

## TESTING LINEAR HYPOTHESES ABOUT $\beta$

Consider the following regression

$$Y_i = b_1 + b_2 X_{2i} + b_3 X_{3i} + b_4 X_{4i} + e_i$$

We want to test 3 linear restrictions regarding the 4  $b$ 's

The restrictions can be written in the following form

$$\underbrace{\begin{bmatrix} R_{11} & R_{12} & R_{13} & R_{14} \\ R_{21} & R_{22} & R_{23} & R_{24} \\ R_{31} & R_{32} & R_{33} & R_{34} \end{bmatrix}}_{(3 \times 4)} \underbrace{\begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix}}_{(4 \times 1)} = \underbrace{\begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix}}_{(3 \times 1)}$$

or

$$\underbrace{R}_{(3 \times 4)} \underbrace{b}_{(4 \times 1)} = \underbrace{r}_{(3 \times 1)}$$

The order of the matrix  $R$  is  $3 \times 4$ : 3 (the rows) is the number of restrictions and 4 is the number of coefficients to be estimated

## EXAMPLE 1

We want to test the following hypothesis

$$H_0 : b_2 = b_3 = b_4 = 0$$

Notice that we have three restrictions. These can be written as follows

$$\underbrace{\begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}}_{(3 \times 4)} \underbrace{\begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix}}_{(4 \times 1)} = \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}}_{(3 \times 1)}$$

Notice that the first equation is the first restriction:  $b_2 = 0$

## EXAMPLE 2

We want to test the following restrictions:

$$b_2 - b_3 = 0, \quad b_4 = 0$$

We can write the two restrictions as follows:

$$\underbrace{\begin{bmatrix} 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}}_{(2 \times 4)} \underbrace{\begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix}}_{(4 \times 1)} = \underbrace{\begin{bmatrix} 0 \\ 0 \end{bmatrix}}_{(2 \times 1)}$$

Notice that the first equation is the first restriction:  $b_2 - b_3 = 0$

### EXAMPLE 3

We want to test the hypothesis:  $b_2 + b_3 = 1$

If  $b_2$  and  $b_3$  are labor and capital elasticities in a production function, this formulation hypothesizes constant returns to scale

We can write this restriction as

$$\underbrace{\begin{bmatrix} 0 & 1 & 1 & 0 \end{bmatrix}}_{(1 \times 4)} \underbrace{\begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix}}_{(4 \times 1)} = \underbrace{\begin{bmatrix} 1 \end{bmatrix}}_{(1 \times 1)}$$

## GENERAL CASE

In general if we have  $q$  restrictions and  $k$  coefficients to be estimated we can write

$$\begin{matrix} R & b & = & r \\ (qxk) & (kx1) & & (qx1) \end{matrix}$$

or

$$\underbrace{\begin{bmatrix} R_{11} & R_{12} & \cdots & R_{1k} \\ R_{21} & R_{22} & \cdots & R_{2k} \\ \vdots & \vdots & \cdots & \vdots \\ R_{q1} & R_{q2} & \cdots & R_{qk} \end{bmatrix}}_{(qxk)} \underbrace{\begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_k \end{bmatrix}}_{(kx1)} = \underbrace{\begin{bmatrix} r_1 \\ r_2 \\ \vdots \\ r_q \end{bmatrix}}_{(qx1)}$$

Thus the first restriction is given by

$$R_{11}b_1 + R_{12}b_2 + \cdots + R_{1k}b_k = r_1$$

and the  $q$ th one by

$$R_{q1}b_1 + R_{q2}b_2 + \cdots + R_{qk}b_k = r_q$$

Under the null

$$\underset{(qx1)}{(R\beta - r)} \sim N[0, \sigma^2 \underset{(qxq)}{R(X'X)^{-1}R'}]$$

$R\beta - r$  are the estimated restrictions

*F* DISTRIBUTION

It can be shown that

$$\frac{\underset{(1xq)}{(R\beta - r)'} \underset{(qxq)}{[s^2 R(X'X)^{-1}R']^{-1}} \underset{(qx1)}{(R\beta - r)}}{q} \sim F(q, N-k)$$

## EXAMPLE 1

Assume that we have three coefficients to estimate:  $b_1, b_2, b_3$  and we want to test the null hypothesis that  $b_2 = 0$

Since we have only one restriction:

$$R\beta - r = \begin{bmatrix} 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \end{bmatrix} = \beta_2$$

Further, recall that the estimated variance-covariance matrix of  $\beta$  is given by

$$\widehat{Var}(\beta) = s^2(X'X)^{-1} = \begin{bmatrix} Var(\beta_1) & \dots & \dots \\ \dots & Var(\beta_2) & \dots \\ \dots & \dots & Var(\beta_3) \end{bmatrix}$$

Since  $R = \begin{bmatrix} 0 & 1 & 0 \end{bmatrix}$ , it follows that

$$Rs^2(X'X)^{-1}R' = \text{Var}(\beta_2)$$

Thus

$$\begin{aligned} & (R\beta - r)'[Rs^2(X'X)^{-1}R']^{-1}(R\beta - r)/q \\ &= \frac{\beta_2^2}{\text{Var}(\beta_2)} \sim F(1, N - k) \end{aligned}$$

Note that  $\frac{\beta_2^2}{\text{Var}(\beta_2)} = \left(\frac{\beta_2}{\widehat{se}(\beta_2)}\right)^2$ , and recall that  $\frac{\beta_2}{\widehat{se}(\beta_2)} \sim t(N - k)$



## EXAMPLE 2

We want to test the hypothesis:  $b_2 + b_3 = 1$ .

