

# Principal Asset Management

**Project Lab – The University of Chicago**  
Equity Factor Construction and Analysis

*Leah Li*

*Matheus Raka Pradnyatama*

*Nicolas Donato*



# Deliverables

1. Factor Construction
2. Back-Testing Pipeline
  - a) Formula Parser in R
  - b) Point-in-Time Data Collector in Excel
3. Sensitivity Analysis on Factor Measures





# Factor Construction

# Factor Construction

- Using the paper titled “**Global Factor Data Documentation**” written by *Theis Ingerslev Jensen, Bryan Kelly, Lasse Heje Pedersen*
- The constructed factors are mapped into relevant Bloomberg fields
- This include:
  - Accounting variables: Gross Profit, Operating Income, Long-Term Debt, etc.
  - Market variables: Price, Return, Cash Dividend, Trading Volume, etc.
- Lag variables were also constructed manually for Momentum Factor, modeled with prices across monthly durations
- The constructed formula mappings are compiled into an excel file titled: “**Formulas**”
  - A previous version: “**Formula Mappings - Paper to Bloomberg FLDS**” was sent to Rajat Kathuria on April 16, 2025

# Factor Construction

Table	Type	Name	Abbreviation	Construction
Accounting	Growth - Percentage	Asset Growth 1yr	at gr1	$BS\_TOT\_ASSET / LAG(BS\_TOT\_ASSET, "12M") - 1$
Accounting	Growth - Percentage	Sales Growth 1yr	sale gr1	$SALES\_REV\_TURN / LAG(SALES\_REV\_TURN, "12M") - 1$
Accounting	Growth - Percentage	Sales Growth 3yr	sale gr3	$SALES\_REV\_TURN / LAG(SALES\_REV\_TURN, "36M") - 1$
Accounting	Growth - Percentage	Total Debt Growth 3yr	debt gr3	$SHORT\_AND\_LONG\_TERM\_DEBT / LAG(SHORT\_AND\_LONG\_TERM\_DEBT, "36M") - 1$
Accounting	Growth - Percentage	CAPX 1 year growth	capx gr1	$CAPITAL\_EXPEND / LAG(CAPITAL\_EXPEND, "12M") - 1$
Accounting	Growth - Percentage	CAPX 2 year growth	capx gr2	$CAPITAL\_EXPEND / LAG(CAPITAL\_EXPEND, "24M") - 1$
Accounting	Growth - Percentage	CAPX 3 year growth	capx gr3	$CAPITAL\_EXPEND / LAG(CAPITAL\_EXPEND, "36M") - 1$
Accounting	Growth - Percentage	Quarterly Sales Growth	sales gr1	$TRAIL\_12M\_NET\_SALES / LAG(TRAIL\_12M\_NET\_SALES, "12M") - 1$
Table 7: Market Variables	CRSP/Compustat	Share Adjustment Factor	shares*	$EQY\_SH\_OUT$
Table 7: Market Variables	CRSP/Compustat	Shares	prc*	$PX\_LAST$
Table 7: Market Variables	CRSP/Compustat	Price	prc.high	$MAX(PX\_LAST, PX\_HIGH)$
Table 7: Market Variables	CRSP/Compustat	Highest Daily Price	prc.low	$MIN(PX\_LOW, PX\_LAST)$
Table 7: Market Variables	CRSP/Compustat	Lowest Daily Price	me*	$PX\_LAST * EQY\_SH\_OUT$
Table 7: Market Variables	CRSP/Compustat	Market Equity		

Source: Formulas.xlsx

**Table/Type** - These columns determine the category of the factor. The factor can be from the accounting or market table, and there are various types within each.

**Name/Abbreviation** - These columns are the same factor names referenced in the Global Factor Data paper Documentation, Tables 5 - 8.

**Construction** - This column contains the formula to construct the factor using Bloomberg field names. This is what's processed in the R parser script to calculate the factors.



# Back-Testing Pipeline

Formula Parser in R

# Formula Parser R - Load Data

## Workflow Description

### Load Formula Definitions

- Read constructed formulas from Formulas.xlsx and recognizes variables

### Ingest & Clean Raw Data

- Iterate over each ticker, reading its “Accounting” and “Market” sheets
- Standardize dates, strip non-numeric characters, lag accounting metrics by 4 months
- Merge all tickers’ data into one long format ready for analysis

```
# =====  
# 1) Read Excel File for Formulas  
# =====  
path.dir      <- dirname(rstudioapi::getSourceEditorContext()$path)  
file_formulas <- file.path(path.dir, "Formulas.xlsx")  
formulas_dt   <- as.data.table(read_excel(file_formulas, sheet = "Formulas"))  
setnames(formulas_dt, old = "Construction", new = "Formula")  
  
# =====  
# 2) Read real data from Excel files (skip first row header; skip missing tickers)  
# =====  
data_dir      <- file.path(path.dir, "data")  
tickers       <- c("CIMB MK Equity", "GAM MK Equity")  
missing_tickers <- character(0)  
data_list     <- list() |  
for (ticker_full in tickers) {  
  ticker_code <- sub(".*", "", ticker_full)  
  acc_file    <- file.path(data_dir, paste0("Data_Accounting_", ticker_code, ".xlsx"))  
  mkt_file    <- file.path(data_dir, paste0("Data_Market_", ticker_code, ".xlsx"))  
}
```

