

**Suggested Answers, Problem Set 4**  
**ECON 30331**

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1. The F-test is defined as  $\hat{F} = \frac{(SSE_r - SSE_u)/q}{SSE_u/(n-k-1)}$ . The  $R^2$  for the unrestricted model is by definition

$R_u^2 = 1 - (SSE_u / SST)$  so therefore,  $SSE_u = SST(1 - R_u^2)$  and likewise  $SSE_r = SST(1 - R_r^2)$ . Note that SST is the same in both the restricted and unrestricted models. Substituting these values into the definition of the F-test

$$\hat{F} = \frac{(SSE_r - SSE_u)/q}{SSE_u/(n-k-1)} = \frac{[SST(1 - R_r^2) - SST(1 - R_u^2)]/q}{(SST(1 - R_u^2))/(n-k-1)} = \frac{[(1 - R_r^2) - (1 - R_u^2)]/q}{(1 - R_u^2)/(n-k-1)} = \frac{(R_u^2 - R_r^2)/q}{(1 - R_u^2)/(n-k-1)}$$

2. The results from STATA are below. Given a null hypothesis that  $H_0: \beta_1 = a$ , the t-statistic is defined as

$$\hat{t} = \frac{\hat{\beta}_1 - a}{se(\hat{\beta}_1)}. \text{ We are given } a=1 \text{ and in the printout below, } \hat{\beta}_1 = 1.111685 \text{ and } se(\hat{\beta}_1) = 0.0881146 \text{ so}$$

$$\hat{t} = \frac{\hat{\beta}_1 - a}{se(\hat{\beta}_1)} = \frac{1.111685 - 1}{0.0881146} = 1.267$$

There are 49 degrees of freedom in this model and the 95% critical value of a t with 49 degrees of freedom is 2.010. In this case, we cannot reject the null the coefficient is 1.

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. use state_cig_data
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. keep if year==1988
(969 observations deleted)
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```
. reg retail_price state_tax
```

Source	SS	df	MS	Number of obs =	51
Model	3880.45446	1	3880.45446	F( 1, 49) =	159.17
Residual	1194.56711	49	24.3789206	Prob > F =	0.0000
				R-squared =	0.7646
				Adj R-squared =	0.7598
Total	5075.02157	50	101.500431	Root MSE =	4.9375

retail_price	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
state_tax	1.111685	.0881146	12.62	0.000	.9346125 1.288758
_cons	98.35109	1.855326	53.01	0.000	94.62267 102.0795

```
. test state_tax=1
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```
( 1) state_tax = 1
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F( 1, 49) = 1.61
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Prob > F = 0.2110

3. a) We are given the model  $y_i = \beta_0 + x_{1i}\beta_1 + x_{2i}\beta_2 + x_{3i}\beta_3 + x_{4i}\beta_4 + \epsilon_i$  and the null  $H_0: \beta_1=(1/2)\beta_2=3\beta_3$ . Note that  $2\beta_1=\beta_2$  and  $(1/3)\beta_1=\beta_3$  so substitute these values in above and collect like terms.

$$y_i = \beta_0 + x_{1i}\beta_1 + x_{2i}2\beta_1 + x_{3i}(1/3)\beta_1 + x_{4i}\beta_4 + \epsilon_i$$

$$y_i = \beta_0 + (x_{1i} + 2x_{2i} + (1/3)x_{3i})\beta_1 + x_{4i}\beta_4 + \epsilon_i$$

$$y_i = \beta_0 + (x_{5i})\beta_1 + x_{4i}\beta_4 + \epsilon_i$$

where  $x_{5i} = x_{1i} + 2x_{2i} + (1/3)x_{3i}$

- b) The null in this case is  $H_0: \beta_4=1-4\beta_1 - \beta_2-2\beta_3$  so substitute  $1-4\beta_1 - \beta_2-2\beta_3$  in for  $\beta_4$  and collect like term

$$y_i = \beta_0 + x_{1i}\beta_1 + x_{2i}\beta_2 + x_{3i}\beta_3 + x_{4i}\beta_4 + \epsilon_i$$

$$y_i = \beta_0 + x_{1i}\beta_1 + x_{2i}\beta_2 + x_{3i}\beta_3 + x_{4i}(1-4\beta_1 - \beta_2 - 2\beta_3) + \epsilon_i$$

$$y_i = \beta_0 + (x_{1i} - 4x_{4i})\beta_1 + (x_{2i} - x_{4i})\beta_2 + (x_{3i} - 2x_{4i})\beta_3 + x_{4i} + \epsilon_i$$

$$y_i - x_{4i} = \beta_0 + (x_{1i} - 4x_{4i})\beta_1 + (x_{2i} - x_{4i})\beta_2 + (x_{3i} - 2x_{4i})\beta_3 + \epsilon_i$$

$$y_i^* = \beta_0 + x_{1i}^*\beta_1 + x_{2i}^*\beta_2 + x_{3i}^*\beta_3 + \epsilon_i$$

where  $y_i^* = y_i - x_{4i}$ ,  $x_{1i}^* = (x_{1i} - 4x_{4i})$ ,  $x_{2i}^* = (x_{2i} - x_{4i})$ ,  $x_{3i}^* = (x_{3i} - 2x_{4i})$

