

Lecture 14 - Economics 113

Professor Spearot

- ▶ Agenda
 1. P-values
 2. Comparing different parameters
 3. Prediction Intervals
- ▶ Homework #1 is posted
- ▶ Exam #3 on Friday

P-values

A different approach

- ▶ How likely is it that I falsely reject the null?
- ▶ P-value:

The probability that the null hypothesis is falsely rejected

- ▶ In the crime and enrollment example

$$H_0 : \beta_{Enroll} = 1, H_A : \beta_{Enroll} \neq 1$$

- ▶ T-statistic

$$t_{stat} = \frac{1.27 - 1}{0.11} = 2.45$$

- ▶ The p-value is the value of the following

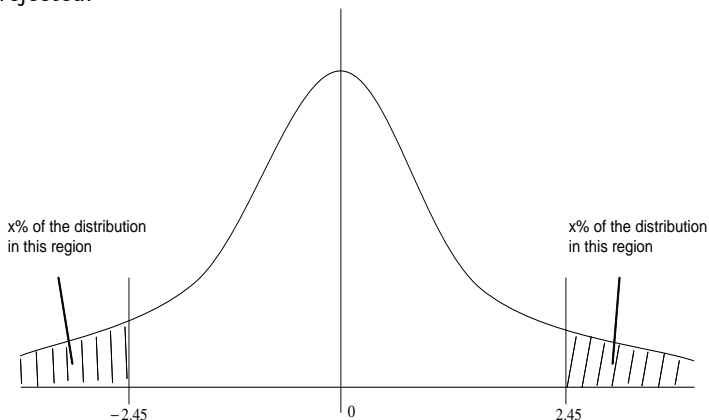
$$pvalue = \Pr(|T| > 2.45)$$

- ▶ The p-value is the probability that I randomly draw a value from the t-distribution that is larger (in absolute terms) than the estimated T-statistic

P-values

An illustration

- ▶ P-value: The smallest significance level such that the null would be rejected.



- ▶ The p-value for this is the probability of being in the shaded region.

P-values

A different approach

- ▶ To find the p-value
 1. Calculate the t-statistic
 2. Find the closest match to your t-statistic on the normal table
 3. The p-value is the significance level that generates this t-statistic.
- ▶ In the crime and enrollment example

1. $t_{stat} = 2.45$
2. $\Pr(|T| > 2.45)$?

$$\begin{aligned}\Pr(|T| > 2.45) &= \Pr(T > 2.45 \cup T < -2.45) \\ &= \Pr(T > 2.45 \cup T < -2.45) \\ &= \Pr(T > 2.45) + \Pr(T < -2.45) = 2 * \Pr(T > 2.45) \\ &= 2 * (1 - 0.9929) = 0.0142\end{aligned}$$

- ▶ Thus, we will falsely reject the null less than 1.42% of the time.

Linear combinations of parameters

Returns to schooling

- ▶ Suppose that we estimated

$$\widehat{\log(\text{Wage})} = \underset{(0.021)}{1.472} + \underset{(0.0068)}{0.0667} JC + \underset{(0.0023)}{0.0769} Univ + \underset{(0.0002)}{0.0049} exper$$
$$obs = 6763, R^2 = 0.222$$

1. *JC* is years at a junior college
 2. *Univ* is years at a university
 3. *exper* is experience.
- ▶ Clearly, $\hat{\beta}_1 < \hat{\beta}_2$.
 - ▶ What does this mean?
 - ▶ Is this significant?
 - ▶ How do we test $H_0 : \beta_1 = \beta_2$
 - ▶ $\beta_1 = \beta_2$ is the same as $\beta_1 - \beta_2 = 0$

Linear combinations of parameters

Returns to schooling

- ▶ Define $\theta = \beta_1 - \beta_2$.
- ▶ $H_0 : \theta = 0$.
- ▶ $H_A : \theta \neq 0$.
- ▶ How do we integrate θ into our regression?
 - ▶ Solve for $\beta_1 = \theta + \beta_2$
 - ▶ Substitute for β_1

$$\log(\text{Wage}) = \beta_0 + \beta_1 JC + \beta_2 Univ + \beta_3 exper + u$$

$$\log(\text{Wage}) = \beta_0 + (\theta + \beta_2) JC + \beta_2 Univ + \beta_3 exper + u$$

- ▶ Collect parameters

$$\log(\text{Wage}) = \beta_0 + \theta JC + \beta_2 (Univ + JC) + \beta_3 exper + u$$

- ▶ $Univ + JC$ is total schooling
- ▶ θ the effect of a JC year, controlling for total schooling.

Linear combinations of parameters

Returns to schooling

- ▶ Modified Model:

$$\log(\widehat{Wage})_{obs} = \underset{(0.021)}{1.472} - \underset{(0.0069)}{0.0102} JC + \underset{(0.0023)}{0.0769} (Univ + JC) + \underset{(0.0002)}{0.0049} exper$$

$obs = 6763, R^2 = 0.222$

- ▶ Can we reject $H_0 : \theta = 0$ in favor of $H_A : \theta \neq 0$?
- ▶ 95% Confidence interval?
- ▶ Use the formula

$$\begin{aligned} -0.0102 - 1.96 * 0.0069 &< \theta < -0.0102 + 1.96 * 0.0069 \\ -0.023724 &< \theta < 0.003324 \end{aligned}$$

- ▶ Interpret?

Predictions

Introduction

- ▶ Suppose you start with the equation:

$$y = \beta_0 + \beta_1 x_1 + \cdots + \beta_k x_k + u$$

- ▶ You may wish to predict y for various individuals which are not in the sample.
- ▶ Solving for predictions is easy
 - Plug $x_1 = c_1, \dots, x_k = c_k$ into equation
 - Produce prediction, $\hat{\theta}$
- ▶ Also might want Standard Errors. Why?
- ▶ The prediction may be precise or imprecise - need standard errors to figure this out.

Predictions

Getting standard errors

- ▶ Prediction:

$$\theta = \beta_0 + \beta_1 c_1 + \cdots + \beta_k c_k$$

- ▶ Solve for β_0 :

$$\beta_0 = \theta - \beta_1 c_1 - \cdots - \beta_k c_k$$

- ▶ Plug into estimating equation

$$y = (\theta - \beta_1 c_1 - \cdots - \beta_k c_k) + \beta_1 x_1 + \cdots + \beta_k x_k + u$$

- ▶ Simplify:

$$y = \theta + \beta_1 (x_1 - c_1) + \cdots + \beta_k (x_k - c_k) + u$$

- ▶ Estimate \rightarrow gives us prediction θ and standard error.

Predictions

Earnings Example

- ▶ Predicted earnings of a person with
 - 10 years of education
 - 2 years of experience
 - 1 year on tenure
- ▶ Want prediction and standard error.
- ▶ Estimate:

$$wage = \theta + \beta_1 (educ - 10) + \beta_2 (exper - 2) + \beta_3 (tenure - 1) + u$$

- ▶ Confidence interval for θ ?