

$$t_0 = \frac{\Gamma \hat{\beta} - \Gamma \beta_0}{\sqrt{\hat{\sigma}^2 \Gamma (\mathcal{X}' \mathcal{X})^{-1} \Gamma'}} = \frac{\Gamma \hat{\beta} - \Gamma \beta_0}{se} \sim t(n-k) \quad \text{under } H_0: \Gamma \beta = \Gamma \beta_0 \\ \text{vs } H_1: \Gamma \beta \neq \Gamma \beta_0$$

Let $\Gamma \beta_0 = 0$, $\Gamma = [0 \dots 0 \ 1 \ 0 \dots 0]$

$$t_0 = \frac{\hat{\beta}_j}{se(\hat{\beta}_j)} \sim t(n-k) \quad \text{under } H_0$$

Under H_1 , however,

$$t_0 = \frac{\hat{\beta}_j - \beta_j}{se(\hat{\beta}_j)} + \frac{\beta_j}{se(\hat{\beta}_j)} = \frac{\hat{\beta}_j}{se(\hat{\beta}_j)} \quad \text{How does } t_0 \text{ behave?}$$

$$se(\hat{\beta}_j) = \sqrt{\hat{\sigma}^2 \Gamma (\mathcal{X}' \mathcal{X})^{-1} \Gamma'} = \sqrt{\hat{\sigma}^2 \Gamma \left(\frac{1}{n} \mathcal{X}' \mathcal{X} \right)^{-1} \cdot \frac{1}{n}}$$

$$\frac{1}{n} \mathcal{X}' \mathcal{X} = \frac{1}{n} \sum_{i=1}^n x_i' x_i \xrightarrow{P} E[x_i' x_i] \quad \text{by WLLN}$$

$$\Rightarrow \frac{\beta_j}{se(\hat{\beta}_j)} = \frac{\sqrt{n} \beta_j}{\underbrace{\sqrt{\hat{\sigma}^2 \Gamma \left(\frac{1}{n} \mathcal{X}' \mathcal{X} \right)^{-1} \Gamma'}}_{\rightarrow \text{constant}}} \rightarrow \pm \infty$$

Thus, under H_1 , $t_0 \rightarrow \pm \infty$. This is why the t -test is consistent. (Explains why we can reject the null when the null is not true).

Restricted Least Squares

MIT 2000, Q4)

unrestricted model: $\mathcal{Y} = \mathcal{X} \beta + \epsilon \Rightarrow \hat{\beta}_{OLS} = (\mathcal{X}' \mathcal{X})^{-1} \mathcal{X}' \mathcal{Y}$

restricted model: $\mathcal{Y} = \mathcal{X} \beta + \epsilon$, $\Gamma \beta = \gamma_0$

e.g. $\beta_1 = 0 \Leftrightarrow \Gamma = [1 \ 0 \ \dots \ 0]_{1 \times k}$, $\gamma_0 = 0$

Constrained optimization:

$$\tilde{\beta} = \hat{\beta}_R \equiv \underset{b: \Gamma b = \gamma_0}{\operatorname{argmin}} \sum_{i=1}^n (\bar{Y}_i - \bar{X}_i b)^2 \stackrel{\substack{\text{monotonic trans.} \\ \downarrow \text{ of obj. fn.}}}{=} \underset{b: \Gamma b = \gamma_0}{\operatorname{argmin}} \frac{1}{2} \sum_{i=1}^n (\bar{Y}_i - \bar{X}_i b)^2$$

$$\Rightarrow \min \frac{1}{2} (\bar{Y} - \bar{X}b)'(\bar{Y} - \bar{X}b) + \lambda (\Gamma b - \gamma_0)$$

$1 \times p$ $p \times k$ $k \times 1$

$$\Leftrightarrow \min \frac{1}{2} (\bar{Y}'\bar{Y} - \bar{Y}'\bar{X}b - b'\bar{X}'\bar{Y} + b'\bar{X}'\bar{X}b) + \lambda (\Gamma b - \gamma_0)$$

FOCs:

$$(b): \frac{1}{2} (-2\bar{X}'\bar{Y} + 2\bar{X}'\bar{X}\hat{\beta}_R) + \Gamma'\lambda' = 0$$

$$\Leftrightarrow \frac{1}{2} (-2\bar{X}'\bar{Y} + 2\bar{X}'\bar{X}\hat{\beta}_R) = -\Gamma'\lambda'$$

$$\Leftrightarrow \bar{X}'\bar{Y} - \bar{X}'\bar{X}\hat{\beta}_R = -\Gamma'\lambda'$$

$$\Leftrightarrow \bar{X}'\bar{X}\hat{\beta}_R = \bar{X}'\bar{Y} - \Gamma'\lambda'$$

$$\Leftrightarrow \hat{\beta}_R = (\bar{X}'\bar{X})^{-1}\bar{X}'\bar{Y} - (\bar{X}'\bar{X})^{-1}\Gamma'\lambda'$$

$$= \hat{\beta}_{UR} - (\bar{X}'\bar{X})^{-1}\Gamma'\lambda' \quad (1)$$

$$\text{Also, } \Gamma\hat{\beta}_R - \gamma_0 = 0$$

$$\Rightarrow \Gamma\hat{\beta}_{UR} - \Gamma(\bar{X}'\bar{X})^{-1}\Gamma'\lambda' - \gamma_0 = 0$$

$$\Leftrightarrow \lambda' = (\Gamma(\bar{X}'\bar{X})^{-1}\Gamma')^{-1}(\Gamma\hat{\beta}_{UR} - \gamma_0) \quad (2)$$

