

### Chapter 6 (Ch. 7 of the textbook)

Wooldridge: Introductory Econometrics:

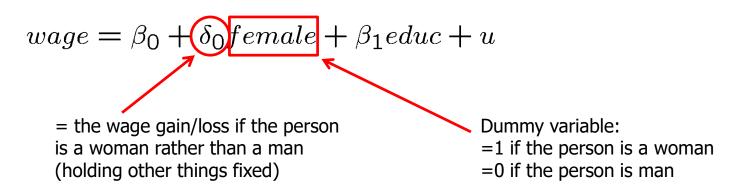
A Modern Approach, 5e



#### Qualitative Information

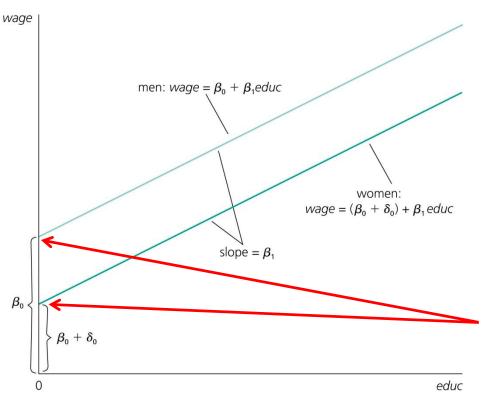
- Examples: gender, race, industry, region, rating grade, ...
- A way to incorporate qualitative information is to use dummy variables
- They may appear as the dependent or as independent variables

#### A single dummy independent variable





#### Graphical Illustration



Alternative interpretation of coefficient:

$$\delta_0 = E(wage|female = 1, educ)$$
 $-E(wage|female = 0, educ)$ 

i.e. the difference in mean wage between men and women with the same level of education.

Intercept shift



#### Dummy variable trap

This model cannot be estimated (perfect collinearity)

$$wage = \beta_0 + \gamma_0 male + \delta_0 female + \beta_1 educ + u$$

When using dummy variables, one category always has to be omitted:

$$wage=eta_0+\delta_0female+eta_1educ+u$$
 — The base category are men 
$$wage=eta_0+\gamma_0male+eta_1educ+u$$
 — The base category are women

Alternatively, one could omit the intercept: <

$$wage = \gamma_0 male + \delta_0 female + \beta_1 educ + u$$

#### Disadvantages:

- 1) More difficult to test for differences between the parameters
- 2) R-squared formula only valid if regression contains intercept

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#### Estimated wage equation with intercept shift

$$\widehat{wage} = -1.57 \leftarrow 1.81 \ female + .572 \ educ$$
 
$$(.72) \leftarrow (.26) \leftarrow (.049)$$
 
$$+ .025 \ exper + .141 \ tenure$$
 
$$(.012) \leftarrow (.021)$$
 Holding education, experience, and tenure fixed, women earn 1.81\$ less per hour than men

$$n = 526, R^2 = .364$$

#### Does that mean that women are discriminated against?

Not necessarily. Being female may be correlated with other productivity characteristics that have not been controlled for.

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#### Comparing means of subpopulations described by dummies

$$\widehat{wage} = 7.10 - 2.51 female$$
 (.21) (.26)

$$n = 526, R^2 = .116$$

Not holding other factors constant, women earn 2.51\$ per hour less than men, i.e. the difference between the mean wage of men and that of women is 2.51\$.

#### Discussion

- It can easily be tested whether difference in means is significant
- The wage difference between men and women is larger if no other things are controlled for; i.e. part of the difference is due to differences in education, experience and tenure between men and women

Further example: Effects of training grants on hours of training

Hours training per employee Dummy indicating whether firm received training grant 
$$hr\widehat{semp} = 46.67 + 26.25 \underbrace{grant}_{0.98} - 0.98 \underbrace{\log(sales)}_{0.54}$$
 
$$- 6.07 \underbrace{\log(employ)}_{0.59}, \ n = 105, R^2 = .237$$
 
$$(3.88)$$

- This is an example of <u>program evaluation</u>
  - Treatment group (= grant receivers) vs. control group (= no grant)
  - Is the effect of treatment on the outcome of interest <u>causal</u>?



