## SPL: Fama French Factor Model

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### Structure

- 1. Model
- 2. Data
- 3. Results

#### 1. Model

- ightharpoonup CAPM  $R_i R_F = \beta \cdot (R_M R_F)$
- ► Fama and French (1993) 3 Factors  $R_i R_F = \beta_M \cdot (R_M R_F) + \beta_S \cdot SMB + \beta_V \cdot HML$
- ► Fama and French (2015) 5 Factors  $R_i R_F = \beta_M \cdot (R_M R_F) + \beta_S \cdot SMB + \beta_V \cdot HML + \beta_P \cdot RMW + \beta_I \cdot CMA$

#### 2. Data

- ► Model factors including the estimated market return and the risk-free rate from Kenneth French's data library: http://mba.tuck.dartmouth.edu/pages/faculty/ken.french/data\_library.html
- ► S&P 500 stock prices downloaded with BatchGetSymbols library.

#### 2. Data: S&P 500

- GetSP500Stocks() returns company info like name and sector.
- Downloaded stock data contains two dataframes: df.control descriptive info. e.g. whether the download for the ticker is successful.
  - **df.tickers** downloaded price data, stacked in rows.
- ▶ BatchGetSymbols() function **keeps** stocks with 75% or more price points compared to the benchmark.
  - We took 3M Company's dates as they are complete.

```
good.tickers <- Stocks$df.control$
    ticker[Stocks$df.control$threshold.decision=="KEEP"]

SP500.data<-data.frame(date = Stocks$df.tickers$
    ref.date[Stocks$df.tickers$ticker=="MMM"])</pre>
```

#### 2. Data: S&P 500

- Loop over the downloaded data **df.tickers**.
- Subset each company by its ticker.
- ▶ Rename colname of the price series to the ticker symbol.
- Merge to the existing dataframe by matching dates.

```
for(i in 1:length(good.tickers))
  X <- data.frame(date =</pre>
    Stocks$df.tickers$
      ref.date[Stocks$df.tickers$ticker==good.tickers[i]],
    Stocks$df tickers$
      price.adjusted[Stocks$df.tickers$ticker
                                 ==good.tickers[i]])
  colnames(X)[2] <- Stocks$df.tickers$</pre>
        ticker[Stocks$df.tickers$ticker==good.tickers[i]]
  SP500.data <- merge.data.frame(SP500.data, X,
                                 by = "date", all.x = TRUE)
```

# 2. Data: from Daily Price to Monthly Return

quantmod::monthlyReturn() requires non-NA daily prices in xts format.

```
Stock.Prices.Daily <-
    xts(Stock.Prices.Daily[,-1],
        order.by = as.POSIXct(Stock.Prices.Daily$date))

Stock.Prices.Daily <-
    Stock.Prices.Daily[!is.na(Stock.Prices.Daily)]

Stock.Prices.Monthly <- monthlyReturn(Stock.Prices.Daily)</pre>
```

How to chose?

- ▶ Remove stocks with NA ensures remaining stocks have same number of observations.
- ▶ Remove NAs in each series results in a larger sample size.

Price data with NAs in the middle might result in strange monthly returns. ("BHY" Brighthouse Financial Inc. removed for 2015-2017 runs)

## 2. Data: Linear Regression

Linear Regression in R

```
y <- lm(rirf ~ rmrf + smb + hml);
round(summary(y)$coefficients, digits = 4)</pre>
```

- We can read out the regression summary from the coefficients matrix.
- Similarly R<sup>2</sup> from summary(y)\$r.squared and standard errors from summary(y)\$sigma.

#### 3. Results

#### Overview:

- ▶ Replicating Fama and French (1993)
- ► Regression over S&P500 stocks
- ▶ 3 Factors vs. 5 Factors
- ► Stock selection?