Source: formula\_parser.r



# Formula Parser R - Factor Computation

## Formula Engine

- Balances syntax, cleans up formulas, and parses text into R expressions
- Utilizes helper functions built to accurately parse constructed formulas
- For each formula and ticker, builds a per-date data frame of inputs, then evaluates in a custom environment
- Automatically skips any formulas that error or yield no results

## Export Step

- Run all parsed formulas and output combined factor results to **computed\_factors.csv**

```
# =====  
# 3) Custom Functions  
# =====  
LAG    <- function(x, k) { ks <- as.character(k); n <- as.numeric(k);  
LEAD   <- function(x, k) { ks <- as.character(k); n <- as.numeric(k);  
STD     <- function(x, window, na.rm = TRUE) { ws <- as.character(window);  
CHG_TO_EXP <- function(x) { l12 <- dplyr::lag(x,12*21); l24 <- dplyr::lag(x,24*21);  
SUR     <- function(x, window = 12*21) { x3 <- dplyr::lag(x,3*21);  
MAX     <- function(x,y) pmax(x,y,na.rm=TRUE)  
MIN     <- function(x,y) pmin(x,y,na.rm=TRUE)  
CUMPROD<- function(x) cumprod(x)
```

Source: formula\_parser.r

```
83250 GAM MK Equity,2018-10-22T00:00:00Z,2.90135491621649,Total Assets scaled by Market Equity  
83251 GAM MK Equity,2018-10-23T00:00:00Z,2.86478321559192,Total Assets scaled by Market Equity  
83252 GAM MK Equity,2018-10-24T00:00:00Z,2.90135491621649,Total Assets scaled by Market Equity  
83253 GAM MK Equity,2018-10-25T00:00:00Z,2.93887243668481,Total Assets scaled by Market Equity  
83254 GAM MK Equity,2018-10-26T00:00:00Z,2.76039840206832,Total Assets scaled by Market Equity
```

Source: computed\_factors.csv



# Formula Parser R - Factor Computation

Feature	Functionality	Usefulness	Adaptability
<b>Time-Aware Functions (LAG, LEAD, STD)</b>	Recognizes flexible time units like “3M”, “1Y” and converts to trading days	Handles financial data frequency naturally	Adapts to various periodicity in factor design
<b>Nested functions</b>	Supports formulas with compound functions STD ( EXP ( LOG ())) including custom functions	Enables mathematically complex and compact expressions	No need to manually split expressions
<b>Formula Chaining</b>	Allows one formula to reference results from previous ones	Enables complex multi-step modeling	Easily supports derived formulas without recomputation
<b>Ticker-wise evaluation</b>	Processes each equity's data independently	Maintains clean separation across securities and records error	Scales well to different groups of tickers of interest



# Back-Testing Pipeline

Point-in-Time Data Collector in Excel

# Point-in-Time Data Collector – Description



- We collected Point-in-Time data using Excel
- We used the BQL.QUERY that can be enabled in Excel
- Syntax:  
= @BQL.QUERY("get(" & B\$2 & ") for(" & B\$1 & " & ") with(DATES=" & TEXT(\$A3,"yyyy-mm-dd") & ")")
- Users need only to change the ticker name for the company (cell B1) to the company they are interested in

	A	B	C
1	Company Name	AAPL US Equity	
2	Date	PX_LAST	PE_RATIO
3	12/13/17	43	19
4	12/14/17	43	19
5	12/15/17	43	19
6	12/16/17	#N/A	#N/A
7	12/17/17	#N/A	#N/A

# Point-in-Time Data Collector – Submitted Files

- We collected 2-years worth of point-in-time data for 3 companies: Apple US Equity (US Company), GAM MK Equity (Malaysian company), CIMB MK Equity (Malaysian company)
- The data collector is split between accounting and market variables to reduce the amount of data saved in one file
- For each company and variable type, there are two excel files:
  - The file containing the formula can be run on a computer connected to an active Bloomberg Terminal to observe how the BQL.QUERY formula functions in practice and to retrieve new data
  - The file that doesn't contain the formulas contains only the data extracted in the past