$$\widehat{\log}(price) = -1.35 + .168 \log(lotsize) + .707 \log(sqrft)$$

$$(.65) \quad (.038) \qquad (.093)$$

$$+ .027 \quad bdrms + .054 \quad colonial \quad whether house is of colonial style$$

$$n = 88, R^2 = .649$$

$$\Rightarrow \frac{\partial \log(price)}{\partial colonial} = \frac{\%\partial price}{\partial colonial} = 5.4\%$$

As the dummy for colonial style changes from 0 to 1, the house price increases by 5.4%



#### Using dummy variables for multiple categories

- 1) Define membership in each category by a dummy variable
- 2) Leave out one category (which becomes the base category)

$$\widehat{\log}(wage) = .321 + .213 \ marrmale = .198 \ marrfem \ (.100) \ (.055)$$

$$- .110 \ singfem + .079 \ educ + .027 \ exper - .00054 \ exper^2 \ (.056)$$

$$- (.007) \ (.005) \ (.00011)$$
Holding other things fixed, married women earn 19.8% less than single men (= the base category) 
$$+ .029 \ tenure - .00053 \ tenure^2 \ (.00023)$$



- Using dummy variables for multiple categories
- **Exact variations:**  $(e^{0.213} 1) \times 100\% = 23.74\%$

$$(e^{-0.198}-1)\times 100\% = -17.96\%$$
  $(e^{-0.11}-1)\times 100\% = -10.42\%$ 

$$\widehat{\log}(wage) = .321 + .213 \frac{marrmale}{(.100)} - .198 \frac{marrfem}{(.058)}$$

$$-.110 \ singfem + .079 \ educ + .027 \ exper - .00054 \ exper^2$$
 (.056) (.007) (.005) (.00011)

Holding other things fixed, married women earn 19.8% less than single men (= the base category)

$$+ .029 \ tenure - .00053 \ tenure^{2}$$
 (.007) (.00023)



<u>Using dummy variables for multiple categories</u>

**Changing the base category** 

$$log(wage) = 0.123 + 0.411 marrmale + 0.198 singmale + 0.088 singfemale + 0.079 educ + ...$$

Holding other things fixed, single men earn 19.8% more than married women (= the base caegory)



- Incorporating ordinal information using dummy variables
- Example: City credit ratings and municipal bond interest rates

Municipal bond rate 
$$\text{Credit rating from 0-4 (0=worst, 4=best)}$$
 
$$MBR = \beta_0 + \beta_1 CR + other \ factors$$

This specification would probably not be appropriate as the credit rating only contains ordinal information. A better way to incorporate this information is to define dummies:

$$MBR = \beta_0 + \delta_1 CR_1 + \delta_2 CR_2 + \delta_3 CR_3 + \delta_4 CR_4 + other \ factors$$

Dummies indicating whether the particular rating applies, e.g.  $CR_1=1$  if CR=1 and  $CR_1=0$  otherwise. All effects are measured in comparison to the worst rating (= base category).



Interaction term

#### Interactions involving dummy variables

**Allowing for different slopes** 

$$\log(wage) = \beta_0 + \delta_0 female + \beta_1 educ + \delta_1 female \cdot educ + u$$

$$\beta_0$$
 = intercept men

$$\beta_1$$
 = slope men

$$\beta_0 + \delta_0 = \text{intercept women}$$
  $\beta_1 + \delta_1 = \text{slope women}$ 