4. In the simple bivariate regression  $y_i = \beta_0 + x_i\beta_1 + \epsilon_i$  we know the estimate for  $\beta_1$  can be written as

$$\hat{\beta}_1 = \frac{\sum_{i=1}^n (y_i - \bar{y})(x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

but in this case  $x_i=1$  or  $0$ . There are  $n$  observations in the sample and

$n_1 = \sum_{i=1}^n x_i$  observations for which  $x_i=1$  and  $n_0 = \sum_{i=1}^n (1-x_i)$  for which  $x_i=0$  and  $n_1+n_0=n$ . Recall also that

$$\bar{y}_1 = \frac{\sum_{i=1}^n y_i x_i}{\sum_{i=1}^n x_i} \text{ and } \bar{y}_0 = \frac{\sum_{i=1}^n y_i (1-x_i)}{\sum_{i=1}^n (1-x_i)}$$

Work with the numerator for  $\hat{\beta}_1$  first.

$$\sum_{i=1}^n (y_i - \bar{y})(x_i - \bar{x}) = \sum_{i=1}^n (y_i - \bar{y})x_i - \bar{y} \sum_{i=1}^n x_i = \sum_{i=1}^n y_i x_i - \bar{y} n_1$$

Note that  $\sum_{i=1}^n y_i x_i = n_1 \bar{y}_1$  and  $\bar{y}$ , the sample mean of  $y$ , is simply a weighted average of  $\bar{y}_1$  and  $\bar{y}_0$  where

$$\bar{y} = \frac{n_1}{n} \bar{y}_1 + \frac{n_0}{n} \bar{y}_0$$

Therefore, the numerator can be written as

$$n_1 \bar{y}_1 - n_1 \left( \frac{n_1}{n} \bar{y}_1 + \frac{n_0}{n} \bar{y}_0 \right) = n_1 \bar{y}_1 - \frac{n_1^2}{n} \bar{y}_1 - \frac{n_1 n_0}{n} \bar{y}_0 = \frac{n n_1 \bar{y}_1 - n_1^2 \bar{y}_1 - n_1 n_0 \bar{y}_0}{n} = \frac{n_1 (n - n_1) \bar{y}_1 - n_1 n_0 \bar{y}_0}{n}$$

and because  $n = n_1 + n_0$  then  $n_0 = n - n_1$  and the numerator equals

$$\frac{n_1 n_0}{n} (\bar{y}_1 - \bar{y}_0)$$

Now work with the denominator. Note that  $\sum_{i=1}^n (x_i - \bar{x})^2 = \sum_{i=1}^n (x_i - \bar{x}) x_i = \sum_{i=1}^n x_i^2 - \bar{x} \sum_{i=1}^n x_i$

Remember that  $n_1 = \sum_{i=1}^n x_i$  and since  $x_i = 1$  or zero then  $\sum_{i=1}^n x_i^2 = \sum_{i=1}^n x_i = n_1$  so

$$\sum_{i=1}^n x_i^2 - \bar{x} n_1 = n_1 - \frac{n_1}{n} (n_1) = n_1 - \frac{n_1^2}{n} = \frac{n_1 n - n_1^2}{n} = \frac{n_1 (n - n_1)}{n} = \frac{n_1 n_0}{n} \text{ and therefore}$$

$$\hat{\beta}_1 = \frac{\frac{n_1 n_0}{n} (\bar{y}_1 - \bar{y}_0)}{\frac{n_1 n_0}{n}} = (\bar{y}_1 - \bar{y}_0)$$

5. a. The confidence interval is by definition  $\hat{\beta}_1 \pm t_{\alpha/2}(t-k-1)se(\hat{\beta}_1)$ . Looking at the printout,  $\hat{\beta}_1 = 34.781$  and  $se(\hat{\beta}_1) = 13.244$ . The regressions has  $n=24$   $k=3$  and  $n-k-1=20$ . The appropriate critical value of the t-distribution is therefore 2.086. Therefore, the 95% confidence interval is  $34.781 \pm 2.086(13.244) = (7.15, 62.41)$ . Since the interval does not contain zero, we can reject the null.