# 3.1 Results: Replicating the 3 Factors Model

b	LOW	2	3	4	HIGH		t(b)	LOW	2	3	4	HIGH
SMALL	1.03	0.97	0.94	0.89	0.95		SMALL	39.23	50.60	58.42	57.99	57.76
2	1.10	1.02	0.96	0.97	1.07		2	53.20	58.56	59.98	62.77	63.25
3	1.10	1.02	0.97	0.97	1.06		3	59.68	56.81	53.35	58.93	51.14
4	1.06	1.07	1.04	1.03	1.15		4	57.16	52.61	50.34	51.30	46.30
BIG	0.96	1.02	0.96	1.01	1.03		BIG	57.20	56.98	42.80	55.04	37.70
5	LOW	2	3	4	HIGH	_	t(s)	LOW	2	3	4	HIGH
SMALL	1.40	1.27	1.16	1.10	1.19		SMALL	35.61	44.82	48.65	48.10	48.63
2	1.00	0.94	0.83	0.71	0.85		2	32.62	36.36	34.80	30.78	33.82
3	0.70	0.63	0.54	0.45	0.65		3	25.53	23.41	20.04	18.46	21.03
4	0.30	0.27	0.25	0.22	0.36		4	10.92	8.75	8.06	7.49	9.64
BIG	(0.20)	(0.19)	(0.27)	(0.19)	(0.04)		BIG	(8.10)	(7.08)	(7.99)	(6.91)	(1.05)
h	LOW	2	3	4	HIGH	_	t(h)	LOW	2	3	4	HIGH
SMALL	(0.30)	0.08	0.27	0.38	0.62		SMALL	(6.77)	2.43	9.92	14.93	22.43
2	(0.48)	0.03	0.23	0.47	0.70		2	(13.93)	0.88	8.73	18.32	24.74
3	(0.43)	0.04	0.31	0.50	0.71		3	(14.04)	1.39	10.27	18.28	20.34
4	(0.44)	0.03	0.30	0.56	0.74		4	(14.24)	0.79	8.77	16.68	17.79
BIG	(0.44)	(0.02)	0.20	0.56	0.76		BIG	(15.96)	(0.68)	5.25	18.41	16.65

16.24

18.34

## 3.1 Results: Replicating the 3 Factors Model

#### Results from Fama and French (1993):

0.21

0.00

-0.46

Table 6

Regressions of excess stock and bond returns (in percent) on the excess market return (RM-RF) and the mimicking returns for the size (SMB) and bookto-market equity (HML) factors: July 1963 to December 1991, 342 months.\*

R(t) - RF(t) = a + b[RM(t) - RF(t)] + sSMB(t) + hHML(t) + e(t)

Dependent variable: Excess returns on 25 stock portfolios formed on size and book-to-market equity

Book-to-market equity (BE/ME) quintiles Size High auintile Low Low High b 1(b) Small 1.04 1.02 0.95 0.91 0.96 39.37 51.80 60.44 59.73 57.89 52.49 61.18 55.88 61.54 65.52 1.11 1.06 1.00 0.97 1.09 1.12 53.17 50.78 52.52 1.02 0.98 0.97 1.09 56.88 54 38 1.07 1.08 1.04 1.05 1 18 53 94 53 51 51.21 47 09 46 10 0.96 1.02 0.98 0.99 1.06 60.93 56.76 46.57 53.87 38.61 1(s) Small 1.46 1.26 1.19 1.17 1.23 37.92 44 11 52 03 52.85 50.97 0.98 1.00 0.88 0.73 0.89 32.73 38.79 34.03 31.66 36.78 0.76 26.40 23.39 21.23 0.65 0.60 0.480.66 18.62 21.91 0.37 0.33 0.29 0.24 0.41 12.73 11.11 9.81 7.38 11.01 Big -0.17-0.12-0.23-7.18-7.58-6.27-1.18-0.17-0.05-4.51h t(h)-0.290.26 22.24 Small 0.08 0.400.62 -6.472.35 9.66 15.53 - 0.52 0.26 0.70 -- 14.57 8.56 17.24 24.80 0.01 0.46 0.41-0.38-11.269.75 . 0.00 0.32 0.51 0.68 0.05 16.88 19.39 -0.420.04 0.30 0.56 0.74 - 12.51 1.04 8.83 14.84 17.09