$$\beta_1 + \delta_1 = \text{slope women}$$

**Interesting hypotheses** 

$$H_0:\delta_1=0$$

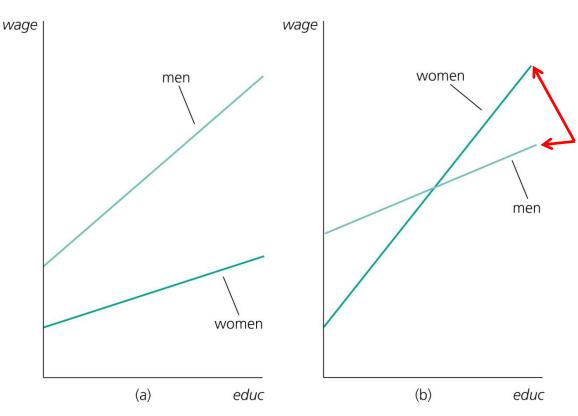
$$H_0: \delta_0 = 0, \delta_1 = 0$$

The return to education is the same for men and women

The whole wage equation is the same for men and women



#### Graphical illustration



Interacting both the intercept and the slope with the female dummy enables one to model completely independent wage equations for men and women



#### Estimated wage equation with interaction term

$$\widehat{\log}(wage) = .389 - .227 female - .082 educ$$

$$(.119) (.168) (.008)$$

$$- .0056 female \cdot educ + .029 exper - .00058 exper^{2}$$

$$(.00131) (.00011)$$

$$+ .032 tenure - .00059 tenure^{2}, n = 526, R^{2} = .441$$

$$(.007) (.00024)$$
Does this mean that there is no significant eviden

No evidence against hypothesis that the return to education is the same for men and women Does this mean that there is no significant evidence of lower pay for women at the same levels of educ, exper, and tenure? No: this is only the effect for educ = 0. To answer the question one has to recenter the interaction term, e.g. around educ = 12.5 (= average education).

- Testing for differences in regression functions across groups
- Unrestricted model (contains full set of interactions)

College grade point average Standardized aptitude test score High school rank percentile 
$$cumgpa = \beta_0 + \delta_0 female + \beta_1 sat + \delta_1 female sat + \beta_2 hsperc \\ + \delta_2 female hsperc + \beta_3 tothrs + \delta_3 female tothrs + u$$

Restricted model (same regression for both groups)

Total hours spent in college courses

$$cumgpa = \beta_0 + \beta_1 sat + \beta_2 hsperc + \beta_3 tothrs + u$$



#### Null hypothesis

$$H_0: \delta_0 = 0, \delta_1 = 0, \delta_2 = 0, \delta_3 = 0$$

All interaction effects are zero, i.e. the same regression coefficients apply to men and women

#### Estimation of the unrestricted model

$$cu\widehat{mgpa} = 1.48 - .353 \ female + .0011 \ sat + .00075 \ female \cdot sat$$
(.21) (.411) (.0002) (.00039)

$$-.0085 \ hisperc - .00055 \ female \cdot hisperc$$
 $(.0014)$ 
 $(.00316)$ 

Tested individually, the hypothesis that the interaction effects are zero cannot be rejected

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#### Joint test with F-statistic

$$F = \frac{(SSR_r - SSR_{ur})/q}{SSR_{ur}/(n-k-1)} = \frac{(85.515 - 78.355)/4}{78.355/(366 - 7 - 1)} \approx 8.18$$

#### Alternative way to compute F-statistic

■ Run separate regressions for men and for women; the unrestricted SSR is given by the sum of the SSR of these two regressions: SSR<sub>1</sub> + SSR<sub>2</sub>

Null hypothesis is rejected

- Run regression for the restricted model and store SSR: SSR<sub>P</sub>
- If the test is computed in this way it is called the <u>Chow-Test</u>
- Important: Test assumes a constant error variance accross groups



**CHOW test:** 
$$H_0: \beta_{01} = \beta_{02}, \beta_{11} = \beta_{12}, \beta_{21} = \beta_{22}, ..., \beta_{k1} = \beta_{k2}$$

$$F = \frac{SSR_p - (SSR_1 + SSR_2)}{SSR_1 + SSR_2} \times \frac{n - 2(k+1)}{k+1} \sim F(k+1, n-2(k+1))$$

If the <u>restricted regression contains a dummy for an intercept shift</u> the statistic to test <u>changes only in the slope coefficients</u> is:

$$H_0: \beta_{11} = \beta_{12}, \beta_{21} = \beta_{22}, ..., \beta_{k1} = \beta_{k2}$$

$$F = \frac{SSR_p^* - (SSR_1 + SSR_2)}{SSR_1 + SSR_2} \times \frac{n - 2(k+1)}{k} \sim F(k, n - 2(k+1))$$

