## Math 5870—Act 4

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# Mathematical Explanation of $\beta_1$ in some different scenarios

Let's first consider a SLR model where all assumptions are met;

$$Y = \beta_0 + \beta_1 x + \epsilon, \epsilon \sim \mathcal{N}(0, \sigma^2)$$
$$y_x = b_0 + b_1 x$$

In the case where there is a unit increase in x;

$$y_{x+1} = b_0 + b_1(x+1)$$

Measuring the corresponding amount of change in y when for a unit increase in x;

$$y_{x+1} - y_x = b_1$$

What do we do when these assumptions are violated?

#### 1)

Transformation for a model that violates the linearity assumption (between x and y) or big x values.

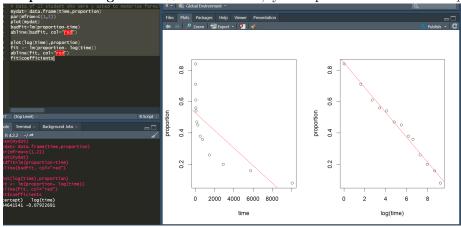
$$Y = \beta_0 + \beta_1 log(x) + \epsilon, \epsilon \sim \mathcal{N}(0, \sigma^2)$$

$$y_{x} = b_{0} + b_{1}log(x)$$

$$y_{x+1} = b_{0} + b_{1}log(x+1) + \epsilon$$

$$y_{x+1} - y_{x} = b_{1}log(\frac{x+1}{x})$$

$$y_{j} - y_{i} = log(\frac{x+1}{x})^{b_{1}}$$



Simple Meaning: For a 1% increase in x, y is expected to increase in  $b_1$ 

## 2)

Transformation for a model that violates normality and homoscedasticity or big Y values .

$$log(Y) = \beta_0 + \beta_1 x + \epsilon, \epsilon \sim \mathcal{N}(0, \sigma^2)$$
$$log(y_x) = b_0 + b_1 x$$
$$log(y_{x+1}) = b_0 + b_1(x+1)$$
$$log(\frac{y_{x+1}}{y_x}) = b_1$$

Therefore,

$$\frac{y_{x+1}}{y_x} = e^{b_1}$$

**Simple Meaning:** For a unit increase in x, log(y) is expected to increase by  $b_1$ 

## 3)

Transformation for a model that violates normality, homoscedasticity and linearity or big y and x values .

$$log(Y) = \beta_0 + \beta_1 log(x) + \epsilon, \epsilon \sim \mathcal{N}(0, \sigma^2)$$
$$log(y_x) = b_0 + b_1 log(x)$$
$$log(y_{x+1}) = b_0 + b_1 log(x+1)$$
$$log(\frac{y_{x+1}}{y_x}) = b_1 log(\frac{x+1}{x})$$

$$\frac{y_{x+1}}{y_x} = (\frac{x+1}{x})^{b_1}$$

Simple Meaning:1% increase in **x** will cause a  $b_1$  increase in log(y)