Questions and Answers on Hypothesis Testing and Confidence Intervals

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1. Using 25 observations and 5 regressors, including the constant term, a researcher estimates a linear regression model by OLS and finds

$$b_2 = 4.21$$
 , st.err. of $b_2 = 1.80$ $b_3 = -0.116$, st.err. of $b_3 = 0.0388$

Let t_{20df} be a random variable having a t distribution with n - K = 25 - 5 = 20 d.f.. Use the following probabilities:

$$Prob(t_{20df} > 1.75) = .05$$

 $Prob(t_{20df} > 2.086) = .025$

For each of the pairs of hypotheses given below, use a t-test to accept or reject H_0 at the 5% significance level. Show your reasoning.

- (a) $H_0: \beta_2 = 0$; $H_a: \beta_2 < 0$
- (b) $H_0: \beta_2 = 0$; $H_a: \beta_2 \neq 0$
- (c) $H_0: \beta_2 = 0$; $H_a: \beta_2 > 0$
- (d) $H_0: \beta_3 = 0$; $H_a: \beta_3 < 0$
- (e) $H_0: \beta_3 = 0$; $H_a: \beta_3 \neq 0$
- (f) $H_0: \beta_3 = 0$; $H_a: \beta_3 > 0$
- (g) $H_0: \beta_2 = 1$; $H_a: \beta_2 \neq 1$
- 2. Suppose that you estimate the model

$$y = \beta_1 \iota + \beta_2 x_2 + \beta_3 x_3 + \beta_4 x_4 + \epsilon \tag{1}$$

using 13 observations (n = 13), and compute the sum of squared residuals to be $SSE_1 = 126.1$. Then the following smaller models are estimated using the same 13 observations.

$$y = \beta_1 \iota + \beta_2 x_2 + \beta_3 x_3 + \epsilon, \quad \text{with } SSE_2 = 166.1$$
 (2)

$$y = \beta_1 \iota + \beta_2 x_2 + \beta_4 x_4 + \epsilon, \quad \text{with } SSE_3 = 204.4$$
 (3)

$$y = \beta_1 \iota + \beta_2 x_2 + \epsilon, \quad \text{with } SSE_4 = 221.4$$
 (4)

- (a) Compute F statistics for each of the following
 - (i) $H_0: \beta_3 = 0$; $H_a: \beta_3 \neq 0$
 - (ii) $H_0: \beta_4 = 0$; $H_a: \beta_4 \neq 0$
 - (iii) $H_0: \beta_4 = 0$; $H_a: \beta_4 \neq 0$ as in part (ii), but this time impose the restriction $\beta_3 = 0$ in both H_0 and H_a .
 - (iv) $H_0: \beta_3 = 0$ and $\beta_4 = 0$; $H_a: \beta_3 \neq 0$ and/or $\beta_4 \neq 0$

Parts (b) and (c) require the probabilities:

$$Prob(t_{9df} > 2.262) = .025$$

$$Prob(t_{10df} > 2.228) = .025$$

$$Prob(t_{9df} > 1.383) = .10$$

$$Prob(t_{9df} > 1.833) = .05$$

- (b) What are the absolute values of the t statistics for testing the hypotheses in parts (i), (ii) and (iii)? What would the t-test accept/reject conclusions be at the 5% level?
- (c) Suppose the OLS estimate of β_4 in equation (1) above is $b_4 = 6.13$.
 - (i) Use some of the above results to obtain a 90% confidence interval for β_4 . (Use the t-statistic from part (b) to get the standard error of b_4 .)
 - (ii) Use some of the above results to test $H_0: \beta_4 = 4$ against $H_a: \beta_4 > 4$ at the 10% significance level.

3. The model

$$y = \beta_1 \iota + \beta_2 x_2 + \beta_3 x_3 + \beta_4 x_4 + \epsilon \tag{5}$$

is estimated by OLS using 15 observations (n = 15), and compute the sum of squared residuals to be $SSE_1 = 55$. Then the following smaller models are estimated using the same 15 observations.

$$y = \beta_1 \iota + \beta_2 x_2 + \beta_3 x_3 + \epsilon, \quad \text{with } SSE_2 = 75$$
 (6)

$$y = \beta_1 \iota + \beta_2 x_2 + \beta_4 x_4 + \epsilon, \quad \text{with } SSE_3 = 72$$
 (7)

$$y = \beta_1 \iota + \beta_2 x_2 + \epsilon, \quad \text{with } SSE_4 = 79$$
 (8)

- (a) Using the above information, compute F statistics for each of the following hypotheses
 - (i) $H_0: \beta_4 = 0$; $H_a: \beta_4 \neq 0$
 - (ii) $H_0: \beta_4 = 0$; $H_a: \beta_4 \neq 0$, impose the restriction $\beta_3 = 0$ in both H_0 and H_a
 - (iii) $H_0:\beta_3=0$ and $\beta_4=0$; $H_a:\beta_3\neq 0$ and/or $\beta_4\neq 0$
- (b) Suppose you wish to test the restriction $\beta_2 + \beta_3 + \beta_4 = 1$.

