

$$TSS = \underset{\downarrow}{ESS} + \underset{\uparrow}{RSS}$$

(II) To test: $H_0: \beta_2 = \beta_3 = \dots = \beta_k = 0$ vs $H_1: \text{at least one } \beta_i \neq 0, i=2,3,\dots,k$

Obj: To test whether the set of explanatory variables used to model y is actually meaningful or not.

Recall from ANOVA: $ESS = \hat{y}'\hat{y} = f(\hat{\beta}_2, \hat{\beta}_3, \dots, \hat{\beta}_k)$

If H_0 is false \Rightarrow we expect ESS to be high [RSS to be low]

If H_0 is true \Rightarrow we expect ESS to be low [RSS to be high]

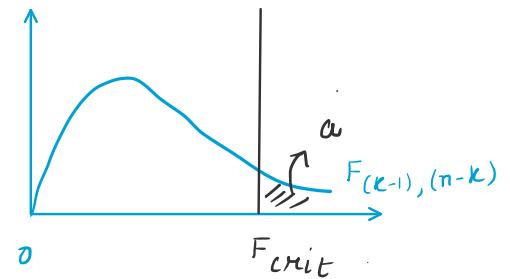
& we know:-

$$\frac{\sum e_i^2}{\sigma^2} \sim \chi^2_{(n-k)} \Rightarrow \frac{e'e}{\sigma^2} \sim \chi^2_{(n-k)}$$

$$\frac{\sum \hat{y}_i^2}{\sigma^2} \sim \chi^2_{(k-1)} \Rightarrow \frac{\hat{y}'\hat{y}}{\sigma^2} \sim \chi^2_{(k-1)}$$

$$F = \frac{\frac{\hat{y}'\hat{y}}{\sigma^2}/(k-1)}{\frac{e'e}{\sigma^2}/(n-k)} \sim F_{(k-1), (n-k)}$$

$$\bar{F} = \frac{\left\{ \begin{array}{l} ESS/(k-1) \\ RSS/(n-k) \end{array} \right\}}{\left\{ \begin{array}{l} ESS/(k-1) \\ RSS/(n-k) \end{array} \right\}} \sim F_{(k-1), (n-k)}$$



Note: If H_0 is true, F will be low

If H_0 is false, F will be high \Rightarrow we will reject H_0 if F is high.

Testing Rule: We will reject H_0 at $\alpha\%$ L.O.S if $F > F_{\text{crit}}$.

- B. Consider a sample data of size (n) collected from a firm which includes output level (q), Labour employed (L) & Capital (K). If the firm implements a Cobb-Douglas prodn fn: $q = A L^{\beta_2} K^{\beta_3}$

If the firm implements a Cobb-Douglas prodn fn: $q = AL^{\beta_2}K^{\beta_3}$ then estimate A, β_2, β_3 from the data & formulate a test to check if the prodn fn satisfies CRS.

$$q_i = A L_i^{\beta_2} K_i^{\beta_3}$$

q	L	K
q_1	L_1	K_1
q_2	L_2	K_2
\vdots	\vdots	\vdots
q_n	L_n	K_n

Take log: $\ln q_i = \underbrace{\ln A}_{\beta_1} + \beta_2 \underbrace{\ln L_i}_{x_{2i}} + \beta_3 \underbrace{\ln K_i}_{x_{3i}}$

True Model: $y_i = \beta_1 + \beta_2 x_{2i} + \beta_3 x_{3i} + u_i$

As the model is linear in parameters, use OLS to estimate the model.

Estimated Model: $\hat{y}_i = \hat{\beta}_1 + \hat{\beta}_2 x_{2i} + \hat{\beta}_3 x_{3i}$

$\hat{\beta}_2$: estimate of β_2

$\hat{\beta}_3$: estimate of β_3 .

$\hat{\beta}_1$: estimate of $\ln A \Rightarrow$ estimate of $A = e^{\hat{\beta}_1}$

where

$$y_i = \ln q_i$$

$$x_{2i} = \ln L_i$$

$$x_{3i} = \ln K_i$$

$$\beta_1 = \ln A$$

To check if the prodn fn satisfies CRS,

We will test: $H_0: \beta_2 + \beta_3 = 1$ vs $H_1: \beta_2 + \beta_3 \neq 1$.

$$H_0: \underbrace{\begin{bmatrix} 0 & 1 & 1 \end{bmatrix}}_R \underbrace{\begin{bmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \end{bmatrix}}_\beta = \underbrace{\begin{bmatrix} 1 \end{bmatrix}}_k$$

∴ Rewriting: $\left\{ H_0: R\beta = k \text{ vs } H_1: R\beta \neq k \right\} \Rightarrow \text{General Representation}$

Eq 2: True Model: $Y_i = \beta_1 + \beta_2 X_{2i} + \beta_3 X_{3i} + \beta_4 X_{4i} + u_i$

To test: $H_0: \begin{cases} \beta_1 - \beta_3 = 2 \\ \beta_2 + \beta_4 = 1 \end{cases}$ vs $H_1: H_0$ is not true.

$\begin{cases} \beta_1 - \beta_3 = 2 \\ \beta_2 + \beta_4 = 1 \end{cases} \rightarrow$ composite hypothesis.

$$\left(\begin{bmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \\ \beta_4 \end{bmatrix} \right)_{2 \times 4} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}_{2 \times 1} \quad \begin{array}{l} \text{4 Parameter,} \\ \text{2 condition} \end{array}$$

(III) Customized Hypothesis:

Consider the k -variable model: $Y = X\beta + U$

Suppose, there are ' q ' restrictions on the parameters that we want to test. [$q < k$].

To test: $H_0: R\beta = \kappa$ vs $H_1: R\beta \neq \kappa$.

We know: $U \sim N[0, \sigma^2 I]$.

and we know that $\hat{\beta}$ is linear in U .

$\therefore \hat{\beta} \sim N[E(\hat{\beta}), \text{Var}(\hat{\beta})]$.

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

Note: To do hypothesis testing on $R\beta$, we will use $(R\hat{\beta})$ ^{vector} to develop the test-statistic for the test.

HW $E[R\hat{\beta}] =$

$\text{Var}[R\hat{\beta}] =$