

Advanced Grad Stats

Interaction Effects

Evidence from 12 Latin American Countries

Lab 7

Diego F. Leal

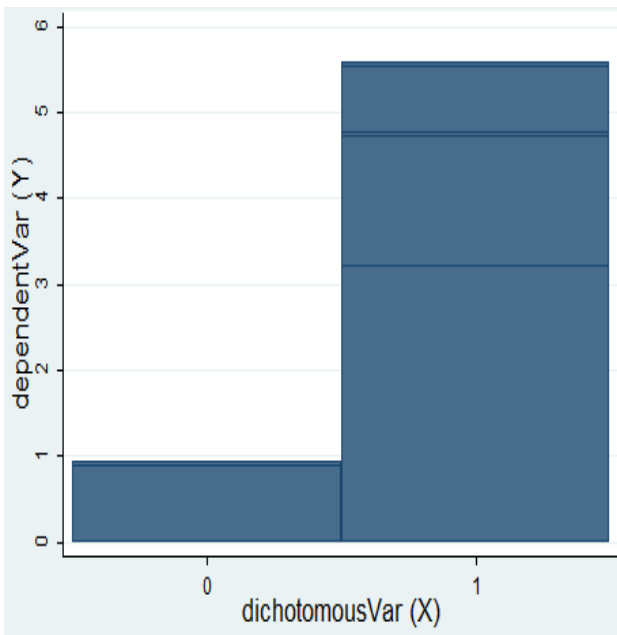
www.diegoleal.info

A Dichotomous/Categorical Independent Variable

```
. reg dependentVar i1.dichotomousVar
```

Source	SS	df	MS	Number of obs = 1000		
Model	45.6895343	1	45.6895343	F(1, 998) =	71.59	
Residual	636.946318	998	.638222764	Prob > F =	0.0000	
Total	682.635852	999	.683319172	R-squared =	0.0669	
				Adj R-squared =	0.0660	
				Root MSE =	.79889	

dependentVar	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
1.dichotomous	2.263343	.2675026	8.46	0.000	1.73841	2.788275
_cons	.7485708	.2662962	2.81	0.005	.2260062	1.271135



Here a “unit increase” represents switching from one category to another. The coefficient is the average difference in Y between the category for which $X = 0$ (the reference group) and the category for which $X = 1$ (the comparison group)

A Dichotomous/Categorical Independent Variable

```
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F( 1, 998) = 71.59
Prob > F = 0.0000
R-squared = 0.0669
Adj R-squared = 0.0660
Root MSE = .79889
```

```
Summary statistics: mean
by categories of: dichotomousVar
```

dichotomousVar	depend~r
0	.7485708
1	3.011913
Total	2.991543

dependentVar	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
1.dichotomou~r	2.263343	.2675026	8.46	0.000	1.73841 2.788275
_cons	.7485708	.2662962	2.81	0.005	.2260062 1.271135

Therefore, compared to the reference group, we would expect people in the comparison group to score 2.26 units higher on the dependent variable, on average.

Additive Model

This is our initial model:

$$E(I) = \alpha + \beta_1 \text{ Education} + \beta_2 \text{ Urban} + \epsilon$$

Where,

I = respondent's wealth

α = slope

β_1 = Respondent's education \rightarrow Main effect 1

β_2 = Respondent's area of residence (urban = 1) \rightarrow Main effect 2

ϵ = error

Additive Model: Stata Output, Evidence from Latin America

```
. xi: svy: reg wealthIndex i.urban education
i.urban          _Iurban_0-1      (naturally coded; _Iurban_0 omitted)
(running regress on estimation sample)
```

