# Big Data Analysis

Lecture 14

2020/1/6

## Time Series Regression

- Factors are portfolios returns, Fama and French (1993)
- Stocks are sorted and grouped to form long short portfolios
- With periodic rebalance, portfolio returns are achieved
- Let  $f_t$  be factor returns during t period,  $R_{it} = \alpha_i + \beta_i f_t + \varepsilon_{it}, t = 1, 2, ..., T$
- Taking expectation:  $E_T[R_i] = \beta_i E_T[f_t] + \alpha_i$
- Comparing to  $E[R_i] = \beta_i \lambda + \alpha_i$ , we have  $\lambda = E_T[f_t]$

## Time Series Regression - Momentum

- Monthly Momentum Factor
- Total 6 portfolios formed monthly on Size and Momentum
- https://mba.tuck.dartmouth.edu/pages/f aculty/ken.french/Data\_Library/det\_mo m\_factor.html
- Mom = 1/2(Small High + Big High) 1/2(Small Low + Big Low)
- Why creating 2x3 portfolios?
- Can we limit the stock to be within a certain industrial sector?

Monthly Returns: January 1927 - November 2019

**Annual Returns:** 1927 - 2018

**Construction:** The portfolios, which are constructed monthly, are the

intersections of 2 portfolios formed on size (market equity, ME) and 3 portfolios formed on prior (2-12) return. The monthly size breakpoint is the median NYSE market equity. The monthly prior (2-12) return breakpoints are the 30<sup>th</sup> and 70<sup>th</sup>

NYSE percentiles.

	Median ME	
70th prior (2-12) percentile - 30th prior (2-12) percentile -	Small Up	Big Up
	Small Medium	Big Medium
	Small Down	Big Down

Stocks: The six portfolios constructed each month include NYSE,

AMEX, and NASDAQ stocks with prior return data. To be included in a portfolio for month t (formed at the end of month t-1), a stock must have a price for the end of month t-13 and a good return for t-2. In addition, any missing returns from t-12 to

t-3 must be -99.0, CRSP's code for a missing price. Each included stock also must have ME for the end of month t-1.

## Cross Sectional Regression

- Factors don't have to be portfolios returns
- Two step regression estimate
- Beta estimation

$$R_{it} = a_i + \beta_i f_t + \varepsilon_{it}, t = 1, 2, \dots, T$$

- Average stock returns to estimate factor return using GLS  $E_T[R_i] = \beta_i \lambda + \alpha_i, i = 1, 2, ..., N$
- Note that  $\beta_i$  are estimated, therefore the covariance matrix of  $\alpha_i$  are not easy to estimate. GMM is needed to test  $\alpha_i$

### Information Coefficient

- Following Barra models,  $\beta_i$  represents characteristics of individual stocks (normalized)
- Let  $\beta_t$  denote the factor exposure vector of stocks at time t, and  $y_{t,t+1}$  denote the stock returns from time t to t+1, we have

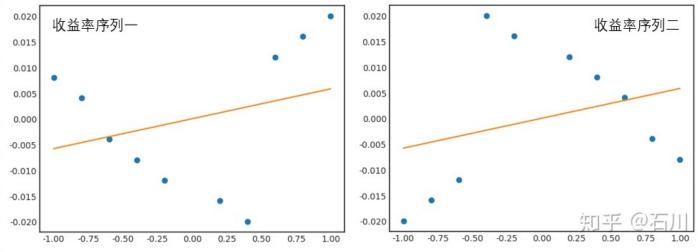
$$IC_t = corr(\beta_t, y_{t,t+1})$$



### Information Coefficient Check

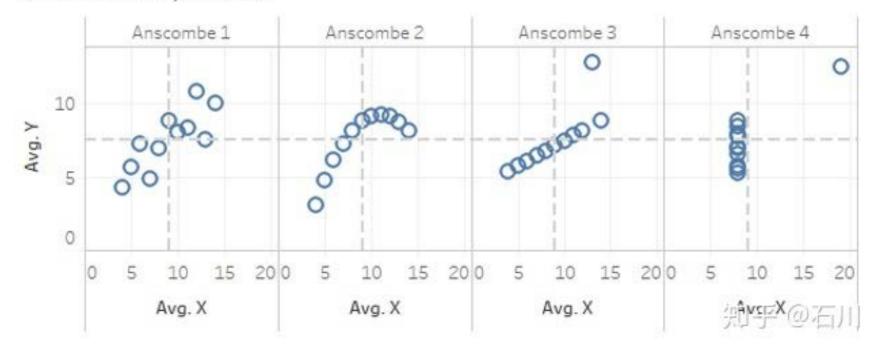
因子取值	收益率序列一	收益率序列二
1.0	2.00 %	-0.80 %
0.8	1.60 %	-0.40 %
0.6	1.20 %	0.40 %
0.4	-2.00 %	0.80 %
0.2	-1.60 %	1.20 %
-0.2	-1.20 %	1.60 %
-0.4	-0.80 %	2.00 %
-0.6	-0.40 %	-1.20 %
-0.8	0.40 %	-1.60 %
-1.0	0.80 %	-2.00;95字 @石川

