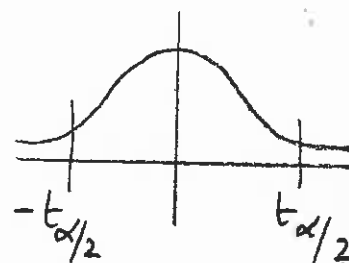


Hypothesis Testing in Classical Regression

Tests of Significance

$$H_0: \beta_j = 0 \quad (H_a: \beta_j \neq 0)$$

$$t_j = \frac{\hat{\beta}_j}{\sqrt{\hat{v}(\hat{\beta}_j)}} \sim t_{T-k}$$



$$|t_j| > t_{\alpha/2} \Rightarrow H_0 \text{ rejected}$$

Tests of Linear Restrictions

$$H_0: R\beta = \alpha; \quad R \text{ (} m \times k \text{) of rank } m$$

$$(H_a: R\beta \neq \alpha)$$

3 approaches: Wald, LM, LR

Wald

$$W = (\hat{\alpha} - \alpha)' [V(\hat{\alpha})]^{-1} (\hat{\alpha} - \alpha) \sim \chi_m^2$$

where $\hat{\alpha} = R\hat{\beta}$, $V(\hat{\alpha}) =$ true variance of $\hat{\alpha}$ (with true σ^2)

$$W > W_{\alpha}^* \Rightarrow H_0 \text{ rejected}$$

If σ^2 unknown, estimate by $\hat{\sigma}^2$

Then

$$F = (\hat{\alpha} - \alpha)' [\hat{V}(\hat{\alpha})]^{-1} (\hat{\alpha} - \alpha) / m \sim F_{m, T-k}$$

$$F > F_{\alpha}^* \Rightarrow \text{reject } H_0$$

(Here $\hat{V}(\hat{\alpha})$ is with $\hat{\sigma}^2$ instead of σ^2)

Also

$$F = \frac{(SS_r - SS) / m}{SS / (T-k)} \sim F_{m, T-k}$$

Lagrange Multiplier (LM)

$$LM_1 = \hat{\lambda}' [V(\hat{\lambda})]^{-1} \hat{\lambda} \sim \chi_m^2$$

where λ = multiplier associated with
($R\beta = \alpha$) in the constrained
estimation procedure

With σ^2 replaced by $\hat{\sigma}^2$,

$$LM_2 = \hat{\lambda}' [\hat{V}(\hat{\lambda})]^{-1} \hat{\lambda} \sim F_{m, T-k}$$

Note: If σ^2 known, $W = LM_1$

Likelihood Ratio (LR)

$$LR = -2 \ln \left(\frac{L_r^*}{L^*} \right) \sim \chi_m^2 \quad (\text{with } \sigma^2 \text{ known})$$

If σ^2 unknown, distribution (exact) of LR not known.

Asymptotically, $LR \sim \chi_m^2$

Note: If σ^2 known, $W = LM_1 = LR$

Zero Slopes test

$$H_0: \beta_2 = \dots = \beta_k = 0$$

$$\text{Restricted Model: } y_t = \beta_1 + \varepsilon_t$$

$$\text{Unrestricted Model: } y_t = \beta' x_t + \varepsilon_t$$

Then the F-test becomes

$$F = \frac{R^2 / (k-1)}{(1-R^2) / (T-k)} \sim F_{k-1, T-k}$$

(R^2 is that of the Unrestricted Model)