

## POLS 6481. Research Design and Quantitative Methods II

### Lecture 26. Interactions I: Multiplying Variables

Readings: Wooldridge, *Introductory Econometrics 5e*, 7.2a + 7.4a

Kam & Franzese, p. 27–29, 54–56, 67–68, 85–86

#### Outline

1. A Single Dummy for a Single Dimension
2. Multiple Dummies for Multiple Dimensions
3. Multiple Dummies for Multiple Dimensions – Interaction Version
4. Gender & Party and Social Welfare Opinions

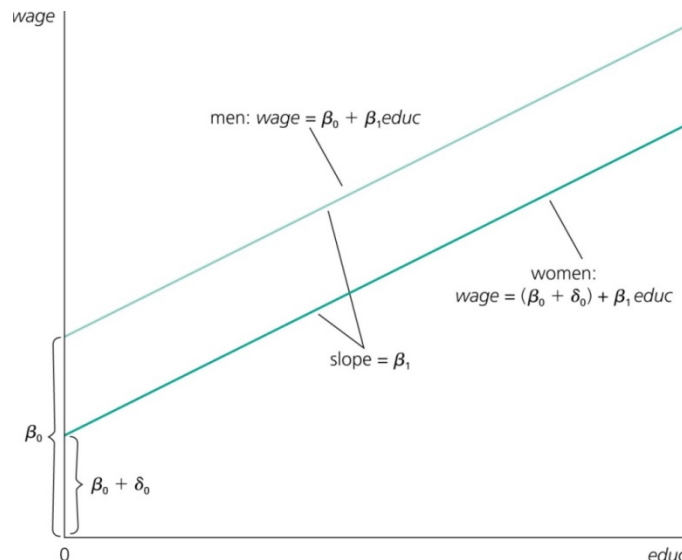
## 1. A Single Dummy for a Single Dimension

Combined model:  $wage = \beta_0 + \delta_0 \cdot female + \beta_1 educ + u$

men only:  $wage = \beta_0 + \beta_1 educ + u$

women only:  $wage = (\beta_0 + \delta_0) + \beta_1 educ + u$

Intercept shift interpretation:  $E(wage \mid female=1, educ = 0) = \beta_0 + \delta_0$   
 $E(wage \mid female=0, educ = 0) = \beta_0$



## 1. A Single Dummy for a Single Dimension

Combined model:  $wage = \beta_0 + \delta_0 \cdot female + \beta_1 educ + u$

men only:  $wage = \beta_0 + \beta_1 educ + u$

women only:  $wage = (\beta_0 + \delta_0) + \beta_1 educ + u$

Treatment effect interpretation:  $\delta_0 = E(wage \mid female=1, educ) - E(wage \mid female=0, educ)$

## 2. Multiple Dummies for Multiple Dimensions

New model

$$\begin{aligned} wage = & \beta_0 + \beta_1 educ + u \\ & + \gamma_0 \textit{married man} \\ & + \delta_0 \textit{single woman} \\ & + \lambda_0 \textit{married woman} \end{aligned}$$

$\gamma_0$ ,  $\delta_0$ , and  $\lambda_0$  represent intercept shifts; each is relative to the omitted category (*single man*)

## 2. Multiple Dummies for Multiple Dimensions

for single men:  $wage = \beta_0 + \beta_1 educ + u$

for married men:  $wage = (\beta_0 + \gamma_0) + \beta_1 educ + u$

for single women:  $wage = (\beta_0 + \delta_0) + \beta_1 educ + u$

for married women:  $wage = (\beta_0 + \lambda_0) + \beta_1 educ + u$

Effects of gender:

Among unmarried:  $E(wage \mid female=1, married=0, educ=0) = \beta_0 + \delta_0$

$$E(wage \mid female=0, married=0, educ=0) = \beta_0$$

(+  $\delta_0$ )

Among married:  $E(wage \mid female=1, married=1, educ=0) = \beta_0 + \lambda_0$

$$E(wage \mid female=0, married=1, educ=0) = \beta_0 + \gamma_0$$

(+  $\lambda_0 - \gamma_0$ )