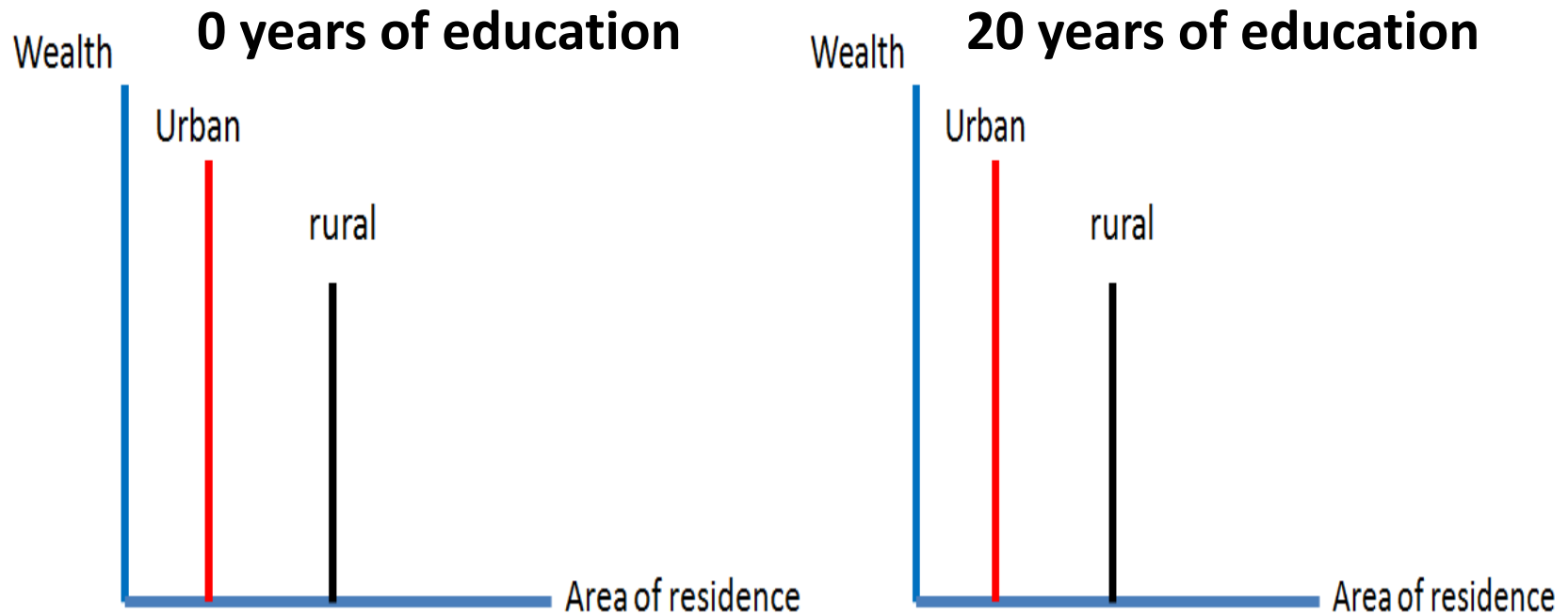
Survey: Linear regression

```
Number of strata =          6          Number of obs      =      8972
Number of PSUs   =         125        Population size     =      8972
                                           Design df         =       119
                                           F(  2,    118)    =     343.34
                                           Prob > F          =      0.0000
                                           R-squared         =      0.3173
```

wealthIndex	Linearized					[95% Conf. Interval]	
	Coef.	Std. Err.	t	P> t			
_Iurban_1	.2116932	.0138543	15.28	0.000	.1842603	.2391261	
education	.029278	.0012547	23.33	0.000	.0267935	.0317624	
_cons	-.1211798	.0121518	-9.97	0.000	-.1452415	-.0971181	

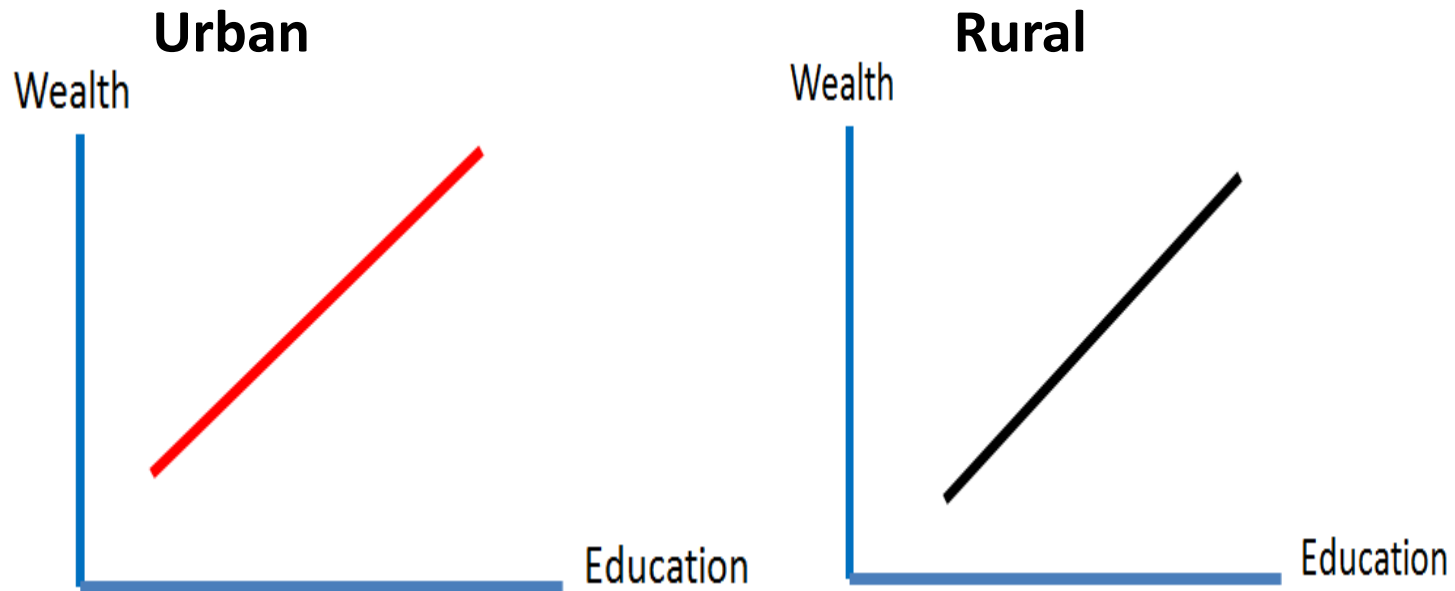
Model Assumptions

Model assumes the difference in wealth between rural and urban residents is constant across **all** levels of education