0.76

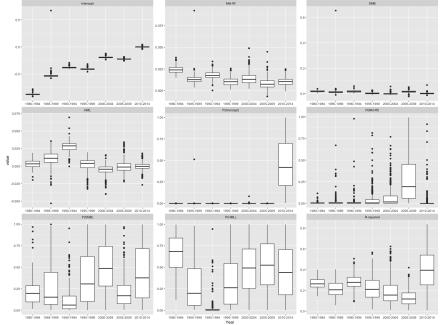
-17.03

0.57

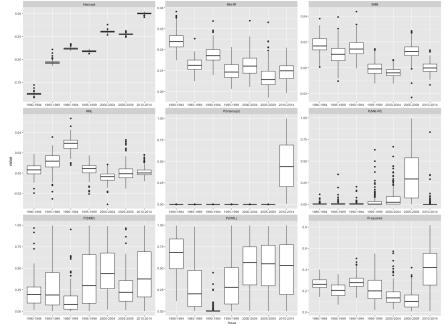
5.80

0.09

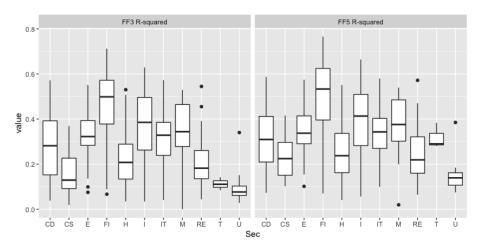
# 3.2 Regression over S&P 500 stocks (All)



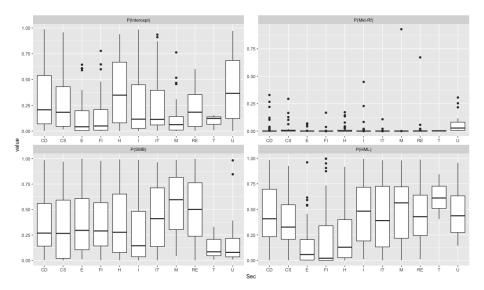
# 3.2 Regression over S&P 500 stocks (Survivors)



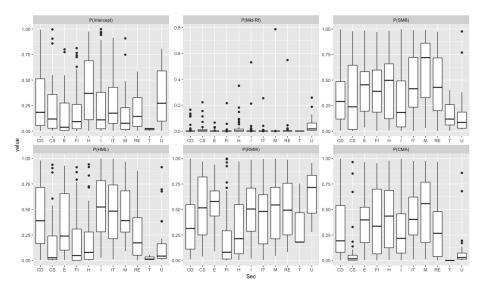
#### ► R<sup>2</sup> Comparison



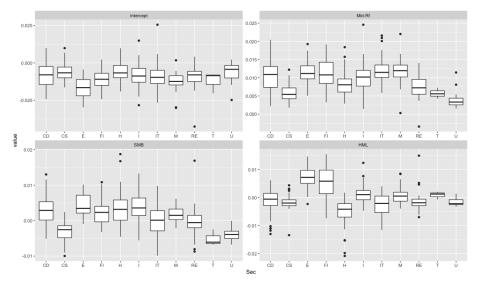
▶ P-Values: 3 Factors



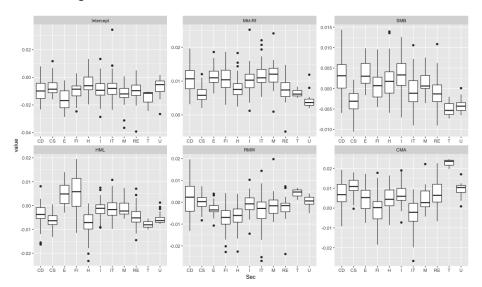
► P-Values: 5 Factors



► Regressed Coefficients: 3 Factors

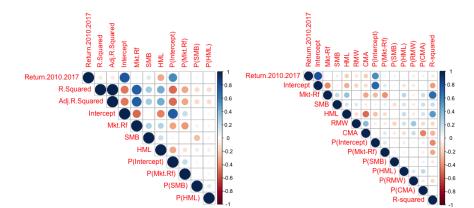


► Regressed Coefficients: 5 Factors



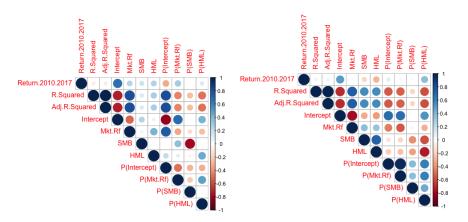
#### 3.4 Stock Selection?

