

# Lecture 17 - Economics 113

Professor Spearot

- ▶ Agenda
  1. F-Tests - General Regression Significance
  2. Quadratics
  3. Interactions
- ▶ Homework 1 due next friday. - LONG!!!
- ▶ Midterm 4 is next Friday, in class.

# Multiple restrictions

## The F-Test

- ▶ Compare a base model with a restricted model
- ▶ Use the F-Statistic

$$F_{stat} = \frac{\text{Average loss in explanatory power due to restrictions}}{\text{Average unexplained variation in base model}}$$

- ▶ If  $F$  is high  $\Rightarrow$  we lose a ton by our restrictions.
- ▶ Under the null,  $F_{stat}$  is distributed according to an "F-distribution"
  - ▶ Even if restrictions are "good", large  $F_{stat}$  is randomly possible, but unlikely
- ▶ Compare  $F_{stat}$  to the "F Distribution"
  1. Choose a significance level (say 5%)
  2. Using the 5% F-table, find the critical value,  $F_{crit}$ .
  3. If  $F > F_{crit}$ , reject the null (reject the restricted model in favor of the unrestricted model)

# The F-Test

## General Regression Significance

- ▶ Unrestricted model

$$\log(\text{wage}) = \beta_0 + \beta_1 \text{educ} + \beta_2 \text{exper} + \beta_3 \text{tenure} + u$$

- ▶ Test  $H_0 : \beta_1 = 0, \beta_2 = 0, \beta_3 = 0$

- ▶ Restricted model

$$\log(\text{wage}) = \beta_0 + u$$

- ▶ Create a new dependent variable
- ▶ Run regressions, conduct F Test.

# Regressions

## Quadratic terms

- ▶ The basic model

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_1^2 + u$$

- ▶ How should we interpret  $\beta_1$ ,  $\beta_2$ ?
  - Together
- ▶ Take the derivative w.r.t  $x_1$ .

$$\frac{\partial y}{\partial x_1} = \beta_1 + 2\beta_2 x_1$$

- ▶ If  $\beta_2 \neq 0$ , then we can solve for the maximum or minimum

$$\begin{aligned}\beta_1 + 2\beta_2 x_1^* &= 0 \\ x_1^* &= -\frac{\beta_1}{2\beta_2}\end{aligned}$$

- ▶ What determines whether it is a maximum or minimum?

# Example

## Wages and Age

- ▶ Regression Equation

$$\log(\text{wage}) = \beta_0 + \beta_{\text{age}} \text{age} + \beta_{\text{age}^2} \text{age}^2 + \beta_{\text{educ}} \text{educ} + \beta_{\text{exper}} \text{exper} + u$$

- ▶ Conditional on education and experience, how does age affect wages?
- ▶ Marginal effect of age on education

$$\frac{\partial \log(\text{wage})}{\partial \text{age}} = \beta_{\text{age}} + 2\beta_{\text{age}^2} \text{age}$$

- ▶ Estimate!

# Regressions

## Quadratic terms

- ▶ Determining wages
- ▶ Base model

$$\log(\text{wage}) = \beta_0 + \beta_1 \text{educ} + \beta_2 \text{exper} + \beta_3 \text{tenure} + u$$

- ▶ Squared model

$$\begin{aligned} \log(\text{wage}) = & \beta_0 + \beta_1 \text{educ} + \beta_2 \text{exper} + \beta_3 \text{tenure} \\ & + \beta_4 \text{exper}^2 + \beta_5 \text{tenure}^2 \end{aligned}$$

- ▶ How do we interpret? Should we keep in the squared terms?
- ▶  $H_0 : \beta_4 = 0, \beta_5 = 0$

# Regressions

## Interaction terms

- ▶ The basic model

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_1 x_2 + \beta_3 x_2 + u$$

- ▶ How do we interpret  $\beta_1$ ,  $\beta_2$ ?
- ▶ Take the derivative w.r.t  $x_1$ .

$$\frac{\partial y}{\partial x_1} = \beta_1 + \beta_2 x_2$$

- ▶ How do we interpret  $\frac{\partial y}{\partial x_1}$ ?

- $\Rightarrow$  If  $\beta_2 > 0$ , then higher  $x_2$  increases  $\frac{\partial y}{\partial x_1}$
- $\Rightarrow$  If  $\beta_2 < 0$ , then higher  $x_2$  decreases  $\frac{\partial y}{\partial x_1}$

# Regressions

## Interaction terms

- ▶ Interact *age* with *educ*

$$\log(\text{wage}) = \beta_0 + \beta_1 \text{educ} + \beta_2 \text{educ} * \text{age} + \beta_3 \text{age} + \beta_4 \text{exper} + u$$

- ▶ How do we interpret these results?
- ▶ Do they make sense?
- ▶ Should we keep in the interaction term?