Backtesting a Trading Strategy in Python

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Outline

Motivation behind Backtesting

- Does the trading strategy you have hypothesised even perform ?

- Backtest is a process of testing the historical performance of a trading strategy. It is used as to check the if a strategy works on past and what performance we can expect in future.

- A backtest is a simulation which can be used for sanity check on the hypothesis behind the strategy and various factors under a given scenario of the market.

- Backtest is not an experiment and shouldn't aimed to prove profitability of a strategy as it is only historical performance.

Entire data start date: 2010-01-04
Entire data end date: 2017-03-21
Backtest months: 86

	Backtest
Annual return	5.7%
Cumulative returns	49.1%
Annual volatility	15.0%
Sharpe ratio	0.44
Calmar ratio	0.37
Stability	0.79
Max drawdown	-15.3%
Omega ratio	1.08
Sortino ratio	0.63
Skew	-0.28
Kurtosis	3.15
Tail ratio	1.03
Common sense ratio	1.09
Daily value at risk	-1.9%
Gross leverage	2.00
Daily turnover	16.2%
Alpha	-0.00
Beta	0.59

Seven Sins of Quantitative Investing

1. *Survivorship bias*: Ignoring the stocks that have gone bust or delisted in past and only considering the existing stocks in the sample. This can lead to overestimation of historical performance. E.g.- Only considering the current composition of SP500.

2. *Look-ahead bias*: Using information that is not available at the current moment.

3. *Storytelling*: Making up a story ex-post to justify some random pattern in the result.

4. *Data snooping*: Using the test sample for tuning the strategy and improving the backtest.

5. *Transaction cost*: Ignoring the transaction cost results inaccurate backtest results. While including transaction cost some strategies that were profitable earlier may fail.

6. *Outliers*: A backtest is severely skewed if it performance (profit or loss) is heavily dependent on few extreme outcomes that is observed in past. E.g.: Financial crash, Tech Boom.

7. *Shorting*: Shorting involves finding lender who can lend securities, which is dependent on inventory, relative demand and markets.

Outline

- Why Backtest ?

- **Methods of Backtesting**

- Types of Backtesting algorithm

- Backtest analysis

- Backtest of Time-series Momentum

- Backtest of Pairs-trading

Types of Backtest algorithms

Vectorized

- Vectorized the backtesting loop using Numpy/Pandas

- Fastest

- Not a reliable backtest. Is used as a sanity check.

- Easy to implement.

For Loop

- The backtest is executed using a for loop.

- Slower than vectorized.

- Not a reliable backtest. Is used as a sanity check.

- Easy to implement.

Event Driven

- Uses market simulation and runs the backtest as events takes place.

- Slowest

- The most reliable backtest.

- Difficult to implement.

Outline

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- Methods of Backtesting

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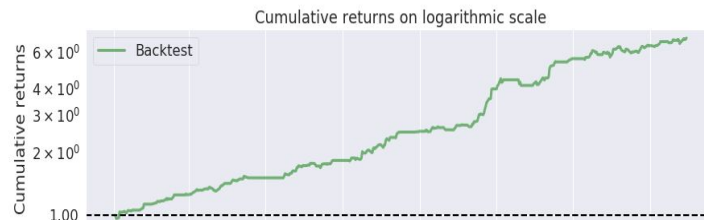- Backtest of Pairs-trading

Risk and Performance Metrics (1)

- **Cumulative Returns**

$$R_c = \prod_{t=0}^{n}(1 + r_t)$$

where

r_t = return at t

n = number of observations

Cumulative returns on logarithmic scale

- **Annualized Return**

$$R_A = R_c^{\frac{m}{n}} - 1$$

where

R_c = Cumulative returns

m = periods per year

n = number of observations

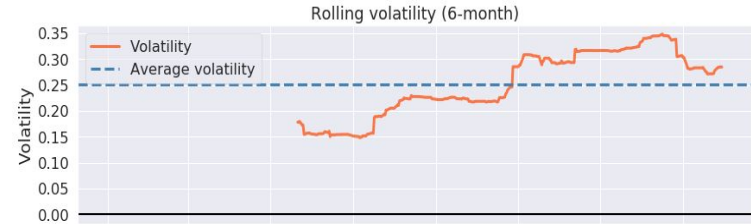Returns

Risk and Performance Metrics (2)

