High Frequency Statistical Arbitrage Model

Pair trading using price movement per second in correlated companies

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MS&E 448

Stanford University | Spring 2019

- Review since midterm
- Model selection
- Simulation
- Results
- Discussion and future work

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Review

Midterm milestone:

- Company selection
 - Two methods discussed:
 - Spherical KMeans on the whole features vectors
 - Euclidean KMeans at every time stamp
- Cointegration of clusters:
 - Chose to focus on just pairs

Review

Defining the universe:

- Final train/test set:
 - Train: one day of data 1/28/19 from 9:30am-4pm
 - Test: following three days
- Final company selection:
 - Two types of results:
 - Clustering
 - <u>19 most correlated pairs, able to execute trades on 14</u>

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Building the models

- Split train set into train and validation sets
- Add features: lags of wmid and volume at 15 and 30 seconds
- Run LASSO, forward stepwise, backward stepwise, and both ways on all features
- Compare methods using MSE

Features: bid, ask, mid, wmid, bsize, asize, anum, bnum, volume, notional, last_price, last_size + lags

Testing the models

NWS PDD Mean square error: 0.002 R2: 0.973 Mean abs error: 0.034



*00S test from 9:30 am - 12 pm

BKNG NWSA Mean square error: 1.143 R2: 0.963 Mean abs error: 0.706



NWS ULTA Mean square error: 0.034 R2: 0.961 Mean abs error: 0.138





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Simulation Strategy

- Best model for each pair selected
- Start with \$10,000, 1 day after training period
- Model used to predict residual return 30 seconds in future
- If prediction is falls outside num_std of residual return, execute a trade
 - Band selection: num_std is a hyperparameter
 - Quantity limited to 1, executed on midprice instead of buy/ask
- Model trained in *online* fashion, i.e. model trained as more data is received

Backend and testing details

- Simulation based on Thesys Technologies simulator
- Difficult to work with simulator
 - Lack of API reference documentation
 - Uptime on server is poor, outages prevented us from testing
 - Missing data needs to be taken into account at runtime
- Performed runs over various timescales, i.e. minutes, hours, weeks after training period
- Performed runs with different bands to optimize num_std

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Results: Pair Trade NWS and ULTA

NWS

Company Name: News Corp

About: Mass Media and Publishing company

Industry: Media

Current Stock Price: \$11.58

ULTA

Company Name: ULTA Beauty

About: Beauty Products, Makeup

Industry: Cosmetics

Current Stock Price: \$335.53

TAKEAWAYS:

- Worried about spurious correlation
- Model has strong signals and high prediction capabilities

Results: Profit/Loss 3 Days (10:00am-12:00pm)



Results: Returns 3 Days



Results: Histogram of Returns 3 Days



Results: Pair Trade VRSK and VIA

VRSK

Company Name: Verisk Analytics

About: Data Analytics and Risk Management

Industry: Data Analytics

Current Stock Price: \$141.84

TAKEAWAYS:

- High Fluctuation from results/Intraday Market Risk
- Lower predictive capabilities
- Overfitting and Multicollinearity
- Different Sector than previous example

VIA

Company Name: Viacom

About: Mass media conglomerate

Industry: Media

Current Stock Price: \$34.01

Results: Profit/Loss 3 Days (10:00am-12:00pm)



Results: Returns 3 Days



Results: Histogram of Returns 3 Days



Results: Across tested Pairs (1/29-1/31) *10am-12pm

Pair of Stocks from NASDAQ 100	Profit (In \$)
BKNG, NWSA	82.785
NTES, NWS	-26.605
NWS, PDD	1.34
NWS, VIA	2.96
PDD, VIA	2.395
HSIC, VIA	-2.595
IDXX, VIA	-1.97

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Discussion

Strengths:

- Made money even using a rudimentary model
- Able to capture returns from highly correlated stocks

Weaknesses:

- Lost money for some of our pairs
- Quantity is unspecified
- Bands are fixed

Future work

- Determine optimal quantities to trade on
- Optimal band selection
- Pull more data to train on
- Update models and pairings (at minimum) biannually
- Build our own simulator

References

[1] Cartea Alvaro, Jaimungal Sebastian, Penalva José (2015). Algorithmic And High-Frequency Trading.

[2] Almgren Robert, Chriss Neil(1999). Optimal Execution of Portfolio Transactions.

[3] Elliott, Robert & van der Hoek, John & P. Malcolm, William. (2005). Pairs Trading. Quantitative Finance.

Final pairs traded: (BKNG, NWSA), (FOX, NWS), (HSIC, VIA), (IDXX, NWS), (IDXX, VIA), (MELI, NWS), (NTES, NWS), (NTES, VIA), (NWS, PDD), (NWS, ULTA), (NWS, VIA), (NWS, VRSK), (PDD, VIA), (VIA, VRSK)



Questions?