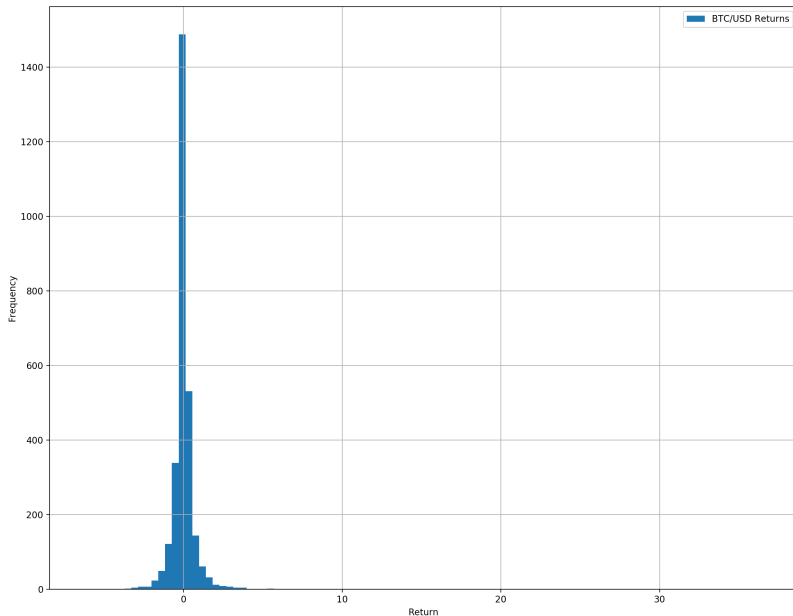




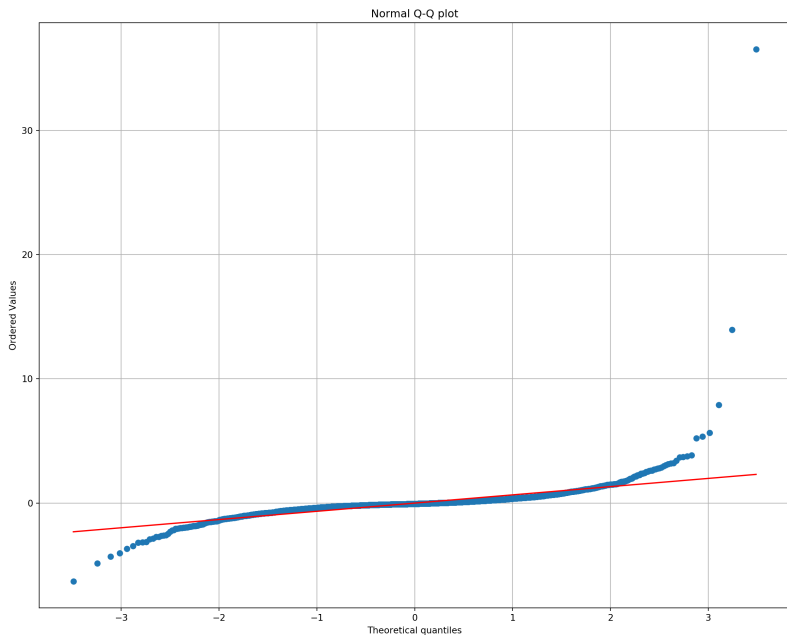
# BTC/USD Daily Returns (2010-2018)



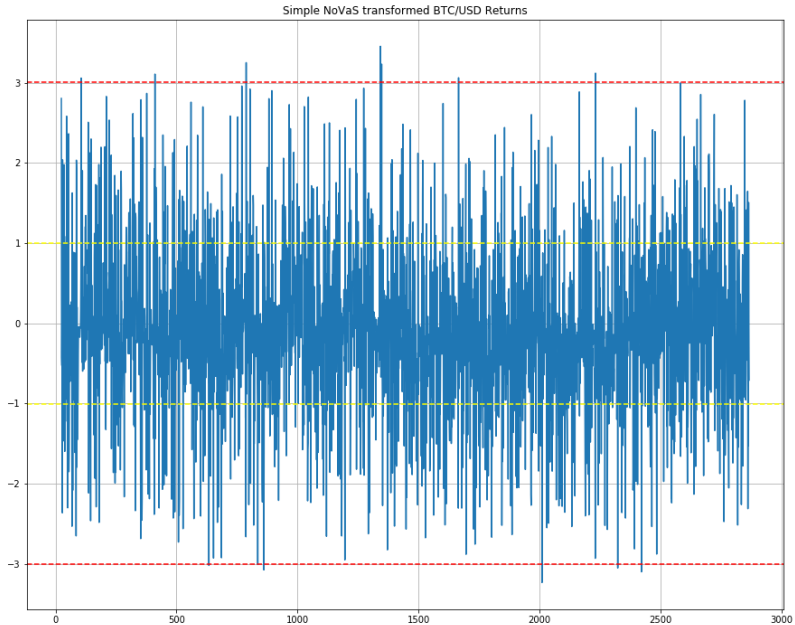
# BTC/USD Daily Returns Histogram (2010-2018)



