

Econ616

Homework #4

Answer key to selected questions

Question 2

Model 1 $\log(Y) = \gamma + \delta \log(X) + \varepsilon$

Model 2 $\log(X) = \alpha + \beta \log(Y) + \varepsilon$

Define variables x : $\log(X)$ in mean deviation form and y : $\log(Y)$ in mean deviation form.

$$(b) \quad d = \frac{\sum_{i=1}^n x_i y_i}{\sum_{i=1}^n y_i^2} \quad \text{and} \quad b = \frac{\sum_{i=1}^n x_i y_i}{\sum_{i=1}^n x_i^2} \quad \text{and thus} \quad d/b = \frac{\sum_{i=1}^n x_i^2}{\sum_{i=1}^n y_i^2} \neq 1$$

(c) The regressor is assumed not to be correlated with an error term.

$$(d) \quad R^2 = \frac{\sum_{i=1}^n \hat{y}_i^2}{\sum_{i=1}^n y_i^2} = \frac{d^2 \sum_{i=1}^n x_i^2}{\sum_{i=1}^n y_i^2} = \frac{d \sum_{i=1}^n x_i y_i}{\sum_{i=1}^n y_i^2} = d \cdot b$$

(e) Define a variable $X^* = pX$, where X denotes E_n and apply the above formulas. The rest follows...

Question 3

Model 1 $E_n = \alpha + \beta \text{RGDP} + u$

Model 2 $E_n^* = \beta \text{RGDP}^* + u^*$

Model 2 is Model 1 in mean deviation form.

Define $Y = E_n$ and $X = \text{RGDP}$, small x and y denote respective variables in mean deviation form.

Solution

$$(a) \quad \text{Model 1 OLS regression gives } b_1 = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_{i=1}^n (Y_i - \bar{Y})^2} = \frac{\sum_{i=1}^n x_i y_i}{\sum_{i=1}^n y_i^2}$$

$$\text{Model 2 OLS through the origin gives } b_2 = \frac{\sum_{i=1}^n x_i y_i}{\sum_{i=1}^n y_i^2}$$

Thus $b_1 = b_2$.

$$(b) \quad \text{Model 1 } R_1^2 = \frac{\sum_{i=1}^n (\hat{Y}_i - \bar{Y})^2}{\sum_{i=1}^n (Y_i - \bar{Y})^2} = \frac{\sum_{i=1}^n \hat{y}_i^2}{\sum_{i=1}^n y_i^2}$$

$$\text{Model 2 } R_2^2 = \frac{\sum_{i=1}^n \hat{y}_i^2}{\sum_{i=1}^n y_i^2}$$

Since $b_1 = b_2$ predicted y will be the same in both models thus $R_1^2 = R_2^2$.

Question 4

Model 1 $E_n = \alpha + \beta \text{RGDP} + u$

Model 2 $E_n^* = \beta \text{RGDP}^* + u^*$

Model 3 $E_n = \beta \text{RGDP} + u$

Define $Y = E_n$ and $X = \text{RGDP}$, small x and y denote respective variables in mean deviation form.

$$(a) \quad \text{Model 3 OLS through the origin gives } b_3 = \frac{\sum_{i=1}^n X_i Y_i}{\sum_{i=1}^n Y_i^2} \neq b_1 = b_2$$

$$(b) \quad \text{Model 3 } R_3^2 = \frac{\sum_{i=1}^n \hat{Y}_i^2}{\sum_{i=1}^n Y_i^2} \quad \text{First, note } R_1^2 = R_2^2 \neq R_3^2. \quad \text{Second, note that } TSS_3 = \sum_{i=1}^n Y_i^2$$

and $TSS_1 = TSS_2 = \sum_{i=1}^n y_i^2$ and thus $TSS_1 = TSS_2 \neq TSS_3$. The coefficients of determination are not comparable.

Question 5

Model $\ln(C) = \beta_1 + \beta_2 \ln(P) + \beta_3 \ln(Y) + u$

Solution

(b) $\beta_2 + \beta_3 = -1$ thus $\beta_3 = -(1 + \beta_2)$. One may rewrite the regression equation as follows

$$\ln(C) = \beta_1 + \beta_2 \ln(P) + \beta_3 \ln(Y) + u$$

$$\ln(C) = \beta_1 + \beta_2 \ln(P) - (1 + \beta_2) \ln(Y) + u$$

$$\ln(C) + \ln(Y) = \beta_1 + \beta_2 \ln(P) - \beta_2 \ln(Y) + u$$

$$\ln(C \cdot Y) = \beta_1 + \beta_2 \ln(P/Y) + u$$

Thus the restriction is built into the model.

(c) Built into the model one restriction at a time then reestimate the model and construct a F- test.

(i) $\beta_3 = 0$ and the model becomes

$$\ln(C) = \beta_1 + \beta_2 \ln(P) + u$$

(ii) see Question 5 (b)

(iii) $\beta_3 = 0$ and $\beta_2 = -1$

$$\ln(C) = \beta_1 + \beta_2 \ln(P) + \beta_3 \ln(Y) + u$$

$$\ln(C) = \beta_1 - \ln(P) + u$$

$$\ln(C \cdot P) = \beta_1 + u$$