

Task 1

$$\log w_i = \alpha_1 + \alpha_2 S_i + \varepsilon_i$$

$$\hat{\alpha}_2 > 0 \text{ \& signif}$$

- omitted relevant Z_i

problem \downarrow

$$\text{cov}(S_i, Z_i) \neq 0 \quad \text{cov}(\log w_i, Z_i) \neq 0$$

Z_i is connected with w_i

Z_i : country; exp; abilities;

- $\rightarrow \log w_i = \alpha_1 + \alpha_2 S_i + \alpha_3 C_i + u_i$

$E(u_i | S_i, C_i) = 0$

\downarrow

$\text{cov}(u_i, S_i) = 0$
 $\text{cov}(u_i, C_i) = 0$

$$\rightarrow \log w_i = \alpha_1 + \alpha_2 S_i + \varepsilon_i$$

$$\hat{\alpha}_2 = \frac{\sum (\log w_i - \overline{\log w})(S_i - \bar{S})}{\sum (S_i - \bar{S})^2} = \frac{s \text{cov}(\log w_i, S_i)}{s \text{var}(S_i)} =$$

$$= \frac{\text{sample cov}(\log w_i, S_i)}{\text{sample var}(S_i)} \xrightarrow[p]{WLN} \frac{\text{cov}(\log w_i, S_i)}{\text{var}(S_i)}$$

$$E(\hat{\alpha}_2) = \frac{\text{cov}(\log w_i, S_i)}{\text{var}(S_i)} = \frac{\text{cov}(\alpha_1 + \alpha_2 S_i + \varepsilon_i, S_i)}{\text{var}(S_i)} =$$

population cov, var

$$= \frac{\text{cov}(\alpha_1, S_i) + \alpha_2 \text{cov}(S_i, S_i) + \text{cov}(\varepsilon_i, S_i)}{\text{var}(S_i)} =$$

$$= \alpha_2 + \frac{\text{cov}(\varepsilon_i, S_i)}{\text{var}(S_i)} = \alpha_2 + \frac{\text{cov}(u_i + \alpha_3 C_i, S_i)}{\text{var}(S_i)} =$$

$$= \alpha_2 + \frac{\text{cov}(u_i, S_i)}{\text{var}(S_i)} + \alpha_3 \frac{\text{cov}(C_i, S_i)}{\text{var}(S_i)} = \alpha_2 + \text{bias}$$

$$\parallel$$

$$0$$

$$\text{cov}(C_i, S_i) > 0,$$

$$\alpha_3 > 0$$

- $\hat{\alpha}_2 > 0$ signif
- ↑
- ↑ not useful
- bias

Task 2

$$\widehat{\ln price}_i = -8,43 + 1,33 \ln A_i - 0,17 AR_i$$

(0,6) (0,05) (0,13)

$n=430$

$R^2 = 0,336$

$SE(\hat{\beta}_2)$

a) Test joint sign.

$$\ln price_i = \beta_0 + \beta_1 \ln A_i + \beta_2 AR_i + u_i \rightarrow SSR_{UR}$$

$H_0: \beta_1 = \beta_2 = 0$

$H_1: \text{otherwise}$

Fstat

$\begin{cases} \beta_1=0 \\ \beta_2=0 \end{cases}$

$H_0: R^2_{UR} = R^2_R \leftarrow$

$\rightarrow H_1: R^2_{UR} > R^2_R \leftarrow$

$H_0: \beta_1 = \beta_2 \rightarrow \beta_1 - \beta_2 = 0$

$$t_{stat} = \frac{\hat{\beta}_1 - \hat{\beta}_2}{SE(\hat{\beta}_1 - \hat{\beta}_2)} =$$

$$= \frac{\hat{\beta}_1 - \hat{\beta}_2}{\sqrt{\widehat{Var}(\hat{\beta}_1 - \hat{\beta}_2)}} = \frac{\hat{\beta}_1 - \hat{\beta}_2}{\sqrt{\widehat{Var} \hat{\beta}_1 + \widehat{Var} \hat{\beta}_2 - 2 \widehat{Cov}(\hat{\beta}_1, \hat{\beta}_2)}}$$

$$\ln price_i = \beta_0 + v_i \rightarrow SSR_R =$$

$$F_{stat} = \frac{(SSR_R - SSR_{UR}) / 2}{SSR_{UR} / 430 - 3} \quad \begin{matrix} \nearrow \text{9 \# restri.} \\ \nearrow \frac{n-k}{n-k} \end{matrix}$$

$$\left(R^2 = 1 - \frac{SSR}{SST} = 1 - \frac{\sum \hat{u}_i^2}{\sum (y_i - \bar{y})^2} \Rightarrow SSR = (1 - R^2) \cdot SST \Rightarrow$$