- Stock returns from Jan. 2010 to Dec. 2017.
  Corr. plot of the stock returns and the regressed coefficients.
- ▶ alpha or Intercept highly correlated to stock returns.
- Positive corr. to SMB: Small cap premium. Negative corr. to HML: Market seemed to favor **Growth** stocks (instead of **Value**) in the bullish years.
- Negative corr. to RMW and CMA.



# 3.4 Stock Selection: 3 Factor / Top and Bottom 20s

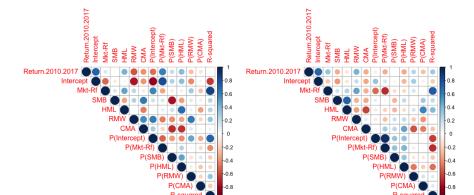
- ▶ Winners have positive correlations to both *SMB* and *HML*, while losers have negative correlations.
- Small Cap and Value premiums work at the extremes.



Although the overall winner, *NETFLIX*, is clearly neither *small cap* nor *value*.

## 3.4 Stock Selection: 5 Factor / Top and Bottom 20s

- ▶ **Profitability** factor *RMW*: negatively corr. to returns of winners and positively corr. to that of losers: might not be good indicator for stock selection.
  - Negative regressed beta for *RMW* for certain sectors: companies might have poor near-term operating performance but very high future potentials, which is *priced-in* to the current stock prices.
- ▶ **Investment** factor *CMA* is negatively correlated to the returns of both winners and losers: both winners and losers invest aggressively?



Thank You!

Below are backup slides

# Melt and Plotting

- Regression results in dataframe Results.
- reshape::melt() prepares the data for plots.
- Each Fama French factor will have 1 sub-plot comparing different sectors

```
df.melt <- melt(
  Results[,c("Intercept", "Mkt-Rf", "SMB", "HML", "RMW", "CMA", "Sec")],
  "Sec"
)</pre>
```

► Boxplot with the melted data

## Helper functions

Resizing the regression output from an 1 dimensional vector to the same format as in the papers with rows for SMB and columns for HML.

## Descriptions copied from K.French's Website

Description of the factors: http://mba.tuck.dartmouth.edu/pages/faculty/ken.french/Data\_Library/f-f\_5\_factors\_2x3.html

- ▶ SMB (Small Minus Big) is the average return on the nine small stock portfolios minus the average return on the nine big stock portfolios.
- ► HML (High Minus Low) is the average return on the two value portfolios minus the average return on the two growth portfolios.
- RMW (Robust Minus Weak) is the average return on the two robust operating profitability portfolios minus the average return on the two weak operating profitability portfolios.
- ► MA (Conservative Minus Aggressive) is the average return on the two conservative investment portfolios minus the average return on the two aggressive investment portfolios.

# Descriptions copied from K.French's Website (cont.)

Rm-Rf, the excess return on the market, value-weight return of all CRSP firms incorporated in the US and listed on the NYSE, AMEX, or NASDAQ that have a CRSP share code of 10 or 11 at the beginning of month t, good shares and price data at the beginning of t, and good return data for t minus the one-month Treasury bill rate (from lbbotson Associates).

#### References

Fama, Eugene F., and Kenneth R. French. 1993. "Common risk factors in the returns on stocks and bonds." *Journal of Financial Economics* 33 (1): 3–56. doi:10.1016/0304-405X(93)90023-5.

———. 2015. "A five-factor asset pricing model." *Journal of Financial Economics* 116 (1). Elsevier: 1–22. doi:10.1016/j.jfineco.2014.10.010.