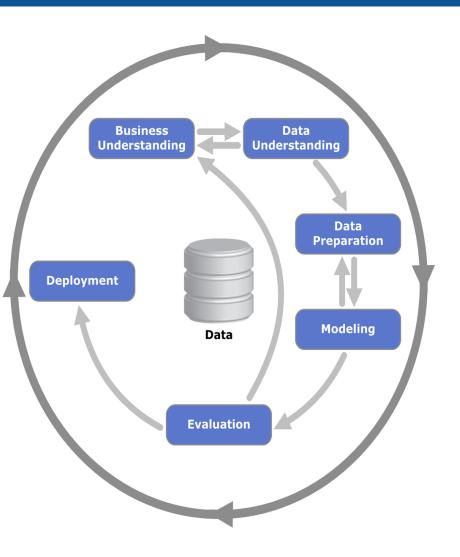




# Agenda



- 1. Problem Statements
- 2. EDA & Hypothesis Definition
- 3. Features selection & Modeling
- 4. Conclusions & Recommendations



#### **Problem Statement**

Over 10 years historical dataset following quant-trading approach is used in integration of market risks with 03 well-known alpha factors: Momentum, Mean reversion & Seasonality. This is the massive trading algorithm that allows investors quantitatively working with an enormous number of stocks at a time. It is considered as a promising approach for investors, which could minimize bad behaviors during trading, beating the benchmark in Vietnam market.

### 1 Context

Vietnam stock exchange is one of the most potential raising up players in frontier markets. The country which has a dynamic, young population structure & a highly open economic regime with more than 20 years stock exchange life-long is explicitly a promising opportunity for domestic & foreign investors making profits. The market is in an early developing phase & lacking massive & advanced financial tools such as algo-trading (quant-trading), an alternative of fundamental & technical analysis.

### 2 Criteria for success

Vietnam stock exchange in a nutshell

Recommending a betas system & potential alpha factors

Build a hybrid model blending market risks & alpha factors in order to form a set of optimum weight schemes that then are used to design trading strategies to defeat the benchmark (VNindex/VN30).

### 3 Scope of solution space

10 years of historical stock data of Vietnam market.

Project only targets to build a multi-factor model, the combination of risk & alpha model. Multi-factor model outcomes are optimum alpha weights for each stock & portfolio net factor exposures for each principal component. Strategy design & backtest are out of scope.

Deliverables include: code. documentation & slice deck

## 4 Constraints within solution space

Solution is just considered as a reference, not financial advice. Solution is time bounded & subject to type of security

### 5 Stakeholders to provide key insight

Data from online open-source platforms

Solution could be advised from: stock traders, quantitative engineers/analysts, portfolio managers

### 6 Key data sources

https://www.vndirect.com.vn/portal/thong-ke-thi-truong-chung-khoan/lich-su-gia.shtml

https://s.cafef.vn/Lich-su-giao-dich-VNINDEX-1.chn#data

### **Data Wrangling**

#### **Dataset:**

- 1. Historical stock data: OHLC Volume Dividend of 621 tickers, from 2006-2021
- 2. Sector: 621 stocks' sector, from 2006-2021

### **Data cleansing:**

Remove duplicate & unnecessary columns

Extract states from locations, correct states 'name convention

Convert data type from object to numeric one

#### Raw features:

'open', 'high', 'low', 'close', 'volume', 'dividend', 'sector'



**Exploration Data Analysis** 

#### **Features definition:**

Key feature: close (stock closing price)

### **Features transforming:**

Create daily simple returns & log returns features

#### **Feature extraction:**

Create Bollinger band feature set

Create breakout feature set

Create Momentum feature set for 1 week, 1 month, 1 quarter, 1 half & 1 year

#### **Features for visualization:**

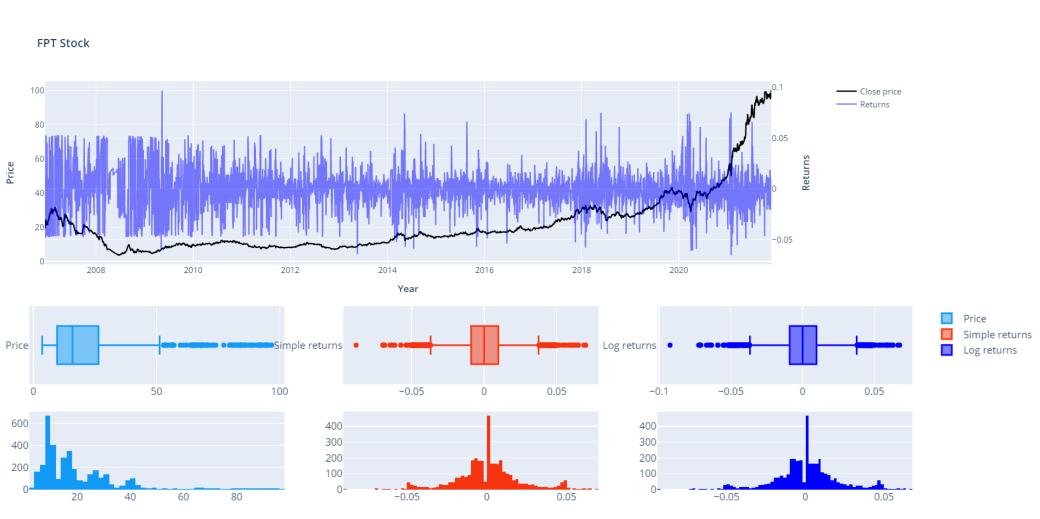
close, simple returns, log returns, Bollinger band feature set, breakout feature set



