

Market and Non-Market Return

- $R_{i,t}$ and $R_{m,t}$: returns in excess of cash ($R_{i,t} = r_{i,t} - r_{f,t}$, $R_{m,t} = r_{m,t} - r_{f,t}$)
- “Characteristic line” regression of asset i ’s excess return on the market excess return:

$$R_{i,t} = \underbrace{\alpha_i + \beta_i R_{m,t}}_{\text{regression line}} + \underbrace{\epsilon_{i,t}}_{\text{residual}} \quad (1)$$

- Random: $R_{i,t}$, $R_{m,t}$, $\epsilon_{i,t}$; constant: α_i , β_i
- By construction,

$$\text{cov}(\epsilon_{i,t}, R_{m,t}) = 0 \quad \text{and} \quad E(\epsilon_{i,t}) = 0 \quad (2)$$

- Return decomposition:

- $\beta_i R_{m,t}$: market-related *realized* return
- $\epsilon_{i,t}$: non-market-related *realized* return
- α_i : part of *expected* return

$$\alpha_i = E(R_{i,t}) - \beta_i E(R_{m,t}) \quad (3)$$

Market and Non-Market Risk

- Variance of a sum: x and y random; a and b constants

$$\text{Var}(ax + by) = a^2\text{Var}(x) + b^2\text{Var}(y) + 2ab\text{Cov}(x, y) \quad (4)$$

- Applying this rule to the regression:

$$\begin{aligned} \text{Var}(R_{i,t}) &= \text{Var}(\alpha_i + \beta_i R_{m,t} + \epsilon_{i,t}) \\ &= \underbrace{\beta_i^2 \text{Var}(R_{m,t})}_{\text{market-related risk}} + \underbrace{\text{Var}(\epsilon_{i,t})}_{\text{non-market risk}} \end{aligned} \quad (5)$$

- The R^2 (“R-squared”) gives the fraction of the asset’s total variance that can be attributed to its market-related variance:

$$R^2 = \frac{\beta_i^2 \text{Var}(R_{m,t})}{\text{Var}(R_{i,t})} \quad (6)$$

- Non-market variance:

$$\text{Var}(\epsilon_{i,t}) = (1 - R^2)\text{Var}(R_{i,t}) \quad (7)$$

Beta and R-Squared

- Slope coefficient is the ratio of the covariance of the “dependent” and “independent” variables to the variance of the independent variable:

$$\beta_i = \frac{\text{Cov}(R_{i,t}, R_{m,t})}{\sigma^2(R_{m,t})} \quad (8)$$

- Beta can also be expressed in terms of correlation and standard deviations:

$$\beta_i = \frac{\sigma(R_i)}{\sigma(R_m)} \rho(R_i, R_m), \quad (9)$$

since correlation (denoted ρ) is defined as

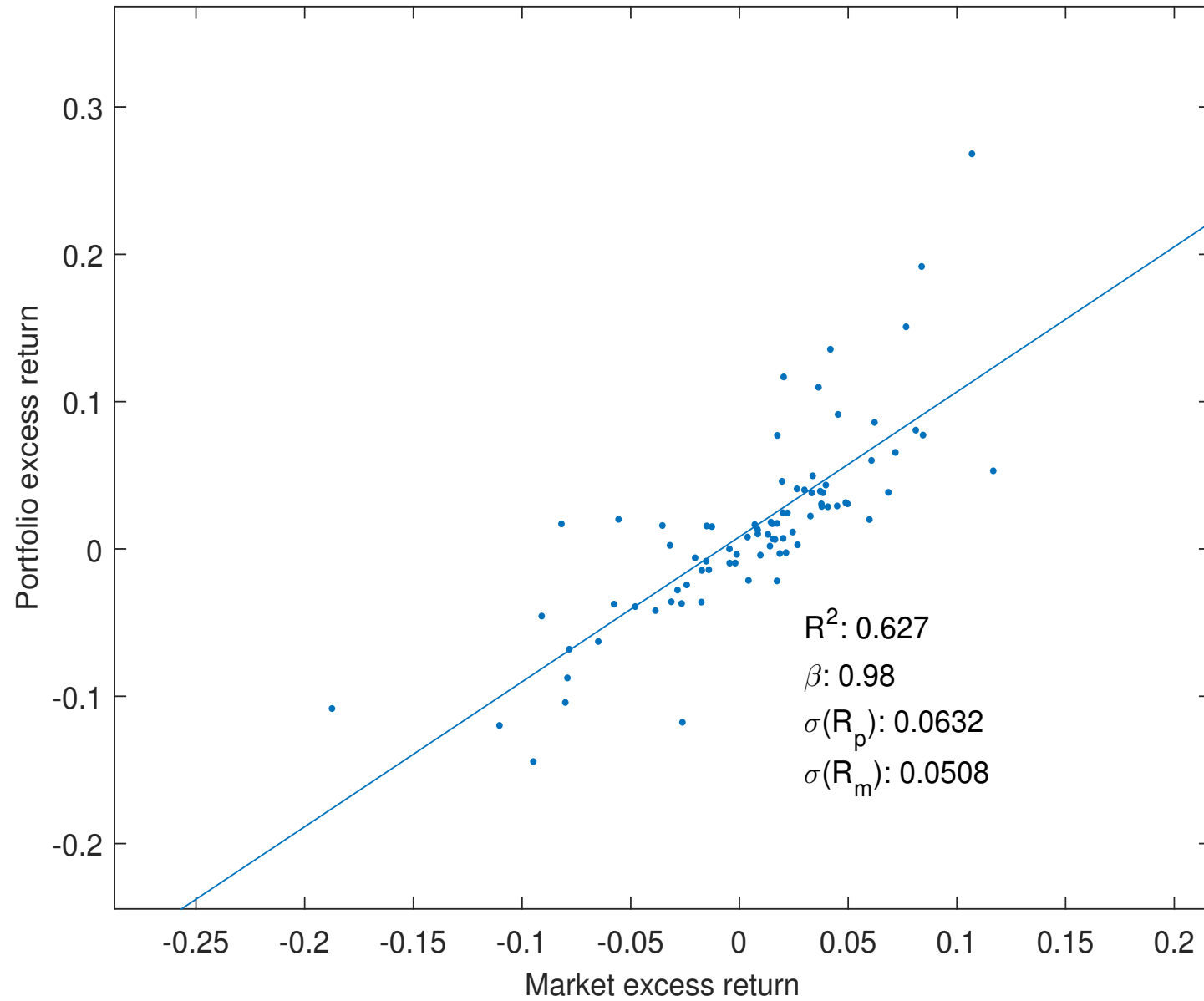
$$\rho(R_i, R_m) = \frac{\text{Cov}(R_{i,t}, R_{m,t})}{\sigma(R_i)\sigma(R_m)}. \quad (10)$$

- In a simple regression (one independent variable), R-squared is squared correlation:

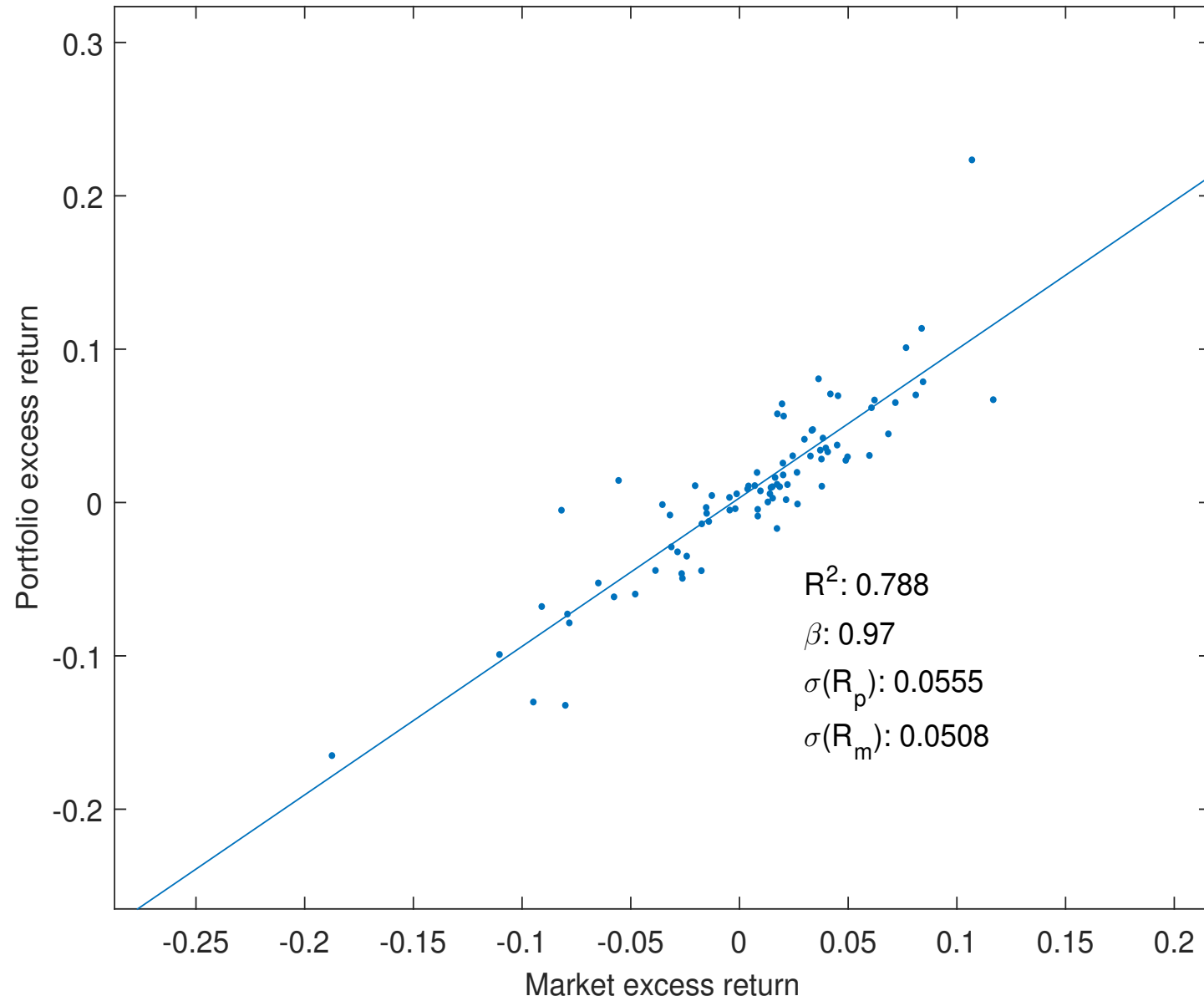
$$R^2 = [\rho(R_i, R_m)]^2. \quad (11)$$

