## **Tables for Multilevel Models in Stata**

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## 1 Introduction

Stata has a number of ways of making tables. There are some alternative commands, and some *tweaks* that may be especially useful for multilevel models.

#### 2 Data Source

The data used in this example are derived from the R package *Functions and Datasets for "Forest Analytics with R"*.

According to the documentation, the source of these data are: "von Guttenberg's Norway spruce (Picea abies [L.] Karst) tree measurement data."



Figure 1: Old Tjikko, a 9,550 Year Old Norway Spruce in Sweden

The documentation goes on to further note that:

"The data are measures from 107 trees. The trees were selected as being of average size from healthy and well stocked stands in the Alps."

# 3 Setup

```
clear all // clear workspace
use "gutten.dta", clear // use tree data as example
describe // describe the data
```

Contains data from gutten.dta Observations: 1,200 Variables: 9

Variables: 9 19 Feb 2020 08:23

Variable name	Storage type	Display format	Value label	Variable label
site	long	%9.0g	site	site
location	long	%9.0g	location	location
tree	long	%9.0g		tree
age_base	long	%9.0g		age.base
height	double	%9.0g		height
dbh_cm	double	%9.0g		dbh.cm
volume	double	%9.0g		volume
age_bh	double	%9.0g		age.bh
tree_ID	long	%9.0g	tree_ID	tree.ID

## 4 Estimate a Multilevel Model

```
mixed height age_base i.site || tree_ID: // mixed model
est store M1 // store the estimates (this would work with multiple stored estimates)
```

```
Performing EM optimization ...
Performing gradient-based optimization:
Iteration 0: Log likelihood = -3051.1192
Iteration 1: Log likelihood = -3051.1192
Computing standard errors ...
Mixed-effects ML regression
                                          Number of obs = 1,200
                                          Number of groups = 107
Group variable: tree_ID
                                          Obs per group:
                                                           5
                                                    min =
                                                    avg = 11.2
                                                    max = 15
                                          Wald chi2(5) = 8651.66
Prob > chi2 = 0.0000
                                          Prob > chi2
Log likelihood = -3051.1192
   height | Coefficient Std. err. z > |z| [95% conf. interval]
age_base | .2144446 .0023691 90.52 0.000 .2098014 .2190879
```

#### **5 Use Estimates Table**

estimates table M1, b(%9.3f) star // nicely formatted table of results

```
Variable | M1

height | age_base | 0.214***

site | 2 | -3.316***
3 | -8.095***
4 | -11.510***
5 | -15.866***

_cons | 8.233***

lns1_1_1 | _cons | 0.387***

lnsig_e | _cons | 1.064***

Legend: * p<0.05; ** p<0.01; *** p<0.001
```

# 6 Use Estimates Table With, variance post

Frustratingly, with multilevel models, the default behavior of estimates table is to report the ln of the random effects. I use the , variance post option to post the variance rather than the logarithm of the variance.

Notice how , variance post essentially *replays* the results, but with the random effects as variances, rather than as the logarithm of the standard deviation.

```
mixed height age_base i.site || tree_ID: // mixed model
estat sd, variance post // post results as variance scale rather than log scale
est store M2 // store the estimates (this would work with multiple stored estimates)
* nicely formatted table of results
estimates table M1 M2, b(%9.3f) star ///
title("M1 and M2 are the Same Model Presented Differently")
```

```
Performing EM optimization ...
Performing gradient-based optimization:
Iteration 0: Log likelihood = -3051.1192
Iteration 1: Log likelihood = -3051.1192
Computing standard errors ...
Mixed-effects ML regression
                                          Number of obs = 1,200
                                          Number of groups = 107
Group variable: tree_ID
                                          Obs per group:
                                                     min = 5
                                                     avg = 11.2
                                                     max =
                                                             15
                                          Wald chi2(5) = 8651.66
Prob > chi2 = 0.0000
                                          Prob > chi2
Log likelihood = -3051.1192
    height | Coefficient Std. err. z P>|z| [95% conf. interval]
age_base | .2144446 .0023691 90.52 0.000 .2098014 .2190879
      site |
       2 | -3.316408 .4738969 -7.00 0.000 -4.245229 -2.387587
        3 | -8.094846 .5358151 -15.11 0.000 -9.145024 -7.044667
        4 | -11.50985 .5291215 -21.75 0.000
                                              -12.54691 -10.47279
```

5	-15.86582	.7116202	-22.30	0.000	-17.26057	-14.47107				
cons	8.233362	4002147	20 12	0 000	7 /31316	0 035408				
_cons	0.233302	.4092147	20.12		7.431310	9.033400				
Random-effec	cts parameters	•								
tree_ID: Ident		 	1508 /1	00///5	1.511891	3 116037				
	۰	+								
	var(Residual	)   8.392	966 .3	586298	7.718693	9.12614				
LR test vs. linear model: $chibar2(01) = 135.90$ Prob >= $chibar2 = 0.0000$										
	Coefficient	Std. err.	7	P> 7	[95% conf.	intervall				
height										
age_base	.2144446	.0023691	90.52	0.000	.2098014	.2190879				
site   2	-3.316408	4738060	-7 00	0.000	-4.245229	-2.387587				
3		.5358151		0.000	-9.145024	-7.044667				
4		.5291215		0.000	-12.54691	-10.47279				
5				0.000	-17.26057	-14.47107				
İ										
_cons	8.233362	.4092147	20.12	0.000	7.431316	9.035408				
tree_ID										
	2.170508	. 4004445			1.511891	3.116037				
Residual										
var(e)	8.392966	. 3586298			7.718693	9.12614				
M1 and M2 are the Same Model Presented Differently										
.,										
Variable	M1	M2								
height	hoight I									
age_base	0.214***	0.21	4***							
5										
site										
2	-3.316***	-3.31	.6***							

```
3 | -8.095*** -8.095***
4 | -11.510*** -11.510***
5 | -15.866*** -15.866***

_cons | 8.233*** 8.233***

lns1_1_1 | __cons | 0.387***

lnsig_e | __cons | 1.064***

tree_ID | var(_cons) | 2.171***

Residual | var(e) | 8.393***

Legend: * p<0.05; ** p<0.01; *** p<0.001
```

#### 7 Use etable

etable is a newer Stata command that is very useful for making nicely formatted tables. etable works with one estimate or multiple estimates.

```
etable, estimates(M1) /// use these estimate(s)
novarlabel /// variable names only
cstat(_r_b) /// beta's only
showstars showstarsnote ///
column(estimate) // column is modelname
```

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

There is also a very helpful export option for exporting these tables to a variety of ouput formats. See help etable in Stata for more information.