HTF

an alternative approach by: Andrew Carpenter



What is HTF?

• High Frequency Trading :

Most commonly known as trades taking place in time intervals ranging from hours to microseconds and the volumes of the stocks traded tend to be quite large \sim around 50,000 shares at a time.

• Additional HTF characteristics:

Exploiting the inefficiencies of the market to make money off of the small fluctuations in price over a short time-interval

Each individual stock sold usually only makes fractions of a dollar or even a single penny.

HTF most often involves the use of an algorithmic trading strategy executed by computer programs written in c++

Key principles involved

- Most often the models for HTF algorithms make use of the inefficiencies of the market.
 - "The relative availability of trading opportunities can be measured as a degree of market inefficiency. "[1]

- "The more inefficient the market, the more predictable trading opportunities become available. Tests for market efficiency help discover the extent of predictable trading opportunities." [1]

- For inefficient markets, price fluctuations for a short period of time are have a degree of nonrandomness and can be correlated to other factors within a certain degree of accuracy

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Conclusion

- If market is *inefficient*: price fluctuations are *predictable*.
- If market is *efficient*: price fluctuations are unpredictable ~ random.

An alternative approach

- If instead we look for efficient markets (markets for certain stocks) then we know that their prices should fluctuate randomly.
- These markets can be found by using certain *test* that check for randomness.
- The *opportunities* that exist in a randomly fluctuating market can be found by:

Identifying the momentary local minimums in the price.

Identifying the momentary local maximums in the price.



- $n_2 = sum of green lines = 7$
- U = sum of consecutive green lines and consecutive blue lines = 9

Test for randomness
[1]
$$Z = \frac{|u - \bar{x}| - 0.5}{s}, \qquad Z < 1.645$$
$$s = \sqrt{\frac{2n_1n_2(2n_1n_2 - n_1 - n_2)}{(n_1 + n_2)^2(n_1 + n_2 - 1)}} \quad \bar{x} = \frac{2n_1n_2}{n_1 + n_2} + 1$$

Denote the total number of runs, both positive and negative, observed in the sample as u

Denote n_1 as the number of positive 1-minute changes Denote n_2 as the number of negative 1-minute changes

If Z < 1.645, then the 1-minute changes are random

Define local minimums:

- The condition that there is a *strong statistical chance* that the next *change* will be *positive*
 - -The chance that you flip a coin to get 10 heads in a row is a bit small, therefore I'd be more willing to but my stakes on 7 heads and 3 tails.
- Determine a rule set in algorithm to define these favorable conditions to buy stocks

-For example: Rule 1 – If n_2 increases 7 times consecutively, buy stocks at current price

A main for loop contains the Buy and Sell Rule set

For(i =1 ; i <362; i ++){ //i represents a minute

- Buy rule for-loop
- Sell rule for-loop

Use of Entropy – Buy Rule

• U = Nn + Pn ;

•

•

```
if(U > 0){probNn = (Nn / U);}
```

//avoids singularities

```
if(U > 0){probPn = (Pn / U);} // ^
```

```
    if( probNn > 0 && probPn > 0 )
{Entropy = -1.0*probNn*log10(probNn) +
-1.0*probPn*log10(probPn);}
```

Sell Rule

- **f**or(n=1; n <= nimax ; n++){
- if(p[i]*1.0 boughtprice5[n]*1.0 > 0.01 && boughtstocks5[n] > 0 && boughtprice5[n] > 0.0){

sellprice5[n] = p[i];

netcash = netcash + boughtstocks5[n]*p[i]; boughtstocks5[n] = 0;

```
    cout << i << " " << "sellprice5[" << n << "]= "</li>
    sellprice5[n] << endl;</li>
```

```
fout << i << " " << sellprice5[n] << endl;
}
}
```

*Green points: Price bought at *Blue points: Price sold at

Symbol DIG



Started off really well...

Money made/lost plotted by tradding days



Trading Days

References

1. Aldridge, Irene. *High-frequency Trading: A Practical Guide to Algorithmic Strategies and Trading Systems*. Hoboken, NJ: Wiley, 2010. Print.