- Important: β_i and R^2 describe different properties

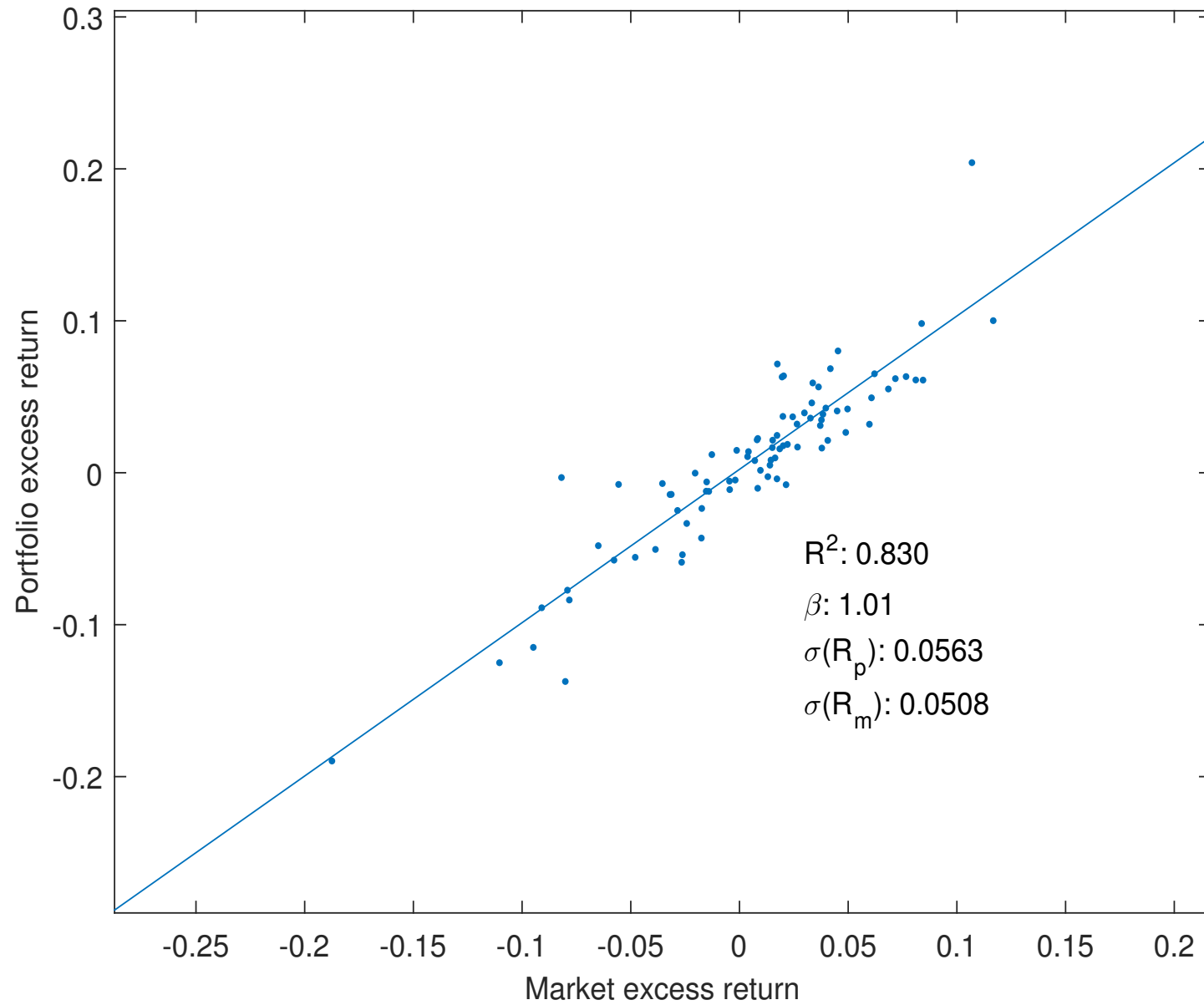
Size Decile 1



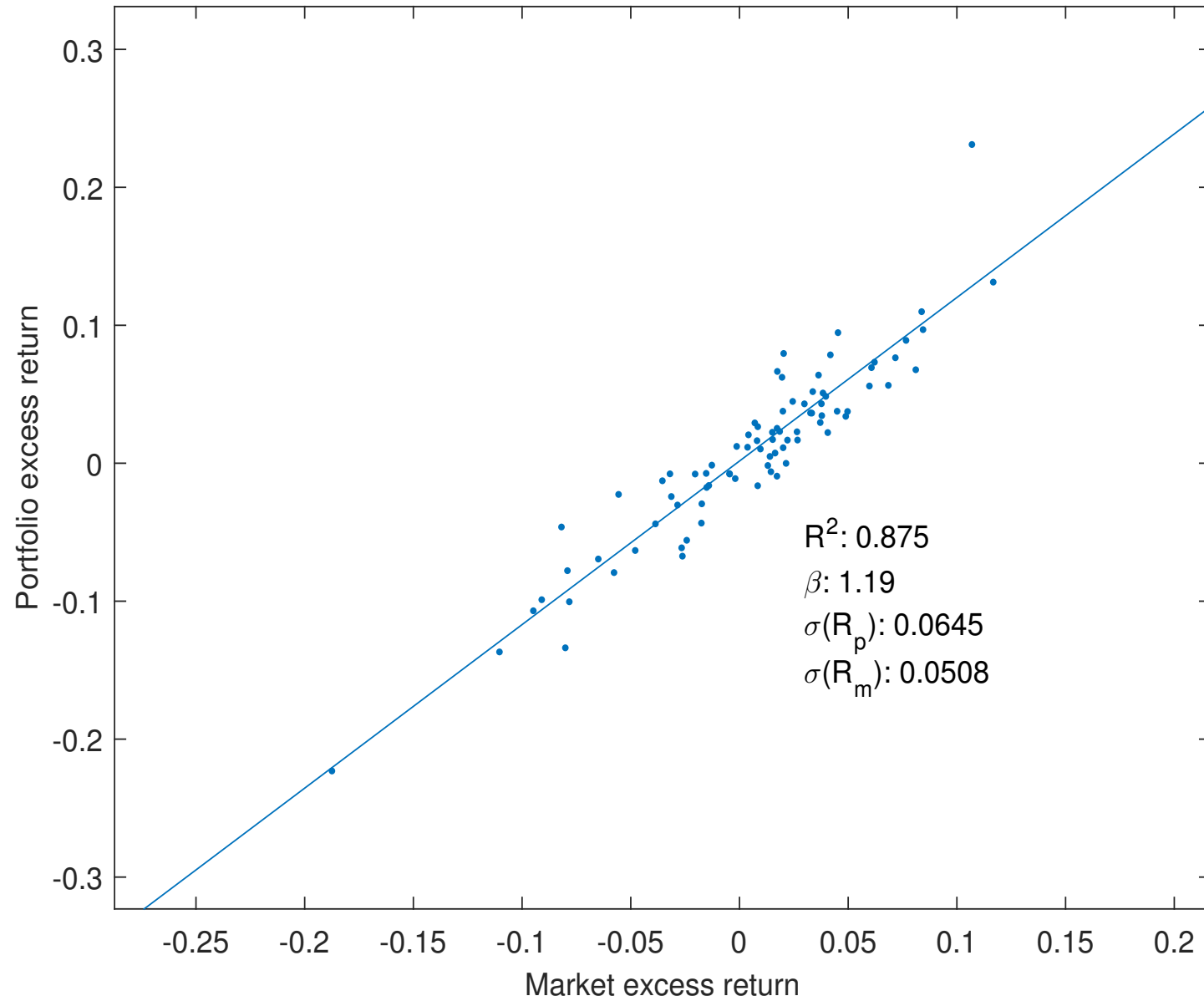
Size Decile 2



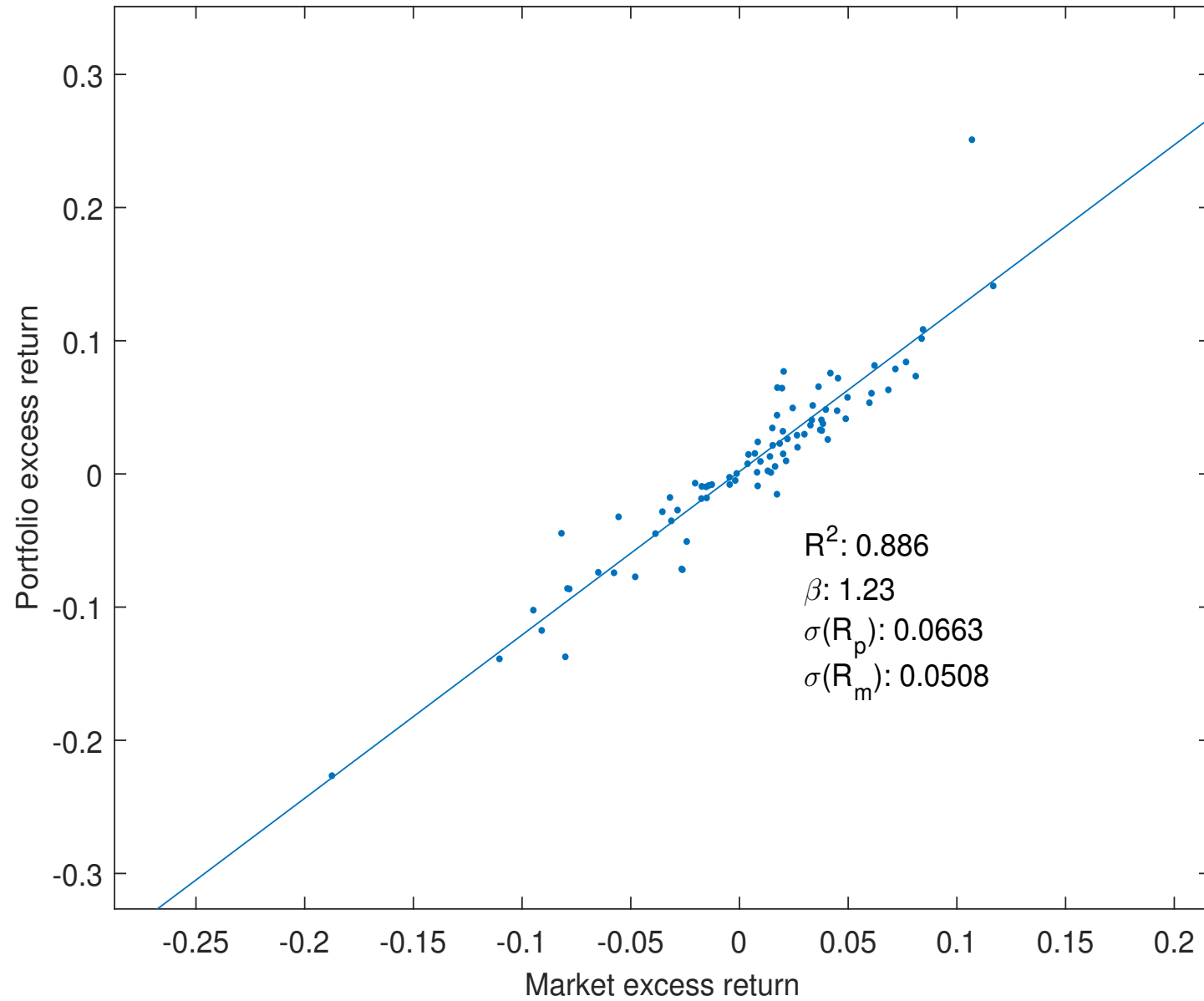
Size Decile 3



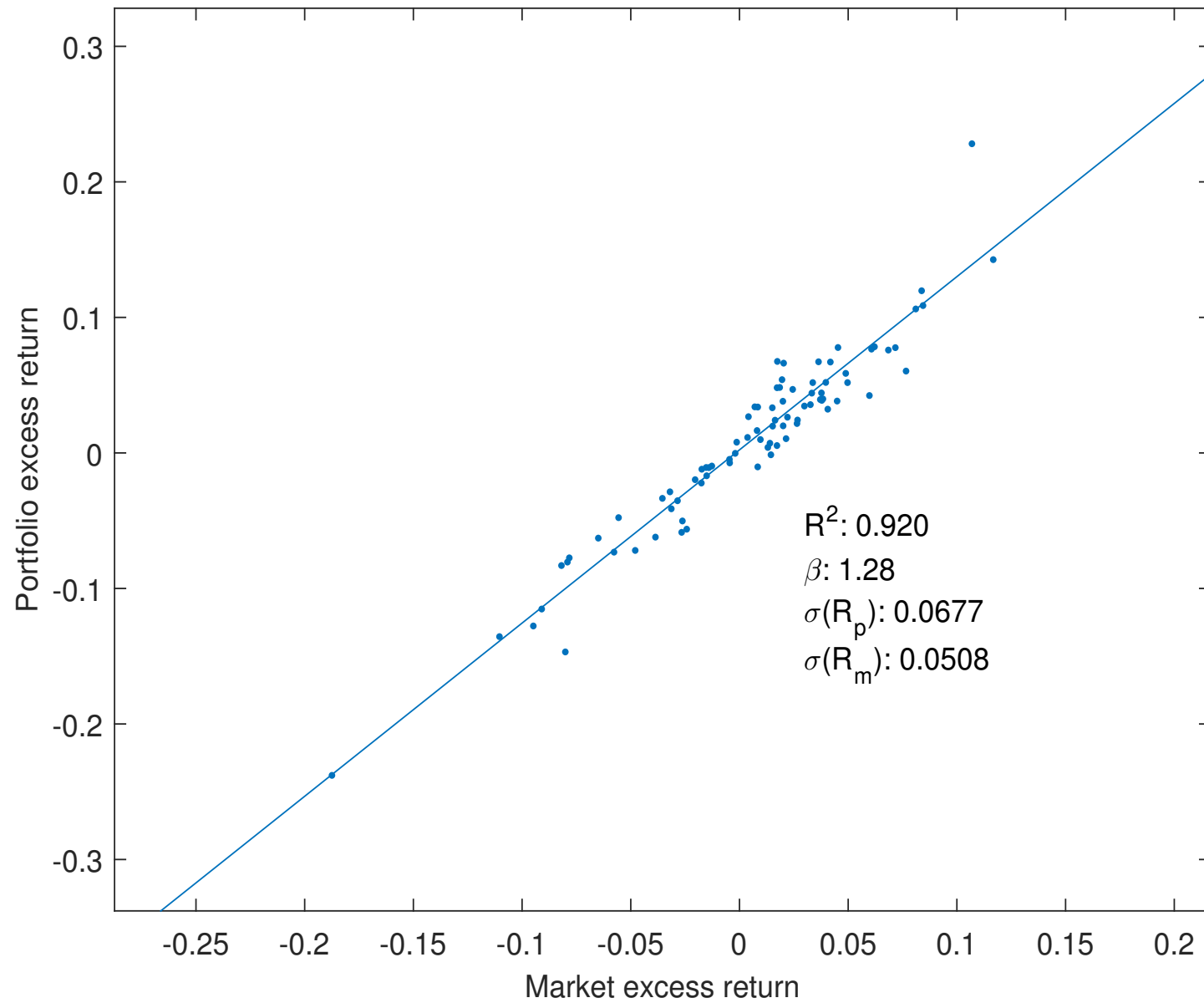
Size Decile 4



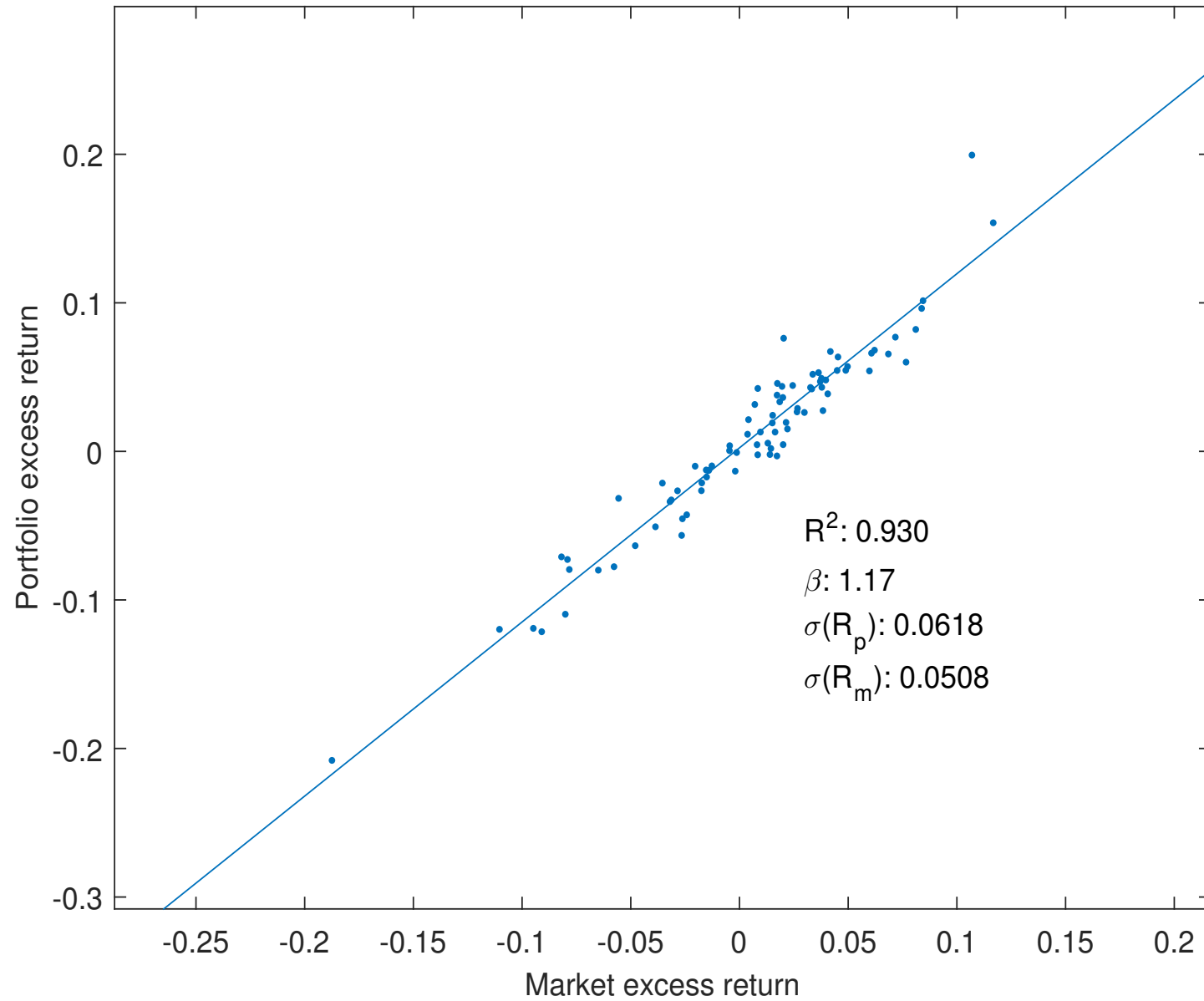
Size Decile 5



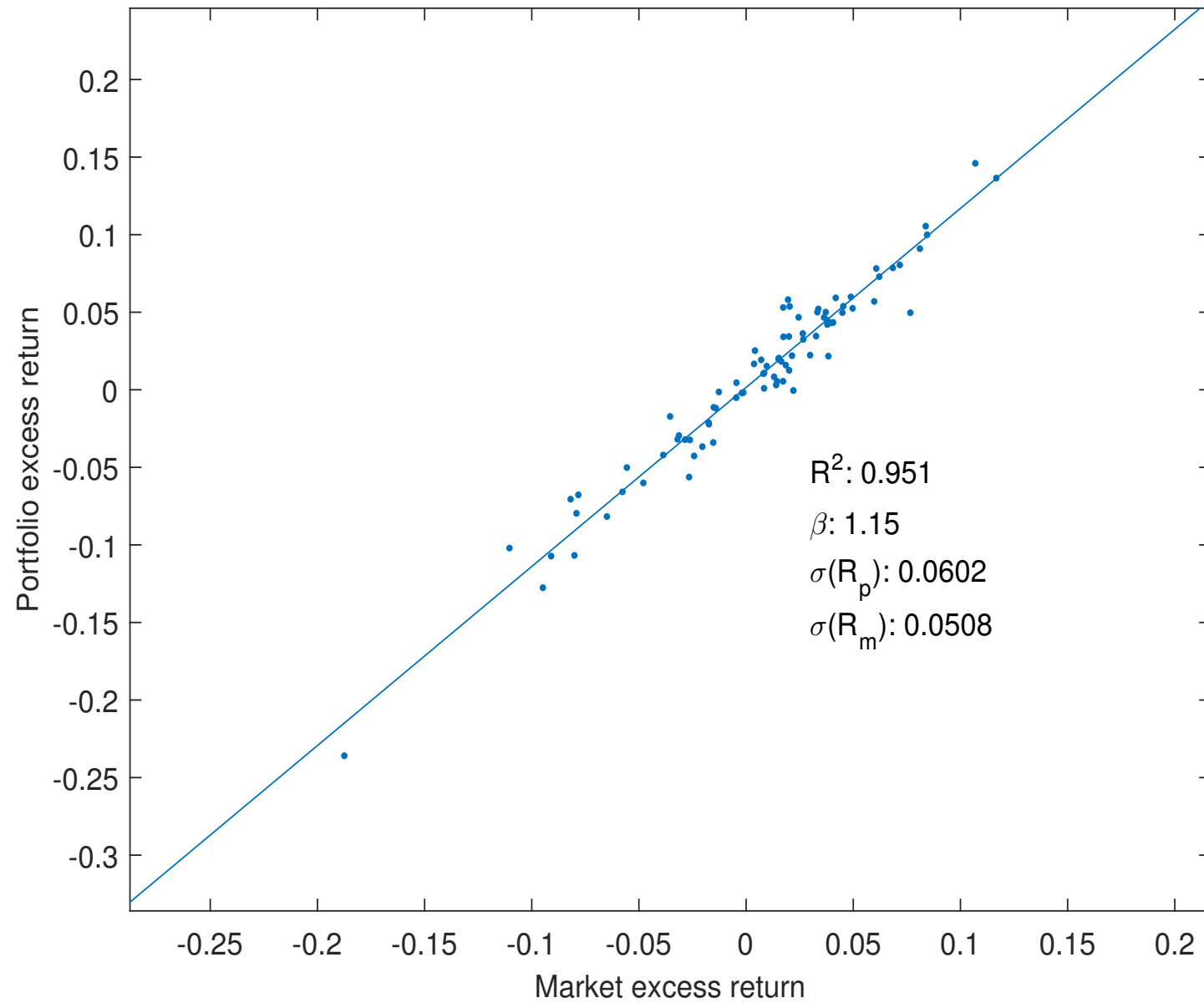
Size Decile 6



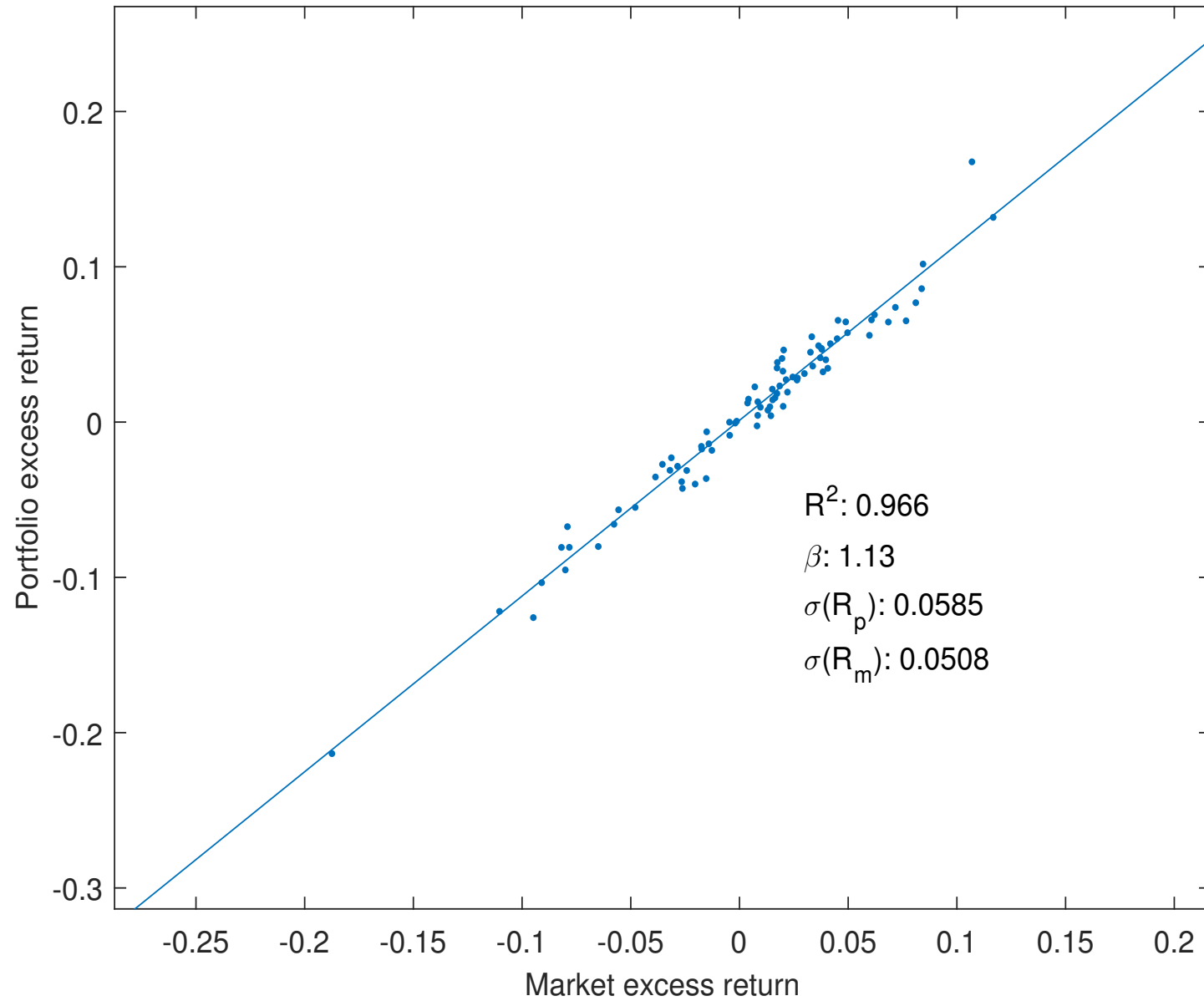
Size Decile 7



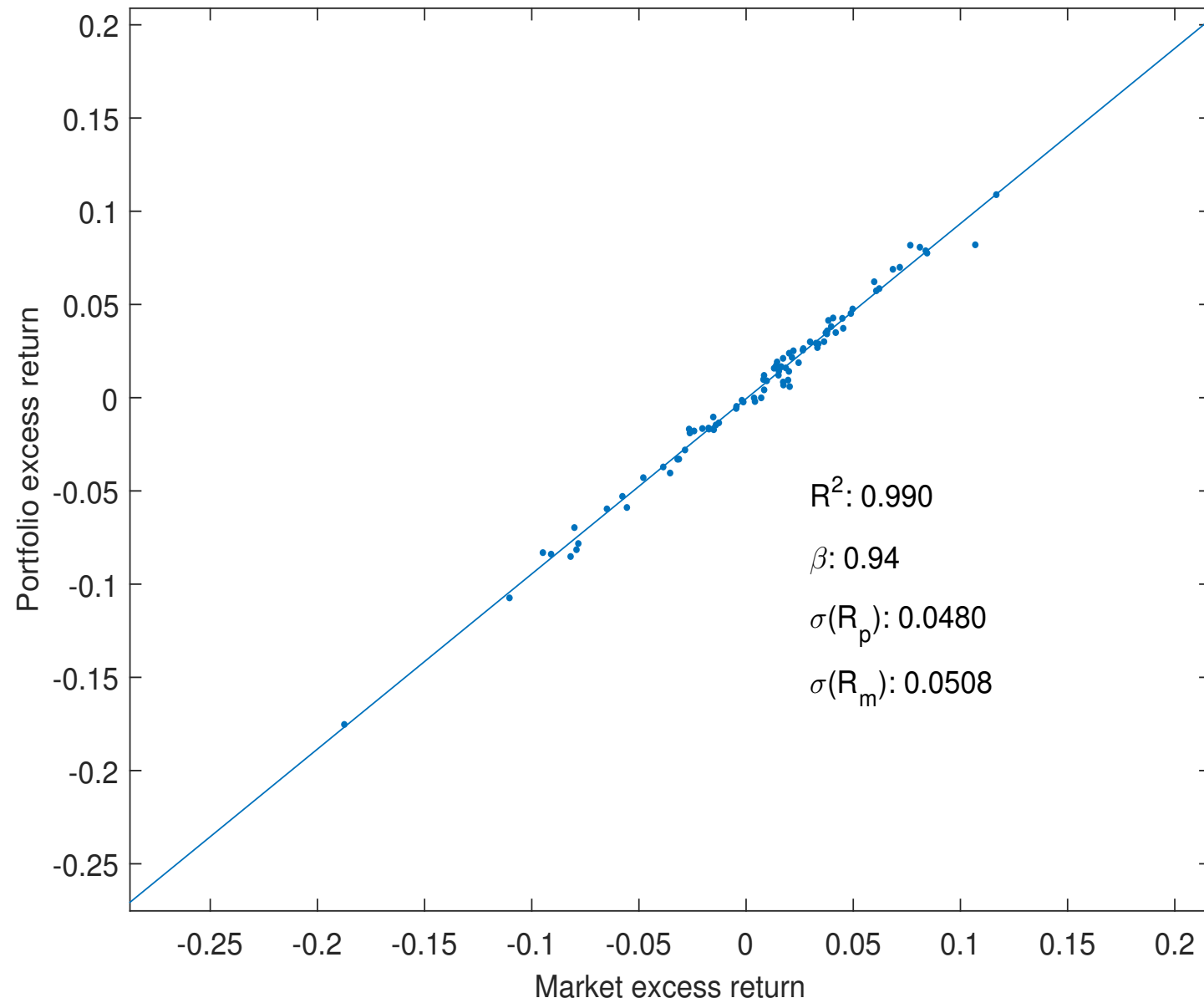
Size Decile 8



Size Decile 9



Size Decile 10



Futures Overlay

- Recall that profit on a market-index futures position of size X_{t-1} is $X_{t-1}R_{m,t}$
(note $R_{m,t}$ denotes the return in excess of cash)
- Consider an overall investment that combines
 - zero-investment futures position of size X_{t-1}
 - positive investment of size W_{t-1} in asset i
- Overall return (in excess of cash return $r_{f,t}$):

$$\begin{aligned} R_{C,t} &= \frac{\text{change in wealth from time } t-1 \text{ to } t}{\text{initial wealth at time } t-1} - r_{f,t} \\ &= \frac{W_{t-1}(R_{i,t} + r_{f,t}) + X_{t-1}R_{m,t}}{W_{t-1}} - r_{f,t} \\ &= R_{i,t} + \frac{X_{t-1}}{W_{t-1}}R_{m,t} \end{aligned} \tag{12}$$

Neutralizing Market Risk Using Futures

