

Part I. Multiple Choice Questions

1. The normality assumption implies that:

- d. ☒ a. the population error u is dependent on the explanatory variables and is normally distributed with mean equal to one and variance σ^2 .
☐ b. the population error u is independent of the explanatory variables and is normally distributed with mean equal to one and variance σ .
☐ c. the population error u is dependent on the explanatory variables and is normally distributed with mean zero and variance σ .
☒ d. the population error u is independent of the explanatory variables and is normally distributed with mean zero and variance σ^2 .

2. Consider the equation, $Y = \beta_1 + \beta_2 X_2 + u$. A null hypothesis, $H_0: \beta_2 = 0$ states that:

- b. ☐ a. X_2 has no effect on the expected value of β_2 .
☒ b. X_2 has no effect on the expected value of Y .
☐ c. β_2 has no effect on the expected value of Y .
☐ d. Y has no effect on the expected value of X_2 .

3. Which of the following statements is true?

- b. ☒ a. If the calculated value of F statistic is higher than the critical value, we reject the alternative hypothesis in favor of the null hypothesis.
☒ b. The F statistic is always nonnegative as SSR_r is never smaller than SSR_{ur} .
☒ c. Degrees of freedom of a restricted model is always less than the degrees of freedom of an unrestricted model.
☒ d. The F statistic is more flexible than the t statistic to test a hypothesis with a single restriction.

a. ☒ 4. Which of the following statements is true?

- ☒ a. The standard error of a regression, $\hat{\sigma}$, is ~~not~~ an unbiased estimator for σ , the standard deviation of the error, u , in a multiple regression model.
☒ b. All estimators that are unbiased are also consistent. *may not converge.*
☒ c. Almost all economists agree that unbiasedness is a minimal requirement for an estimator in regression analysis.
☒ d. All estimators in a regression model that are consistent are also unbiased.

- d. 5. In a multiple regression model, the OLS estimator is consistent if:
- ☒ a. there is no correlation between the dependent variables and the error term.
 - ☒ b. there is a perfect correlation between the dependent variables and the error term.
 - ☒ c. the sample size is less than the number of parameters in the model.
 - ☒ d. there is no correlation between the independent variables and the error term.

- b. 6. If $\delta_1 = \text{Cov}(x_1, x_2) / \text{Var}(x_1)$ where x_1 and x_2 are two independent variables in a regression equation, which of the following statements is true? $\beta_1 + \delta_1 \beta_2$
- ☒ a. If x_2 has a positive partial effect on the dependent variable, and $\delta_1 > 0$, then the inconsistency in the simple regression slope estimator associated with x_1 is negative.
 - ☒ b. If x_2 has a positive partial effect on the dependent variable, and $\delta_1 > 0$, then the inconsistency in the simple regression slope estimator associated with x_1 is positive.
 - ☒ c. If x_1 has a positive partial effect on the dependent variable, and $\delta_1 > 0$, then the inconsistency in the simple regression slope estimator associated with x_1 is negative.
 - ☒ d. If x_1 has a positive partial effect on the dependent variable, and $\delta_1 > 0$, then the inconsistency in the simple regression slope estimator associated with x_1 is positive.

- c. 7. Which of the following statements is true under the Gauss-Markov assumptions?
- a. Among a certain class of estimators, OLS estimators are best linear unbiased, but are asymptotically inefficient.
 - b. Among a certain class of estimators, OLS estimators are biased but asymptotically efficient.
 - ☒ c. Among a certain class of estimators, OLS estimators are best linear unbiased and asymptotically efficient.
 - d. The LM test is independent of the Gauss-Markov assumptions.

- a. 8. Changing the unit of measurement of any independent variable, where log of the independent variable appears in the regression:
- ☒ a. affects only the intercept coefficient.
 - b. affects only the slope coefficient.
 - c. affects both the slope and intercept coefficients.
 - d. affects neither the slope nor the intercept coefficient.

- b. 9. If a regression equation has only one explanatory variable, say x_1 , its standardized coefficient must lie in the range:
- a. -2 to 0.
 - b. -1 to 1.
 - c. 0 to 1.
 - d. 0 to 2.

C 10. Which of the following correctly represents the equation for adjusted R^2 ?