Both return time series have the same IC equal to 0.29



### Statistics Characteristics Are Summaries

#### Anscombe quartet



### Statistics Characteristics Are Summaries

第一	一组	第二	二组	第三	三组	第四	四组
X	У	Х	У	Х	у	Х	У
10.0	8.04	10.0	9.14	10.0	7.46	8.0	6.58
8.0	6.95	8.0	8.14	8.0	6.77	8.0	5.76
13.0	7.58	13.0	8.74	13.0	12.74	8.0	7.71
9.0	8.81	9.0	8.77	9.0	7.11	8.0	8.84
11.0	8.33	11.0	9.26	11.0	7.81	8.0	8.47
14.0	9.96	14.0	8.10	14.0	8.84	8.0	7.04
6.0	7.24	6.0	6.13	6.0	6.08	8.0	5.25
4.0	4.26	4.0	3.10	4.0	5.39	19.0	12.50
12.0	10.84	12.0	9.13	12.0	8.15	8.0	5.56
7.0	4.82	7.0	7.26	7.0	6.42	8.0	7.91
5.0	5.68	5.0	4.74	5.0	5.73	8.0	6.89

性质	数值
x 的均值	9
x 的方差	11
y 的均值	7.50,精确到小数点后两位
y 的方差	4.125 ± 0.003
x 与 y 之间的相 关系数	0.816,精确到小数点后三位
线性回归线	y = 3.00+0.500x 分别精确到小数点后两位和三 位
线性回归 R-squared	0.67,精确到了数点后两位

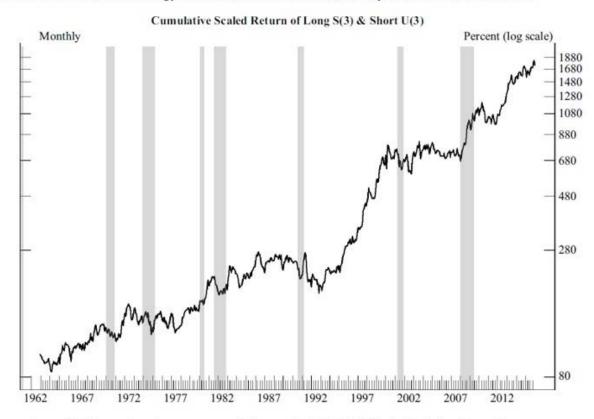
Matejka, J. and G. Fitzmaurice (2017). Same Stats, Different Graphs: Generating Datasets with Varied Appearance and Identical Statistics through Simulated Annealing. CHI 2017 Conference proceedings: ACM SIGCHI Conference on Human Factors in Computing Systems.

## Backtesting is Everything?

EXHIBIT 1

Long-Short Market-Neutral Strategy Based on NYSE Stocks, January 1963 to December 2015

- 1. The same simple factor construction method: S(3) U(3)
- 2. Very low correlation with most well known factors
- 3. Only 10% turnover annually



Notes: Gray areas denote NBER recessions. Strategy returns scaled to match S&P 500 T-bill volatility during this period.

Source: Campbell Harvey, using data from CRSP.

出处: Arnott, Harvey, and Markowitz (2019)

### Research Protocol – Part I Arnott, Harvey, and Markowitz (2019)

#### EXHIBIT 2

#### Seven-Point Protocol for Research in Quantitative Finance

#### 1. Research Motivation

- a. Does the model have a solid economic foundation?
- b. Did the economic foundation or hypothesis exist before the research was conducted?

#### 2. Multiple Testing and Statistical Methods

- a. Did the researcher keep track of all models and variables that were tried (both successful and unsuccessful), and are the researchers aware of the multiple-testing issue?
- b. Is there a full accounting of all possible interaction variables if interaction variables are used?
- c. Did the researchers investigate all variables set out in the research agenda, or did they cut the research as soon as they found a good model?

#### 3. Data and Sample Choice

- a. Do the data chosen for examination make sense? And, if other data are available, is it reasonable to exclude these data?
- b. Did the researchers take steps to ensure the integrity of the data?
- c. Do the data transformations, such as scaling, make sense? Were they selected in advance? And are the results robust to minor changes in these transformations?
- d. If outliers are excluded, are the exclusion rules reasonable?
- e. If the data are winsorized, was there a good reason to do it? Was the winsorization rule chosen before the research was started? Was only one winsorization rule tried (as opposed to many)?

#### 4. Cross-Validation

- a. Are the researchers aware that true out-of-sample tests are only possible in live trading?
- b. Are steps in place to eliminate the risk of out-of-sample iterations (i.e., an in-sample model that is later modified to fit out-of-sample data)?
- c. Is the out-of-sample analysis representative of live trading? For example, are trading costs and data revisions taken into account?

### Research Protocol – Part II Arnott, Harvey, and Markowitz (2019)

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#### 5. Model Dynamics

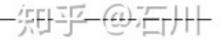
- a. Is the model resilient to structural change, and have the researchers taken steps to minimize the overfitting of the model dynamics?
- b. Does the analysis take into account the risk/likelihood of overcrowding in live trading?
- c. Do researchers take steps to minimize the tweaking of a live model?

#### 6. Complexity

- a. Is the model designed to minimize the curse of dimensionality?
- b. Have the researchers taken steps to produce the simplest practicable model specification?
- c. Has an attempt been made to interpret the predictions of the machine learning model rather than using it as a black box?

#### 7. Research Culture

- a. Does the research culture reward the quality of the science rather than the finding of a winning strategy?
- b. Do the researchers and management understand that most tests will fail?
- c. Are expectations clear (that researchers should seek the truth, not just something that works) when research is delegated?



## Further Readings – Journals

- Academic: JF, JFE, RFS
- Journal of Portfolio Management
- Journal of Financial Data Science

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## Further Readings - Online

- <a href="https://www.zhihu.com/people/mitcshi/activities">https://www.zhihu.com/people/mitcshi/activities</a>
- https://www.zhihu.com/people/Join Quant/activities
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# Thank you!