

Simple tests

$$\begin{aligned}\beta &\in \hat{\beta} \pm s.e.(\hat{\beta}) \cdot t_{crit} \\ F_{k,n-k-1} &= \frac{SSE/k}{RSS/(n-k-1)} = \frac{R^2/k}{(1-R^2)(n-k-1)} \\ t &= \frac{\hat{\beta}_2 - \beta_2^0}{s.e.(\hat{\beta}_2)}, df = n-2 \\ F_{q,n-k} &= \frac{(SSR_r - SSR_{ur})/q}{SSR_{ur}/(n-k)} = \frac{(R_{ur}^2 - R_r^2)/q}{(1-R_{ur}^2)/(n-k)} \\ &\text{r - restricted, ur - unrestricted,} \\ &\text{q - number of linear restrictions on coefs} \\ \chi^2(1) &= \frac{n}{2} \log \frac{SSR_2}{SSR_1} \\ &\text{Chow breakpoint test} \\ F_{k(m-1),n-mk} &= \frac{(SSR_0 - \sum SSR_i)/(k(m-1))}{\sum SSR_i/(n-mk)}\end{aligned}\quad (1)$$

White test heteroscedasticity test

$$H_0 : \text{homoscedastic} \quad (3)$$

- Regress against variables, their squares, and their cross-products, no dup
- Calculate nR^2 using R^2 from this regression. It is $\sim \chi_{n-1}$, where n - number of regressors, including constant.

Breusch-Pagan heteroscedasticity test

Needs less variables.

$$H_0 : \text{homoscedastic} \quad (4)$$

- Regress against variables
- Calculate nR^2 using R^2 from this regression. It is $\sim \chi_{n-1}$, where n - number of regressors, including constant.

Goldfeld-Quandt heteroscedasticity test

Split into three ranges: n_1 with smallest X , n_2 with highest X , and the rest.

$$F_{n_2-k, n_1-k} = \frac{RSS_2/(n_2-k)}{RSS_1/(n_1-k)} \quad (5)$$

Maximum likelihood estimation

$$\begin{aligned}f(Y_j) &= \frac{1}{\sigma\sqrt{2\pi}} e^{-0.5\left(\frac{Y_j - \beta_1 - \beta_2 X_j}{\sigma}\right)^2} \\ \log L &= \log \left(\prod f(Y_i) \right) = n \log \left(\frac{1}{\sigma\sqrt{2\pi}} \right) - \frac{\sigma^{-2}}{2} Z \\ R_{\text{pseudo}}^2 &= 1 - \frac{\log L}{\log L_0}; LR = 2(\log L - \log L_0) \\ &\text{where } L_0 \text{ is the likelihood with only intercept}\end{aligned}\quad (6)$$

Formulas

$$\begin{aligned}\hat{\beta}_1 &= \bar{Y} - \hat{\beta}_2 \bar{X}; \hat{\beta}_2 = \frac{\sum (X_i - \bar{X})(Y_i - \bar{Y})}{\sum (X_i - \bar{X})^2} \\ \hat{\beta}_2 &= \frac{\sum X_i Y_i}{\sum X_i^2} - \text{no intercept} \\ SST &= \sum (Y_i - \bar{Y})^2, \text{ sum of squares total} \\ SSE &= \sum (\hat{Y}_i - \bar{Y})^2, \text{ sum of squares explained} \\ SSR &= \sum (\hat{Y}_i - Y_i)^2, \text{ sum of squares} \\ R^2 &= \frac{SSE}{SST} = \frac{SST - SSR}{SST}; r_{Y, \hat{Y}} = \sqrt{R^2} - \text{correlation} \\ R_{adj}^2 &= 1 - \frac{SSR/(n-k)}{SST/(n-1)} \\ p &= F(Z); Z = \beta_1 + \beta_2 X_3 + \dots + \beta_k X_k \\ F(Z) &= \frac{1}{1 + e^{-Z}}; f(Z) = \frac{dp}{dZ} = \frac{1}{\sqrt{2\pi}} e^{-0.5Z^2} \\ Logit &= \prod Pr(Y = Y_j | X_j, \beta) - \text{max over } \beta \\ P(Y = 1) &= F(Y); P(Y = 0) = 1 - F(Y)\end{aligned}\quad (2)$$

Gauss-Markov conditions

1. $\mathbb{E}(u_i) = 0$
2. $\sigma_{u_i}^2 = \sigma_{u_j}^2$
3. u_i and u_j are independent $\iff i \neq j$
4. u_i is independent of X

4th condition violation

- Reasons: measurement errors, endogeneity
- Consequences: inconsistent OLS estimators, standard stats wrongly calculated, tests invalid
- Remedial measures: instrumental variables. Find another variable independently distributed with the disturbance term, but correlated with the endogenous explanatory variable
- Detection: Durbin-Wu-Hausman test (standard or Davidson-McKinnon version a a)

MLE tests

Test H_0 if all explanatory variables are 0.

$$LR \sim \chi_{k-1}^2 \quad (7)$$

where k - number of parameters estimated (and $k-1$ - number of explanatory variables). Also, F -test:

$$LR = 2(L_{ur} - L_r) \sim \chi_q^2 \quad (8)$$