

Testing Joint hypotheses

2 or more hypotheses to test -
both assumed to be all true under H_0 :

At least 1 of the hypotheses will be false
Then you reject H_0 :

Based on an F-Test.

Recall $\chi^2 \text{ is } \chi^2_m \text{ independent R.V.}$

$$F = \frac{\chi^2_{m-l_m}}{\chi^2_{l_m}} \sim F_{m, l_m}$$

so, F-Tests are basically ratios, just like t-test - except you are taking ratio of 2 chi-squares.

Basic Idea

Hypotheses represent possible restrictions that can be imposed on a model.

$$\epsilon_R = \beta_1 + \beta_2 A_t + \beta_3 P_t + \epsilon_+$$

$$H_0: \beta_2 = 0$$

$$H_A: \beta_2 \neq 0$$

The null says A has no effect. Imposing
this

$$\begin{aligned}\epsilon_R &= \beta_1 + 0 \cdot A_t + \beta_3 P_t + \epsilon_+ \\ &= \beta_1 + \beta_3 P_t + \epsilon_+\end{aligned}$$

OR

suppose A_t offsets P_t in the sense

$$H_0: \beta_2 + \beta_3 = 0$$

$$H_A: \text{not } H_0:$$

$$\text{Restriction} \Rightarrow \beta_2 = -\beta_3$$

$$\epsilon_R = \beta_1 + (-\beta_3 A_t) + \beta_3 P_t + \epsilon_+$$

$$= \beta_1 + \beta_3 (P_t - A_t) + \epsilon_+$$

create new variable

$$X_t = P_t - A_t$$

use as

$$\begin{aligned}\epsilon_R &= \beta_1 + \beta_3 X_t + \epsilon_+ \\ \text{to get } \beta_2 &= -\beta_3\end{aligned}$$

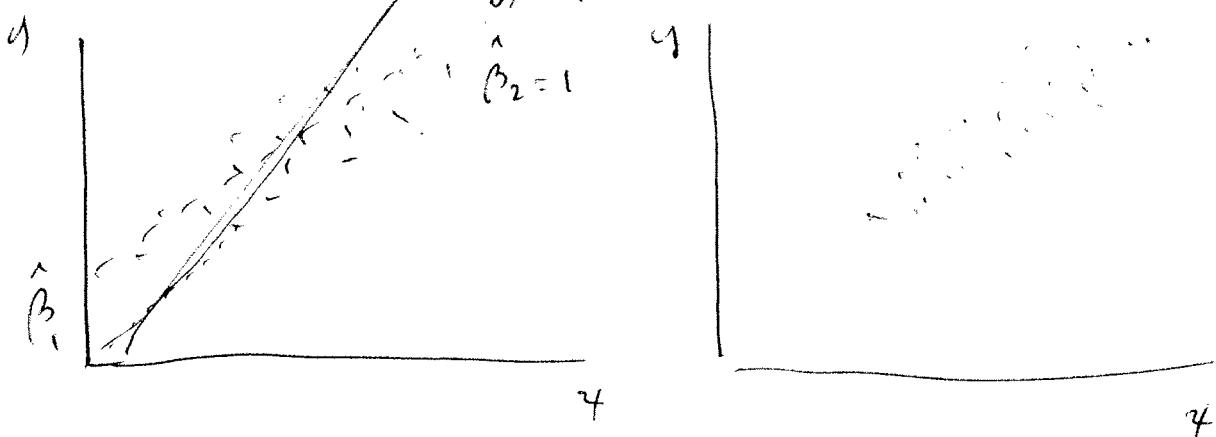
Using LS on Restricted Model is
called Restricted LS.

3

Test

The test of the restriction compares SSE from best and unrestricted models that are estimated by LS.

$$y = \beta_1 + \beta_2 x + e.$$



Suppose I restrict $\beta_1 = 0$

LS fit won't be as good when I impose restriction

$$SSE_R \geq SSE_U$$

The worse the restriction, the bigger the SSE_R will get.

If SSE_R is enough larger than SSE_U then we have evidence that rest. is not true.

Test Stat

4

$$F = \frac{(SSE_R - SSE_u) / J}{SSE_u / T - K}$$

SSE_R - sum of squared errors
from Test Model

SSE_u " " "
from untest.

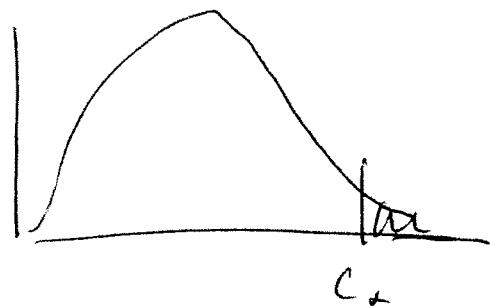
J # of hypotheses.

T sample size

K # of Parameters in untest
Model.

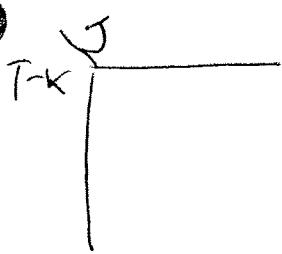
If H_0 true, $F \sim F_{J, T-K}$. Choose α and

find C_α



If $F > C_\alpha$ Reject.

