Multilevel Multilingual

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1 Multilevel Models in Stata, R and Julia

1.1 Introduction

Below, I describe the use of Stata, R, and Julia to estimate multilevel models. Because this document is built by Quarto, I describe calling these programs from within a Quarto environment. However, each piece of software could be used individually and separately.

1.2 The Data

The examples below use the simulated_multilevel_data.dta file from *Multilevel Thinking*. Here is a direct link to download the data.

Table 1.1: Sample of Simulated Multilevel Data

countr	y HDI	family	id	group	physical_	_punis kraren t	thoutcom
1	69	1	1.1	2	2	3	59.18
1	69	2	1.2	2	4	0	61.54
1	69	3	1.3	1	4	4	51.87
1	69	4	1.4	2	0	6	51.71
1	69	5	1.5	2	3	2	55.88
1	69	6	1.6	1	5	3	60.78

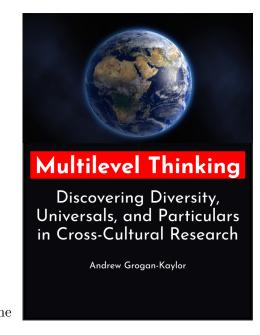


Figure 1.1: Book Cover For Multi-level Thinking

1.3 The Equation

 $\text{outcome}_{ij} = \beta_0 + \beta_1 \text{warmth}_{ij} + \beta_2 \text{physical punishment}_{ij} + \beta_3 \text{group}_{ij} + \beta_4 \text{HDI}_{ij} + u_{0j} + u_{1j} \times \text{warmth}_{ij} + e_{ij} \tag{1.1}$

1.4 Setup

1.4.1 Stata

I need to use the library Statamarkdown to call Stata, or I could run Stata on its own

```
library(Statamarkdown)
```

1.4.2 R

In R, I use the library lme4 to run multilevel models.

```
library(lme4)
```

1.4.3 Julia

I need to call Julia from R.

```
library(JuliaCall)
julia_setup(JULIA_HOME = "/Applications/Julia-1.8.app/Contents/Resources/julia/bin")
```

1.5 Get Data & Run Models

To explain statistical syntax for each software, I consider the more general case of a multilevel model with dependent variable y, independent variables x and z, clustering variable group, and a random slope for x. i is the index for the person, while j is the index for the group.

$$y = \beta_0 + \beta_1 x_{ij} + \beta_2 z_{ij} + u_{0j} + u_{1j} \times x_{ij} + e_{ij}$$
 (1.2)

1.5.1 Stata

In Stata mixed, the syntax for a multilevel model of the form described in Equation ?? is:

```
mixed y x || group: x
```

1.5.1.1 Get The Data



Tip For Running Stata From Quarto

Because I am running Stata from inside a Quarto document, and running Stata in multiple chunks, I need to use the collectcode=TRUE option in the first Stata chunk. i.e. my Quarto chunk needs to begin with "'{stata, collectcode=TRUE}

See Doug Hemken's excellent documentation Statamarkdown here.

```
use simulated_multilevel_data.dta
```

1.5.1.2 Graph

```
twoway scatter outcome warmth, xtitle("warmth") ytitle("outcome") title("Outcome by Parental
quietly graph export scatter.png, replace
```

> arental Warmth")

1.5.1.3 Run The Model

```
mixed outcome warmth physical_punishment group HDI || country: warmth
```

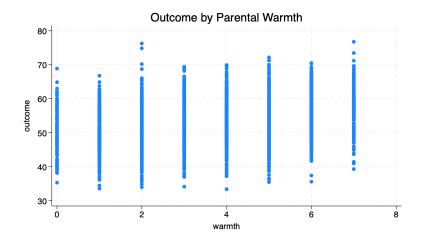


Figure 1.2: Outcome by Parental Warmth

Performing EM optimization ...

```
Performing gradient-based optimization:
```

Iteration 0: Log likelihood = -9668.198 Iteration 1: Log likelihood = -9667.9551 Iteration 2: Log likelihood = -9667.9534

Iteration 3: Log likelihood = -9667.9533 Iteration 4: Log likelihood = -9667.9532

Computing standard errors ...

Mixed-effects ML regression Group variable: country

Number of obs = 3,000Number of groups =

Obs per group:

100 100.0 avg =

100

Wald chi2(4) = 401.26 = 0.0000 Prob > chi2

Log likelihood = -9667.9532

outcome | Coefficient Std. err. P>|z| [95% conf. interval] warmth | .9616447 .0581825 16.53 0.000 .8476091 1.07568

<pre>physical_punishment</pre>	1	8453802	.0798155	-10.59	0.000	-1.001816	6889448
group	1	1.084344	.2200539	4.93	0.000	.6530461	1.515642
HDI	1	.010557	.0204522	0.52	0.606	0295286	.0506426
_cons	1	49.87963	1.436612	34.72	0.000	47.06392	52.69534