- Using the characteristic line for asset i ;

$$\begin{aligned} R_{C,t} &= R_{i,t} + \frac{X_{t-1}}{W_{t-1}} R_{m,t} \\ &= (\alpha_i + \beta_i R_{m,t} + \epsilon_{i,t}) + \frac{X_{t-1}}{W_{t-1}} R_{m,t} \\ &= \alpha_i + \left(\beta_i + \frac{X_{t-1}}{W_{t-1}} \right) R_{m,t} + \epsilon_{i,t} \end{aligned} \quad (13)$$

- To achieve an overall market-neutral investment, choose X_{t-1} so that

$$\left(\beta_i + \frac{X_{t-1}}{W_{t-1}} \right) = 0 \quad (14)$$

or

$$X_{t-1} = -\beta_i W_{t-1}. \quad (15)$$

- Gives

$$R_{C,t} = \underbrace{\alpha_i}_{\text{expected}} + \underbrace{\epsilon_{i,t}}_{\text{unexpected}} \quad (16)$$

Example

- \$20 million portfolio
 - standard deviation: 0.40
 - beta: 0.8
 - R-square: 0.75
- What is the market-neutralizing futures overlay?
- What is the volatility of the neutralized investment return?

Example

- \$20 million portfolio
 - standard deviation: 0.40
 - beta: 0.8
 - R-square: 0.75
- What is the market-neutralizing futures overlay?
- What is the volatility of the neutralized investment return?

$$\begin{aligned} X_{t-1} &= -\beta_i W_{t-1} \\ &= -0.8 \cdot (\$20 \text{ million}) \\ &= -\$16 \text{ million} \end{aligned}$$

Example

- \$20 million portfolio
 - standard deviation: 0.40
 - beta: 0.8
 - R-square: 0.75
