

## Hypothesis Testing

We've tested whether a coefficient in our population model was equal to zero.

- Gave conditions under which coefficients had normal sampling distribution
- Used  $t$ -distribution for hypothesis test
- Used simple  $t$ -test to see if each coefficient is different than zero
  - Automatically provided by R in regression output

**Specification can be used to test a wide variety of hypotheses.**

## Testing If Parameters Are Different

- $\log(\text{wage}) = \beta_0 + \beta_1 \text{jc} + \beta_2 \text{univ} + \beta_3 \text{exper} + u$ 
  - $\text{jc}$  = years at two-year junior college
  - $\text{univ}$  = years at four-year university
- Example: modeling wage as function of years spent at a two-year junior college and a four-year university
- Are the two coefficients different?
- Test  $H_0 : \beta_1 - \beta_2 = 0$  against  $H_0 : \beta_1 - \beta_2 \neq 0$

## Setting Up Test Statistic

$$t = \frac{\hat{\beta}_1 - \hat{\beta}_2}{se(\hat{\beta}_1 - \hat{\beta}_2)}$$

- Estimating standard error, so statistic has  $t$ -distribution.
- R won't give us this statistic if we run the linear regression.
- We can't compute  $t$ -statistic we want from output in R.
- What's the standard error of the difference in the denominator?

$$se(\hat{\beta}_1 - \hat{\beta}_2) = \sqrt{\text{Var}(\hat{\beta}_1 - \hat{\beta}_2)} = \sqrt{\text{Var}(\hat{\beta}_1) + \text{Var}(\hat{\beta}_2) - 2\text{Cov}(\hat{\beta}_1, \hat{\beta}_2)}$$

- R tells us variance of  $\hat{\beta}_1$  and  $\hat{\beta}_2$  but not covariance.

## An Alternate Method

- Change the variables:
  - Define  $\theta_1 = \beta_1 - \beta_2$
  - Then  $H_0 : \beta_1 - \beta_2 = 0$  is equivalent to  $H_0 : \theta_1 = 0$
- Rewrite population model:

$$\begin{aligned}\log(\text{wage}) &= \beta_0 + (\theta_1 + \beta_2)jc + \beta_2\text{univ} + \beta_3\text{exper} + u \\ &= \beta_0 + \theta_1 jc + \beta_2(jc + \text{univ}) + \beta_3\text{exper} + u\end{aligned}$$

- New variable that represents total years in any college:  $jc + \text{univ}$ 
  - Create this variable in our data table, then estimate regression.
  - R estimates  $\theta_1$  and tests whether it's statistically significant.
  - If  $\theta_1$  is statistically significant, reject original null hypothesis.

## OLS Regression Results

$$\begin{aligned}\hat{\log}(\text{wage}) &= 1.472 - .0102jc + .0769\text{totcoll} + .0049\text{exper} \\ &\quad (.021) \quad (.0069) \quad (.0023) \quad (.0002)\end{aligned}$$

$$n = 6,763, R^2 = .222$$

$$t = -.0102/.0069 = -1.48$$

$$p = \Pr(|t| > 1.48) = 0.139$$

- Coefficient for  $jc$  is less than twice its standard error, so not significant.
- $p = 0.139$ , so we cannot reject hypothesis that the effects of either type of college are equal.

**This method can be generalized to test any linear combination of parameters.**