

Economics 1123

Lecture #3

Tuesday September 26, 2006

Linear Regression with a Single Regressor, ctd.

Outline

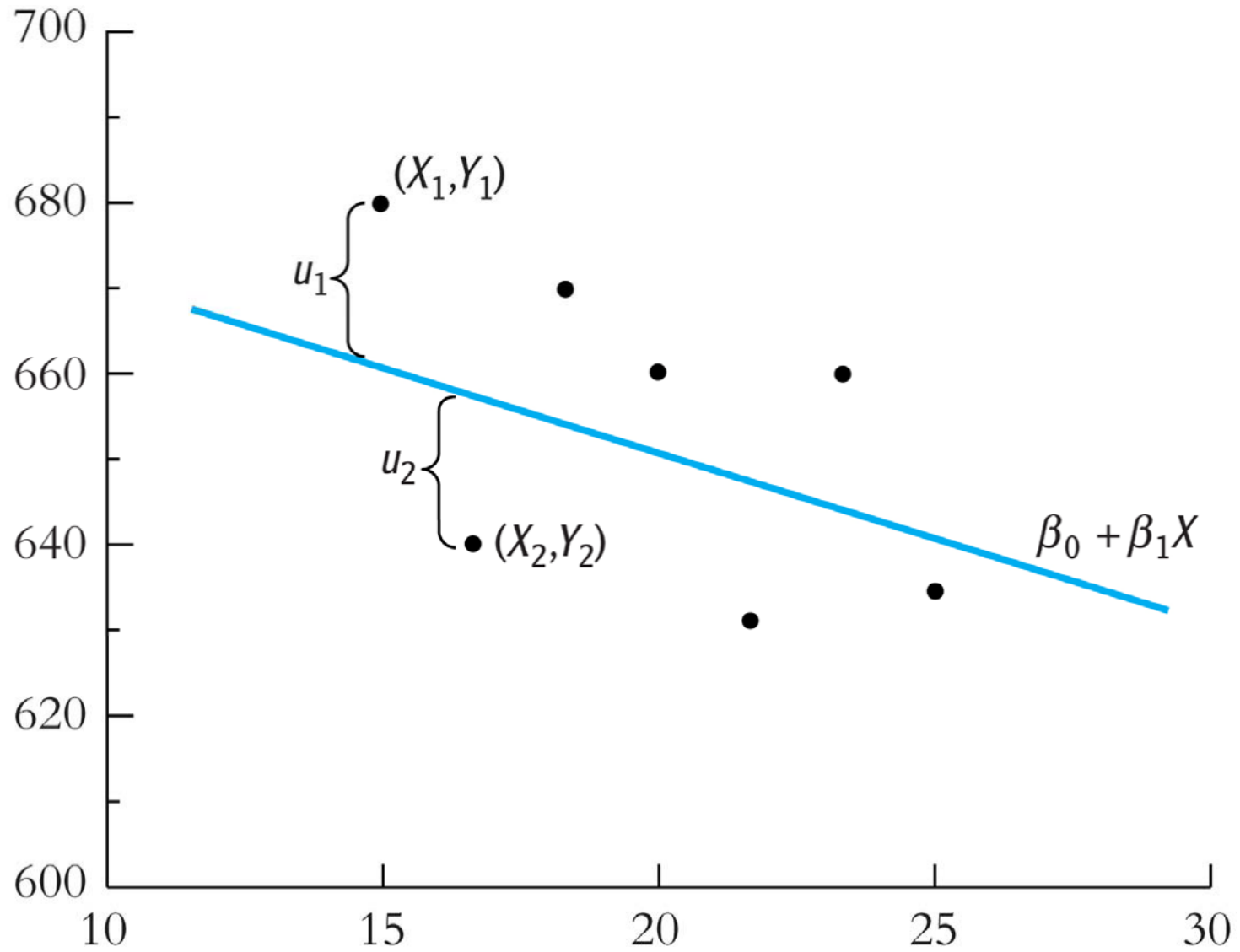
- A. Measures of fit
- B. The OLS sampling distribution
- C. Hypothesis tests and confidence intervals
- D. Misc. additional topics

The Population Linear Regression Model (Review)

$$Y_i = \beta_0 + \beta_1 X_i + u_i, i = 1, \dots, n$$

- X is the *independent variable* or *regressor*
- Y is the *dependent variable*
- $\beta_0 = \text{intercept}$
- $\beta_1 = \text{slope}$
- The *population regression line* is $E(Y|X) = \beta_0 + \beta_1 X$
- $u_i =$ the regression *error*
- The regression error consists of omitted factors, or possibly measurement error in the measurement of Y . In general, these omitted factors are other factors that influence Y , other than the variable X

In a picture:



Measures of Fit of an Estimated Regression Line

OLS divides the observation Y_i into two parts: a part that is “explained” by X_i (the predicted value) and a part that is not (the residual):

$$Y_i = \hat{Y}_i + \hat{u}_i = \text{OLS prediction} + \text{OLS residual}$$

Two important measures of fit of a regression are:

1. The *regression R^2*
2. The *standard error of the regression (SER)*

The *regression* R^2 is the fraction of the sample variance of Y_i “explained” by the regression.

$$Y_i = \hat{Y}_i + \hat{u}_i = \text{OLS prediction} + \text{OLS residual}$$

\Rightarrow sample var (Y) = sample var(\hat{Y}_i) + sample var(\hat{u}_i) (*why?*)

\Rightarrow total sum of squares = “explained” SS + “residual” SS

Definition of R^2 :

$$R^2 = \frac{ESS}{TSS} = \frac{\sum_{i=1}^n (\hat{Y}_i - \bar{\hat{Y}})^2}{\sum_{i=1}^n (Y_i - \bar{Y})^2}$$