**Exploration Data Analysis – Data Overview** 

#### **FPT stock:**

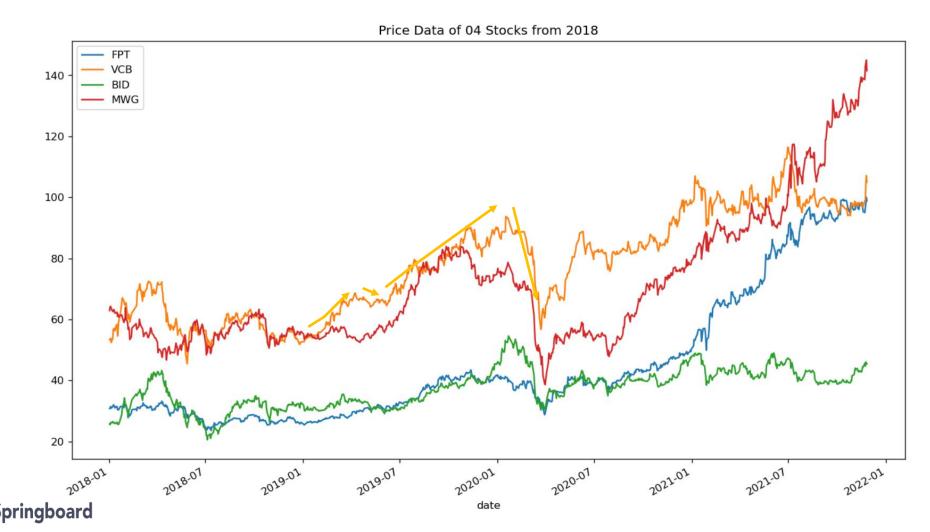
Price is non-stationary data, not normal distribution with mean & variance change by time Returns data is more stationary, normal distribution with mean=0 & variance/standard deviation stable by time Return data will be the input for model construction later on



## **Exploration Data Analysis – Visualization Stock Price**

#### Top 4 stocks of the market:

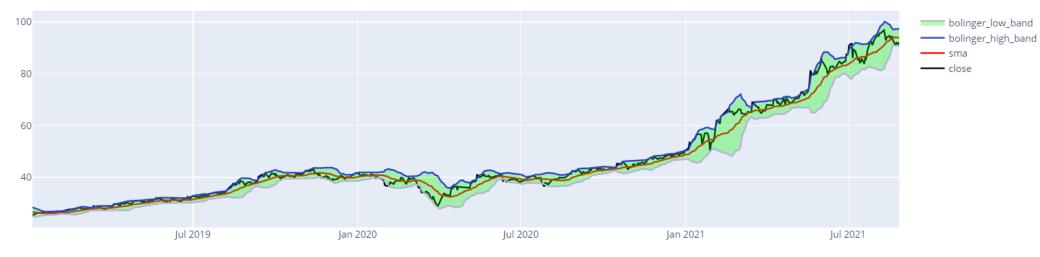
Both momentum & mean reversion effect could be seen



## **Exploration Data Analysis – Bollinger Band Effect**

#### FPT stock Bollinger band effect from 2018-2021:

- 1 month simple moving average curve (sma), upper & lower band cut the ticker price curve which could be long & short signal





## **Exploration Data Analysis – Breakout Effect**

#### Features visualization by time: 2009-2018:

- 1 month lookback low & high curves cut the FPT price one creating potential trading signals

