

# Economics 471: Introductory Econometrics

Department of Economics, Finance and Legal Studies

University of Alabama

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Midterm II

- Key

The exam consists of three questions on four pages. Each question is of equal value.

1. Consider a random sample of data  $\{x_i, y_i\}_{i=1}^n$  and the model  $y_i = \alpha + \beta x_i + u_i$ , where  $E(u_i|x_i) = 0$  and  $V(u_i|x_i) = \sigma^2$ . Consider the estimator of  $\beta$

$$\tilde{\beta} = \frac{\sum_{i=1}^n \hat{r}_i y_i}{\sum_{i=1}^n \hat{r}_i^2},$$

where  $x_i = \gamma_0 + r_i$ . With this information, answer the following questions:

- Derive the least-squares estimator of the intercept parameter  $\alpha$ .
- What model is used to obtain  $\hat{r}_i$ ?
- Derive the least-squares estimator of the intercept parameter  $\gamma_0$ .
- Using the result from part (c), show that  $\tilde{\beta}$  is equivalent to the slope estimator we derived in class (i.e.,  $\hat{\beta} = \sum_{i=1}^n (y_i - \bar{y})(x_i - \bar{x}) / \sum_{i=1}^n (x_i - \bar{x})^2$ ).
- Noting your result from part (d), what is the conditional variance of  $\tilde{\beta}$  (i.e.,  $\text{Var}(\tilde{\beta}|x)$ )?

$$\begin{aligned} (a) \quad \sum_{i=1}^n \hat{u}_i^2 &= \sum_{i=1}^n (y_i - \hat{\alpha} - \hat{\beta} x_i)^2 \\ \frac{\partial}{\partial \hat{\alpha}} &= -2 \sum_{i=1}^n (y_i - \hat{\alpha} - \hat{\beta} x_i) = 0 \Rightarrow \\ \sum_{i=1}^n y_i - \sum_{i=1}^n \hat{\alpha} - \hat{\beta} \sum_{i=1}^n x_i &= 0 \Rightarrow \\ n \hat{\alpha} &= \sum_{i=1}^n y_i - \hat{\beta} \sum_{i=1}^n x_i = 0 \\ &= \bar{y} - \hat{\beta} \bar{x} = 0 \end{aligned}$$

$$(b) \quad x_i = \gamma_0 + r_i \Rightarrow \hat{r}_i = x_i - \hat{\gamma}_0$$

$$\begin{aligned} (c) \quad \sum_{i=1}^n \hat{r}_i^2 &= \sum_{i=1}^n (x_i - \hat{\gamma}_0)^2 \\ \frac{\partial}{\partial \hat{\gamma}_0} &= -2 \sum_{i=1}^n (x_i - \hat{\gamma}_0) = 0 \Rightarrow \sum_{i=1}^n x_i = \sum_{i=1}^n \hat{\gamma}_0 \Rightarrow \\ n \hat{\gamma}_0 &= \sum_{i=1}^n x_i \Rightarrow \hat{\gamma}_0 = \frac{1}{n} \sum_{i=1}^n x_i = \bar{x} \end{aligned}$$

$$\begin{aligned}
 (d) \quad \tilde{\beta} &= \frac{\sum_{i=1}^n \hat{r}_i y_i}{\sum_{i=1}^n \hat{r}_i^2} = \frac{\sum_{i=1}^n (x_i - \hat{r}_0) y_i}{\sum_{i=1}^n (x_i - \hat{r}_0)^2} \\
 &= \frac{\sum_{i=1}^n (x_i - \bar{x}) y_i}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2} = \hat{\beta}
 \end{aligned}$$

$$\begin{aligned}
 (e) \quad \text{since } \tilde{\beta} = \hat{\beta} \Rightarrow V(\tilde{\beta} | x) &= V(\hat{\beta} | x) \\
 &= \frac{\sigma^2}{\sum_{i=1}^n (x_i - \bar{x})^2}
 \end{aligned}$$

2. Consider a random sample of data  $\{x_{1i}, x_{2i}, y_i\}_{i=1}^n$  and the model  $y_i = \alpha + \beta_1 x_{1i} + \beta_2 x_{2i} + u_i$ , where  $E(u_i | x_{1i}, x_{2i}) = 0$ . With this information, answer the following:

- Derive the method of moments estimator of the intercept parameter  $\alpha$ .
- For the null  $H_0 : \beta_1 = 0$ , give the test statistic and the distribution of the test statistic under the null hypothesis.
- Consider the test discussed in part (b), draw the null distribution and indicate the rejection region.
- For the null  $H_0 : \beta_1 = \beta_2 = 0$ , give the test statistic and the distribution of the test statistic under the null hypothesis.
- Consider the test discussed in part (d), draw the null distribution and indicate the rejection region.