- What is the market-neutralizing futures overlay?
- What is the volatility of the neutralized investment return?

$$\begin{aligned}X_{t-1} &= -\beta_i W_{t-1} \\&= -0.8 \cdot (\$20 \text{ million}) \\&= -\$16 \text{ million}\end{aligned}$$

$$\begin{aligned}\sigma(R_{C,t}) &= [\sigma^2(\epsilon_{i,t})]^{1/2} \\&= [(1 - R^2)\sigma^2(R_{i,t})]^{1/2} \\&= \sqrt{0.25 \times 0.40^2} \\&= 0.20\end{aligned}$$

Estimating Betas

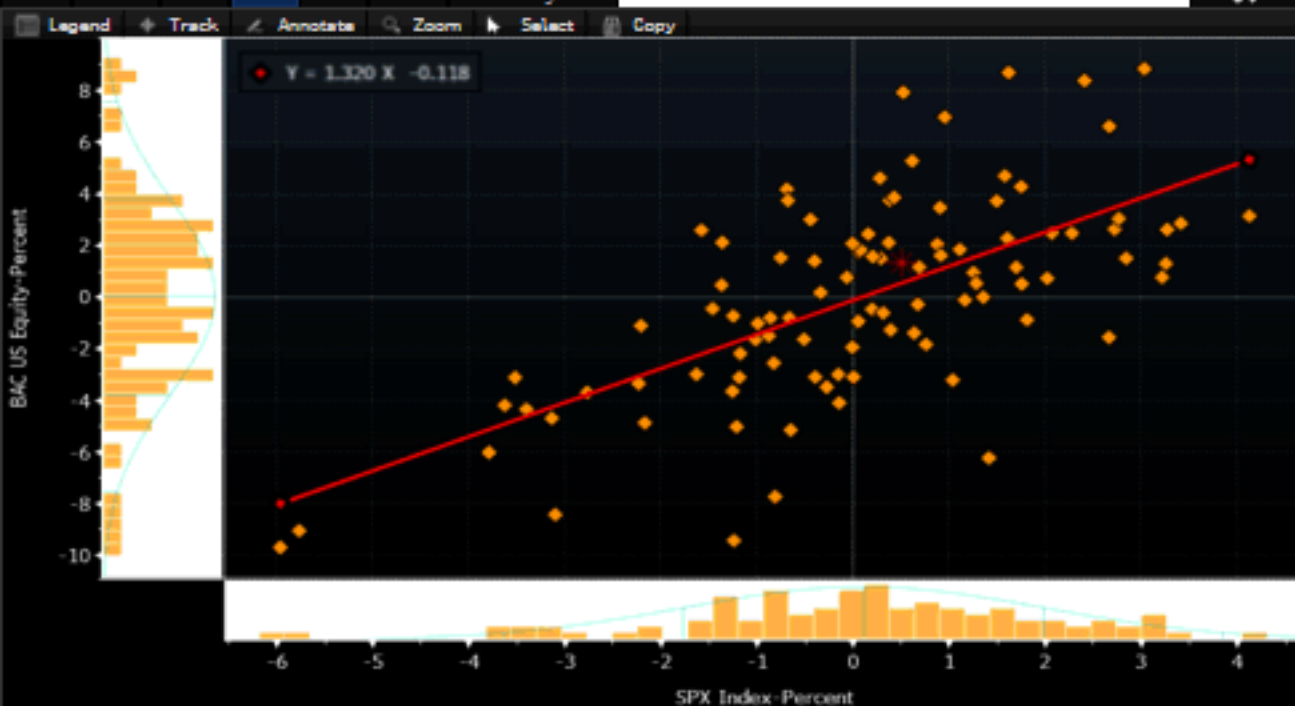
- The OLS regression estimator $\hat{\beta}_i$ often referred to as “historical beta.”
- The standard error of $\hat{\beta}_i$ is given (approximately) by

$$s(\hat{\beta}_i) = \sqrt{\frac{(1 - R^2)}{T} \frac{\sigma^2(R_{i,t})}{\sigma^2(R_{m,t})}} \quad (17)$$

- Thus, the precision of an OLS beta estimate is higher with
 - a larger number of observations (T)
 - a portfolio instead of an individual stock—higher R^2 and lower $\sigma^2(R_{i,t})$
- With monthly data, 5-7 years seems to give good tradeoff between
 - greater accuracy arising from more observations
 - lower accuracy from changes in true beta.
- Increasing T by using higher frequency returns, e.g., daily
 - allows a shorter window in calendar time
 - can be problematic for less liquid stocks (more on this later)