## 2. Multiple Dummies for Multiple Dimensions

for single men:  $wage = \beta_0 + \beta_1 educ + u$

for married men:  $wage = (\beta_0 + \gamma_0) + \beta_1 educ + u$

for single women:  $wage = (\beta_0 + \delta_0) + \beta_1 educ + u$

for married women:  $wage = (\beta_0 + \lambda_0) + \beta_1 educ + u$

Effects of marriage:

Among males:  $E(wage \mid female=0, married = 1, educ = 0) = \beta_0 + \gamma_0$   
 $E(wage \mid female=0, married = 0, educ = 0) = \beta_0$

(+  $\gamma_0$ )

Among females:  $E(wage \mid female=1, married = 1, educ = 0) = \beta_0 + \lambda_0$   
 $E(wage \mid female=1, married = 0, educ = 0) = \beta_0 + \delta_0$

( $\lambda_0 - \delta_0$ )

## 2. Multiple Dummies for Multiple Dimensions

Example 7.6 originally... (Wooldridge 235–236)

$$\log(\widehat{wage}) = .321 + .213 \text{ marr.male} - .110 \text{ sing.fem} - .198 \text{ marr.fem} + .079 \text{ educ} + \dots$$

(.100) (.055) (.056) (.058) (.007)

$$E(wage \mid female=0, married = 0, educ = 0, \dots) = .321 \quad [\text{baseline}]$$

$$E(wage \mid female=0, married = 1, educ = 0, \dots) = .534$$

$$E(wage \mid female=1, married = 0, educ = 0, \dots) = .211$$

$$E(wage \mid female=1, married = 1, educ = 0, \dots) = .123$$

Effects of marriage – for men:	$+ .213 = .534 - .321$
for women:	$-.088 = .123 - .211 \quad (= -.198 - -.110)$

Effects of gender – for singles:	$-.110 = .211 - .321$
for marrieds:	$-.411 = .123 - .534 \quad (= -.198 - .213)$

### 3. Multiple Dummies for Multiple Dimensions – Interaction Version

Example 7.6 continues... (Wooldridge 240)

$$\log(\widehat{wage}) = .321 + .213 \text{ married} - .110 \text{ female} - .301 \text{ marr*female} + .079 \text{ educ} + \dots$$

(.100) (.055) (.056) (.072) (.007)

$$E(wage \mid female=0, married = 0, educ = 0, \dots) = .321 \quad [\text{baseline}]$$

$$E(wage \mid female=0, married = 1, educ = 0, \dots) = .534$$

$$E(wage \mid female=1, married = 0, educ = 0, \dots) = .211$$

$$E(wage \mid female=1, married = 1, educ = 0, \dots) = .123$$

Effects of marriage – for men:	+.213	
for women:	-.088	( = +.213 – .301)

Effects of gender – for singles:	-.110	
for marrieds:	-.411	( = -.110 – .301)



### 3. Multiple Dummies for Multiple Dimensions – Interaction Version

Combining coefficients – suppose single male (*sing.m*) is our baseline category

A. suppose we are using *marr.m*, *sing.f*, and *marr.f*

Effect of marriage for males is simply coefficient for *marr.m* ( $\hat{\beta}_{marr.m}$ )

... which has standard error:

$$\sqrt{var(\hat{\beta}_{marr.m})}$$

Effect of marriage for females is difference of coefficients: ( $\hat{\beta}_{marr.f} - \hat{\beta}_{sing.f}$ )

... which has standard error:

$$\sqrt{var(\hat{\beta}_{marr.f}) + var(\hat{\beta}_{sing.f}) - 2 \cdot cov(\hat{\beta}_{marr.f}, \hat{\beta}_{sing.f})}$$

