Multilevel Workshop

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1 Introduction

Workshop materials for a workshop on Multilevel Modeling

2 Two Level Cross Sectional; And Three Level Longitudinal Models

2.1 Cross Sectional Model

2.1.1 Get Data

use "simulated_multilevel_data.dta", clear

2.1.2 The Equation

 $\text{outcome}_{ij} = \beta_0 + \beta_1 \text{parental warmth} + \beta_2 \text{physical punishment} + \beta_3 \text{time} +$

$$\beta_4 {\rm identity}_2 + \beta_5 {\rm intervention} + \beta_6 HDI +$$

$$u_{0j} + u_{1j} \times \text{parental warmth} + e_{ij}$$

2.1.3 Descriptive Statistics

summarize // descriptive statistics

Variable	Obs	Mean	Std. dev.	Min	Max
country	3,000	15.5	8.656884	1	30
HDI	3,000	64.76667	17.24562	33	87
family	3,000	50.5	28.87088	1	100
id	0				
identity	3,000	.4976667	.5000779	0	1

	-+					
intervention	· 	3,000	.4843333	.4998378	0	1
physical_p~t	1	3,000	2.478667	1.360942	0	5
warmth	1	3,000	3.521667	1.888399	0	7
outcome	1	3,000	52.43327	6.530996	29.60798	74.83553

2.1.4 Spaghetti Plot

```
spagplot outcome warmth, id(country) scheme(stcolor)
graph export spagplot1.png, width(1000) replace
```

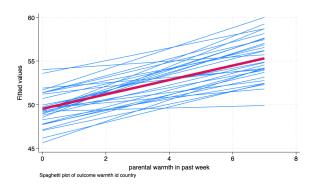


Figure 2.1: Spaghetti Plot of Outcome by Warmth by Country

2.1.5 Unconditional Model

2.1.5.1 Model

```
mixed outcome || country: // unconditional model
```

Performing EM optimization ...

Performing gradient-based optimization: Iteration 0: Log likelihood = -9802.8371 Iteration 1: Log likelihood = -9802.8371

Computing standard errors ...

Mixed-effects ML regression Group variable: country	Number of obs = 3,000 Number of groups = 30 Obs per group:
	min = 100 avg = 100.0 max = 100
Log likelihood = -9802.8371	Wald chi2(0) = . Prob > chi2 = .
outcome Coefficient Std. err. z P> z	[95% conf. interval]
_cons 52.43327 .3451217 151.93 0.000	51.75685 53.1097
Random-effects parameters Estimate Std. err.	[95% conf. interval]
country: Identity var(_cons) 3.178658 .9226737	1.799552 5.614658
var(Residual) 39.46106 1.024013	
LR test vs. linear model: chibar2(01) = 166.31	Prob >= chibar2 = 0.0000
2.1.5.2 ICC	
estat icc	
Intraclass correlation	

Level | ICC Std. err. [95% conf. interval]

country | .0745469 .0201254 .0434963 .1248696

2.1.6 Conditional Model

```
mixed outcome warmth physical_punishment identity i.intervention HDI || country: warmth // me est store crosssectional // store estimates
```

Performing EM optimization ...

 ${\tt Performing\ gradient-based\ optimization:}$

Iteration 0: Log likelihood = -9626.6279
Iteration 1: Log likelihood = -9626.607
Iteration 2: Log likelihood = -9626.607

Computing standard errors ...

Mixed-effects ML regression

Group variable: country

Number of obs = 3,000

Number of groups = 30

Obs per group:

min = 100 avg = 100.0

max = 100Wald chi2(5) = 334.14

Prob > chi2 = 0.0000

Log likelihood = -9626.607

outcome | Coefficient Std. err. z P>|z| [95% conf. interval]

 warmth | .8345368
 .0637213
 13.10
 0.000
 .7096453
 .9594282

 physical_punishment | -.9916657
 .0797906
 -12.43
 0.000
 -1.148052
 -.8352791

 identity | -.3004767
 .2170295
 -1.38
 0.166
 -.7258466
 .1248933

 1.intervention | .6396427
 .2174519
 2.94
 0.003
 .2134448
 1.065841

HDI | -.003228 .0199257 -0.16 0.871 -.0422817 .0358256 _cons | 51.99991 1.371257 37.92 0.000 49.3123 54.68753

Random-effects parameters		Estimate	Std. err.	[95% conf.	interval]
country: Independent	 				
var(warmth)		.0227504	.0257784	.0024689	.2096436
var(_cons)		2.963975	.9737647	1.556777	5.643163

var(Residual) | 34.97499 .9097109 33.23668 36.80422

LR test vs. linear model: chi2(2) = 205.74 Prob > chi2 = 0.0000

Note: LR test is conservative and provided only for reference.

