

# TEST OF SEVERAL LINEAR RESTRICTIONS (CONTD)

CONSIDER A PARTITION OF  $X$  AS  
 $X = [x_1 \ x_2]$  & THE CORRESPONDING  
 PARTITION OF  $\beta$  AS

$$\beta = \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix}$$

THUS THE REGRESSION MODEL  
 $y = X\beta + u$  BECOMES

$$y = x_1\beta_1 + x_2\beta_2 + u$$

NOW CONSIDER THE RESTRICTIONS  $\beta_2 = 0$ .

THUS,  $R = [0 \ I]$  & THE MATRIX FORM OF  
 THE RESTRICTIONS IS

$$R\beta = [0 \ I] \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} = \beta_2 = 0$$

SUPPOSE THAT THE RESTRICTED ESTIMATOR IS

$\beta^* = \begin{bmatrix} \beta_1^* \\ 0 \end{bmatrix}$  AND THE RESIDUALS FROM THE  
 RESTRICTED MODEL IS

$$e^* = y - X\beta^* \\ = y - X\hat{\beta} - X\beta^* + X\hat{\beta} = (y - X\hat{\beta}) - X(\beta^* - \hat{\beta})$$

HENCE, 
$$e^* = e - X(\beta^* - \hat{\beta})$$

FURTHER 
$$e^{*'}e^* = e'e + (\beta^* - \hat{\beta})'X'X(\beta^* - \hat{\beta}) - 2(\beta^* - \hat{\beta})'X'e$$

$$= e'e + (\beta^* - \hat{\beta})'X'X(\beta^* - \hat{\beta}) \quad \left[ \text{BECAUSE } X'e = 0 \right]$$

HENCE, 
$$e^{*'}e^* - e'e = (\beta^* - \hat{\beta})'X'X(\beta^* - \hat{\beta})$$

HOWEVER 
$$\beta^* = \hat{\beta} - (X'X)^{-1}R'[R(X'X)^{-1}R']^{-1}(R\hat{\beta} - q)$$

IN THE GENERAL CASE

HENCE 
$$\begin{aligned} & (\beta^* - \hat{\beta})'X'X(\beta^* - \hat{\beta}) \\ &= (R\hat{\beta} - q)' [R(X'X)^{-1}R']^{-1} R (X'X)^{-1} X'X (X'X)^{-1} R' [R(X'X)^{-1}R']^{-1} (R\hat{\beta} - q) \\ &= (R\hat{\beta} - q)' [R(X'X)^{-1}R']^{-1} (R\hat{\beta} - q) \sim \chi_p^2 \end{aligned}$$

(AS SEEN BEFORE)

THUS

$$\frac{[e^x'e^x - e'e] / p}{\frac{e'e}{n-k}} \sim F_{p, n-k}$$

HERE  $p$  IS THE # OF COLUMNS OF  $X_2$   
 (i.e. THE NUMBER OF VARIABLES EXCLUDED)

DEFINE

$$e'e = SSE_U \quad (\text{UNRESTRICTED})$$

$$e^x'e^x = SSE_R \quad (\text{RESTRICTED})$$

THEN

$$F = \frac{(SSE_R - SSE_U) / p}{\frac{SSE_U}{n-k}}$$

$$= \frac{\left( \frac{SSE_R}{SST} - \frac{SSE_U}{SST} \right) / p}{\frac{SSE_U / SST}{n-k}}$$

$$= \frac{[(1 - R_A^2) - (1 - R_U^2)] / p}{\frac{1 - R_U^2}{n-k}}$$

$$= \frac{(R_U^2 - R_A^2) / p}{\frac{1 - R_U^2}{n-k}} \sim F_{p, n-k}$$

HERE

$R_U^2$  = GOODNESS OF FIT OF UNRESTRICTED MODEL

$R_A^2$  = GOODNESS OF FIT OF RESTRICTED MODEL

R<sup>2</sup> & R<sup>2</sup> :

GOODNESS OF FIT OF A MODEL IS MEASURED BY THE

$$R^2 = 1 - \frac{SSE}{SST} = 1 - \frac{\sum e_i^2}{\sum (y_i - \bar{y})^2}$$

FOR A GIVEN SAMPLE THE VALUES OF THE DEPENDENT VARIABLE ARE GIVEN & DONOT CHANGE WITH THE CHOICE OF EXPLANATORY VARIABLES. THUS SST (i.e.  $\sum (y_i - \bar{y})^2$ ) IS CONSTANT. BUT AS THE NUMBER OF EXPLANATORY VARIABLES INCREASES SSE WILL EITHER DECLINE OR (AT WORST) REMAIN UNCHANGED. HENCE R<sup>2</sup> CAN ONLY INCREASE (STAY THE SAME) AS THE NUMBER OF X VARIABLES INCREASES. THERE IS, THUS, A BUILT IN BIAS IN FAVOR OF A MORE ELABORATE MODEL.

THEIR PROPOSED THE ADJUSTED R<sup>2</sup> :

$$\bar{R}^2 = 1 - \frac{\sum e_i^2 / n - k}{\sum (y_i - \bar{y})^2 / n - 1} = 1 - \frac{SSE / n - k}{SST / n - 1}$$

HERE BOTH SSE & SST ARE SCALED BY THE RESPECTIVE DEGREES OF FREEDOM

AS k INCREASES, R<sup>2</sup> WILL INCREASE ONLY IF SSE/n-k FALLS (SST/n-1 DOES NOT DEPEND ON k).

AS k INCREASES, SSE FALLS. BUT n-k ALSO FALLS. THE RATIO FALLS ONLY IF SSE DECLINES FASTER THAN n-k.