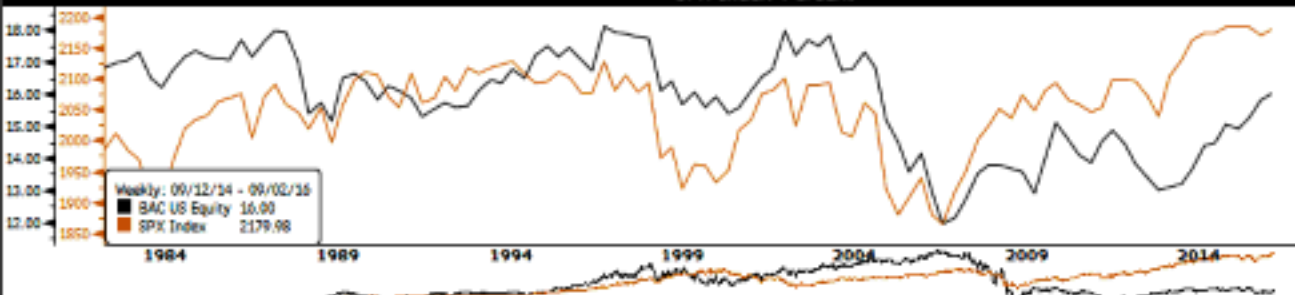
BAC US \$ ↓ 15.925 +.225 N15.92 / 15.93N 890 x 1000
 At 12:57 d Vol 40,070,702 0 15.77K H 15.94D L 15.68N Val 633.618M

BAC US Equity Relative Index SPX Index 96 Actions 97 Edit Historical Beta
 Data Last Price Data Last Price Wkly Linear Beta +/- Non-Param Reg On Percent
 09/09/2014 - 09/08/2016 09/09/2013 - 09/09/2015 Lag 0 Winsorize 2 Std Dev Local
 6M YTD 1Y 2Y 5Y Max Weekly



Y = BANK OF AMERICA CORP
 X = S&P 500 INDEX

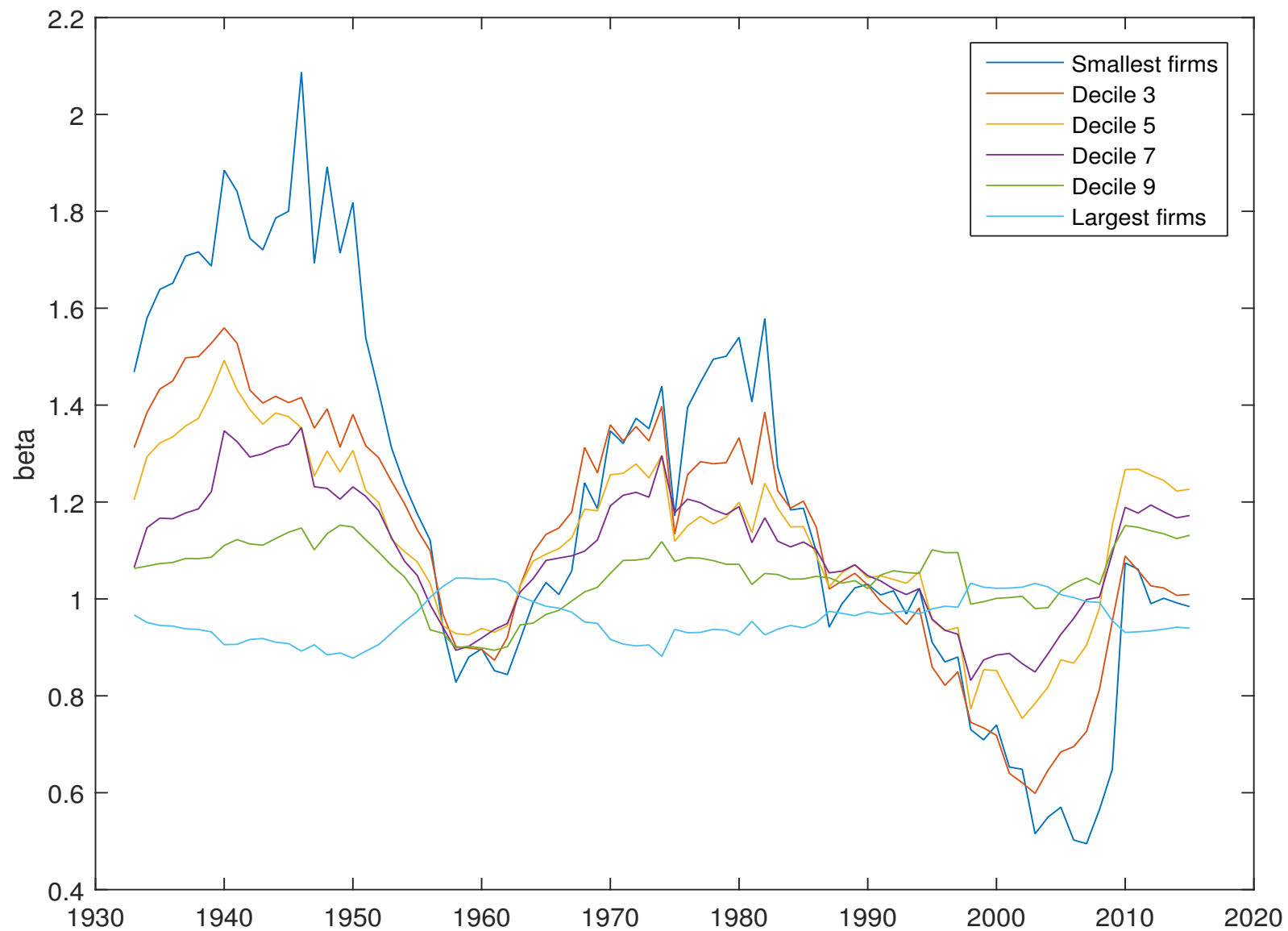
Linear Beta	Range 1
Raw BETA	1.320
Adjusted BETA	1.213
ALPHA (Intercept)	-0.118
R^2 (Correlation^2)	0.428
R (Correlation)	0.654
Std Dev of Error	2.883
Std Error of ALPHA	0.285
Std Error of BETA	0.152
t-Test	8.699
Significance	0.000
Last T-Value	0.270
Last P-Value	0.606
Number of Points	103
Last Spread	2163.98
Last Ratio	0.007



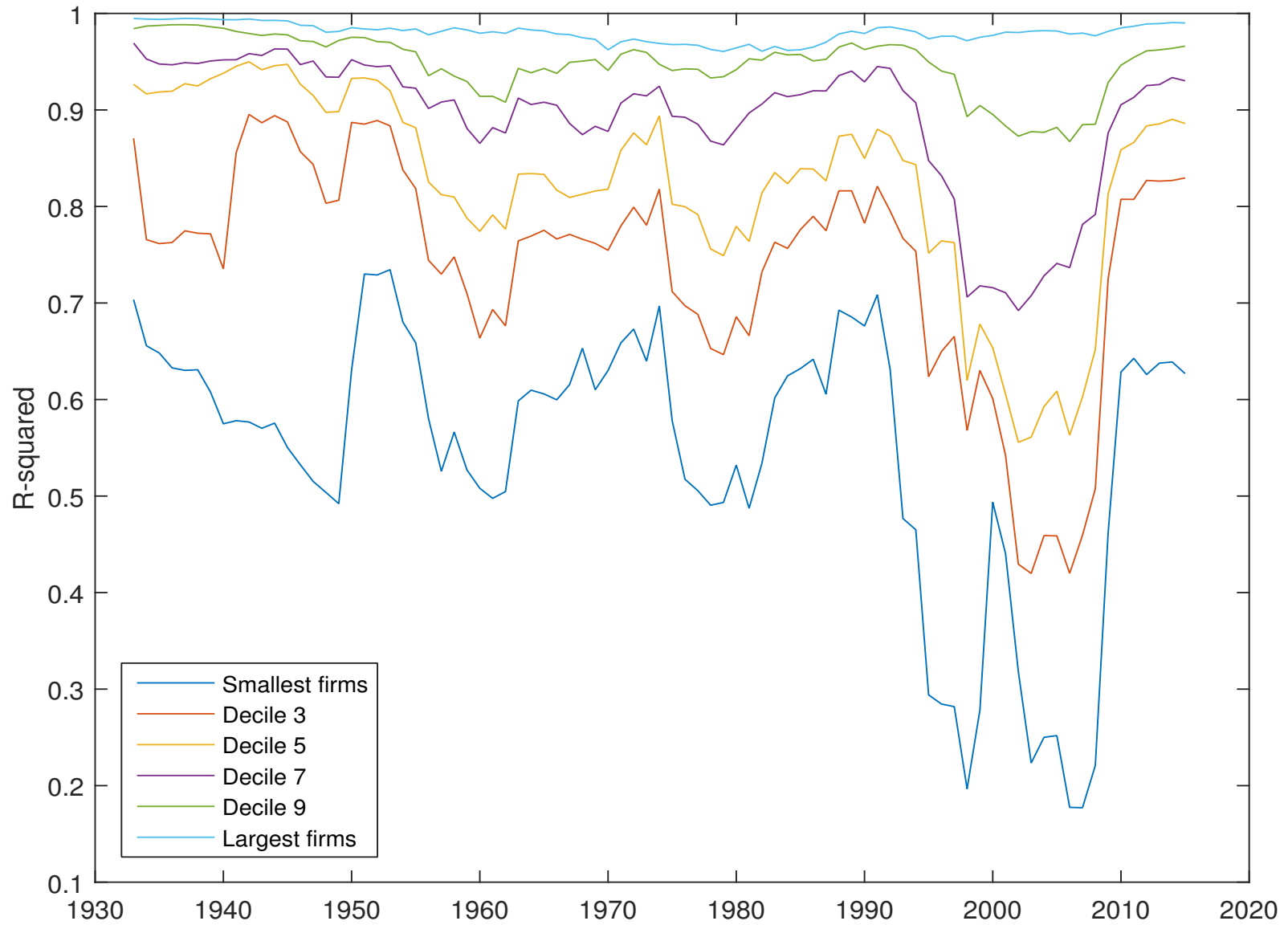
Adjusting Historical Betas on Individual Stocks

