## QUEEN'S UNIVERSITY AT KINGSTON Department of Economics

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#### ECONOMICS 351\* - Fall Term 2008

#### **Introductory Econometrics**

Fall Term 2008 MID-TERM EXAM: ANSWERS

M.G. Abbott

- DATE: Monday October 27, 2008.
- <u>TIME:</u> 80 minutes; 1:00 p.m. 2:20 p.m.
- <u>INSTRUCTIONS</u>: The exam consists of <u>FOUR</u> (4) questions. Students are required to answer ALL FOUR (4) questions.

Answer all questions in the exam booklets provided. Be sure your *student number* is printed clearly on the front of all exam booklets used. Your *name* is optional.

Do not write answers to questions on the front page of the first exam booklet.

**Please label clearly** each of your answers in the exam booklets with the appropriate number and letter.

#### Please write legibly.

Tables of percentage points of the t-distribution and F-distribution are given on the last two pages of the exam.

This exam is **CONFIDENTIAL**. This question paper must be submitted in its entirety with your answer booklet(s); otherwise your exam will not be marked.

# <u>MARKING</u>: The marks for each question are indicated in parentheses immediately above each question. **Total marks** for the exam **equal 100**.

GOOD LUCK!

## **<u>QUESTIONS</u>**: Answer ALL <u>FOUR</u> questions.

All questions pertain to the simple (two-variable) linear regression model for which the population regression equation can be written in conventional notation as:

$$\mathbf{Y}_{i} = \boldsymbol{\beta}_{0} + \boldsymbol{\beta}_{1} \mathbf{X}_{i} + \mathbf{u}_{i} \tag{1}$$

where  $Y_i$  and  $X_i$  are observable variables,  $\beta_0$  and  $\beta_1$  are unknown (constant) regression coefficients, and  $u_i$  is an unobservable random error term. The Ordinary Least Squares (OLS) sample regression equation corresponding to regression equation (1) is

$$Y_i = \hat{\beta}_0 + \hat{\beta}_1 X_i + \hat{u}_i$$
 (i = 1, ..., N) (2)

where  $\hat{\beta}_0$  is the OLS estimator of the intercept coefficient  $\beta_0$ ,  $\hat{\beta}_1$  is the OLS estimator of the slope coefficient  $\beta_1$ ,  $\hat{u}_i$  is the OLS residual for the i-th sample observation, and N is sample size (the number of observations in the sample).

## (15 marks)

1. Show that the OLS slope coefficient estimator  $\hat{\beta}_1$  is a linear function of the  $Y_i$  sample values. Stating explicitly all required assumptions, prove that the OLS slope coefficient estimator  $\hat{\beta}_1$  is an unbiased estimator of the slope coefficient  $\beta_1$ .

## **ANSWER to Question 1**:

## (5 marks)

• Show that the OLS slope coefficient estimator  $\hat{\beta}_1$  is a *linear* function of the Y<sub>i</sub> sample values.

$$\hat{\beta}_{1} = \frac{\sum_{i} x_{i} y_{i}}{\sum_{i} x_{i}^{2}} = \frac{\sum_{i} x_{i} (Y_{i} - \overline{Y})}{\sum_{i} x_{i}^{2}} = \frac{\sum_{i} x_{i} Y_{i}}{\sum_{i} x_{i}^{2}} - \frac{\overline{Y} \sum_{i} x_{i}}{\sum_{i} x_{i}^{2}}$$
$$= \frac{\sum_{i} x_{i} Y_{i}}{\sum_{i} x_{i}^{2}} \qquad \text{because } \sum_{i} x_{i} = 0$$
$$= \sum_{i} k_{i} Y_{i} \qquad \text{where } k_{i} = \frac{x_{i}}{\sum_{i} x_{i}^{2}}.$$

## **ANSWER to Question 1 -- continued:**

## (10 marks)

- Stating explicitly all required assumptions, prove that the OLS slope coefficient estimator  $\hat{\beta}_1$  is an unbiased estimator of the slope coefficient  $\beta_1$ .
- (1) Substitute for  $Y_i$  the expression  $Y_i = \beta_0 + \beta_1 X_i + u_i$  from the population regression equation (or PRE). (5 marks)

$$\begin{split} \hat{\beta}_{1} &= \sum_{i} k_{i} Y_{i} \\ &= \sum_{i} k_{i} (\beta_{0} + \beta_{1} X_{i} + u_{i}) \\ &= \sum_{i} (\beta_{0} k_{i} + \beta_{1} k_{i} X_{i} + k_{i} u_{i}) \\ &= \beta_{0} \sum_{i} k_{i} + \beta_{1} \sum_{i} k_{i} X_{i} + \sum_{i} k_{i} u_{i} \\ &= \beta_{1} + \sum_{i} k_{i} u_{i}, \end{split}$$
 since  $\sum_{i} k_{i} = 0$  and  $\sum_{i} k_{i} X_{i} = 1$ 

(2) Now take expectations of the above expression for  $\hat{\beta}_1$  conditional on the value  $X_i$  of X: (5 marks)

$$\begin{split} E(\hat{\beta}_{1}) &= E(\beta_{1}) + E[\sum_{i} k_{i} u_{i}] \\ &= \beta_{1} + \sum_{i} k_{i} E(u_{i} | X_{i}) \\ &= \beta_{1} + \sum_{i} k_{i} 0 \\ &= \beta_{1} \end{split} \qquad since \beta_{1} \text{ is a constant and the } k_{i} \text{ are nonstochastic} \\ &= \beta_{1} + \sum_{i} k_{i} 0 \\ &= \beta_{1} \end{split}$$

## (15 marks)

2. Give a general definition of the F-distribution. Starting from this definition, derive the F-statistic for the OLS slope coefficient estimator  $\hat{\beta}_1$ . State all assumptions required for the derivation.

## **ANSWER to Question 2**:

## (3 marks)

- *General definition of the F-distribution:* Consider the two random variables  $V_1$  and  $V_2$  such that
  - (1)  $V_1 \sim \chi^2[m_1]$ (2)  $V_2 \sim \chi^2[m_2]$ (3)  $V_1$  and  $V_2$  are *independent*.