- **Annualized Volatility**

$$\sigma_A = \sigma_r \sqrt{m}$$

where

$$\sigma_r = \text{std. dev. of returns}$$
$$m = \text{periods per year}$$

Rolling volatility (6-month)

- **Annualized Sharpe Ratio**

$$SR = \frac{R_A - R_f}{\sigma_A}$$

where

$$R_A = \text{Annualized return}$$
$$R_f = \text{risk-free rate}$$
$$\sigma_A = \text{Annualized volatility}$$
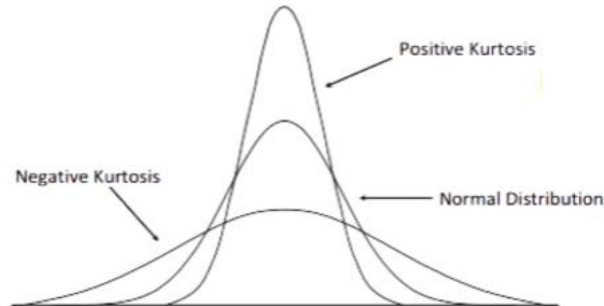

Rolling Sharpe ratio (6-month)
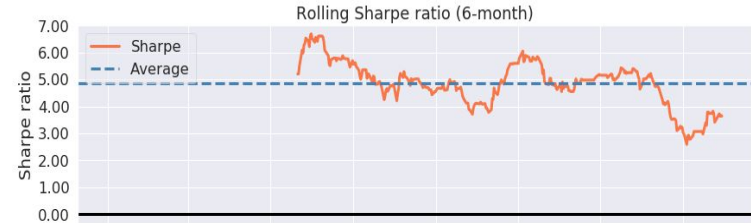
Risk and Performance Metrics (3)

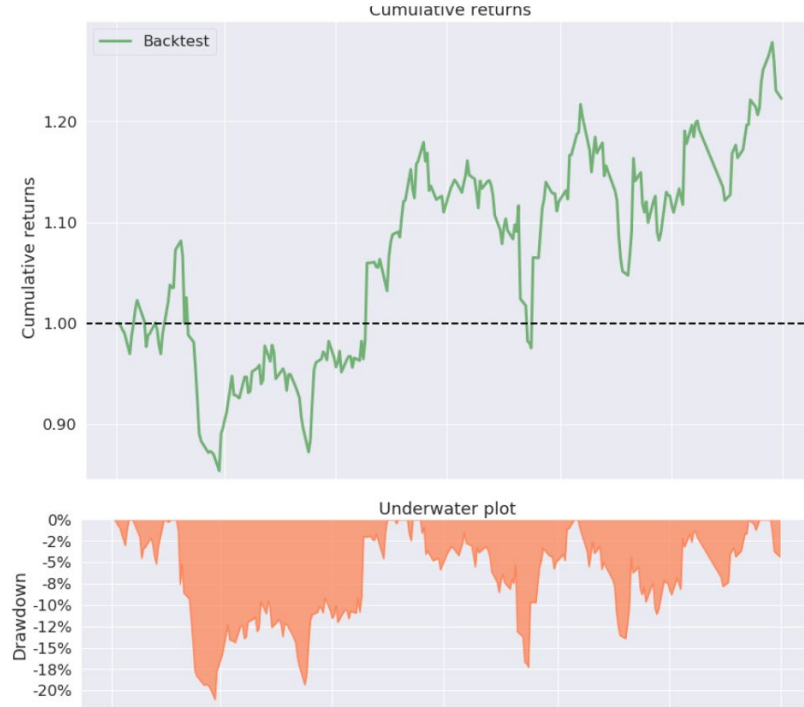- **Skewness**
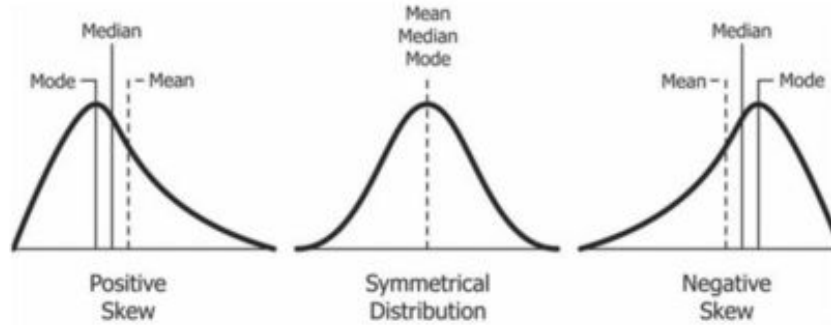
Source: wikipedia