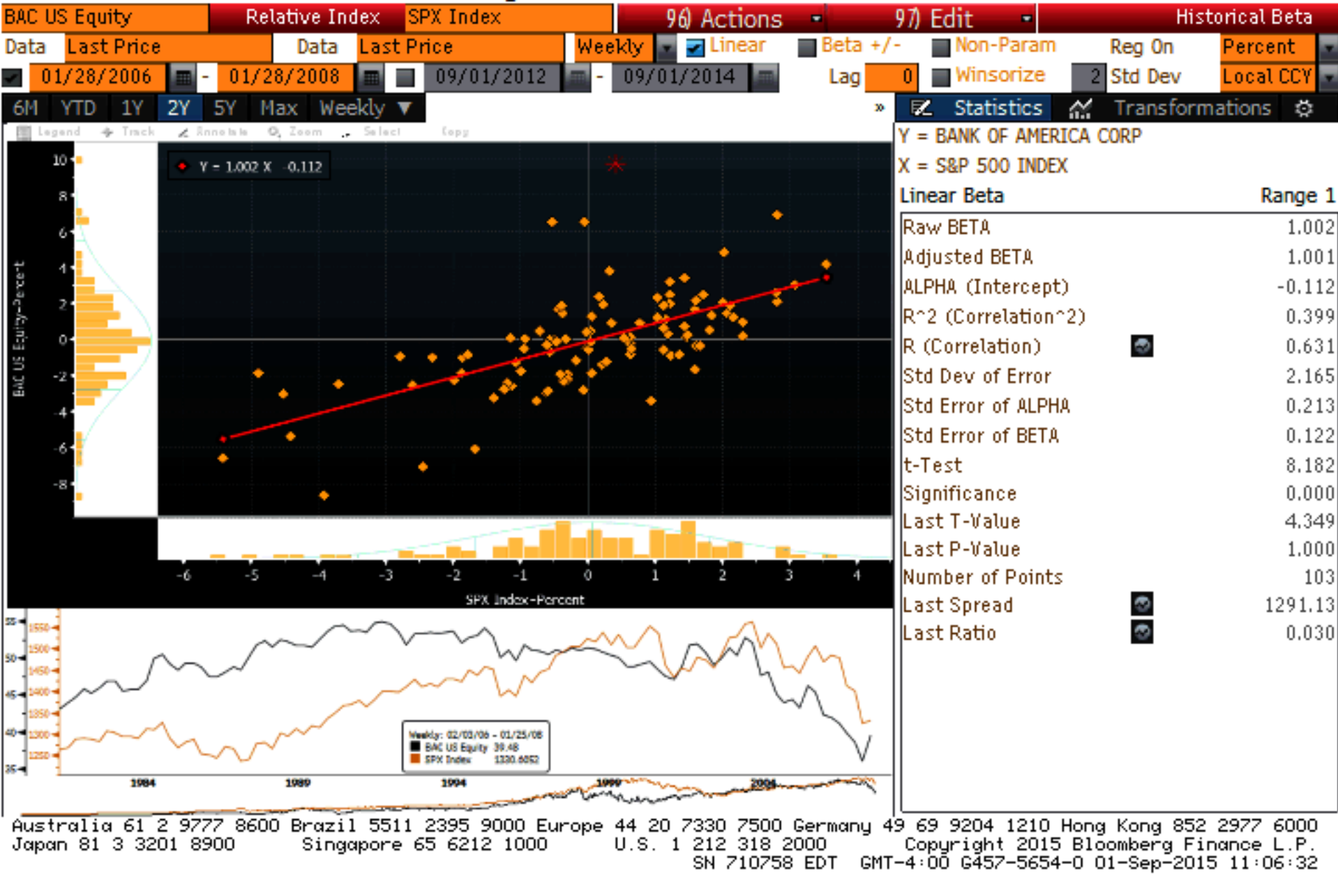
- Historical betas tend to “regress to the mean” from one period to the next.
- Two potential reasons:
 - economic: firms become larger and more “typical” as they mature
 - statistical: historical betas contain errors
- Second reason probably more important
- For any quantity estimated with error
 - Highest values in the cross section are probably too high
 - Lowest values in the cross section are probably too low
- “Raw” historical betas on individual stocks are often adjusted toward 1:
$$\hat{\beta}_{ADJ,i} = w \cdot 1 + (1 - w) \cdot \hat{\beta}_i. \quad (18)$$
- Common specification: $w = 1/3$.
- Historical betas on portfolios are typically not adjusted.

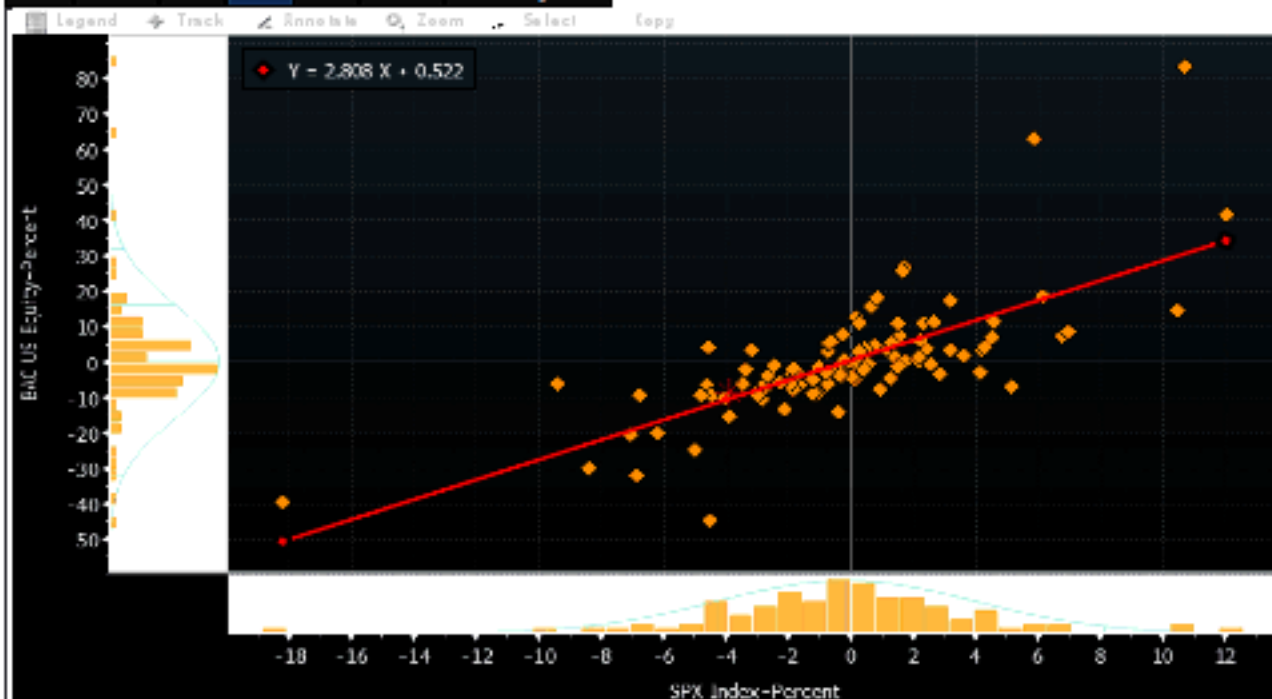
Beta Changes Over Time ...



As Does R-Squared ...

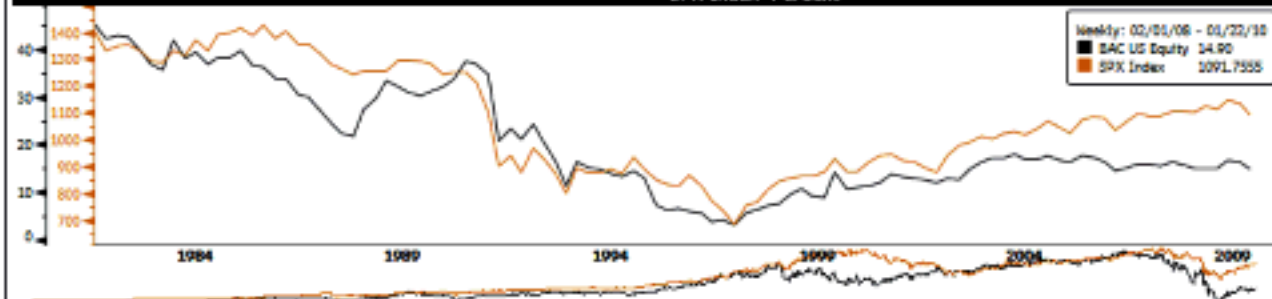


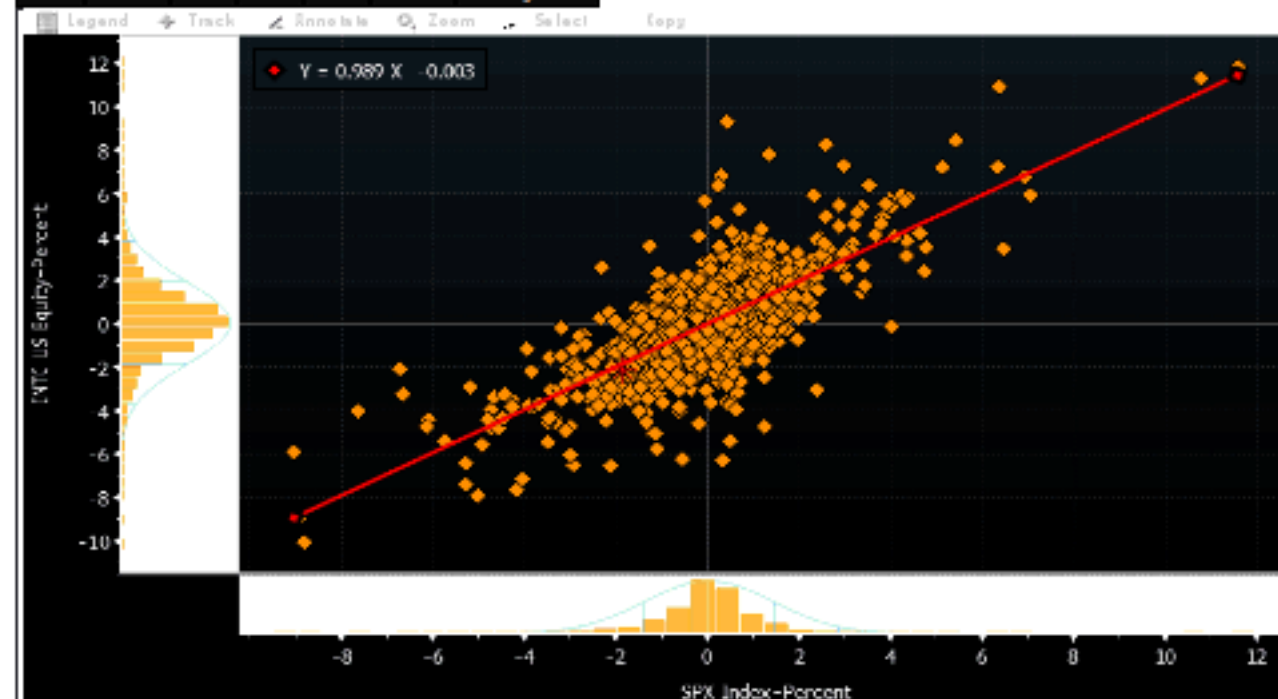




Y = BANK OF AMERICA CORP
X = S&P 500 INDEX

Linear Beta		Range 1
Raw BETA	2.808	
Adjusted BETA	2.205	
ALPHA (Intercept)	0.522	
R ² (Correlation ²)	0.534	
R (Correlation)	0.731	
Std Dev of Error	10.965	
Std Error of ALPHA	1.081	
Std Error of BETA	0.261	
t-Test	10.755	
Significance	0.000	
Last T-Value	0.186	
Last P-Value	0.574	
Number of Points	103	
Last Spread	1076.86	
Last Ratio	0.014	

