

6. from chapter 2 - Econometrics - simple linear regression analysis I found the appropriate test statistic to perform the test on.

⇒ Testing and confidence interval for slope (b):
and σ^2 is unknown

100(1- α)% confidence interval b is

$$\left[b \pm t_{n-2, \alpha/2} \sqrt{\frac{SS_{res}}{n-2}} \right]$$

SS_{res} is the residual sum of squares

$$SS_{res} = \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

⇒ Testing and confidence interval for intercept term (a)

100(1- α)% CI a is

$$\left[a \pm t_{n-2, \alpha/2} \sqrt{\frac{SS_{res}}{n-2} \left(\frac{1}{n} + \frac{\bar{x}^2}{S_{xx}} \right)} \right]$$

$$S_{xx} = \sum_{i=1}^n (x_i - \bar{x})^2$$

$$6.7 \quad b = \frac{\sum_{i=1}^n x_i y_i - n \bar{x} \bar{y}}{\sum_{i=1}^n x_i^2 - n \bar{x}} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

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we need to show $E(b) = \beta$

$$b = \frac{\sum_{i=1}^n (x_i - \bar{x}) y_i - \sum_{i=1}^n (x_i - \bar{x}) \bar{y}}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{\sum_{i=1}^n (x_i - \bar{x}) y_i - \bar{y} \sum_{i=1}^n (x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$b = \frac{\sum_{i=1}^n (x_i - \bar{x}) y_i - \bar{y} \left[\sum_{i=1}^n x_i - \sum_{i=1}^n \bar{x} \right]}{\sum_{i=1}^n (x_i - \bar{x})^2} \Rightarrow 1$$

$$\therefore \bar{x} = \frac{\sum_{i=1}^n x_i}{n} \quad \therefore n\bar{x} = \sum_{i=1}^n x_i \Rightarrow 2 \quad \therefore \sum_{i=1}^n \bar{x} = n\bar{x} \Rightarrow \textcircled{3}$$

apply 2 and 3 in 1

$$b = \frac{\sum_{i=1}^n (x_i - \bar{x}) y_i - [n\bar{x} - n\bar{x}]}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{\sum_{i=1}^n (x_i - \bar{x}) y_i - 0}{\sum_{i=1}^n (x_i - \bar{x})^2} \Rightarrow \textcircled{4}$$

$$\therefore y_i = \beta x_i + \alpha + \epsilon_i$$

$$\therefore b = \frac{\sum_{i=1}^n (x_i - \bar{x}) \cdot (\beta x_i + \alpha + \epsilon_i)}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{\sum_{i=1}^n [\beta (x_i - \bar{x}) x_i + \alpha (x_i - \bar{x}) + \epsilon_i (x_i - \bar{x})]}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$b = \frac{\beta \sum_{i=1}^n (x_i - \bar{x}) \cdot x_i + \alpha \sum_{i=1}^n (x_i - \bar{x}) + \sum_{i=1}^n \epsilon_i (x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2} \Rightarrow \textcircled{5}$$

As we know and proof from (4) that $\sum_{i=1}^n (x_i - \bar{x}) = 0$

we will apply it in 5

$$b = \frac{\beta \sum_{i=1}^n (x_i - \bar{x}) \cdot x_i + \alpha[0] + \sum_{i=1}^n \epsilon_i (x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$b = \frac{\beta \sum_{i=1}^n (x_i - \bar{x}) x_i + \sum_{i=1}^n \epsilon_i (x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$b = \beta \frac{\sum_{i=1}^n (x_i - \bar{x}) x_i}{\sum_{i=1}^n (x_i - \bar{x})^2} + \frac{\sum_{i=1}^n \epsilon_i (x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2} \Rightarrow (6)$$

$$\begin{aligned} \therefore \sum_{i=1}^n (x_i - \bar{x})^2 &= \sum_{i=1}^n (x_i - \bar{x})(x_i - \bar{x}) = \sum_{i=1}^n (x_i^2 - x_i \bar{x} - \bar{x} x_i + \bar{x}^2) \\ &= \sum_{i=1}^n x_i^2 - 2\bar{x} \sum_{i=1}^n x_i + \sum_{i=1}^n \bar{x}^2 \end{aligned}$$

$$= \sum_{i=1}^n x_i^2 - 2\bar{x} (n\bar{x}) + n\bar{x}^2 = \sum_{i=1}^n x_i^2 - n\bar{x}^2 = \sum_{i=1}^n x_i^2 - n\bar{x} \cdot \bar{x}$$

$$= \sum_{i=1}^n x_i^2 - n\bar{x} \frac{\sum_{i=1}^n x_i}{n} = \sum_{i=1}^n x_i^2 - \bar{x} \sum_{i=1}^n x_i = \sum_{i=1}^n (x_i^2 - \bar{x} x_i)$$

$$= \sum_{i=1}^n (x_i - \bar{x}) x_i \Rightarrow (7)$$

apply 7 in 6

$$b = \beta \frac{\sum_{i=1}^n (x_i - \bar{x}) x_i}{\sum_{i=1}^n (x_i - \bar{x}) x_i} + \frac{\sum_{i=1}^n (x_i - \bar{x}) \cdot \epsilon_i}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

complete ϵ_i

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$$b = \beta + \frac{\sum_{i=1}^n (x_i - \bar{x}) \cdot \epsilon_i}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