# Point-in-Time Data Collector – Submitted Files

 Data\_Market\_Apple\_with\_Formula.xlsx 

 Data\_Market\_Apple.xlsx 

 Data\_Market\_CIMB\_with\_Formula.xlsx 

 Data\_Market\_CIMB.xlsx 

 Data\_Market\_GAM\_with\_Formula.xlsx 

 Data\_Market\_GAM.xlsx 



 Data\_Accounting\_Apple\_with\_Formula.xlsx 

 Data\_Accounting\_Apple.xlsx 

 Data\_Accounting\_CIMB\_with\_Formula.xlsx 

 Data\_Accounting\_CIMB.xlsx 

 Data\_Accounting\_GAM\_with\_Formula.xlsx 

 Data\_Accounting\_GAM.xlsx 

# Point-in-Time Data Collector – Submitted Files

	A	B	C	D	E	F	G	H	I	J
1	Company Name	AAPL US Equity								
2	Date	PX_LAST	PE_RATIO	NET_INCOME	PX_HIGH	PX_LOW	PX_VOLUME	RETURN_COM_EQY	CUR_MKT_CAP	EQY_DPS
3	12/13/17	43	19	48,351,000,000	43.385	43	95,273,788	36.86750846	884,487,928,240	#N/A Invalid Parameter: Eri
4	12/14/17	43	19	48,351,000,000	43.2825	42.9125	81,906,164	36.86750846	884,231,212,640	#N/A Invalid Parameter: Eri
5	12/15/17	43	19	48,351,000,000	43.5425	43.115	160,677,228	36.86750846	893,216,258,640	#N/A Invalid Parameter: Eri
6	12/16/17	#N/A	#N/A	48,351,000,000	#N/A	#N/A	#N/A	36.86750846	#N/A	#N/A Invalid Parameter: Eri
7	12/17/17	#N/A	#N/A	48,351,000,000	#N/A	#N/A	#N/A	36.86750846	#N/A	#N/A Invalid Parameter: Eri
8	12/18/17	44	19	48,351,000,000	44.3	43.715	117,684,456	36.86750846	905,795,323,040	#N/A Invalid Parameter: Eri
9	12/19/17	44	19	48,351,000,000	43.8475	43.5225	109,745,788	36.86750846	896,142,816,480	#N/A Invalid Parameter: Eri
10	12/20/17	44	19	48,351,000,000	43.855	43.3125	93,902,596	36.86750846	895,167,297,200	#N/A Invalid Parameter: Eri
11	12/21/17	44	19	48,351,000,000	44.005	43.525	83,799,584	36.86750846	898,555,943,120	#N/A Invalid Parameter: Eri
12	12/22/17	44	19	48,351,000,000	43.856	43.625	65,397,776	36.86750846	898,555,943,120	#N/A Invalid Parameter: Eri
13	12/23/17	#N/A	#N/A	48,351,000,000	#N/A	#N/A	#N/A	36.86750846	#N/A	#N/A Invalid Parameter: Eri
14	12/24/17	#N/A	#N/A	48,351,000,000	#N/A	#N/A	#N/A	36.86750846	#N/A	#N/A Invalid Parameter: Eri

- #N/A:
  - Data is not available during those dates (e.g., not a business day)
- #N/A Invalid Parameter: Error encountered while validating *field\_name*
  - The company does not have any data on that specific variable



# Sensitivity Analysis



# Sensitivity Analysis

## Data Inputs

- Computed Factors: loads a pre-computed CSV of factor values (one row per ticker-date-factor).
- Market Prices: reads raw Excel price files for each ticker, cleans them, and computes forward returns.

## Parameter Controls

- Date Range: restrict the analysis window
- Ticker: pick which instrument to study
- Factor: select which pre-computed metric (e.g. leverage, valuation)
- Horizon: choose the look-ahead period (1 day, week, month, quarter, year)