#### High and Low of FPT Stock

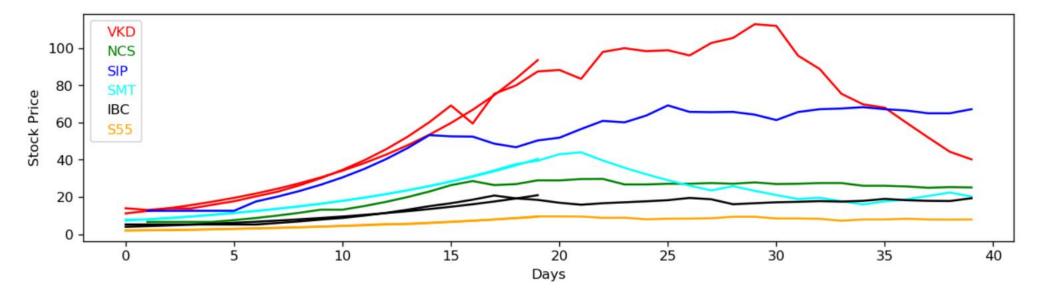




## **Exploration Data Analysis – Momentum Effect**

#### Time series analysis: 2011-2021:

- Top 6 stocks have highest 1month momentum effect
- The first 3 stocks (VKD, SIP, SMT) seems having a significant momentum intensity
- 1month momentum effect for the rest are predicted from fair to weak, even negative.

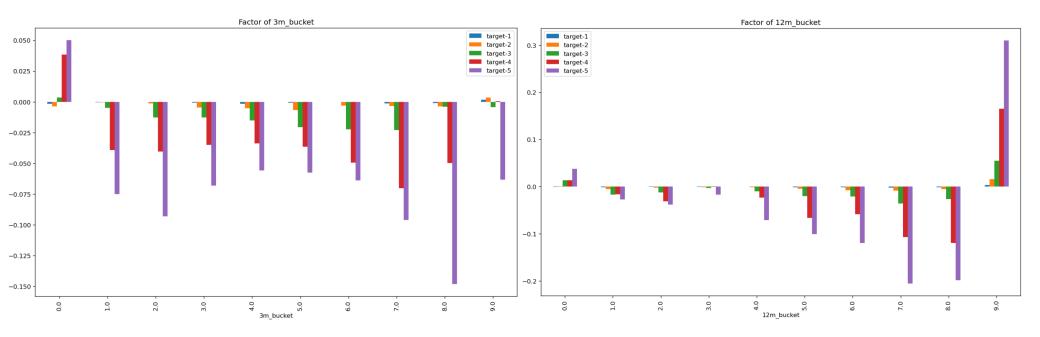




## **Exploration Data Analysis – Momentum Effect**

#### Cross sectional momentum analysis: 04 stocks basket (FPT, VCB, BID, MWG)

- A year momentum works better than quarter one, quantile 10<sup>th</sup> also perform the best

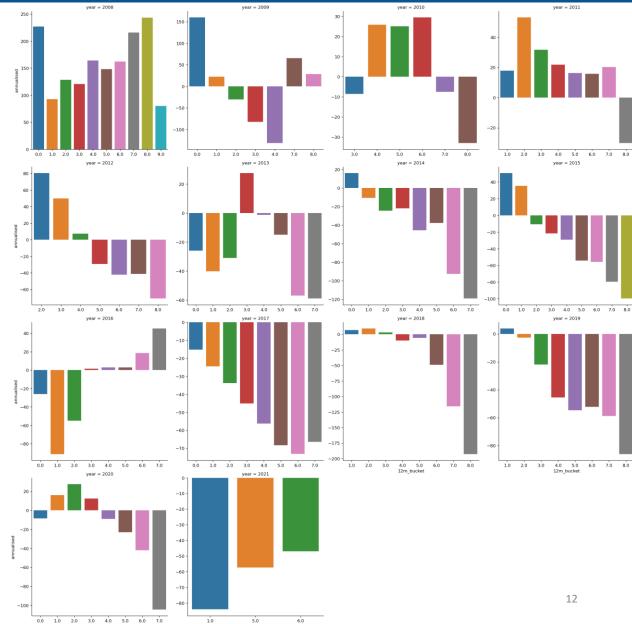




**Exploration Data Analysis – Momentum Effect** 

# Cross sectional momentum analysis: 04 stocks basket (FPT, VCB, BID, MWG) 2008-2021

- 12months momentum displays by year
- The first half of quantiles seem more resilient than the last part





### **Hypotheses Statements**

#### **Momentum**

Weekly, monthly, quarterly, halfly & yearly momentum indicators have been observed both in time series & cross sectional analysis, they are considered as the short, mid & long-term alpha factors.

- Yearly momentum hypothesis: it is "Higher past 12-months (252 days) returns are proportional to future return."
- Halfly momentum hypothesis: it is "Higher past 6-months (120 days) returns are proportional to future return."
- Quarterly momentum hypothesis: it is "Higher past 3-months (60 days) returns are proportional to future return."
- Monthly momentum hypothesis: it is "Higher past 1-month (20 days) returns are proportional to future return."
- Weekly momentum hypothesis: it is "Higher past 1-week (5 days) returns are proportional to future return."