Then the random variable

$$F = \frac{V_1/m_1}{V_2/m_2} ~ \sim ~ F[m_1, m_2]$$

where  $F[m_1, m_2]$  denotes the **F**-distribution with  $m_1$  numerator degrees of freedom and  $m_2$  denominator degrees of freedom.

• *Error Normality Assumption:* The random error term  $u_i$  is normally distributed with mean 0 and variance  $\sigma^2$ ; that is,

 $u_i \, \big| \, X_i \sim \, N \Big[ 0, \sigma^2 \Big] \, \text{for all} \, i \qquad \text{OR} \qquad u_i \, \, \text{is iid as} \, \, N \Big[ 0, \sigma^2 \Big]$ 

• Use **three implications** of the error normality assumption to derive  $F(\hat{\beta}_1)$ :

1. 
$$\hat{\beta}_1 \sim N[\beta_1, Var(\hat{\beta}_1)]$$
 where  $Var(\hat{\beta}_1) = \frac{\sigma^2}{\sum_{i=1}^N x_i^2} = \frac{\sigma^2}{\sum_{i=1}^N (X_i - \overline{X})^2}$  (1 mark)

Implications of normality of  $\hat{\beta}_1$ :

$$Z(\hat{\beta}_1) = \frac{\hat{\beta}_1 - \beta_1}{\operatorname{se}(\hat{\beta}_1)} \sim N[0, 1]$$

## **ANSWER to Question 2 -- continued:**

$$[Z(\hat{\beta}_{1})]^{2} = \frac{(\hat{\beta}_{1} - \beta_{1})^{2}}{Var(\hat{\beta}_{1})} = \frac{(\hat{\beta}_{1} - \beta_{1})^{2}}{\sigma^{2} / \sum_{i} x_{i}^{2}} = \frac{(\hat{\beta}_{1} - \beta_{1})^{2} \sum_{i} x_{i}^{2}}{\sigma^{2}} \sim \chi^{2}[1]$$
(1) (1 mark)

2. 
$$\frac{(N-2)\hat{\sigma}^2}{\sigma^2} \sim \chi^2[N-2] \implies \frac{\hat{\sigma}^2}{\sigma^2} \sim \frac{\chi^2[N-2]}{(N-2)}$$
 (2) (1 mark)

- 3. The estimators  $\hat{\beta}_1$  and  $\hat{\sigma}^2$  are statistically independent *OR* (1 mark) The statistics  $Z(\hat{\beta}_1)$  and  $\hat{\sigma}^2/\sigma^2$  are statistically independent.
- The *F*-statistic for  $\hat{\beta}_1$ . The F-statistic for  $\hat{\beta}_1$  is therefore the ratio of (1) to (2): (6 marks)

$$\begin{split} F(\hat{\beta}_{1}) &= \frac{\left(Z(\hat{\beta}_{1})\right)^{2}}{\hat{\sigma}^{2}/\sigma^{2}} \\ &= \frac{\left(\hat{\beta}_{1} - \beta_{1}\right)^{2} \left(\sum_{i} x_{i}^{2}\right)/\sigma^{2}}{\hat{\sigma}^{2}/\sigma^{2}} \\ &= \frac{\left(\hat{\beta}_{1} - \beta_{1}\right)^{2} \left(\sum_{i} x_{i}^{2}\right)}{\hat{\sigma}^{2}} \\ &= \frac{\left(\hat{\beta}_{1} - \beta_{1}\right)^{2}}{\hat{\sigma}^{2}/\sum_{i} x_{i}^{2}} \\ &= \frac{\left(\hat{\beta}_{1} - \beta_{1}\right)^{2}}{V\hat{a}r(\hat{\beta}_{1})} \qquad \text{since } \hat{\sigma}^{2}/\sum_{i} x_{i}^{2} = V\hat{a}r(\hat{\beta}_{1}). \end{split}$$

 $\square \quad \underline{\textit{Result}:} \quad \text{The F-statistic for } \hat{\beta}_1 \text{ takes the form}$ 

$$F(\hat{\beta}_{1}) = \frac{(\hat{\beta}_{1} - \beta_{1})^{2}}{\hat{\sigma}^{2} / (\sum_{i} x_{i}^{2})} = \frac{(\hat{\beta}_{1} - \beta_{1})^{2}}{V\hat{a}r(\hat{\beta}_{1})} \sim F[1, N-2].$$

## (32 marks)

3. A researcher is using data for a sample of 3240 female employees 25 years of age and over to investigate the relationship between employees' hourly wage rates  $Y_i$  (measured in *dollars per hour*) and their age  $X_i$  (measured in *years*). The population regression equation takes the form of equation (1):  $Y_i = \beta_0 + \beta_1 X_i + u_i$ . Preliminary analysis of the sample data produces the following sample information:

$$\begin{split} \mathbf{N} &= 3240 \qquad \sum_{i=1}^{N} y_{i}^{2} = 78434.97 \qquad \sum_{i=1}^{N} x_{i}^{2} = 25526.17 \qquad \sum_{i=1}^{N} x_{i}y_{i} = 3666.426 \\ &\sum_{i=1}^{N} Y_{i} = 34379.16 \qquad \sum_{i=1}^{N} X_{i} = 96143.00 \qquad \sum_{i=1}^{N} Y_{i}^{2} = 443227.1 \\ &\sum_{i=1}^{N} X_{i}^{2} = 2878451.0 \qquad \sum_{i=1}^{N} X_{i}Y_{i} = 1023825.0 \qquad \sum_{i=1}^{N} \hat{u}_{i}^{2} = 77908.35 \end{split}$$

where  $x_i \equiv X_i - \overline{X}$  and  $y_i \equiv Y_i - \overline{Y}$  for i = 1, ..., N. Use the above sample information to answer all the following questions. Show explicitly all formulas and calculations.

## (10 marks)

(a) Use the above information to compute OLS estimates of the intercept coefficient  $\beta_0$  and the slope coefficient  $\beta_1$ .