Not sure what you mean by iterated expectations; should give a reference. Here you can just state that $E[\sum(z)] = \sum(E[z])$ - the proof is trivial and does not need to be shown.

$$E(b) = E \left[\beta + \frac{\sum_{i=1}^n (x_i - \bar{x}) \cdot \epsilon_i}{\sum_{i=1}^n (x_i - \bar{x})^2} \right]$$

$$E(b) = E(\beta) + E \left[\frac{\sum_{i=1}^n (x_i - \bar{x}) \epsilon_i}{\sum_{i=1}^n (x_i - \bar{x})^2} \right]$$

From the law of Iterated expectations

$$E(b) = E[E(b|x)] \text{ which com } E(x) = E[E(x|y)]$$

$$E(b) = \beta + E \left[E \left[\frac{\sum_{i=1}^n (x_i - \bar{x}) \epsilon_i}{\sum_{i=1}^n (x_i - \bar{x})^2} \mid x \right] \right]$$

$$E(b) = \beta + E \left[\frac{E \left[\sum_{i=1}^n (x_i - \bar{x}) \epsilon_i \mid x \right]}{\sum_{i=1}^n (x_i - \bar{x})^2} \right]$$

$$E(b) = \beta + E \left[\frac{\sum_{i=1}^n (x_i - \bar{x}) \cdot E[\epsilon_i | x]}{\sum_{i=1}^n (x_i - \bar{x})^2} \right]$$

$\therefore E[\epsilon_i | x] = 0$ The expected value of the error is zero conditional on $x = 0$

$$\therefore E(b) = \beta + E \left[\frac{\sum_{i=1}^n (x_i - \bar{x}) (0)}{\sum_{i=1}^n (x_i - \bar{x})^2} \right]$$

$$E(b) = \beta + E \left[\frac{0}{\sum_{i=1}^n (x_i - \bar{x})^2} \right]$$

$$E(b) = \beta$$

$$\therefore \boxed{E(b) = \beta}$$

Another way

$$b = \frac{\sum_{i=1}^n x_i y_i - n \bar{x} \bar{y}}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2} = \frac{\sum_{i=1}^n x_i y_i - \frac{1}{n} \sum_{i=1}^n x_i \sum_{i=1}^n y_i}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2} \left\{ \begin{array}{l} \sum x_i = \bar{x} \\ \sum y_i = \bar{y} \end{array} \right.$$

$$b = \frac{\sum_{i=1}^n x_i y_i - \bar{x} \sum_{i=1}^n y_i}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2} = \frac{\sum_{i=1}^n (x_i y_i - \bar{x} y_i)}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$b = \frac{\sum_{i=1}^n (x_i - \bar{x}) y_i}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$E[b|x_i] = \frac{\sum_{i=1}^n (x_i - \bar{x}) E[y_i | x_i]}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$E[b|x_i] = \frac{\sum_{i=1}^n (x_i - \bar{x}) E[\beta x_i + \alpha + \epsilon_i | x_i]}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$E[b|x_i] = \frac{\sum_{i=1}^n (x_i - \bar{x}) [E[\beta x_i | x_i] + E[\alpha | x_i] + E[\epsilon_i | x_i]]}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$E[b|x_i] = \frac{\sum_{i=1}^n (x_i - \bar{x}) [\beta x_i + \alpha + 0]}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$E[b|x_i] = \frac{\sum_{i=1}^n (x_i - \bar{x}) (\beta x_i + \alpha)}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$E[b|x_i] = \frac{\sum_{i=1}^n (\beta x_i^2 + \alpha x_i - \beta x_i \bar{x} - \alpha \bar{x})}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$E[b|x_i] = \frac{\sum_{i=1}^n \alpha x_i + \sum_{i=1}^n \alpha \bar{x} + \beta \left[\sum_{i=1}^n (x_i^2 - x_i \bar{x}) \right]}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$\therefore \frac{\sum_{i=1}^n x_i}{n} = \bar{x}$$

$$\therefore E[b|x_i] = \frac{\alpha \left[\sum_{i=1}^n x_i - n \bar{x} \right] + \beta \left(\sum_{i=1}^n x_i^2 - \bar{x} \sum_{i=1}^n x_i \right)}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$E[b|x_i] = \frac{\alpha \left[\sum_{i=1}^n x_i - n \left(\frac{\sum_{i=1}^n x_i}{n} \right) \right] + \beta \left[\sum_{i=1}^n x_i^2 - n \bar{x}^2 \right]}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$E[b|x_i] = \frac{\alpha \left[\sum_{i=1}^n x_i - \sum_{i=1}^n x_i \right] + \beta \left[\sum_{i=1}^n x_i^2 - n \bar{x}^2 \right]}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$E[b|x_i] = \frac{\alpha [0] + \beta \left[\sum_{i=1}^n x_i^2 - n \bar{x}^2 \right]}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$E[b|x_i] = \frac{\beta \sum_{i=1}^n (x_i)^2 - n \bar{x}^2}{\sum_{i=1}^n (x_i)^2 - n \bar{x}^2}$$

$$E[b|x_i] = \beta$$

$$\therefore E[b|x_i] = E[b]$$

$$\therefore E[b] = \beta$$