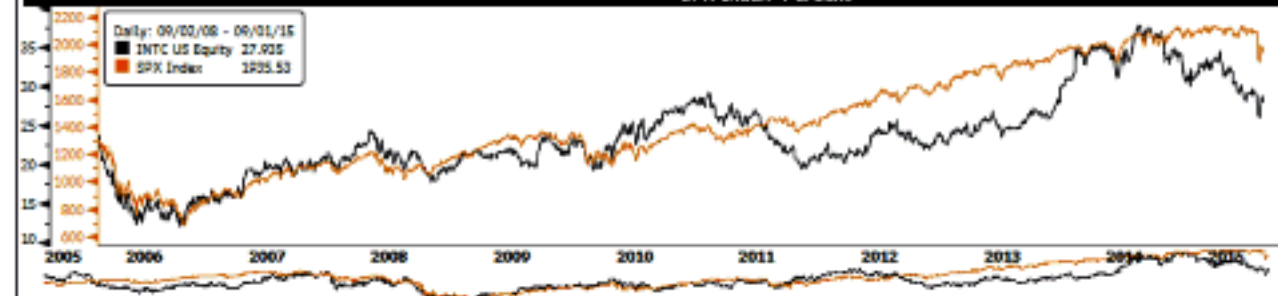


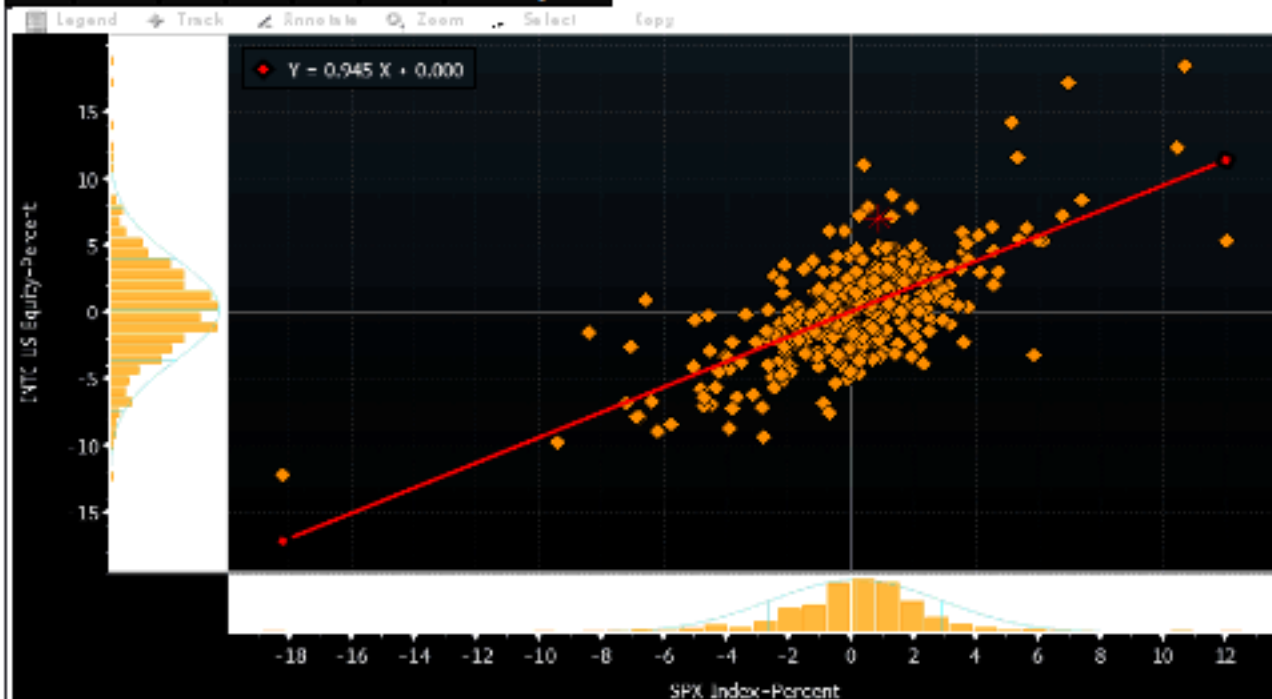


Y = INTEL CORP
X = S&P 500 INDEX

Linear Beta

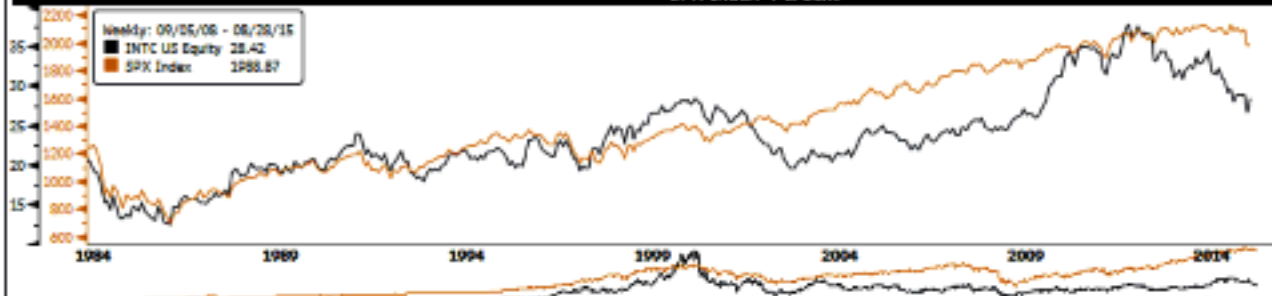
	Range 1
Raw BETA	0.989
Adjusted BETA	0.992
ALPHA (Intercept)	-0.003
R^2 (Correlation^2)	0.549
R (Correlation)	0.741
Std Dev of Error	1.274
Std Error of ALPHA	0.030
Std Error of BETA	0.021
t-Test	46.246
Significance	0.000
Last T-Value	-0.219
Last P-Value	0.413
Number of Points	1762
Last Spread	1907.60
Last Ratio	0.014

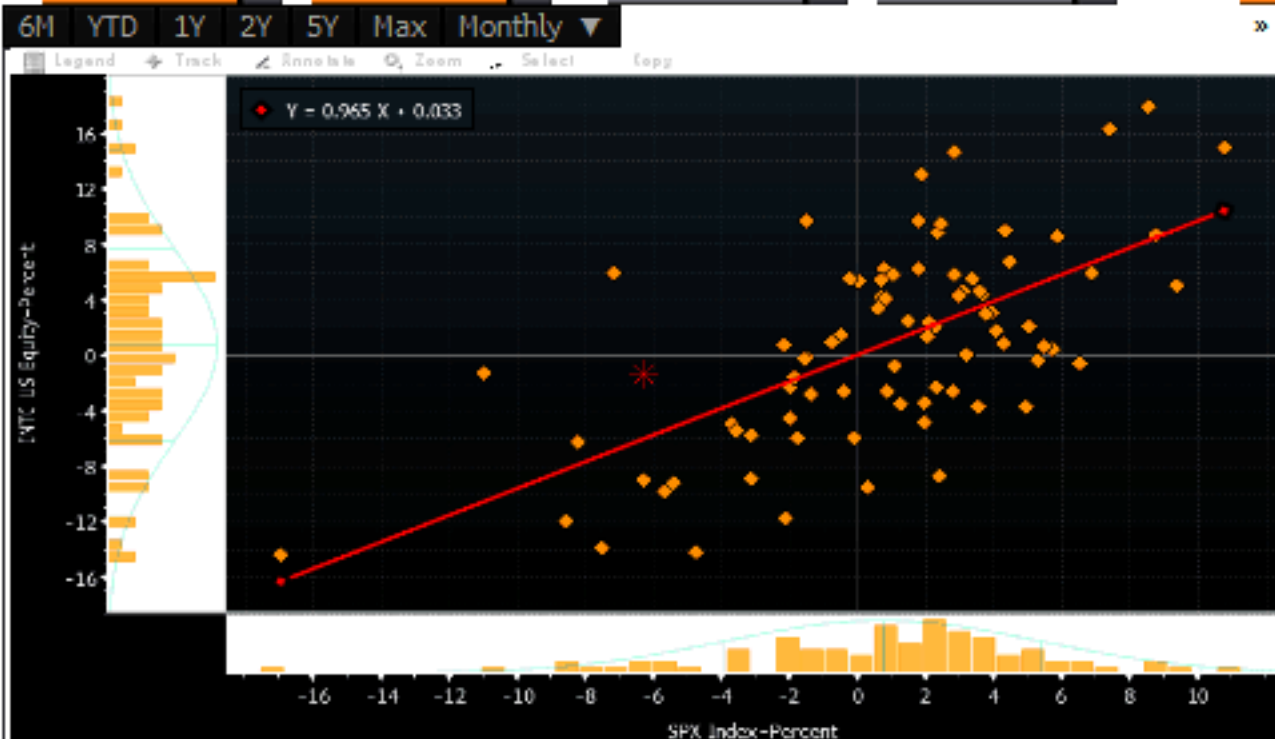




Y = INTEL CORP
X = S&P 500 INDEX

Linear Beta	Range 1
Raw BETA	0.945
Adjusted BETA	0.963
ALPHA (Intercept)	0.000
R^2 (Correlation^2)	0.477
R (Correlation)	0.691
Std Dev of Error	2.754
Std Error of ALPHA	0.145
Std Error of BETA	0.052
t-Test	18.169
Significance	0.000
Last T-Value	2.226
Last P-Value	0.987
Number of Points	364
Last Spread	1960.45
Last Ratio	0.014