• 
$$\hat{\beta}_1 = \frac{\sum_i x_i y_i}{\sum_i x_i^2} = \frac{3666.426}{25526.17} = 0.143634 = 0.143634$$
 (5 marks)

• 
$$\hat{\beta}_0 = \overline{Y} - \hat{\beta}_1 \overline{X}$$
  
 $\overline{Y} = \frac{\sum_{i=1}^N Y_i}{N} = \frac{34379.16}{3240} = 10.61085$  and  $\overline{X} = \frac{\sum_{i=1}^N X_i}{N} = \frac{96143.00}{3240} = 29.67377$ 

Therefore

$$\hat{\beta}_0 = \overline{Y} - \hat{\beta}_1 \overline{X} = 10.610852 - (0.143634)(29.67377) = 10.610852 - 4.262162 = \underline{6.34869} = \underline{6.349}$$
(5 marks)

## (4 marks)

(b) Interpret the slope coefficient estimate you calculated in part (a) – i.e., explain in words what the numeric value you calculated for  $\hat{\beta}_1$  means.

<u>*Note:*</u>  $\hat{\beta}_1 = 0.1436$ . Y<sub>i</sub> is measured in <u>*dollars per hour*</u>, and X<sub>i</sub> is measured in <u>*years*</u>.

The estimate **0.143634** of  $\beta_1$  means that a **1-year** *increase* (decrease) in age  $X_i$  is associated on average with an *increase* (decrease) in hourly wage rate equal to 0.1436 *dollars per hour*, or 14.36 *cents per hour*.

## (4 marks)

(c) Calculate an estimate of  $\sigma^2$ , the error variance.

RSS = 
$$\sum_{i=1}^{N} \hat{u}_{i}^{2}$$
 = 77908.35; N-2 = 3240 - 2 = 3238  
 $\hat{\sigma}^{2} = \frac{RSS}{N-2} = \frac{\sum_{i=1}^{N} \hat{u}_{i}^{2}}{N-2} = \frac{77908.35}{3240-2} = \frac{77908.35}{3238} = \underline{24.060639} = \underline{24.0606}$  (4 marks)

## (6 marks)

(d) Compute the value of  $R^2$ , the coefficient of determination for the estimated OLS sample regression equation. Briefly explain what the value you have calculated for  $R^2$  means.

(4 marks)

$$R^{2} = \frac{ESS}{TSS} = \frac{\sum_{i=1}^{N} y_{i}^{2} - \sum_{i=1}^{N} \hat{u}_{i}^{2}}{\sum_{i=1}^{N} y_{i}^{2}} = \frac{78434.97 - 77908.35}{78434.97} = \frac{526.62}{78434.97} = 0.006714$$

OR

$$R^{2} = 1 - \frac{RSS}{TSS} = 1 - \frac{\sum_{i=1}^{N} \hat{u}_{i}^{2}}{\sum_{i=1}^{N} y_{i}^{2}} = 1 - \frac{77908.35}{78434.97} = 1 - 0.993286 = 0.006714$$

(2 marks)

**Interpretation of R<sup>2</sup> = 0.006714:** The value of 0.006714 indicates that **0.67 percent of** the total sample (or observed) variation in  $Y_i$  (hourly wage rates) is *attributable to*, or *explained by*, the regressor  $X_i$  (age) or the sample regression function.

## (8 marks)

- (e) Calculate the sample value of the t-statistic for testing the null hypothesis  $H_0$ :  $\beta_1 = 0$  against the alternative hypothesis  $H_1$ :  $\beta_1 \neq 0$ . (Note: You are not required to obtain or state the inference of this test.)
- t-statistic for  $\hat{\beta}_1$  is  $t(\hat{\beta}_1) = \frac{(\hat{\beta}_1 \beta_1)}{\hat{se}(\hat{\beta}_1)}$  (1) (2 marks)
- From part (a),  $\hat{\beta}_1 = 0.143634$ ;  $\hat{Var}(\hat{\beta}_1) = \frac{\hat{\sigma}^2}{\sum_{i=1}^N x_i^2} = \frac{24.060639}{25526.17} = 0.000942587.$
- The estimated standard error of  $\hat{\beta}_1$  is:

$$\hat{se}(\hat{\beta}_1) = \sqrt{\hat{Var}(\hat{\beta}_1)} = \sqrt{0.000942587} = 0.03070158 = 0.0307016$$

• Calculate the *sample value* of the t-statistic (1) under H<sub>0</sub>: set  $\beta_1 = 0$ ,  $\hat{\beta}_1 = 0.143634$ and  $\hat{se}(\hat{\beta}_1) = 0.0307016$  in (1) above.

$$t_0(\hat{\beta}_1) = \frac{(\hat{\beta}_1 - \beta_1)}{\hat{se}(\hat{\beta}_1)} = \frac{0.143634 - 0}{0.0307016} = \frac{0.143634}{0.0307016} = \frac{4.678388}{4.678388} = \frac{4.68}{4.68}$$
(6 marks)

## (38 marks)

4. You have been commissioned to investigate the relationship between the birth weights of newborn females and the number of prenatal visits to a physician or midwife that their mothers made during pregnancy. The dependent variable is *bwght<sub>i</sub>*, the birth weight of the i-th newborn female, measured in *grams*. The explanatory variable is *pnvisits<sub>i</sub>*, the number of prenatal visits of the i-th newborn's mother during pregnancy, measured in *number of visits*. The model you propose to estimate is given by the population regression equation

 $bwght_i = \beta_0 + \beta_1 pnvisits_i + u_i$ .

Your research assistant has used 857 sample observations on  $bwght_i$  and  $pnvisits_i$  to estimate the following OLS sample regression equation, where the figures in parentheses below the coefficient estimates are the *estimated standard errors* of the coefficient estimates:

bwght<sub>i</sub> = 3199.02 + 14.1219 pnvisits<sub>i</sub> +  $\hat{u}_i$  (i = 1, ..., N) N = 857 (3) (65.6909) (5.36347)  $\leftarrow$  (standard errors)

## (8 marks)

(a) Perform a test of the null hypothesis H<sub>0</sub>:  $\beta_1 = 0$  against the alternative hypothesis H<sub>1</sub>:  $\beta_1 \neq 0$  at the 1% significance level (i.e., for significance level  $\alpha = 0.01$ ). Show how you calculated the test statistic. State the decision rule you use, and the inference you would draw from the test. What would you conclude from the results of the test?