Substitute (2) into (1):

$$\hat{\beta}_R = \hat{\beta}_{UR} - (\bar{X}'\bar{X})^{-1}\Gamma'(\Gamma(\bar{X}'\bar{X})^{-1}\Gamma')^{-1}(\Gamma\hat{\beta}_{UR} - \gamma_0)$$

Examples:

1] Suppose $\Gamma = I_k$, $\gamma_0 = 0$

$$\Rightarrow \hat{\beta}_R = \hat{\beta}_{UR} - (\bar{X}'\bar{X})^{-1}\bar{X}'\bar{X}(\hat{\beta}_{UR}) = 0$$

2] Suppose $\Gamma = \gamma_0 = 0$

$$\Rightarrow \hat{\beta}_R = \hat{\beta}_{UR} - (\bar{X}'\bar{X})^{-1}0(0(\bar{X}'\bar{X})^{-1}0)^{-1}(0\hat{\beta}_{UR} - 0) = \hat{\beta}_{UR}$$

$$b) \hat{\epsilon}_R \equiv Y - X\hat{\beta}_R$$

$$\hat{\epsilon}_{UR} \equiv Y - X\hat{\beta}_{UR}$$

$$\Rightarrow \hat{\epsilon}_R - \hat{\epsilon}_{UR} = X(\hat{\beta}_{UR} - \hat{\beta}_R) \quad (1)$$

$$\text{Also, } \hat{\epsilon}_R' \hat{\epsilon}_R - \hat{\epsilon}_{UR}' \hat{\epsilon}_{UR} = (\hat{\epsilon}_R - \hat{\epsilon}_{UR})' (\hat{\epsilon}_R - \hat{\epsilon}_{UR}) + 2 \hat{\epsilon}_{UR}' (\hat{\epsilon}_R - \hat{\epsilon}_{UR})$$

$$\text{From (1):} \quad = (\hat{\epsilon}_R - \hat{\epsilon}_{UR})' (\hat{\epsilon}_R - \hat{\epsilon}_{UR}) + 2 \underbrace{\hat{\epsilon}_{UR}' X(\hat{\beta}_{UR} - \hat{\beta}_R)}_{=0 \text{ by orthogonality}}$$

$$= (\hat{\epsilon}_R - \hat{\epsilon}_{UR})' (\hat{\epsilon}_R - \hat{\epsilon}_{UR})$$

$$= (\hat{\beta}_{UR} - \hat{\beta}_R)' X' X (\hat{\beta}_{UR} - \hat{\beta}_R)$$

\Rightarrow From part (a):

$$\hat{\epsilon}_R' \hat{\epsilon}_R - \hat{\epsilon}_{UR}' \hat{\epsilon}_{UR} = (\Gamma \hat{\beta}_{UR} - \gamma_0)' (\Gamma (X'X)^{-1} \Gamma)' \Gamma (X'X)^{-1} (X'X)$$

$$\cdot (X'X)^{-1} \Gamma' (\Gamma (X'X)^{-1} \Gamma')^{-1} (\Gamma \hat{\beta}_{UR} - \gamma_0)$$

$$= (\Gamma \hat{\beta}_{UR} - \gamma_0)' (\Gamma (X'X)^{-1} \Gamma')^{-1} (\Gamma \hat{\beta}_{UR} - \gamma_0)$$

$$= \frac{\hat{\epsilon}_{UR}' \hat{\epsilon}_{UR} \cdot p F_0}{n-k}$$

$$\Rightarrow F_0 = \frac{(\hat{\epsilon}_R' \hat{\epsilon}_R - \hat{\epsilon}_{UR}' \hat{\epsilon}_{UR}) / p}{\hat{\epsilon}_{UR}' \hat{\epsilon}_{UR} / (n-k)}$$

$$\text{Remark: } F_0 = \frac{(R_{UR}^2 - R_R^2) / p}{(1 - R_{UR}^2) / (n-k)}$$

$$\text{since } R_{UR}^2 = 1 - \frac{\hat{\epsilon}_{UR}' \hat{\epsilon}_{UR}}{Y' M_U Y}$$

$$\Leftrightarrow \hat{\epsilon}_{UR}' \hat{\epsilon}_{UR} = (1 - R_{UR}^2) (Y' M_U Y)$$

STATA

$$se = \sqrt{\hat{\sigma}^2 \Gamma (\mathbf{X}'\mathbf{X})^{-1} \Gamma'}$$

$$edoc \Rightarrow \Gamma = [1 \ 0 \ 0 \ 0]$$

$$t_0 \equiv \frac{\hat{\beta}_j}{se(\hat{\beta}_j)}$$

For a hypothesis test of $\Gamma\beta = 0$, compare $|t_0|$ and $c_{0.05, (n-k)}^*$

What is the $P > |t|$ term?

Let $T \sim t(n-k)$. Then $(P > |t|) \equiv 2 \Pr[T > |t_0|]$ where t_0 is the realization, i.e. is a constant in this calculation

If $(P > |t|) < 0.05$, then we reject the null and conclude statistical significance.