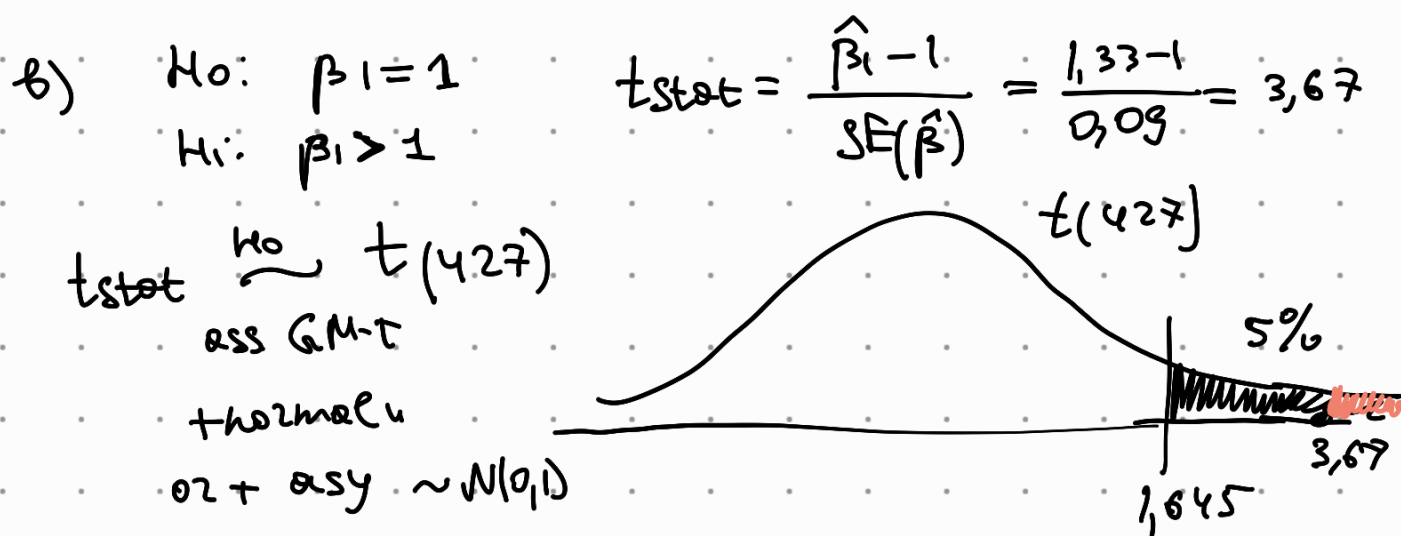
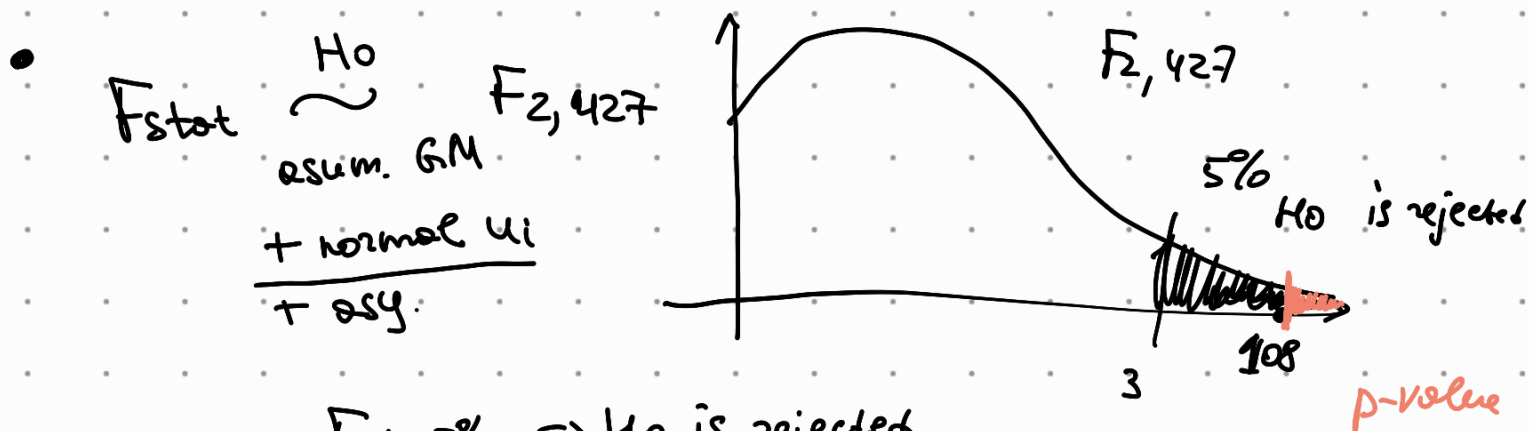
$$\Rightarrow F_{stat} = \frac{1 - R^2_R - (1 - R^2_{UR})}{1 - R^2_{UR}} \cdot \frac{427}{2} = \frac{R^2_{UR} - R^2_R}{1 - R^2_{UR}} \cdot \frac{427}{2}$$