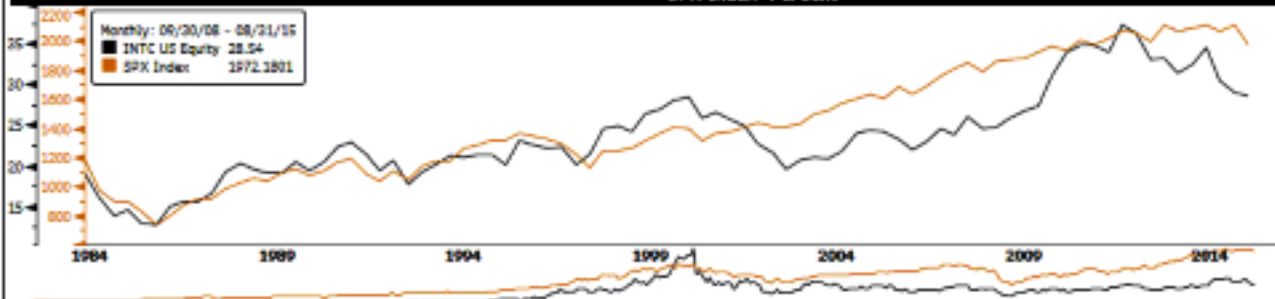


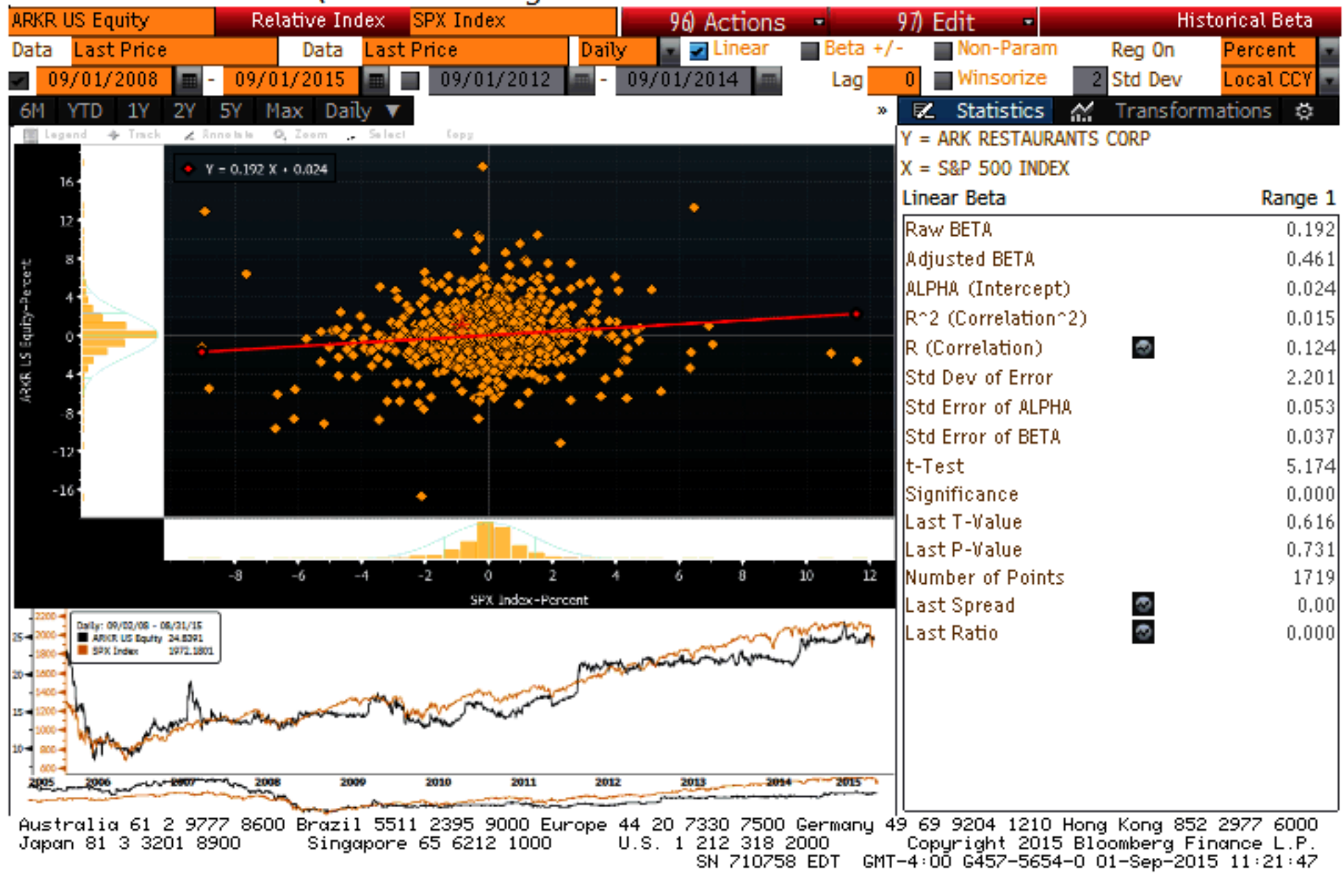


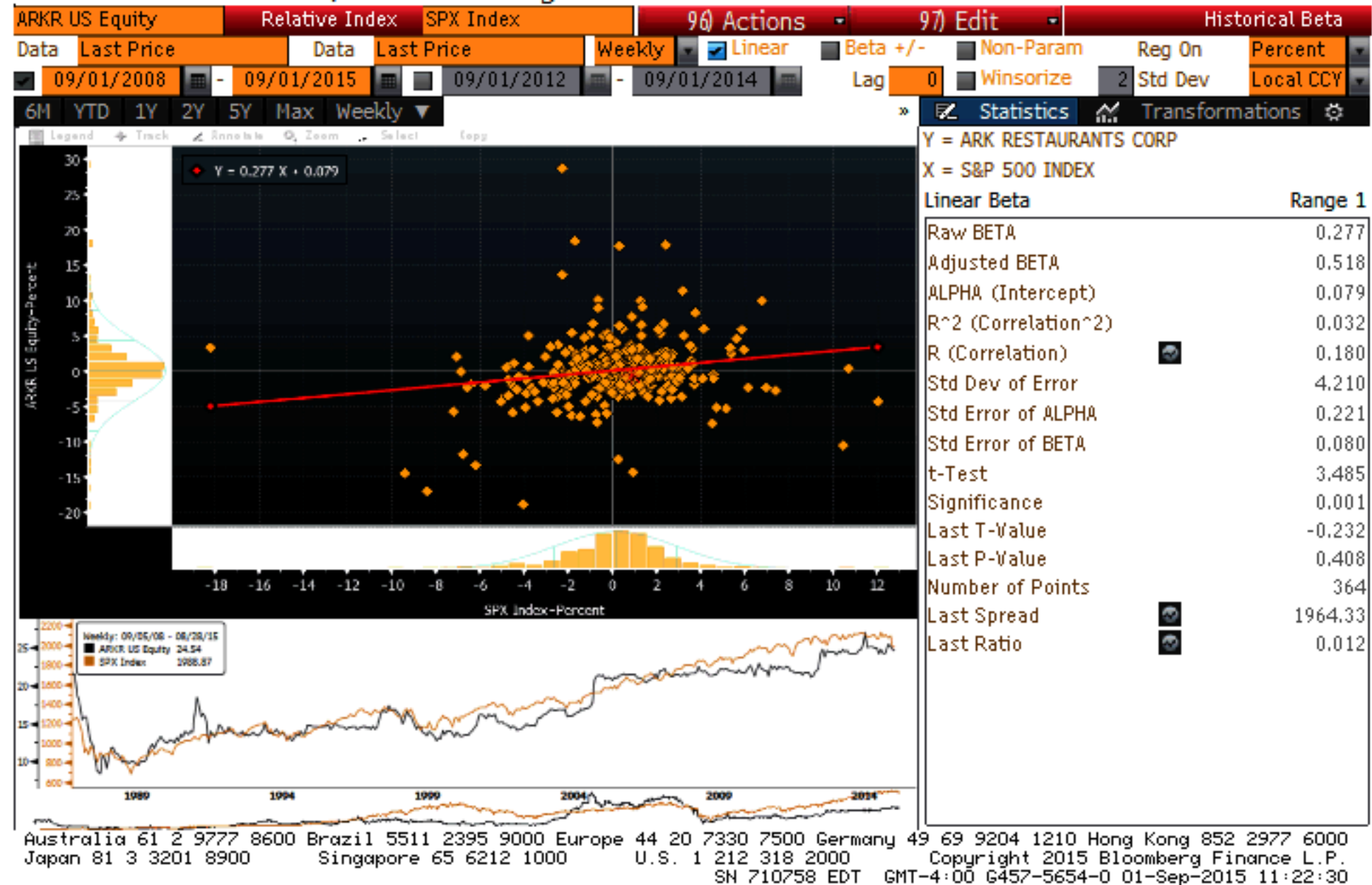
Statistics Transformations

Y = INTEL CORP
X = S&P 500 INDEX

Linear Beta	Range 1
Raw BETA	0.965
Adjusted BETA	0.977
ALPHA (Intercept)	0.033
R^2 (Correlation^2)	0.412
R (Correlation)	0.642
Std Dev of Error	5.427
Std Error of ALPHA	0.603
Std Error of BETA	0.128
t-Test	7.539
Significance	0.000
Last T-Value	0.830
Last P-Value	0.795
Number of Points	83
Last Spread	1943.64
Last Ratio	0.014







Australia 61 2 9777 8600

Brazil 5511 2395 9000

Europe 44 20 7330 7500

Germany 49 69 9204 1210

Hong Kong 852 2977 6000

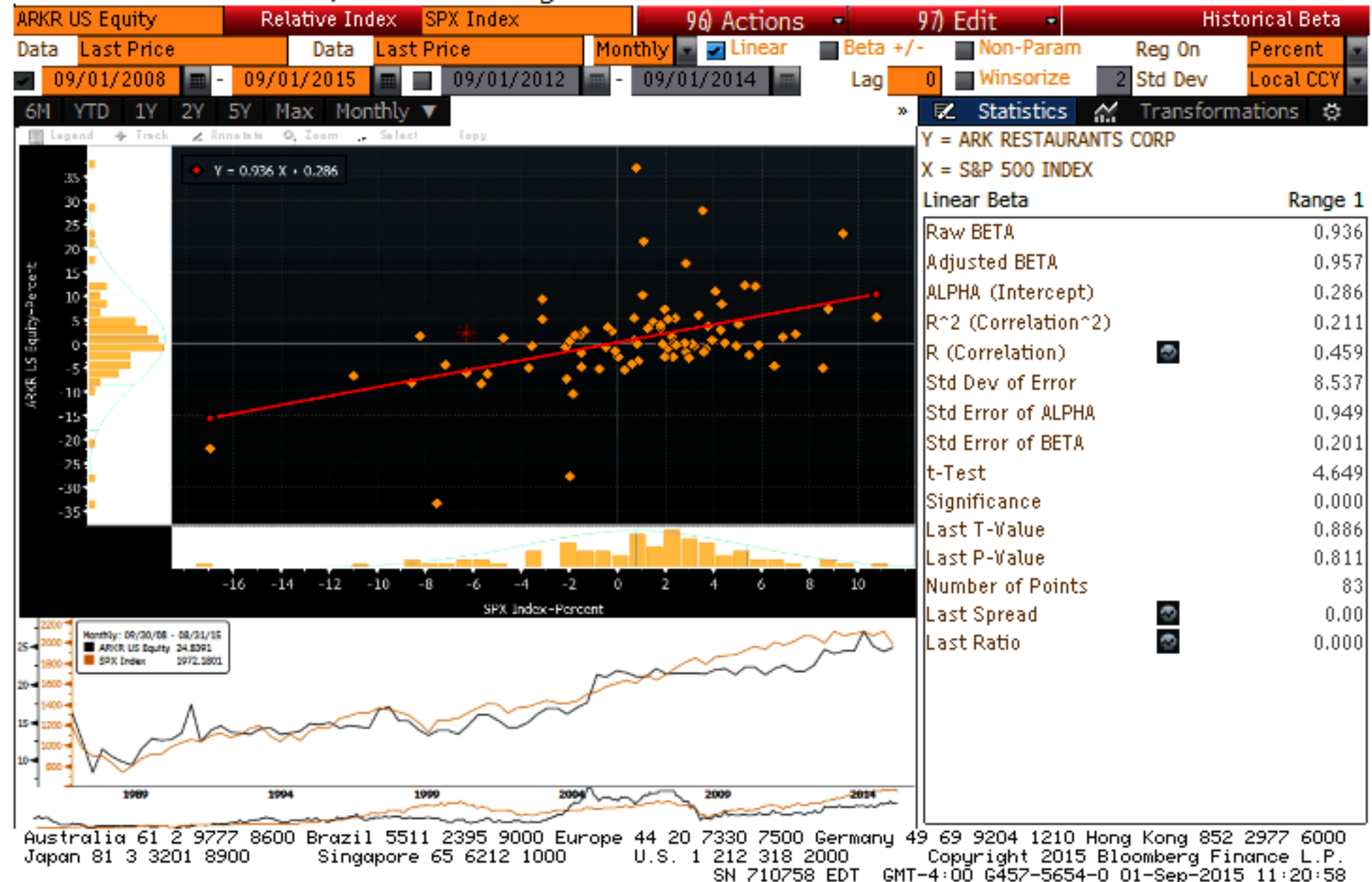
Japan 81 3 3201 8900

Singapore 65 6212 1000

U.S. 1 212 318 2000

Copyright 2015 Bloomberg Finance L.P.

SN 710758 EDT GMT-4:00 G457-5654-0 01-Sep-2015 11:22:30



ARKR US Equity-Percent

SPX Index-Percent

Y = 0.936 X + 0.286

Monthly: 09/30/08 - 08/31/15

ARKR US Equity 24.8391

SPX Index 1972.1801

Australia 61 2 9777 8600

Brazil 5511 2395 9000

Europe 44 20 7330 7500

Germany 49 69 9204 1210

Hong Kong 852 2977 6000

Japan 81 3 3201 8900

Singapore 65 6212 1000

U.S. 1 212 318 2000

Copyright 2015 Bloomberg Finance L.P.

SN 710758 EDT GMT-4:00 G457-5654-0 01-Sep-2015 11:20:58

Liquidity and Beta Estimation

- Estimating betas with higher frequency (e.g., daily returns)
 - works well for stocks traded fairly heavily
 - generally under-estimates betas of less liquid stocks
- For a less liquid stock
 - prices don't immediately reflect relevant market-wide information
 - observed day t return, $R_{i,t}$, includes market effects before day t
 - regressing R_{it} on $R_{m,t}$ misses some market sensitivity
- Common fix (Dimson):
 - Run the multiple regression

$$R_{i,t} = a_i + \beta_{i,0}R_{m,t} + \beta_{i,1}R_{m,t-1} + \cdots + \beta_{i,K}R_{m,t-K} + e_{i,t}$$

- Estimate the stock's beta as the sum of the estimated slopes:

$$\hat{\beta}_i = \hat{\beta}_{i,0} + \hat{\beta}_{i,1} + \cdots + \hat{\beta}_{i,K}$$

Combining Assets

- Consider a portfolio p of two assets, i and j .

$$\begin{aligned} R_{p,t} &= w_i R_{i,t} + w_j R_{j,t} \\ &= w_i (\alpha_i + \beta_i R_{m,t} + \epsilon_{i,t}) + w_j (\alpha_j + \beta_j R_{m,t} + \epsilon_{j,t}) \\ &= \alpha_p + \beta_p R_{m,t} + \epsilon_{p,t}, \end{aligned} \tag{19}$$

where

$$\alpha_p = w_i \alpha_i + w_j \alpha_j$$

$$\beta_p = w_i \beta_i + w_j \beta_j$$

$$\epsilon_{p,t} = w_i \epsilon_{i,t} + w_j \epsilon_{j,t}$$

- Same properties as before:

$$\text{COV}(\epsilon_{p,t}, R_{m,t}) = 0 \tag{20}$$

$$\beta_p = \frac{\text{Cov}(R_{p,t}, R_{m,t})}{\sigma^2(R_{m,t})} \tag{21}$$

$$\text{Var}(R_{p,t}) = \beta_p^2 \text{Var}(R_{m,t}) + \text{Var}(\epsilon_{p,t}) \tag{22}$$