

Introduction to Regression

Rob Hayward

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Outline

- 1 Modelling
- 2 Ordinary Least Squares
- 3 OLS Assumptions
- 4 Identifying and dealing with problems

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- What is the idiosyncratic or individual performance of the security?

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- ε_t is the error that covers omitted variables, measurement error and other stochastic or random elements

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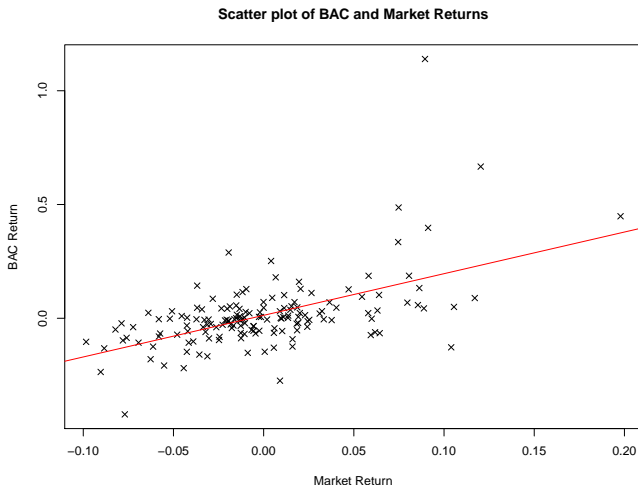
- y_t is the return of Bank of America
- α is an intercept or constant
- x_t is the return of the market (S&P 500)
- β is the relationship between BAC returns and the market returns
- ε_t is all the other factors that affect BAC returns

Caution!

“Essentially all models are wrong, but some are useful”

(Box, 1987, p. 424)

S&P 500 and BAC



Solution 1

$$y_t = a + bx_t + u_t$$

Minimise the residuals

$$\text{Min} \sum_{t=1}^{t=T} u_t^2$$

$$\text{Min} \sum_{t=1}^{t=T} (y_t - a - bx_t)^2$$

Take, partial derivative to get the conditions.

Solution 2

$$\frac{\delta u}{\delta a} = \sum_{t=1}^{t=T} 2(y_t - a - bx_t) = 0$$

$$\frac{\delta u}{\delta b} = \sum_{t=1}^{t=T} 2x_t(y_t - a - bx_t) = 0$$

The following can be skipped. It comes from the appendix to chapter 4 of Econometrics (Stock and Watson)

Solution 3

collecting terms and dividing by n . Remember that summing and dividing by T is equal to the mean.

$$\bar{y} = a + b\bar{x} = 0 \quad (1)$$

and

$$\frac{1}{T} \sum_{t=1}^T x_t y_t - a\bar{x} - b\frac{1}{T} \sum_{t=1}^T x_t^2 = 0 \quad (2)$$

Substitute Equation 1 into 2

$$\frac{1}{T} \sum x_t y_t - (\bar{y} - b\bar{x}) \bar{x} - b\frac{1}{T} \sum_{t=1}^T x_t^2$$

Solution 4

$$\begin{aligned} & \frac{1}{T} \sum x_t y_t - \bar{y} \bar{x} + b \bar{x}^2 - b \frac{1}{T} \sum x^2 \\ & \frac{1}{T} \sum x_t y_t - \bar{y} \bar{x} + b \left(\bar{x}^2 - \frac{1}{T} \sum x^2 \right) \\ & b = \frac{\frac{1}{T} \sum x_t y_t - \bar{y} \bar{x}}{\frac{1}{T} \sum x^2 - \bar{x}^2} \\ & b = \frac{\sum (x_t - \bar{x})(y_t - \bar{y})}{\sum (x_t - \bar{x})^2} \end{aligned}$$

Solution: matrix form

In matrix form

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{u}$$

$$\mathbf{u} = \mathbf{y} - \mathbf{X}\boldsymbol{\beta}$$

$$\mathbf{u}'\mathbf{u} = (\mathbf{X}\boldsymbol{\beta} + \mathbf{u})'(\mathbf{X}\boldsymbol{\beta} + \mathbf{u})$$

Taking derivative and re-arranging (see textbook for proof)

$$\boldsymbol{\beta} = (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{y}$$

Regression Table

	Estimate	Std. Error	t value	$\Pr(> t)$
(Intercept)	0.0130	0.0105	1.23	0.2203
SPY.R	1.8303	0.2240	8.17	0.0000

The Adjusted R^2 is 0.29, therefore nearly 30% of the BAC returns are explained by the returns of the market. 95% confidence intervals for the β are 1.39 to 2.27.

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Given a number of assumptions OLS is the BLUE **B**est, **L**inear, **U**nbiased, **E**stimator.

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 - Heteroskedasticity (some errors are systematically larger than others)
- Explanatory variables are not related to the error
- Additionally, assume *normal errors* if we want to use normal assumption to compute *t-tests* of coefficients

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Solution: Add missing variable or a proxy

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Solution: Remove superfluous variable. Be careful of *the dummy variable problem*

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Textbooks

All in the library

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- C. Dougherty, "Introduction to Econometrics", OUP

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- JH Stock and M Watson "Introduction to Econometrics", Pearson

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- D Gujarati, "Basic Econometrics", McGraw-Hill

Bibliography

Box, G. E. (1987), *Empirical Model Building and Response Surfaces*, John Wiley and sons.