- (i) Define R and q in such a way that this restriction can be expressed in the form $R\beta q = 0$.
- (ii) Write a model that could be estimated by OLS to get the restricted sum of squared residuals, SSE_R , that could be used in an F test of this restriction.
- 4. Let $\beta = \begin{bmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_5 \end{bmatrix}$. State whether each of the following restrictions or sets of restrictions are linear

or nonlinear. They are linear if an equivalent set of restrictions can be written in the form $R\beta - q = 0$ with matrix R and vector q being known constants. For the ones that are linear, give numerical values of R and q.

- (a) $\beta_4^3 = 27$
- (b) $\beta_2 \beta_3 = 1$
- (c) $\frac{\beta_2}{\beta_3} = 1$
- (d) $\frac{\beta_2}{\beta_3} = \beta_4$
- (e) $\beta_2 = \beta_3 = \beta_4$
- **5.** (a) Explain what is meant when a coefficient estimate is said to be "statistically significant at the 5% significance level".
 - (b) Briefly describe a situation in which a coefficient estimate is statistically significant at the 5% level, but is not significant in a practical or economic sense.
- **6.** A data set contains the following variables on individuals. They live in one of two cities, city #1 or city #2.

VARIABLE DESCRIPTION

- y log of hourly wage
- x equals 1 if the individual is an immigrant and x = 0 otherwise
- c equals 1 if the individual lives in city #2, otherwise c = 0
- $x \times c$ an interaction variable which equals one only if both x and c equal one.

The sums of squared residuals (SSEs) of four OLS regressions, each based on n=20 observations,

are

$$y = b_1 + b_2 x + e_1,$$
 with $SSE_1 = 59$ (9)

$$y = \tilde{b}_1 + \tilde{b}_3 c + e_2,$$
 with $SSE_2 = 54$ (10)

$$y = b_1^* + b_2^* x + b_3^* c + e_3, \quad \text{with } SSE_3 = 49$$
 (11)

$$y = b_1^{**} + b_2^{**}x + b_3^{**}c + b_4^{**}(x \times c) + e_4, \text{ with } SSE_4 = 46$$
 (12)

Compute F statistics that could be used for testing the following null hypotheses.

- (a) City of residence has no effect on log wages, using model (11) as the unrestricted model.
- (b) City of residence has no effect on log wages, using model (12) as the unrestricted model.
- (c) Immigration status has no effect on log wages, using model (12) as the unrestricted model.
- 7. Using 100 observations and 10 regressors, including the constant term, a researcher estimates a linear regression model by OLS and finds

$$b_2 = 3.1$$
 , st.err. of $b_2 = 1.7$

Let t_{90df} be a random variable having a t distribution with 90 d.f.. Use the following probabilities:

$$Prob(t_{90df} > 1.66) = .05$$

$$Prob(t_{90df} > 1.99) = .025$$

- (a) For each of the pairs of hypotheses given below, use a t-test to accept or reject H_0 at the 5% significance level. Show your reasoning.
 - (i) $H_0: \beta_2 = 0$; $H_a: \beta_2 < 0$
 - (ii) $H_0: \beta_2 = 0$; $H_a: \beta_2 \neq 0$
 - (iii) $H_0: \beta_2 = 0$; $H_a: \beta_2 > 0$
 - (iv) $H_0: \beta_2 = 1$; $H_a: \beta_2 \neq 1$
- (b) Which one of the four tests in part (a) has the smallest P-value?
- (c) Compute the 95% confidence interval for β_2 .
- **8.** (13 marks: 9 for a, 4 for b)

Four regression models are estimated by OLS using 20 observations. The equations and their

sums of squared residuals are

$$y_i = \beta_1 + \beta_2 w_i + \epsilon_i, \qquad SSE_1 = 400 \tag{13}$$

$$y_i = \beta_1 + \beta_2 w_i + \beta_3 g_i + \epsilon_i, \qquad SSE_2 = 360$$

$$(14)$$

$$y_i = \beta_1 + \beta_2 w_i + \beta_3 g_i + \beta_4 (w_i g_i) + \epsilon_i, \quad SSE_3 = 350$$
 (15)

$$y_i = \beta_1 + \beta_2 w_i + \beta_3 h_i + \epsilon_i,$$
 $SSE_4 = 340$ (16)

In equations (2) and (3), $g_i = 1$ if $w_i > 10$ and $g_i = 0$ if $w_i \le 10$.