- b. Given a null hypothesis that  $H_0: \beta_1 = a$ , the t-statistic is defined as  $\hat{t} = \frac{\hat{\beta}_1 - a}{se(\hat{\beta}_1)}$ . In the problem, we are

$$\text{given that } a=0, \hat{\beta}_1 = 34.781 \text{ and } se(\hat{\beta}_1) = 13.244 \text{ so } \hat{t} = \frac{\hat{\beta}_1 - a}{se(\hat{\beta}_1)} = \frac{34.781}{13.244} = 2.626. \text{ Since}$$

$|\hat{t}| > t_{\alpha/2}(n-k-1)$  we can reject the null that  $\beta_1 = 0$ .

- c. With a 99% confidence level, the critical value of the t-distribution with 20 degrees of freedom is 2.845. In this case,  $|\hat{t}| < t_{\alpha/2}(n-k-1)$  so we cannot reject the null.

- d. Panel A contains the unrestricted model and Panel B is the restricted model. The F-test is by

$$\hat{F} = \frac{(SSE_r - SSE_u) / q}{SSE_u / (n - k - 1)} \text{ and note that the denominator in the f-test is simple } \hat{\sigma}_\varepsilon^2 \text{ in the unrestricted model,}$$

which is label as the MSE or mean squared residual on the printout (46918.9833). In this case,  $SSE_u = 938379.666$ ,  $SSE_r = 1027703.99$ ,  $q=2$ ,  $n-k-1=20$ .

$$\hat{F} = \frac{(SSE_r - SSE_u) / q}{SSE_u / (n - k - 1)} = \frac{(1027703.99 - 938379.67) / 2}{46918.9833} = 0.95$$

The 95% critical value of the F-distribution with 2 and 20 degrees of freedom is 3.49, so since

$\hat{F} < F_\alpha$ , we cannot reject the null hypothesis.

6. A sample program that generates results and the log from this program are included on the web page.
- SSE=10,978.99,  $R^2=0.1193$
  - Males have 27.7 percent lower spending than female  
a one unit increase in the BMI will increase spending by 2.6%  
a 10% increase in income will reduce spending by  $(0.1)(-0.168)=-0.017$  or by 1.7 percent
  - $\hat{t}$  on income is -1.57 and the 95% critical value of the t-distribution with over 3000 degrees of freedom is 1.96 so since  $|\hat{t}| < t_{\alpha/2}(n-k-1)$  we cannot reject the null the true parameter is zero.
  - After running the unrestricted model, add the line  
  

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test midwest south west
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To perform the f-test. You will see the F-statistic is 3.41. If the null is correct, the test statistic is distributed as an F-distribution with 3 and infinite degrees of freedom and the 95% critical value is 2.60 so we can reject the null.
  - I must admit this is a stupid question on my part. Since you rejected the null at the 95% level, you can also reject the null at the 80% level.
- 7.
- The  $R^2$  measures the fraction of the variation in Y explained by the model  
In this case,  $R^2=1-(SSE/SST)=1-(3134.42/14370.98) = 0.782$
  - $\hat{\sigma}_\varepsilon^2 = SSE/(n-k-1)$ . In this case,  $SSE = 3134.42$ ,  $n=29$ ,  $k=3$ , so  $n-k-1=25$ , so  
 $\hat{\sigma}_\varepsilon^2 = SSE / (n - k - 1) = 3134.42 / 25 = 125.38$
  - 95% confidence interval is  $\hat{\beta}_1 \pm t_{\alpha/2}(n-k-1)[se(\hat{\beta}_1)]$ .  $\hat{\beta}_1=-0.449$ ,  $se(\hat{\beta}_1) = 0.164$  and with 25 degrees of freedom and  $\alpha=0.05$ , the appropriate critical value of the t-distribution is 2.060. So  $\hat{\beta}_1 \pm t_{\alpha/2}(n-k-1)[se(\hat{\beta}_1)] = -0.449 \pm 2.060(.164) = [-0.787, -0.111]$ . Since the 95% confidence does not contain 0, we CAN REJECT the null hypothesis.
  - $\hat{t} = \hat{\beta}_2 / se(\hat{\beta}_2) = 0.668 / 0.323 = 2.068$ . With 25 degrees of freedom and  $\alpha=0.01$ , the appropriate critical value of the t-distribution is 2.787 so, since  $|\hat{t}| < t_{\alpha/2}(n-k-1)$  at the 99% confidence level, one CANNOT REJECT the null that  $\beta_2=0$ .
  - $\hat{F} = \frac{(SSE_r - SSE_u) / q}{SSE_u / (n - k - 1)}$ .  $SSE_r=3673.06$ ,  $SSE_u=3134.42$ ,  $q=2$ ,  $n-k-1=25$ , so

$\hat{F} = \frac{(SSE_r - SSE_u) / q}{SSE_u / (n - k - 1)} = \frac{(3673.06 - 3134.42) / 2}{3134.32 / 25} = 2.15$ . If the null is correct, the F-test statistic is distributed as an F distribution with 2 and 25 degrees of freedom. The 95% critical value would then be 3.39 and since  $\hat{F} < F_\alpha$ , we CANNOR REJECT the null.