- $R^2 = 0$ means...
- $R^2 = 1$ means...
- $0 \leq R^2 \leq 1$
- For regression with a single X , $R^2 =$ the square of the correlation coefficient between X and Y

The Standard Error of the Regression (SER) and the ***Root Mean Square Error (RMSE)*** of the residual measure the spread of the distribution of u .

$$SER = \sqrt{\frac{1}{n-2} \sum_{i=1}^n (\hat{u}_i - \bar{\hat{u}})^2} = \sqrt{\frac{1}{n-2} \sum_{i=1}^n \hat{u}_i^2}$$

$$RMSE = \sqrt{\frac{1}{n} \sum_{i=1}^n \hat{u}_i^2}$$

Notes:

- for the OLS residual, $\bar{\hat{u}} = \frac{1}{n} \sum_{i=1}^n \hat{u}_i = 0$.
- units...
- d.f. adjustment...

Example of the R^2 and the *SER*:

Teaching evaluations and instructor “beauty”

Data:

Y = teaching evaluations (course overall), scale of 1-5

X = standardized “Beauty” ratings of instructors

(other variables too – we’ll look at them later)

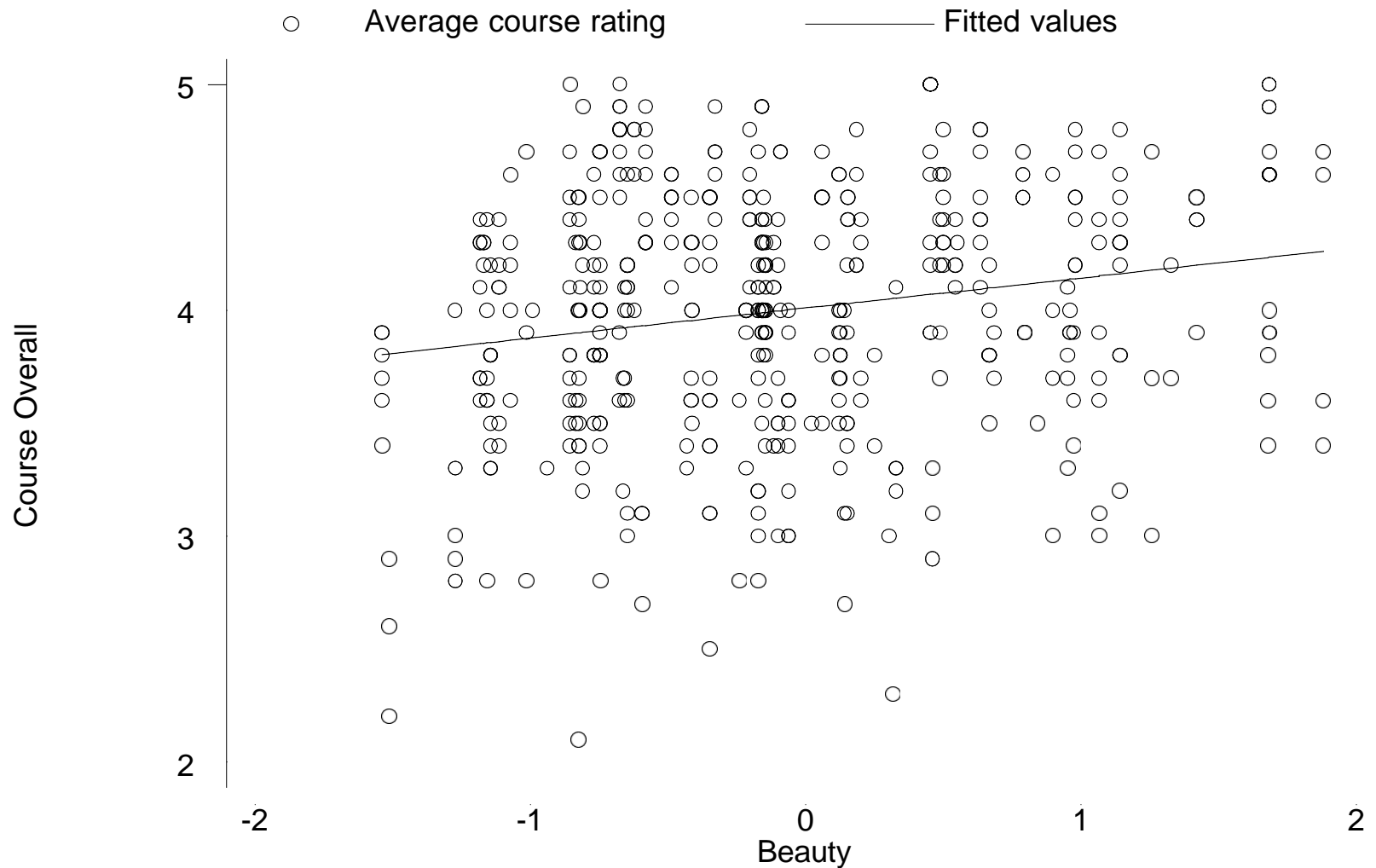
Sample: $n = 463$ courses at U.T. Austin, academic years
2000-2002

(Source: Hamermesh and Parker (2005))

Summary statistics

Variable	Obs	Mean	Std. Dev.	Min	Max
courseeval~n	463	3.998272	.	2.1	5
profevalua~n	463	4.17473	.	2.3	5
btystdave	463	-.	.	-1.538843	1.881674
lower	463	.	.	0	1
tenured	463	.	.	0	1
female	463	.	.	0	1
nonenglish	463	.	.	0	1
tenuretrack	463	.	.	0	1

Scatterplot of “course overall” ratings v. Beauty:



Scatterplot and OLS Regression Line

correlation coefficient = 0.189

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