- a. $\bar{R}^2 = 1 - [SSR/(n-1)]/[SST/(n+1)]$
- b. $\bar{R}^2 = 1 - [SSR/(n-k-1)]/[SST/(n+1)]$
- c. $\bar{R}^2 = 1 - [SSR/(n-k-1)]/[SST/(n-1)]$
- d. $\bar{R}^2 = 1 - [SSR]/[SST/(n-1)]$

11. The following simple model is used to determine the annual savings of an individual on the basis of his annual income and education.

$$\text{Savings} = \beta_0 + \beta_1 \text{Edu} + \beta_2 \text{Inc} + u$$

The variable 'Edu' takes a value of 1 if the person is educated and the variable 'Inc' measures the income of the individual. The inclusion of another binary variable in this model that takes a value of 1 if a person is uneducated, will give rise to the problem of _____.

- a. omitted variable bias
- b. self-selection
- c. dummy variable trap
- d. heteroskedasticity

虚拟变量
异方差

Part 2. Quantitative Questions

1. The F -statistic for testing a set of linear hypotheses is given by the formula

$$F = \frac{(SSR_r - SSR_{ur})/q}{SSR_{ur}/(n - k - 1)}$$

where SSR_r is the sum of squared residuals from the restricted regression, SSR_{ur} is the sum of squared residuals from the unrestricted regression, q is the number of restrictions under the null hypothesis, and k is the number of regressors in the unrestricted regression. Prove that this formula is the same as the following formula based on the regression R^2 of the restricted and unrestricted regression:

$$R^2 = 1 - \frac{SSR}{SST}$$

$$F = \frac{(R_{ur}^2 - R_r^2)/q}{(1 - R_{ur}^2)/(n - k - 1)}$$

$$SSR = (1 - R^2)SST.$$

2. Consider the following multiple regression model

$$Y_i = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \beta_3 X_{3i} + u_i$$

You want to consider certain hypotheses involving more than one parameter. You decide to test the joint hypotheses using the F -statistics. For each of the cases below specify a restricted model and indicate how you would compute the F -statistic to test for the validity of the restrictions.

- (a) $\beta_1 = -\beta_2; \beta_3 = 0$ $Y_i = \beta_0 + X_{1i} + \beta_2(X_{2i} - X_{1i}) + u_i$
- (b) $\beta_1 + \beta_2 + \beta_3 = 1$ $F = \frac{(SSR_r - SSR_{ur}) / 2}{SSR_{ur} / (n - 3 - 1)}$
- (c) $\beta_1 = \beta_2 / \beta_3$

$$Y_i = \beta_0 + X_{1i} + \beta_2(X_{2i} - X_{1i}) + \beta_3(X_{3i} - X_{1i}) + u_i$$

$$F = \frac{\text{---} / 1}{\text{---} / (n - 3 - 1)}$$

(C) is not a linear restriction

can't use F -test.

3. In section 4.5 in the textbook, we used the following log-log model for testing the rationality of assessment of housing prices:

$$\log(\text{price}) = \beta_0 + \beta_1 \log(\text{assess}) + \beta_2 \log(\text{lotsize}) + \beta_3 \log(\text{sqrft}) + \beta_4 \text{bdrms} + u,$$

where

price = house price.

assess = the assessed housing value (before the house was sold).

lotsize = size of the lot, in square feet.

sqrft = square footage.

bdrms = number of bedrooms.

If the assessed housing price is a rational valuation, then a 1% change in *assess* should be associated with a 1% change in *price*.

We now use a level-level formulation to test for the rationality of the assessed valuation.

- (i) In the simple regression model

$$\text{price} = \beta_0 + \beta_1 \text{assess} + u,$$

the assessment is rational if $\beta_1 = 1$ and $\beta_0 = 0$. The estimated equation is

$$\widehat{\text{price}} = -14.47 + 0.976 \text{ assess}$$

$$(16.27) (0.049)$$

$$n = 88, SSR = 165,644.51, R^2 = 0.820.$$

First, test the hypothesis that $H_0: \beta_0 = 0$ against the two-sided alternative. Then, test $H_0: \beta_1 = 1$ against the two-sided alternative. What do you conclude?