The other indicators such as mean reversion & overnight sentiment - seasonality, both are short-time factors, are also taken into account in the model.

#### Mean reversion

Mean reversion theory states that security prices and economic indicators such as interest rates will tend to revert to the historical mean prices.

- Mean reversion 5day "Short-term outperformers (underperformers) compared to their sector will revert."
- Mean\_reversion\_5days\_smoothed is a smoothed version of Mean\_reversion\_5day factor
- Mean reversion 20day "Short-term outperformers (underperformers) compared to their sector will revert.

#### **Overnight sentiment**

The US stocks have significantly higher return during the night session compared to the daily session.

- Overnight Sentiment
- Overnight Sentiment Smoothed



### **Model Construction**

#### **Candidate selection**

This is a pretty high level research, it covers the whole VN stock exchange. 621 exchange-traded stocks, from 2011-09-01 to 2021-09-01 will be used as input for modeling process.

Poor candidate selection could lead to scattering/divergent in analysis & low performance of the models.

$$R_{p}(t) = \sum (B_{p,k} * R_{m}(t) + \alpha_{p}(t))$$

Risk factors are well-known and significant contributors to the variance of asset returns. **Risk model** actually is **Portfolio Covariance Matrix** will be built on stock returns data under CAPM, EMH & Modern Portfolio Theory-Effective Frontier theories to compute model factors: factor\_betas, factor\_returns, factor\_cov\_matrix, idiosyncratic\_var\_matrix & idiosyncratic\_var\_vector to estimate the portfolio risk. It will then be used as the constrains for alpha operation in the weight optimization process of multi-factor model

Factors that are proprietary and significant in describing the mean of asset returns can be candidates for alpha factors. **Alpha model** is built under Fundament law. Momentum, mean-reversion and seasonality effects are the key resilient factors for alpha model.

Finally, the two models are combined as a **Multi-factor model** in order to derive optimum weight schemes that neutralize market risks and maximize alpha returns. From that, investors/trader/portfolio managers can quantitatively design their own strategies & backtests that allows them quantitatively make up their further decision.



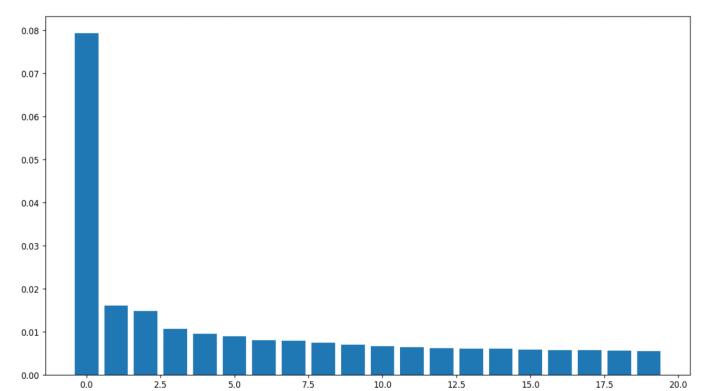
### **Model Construction – Risk Model**

PCA squeezes stock dataset from (2517x621) matrix to (621x20) components.

Total variance of data is explained by 20 components (factor\_betas) is about 23% which is low. It because of missing values from at least 30% of the universe has less than 5 year life long.

Top 5 stocks contribute the most in the variance of the first principal component Equity(572 [VIG]) 9.6%, Equity(237 [HUT]) 10.3%, Equity(420 [SCR]) 9.8%, Equity(443 [SHS]) 11.7%, Equity(355 [PFL]) 10.2%

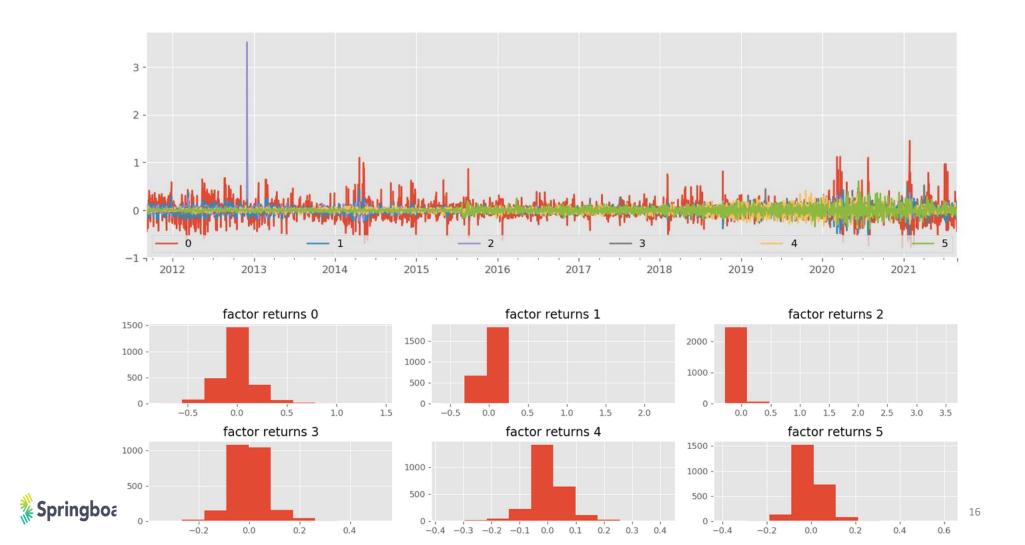
Sum square of stocks' contribution in each component approximates 1