With  $SSR_P^*$  the sum of squared residuals of the restricted regression



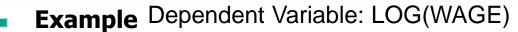
#### Example

Dependent Variable: LOG(WAGE)

Method: Least Squares

Included observations: 526

| Variable                      | Coefficient                                   | Std. Error                                   | t-Statistic                                   | Prob.                                |
|-------------------------------|-----------------------------------------------|----------------------------------------------|-----------------------------------------------|--------------------------------------|
| C<br>EDUC<br>EXPER<br>EXPER^2 | 0.127998<br>0.090366<br>0.041009<br>-0.000714 | 0.105932<br>0.007468<br>0.005197<br>0.000116 | 1.208296<br>12.10041<br>7.891606<br>-6.163888 | 0.2275<br>0.0000<br>0.0000<br>0.0000 |
| Sum squared resid             | 103.7904                                      |                                              |                                               |                                      |



Method: Least Squares Sample: IF FEMALE=1

Included observations: 252

| Variable | Coefficient | Std. Error | t-Statistic | Prob.  |
|----------|-------------|------------|-------------|--------|
| C        | 0.266084    | 0.141535   | 1.879985    | 0.0613 |
| EDUC     | 0.079195    | 0.010369   | 7.637838    | 0.0000 |
| EXPER    | 0.022372    | 0.006664   | 3.356971    | 0.0009 |
| EXPER^2  | -0.000423   | 0.000148   | -2.853701   | 0.0047 |

Sum squared resid

38.38393

Example Dependent Variable: LOG(WAGE)

Method: Least Squares Sample: IF FEMALE=0

Included observations: 274

| Variable | Coefficient | Std. Error | t-Statistic | Prob.  |
|----------|-------------|------------|-------------|--------|
| C        | 0.157291    | 0.136403   | 1.153132    | 0.2499 |
| EDUC     | 0.090354    | 0.009267   | 9.750477    | 0.0000 |
| EXPER    | 0.054017    | 0.006743   | 8.011343    | 0.0000 |
| EXPER^2  | -0.000914   | 0.000150   | -6.079015   | 0.0000 |

Sum squared resid

47.35130



Dependent Variable: LOG(WAGE)

Method: Least Squares

Included observations: 526

| Variable | Coefficient | Std. Error | t-Statistic | Prob.  |
|----------|-------------|------------|-------------|--------|
|          | 0.000.400   | 0.400040   | 0.000440    | 0.0004 |
| С        | 0.390483    | 0.102210   | 3.820413    | 0.0001 |
| FEMALE   | -0.337187   | 0.036321   | -9.283424   | 0.0000 |
| EDUC     | 0.084136    | 0.006957   | 12.09407    | 0.0000 |
| EXPER    | 0.038910    | 0.004824   | 8.066683    | 0.0000 |
| EXPER^2  | -0.000686   | 0.000107   | -6.388842   | 0.0000 |

Sum squared resid 89.05862



#### **The CHOW test for the example:**

$$H_0: \beta_{01} = \beta_{02}, \beta_{11} = \beta_{12}, \beta_{21} = \beta_{22}, \beta_{31} = \beta_{32}$$

$$F = \frac{103.7904 - (38.3839 + 47.3513)}{38.3839 + 47.3513} \times \frac{526 - 2(3+1)}{3+1} = 27.272 > F_{0.05}(4,518) = 2.39$$

#### To test changes only in the slope coefficients in the example:

$$H_0: \beta_{11} = \beta_{12}, \beta_{21} = \beta_{22}, \beta_{31} = \beta_{32}$$

$$F = \frac{89.0586 - (38.3839 + 47.3513)}{38.3839 + 47.3513} \times \frac{526 - 2(3+1)}{3} = 6.6931 > F_{0.05}(3,518) = 2.62$$