## 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 ~ 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


## Conclusion

- If market is inefficient: price fluctuations are predictable.
- If market is efficient: price fluctuations are unpredictable $\sim$ 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.

- $\mathrm{n}_{1}=$ sum blue lines $=8$
- $\mathrm{n}_{2}=$ sum of green lines $=7$
- $\mathrm{U}=$ sum of consecutive green lines and consecutive blue lines $=9$


## Test for randomness [1]

$$
Z=\frac{|u-\bar{x}|-0.5}{s}
$$

$$
Z<1.645
$$

$s=\sqrt{\frac{2 n_{1} n_{2}\left(2 n_{1} n_{2}-n_{1}-n_{2}\right)}{\left(n_{1}+n_{2}\right)^{2}\left(n_{1}+n_{2}-1\right)}}$

$$
\bar{x}=\frac{2 n_{1} n_{2}}{n_{1}+n_{2}}+1
$$

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 l'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 setFor( i =1; i <362; i ++) $/ / \mathrm{i}$ represents a minute

- Buy rule for-loop
- Sell rule for-loop
\}


## Use of Entropy - Buy Rule

- $\mathrm{U}=\mathrm{Nn}+\mathrm{Pn}$;
- 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);}
if( (1-pow( (Entropy/EnMax),2.0 ) ) > 0.8 && Nn > Pn && Nn > 9 && i < 362 &&
floor((netcash*0.5)/p[i]) >= 1.0 ){
    n++;
    nimax = n;
    boughtstocks5[n] = floor( (netcash*0.5)/p[i] );
    netcash = (netcash - boughtstocks5[n]*p[i]) ;
    boughtprice5[n] = p[i];
    fout << i << " " << boughtprice5[n] << endl;
    cout << i << " " << "boughtprice5["<< n << "]=" <<
boughtprice5[n] << " " << p[i] << endl;
    }
```


## Sell Rule

- $\operatorname{for}(\mathrm{n}=1 ; \mathrm{n}<=\operatorname{nimax} ; \mathrm{n}++)\{$
if( $\mathrm{p}[\mathrm{i}]^{* 1} 1.0$ - boughtprice5[n]*1.0 $>0.01$ \&\& boughtstocks5[n] > 0 \& \& boughtprice5[n] > 0.0) \{
- boughtstocks5[n] = 0;

$$
\begin{aligned}
& \text { sellprice } 5[\mathrm{n}]=\mathrm{p}[\mathrm{i}] ; \\
& \text { netcash }=\text { netcash + boughtstocks5[n]*p[i]; }
\end{aligned}
$$

cout << i << " " << "sellprice5[" << n << "]= " << sellprice5[n] << endl;
fout << i << " " << sellprice5[n] << endl;

$$
\}
$$

- 

\}

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



## Started off really well...

Money made/lost plotted by tradding days


## References

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