Recall that

$$\begin{aligned}\widehat{Var}(\beta) &= s^2(X'X)^{-1} \\ &= \begin{bmatrix} Var(\beta_1) & \dots & \dots \\ \dots & Var(\beta_2) & \dots \\ \dots & \dots & Var(\beta_3) \end{bmatrix}\end{aligned}$$

So we have only one restriction.

Notice that  $R = \begin{bmatrix} 0 & 1 & 1 \end{bmatrix}$ . Thus

$$\begin{aligned}Rs^2(X'X)^{-1}R' &= Var(\beta_2) + Var(\beta_3) + 2Cov(\beta_2, \beta_3) \\ &= Var(\beta_2 + \beta_3)\end{aligned}$$

In addition  $R\beta - r = \beta_2 + \beta_3 - 1$ .

Finally,

$$\begin{aligned}& (R\beta - r)'[Rs^2(X'X)^{-1}R']^{-1}(R\beta - r)/q \\ &= \frac{(\beta_2 + \beta_3 - 1)^2}{Var(\beta_2 + \beta_3)} \sim F(1, N - k)\end{aligned}$$

Two other forms of the  $F$  test

$$\frac{\text{ESS}/(k - 1)}{\text{RSS}/(N - k)} \sim F(k - 1, N - k),$$

Since  $r^2 = \text{ESS}/\text{TSS} = 1 - \frac{\text{RSS}}{\text{TSS}}$ , the above equation yields

$$\frac{r^2/k - 1}{(1 - r^2)/N - k} \sim F(k - 1, N - k)$$

Another form of the  $F$  test:

First, regress the restricted regression. For example, let the restriction be:  $b_2 + b_3 = 1$  or  $b_3 = 1 - b_2$

The unrestricted regression is

$$y_i = b_1 + b_2x_{2i} + b_3x_{3i} + e_i$$

whereas the restricted one is

$$\begin{aligned} y_i &= b_1 + b_2x_{2i} - b_2x_{3i} + x_{3i} + e_i \Rightarrow \\ (y_i - x_{3i}) &= b_1 + b_2(x_{2i} - x_{3i}) + e_i \end{aligned}$$

The  $F$  test statistic is given by

$$\begin{aligned} \frac{(\text{RSS}_R - \text{RSS}_{UR})/q}{\text{RSS}_{UR}/(N - k)} &\sim F(q, N - k), \text{ or} \\ \frac{(e'e_R - e'e_{UR})/q}{e'e_{UR}/(N - k)} &\sim F(q, N - k) \end{aligned}$$

**PROOFS FROM WEEK 7 ONWARDS,  
ONLY FOR EC5501**

$\beta, R\beta$

Recall that

$$\beta \sim N[b, \sigma^2(X'X)^{-1}]$$

The expected value of  $R\beta$  is:  $E(R\beta) = Rb$

Next we want to calculate the variance-covariance matrix of  $R\beta$ :

$$\begin{aligned} \text{Var}(R\beta) &= E[(R\beta - Rb)(R\beta - Rb)'] \\ &= E[R(\beta - b)(\beta - b)'R'] \\ &= RE[(\beta - b)(\beta - b)']R' \\ &= R\text{Var}(\beta)R' = \underset{(qxk)}{R} \underset{(kxk)}{\sigma^2(X'X)^{-1}} \underset{(kxq)}{R'} \end{aligned}$$

Thus

$$R\beta \sim N[Rb, \sigma^2 R(X'X)^{-1} R']$$

and

$$(R\beta - Rb) \sim N[0, \sigma^2 R(X'X)^{-1}R']$$

Rewrite:

$$(R\beta - Rb) \sim N[0, \sigma^2 R(X'X)^{-1}R']$$

Under the null hypothesis,  $H_0 : Rb = r$ . Thus under the null

$$\begin{matrix} (R\beta - r) \sim N[0, \sigma^2 R(X'X)^{-1}R'] \\ (qx1) \qquad \qquad \qquad (qxq) \end{matrix}$$

$R\beta - r$  are the estimated restrictions

## $\chi^2$ DISTRIBUTION

We have  $q$  estimated restrictions  $R\beta - r$ . Each follows the  $N$  distribution.

Their sum of squares is given by  $(R\beta - r)'(R\beta - r)$

Recall that  $Var(R\beta - r) = \sigma^2 R(X'X)^{-1}R'$

The sum of squares,  $(R\beta - r)'(R\beta - r)$ , standardized by the variance of  $R\beta - r$  is given by

$$(R\beta - r)'[\sigma^2 R(X'X)^{-1}R']^{-1}(R\beta - r) \sim \chi^2(q)$$

and it follows the  $\chi^2(q)$

## $F$ DISTRIBUTION

It can also be shown that the RSS divided by  $\sigma^2$  follows the  $\chi^2(N - k)$

$$\frac{\widehat{e}'\widehat{e}}{\sigma^2} \sim \chi^2(N - k)$$

It can also be shown that the ratio of two  $\chi^2$  distributions with  $q$  and  $N - k$  degree of freedoms divided by  $\frac{q}{N - k}$  follows the  $F$  distribution with  $q$  and  $N - k$  degrees of freedom

That is

$$\begin{aligned} & \frac{\begin{matrix} (R\beta - r)' & [\sigma^2 R(X'X)^{-1}R']^{-1} & (R\beta - r) \\ (1xq) & (qxq) & (qx1) \end{matrix}}{\frac{\widehat{e}'\widehat{e}}{\sigma^2} \frac{q}{N - k}} \\ &= \frac{\begin{matrix} (R\beta - r)' & [R(X'X)^{-1}R']^{-1} & (R\beta - r) \\ (1xq) & (qxq) & (qx1) \end{matrix}}{s^2 q} \sim F(q, N - k) \end{aligned}$$

since  $\frac{\widehat{e}'\widehat{e}}{N - k} = s^2$ .