Table 3 for Cut Value



Example

Overall - F Test

Test for overall Regression Significance

$$y_t = \beta_1 + \beta_2 x_{t2} + \cdots + \beta_k x_{tk} + \epsilon_t$$

If all slopes are zero \Rightarrow no Model

$$y_t = \beta_1 + \epsilon_t$$

$$H_0: \beta_2 = \beta_3 = \cdots = \beta_k = 0$$

$$H_A: \text{At least } 1 \beta_i \neq 0 \quad i=2, \dots, k$$

This amounts to $k-1$ restrictions

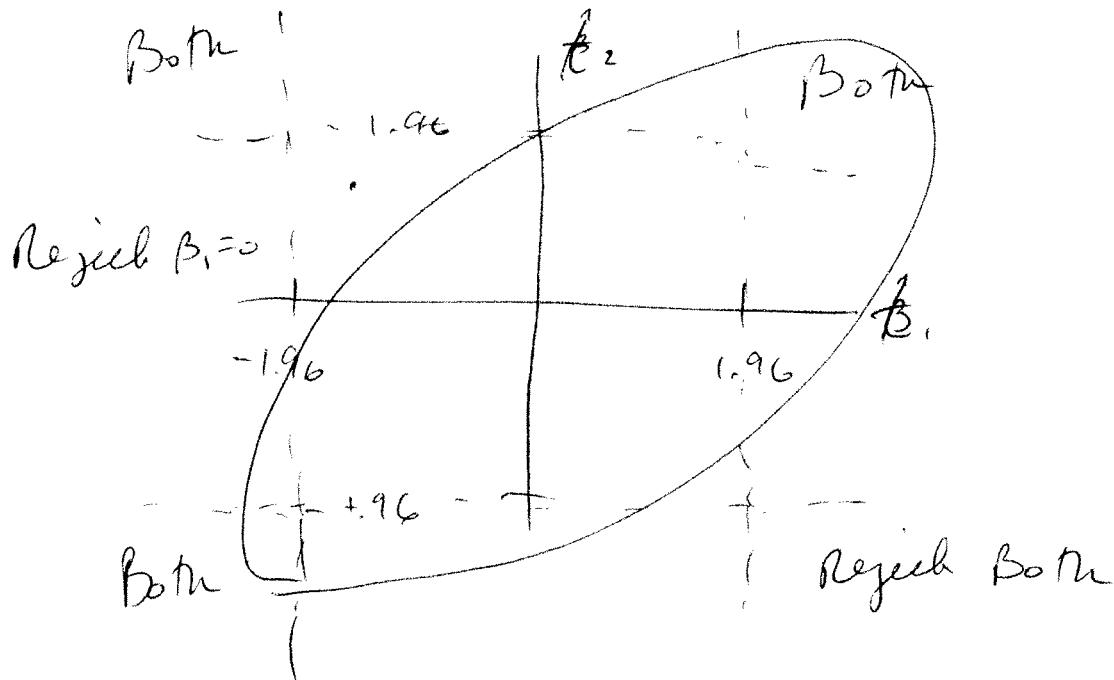
$$\begin{aligned} \text{Note: } J &= \# \text{params in overall} - \# \text{params in} \\ &\quad \text{rest.} \\ &= k - 1 \end{aligned}$$

NOT the same as individual t-tests

t

$$T - k > 120$$

Reject $\beta_2 = 0$



F stat is ellipse

and takes into account

$$\text{that } \text{Cov}(\hat{\beta}_1, \hat{\beta}_2) \neq 0$$

So, you could not reject $\beta_1, \beta_2 = 0$

By t-stats But be in
reject region of joint test.

or in not reject region of F
But Reject Both in t-tests.

Gretl

restrict

$$b_1 = 0$$

$$b_1 + 2 \cdot b_3 = 1$$

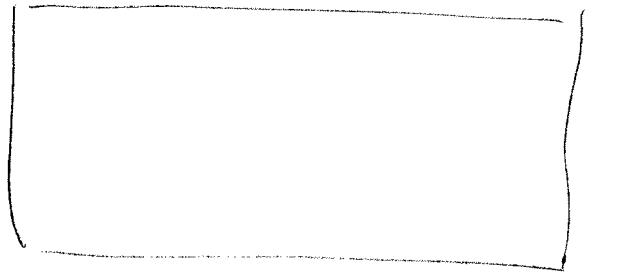
$$b_4 - b_5 = 0$$

end restrict

Estimate Model

Tests > Linear Restrictions

Gretl: linear restr.



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restriction set

$$1: b(\text{CPI}) = 0$$

$$2: b(\text{Pop}) + 2 \cdot b(\text{packPC}) = 1$$

$$3: b($$

Test Stat $F(3, 91) = 1625.5$

$P_{\text{cal}} = 6.95 \times 10^{-9}$

λ
Reject