 $\begin{array}{l} H_0: \ \beta_1 \ = \ 0 \\ H_1: \ \beta_1 \ \neq \ 0 \\ \end{array} \ a \ \textit{two-sided} \ \textit{alternative hypothesis} \ \Rightarrow \ a \ \textit{two-tailed test} \end{array}$ 

• Test statistic is either 
$$t(\hat{\beta}_1) = \frac{\hat{\beta}_1 - \beta_1}{\hat{se}(\hat{\beta}_1)} \sim t[N-2] \text{ or } F(\hat{\beta}_1) = \frac{(\hat{\beta}_1 - \beta_1)^2}{V\hat{a}r(\hat{\beta}_1)} \sim F[1, N-2].$$

- $\hat{\beta}_1 = 14.1219;$   $\hat{se}(\hat{\beta}_1) = 5.36347;$   $\hat{Var}(\hat{\beta}_1) = (\hat{se}(\hat{\beta}_1))^2 = 28.76681$
- Calculate the *sample value* of *either* the t-statistic *or* the F-statistic under H<sub>0</sub>: set  $\beta_1 = 0$ ,  $\hat{\beta}_1 = 14.1219$ ,  $\hat{se}(\hat{\beta}_1) = 5.36347$ , and  $\hat{Var}(\hat{\beta}_1) = 28.76681$ .

$$t_{0}(\hat{\beta}_{1}) = \frac{\hat{\beta}_{1} - \beta_{1}}{\hat{se}(\hat{\beta}_{1})} = \frac{14.1219 - 0.0}{5.36347} = \frac{14.1219}{5.36347} = 2.63298 = 2.63$$
  
or (3 marks)

## **ANSWER to Question 4(a) -- continued:**

$$F_0(\hat{\beta}_1) = \frac{(\hat{\beta}_1 - \beta_1)^2}{V\hat{a}r(\hat{\beta}_1)} = \frac{(14.1219 - 0.0)^2}{28.76681} = \frac{199.42806}{28.76681} = 6.9326 = 6.9326$$

- Null distribution of  $t_0(\hat{\beta}_1)$  is t[N-2] = t[857-2] = t[855]
- Null distribution of  $F_0(\hat{\beta}_1)$  is F[1, N-2] = F[1, 857 2] = F[1, 855]

**Decision Rule:** At significance level  $\alpha$ ,

- reject  $\mathbf{H}_0$  if  $F_0(\hat{\beta}_1) > F_{\alpha}[1,855]$  or  $|\mathbf{t}_0(\hat{\beta}_1)| > \mathbf{t}_{\alpha/2}[855]$ , i.e., if either (1)  $\mathbf{t}_0(\hat{\beta}_1) > \mathbf{t}_{\alpha/2}[855]$  or (2)  $\mathbf{t}_0(\hat{\beta}_1) < -\mathbf{t}_{\alpha/2}[855]$ ;
- retain  $\mathbf{H}_{0}$  if  $F_{0}(\hat{\beta}_{1}) \leq F_{\alpha}[1,855]$  or  $|t_{0}(\hat{\beta}_{1})| \leq t_{\alpha/2}[855]$ , i.e., if  $-t_{\alpha/2}[855] \leq t_{0}(\hat{\beta}_{1}) \leq t_{\alpha/2}[855]$ .

**Critical values of t[855]-distribution or F[1, 855]-distribution:** in t-table, use  $df = \infty$ .

• *two-tailed* <u>1 percent</u> critical value =  $t_{\alpha/2}[855] = t_{0.005}[855] = 2.5816 = 2.58$  (1 mark) =  $t_{0.005}[\infty] = 2.576 = 2.58$ 

Critical values of F[1, 855]-distribution: in F-table, use denominator  $df = \infty$ .

• <u>**1** percent</u> critical value =  $F_{\alpha}[1, 855] = F_{0.01}[1, 855] = 6.6646 = <u>6.66</u>$  $= <math>F_{0.01}[1, \infty] = \underline{6.63}$ 

## Inference:

• At **1 percent significance level**, i.e., for  $\alpha = 0.01$ ,

 $|t_0(\hat{\beta}_1)| = 2.63 > 2.58 = t_{0.005}[855] \implies reject H_0 \text{ vs. } H_1 \text{ at 1 percent level.}$  $F_0(\hat{\beta}_1) = 6.93 > 6.63 = F_{0.01}[1, 855] \implies reject H_0 \text{ vs. } H_1 \text{ at 1 percent level.}$ 

## (2 marks)

(1 mark)

#### **ANSWER to Question 4(a) -- continued:**

• **Inference:** At the 1% significance level, the null hypothesis  $H_0$ :  $\beta_1 = 0$  is *rejected* in favour of the alternative hypothesis  $H_1$ :  $\beta_1 \neq 0$ .

#### **Conclusion implied by test outcome:**

#### (1 mark)

Rejection of the null hypothesis  $\beta_1 = 0$  against the alternative hypothesis  $\beta_1 \neq 0$  means that **the sample evidence favours the existence of a relationship between the** *number of prenatal visits* of mothers and the *birth weights* of their female newborns.

#### Question 4(a) - Alternative Answer -- uses confidence interval approach

• The two-sided  $(1 - \alpha)$ -level, or  $100(1 - \alpha)$  percent, confidence interval for  $\beta_1$  is:

$$\hat{\beta}_1 - t_{\alpha/2}[N-2]\hat{se}(\hat{\beta}_1) \le \beta_1 \le \hat{\beta}_1 + t_{\alpha/2}[N-2]\hat{se}(\hat{\beta}_1)$$
$$\hat{\beta}_{1L} \le \beta_1 \le \hat{\beta}_{1U}$$

• Required results and intermediate calculations:

$$N - K = 857 - 2 = 855; \qquad \hat{\beta}_1 = 14.1219; \qquad s\hat{e}(\hat{\beta}_1) = 5.36347$$

$$1 - \alpha = 0.99 \implies \alpha = 0.01 \implies \alpha/2 = 0.005; \ t_{\alpha/2}[N - 2] = t_{0.005}[855] = 2.576 \qquad (1 \text{ mark})$$

$$t_{\alpha/2}[N - 2]s\hat{e}(\hat{\beta}_1) = t_{0.005}[855]s\hat{e}(\hat{\beta}_1) = 2.576(5.36347) = 13.816299$$

• Lower 99% confidence limit for  $\beta_1$  is:

#### (2 marks)

(2 marks)

$$\hat{\beta}_{1L} = \hat{\beta}_1 - t_{\alpha/2} [N-2] \hat{se}(\hat{\beta}_1) = \hat{\beta}_1 - t_{0.005} [855] \hat{se}(\hat{\beta}_1)$$

$$= 14.1219 - 2.576(5.36347) = 14.1219 - 13.816299 = 0.305601 = \underline{0.3056}$$

$$= 14.1219 - 2.617(5.36347) = 14.1219 - 14.036201 = 0.0696086 = \underline{0.08570}$$

$$= 14.1219 - 2.58(5.36347) = 14.1219 - 13.837753 = 0.284147 = \underline{0.2841}$$

• Upper 99% confidence limit for  $\beta_1$  is:

$$\begin{aligned} \hat{\beta}_{1U} &= \hat{\beta}_1 + t_{\alpha/2} [N-2] \hat{se}(\hat{\beta}_1) = \hat{\beta}_1 + t_{0.005} [855] \hat{se}(\hat{\beta}_1) \\ &= 14.1219 + \textbf{2.576} (5.36347) = 14.1219 + 13.816299 = 27.938199 = \textbf{27.9382} \\ &= 14.1219 + \textbf{2.617} (5.36347) = 14.1219 + 14.036201 = 28.158101 = \textbf{28.1581} \\ &= 14.1219 + \textbf{2.58} (5.36347) = 14.1219 + 13.837753 = 27.959653 = \textbf{27.9597} \end{aligned}$$

## **Question 4(a) – Alternative Answer (continued)**

• **Two-sided 99% confidence interval for**  $\beta_1$  is therefore:

 $0.3056 \le \beta_1 \le 27.9382$  or  $0.08570 \le \beta_1 \le 28.1581$  or  $0.2841 \le \beta_1 \le 27.9597$ 

- **Decision Rule:** At significance level  $\alpha$ ,
- *reject* H<sub>0</sub> if the *hypothesized* value b<sub>1</sub> of β<sub>1</sub> specified by H<sub>0</sub> lies *outside* the two-sided (1-α)-level confidence interval for β<sub>1</sub>, i.e., if either
   (1) b<sub>1</sub> < β<sub>1</sub> t<sub>α/2</sub>[855]sê(β<sub>1</sub>) or (2) b<sub>1</sub> > β<sub>1</sub> + t<sub>α/2</sub>[855]sê(β<sub>1</sub>).
- retain H<sub>0</sub> if the hypothesized value b<sub>1</sub> of β<sub>1</sub> specified by H<sub>0</sub> lies inside the two-sided (1-α)-level confidence interval for β<sub>1</sub>, i.e., if β<sub>1</sub> t<sub>α/2</sub>[855]sê(β<sub>1</sub>) ≤ b<sub>1</sub> ≤ β<sub>1</sub> + t<sub>α/2</sub>[855]sê(β<sub>1</sub>).

## **Inference**:

• At **1 percent significance level**, i.e., for  $\alpha = 0.01$ ,

$$b_{1} = 0 < 0.3056 = \hat{\beta}_{1L} = \hat{\beta}_{1} - t_{0.005}[855]\hat{se}(\hat{\beta}_{1})$$
  

$$\Rightarrow reject H_{0} \text{ vs. } H_{1} \text{ at 1 percent level.}$$

• **Inference:** At the 1% significance level, the null hypothesis  $H_0$ :  $\beta_1 = 0$  is *rejected* in favour of the alternative hypothesis  $H_1$ :  $\beta_1 \neq 0$ .

## Conclusion implied by test outcome:

Rejection of the null hypothesis  $\beta_1 = 0$  against the alternative hypothesis  $\beta_1 \neq 0$  means that **the sample evidence favours the existence of a relationship between the** *number of prenatal visits* of mothers and the *birth weights* of their female newborns.

#### (1 mark)

(1 mark)

(1 mark)

## (8 marks)

(b) Perform a test of the proposition that newborn females' birth weights are positively related to their mother's number of prenatal visits during pregnancy, i.e., that an increase in the number of prenatal visits *increases* the birth weights of newborn females. Use the 1 percent significance level (i.e.,  $\alpha = 0.01$ ). State the null hypothesis H<sub>0</sub> and the alternative hypothesis H<sub>1</sub>. Show how you calculated the test statistic. State the decision rule you use, and the inference you would draw from the test.

## ANSWER to Question 4(b):

## Null and Alternative Hypotheses:

- $\begin{array}{ll} H_0: \ \beta_1 \ = \ 0 & (1 \ \text{mark}) \\ H_1: \ \beta_1 \ > \ 0 & \Rightarrow \ a \ \textit{right-tailed t-test} & (1 \ \text{mark}) \end{array}$
- Test statistic is  $t(\hat{\beta}_1) = \frac{\hat{\beta}_1 \beta_1}{\hat{se}(\hat{\beta}_1)} \sim t[N-2]; \hat{\beta}_1 = 14.1219; \hat{se}(\hat{\beta}_1) = 5.36347.$
- Calculate the *sample value* of the t-statistic under H<sub>0</sub>: set  $\beta_1 = 0$ ,  $\hat{\beta}_1 = 14.1219$ , and  $\hat{se}(\hat{\beta}_1) = 5.36347$ .

$$t_0(\hat{\beta}_1) = \frac{\hat{\beta}_1 - \beta_1}{\hat{se}(\hat{\beta}_1)} = \frac{14.1219 - 0.0}{5.36347} = \frac{14.1219}{5.36347} = 2.63298$$

• Null distribution of  $t_0(\hat{\beta}_1)$  is t[N-2] = t[857-2] = t[855]

**Decision Rule:** At significance level α,

- *reject*  $\mathbf{H}_0$  if  $t_0(\hat{\beta}_1) > t_{\alpha}[855]$ ,
- retain  $\mathbf{H}_{\mathbf{0}}$  if  $t_0(\hat{\beta}_1) \leq t_{\alpha}[855]$ .