Model Assumptions

Model assumes the slope of education is the same for urban and rural residents.



Non-Additive Way of Thinking

Slope line for education could depend on the group (rural vs. urban residents) one is analyzing

Categorical variable as the moderator

Difference in income between urban and rural residents could depend on the amount of education people have

Continuous variable as the moderator

Definition

An interaction exists when the relationship between X and Y changes (in magnitude or direction) when examined at different levels of a third variable

Non-Additive Model

This is our initial model:

$$E(I) = \alpha + \beta_1 \text{ Education} + \beta_2 \text{ Urban} + \beta_3 \text{ Urban} * \text{Education} + \epsilon$$

Where,

I = respondent's wealth

α = slope

β_1 = Respondent's education \rightarrow conditional effect 1

β_2 = Respondent's area of residence (urban = 1) \rightarrow conditional effect 2

β_3 = Respondent's area of residence (urban = 1) * respondent's education \rightarrow cross-product term

ϵ = error

Something to Consider First

```
. xi: svy: reg wealthIndex education if urban == 0
(running regress on estimation sample)
```

Survey: Linear regression

```
Number of strata =      5      Number of obs   =   2349
Number of PSUs  =     70      Population size =   2349
                                   Design df       =     65
                                   F( 1, 65)         =   61.47
                                   Prob > F          =   0.0000
                                   R-squared          =   0.0737
```

wealthIndex	Linearized				
	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
education	.0109204	.0013929	7.84	0.000	.0081386 .0137023
_cons	.0030168	.0086096	0.35	0.727	-.0141777 .0202113

Rural residents' education effect (slope)
= **0.0109**

Rural residents' slope is not as steeper
as urban residents' slope

```
. xi: svy: reg wealthIndex education if urban == 1
(running regress on estimation sample)
```

Survey: Linear regression

```
Number of strata =      6      Number of obs   =   6623
Number of PSUs  =    124      Population size =   6623
                                   Design df       =    118
                                   F( 1, 118)        =  640.78
                                   Prob > F          =   0.0000
                                   R-squared          =   0.2102
```

wealthIndex	Linearized				
	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
education	.0349324	.00138	25.31	0.000	.0321996 .0376651
_cons	.0336815	.0169335	1.99	0.049	.0001485 .0672144

Urban residents' education effect
(slope) = **0.0349**

Wealth returns to education are
stronger for urban residents

Interpreting the Results

```
. xi: svy: reg wealthIndex i.urban education urban_education
i.urban      _Iurban_0-1      (naturally coded; _Iurban_0 omitted)
(running regress on estimation sample)
```

Survey: Linear regression

```
Number of strata   =           6      Number of obs       =      8972
Number of PSUs    =          125     Population size     =      8972
                                           Design df          =        119
                                           F(   3,    117)    =      333.75
                                           Prob > F           =      0.0000
                                           R-squared          =      0.3352
```

wealthIndex	Linearized		t	P> t	[95% Conf. Interval]	
	Coef.	Std. Err.				
_Iurban_1	.0306647	.0176263	1.74	0.084	-.0042373	.0655666
education	.0109204	.0014191	7.70	0.000	.0081105	.0137304
urban_education	.0240119	.0019381	12.39	0.000	.0201743	.0278496
_cons	.0030168	.0085909	0.35	0.726	-.0139941	.0200277

- **Education:** effect of education for rural residents (reference group)
- **_Iurban_1:** predicted income gap between rural and urban residents when education is zero in both groups
- **Urban_education:** part of education effect for urban residents

The Math

$$E(Y) = \alpha + \beta_1 (\text{Education}) + \beta_2 (\text{Urban}) + \beta_3 (\text{Urban} * \text{Education}) + \epsilon$$
$$E(I) = .003 + .01 (\text{Education}) + .03 (\text{Urban}) + .02 (\text{Urban} * \text{Education})$$