NOTE  $\frac{SSE}{n-k} = \hat{\sigma}^2$ . R<sup>2</sup> ~~FALLS~~ INCREASES WHEN  $\hat{\sigma}^2$  FALLS. IN GENERAL AS k INCREASES

R<sup>2</sup> MAY EITHER RISE OR FALL. THE MODEL WITH A HIGHER R<sup>2</sup> IS REGARDED

$$\bar{R}^2 = 1 - \left( \frac{SSE}{SST} \right) \left( \frac{n-1}{n-k} \right)$$

$$= 1 - (1-R^2) \left( \frac{n-1}{n-k} \right)$$

$$\Rightarrow 1 - \bar{R}^2 = (1-R^2) \left( \frac{n-1}{n-k} \right)$$

$$\Rightarrow \frac{1-R^2}{1-\bar{R}^2} = \frac{n-k}{n-1} < 1 \quad (\text{BECAUSE } k \geq 2)$$

$$\Rightarrow 1-R^2 < 1-\bar{R}^2 \Rightarrow \boxed{\bar{R}^2 < R^2}$$

NOTE :  $0 \leq R^2 \leq 1$   
 BUT  $\bar{R}^2$  CAN BE NEGATIVE.

RELATION BETWEEN  $R^2$ ,  $\bar{R}^2$ , & F:

RECALL THAT FOR THE SIGNIFICANCE OF OVERALL REGRESSION

$$F = \frac{R^2 / k - 1}{(1-R^2) / (n-k)} \Rightarrow \frac{1-R^2}{R^2} = \frac{n-k}{k} \cdot \frac{1}{F}$$

$$\frac{1}{R^2} = 1 + \left( \frac{n-k}{k} \right) \frac{1}{F}$$

ASIDE: t, F &  $R^2$  IN THE 2 VARIABLE REGRESSION

$$y_i = \beta_1 + \beta_2 x_i + u_i$$

$$\hat{y}_i = \hat{\beta}_1 + \hat{\beta}_2 x_i$$

$$\hat{y}_i - \bar{y} = \hat{\beta}_2 (x_i - \bar{x})$$

$$\sum (\hat{y}_i - \bar{y})^2 = (\hat{\beta}_2)^2 \sum (x_i - \bar{x})^2$$

FOR THE  $H_0: \beta_2 = 0$

$$t = \frac{\hat{\beta}_2}{se(\hat{\beta}_2)} \Rightarrow t^2 = \frac{(\hat{\beta}_2)^2}{\widehat{Var}(\hat{\beta}_2)}$$

$$= \frac{(\hat{\beta}_2)^2}{\left[ \frac{\sum e_i^2 / (n-2)}{\sum (x_i - \bar{x})^2} \right]} = \left[ \frac{(\hat{\beta}_2)^2 \sum (x_i - \bar{x})^2}{\sum e_i^2} \right]^{n-2}$$

$$t^2 = \left( \frac{SSR}{SSE} \right)^{n-2} = \left[ \frac{SSR/SST}{SSE/SST} \right]^{n-2}$$

$$= \left[ \frac{R^2}{1-R^2} \right]^{(n-2)}$$

$$\frac{1-R^2}{R^2} = \frac{n-2}{t^2}$$

$$\frac{1}{R^2} - 1 = \frac{n-2}{t^2} \Rightarrow \frac{1}{R^2} = \frac{n-2}{t^2} + 1 = \frac{(n-2) + t^2}{t^2}$$

$$\Rightarrow \boxed{R^2 = \frac{t^2}{t^2 + (n-2)}}$$

UNRESTRICTED MODEL :  $y = x_1 \beta_1 + x_2 \beta_2 + u_2$  (1)

RESTRICTED MODEL :  $y = x_1 \beta_1 + u$  (2)

SUPPOSE THAT (1) HAS  $k$  EXPLANATORY VARIABLES & (2) HAS  $k_1$  EXPLANATORY VARIABLES. THUS, (2) EXCLUDES

$k_2 = k - k_1$  VARIABLES.

$$H_0: \beta_2 = 0$$

$$F = \frac{(SSE_R - SSE_U) / R_2}{SSE_U / (n - K)}$$

REJECT  $H_0$  IF  $F > F_{R_2, n-K}^*$ .

NOW CONSIDER  $\bar{R}^2$  CRITERION.

SUPPOSE  $\bar{R}_U^2 > \bar{R}_R^2$ .

$$1 - \frac{SSE_U / (n - K)}{SST / (n - 1)} > 1 - \frac{SSE_R / (n - K_1)}{SST / (n - 1)}$$

$$\Rightarrow \frac{SSE_R / (n - K_1)}{SST / (n - 1)} > \frac{SSE_U / (n - K)}{SST / (n - 1)}$$

$$\Rightarrow \frac{SSE_R}{n - K_1} > \frac{SSE_U}{n - K}$$

$$\Rightarrow \frac{SSE_R}{SSE_U} > \frac{n - K_1}{n - K}$$

$$\Rightarrow \frac{SSE_R}{SSE_U} - 1 > \frac{n - K_1}{n - K} - 1$$

$$\Rightarrow \frac{SSE_R - SSE_U}{SSE_U} > \frac{K - K_1}{n - K} = \frac{R_2}{n - K}$$

$$\Rightarrow \frac{(SSE_R - SSE_U) / R_2}{SSE_U / (n - K)} > 1.$$

THUS  $\bar{R}_U^2 > \bar{R}_R^2 \Rightarrow F > 1$ .

HENCE THE  $\bar{R}^2$  CRITERION SELECTS THE UNRESTRICTED MODEL WHENEVER  $F > 1$  IRRESPECTIVE OF DEGREES OF FREEDOM.