## Forward Return Calculation

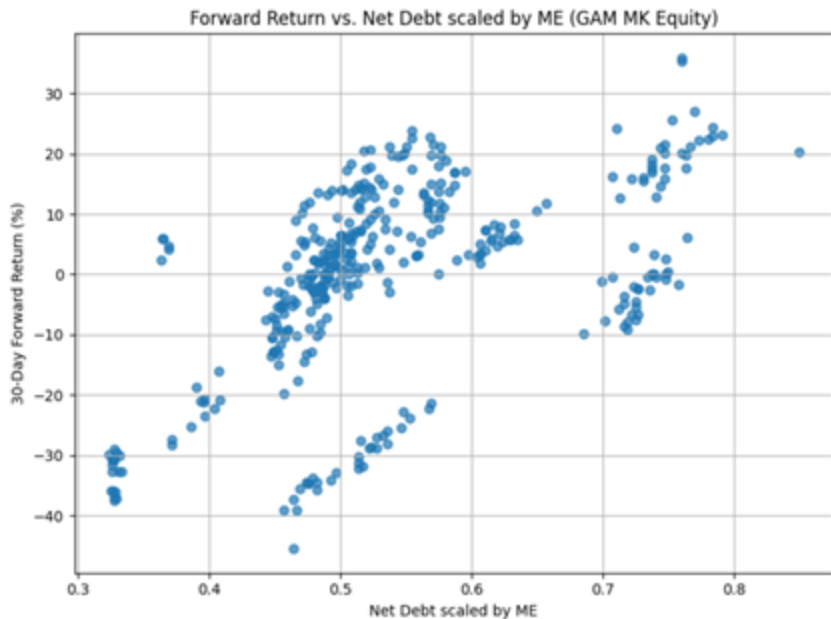
- Forward return measures how much the price increases or decreases after a chosen number of trading days. It reflects the future price movement from a given starting point.

## Visualization

- The app creates a scatter plot showing the relationship between the selected factor and the forward return. Each point represents a historical observation for the selected ticker and date range.

# Sensitivity Analysis

## Jupyter Notebook



Source: sensitivity\_analysis\_notebook.ipynb

## Python Application UI

Parameters

Start Date  
2017/12/13

End Date  
2020/04/17

Ticker  
GAM MK Equity

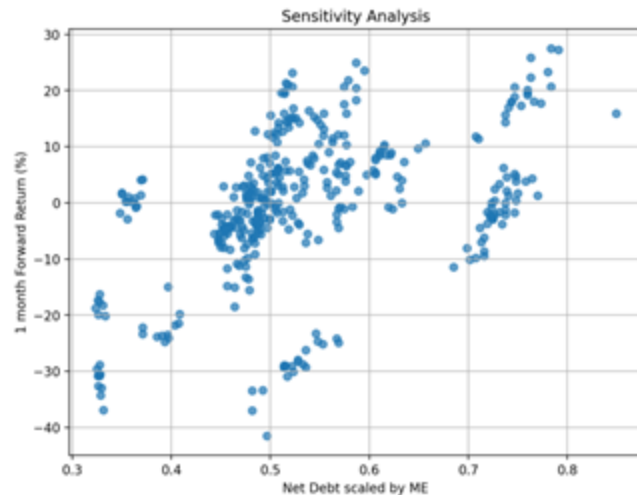
Factor  
Net Debt scaled by ME

Horizon  
1 month

Analyze

Source: sensitivity\_analysis\_UI.py

## Sensitivity Analysis



# Sensitivity Analysis

## Note:

- Sensitivity analysis is available in both the **sensitivity\_analysis\_notebook.ipynb**, for a python notebook version, and **sensitivity\_analysis\_UI.py**, for a clean browser-UI program.
- To run the **sensitivity\_analysis\_UI.py** version
  - Make sure streamlit is installed (“pip install streamlit”)
  - Run the program via terminal using “streamlit run sensitivity\_analysis\_UI.py”
  - The UI will then open in your browser
- Ensure the following data files are in the directory of the sensitivity analysis program:
  - `computed_factors.csv` - For all the computed factors calculated from `formula_parser.r`
  - `Data_Market_{ticker}.xlsx` - Which contains Bloomberg PX\_LAST data for the specified ticker



# Challenges

# Challenges – Data Collection

- There are two alternative methods for data collection that may prove to be more effective
- **EE (Earnings Estimates)** and **Earnings Summary (ERN)** functions in BBG
  - We needed to learn this with a BBG representative
  - The Live Help feature is disabled in the Finmath student BBG subscription
  - We can only communicate with BBG with a one-business-day delay
- **Bloomberg BLP API in Python**
  - Succeeded in implementing blp/refdata using Python script, but blp/bql is disabled in the Finmath student BBG subscription on campus
  - We could only collect historical instead of point-in-time data as needed
- **Conclusion:** We used the **BQL.QUERY** directly in **Excel**

# Challenges – Data

- There is often missing data from specific fields for particular tickers
- Accounting variables and market variables are recorded at different cadences (daily vs quarterly for example)
- Further consideration is needed for how to deal with accounting variables in the sensitivity analysis, perhaps using only quarterly data for accounting factors
- Simplification assumptions were made for market factors that did not have direct Bloomberg fields. Replication of those factors were done using basic variables.
- More complex market factors can be added to list of constructed variables with further attention to different time lags

# Thank you

