

## Simple tests

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

## White test heteroscedasticity test

$$H_0 : \text{homoscedastic} \tag{3}$$

- 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.

$$H_0 : \text{homoscedastic} \tag{4}$$

- 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.

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

## Maximum likelihood estimation

$$\begin{aligned}
 f(Y_j) &= \frac{1}{\sigma\sqrt{2\pi}} e^{-0.5\left(\frac{Y_j - \beta_1 - \beta_2 X_j}{\sigma}\right)^2} \\
 \log L &= \log \left( \prod f(Y_i) \right) = n \log \left( \frac{1}{\sigma\sqrt{2\pi}} \right) - \frac{\sigma^{-2}}{2} Z
 \end{aligned} \tag{6}$$

$$R_{\text{pseudo}}^2 = 1 - \frac{\log L}{\log L_0}; LR = 2(\log L - \log L_0)$$

where  $L_0$  is the likelihood with only intercept

## Formulas

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

## 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)