

## **Algorithmic-Trading - Summary Paper - Motivations, Foundations, Techniques**

### **Introduction and Foundations**

Algorithmic-Trading (AT), is defined as follows:

“In algorithmic trading (AT), computers directly interface with trading platforms, placing orders without immediate human intervention. The computers observe market data and possibly other information at very high frequency, and, based on a built-in algorithm, send back trading instructions, often within milliseconds. A variety of algorithms are used: for example, some look for arbitrage opportunities, including small discrepancies in the exchange rates between three currencies; some seek optimal execution of large orders at the minimum cost; and some seek to implement longer-term trading strategies in search of profits.” (Chaboud et al. 2009).

In other words, AT is the generation and submission of buy and sell orders by an algorithm, or set of instructions that processes current market data and places orders in stock marketplaces without human interaction.

High Frequency Trading (HFT), is a subset of algorithmic trading and a far newer phenomenon that has been made possible by the rapid improvement of computerized trading speed. This is the primary form of algorithmic trading found in financial markets today, with billions of dollars constantly traded by machines every second.

AT is largely unregulated. There are absolutely no restrictions on electronic trading, which has resulted in various performance throughout entire markets. See the Flash Crash below. Especially with the rise of costless transactions with platforms like Robinhood, it has never been easier for individuals to engage in AT.

Rapid trading traces its roots back to the early 1930s, with specialists and pit traders buying and selling positions at a physical location of the exchange and broadcasting it via telegram services. Computerization of trades started in the 1980s when the NASDAQ introduced purely electronic trading. Today, HFT trading time has changed from a matter of seconds to microseconds.

The May 6th, 2010 “Flash Crash” brought the public’s attention to the little publicized, but very heavily used algorithmic trading in financial markets. This happened with E-mini, denoted by ES, which is a stock market index futures contract that trades for around 50 times the value of the S&P. A mutual fund complex sold 75,000 of these contracts valued at approximately 4.1 billion dollars - resulting in the largest net change in daily position of any trader in the E-mini since the beginning of the year. This caused a cascading effect, although initially absorbed by HFT’s, with over 20,000 trades across 300 separate securities executing at prices 60 percent away from their prices a mere half hour earlier.

### **Motivations for Research**

AT is a relatively secretive area of financial trading, but can result in massive profits if performed correctly with proper portfolio management, risk and alpha models.

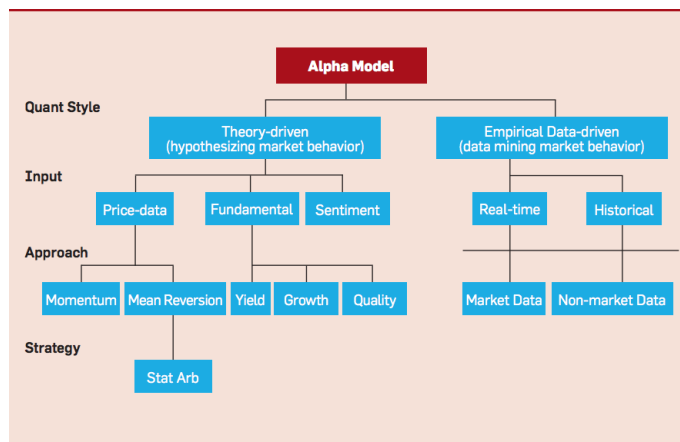


Figure 1: Alpha Models

## Key Techniques

### Simple Moving Average (SMA)

This is calculated by adding a certain amount of closing prices and dividing by the total amount of days. Short term averages are often compared against longer term averages, often signaling uptrends and downtrends. Popular methods include the golden cross and death cross, with the 50 day SMA crosses below or above the 200 day moving average. The former is often a buy signal reinforced by high trading volumes while the latter is considered bearish.

### Pairs Trading and Arbitrage

Statistical arbitrage uses time series methods to identify brief pricing errors of stocks. HFT takes advantage of this because these discrepancies in price exist for mere fractions of a second. Statistical arbitrage is something that is far more practical to implement in the bitcoin market, especially without the tools that prop-trading firms have. Pairs trading uses arbitrage by choosing stocks that are correlated and move similarly.

### Market Making

Another common form of algorithmic trading often used by large funds or market makers in a financial market. It provides both advantages to the market making institution as well as the market participants. Automated scripts will never deviate into risky actions that human traders may complete and further reduces the incidence of market crashes and negative surprises for the market maker's bottom line. A sample naive market making strategy is fixed offset, which continuously places limit orders on both sides a specified number of ticks away from the market price.

### Mean Reversion

This strategy assumes that prices and returns eventually move back towards the mean. Pairs trading takes advantage of this by choosing stocks that are historically correlated that have diverged and will most likely converge in the future.

## **Machine Learning - Bayesian networks**

A bayesian network is a dag that represents a set of variables and their conditional dependencies. In this case, it will give decisions based on certain market conditions and prices to sell or buy a particular stock based on this learned conditional dependencies. Interesting research suggests that gated bayesian networks are far more effective in AT.

## **Momentum Trading Strategies**

General strategies that aim at capturing trends in the market fall under this umbrella. SMA is an example of a momentum trading strategy.

## **Risk Model**

This method is less of a trading strategy and more of a stop method used to evaluate portfolios. A sample model may evaluate the VAR (value at risk) for everything owned in the portfolio. If the VAR reaches a certain threshold in comparison to its value, close all positions in the portfolio.