### 3. Multiple Dummies for Multiple Dimensions – Interaction Version

Combining coefficients – suppose single male (*sing.m*) is our baseline category

B. suppose we are using *married*, *female*, and *married\*female*

Effect of marriage for males is simply coefficient for *married* ( $\hat{\beta}_{\text{married}}$ )

... which has standard error:

$$\sqrt{\text{var}(\hat{\beta}_{\text{married}})}$$

Effect of marriage for females is sum of coefficients: ( $\hat{\beta}_{\text{married}} + \hat{\beta}_{\text{married*female}}$ )

... which has standard error:

$$\sqrt{\text{var}(\hat{\beta}_{\text{married}}) + \text{var}(\hat{\beta}_{\text{married*female}}) + 2 \cdot \text{cov}(\hat{\beta}_{\text{married}}, \hat{\beta}_{\text{married*female}})}$$

### 3. Multiple Dummies for Multiple Dimensions – Interaction Version

Shortcut for finding or verifying calculations on previous page: change baseline group!

In A., replace *sing.f* with *sing.m*;

then coefficient for *marr.f* is effect of marriage for women

In B., replace *female* with *male* and *married\*female* with *married\*male*;

then coefficient for *married* is effect of marriage for women

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Readings: Kam & Franzese, p. 27–29, 54–56, 67–68, 85–86

#### 4. Gender & Party and Social Welfare Opinions

Data: *socialwelfare.dta*

Script: *Lecture 26 socialwelfare.R*

N = 1,077 respondents in 2004 American National Election Study

Dependent variable : *socwel* (index described in fn. 16; range [0,1])

The dependent variable is compiled from support for services and spending; government provision of jobs and a standard of living; and support for federal spending on welfare programs, social security, public schools, child care, and assistance to the poor, rescaled to range from zero (least supportive) to one (most supportive).

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#### 4. Gender & Party and Social Welfare Opinions

Data: *socialwelfare.dta*

Script: *Lecture 26 socialwelfare.R*

N = 1,077 respondents in 2004 American National Election Study

Dependent variable : *socwel* (a.k.a *Opinion*)

Independent variables: *Female* {0,1}  
*Republican* {0,1}

$$Opinion = \beta_0 + \beta_F Female + \beta_R Republican + \beta_{FR} Female \times Republican \quad (16)$$

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#### 4. Gender & Party and Social Welfare Opinions

TABLE 3. OLS Regression Results, *Support for Social Welfare*

	Coefficient (standard error) <i>p</i> -Value
<i>Female</i>	–0.0031 (0.0144) <i>0.828</i>
<i>Republican</i>	–0.2205 (0.0155)
<i>Female × Republican</i>	0.0837 (0.0214) <i>0.000</i>
Intercept	0.7451 (0.0110) <i>0.000</i>
<i>N (df)</i>	1,077 (1,073)
Adjusted $R^2$	0.223
$P > F$	0.000

*Note:* Cell entries are the estimated coefficient, with standard error in parentheses, and two-sided *p*-level (probability  $|T| > t$ ) referring to the null hypothesis that  $\beta = 0$  in italics.

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#### 4. Gender & Party and Social Welfare Opinions

Marginal Effects – easiest to start with predicted values, then calculate deltas:

TABLE 4. Predicted Support for Social Welfare

	Democrats (Republican = 0)	Republicans (Republican = 1)
Males (Female = 0)	0.745	0.525
Females (Female = 1)	0.742	0.605

Compare down columns:

	$\partial \hat{y} / \partial F$	<i>s.e.</i> ( $\partial \hat{y} / \partial F$ )	<i>t</i> -Statistic
Republican = 0	-0.003	0.0144	-0.218
Republican = 1	0.081	0.0158	5.109

Compare across rows:

	$\partial \hat{y} / \partial R$	<i>s.e.</i> ( $\partial \hat{y} / \partial R$ )	<i>t</i> -Statistic
Female = 0	-0.220	0.0155	-14.18
Female = 1	-0.137	0.0147	-9.33

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#### 4. Gender & Party and Social Welfare Opinions