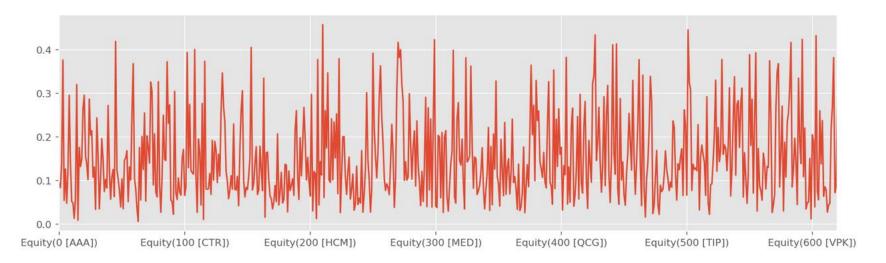
### **Model Construction – Risk Model**

**factor\_returns** is the returns time series for some kind of latent or unknown driver of return variance. The first 06 components fator\_returns



### **Model Construction – Risk Model**

**idiosyncratic variance vector** over 621 stocks, they are alpha risks that are related to each individual stocks, not the market.



Portfolio risk prediction for equal weights  $SQRT(X^T(BFB^T + S)X) = 10.52\%$ 



### **Model Construction – Alpha Model**

Alpha factors and risk factors can be interchangeable by time and markets. Once alpha becomes publicly known and available, its profitable power will depreciate and turn to be a risk factor and vice versa. Alpha model is built & assessed using **Quantopian** API.

Raw alpha factors are created by **Zipline** module then processed for sector neutralizing, ranking & zscore

Processed alpha factors are modeled & assess the performances by Alphalens module



**Model Construction – Alpha Model – Alpha Factors Processing** 

Raw alpha factors are processed for sector neutralization, ranking & zscore

10 alpha factors are computed for a 5 year period:

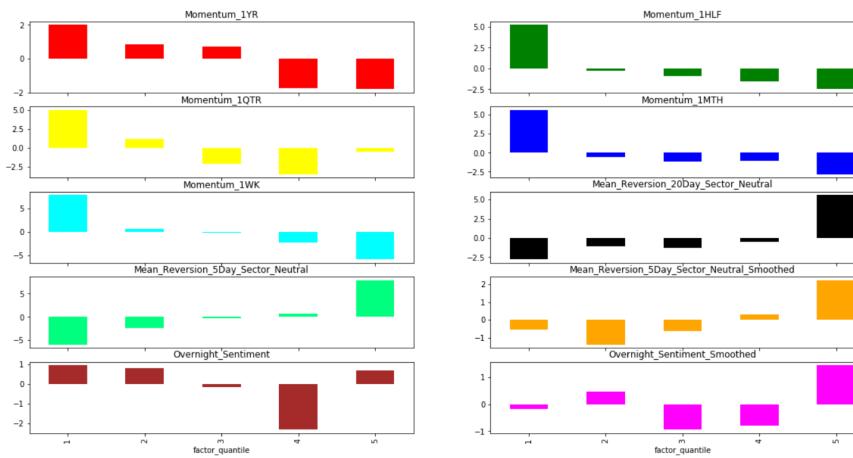
- Momentum 1 Year, 1 Half Year, 1 Quarter, 1 Month, 1 Week Factors
- Mean Reversion 5Days, 20Days Sector Neutral Factors
- Mean Reversion 5 Day Sector Neutral Smoothed Factor
- Overnight Sentiment Factor
- Overnight Sentiment Smoothed Factor



### **Model Construction – Alpha Model – Factor Operation**

Processed alpha factors & pricing data are used to compute 1 day forward returns & classify to 5 quantiles:

- Momentum returns performances seems working well in the 1st quantile
- Mean reversion & sentiment returns performances are good at the 5<sup>th</sup> quantile