Substitute 1, 2 & 3 as possible values for education and the interaction; and 0 for urban

$$E(I) = .003 + .01 (1) + .03 (0) + .02 (0) = 0.4$$

$$E(I) = .003 + .01 (2) + .03 (0) + .02 (0) = 0.5$$

$$E(I) = .003 + .01 (3) + .03 (0) + .02 (0) = 0.6$$

The difference between predicted values is (e.g. $0.5 - 0.4$) = .01

So, .01 is the change in slope for rural residents → increment in wealth for rural residents for one unit increase in education

The Math

$$E(Y) = \alpha + \beta_1 (\text{Education}) + \beta_2 (\text{Urban}) + \beta_3 (\text{Urban} * \text{Education}) + \epsilon$$

$$E(I) = .003 + .01 (\text{Education}) + .03 (\text{Urban}) + .02 (\text{Urban} * \text{Education})$$

Substitute 1, 2 & 3 for education; 1 urban; and 1, 2 & 3 for the interaction

$$E(I) = .003 + .01 (1) + .03 (1) + .02 (1) = 0.9$$

$$E(I) = .003 + .01 (2) + .03 (1) + .02 (2) = 1.2$$

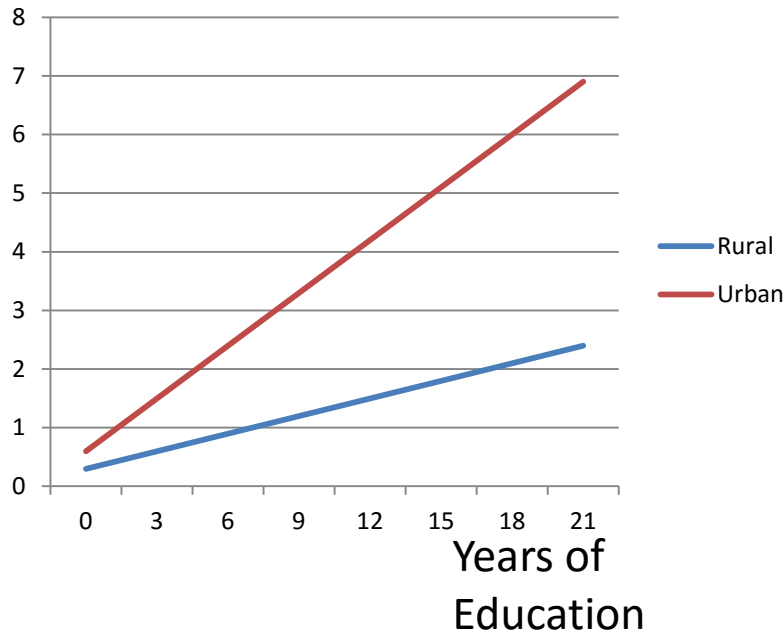
$$E(I) = .003 + .01 (3) + .03 (1) + .02 (3) = 1.5$$

The difference between predicted values is (e.g. $1.2 - 0.9$) = 0.3

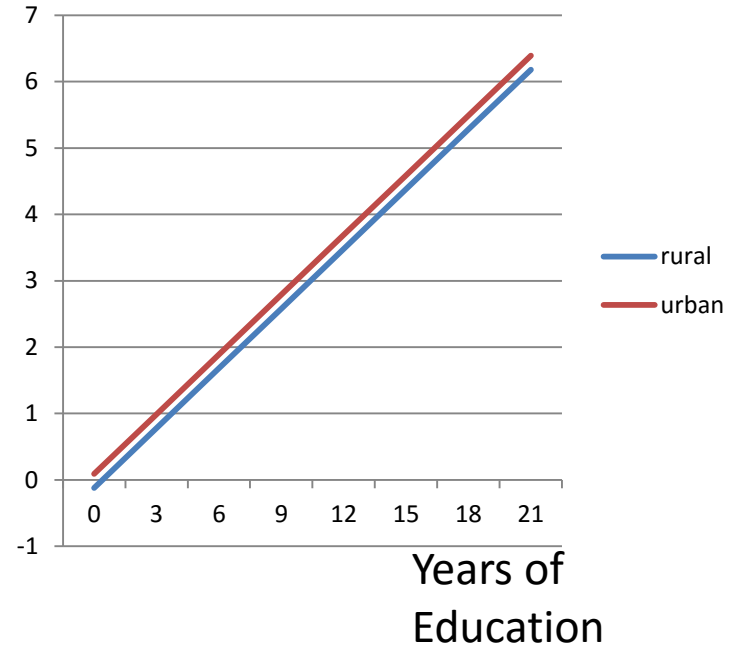
So, .03 is the change in slope for urban residents → increment in wealth for urban residents for one unit increase in education

A Graphic Result: Non-Additive vs. Additive

Predicted Wealth



Predicted Wealth



Conclusion

When people have more education, the urban vs. rural gap in wealth is large.

If you live in a rural area, it does not pay that much to get more education (when compared to the situation in urban areas).

Given the size of the coefficients and their signs, a positive coefficient for the interaction means both slopes for area of residence are positive.