In equation (4), $h_i = 0$ if $w_i < 10$ and $h_i = w_i - 10$ if $w_i \ge 10$. The β values can differ across models.

- (a) Compute F statistics for testing the following pairs of hypotheses.
 - (i) H_0 : The same linear relationship between y_i and w_i applies when $w_i \leq 10$ as when $w_i > 10$
 - H_a : The linear relationship between y_i and w_i jumps to a different level at $w_i = 10$, but the slope of the linear relationship is the same, regardless of whether $w_i \leq 10$ or $w_i > 10$
 - (ii) H_0 : The same linear relationship between y_i and w_i applies when $w_i \le 10$ as when $w_i > 10$
 - H_a : The linear relationship between y_i and w_i jumps to a different level at $w_i = 10$, and/or the slope of the linear relationship depends on whether $w_i \leq 10$ or $w_i > 10$
 - (iii) H_0 : The same linear relationship between y_i and w_i applies when $w_i \leq 10$ as when $w_i > 10$
 - H_a : The slope of the linear relationship between y_i and w_i depends on whether $w_i \leq 10$ or $w_i > 10$, but these two lines join at $w_i = 10$
- (b) Sketch four diagrams, one for each model, each with w_i on the horizontal axis and $E(y_i|w_i)$ on the vertical axis. On each one, draw an example of the type of relation between y_i and w_i that is allowed by that model.

Answers

1. (a) t = (4.21 - 0)/1.80 = 2.34

Reject if t < -1.725. Since t > -1.725, then accept H_0 at the 5% significance level

(b) t = 2.34

Reject if |t| > 2.086. Since t = 2.34, then |t| > 2.086, so reject H_0 at the 5% level

(c) t = 2.34

Reject if t > 1.725. Since t > 1.725, then reject H_0 at the 5% level

- (d) t = (-0.116 0)/.0388 = -2.99Reject if t < -1.725. Since t < -1.725, then reject H_0 at the 5% level
- (e) t = -2.99Reject if |t| > 2.086. Since t = -2.99, then |t| > 2.086, so reject H_0 at 5% level
- (f) t = -2.99Reject if t > 1.725. Therefore accept H_0 at the 5% level
- (g) t = (4.21 1)/1.80 = 1.78Reject if |t| > 2.086. Since t = 1.78, then |t| < 2.086, so accept H_0 at the 5% level
- (a) (i) $F = \frac{(SSE_3 SSE_1)/1}{SSE_1/(13-4)} = \frac{(204.4 126.1)/1}{126.1/9} = 5.59$ (ii) $F = \frac{(SSE_2 SSE_1)/1}{SSE_1/(13-4)} = \frac{(166.1 126.1)/1}{126.1/9} = 2.86$
 - (iii) Because of the restriction that $\beta_3=0$, the unrestricted model is now (11) and the restricted model is (12), so $F = \frac{(SSE_4 - SSE_3)/1}{SSE_3/(13-3)} = \frac{(221.4 - 204.4)/1}{204.4/10} = 0.83$
 - (iv) $F = \frac{(SSE_4 SSE_1)/2}{SSE_1/(13-4)} = \frac{(221.4 126.1)/2}{126.1/9} = 3.40$
 - (b) When testing one restriction, the t statistic and the F statistic are related by $t^2 = F$. So from the F statistics in (a), we know: in (i) |t| = 2.36; in (ii) |t| = 1.69; and in (iii) |t| = 0.91.
 - Reject H_0 when |t| > 2.262 at the 5% level for a two-tailed test with 9 d.f. Reject H_0 when |t| > 2.228 at the 5% level for a two-tailed test with 10 d.f. So the test in (i) rejects H_0 , and the tests in (ii) and (iii) accept H_0 .
 - (c) (i) From (b)(ii), |t| = 1.69 for testing $H_0: \beta_4 = 0$. Since $b_4 = 6.13$ and $t = (b_4 b_4)$ $0)/[\text{st.err}(b_4)]$ then $1.69 = 6.13/[\text{st.err}(b_4)]$ so that $\text{st.err}(b_4) = 3.63$. The 90% c.i. is $6.13 \pm t_{.05} \times 3.63 = 6.13 \pm 1.833 \times 3.63 = 6.13 \pm 6.65 = (-0.52, 12.78)$
 - (ii) $t = (b_4 4)/\text{st.err}(b_4) = (6.13 4)/3.63 = 0.59.$ Reject H_0 if $t > t_{.10}$ \Rightarrow Reject if t > 1.383. Accept H_0 at the 10% level, since t = 0.59 < 1.383.
- 3. (a) Use $F = \frac{(SSE_R SSE_U)/J}{SSE_U/(n-K)}$
 - (i) Compare (5) and (6). $F = \frac{(75-55)/1}{55/(15-4)} = 4$ (ii) Compare (7) and (8). $F = \frac{(79-72)/1}{72/(15-3)} = 7/6$ (iii) Compare (5) and (8). $F = \frac{(79-55)/2}{55/(15-4)} = 2.4$