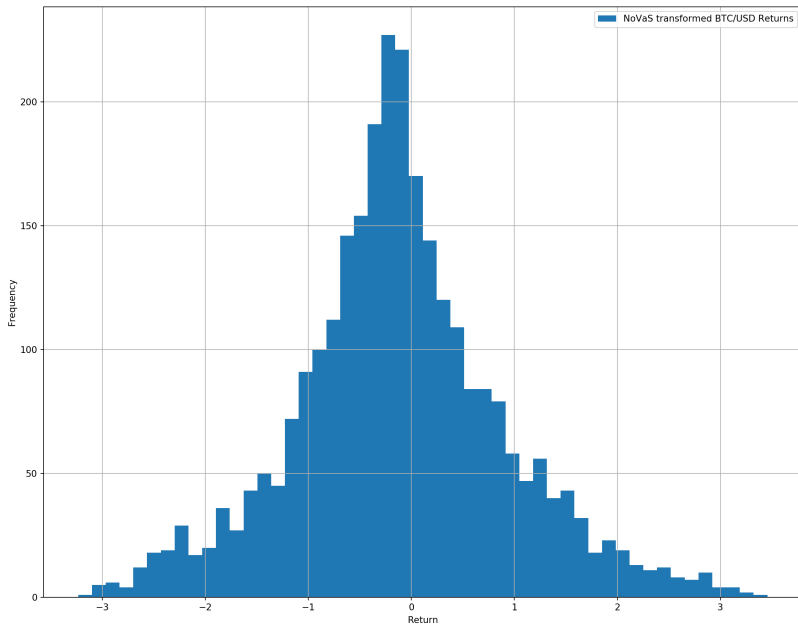
# BTC/USD Daily Returns QQ-Plot (2010-2018)



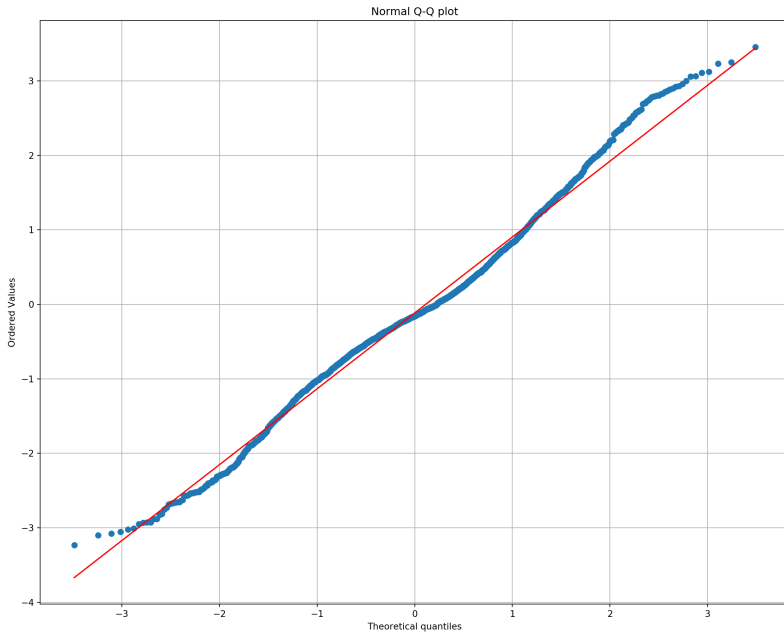
# NoVaS Transformed BTC/USD Returns ( $p=12$ )



# NoVaS Transformed BTC/USD Histogram (p=12)



# NoVaS Transformed BTC/USD QQ-Plot ( $p=12$ )



# Treasury Futures

pre-NoVas Treasury Futures returns plot

# Treasury Futures

pre-NoVas Treasury Futures returns histogram



# Treasury Futures

pre-NoVas Treasury Futures returns q-q plot

# Treasury Futures

post-NoVas Treasury Futures returns plot

# Treasury Futures

post-NoVas Treasury Futures returns histogram

# Treasury Futures

post-NoVas Treasury Futures returns q-q plot

## SP500 not perfect transform with NoVaS

Simply show an imperfect transformation to make the point that financial time series over long periods are not necessary stationary (only locally stationary) and thus we should use time-varying versions of NoVaS where the window size isn't too big.

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# One-Step Ahead Prediction

Define the volatility prediction problem. Outline that you're using squared returns  $Y_t^2$  as a noisy proxy for Volatility What loss function should you use? Will use a window size of 250 days which is approx. 1 year

# Infinite Kurtosis?

Do financial returns have infinite kurtosis? If this is the case you, predicting under L2 is incorrect. Instead you should L1 loss where the median is optimal



# Infinite Kurtosis Plot SP500

# Infinite Kurtosis Plot BTC

# Infinite Kurtosis Plot Treasury Futures

# Prediction Intervals

Steps for deriving prediction intervals - same for GARCH and NoVaS

# SP500 Feb 2018 One Step Ahead Prediction

Plot predicting SP500 Feb 2018 Volatility spike, along with prediction intervals Shows that Simple NoVaS is better than GARCH(1,1)

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## Can predict $\sigma^2$ using NoVaS under special conditions

Talk about the conditions under which you can actually predict  $\sigma^2$ , plot the ACF to confirm that transformed series is uncorrelated and independent.

$$RV(t+1)-IV(t)$$

Outline strategy that if  $RV(t+1)-IV(t) > 0$  you buy VXX and vice versa.



# Strategy Results

Cumulative returns plot, legend contains CAGR and Sharpe Ratio