Random-effects parameters	Estimate	Std. err.	[95% conf.	_
country: Independent var(warmth) var(_cons)	1.83e-06 3.370262	.0000173 .9633726	1.76e-14 1.924651	190.9774 5.901676
var(Residual)	36.01906	.9346936	34.23291	37.89842
IR test vs linear model: chi2(2) = 198 01		Prob > chi	2 = 0 0000

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

1.5.2 R

In R lme4, the general syntax for a multilevel model of the form described in Equation ?? is:

```
lmer(y \sim x + z + (1 + x \mid | group), data = ...)
```

1.5.2.1 Get The Data

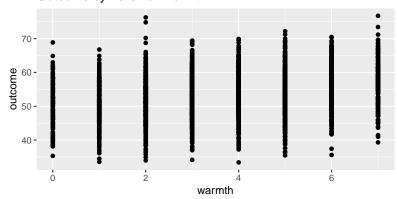
```
library(haven)

df <- read_dta("simulated_multilevel_data.dta")</pre>
```

1.5.2.2 Graph

```
library(ggplot2)
ggplot(df,
```

Outcome by Parental Warmth



1.5.2.3 Run The Model

```
Linear mixed model fit by REML ['lmerMod']
Formula: outcome ~ warmth + physical_punishment + group + HDI + ((1 |
    country) + (0 + warmth | country))
Data: df
```

REML criterion at convergence: 19350.3

Scaled residuals:

```
Min 1Q Median 3Q Max -3.4496 -0.6807 0.0016 0.6864 3.1792
```

Random effects:

```
Groups Name Variance Std.Dev.
country (Intercept) 3.611568 1.90041
country.1 warmth 0.001876 0.04331
Residual 36.049124 6.00409
Number of obs: 3000, groups: country, 30
```

Fixed effects:

	Estimate	Std.	Error	t value
(Intercept)	49.88754	1	.48203	33.662
warmth	0.96155	0	.05875	16.367
<pre>physical_punishment</pre>	-0.84556	0	.07986	-10.588
group	1.08471	0	.22017	4.927
HDI	0.01044	0	.02116	0.493

Correlation of Fixed Effects:

```
(Intr) warmth physc_ group
warmth -0.126
physcl_pnsh -0.135 -0.025
group -0.218 -0.010 -0.019
```

HDI -0.925 -0.006 0.008 -0.001

1.5.3 Julia

In Julia MixedModels, the general syntax for a multilevel model of the form described in Equation ?? is:

```
fit(MixedModel, @formula(y \sim x + z + (1 + x \mid group)), data)
```

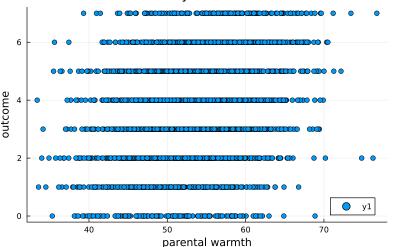
1.5.3.1 Load The Needed Packages And Load The Data

```
using Tables, MixedModels, StatFiles, DataFrames, CategoricalArrays, DataFramesMeta

df = DataFrame(load("simulated_multilevel_data.dta"))
```

1.5.3.2 Graph

Outcome by Parental Warmth



1.5.3.3 Change Country To Categorical

```
@transform!(df, :country = categorical(:country))
```

1.5.3.4 Run The Model

Linear mixed model fit by maximum likelihood

outcome ~ 1 + warmth + physical_punishment + group + HDI + (1 + warmth | country) logLik -2 logLik AIC AICc BIC -9667.9392 19335.8783 19353.8783 19353.9385 19407.9357

Variance components:

Column Variance Std.Dev. Corr.

country (Intercept) 3.2369484 1.7991521

warmth 0.0001080 0.0103903 +1.00

Residual 36.0187144 6.0015593

Number of obs: 3000; levels of grouping factors: 30

Fixed-effects parameters:

	Coef.	Std. Error	z	Pr(> z)
(Intercept)	49.9018	1.43435	34.79	<1e-99
warmth	0.961545	0.0582135	16.52	<1e-60
physical_punishment	-0.845389	0.0798149	-10.59	<1e-25
group	1.08524	0.220055	4.93	<1e-06
HDI	0.0101984	0.0204401	0.50	0.6178

2 Cross-Classified Models in Stata, R and Julia

2.1 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.

In this data, events are nested inside persons which are in turn nested in countries, since in this data, individuals never change countries. However, the use of a cross-classified framework would allow for a situation in which persons moved from country to country, and experienced different events in different countries.

Below, I describe the use of Stata, R, and Julia to estimate cross-classified models. Because this document is built by Quarto, I describe calling these programs from within a Quarto environment. However, each piece of software could be used individually and separately.

2.2 The Data

The examples below use the simulated_multilevel_longitudinal file from *Multilevel Thinking*. Here is a direct link to download the data.

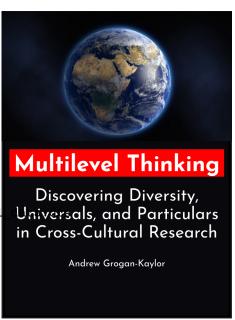


Figure 2.1: Book Cover For Multilevel Thinking