Model: 
$$y = \gamma_0 + \beta_x x + \beta_z z + \beta_{xz} xz + \varepsilon \quad (2)$$

Marginal Effects: 
$$\partial y / \partial x = \beta_x + \beta_{xz} z \quad (9)$$

$$\partial y / \partial z = \beta_z + \beta_{xz} x \quad (10)$$

Variances 
$$V(\partial \hat{y} / \partial x) = V(\hat{\beta}_x) + z^2 V(\hat{\beta}_{xz}) + 2zC(\hat{\beta}_x, \hat{\beta}_{xz}) \quad (26)$$

$$V(\partial \hat{y} / \partial z) = V(\hat{\beta}_z) + x^2 V(\hat{\beta}_{xz}) + 2xC(\hat{\beta}_z, \hat{\beta}_{xz})$$

Standard Errors: 
$$s.e.(\partial \hat{y} / \partial x) = \sqrt{V(\partial \hat{y} / \partial x)}$$
$$s.e.(\partial \hat{y} / \partial z) = \sqrt{V(\partial \hat{y} / \partial z)}$$

Hypothesis Tests: 
$$t = \frac{\partial \hat{y} / \partial x}{s.e.(\partial \hat{y} / \partial x)}$$

$$t = \frac{\partial \hat{y} / \partial z}{s.e.(\partial \hat{y} / \partial z)}$$



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#### 4. Gender & Party and Social Welfare Opinions

Model:  $SW = .7451 - .0031 F - .2205 R + .0837 FR$

Marginal Effects:  $\partial SW / \partial F = - .0031 + .0837 R$       = - .0031 among Democrats  
= + .0806 among Republicans

$\partial SW / \partial R = - .2205 + .0837 F$       = - .2205 among males  
= - .1368 among females

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#### 4. Gender & Party and Social Welfare Opinions

TABLE 26. OLS Regression Results, *Support for Social Welfare*, Pooled and Split Samples

	Pooled Sample Coefficient (standard error) <i>p</i> -Value	Males Only Coefficient (standard error) <i>p</i> -Value	Females Only Coefficient (standard error) <i>p</i> -Value
<i>Female</i>	−0.0031 (0.0144) <i>0.828</i>	—	—
<i>Republican</i>	−0.2205 (0.0155) <i>0.000</i>	−0.2205 (0.0154) <i>0.000</i>	−0.1368 (0.0148) <i>0.000</i>
<i>Female × Republican</i>	0.0837 (0.0214) <i>0.000</i>	—	—
Intercept	0.7451 (0.0110) <i>0.000</i>	0.7451 (0.0109) <i>0.000</i>	0.7420 (0.0094) <i>0.000</i>
<i>N (df)</i>	1,077 (1,073)	498 (496)	579 (577)
Adjusted <i>R</i> <sup>2</sup>	0.223	0.290	0.128
<i>P &gt; F</i>	0.000	0.000	0.000

*Note:* Cell entries are the estimated coefficient, with standard error in parentheses, and two-sided *p*-level (probability  $|T| > t$ ) referring to the null hypothesis that  $\beta = 0$  in italics.

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#### 4. Gender & Party and Social Welfare Opinions

TABLE 16. Estimated Variance-Covariance Matrix of Coefficient Estimates,  
 Predicting *Support for Social Welfare*

	<i>Female</i>	<i>Republican</i>	<i>Female × Republican</i>	Intercept
<i>Female</i>	0.00021			
<i>Republican</i>	0.00012	0.00024		
<i>Female × Republican</i>	−0.00021	−0.00024	0.00046	
Intercept	−0.00012	−0.00012	0.00012	0.00012

Variances:  $V(\partial\hat{y}/\partial F) = V(\hat{\beta}_F) + R^2 \cdot V(\hat{\beta}_{FR}) + 2R \cdot C(\hat{\beta}_F, \hat{\beta}_{FR})$   
 $= .00021 + R^2 \cdot (.00046) + 2R \cdot (-.00021)$   
 $= .00021$  if  $R = 0$   
 $= .00025$  if  $R = 1$