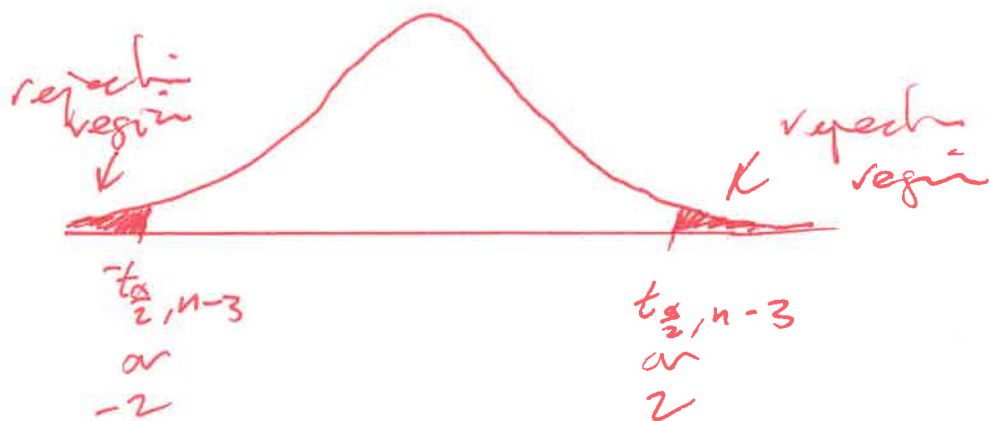
$$\begin{aligned}
 (a) \quad E(u) = 0 &\Rightarrow \frac{1}{n} \sum_{i=1}^n \hat{u}_i = 0 \Rightarrow \\
 &\frac{1}{n} \sum_{i=1}^n (y_i - \hat{\alpha} - \hat{\beta}_1 x_{1i} - \hat{\beta}_2 x_{2i}) = 0 \\
 \bar{y} - \hat{\alpha} - \hat{\beta}_1 \bar{x}_1 - \hat{\beta}_2 \bar{x}_2 &= 0 \\
 \hat{\alpha} &= \bar{y} - \hat{\beta}_1 \bar{x}_1 - \hat{\beta}_2 \bar{x}_2
 \end{aligned}$$

$$(b) \quad H_0 : \beta_1 = 0$$

$$H_1 : \beta_1 \neq 0$$

$$t = \frac{\hat{\beta}_1 - 0}{\text{se}(\hat{\beta}_1)} \sim t_{n-3}$$

(c)



(d)  $H_0: \beta_1 = \beta_2 = 0$

vs  $H_1: H_0$  is not true

restricted

$$y = \alpha + u$$

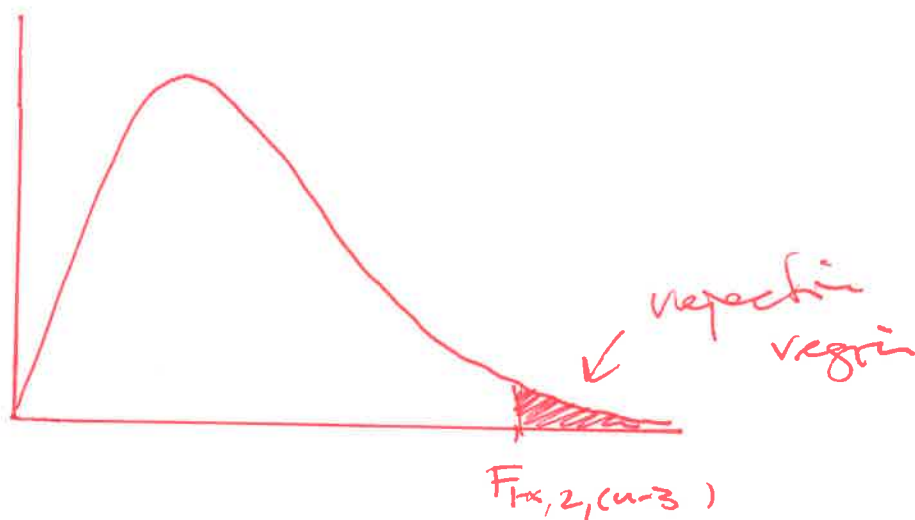
unrestricted

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

$$F = \frac{(SSR_R - SSR_U) / 2}{SSR_U / (n-3)} \sim F_{2, (n-3)}$$

$R^2$  sample is also de

(e)



3. Consider the two pieces of gretl output below which relates semester GPA (*termgpa*) to the previous semesters GPA (*priGPA*), ACT score (*ACT*), class attendance rate (*atndrte*), homework completion rate (*hwrte*), and in Model 2, the squares of *atndrte* (*sqatndrte*) and *hwrte* (*sqhwrte*) as well as the interaction of *atndrte* and *hwrte* (*atndrtehwrte*). Using this information, answer the following (be specific):

Model 1: OLS, using observations 1-680 ( $n = 674$ )