**Critical value of t[855]-distribution:** from t-table, use df = 120 or  $df = \infty$  or any number in between.

• right-tail <u>1 percent</u> critical value =  $t_{0.01}[855] = \underline{2.331} = \underline{2.33}$  (1 mark) =  $t_{0.01}[855] = \underline{2.358} = \underline{2.36}$ =  $t_{0.01}[\infty] = \underline{2.326} = \underline{2.33}$ 

(2 marks)

## **ANSWER to Question 4(b) -- continued:**

## Inference:

## (1 mark)

• At **1 percent significance level**, i.e., for  $\alpha = 0.01$ ,

 $t_0(\hat{\beta}_1) = 2.63 > 2.33 = t_{0.01}[855] \implies reject H_0 \text{ vs. } H_1 \text{ at 1 percent level.}$ 

• <u>Inference</u>: At the 1% significance level, the null hypothesis  $H_0$ :  $\beta_1 = 0$  is *rejected* in favour of the alternative hypothesis  $H_1$ :  $\beta_1 > 0$ .

#### (14 marks)

(c) Compute the two-sided 95% confidence interval for the intercept coefficient  $\beta_0$ . Use this two-sided 95% confidence interval for  $\beta_0$  to test the hypothesis that the mean birth weight of newborn females whose mothers made no prenatal visits to a physician or midwife equals 3,000 grams. State the null hypothesis H<sub>0</sub> and the alternative hypothesis H<sub>1</sub>. State the decision rule you use, and the inference you would draw from the test.

## **ANSWER to Question 4(c):**

• The two-sided  $(1 - \alpha)$ -level, or  $100(1 - \alpha)$  percent, confidence interval for  $\beta_1$  is computed as

$$\hat{\beta}_0 - t_{\alpha/2}[N-2]\hat{se}(\hat{\beta}_0) \le \beta_0 \le \hat{\beta}_0 + t_{\alpha/2}[N-2]\hat{se}(\hat{\beta}_0)$$
(2 marks)

where

- $\hat{\beta}_{0L} = \hat{\beta}_0 t_{\alpha/2}[N-2]\hat{se}(\hat{\beta}_0) = \text{the lower } 100(1-\alpha)\%$  confidence limit for  $\beta_0$
- $\hat{\beta}_{0U} = \hat{\beta}_0 + t_{\alpha/2}[N-2]\hat{se}(\hat{\beta}_0) = \text{the upper } 100(1-\alpha)\%$  confidence limit for  $\beta_0$
- $t_{\alpha/2}[N-2] =$  the  $\alpha/2$  critical value of the t-distribution with N–2 degrees of freedom.
- Required results and intermediate calculations:
  - $N K = 857 2 = 855; \qquad \hat{\beta}_0 = 3199.02; \qquad s\hat{e}(\hat{\beta}_0) = 65.6909$   $1 \alpha = 0.95 \implies \alpha = 0.05 \implies \alpha/2 = 0.025; \qquad t_{\alpha/2}[N 2] = t_{0.025}[855] = 1.963$   $= t_{0.025}[855] = 1.980$   $= t_{0.025}[\infty] = 1.960$

 $t_{\alpha/2}[N-2]\hat{se}(\hat{\beta}_0) = t_{0.025}[855]\hat{se}(\hat{\beta}_0) = 1.963(65.6909) = 128.95124$ 

• Lower 95% confidence limit for  $\beta_0$  is:

#### (3 marks)

$$\hat{\beta}_{0L} = \hat{\beta}_0 - t_{\alpha/2}[N-2]\hat{s}\hat{e}(\hat{\beta}_0) = \hat{\beta}_0 - t_{0.025}[855]\hat{s}\hat{e}(\hat{\beta}_0)$$

$$= 3199.02 - \mathbf{1.963}(65.6909) = 3199.02 - 128.95124 = 3070.069 = \mathbf{3070.07}$$

$$= 3199.02 - \mathbf{1.980}(65.6909) = 3199.02 - 130.06798 = 3068.952 = \mathbf{3068.955}$$

$$= 3199.02 - \mathbf{1.960}(65.6909) = 3199.02 - 128.75416 = 3070.266 = \mathbf{3070.27}$$

#### **ANSWER to Question 4(c) -- continued:**

- Upper 95% confidence limit for  $\beta_0$  is:  $\hat{\beta}_{0U} = \hat{\beta}_0 + t_{\alpha/2} [N-2] \hat{se}(\hat{\beta}_0) = \hat{\beta}_0 + t_{0.025} [855] \hat{se}(\hat{\beta}_0)$  = 3199.02 + 1.963(65.6909) = 3199.02 + 128.95124 = 3327.971 = 3327.97 = 3199.02 + 1.980(65.6909) = 3199.02 + 130.06798 = 3329.088 = 3329.09= 3199.02 + 1.960(65.6909) = 3199.02 + 128.75416 = 3327.774 = 3327.77
- <u>Result</u>: The two-sided 95% confidence interval for  $\beta_0$  is: [3070.07, 3327.97]

 $3070.07 \le \beta_0 \le 3327.97$  or  $3068.95 \le \beta_0 \le 3329.09$  or  $3070.27 \le \beta_0 \le 3327.77$ 

- Null and Alternative Hypotheses:
  - $H_0: \beta_0 = 3000$  $H_1: \beta_0 \neq 3000$ a two-sided alternative hypothesis  $\Rightarrow$  a two-tailed test
- **Decision Rule:** At significance level  $\alpha$ ,
- *reject* H<sub>0</sub> if the *hypothesized* value b<sub>0</sub> of β<sub>0</sub> specified by H<sub>0</sub> lies *outside* the two-sided (1-α)-level confidence interval for β<sub>0</sub>, i.e., if either
   (1) b<sub>0</sub> < β̂<sub>0</sub> t<sub>α/2</sub>[855]sê(β̂<sub>0</sub>) or (2) b<sub>0</sub> > β̂<sub>0</sub> + t<sub>α/2</sub>[855]sê(β̂<sub>0</sub>).
- retain H<sub>0</sub> if the hypothesized value b<sub>0</sub> of β<sub>0</sub> specified by H<sub>0</sub> lies inside the two-sided (1-α)-level confidence interval for β<sub>0</sub>, i.e., if
   β<sub>0</sub> t<sub>α/2</sub>[855]sê(β̂<sub>0</sub>) ≤ b<sub>0</sub> ≤ β̂<sub>0</sub> + t<sub>α/2</sub>[855]sê(β̂<sub>0</sub>).