 - (b) (i) $R = \begin{bmatrix} 0 & 1 & 1 & 1 \end{bmatrix}$; q = 1
 - (ii) Substitute out one of the β 's from the original model, then re-arrange the variables. There should be one less regressor since there is one restriction imposed. For example,

replacing β_4 with $1 - \beta_2 - \beta_3$ gives

$$y = \beta_1 \iota + \beta_2 x_2 + \beta_3 x_3 + (1 - \beta_2 - \beta_3) x_4 + \epsilon$$
$$y - x_4 = \beta_1 \iota + \beta_2 (x_2 - x_4) + \beta_3 (x_3 - x_4) + \epsilon$$

- **4.** (a) linear. $R = \begin{bmatrix} 0 & 0 & 0 & 1 & 0 \end{bmatrix}$; q = 3
 - (b) nonlinear
 - (c) linear. $R = \begin{bmatrix} 0 & 1 & -1 & 0 & 0 \end{bmatrix}$; q = 0
 - (d) nonlinear

(e) linear.
$$R = \begin{bmatrix} 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 \end{bmatrix}$$
; $q = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

- 5. (a) Let the coefficient be β and its estimate be $\hat{\beta}$. "Statistically significant at the 5% significance level" means "The null hypothesis $H_0: \beta = 0$ is rejected at the 5% significance level." Usually this means that $|\frac{\hat{\beta}}{\text{st.err.}(\hat{\beta})}| > \text{(critical value)}, \text{ or } P\text{-value} < .05, \text{ where } P\text{-value} = \text{Prob}(|t| > |\frac{\hat{\beta}}{\text{st.err.}(\hat{\beta})}|)$ and t represents a random variable with a distribution that one hopes is close to the one that the test statistic would have when H_0 is true. (This is the "null distribution".)
 - (b) Suppose that if β really was equal to $\hat{\beta}$ in the model, the economic implications would be much the same as if $\beta = 0$. In that case, $\hat{\beta}$ is not significant in an economic sense. For example, suppose $\hat{\beta}$ is estimating an income elasticity and $\hat{\beta} = .01$. Such a $\hat{\beta}$ still could be statistically significant, because its standard error might also be very small. In other words, $|\frac{\hat{\beta}}{\text{st.err.}(\hat{\beta})}|$ might be large enough for statistical significance, even when $\hat{\beta}$ is small enough to be economically insignificant.
- **6.** (a) Compare (11) with (9). $F = \frac{(59-49)/1}{49/(20-3)} = 3.47$
 - (b) Compare (12) with (9). $F = \frac{(59-46)/2}{46/(20-4)} = 2.26$
 - (c) Compare (12) with (10). $F = \frac{(54-46)/2}{46/(20-4)} = 1.39$
- 7. (a) t = 3.1/1.7 = 1.82
 - (i) Reject if t < -1.66. Therefore, do not reject H_0 .
 - (ii) Reject if |t| > 1.99. Therefore, do not reject H_0 .
 - (iii) Reject if t > 1.66. Therefore, reject H_0 .
 - (iv) Now t = (3.1 1)/1.7 = 1.24. Reject if |t| > 1.99. Therefore, do not reject H_0 .

- (b) The only test that rejected H_0 at the 5% level was the one in part (iii). So it is the only one with a P-value less than .05, and so it must have the smallest P-value of the four tests.
- (c) $b_2 \pm t_{.025} \times \text{st.err.}(b_2) = 3.1 \pm 1.99 \times 1.7 = 3.1 \pm 3.383 = (-0.283, 6.483)$
- **8.** (a) (i) Compare (1) and (2)

$$F = \frac{(400 - 360)/1}{360/(20 - 3)} = 1.89$$

(ii) Compare (1) and (3)

$$F = \frac{(400 - 350)/2}{350/(20 - 4)} = 1.14$$

(iii) Compare (1) and (4)

$$F = \frac{(400 - 340)/1}{340/(20 - 3)} = 3.00$$

(b) see next page