2.2 Longitudinal Model

2.2.1 Get Data

use "simulated_multilevel_longitudinal_data.dta", clear

2.2.2 The Equation

 $\text{outcome}_{ij} = \beta_0 + \beta_1 \text{parental warmth} + \beta_2 \text{physical punishment} + \beta_3 \text{time} +$

$$\beta_4$$
identity₂ + β_5 intervention + $\beta_5 HDI$ +

$$u_{0j} + u_{1j} \times \text{parental warmth} +$$

$$v_{0i} + v_{1i} \times t + e_{ij}$$

2.2.3 Descriptive Statistics

summarize // descriptive statistics

Variable	0bs	Mean	Std. dev.	Min	Max
country	9,000	15.5	8.655922	1	30
HDI	9,000	64.76667	17.2437	33	87
family $ $	9,000	50.5	28.86767	1	100
id	0				

identity		9,000	.4976667	.5000223	0	1
intervention		9,000	. 4843333	.4997823	0	1
t	1	9,000	2	.8165419	1	3
physical_p~t	1	9,000	2.485333	1.373639	0	5
warmth	1	9,000	3.514222	1.8839	0	7
outcome	1	9,000	53.37768	6.572285	29.60798	79.02199

2.2.4 Alternate Plot

```
encode id, generate(idNUMERIC) // numeric version of id

* spagplot outcome t if idNUMERIC <= 10, id(idNUMERIC) scheme(stcolor)

twoway (lfit outcome t) (scatter outcome t) if idNUMERIC <= 10, by(idNUMERIC) scheme(stcolor)

graph export spagplot2.png, width(1000) replace</pre>
```

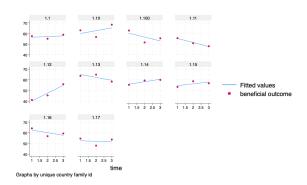


Figure 2.2: Alternate Plot of Outcome by Time by Individual; First 10 Observations

2.2.5 Unconditional Model

2.2.5.1 Model

```
mixed outcome || country: || id: // unconditional model
```

2.2.5.2 ICC

estat icc

Intraclass correlation

Level	l ICC			_
country	.0748336 .3462837	.0190847	.0450028 .3134867	.1219141

2.2.6 Conditional Model

```
mixed outcome t warmth physical_punishment i.identity i.intervention HDI || country: warmth est store longitudinal // store estimates
```

Performing EM optimization ...

Performing gradient-based optimization:

Iteration 0: Log likelihood = -28523.49
Iteration 1: Log likelihood = -28499.987
Iteration 2: Log likelihood = -28499.739
Iteration 3: Log likelihood = -28499.604
Iteration 4: Log likelihood = -28499.603

Computing standard errors \dots

Mixed-effects ML regression

Number of obs = 9,000

Grouping information

	No. of	Obser	vations per	group
Group variable	groups	Minimum	Average	Maximum
country	30	300	300.0	300

id	3,000	3	3.0	3

Wald chi2(6) = 1096.15Log likelihood = -28499.603 Prob > chi2 = 0.0000

outcome	Coefficient	Std. err.	z	P> z	[95% conf.	interval]
t warmth	.943864 .9134959	.0658716	14.33 21.56	0.000	.814758 .830446	1.07297
physical_punishment 1.identity 1.intervention	1276926	.0497622 .1515835 .1519095	-20.25 -0.84 5.65	0.000 0.400 0.000	-1.105429 4247908 .5612596	9103647 .1694057 1.156734
HDI _cons	0005657 50.46724	.0196437	-0.03 37.71	0.977	0390666 47.84418	.0379352

var(Residual) | 26.02733 .4753701 25.11211 26.97592

Prob > chi2 = 0.0000

Note: LR test is conservative and provided only for reference.

LR test vs. linear model: chi2(4) = 1247.03

2.3 Nice Table of Results

```
etable, estimates(crosssectional longitudinal) ///
showstars showstarsnote /// show stars and note
column(estimate) // column is modelname
```

	crosssectional	longitudinal
parental warmth in past week	0.835 **	0.913 **
	(0.064)	(0.042)
physical punishment in past week	-0.992 **	-1.008 **
	(0.080)	(0.050)
hypothetical identity group variable	-0.300	
	(0.217)	
recieved intervention		
1	0.640 **	0.859 **
	(0.217)	(0.152)
Human Development Index	-0.003	-0.001
	(0.020)	(0.020)
time		0.944 **
		(0.066)
hypothetical identity group variable		
1		-0.128
		(0.152)
Intercept	52.000 **	50.467 **
	(1.371)	(1.338)
var(warmth)	0.023	0.011
	(0.026)	(0.013)
<pre>var(_cons)</pre>	2.964	3.167
	(0.974)	(0.915)
var(e)	34.975	26.027
	(0.910)	(0.475)
<pre>var(_cons)</pre>		8.387
		(0.472)
var(t)		0.000
		(0.000)

^{**} p<.01, * p<.05

2.4 QUESTIONS???

Number of observations

3000

9000

3 Cross-Classified Models

3.1 Introduction

A two level multilevel model imagines that *Level 1* units are nested in *Level 2* units. A three level multilevel model imagines that *Level 1* units are nested in *Level 2* units, which are in turn nested in *Level 3*.

A cross-classified model imagines that the nesting is not hierarchical, but rather that there are two sets of clusters or nestings in which individuals may be nested.