#### **Inference**:

• At **5 percent significance level**, i.e., for  $\alpha = 0.05$ ,

 $b_0 = 3000 < 3070.07 = \hat{\beta}_{0L} = \hat{\beta}_0 - t_{0.025}[855]\hat{s}\hat{e}(\hat{\beta}_0)$  $\Rightarrow reject H_0 vs. H_1 at 5 percent level.$ 

• <u>Inference</u>: At the 5% significance level, the null hypothesis  $H_0$ :  $\beta_0 = 3000$  is *rejected* in favour of the alternative hypothesis  $H_1$ :  $\beta_0 \neq 3000$ .

## (2 marks)

(2 marks)

(2 marks)

## (8 marks)

(d) Perform a test of the proposition that each additional prenatal visit made by the mother is associated on average with an increase in their newborn females' birth weight of less than 25 grams. Use the 5 percent significance level (i.e.,  $\alpha = 0.05$ ). State the null hypothesis H<sub>0</sub> and the alternative hypothesis H<sub>1</sub>. Show how you calculated the test statistic. State the decision rule you use, and the inference you would draw from the test.

## ANSWER to Question 4(d):

## Null and Alternative Hypotheses:

- Test statistic is  $t(\hat{\beta}_1) = \frac{\hat{\beta}_1 \beta_1}{\hat{se}(\hat{\beta}_1)} \sim t[N-2]; \ \hat{\beta}_1 = 14.1219; \ \hat{se}(\hat{\beta}_1) = 5.36347.$
- Calculate the *sample value* of the t-statistic under H<sub>0</sub>: set  $\beta_1 = 25$ ,  $\hat{\beta}_1 = 14.1219$ , and  $\hat{se}(\hat{\beta}_1) = 5.36347$ .

$$t_0(\hat{\beta}_1) = \frac{\hat{\beta}_1 - \beta_1}{\hat{se}(\hat{\beta}_1)} = \frac{14.1219 - 25.0}{5.36347} = \frac{-10.8781}{5.36347} = -2.02818$$

• Null distribution of  $t_0(\hat{\beta}_1)$  is t[N-2] = t[857-2] = t[855]

**Decision Rule:** At significance level  $\alpha$ ,

- reject  $\mathbf{H}_0$  if  $t_0(\hat{\beta}_1) < -t_{\alpha}[855]$ ,
- retain  $\mathbf{H}_{\mathbf{0}}$  if  $t_0(\hat{\beta}_1) \geq -t_{\alpha}[855]$ .

Critical value of t[855]-distribution: from t-table, use df = 120 or  $df = \infty$  or any number in between.

• left-tail 1 percent critical value = 
$$-t_{0.05}[855] = -\underline{1.6466} = -\underline{1.65}$$
 (1 mark)  
=  $-t_{0.05}[855] = \underline{-1.658} = \underline{-1.66}$   
=  $-t_{0.05}[\infty] = \underline{-1.645} = \underline{-1.65}$ 

(2 marks)

## **ANSWER to Question 4(d) -- continued:**

## Inference:

## (1 mark)

• At **5 percent significance level**, i.e., for  $\alpha = 0.05$ ,

 $t_0(\hat{\beta}_1) = -2.03 < -1.65 = -t_{0.05}[855] \implies reject H_0 \text{ vs. } H_1 \text{ at 5 percent level.}$ 

• <u>Inference</u>: At the 5% significance level, the null hypothesis  $H_0$ :  $\beta_1 = 25$  is *rejected* in favour of the alternative hypothesis  $H_1$ :  $\beta_1 < 25$ .

## **Percentage Points of the t-Distribution**

TABLE D.2 Percentage points of the t distribution

Example	0.05
$\Pr(t > 2.086) = 0.025$	
Pr(t > 1.725) = 0.05 for df = 20	
$\Pr( t  > 1.725) = 0.10$	0 1.725

Pr df	0.25 0.50	0.10 0.20	0.05 0.10	0.025 0.05	0.01 0.02	0.005 0.010	0.001 0.002	
1	1.000	3.078	6.314	12.706 31.821 63.657		318.31		
2	0.816	1.886	2.920	4.303 6.965 9.925		22.327		
3	0.765	1.638	2.353	3.182 4.541 5.841		10.214		
4	0.741	1.533	2.132	2.776 3.747 4.604		7.173		
5	0.727	1.476	2.015	2.571	3.365	4.032	5.893	
6	0.718	1.440	1.943	2.447	3.143	3.707	5.208	
7	0.711	1.415	1.895	2.365	2.998	3.499	4.785	
8	0.706	1.397	1.860	2.306	2.896	3.355	4.501	
9	0.703	1.383	1.833	2.262	2.821	3.250	4.297	
10	0.700	1.372	1.812	2.228	2.764	3.169	4.144	
11	0.697	1.363	1.796	2.201	2.718	3.106	4.025	
12	0.695	1.356	1.782	2.179	2.681	3.055	3.930	
13	0.694	1.350	1.771	2.160	2.650	3.012	3.852	
14	0.692	1.345	1.761	2.145	2.624	2.977	3.787	
15	0.691	1.341	1.753	2.131	2.602	2.947	3.733	
16	0.690	1.337	1.746	2.120	2.583	2.921	3.686	
17	0.689	1.333	1.740	2.110	2.567	2.898	3.646	
18	0.688	1.330	1.734	2.101	2.552	2.878	3.610	
19	0.688	1.328	1.729	2.093	2.539	2.861	3.579	
20	0.687	1.325	1.725	2.086	2.528	2.845	3.552	
21	0.686	1.323	1.721	2.080	2.518	2.831	3.527	
22	0.686	1.321	1.717	2.074	2.508	2.819	3.505	
23	0.685	1.319	1.714	2.069 2.500 2.807		2.807	3.485	
24	0.685	1.318	1.711	2.064	2.492	2.797	3.467	
25	0.684	1.316	1.708	2.060	2.485	2.787	3.450	
26	0.684	1.315	1.706	2.056	2.479	2.779	3.435	
27	0.684	1.314	1.703	2.052	2.473	2.771	3.421	
28	0.683	1.313	1.701	2.048 2.467 2.763		3.408		
29	0.683	1.311	1.699	2.045	2.045 2.462 2.756		3.396	
30	0.683	1.310	1.697	2.042	2.457	2.750	3.385	
40	0.681	1.303	1.684	2.021	2.423	2.704	3.307	
60	0.679	1.296	1.671	2.000	2.390	2.660	3.232	
120	0.677	1.289	1.658	1.980	2.358	2.617	3.160	
∞	0.674	1.282	1.645	1.960	2.326	2.576	3.090	

Note: The smaller probability shown at the head of each column is the area in one tail; the larger probability is the area in both tails.

