

# Volatility Trading using Machine Learning

Amrit Prasad, Nathan Johnson, Salman Khan Pathan, Saurabh Kelkar

# Problem Statement

- Predict the volatility of the S&P 500 index
- Compare predicted volatility levels over a specified time horizon with market implied volatility and select a portfolio of options + index to capture movements in the premium
- Metric: 1. MSE 2. Trading Profitability

## Benchmark: GARCH (1,1)

- Use GARCH(1,1) estimates of realized volatility
- Backtest trading strategy as benchmark return

## ML Application for Volatility Prediction

- Multiple neural networks (e.g., Jordan Neural Network) demonstrate superiority to GARCH(1,1) models

## Trading Strategy

- Trade S&P options based on forecast volatility
- Ex: Expected premium contraction, short option and delta-hedge with daily rebalancing

# Data Sets: Stock Domain and Features

- S&P 500 Index realized volatility
- Set of traded options' implied volatilities
- Date: 1st Jan 2008 - 31st Dec 2017

## Data Cleaning:

- News Headlines scraped in JSON format
- Time Series Data: Is clean

### Technical/Price data

- Source: Bloomberg
- Usage: Technical Indicators based on OHLC Data.

### Google Trends

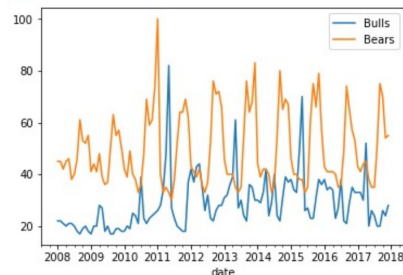
- Source: Google Trends API
- Usage: Word count occurrence in searches

### Implied Volatility

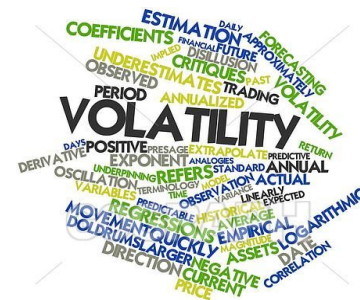
- Source: Option Metrics
- Usage: Trading Backtester

### News Headlines

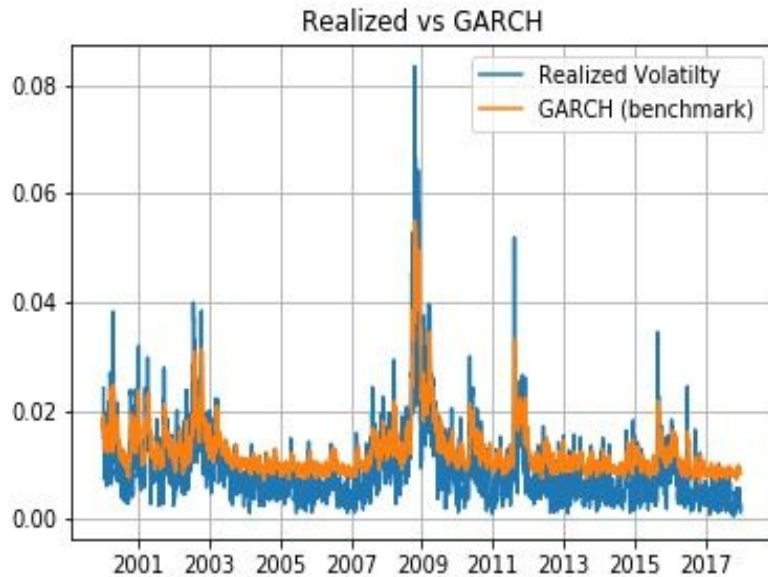
- Source: Scraping (WSJ Archives)
- Usage: NLP



Historical Volatility	Implied Volatility
<p>Look back in time to show where volatility has been in the past.</p> <p>*****</p>	<p>Traders view of expected future volatility based on current option prices.</p> <p>*****</p>
Shows expected trading range of market	Indicator of the current sentiment of the market



# Benchmark



## Key References:

- [1] Tae Roh, Forecasting the volatility of stock price index
- [2] Arnerić, et al, GARCH based artificial neural networks in forecasting conditional variance of stock returns

- Volatility has various characteristics:
  - Clustering
  - Asymmetry
  - Regimes Behaviour
  - Long-Term Memory
- GARCH(1,1) model is parsimonious and captures the clustering pretty well. However the asymmetric and regimes behaviour isn't described desirably. The decay of the coefficient on older lags also leads to a loss of long-term memory.
- Neural Networks and other ML algorithms have the ability to capture the missing aspects. They are particularly flexible as information apart from price returns can also be incorporated.