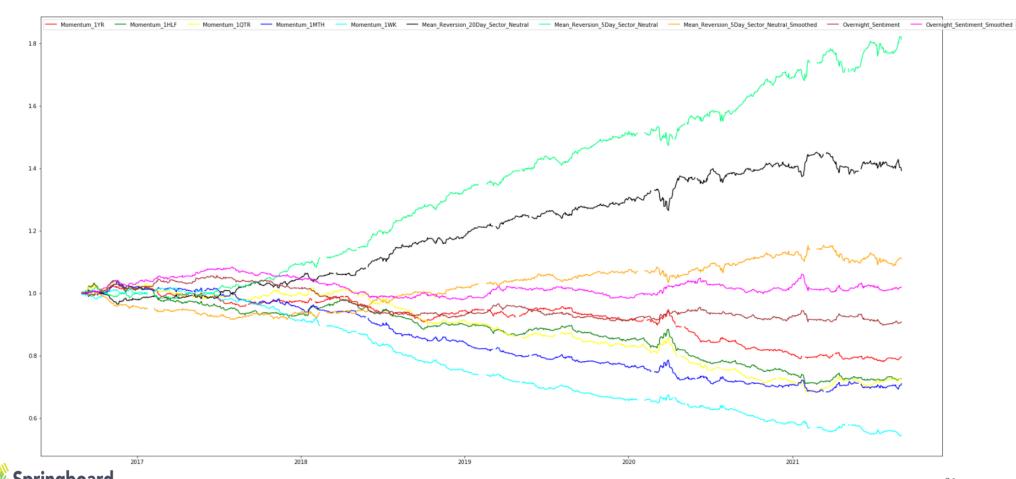




## **Model Construction – Alpha Model – Factor Performance**

Alpha factors cumulative returns for all quantiles.

- Momentum factors perform from weak to negative
- Sentiment factors show a fair to week performance
- Mean reversion factors perform the best



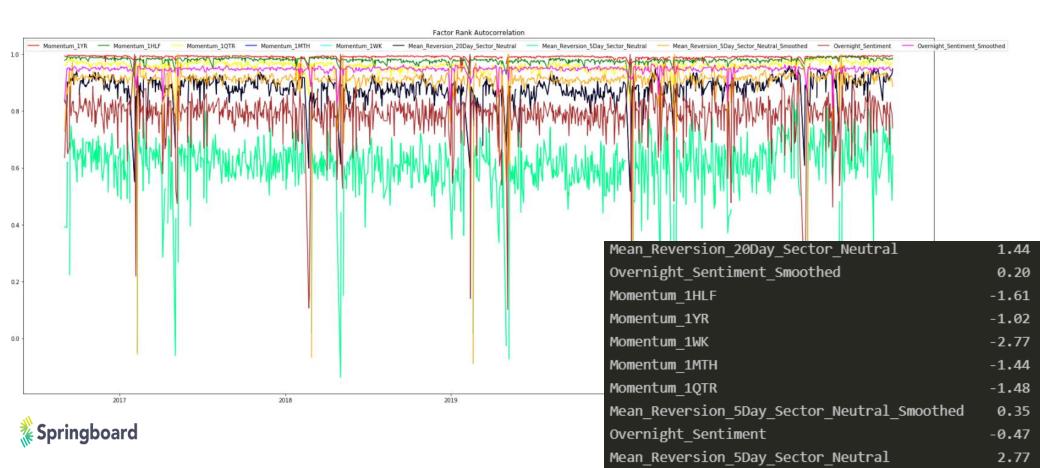
### **Model Construction – Alpha Model – Factor Assessment**

Turnover & Sharpe Ratio are two key indicators to assess alpha factors performances:

Turnover or Factor Rank Autocorrelation expresses how persistent the factors are for future prediction. The closer to 1 the more consistent performance of a factor is & less trading

- Momentum & smoothed factors are showing a stable status
- Mean reversion & sentiment factors are unstable

Sharp Ratio shows how well the factor making profit. Mean reversion factors are good candidates

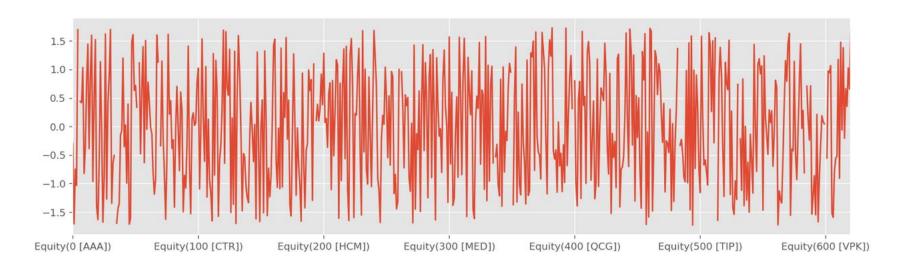


### **Model Construction – Alpha Model – Factor Assessment**

Regardless of low FRA, Mean\_Reversion\_5Day\_Sector\_Neutral & Mean\_Reversion\_20Day\_Sector\_Neutral Factors are still selected as input for Multi-Factor Model.