$V(\partial\hat{y}/\partial R) = V(\hat{\beta}_R) + F^2 \cdot V(\hat{\beta}_{FR}) + 2F \cdot C(\hat{\beta}_R, \hat{\beta}_{FR})$   
 $= .00024 + R^2 \cdot (.00046) + 2R \cdot (-.00024)$   
 $= .00024$  if  $R = 0$   
 $= .00022$  if  $R = 1$

Standard Errors:

$\sqrt{.00021} = .0145$   
 $\sqrt{.00025} = .0158$

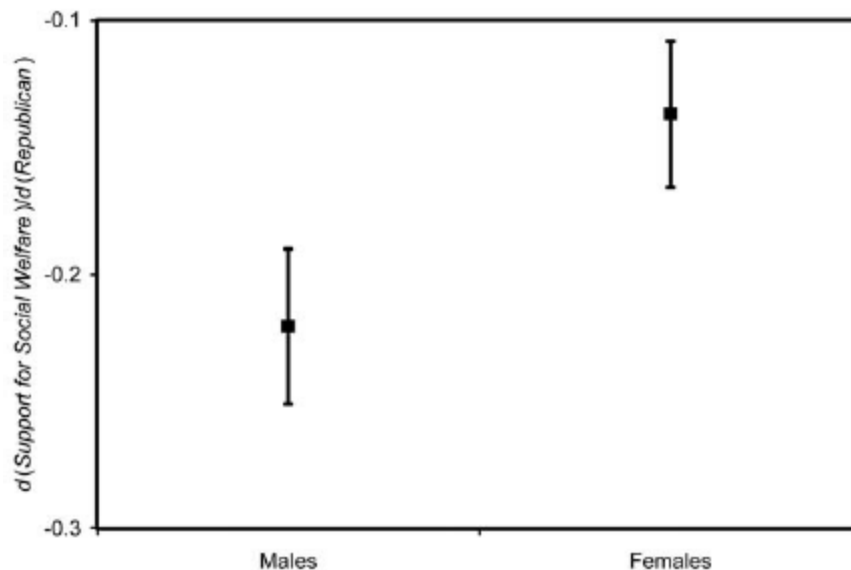
$\sqrt{.00024} = .0155$   
 $\sqrt{.00022} = .0148$

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#### 4. Gender & Party and Social Welfare Opinions

TABLE 18. Hypothesis Tests of whether *Republican* Affects *Support for Social Welfare*

	$\partial \hat{y} / \partial R$	<i>s.e.</i> ( $\partial \hat{y} / \partial R$ )	<i>t</i> -Statistic	One-Tailed <i>p</i> -Value $H_0: \beta_R + \beta_{FR} \text{Female} \leq 0$	One-Tailed <i>p</i> -Value $H_0: \beta_R + \beta_{FR} \text{Female} \geq 0$	95% Confidence Interval
<i>Female</i> = 0	-0.220	0.0155	-14.18	0.999	0.000	[-0.251, -0.190]
<i>Female</i> = 1	-0.137	0.0147	-9.33	0.999	0.000	[-0.166, -0.108]



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#### 4. Gender & Party and Social Welfare Opinions

TABLE 17. Hypothesis Tests of whether *Female Affects Support for Social Welfare*

	$\partial \hat{y} / \partial F$	<i>s.e.</i> ( $\partial \hat{y} / \partial F$ )	<i>t</i> -Statistic	One-Tailed <i>p</i> -Value $H_0: \beta_F + \beta_{FR} \text{Republican} \leq 0$	One-Tailed <i>p</i> -Value $H_0: \beta_F + \beta_{FR} \text{Republican} \geq 0$	95% Confidence Interval
<i>Republican</i> = 0	-0.003	0.0144	-0.218	0.586	0.414	[-0.031, 0.025]
<i>Republican</i> = 1	0.081	0.0158	5.109	0.000	0.999	[0.050, 0.111]