• $F_{stat} = \frac{0,336 - R^2_R}{1 - 0,336} \cdot \frac{427}{2} = \frac{0,336}{1 - 0,336} \cdot \frac{427}{2} = 108$

$$\left[\ln price_i = \beta_0 + v_i \Rightarrow \hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} \Big|_{\hat{\beta}_1=0} = \bar{y} \Rightarrow \hat{y} = \bar{y} \right]$$

$\Rightarrow SSE = \sum (\hat{y}_i - \bar{y})^2 = \sum (\bar{y} - \bar{y})^2$

$$R^2 = \frac{SSR}{SST} = \frac{\sum (y_i - \bar{y})^2}{\sum (y_i - \bar{y})^2} = 0$$



Task 3 $y_i = \beta_0 + \beta_1 x_i + u_i$

a) without restrictions:

$$\sum \hat{u}_i^2 = \sum (y_i - \hat{y}_i)^2 = \sum (y_i - \underbrace{\beta_0 + \beta_1 x_i}_{\hat{y}_i})^2 \rightarrow \min_{\beta_0, \beta_1}$$

with restrictions $\beta_1 = 1$

$$\sum \tilde{u}_i^2 = \sum (y_i - \tilde{y}_i)^2 = \sum (y_i - \beta_0 - x_i)^2 \rightarrow \min_{\beta_0}$$

$$-2 \sum (y_i - \tilde{\beta}_0 - x_i) = 0 \Rightarrow \sum y_i - n\tilde{\beta}_0 - \sum x_i = 0$$

$$\bar{y} - \tilde{\beta}_0 - \bar{x} = 0$$

$$\tilde{\beta}_0 = \bar{y} - \bar{x}$$

$$\textcircled{b} \quad E(\tilde{\beta}_0) = E(\bar{y} - \bar{x}) = E\left(\frac{\sum y_i}{n}\right) - \bar{x} =$$

$$= E\left(\frac{\sum (\beta_0 + \beta_1 x_i + u_i)}{n}\right) - \bar{x} = E\left(\frac{n\beta_0 + \beta_1 \sum x_i + \sum u_i}{n}\right) - \bar{x} =$$

$$= \beta_0 + \beta_1 \bar{x} + E\left(\frac{\sum u_i}{n}\right) - \bar{x} = \beta_0 + (\beta_1 - 1)\bar{x} + \underbrace{E u_i}_{\substack{\text{iid} \\ 0}} =$$

$$\Rightarrow E(\tilde{\beta}_0) = \beta_0 + (\beta_1 - 1)\bar{x} \quad \left| \begin{array}{l} \beta_1 = 1 \\ \bar{x} = 0 \end{array} \right. = \beta_0 \Rightarrow$$

$\Rightarrow \tilde{\beta}_0$ unbiased

$$E(\tilde{\beta}_0) = \beta_0 + \underbrace{(\beta_1 - 1)\bar{x}}_{\text{bias}} \quad \left| \begin{array}{l} \bar{x} = 0 \end{array} \right. = \beta_0 \rightarrow \text{unbiased}$$

Lemma: If $E(\beta^1|x) = \beta$ (unbiased) and $\text{Var}(\beta^1|x) \rightarrow 0 \Leftrightarrow n \rightarrow \infty$

$$\Rightarrow \beta^1 \xrightarrow{P} \beta \quad (\text{consistent})$$

But: $\bar{y} \xrightarrow{P} E y_i = \mu_y$

$$\boxed{(\bar{y})^2 \xrightarrow{P} \mu_y^2}$$

$$E(\bar{y}^2) = \text{Var} \bar{y} + E^2 \bar{y} = \frac{\sigma_y^2}{n} + \mu_y^2 \neq \mu_y^2$$

↑ biased est. of μ_y^2

asym. unbiased (bias = $\frac{\sigma_y^2}{n} \rightarrow 0$)

$$E(\hat{\beta}_1 | x) = \frac{\text{scov}(y_i, x_i)}{\text{var}(x_i)} \xrightarrow{p} \frac{\text{cov}(y_i, x_i)}{\text{var}(x_i)} =$$

$$\Rightarrow \frac{\text{cov}(y_i, x_i)}{\text{var}(x_i)} = \frac{\text{cov}(\beta_0 + \beta_1 x_i + u_i, x_i)}{\text{var}(x_i)} = \beta_1$$

$$\boxed{\hat{\beta}_1 \xrightarrow{p} \beta_1}$$