They are simply blended by averaging to the alphas. Below is the alpha vector of the last day 2021-08-31 of all 621 stocks that we think will be proportional to the next day returns

The higher alpha, the better arbitrage opportunities. Multi factor model need to preserve these alpha from market & idiosyncratic risks by weight optimizing process.





### **Model Construction – Multi Factor Model**

Alpha model and risk model are available. Let's find a portfolio that trades as close as possible to the alpha model but limiting the risks as measured by the risk model by optimizing the weights under risks constrain.

#### Multi factor model comprises objective function & its constraints

#### Where:

- Objective function is maximizing ( $\alpha^T * x$ ) with x is the portfolio weights and  $\alpha$  is the alpha vector.
- Constraints are risk model as following:

r is the portfolio risk, risk<sup>2</sup><sub>cap</sub> is market (vnindex/ETF...) volatility
B is the factor betas, factor<sub>max,min</sub> are max & min portfolio factor exposures
sum of weights = 0 (market neutral)
leverage (borrow cash/stocks) constraint <= 1

minimum and maximum limits on individual holdings, kind of insurance.

 $B^T * x \preceq factor_{ ext{max}} \ B^T * x \succeq factor_{ ext{min}} \ x^T 1 = 0 \ \|x\|_1 \leq 1 \ x \succeq weights_{ ext{min}} \ x \preceq weights_{ ext{max}}$ 

Output is optimum weights scheme for a period of time (1 day forward for this case). Negative weight is short position, positive is long position



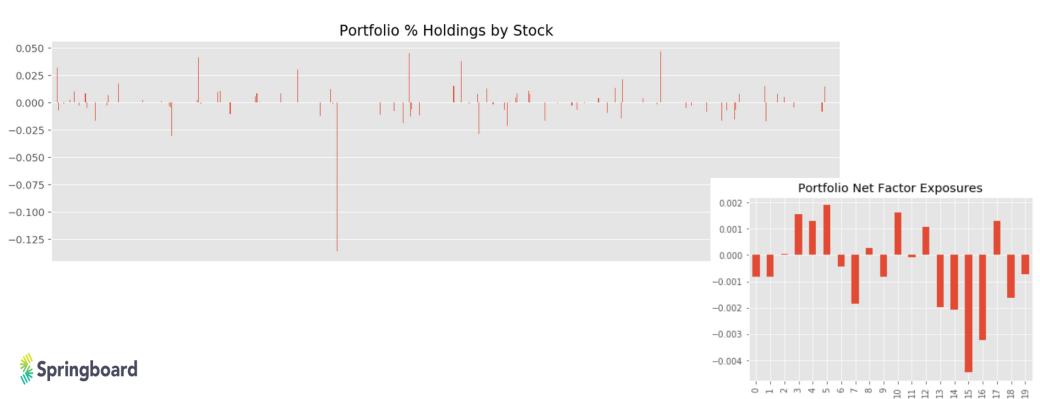
### **Model Construction – Multi Factor Model**

#### Optimizing object function $(\alpha^T * x)$

Default portfolio constraints: risk\_cap=0.05, factor\_max=10, factor\_min=-10, weights\_max=0.55, weights\_min=-0.55

- Weights allocation seems not diversity, not optimum for lowering risks
- Weights allocation for the case of portfolio contains 20 components

Portfolio risk prediction for optimum\_weights is SQRT(X<sup>T</sup>(BFB<sup>T</sup> + S)X) = 5.0% Transfer coefficient = 20.4%, optimum\_weights is limited match with alpha\_vector



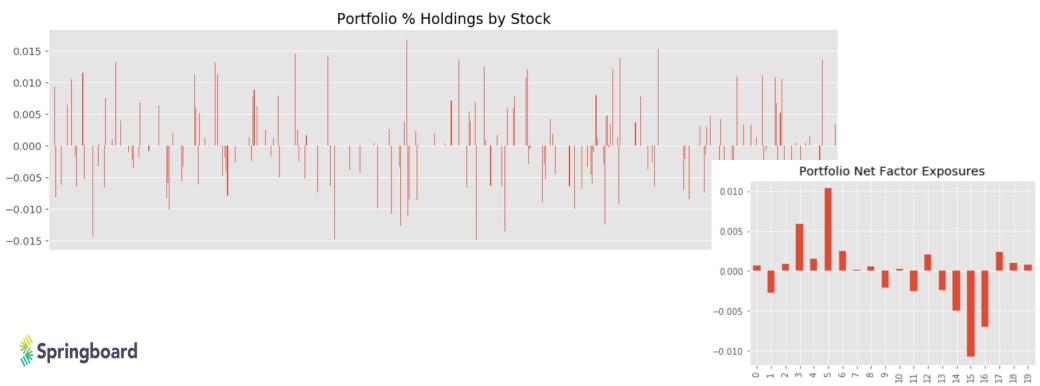
### **Model Construction – Multi Factor Model**