Missing or incomplete observations dropped: 6

Dependent variable: termgpa

		Coefficient	Std. Error	t-ratio	p-value
$\alpha$	const	-1.28748	0.165892	-7.761	0.0000
$\beta_1$	priGPA	0.557014	0.0423614	13.15	0.0000
$\beta_2$	ACT	0.0358472	0.00604693	5.928	0.0000
$\beta_3$	atndrte	0.0101788	0.00155644	6.540	0.0000
$\beta_4$	hwrte	0.00928213	0.00122704	7.565	0.0000
	Mean dependent var	2.613838	S.D. dependent var	0.725003	
	Sum squared resid	152.2166	S.E. of regression	0.477000	
	$R^2$	0.569704	Adjusted $R^2$	0.567131	
	$F(4, 669)$	221.4360	P-value( $F$ )	6.0e-121	
	Log-likelihood	-454.9336	Akaike criterion	919.8672	
	Schwarz criterion	942.4334	Hannan-Quinn	928.6056	

Model 2: OLS, using observations 1-680 ( $n = 674$ )

Missing or incomplete observations dropped: 6

Dependent variable: termgpa

		Coefficient	Std. Error	t-ratio	p-value
$\alpha$	const	-1.26265	0.295228	-4.277	0.0000
$\beta_1$	priGPA	0.551005	0.0426538	12.92	0.0000
$\beta_2$	ACT	0.0358668	0.00604662	5.932	0.0000
$\beta_3$	atndrte	0.00486132	0.00653044	0.7444	0.4569
$\beta_{33}$	sqatndrte	0.000108157	6.43411e-005	1.681	0.0932
$\beta_4$	hwrte	0.0121354	0.00570628	2.127	0.0338
$\beta_{44}$	sqhwrte	4.61294e-005	4.81473e-005	0.9581	0.3384
$\beta_{34}$	atndrtehwrte	-0.000128909	8.08380e-005	-1.595	0.1113
	Mean dependent var	2.613838	S.D. dependent var	0.725003	
	Sum squared resid	151.4907	S.E. of regression	0.476931	
	$R^2$	0.571756	Adjusted $R^2$	0.567255	
	$F(7, 666)$	127.0269	P-value( $F$ )	3.5e-118	
	Log-likelihood	-453.3228	Akaike criterion	922.6455	
	Schwarz criterion	958.7514	Hannan-Quinn	936.6269	

- (a) What is the marginal impact of a 1% increase in the attendance rate (*atndrte*) on semester GPA (*termgpa*) in Model 1? In Model 2?
- (b) Test the null hypothesis that the square terms (i.e., *sqatndrte* and *sqhwrt*) and interaction term (*atndrtehwrt*) are jointly irrelevant in the prediction of semester GPA (*termgpa*).
- (c) Test the null hypothesis that all of the regressors (*priGPA*, *ACT*, *atndrte* and *hwrt*) are (jointly) irrelevant in the prediction of semester GPA (*termgpa*) in Model 1.
- (d) Using any three selection criteria discussed in class, argue whether Model 1 or Model 2 is preferable?
- (e) What is the total sum of squares (SST) from Model 1? From Model 2?

(a) 
$$\frac{d\hat{y}}{datndrte} = \hat{\beta}_3 = 0.0101788$$

$$\frac{d\hat{y}}{datndrte} = \hat{\beta}_3 + 2\hat{\beta}_{33}atndrte + \hat{\beta}_{34}hwrt$$

$$= 0.00486 + 2 \cdot 0.000108atndrte + (-0.0001289)hwrt$$

(b)  $H_0: \beta_{33} = \beta_{34} = \beta_{34} = 0$

$H_1: H_0$  is not true

model 1 : restricted model

model 2 : unrestricted model

$$F = \frac{(SSR_R - SSR_U) / 3}{SSR_U / (674 - 8)} \sim F_{3, 666}$$

$$= \frac{(152.266 - 151.4907) / 3}{151.4907 / 666}$$

if  $F > F_{(1-\alpha), 3, 666} \Rightarrow$  reject  $H_0$

$$(c) H_0: \beta_1 = \beta_2 = \beta_3 = \beta_4 = 0$$

$H_1$ :  $H_0$  is not true

restricted model

$$y = \alpha + u$$

unrestricted model

model 1

$$F = \frac{(SSR_R - SSR_U) / 4}{SSR_U / (674 - 5)} \sim F_{4, 669}$$

$$= 221.436 \quad (\text{from the table})$$

w/  $p$ -value  $< 0.05 \Rightarrow$  reject  $H_0$

(d)  $R^2, SSR, \hat{\sigma}^2, LL \Rightarrow$  model 2

AIC, SC, HQ  $\Rightarrow$  model 1

(e)  $SST = \sum_{i=1}^n (y_i - \bar{y})^2$  does not depend  
upon the model so the same in both

$$\hat{\sigma}_y^2 = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2} \Rightarrow (n-1) \hat{\sigma}^2 = \sum_{i=1}^n (y_i - \bar{y})^2$$

$$(673) \left( \frac{0.477^2}{\cancel{0.477^2}} \right) = SST$$