$$H_0: \beta_0 = 0 : t = \frac{-14.47}{16.27} = -0.889367 \Rightarrow \text{accept.}$$

df=86. |t| < 1.987

$$H_0: \beta_1 = 1 : t = \frac{0.976 - 1}{0.049} = -0.489796 \Rightarrow \text{accept.}$$

|t| < 1.987

- (ii) To test the joint hypothesis that $\beta_0 = 0$ and $\beta_1 = 1$, we need the SSR in the restricted model. This amounts to computing $\sum_{i=1}^n (\text{price}_i - \text{assess}_i)^2$, since the residuals in the restricted model are just $\text{price}_i - \text{assess}_i$. (No estimation is needed for the restricted model because both parameters are specified under H_0 .) This turns out to yield SSR = 209,448.99. Carry out the F test for the joint hypothesis.

$$F = \frac{(209448.99 - 165644.51) / 2}{165644.51 / 86} \approx 11.37$$

\Rightarrow reject

- (iii) Now, test $H_0: \beta_2 = 0, \beta_3 = 0, \beta_4 = 0$ in the model

$$\text{price} = \beta_0 + \beta_1 \text{assess} + \beta_2 \text{lotsize} + \beta_3 \text{sqrft} + \beta_4 \text{bdrms} + u.$$

The R-squared from estimating this model using the same 88 houses is 0.829.

$$F = \frac{(0.829 - 0.820) / 3}{(1 - 0.829) / 83} = 1.46$$

\Rightarrow accept

- (iv) If the variance of *price* changes with *assess*, *lotsize*, *sqft*, or *bdrms*, what can you say about the *F* test from part (iii)?

heteroskedasticity

not have \bar{F} distribution under null hypothesis.

Can't use \bar{F} test.

4. In the simple regression model $y = \beta_0 + \beta_1 x + u$ it was shown in class that under MLR.1 through MLR.4 assumptions the slope coefficient $\hat{\beta}_1$ is consistent for β_1 . Using $\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$, show that $\text{plim} \hat{\beta}_0 = \beta_0$. [You need to use the consistency of $\hat{\beta}_1$ and the law of large numbers, along with the fact that $\beta_0 = E(y) - \beta_1 E(x)$].

$$\begin{aligned}\text{plim } \hat{\beta}_0 &= \text{plim } (\bar{y} - \hat{\beta}_1 \bar{x}) \\ &= E(y) - E(x) \cdot \text{plim } \hat{\beta}_1 \\ &= E(y) - \beta_1 E(x). \\ &= \beta_0\end{aligned}$$

5. The following equations were estimated using the data in BWGHT.TXT:

$$\log(\widehat{bwght}) = 4.66 - .0044cigs + .0093 \log(faminc) + .016parity$$

$b\% = 100 \times 0.0044$ (22) (0.0009) (0.0059) (0.006)
 $cigs.$
 $+ .027male + .055white$
 (.010) (.013)

$n = 1,388, R^2 = 0.0472$

and

$$\log(\widehat{bwght}) = 4.65 - .0052cigs + .0110 \log(faminc) + .017parity$$

(.38) (0.0010) (0.0085) (0.006)

$$+ .034male + .045white - .0030motheduc + .0032fatheduc$$

(.011) (.015) (.0030) (.0026)

$n = 1,190, R^2 = 0.0493,$

where

bwght = birth weight, in pounds.

cigs = average number of cigarettes the mother smoked per day during pregnancy.

parity = the birth order of this child.

faminc = annual family income.

motheduc = years of schooling for the mother.

fatheduc = years of schooling for the father.

male = dummy variable, equals 1 if child is male.

white = dummy variable, equals 1 if child is classified as white.

10 cig per days means 4.4% lower.

- (i) In the first equation, interpret the coefficients on the variable *cigs*. In particular, what is the effect on birth weight from smoking 10 cigarettes per day?
- (ii) How much more is a white child predicted to weigh than a nonwhite child, holding the other factors in the first equation fixed? Is the difference statistically significant?
- (iii) Comment on the estimated effect and statistical significance of *motheduc*.
- (iv) From the given equation, why are you unable to compute the *F* statistic for joint significance of *motheduc* and *fatheduc*? What would you have to do to compute the *F* statistic?

5.5% $t = \frac{0.055}{0.013} = 4.23 \cdot 77. \Rightarrow \text{reject.}$
 $\Rightarrow \text{significant.}$

not significant.