Source: From E. S. Pearson and H. O. Hartley, eds., Biometrika Tables for Statisticians, vol. 1, 3d ed., table 12, Cambridge University Press, New York, 1966. Reproduced by permission of the editors and trustees of Biometrika.

Source: Damodar N. Gujarati, Basic Econometrics, Third Edition. New York: McGraw-Hill, 1995, p. 809.

## **Selected Upper Percentage Points of the F-Distribution**

TABLE D.3			
Upper percentage	points of the	F distribution	(continued)

df for denom-		df for numerator Ni											
inator N <sub>2</sub>	Pr	t	2	3	4	5	6	7	8	9	10	11	12
	.25	1.40	1.48	1.47	1.45	1.44	1.42	1.41	1.40	1.39	1.39	1.38	1.37
22	.10	2.95	2.56	2.35	2.22	2.13	2.06	2.01	1.97	1.93	1.90	1.88	1.86
	.05	4.30	3.44	3.05	2.82	2.66	2.55	2.46	2.40	2.34	2.30	2.26	2.23
	.01	7.95	5.72	4.82	4.31	3.99	3.76	3.59	3.45	3.35	3.26	3.18	3.12
	.25	1.39	1.47	1.46	1.44	1.43	1.41	1.40	1.39	1.38	1.38	1.37	1.36
24	.10	2.93	2.54	2.33	2.19	2.10	2.04	1.98	1.94	1.91	1.88	1.85	1.83
	.05	4.26	3.40	3.01	2.78	2.62	2.51	2.42	2.36	2.30	2.25	2.21	2.18
	.01	, 7.82	5.61	4.72	4.22	3.90	3.67	3.50	3.30	3.26	3.17	3.09	3.03
	.25	1.38	1.46	1.45	1.44	1.42	1.41	1.39	1.38	1.37	1.37	1.36	1.35
26	.10	2.91	2.52	2.31	2.17	2.08	2.01	1.96	1.92	1.88	1.86	1.84	1.81
	.05	4.23	3.37	2.98	2.74	2.59	2.47	2.39	2.32	2.27	2.22	2.18	2.15
	.01	7.72	5.53	4.64	4.14	3.82	3.59	3.42	3.29	3.18	3.09	3.02	2.96
	.25	1.38	1.46	1.45	1.43	1.41	1.40	1.39	1.38	1.37	1.36	1.35	1.34
28	.10	2.89	2.50	2.29	2.16	2.06	2.00	1.94	1.90	1.87	1.84	1.81	1.79
	.05	4.20	3.34	2.95	2.71	2.56	2.45	2.36	2.29	2.24	2.19	2.15	2.12
	.01	7.64	5.45	4.57	4.07	3.75	3.53	3.36	3.23	3.12	3.03	<b>2.9</b> 6	2.90
	.25	1.38	1.45	1.44	1.42	1.41	1.39	1.38	1.37	1.36	1.35	1.35	1.34
30	.10	2.88	2.49	<b>2.2<u></u>8</b>	2.14	2.05	1.98	1.93	1.88	1.85	1.82	1.79	1.77
	.05	4.17	3.32	2.92	2.69	2.53	2.42	2.33	2.27	2.21	2.16	2.13	2.09
	.01	7.56	5.39	4.51	4.02	3.70	3.47	3.30	3.17	3.07	2.98	2.91	2.84
	.25	1.36	1.44	1.42	1.40	1.39	1.37	1.36	1.35	1.34	1.33	1.32	1.31
40	.10	2.84	2.44	2.23	2.09	2.00	1.93	1.87	1.83	1.79	1.76	1.73	1.71
	.05	4.08	3.23	2.84	2.61	Z.45	2.34	2.25	2.18	2.12	2.08	Z.04	2.00
	.01	7.31	5.18	4.31	3.83	3.51	3.29	3.12	2.99	2.89	2.80	2.15	(2.00
	.25	1.35	1.42	1.41	1.38	1.37	1.35	1.33	1.32	1.31	1.30	1.29	1.29
60	.10	2.79	2.39	2.18	2.04	1.95	1.87	1.82	1.77	1.74	1.71	1.68	1.66
	.05	4.00	3.15	2.76	2.53	2.37	2.25	2.17	2.10	2.04	1.99	1.95	1.92
	10.	7.08	4.98	4.13	3.65	3.34	3.12	2.95	<b>Z.8</b> 2	2.72	2.63	2.56	2.50
	.25	1.34	1.40	1.39	1.37	1.35	1.33	1.31	1.30	1.29	1.28	1.27	1.26
120	.10	2.75	2.35	2.13	1.99	1.90	1.82	1.77	1.72	1.68	1.65	1.62	1.60
	.05	3.92	3.07	2.68	2.45	2.29	2.17	2.09	2.02	1.96	1.91	1.87	1.83
	.01	6.85	4.79	3.95	3.48	3.17	2.90	2.79	2.00	2.56	2.47	2.40	2.34
	.25	1.33	1.39	1.38	1.36	1.34	1.32	1.31	1.29	1.28	1.27	1.26	1.25
200	.10	2.73	2.33	2.11	1.97	1.88	1.80	1.75	1.70	1.66	1.63	1.60	1.57
	.05	3.89	3.04	2.65	2.42	2.20	2.14	2.00	1.98	1.93	1.88	1.84	1.80
	.01	6.76	4.71	3.88	3.41	3.11	2.89	2.73	2.60	2.50	2.41	2.34	2.27
	.25	1.32	1.39	1.37	1.35	1.33	1.31	1.29	1.28	1.27	1.25	1.24	1.24
<b>e</b> 0	.10	2.71	2.30	2.08	1.94	1.85	1.77	1.72	1.67	1.63	1.60	1.57	1.55
	.05	3.84	3.00	2.60	2.37	2.21	2.10	2.01	1.94	1.88	1.83	1.79	1.75
	.01	6.63	4.61	3.78	3.32	3.02	2.80	2.64	2.51	2.41	2.32	2.25	2.18

Source: Damodar N. Gujarati, Basic Econometrics, Third Edition. New York: McGraw-Hill, 1995, p. 814.