- **Kurtosis**

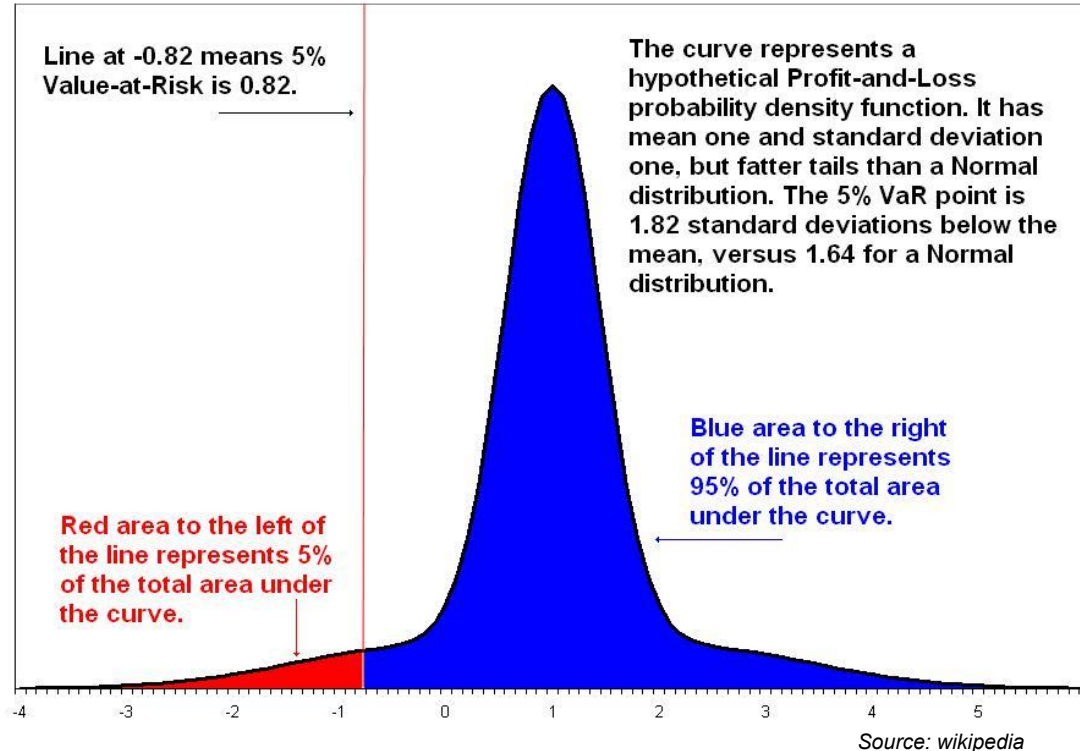# Risk and Performance Metrics (4)

- **Maximum Drawdown**

$$MDD = \frac{\text{Trough Value} - \text{Peak Value}}{\text{Peak Value}}$$

Risk and Performance Metrics (5)

- **VaR (Value at Risk)**

Line at -0.82 means 5% Value-at-Risk is 0.82.

The curve represents a hypothetical Profit-and-Loss probability density function. It has mean one and standard deviation one, but fatter tails than a Normal distribution. The 5% VaR point is 1.82 standard deviations below the mean, versus 1.64 for a Normal distribution.

Blue area to the right of the line represents 95% of the total area under the curve.

Red area to the left of the line represents 5% of the total area under the curve.

Source: wikipedia

Backtest : Time-series Momentum

Notebook Link: https://bit.ly/3xOvxN8

Backtest : Pairs-Trading

Notebook Link: https://bit.ly/3v7KWWK

References

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QNA

Task for Day 2

Backtest: Time-series Momentum

- Backtest the strategy discussed in the video :
  https://www.youtube.com/watch?v=D_WhInJePC8

- Perform the backtest on a stock of your choice and collect the historical data using Alpaca Trading API.

- Vectorized backtest is preferable but for-loop can also be used.

- Calculate the risk and performance metrics.

- Plot cumulative returns against a benchmark, drawdown and rolling Sharpe ratio.

- Extra points: Include transaction cost.

- Use Google Colab

- Last date for submission : 05-05-2022

- Submission form link : https://forms.gle/JgrH7U7sdoSMVLUU8