### Optimizing object function $(\alpha^T * x) + \lambda ||x||_2$

Regularization parameter penalize the extreme weights & re-attribute to the whole portfolio Default portfolio constraints:  $\lambda$  =0.5, risk\_cap=0.05, factor\_max=10.0, factor\_min=-10.0, weights\_min=-0.55

- Weights allocation is now pretty diversified
- Weights allocation for the case of portfolio contains 20 components

Portfolio risk prediction for optimum\_weights\_1 is SQRT(X<sup>T</sup>(BFB<sup>T</sup> + S)X) = 4.5% Transfer coefficient = 35.5%, optimum\_weights\_1 fairly matched with alpha\_vector



## **Model Construction – Multi Factor Model**

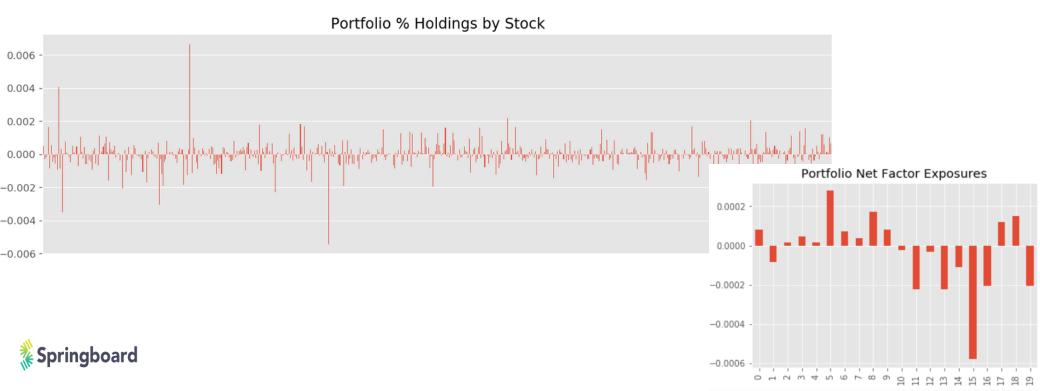
### Optimizing object function $-(\alpha^T * x) + ||x - x^*||_2$

Specific portfolio constraints: weights\_max=0.02, weights\_min=-0.02, risk\_cap=0.0015, factor\_max=0.015, factor\_min=-0.015

- Portfolio weights scheme is much more diversified
- Weights allocation for the case of portfolio contains 20 components

**Portfolio risk** prediction for optimum\_weights\_2 is  $SQRT(X^T(BFB^T + S)X) = 0.4\%$ , much lower than equal weight scheme (10.5%)

**Transfer coefficient = 31.3%,** optimum\_weights\_2 fairly matched with alpha\_vector



### **Conclusions**

#### **Project achievements:**

- Successfully ingest Vietnam stock data to Quantopian API format
- Conventional quant effect analysis: Bollinger band, breakout & momentum
- Fully create risk model for universe of 621 stocks
- Create 10 alpha factors model, operating & performance assessment: **Mean reversion factors outperform** for the universe of 621 stocks working for 1 day looking forward
- 03 cases of weight optimization under 02 risks constraints cases. The 3<sup>rd</sup> weights scheme reflects the best real trading condition.

#### **Discussion:**

- Predicted portfolio risk for non-managed case is 10.52% and fully managed case is 0.4% which is a significant market risk neutral
- Selected alpha factors have low turnover indicator(FRA) which is not good for future prediction
- Full-managed case's weights allocation is well diversified, but fair transfer coefficient (31.4%) which is a not good indicator of well-preserved alpha.



### Recommendations

#### for Retail investors, Fund managers, Portfolio managers:

Stock universe (candidate selection) should be carefully selected ex: filtered by

- size (volume),
- value (price),
- other fundamental factors : dividend, P/E...
- or even following ETF indexes: vnindex, vn30index...

#### Recommended further steps:

- create more alphas,
- apply weight scheme for alphas selected by sharpe ratio or other alpha performance indicators,
- back-testing under real trading condition in order to fully assess how effective, persistent & resilient the factor models are