3.2 Get Data

use "simulated_multilevel_longitudinal_data.dta", clear

3.3 Cross Classified Model

We can treat these random effects as being cross classified.

This might be useful if we had data where individuals lived in different countries at different times.

However, because id is in fact nested inside country, in this case, estimating the random effects as cross classified will be more time consuming, but will give us equivalent results to a three level model.

3.3.1 Standard (Less Computationally Efficient) Syntax

The below syntax will take a very long time to run with the full sample, and thus we have commented it out.

```
* mixed outcome t warmth physical_punishment || _all: R.country || _all: R.id
* est store crossed1
```

The documentation notes that we can use a *much* more computationally efficient version of the above command, which is what we do in these notes. The user can verify that both versions of the command will produce equivalent results.

In fact, at the end of handout we verify the similarity of both sets of syntax using a random sample.

3.3.2 Cross Classified With Computationally Efficient Syntax

```
mixed outcome t warmth physical_punishment || _all: R.country || id:
est store crossed2 // store crossed effects result
```

Performing EM optimization ...

Performing gradient-based optimization:

Iteration 0: Log likelihood = -28516.314
Iteration 1: Log likelihood = -28516.277
Iteration 2: Log likelihood = -28516.277

Computing standard errors ...

Mixed-effects ML regression

Number of obs = 9,000

Grouping information

Group variable	No. of groups	Obser Minimum	vations per Average	group Maximum
_all id	1 3,000	9,000	9,000.0	9,000

Wald chi2(3) = 1168.69Prob > chi2 = 0.0000

Log likelihood = -28516.277

outcome	•	Coefficient					interval]
t		.9434605	.065866	14.32	0.000	.8143654	1.072556
warmth		.9053924	.0380439	23.80	0.000	.8308277	.9799572
physical_punishment		-1.014385	.0499354	-20.31	0.000	-1.112257	916514
_cons		50.8301	.4123007	123.28	0.000	50.022	51.63819

Random-effects par					_
_all: Identity	country)	3.429974	.930313	2.015668	5.836634
id: Identity	r(_cons)	8.608872	. 4757699	7.725107	9.59374
	·	26.02862		25.11363	26.97695
LR test vs. linear model: chi2(2) = 1260.84			:	Prob > chi2	= 0.0000

Note: LR test is conservative and provided only for reference.

3.4 Three Level Model

```
mixed outcome t warmth physical_punishment || country: || id: // 3 level w/ random intercept est store threelevel // store random intercept model
```

Performing EM optimization ...

Performing gradient-based optimization:

Iteration 0: Log likelihood = -28516.314
Iteration 1: Log likelihood = -28516.277
Iteration 2: Log likelihood = -28516.277

Computing standard errors ...

 ${\tt Mixed-effects}\ {\tt ML}\ {\tt regression}$

Number of obs = 9,000

Grouping information

Group variable	No. of groups	Obser Minimum	rvations per Average	group Maximum
country id	30 3,000	300 3	300.0	300

Log likelihood = -28516.277

Wald chi2(3) = 1168.69Prob > chi2 = 0.0000

	Coefficient			• •	[95% conf.	interval]
t warmth	.9434605 .9053924	.065866 .0380439	14.32 23.80	0.000	.8143654 .8308277	1.072556 .9799572
physical_punishment cons	-1.014385 50.8301	.0499354	-20.31 123.28	0.000	-1.112257 50.022	916514 51.63819

LR test vs. linear model: chi2(2) = 1260.84

Prob > chi2 = 0.0000

Note: LR test is conservative and provided only for reference.

3.5 Nice Table of Results of Three Level and Cross Classified Model

```
etable, estimates(threelevel crossed2), ///
showstars showstarsnote /// show stars and note
column(estimate) // column is modelname

invalid 'showstars'
r(198);
```

3.6 Verification of Syntax Equivalence for Cross Classified Model

```
keep if family <= 5 // random sample of families

quietly mixed outcome t warmth physical_punishment || _all: R.country || _all: R.id

est store crossed1A // less efficient syntax

quietly mixed outcome t warmth physical_punishment || _all: R.country || id:

est store crossed2A // more efficient syntax

etable, estimates(crossed1A crossed2A) ///

showstars showstarsnote /// show stars and note

column(estimate) // column is modelname
```

(8,550 observations deleted)

-----crossed1A crossed2A

time	0.745	**	0.745	**
	(0.281)		(0.281)	
parental warmth in past week	0.871	**	0.871	**
	(0.160)		(0.160)	
physical punishment in past week	-1.262	**	-1.262	**
	(0.206)		(0.206)	
Intercept	51.755	**	51.755	**
	(1.009)		(1.009)	
<pre>var(R_country)</pre>	2.245		2.245	
	(1.319)		(1.319)	
<pre>var(R_id)</pre>	5.425			
	(1.843)			
var(e)	23.638		23.638	
	(1.933)		(1.933)	
var(_cons)			5.425	
			(1.843)	
Number of observations	450		450	

** p<.01, * p<.05

3.7 QUESTIONS???