

Tables for Multilevel Models in Stata

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1 Introduction

Stata has a number of ways of making tables. Here 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	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
Sorted by:				

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
Group variable: tree_ID	Number of groups	=	107
	Obs per group:		
	min	=	5
	avg	=	11.2
	max	=	15
	Wald chi2(5)	=	8651.66
Log likelihood = -3051.1192	Prob > chi2	=	0.0000

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 .4092147 20.12 0.000 7.431316 9.035408
-----
Random-effects parameters | Estimate Std. err. [95% conf. interval]
-----+-----
tree_ID: Identity |
      var(_cons) | 2.170508 .4004445 1.511891 3.116037
-----+-----
      var(Residual) | 8.392966 .3586298 7.718693 9.12614
-----
LR test vs. linear model: chibar2(01) = 135.90 Prob >= chibar2 = 0.0000

```

5 Use estimates table {#sec-esttable}

```
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 store With , variance post

Frustratingly, as you can see in ?@sec-esttable, with multilevel models, the default behavior of estimates table is to report the *ln* of the random effects. Below, 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)
```

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

Group variable: tree_ID

Number of obs = 1,200

Number of groups = 107

Obs per group:

min = 5

avg = 11.2

max = 15

Wald chi2(5) = 8651.66

Prob > chi2 = 0.0000

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	.4092147	20.12	0.000	7.431316	9.035408

Random-effects parameters				Estimate	Std. err.	[95% conf. interval]		
tree_ID: Identity								
	var(_cons)		2.170508	.4004445		1.511891	3.116037	
	var(Residual)		8.392966	.3586298		7.718693	9.12614	

LR test vs. linear model: chibar2(01) = 135.90						Prob >= chibar2 = 0.0000		

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

height								
	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	.4092147	20.12	0.000	7.431316	9.035408

tree_ID								
	var(_cons)		2.170508	.4004445			1.511891	3.116037

Residual								
	var(e)		8.392966	.3586298			7.718693	9.12614

7 Use estimates table To Compare These Approaches

⚠ We Usually Use estimates table for *Different* Models

When used with multiple sets of estimates, we usually use estimates table to present the results of *different* models, rather than the same model presented in different ways. Below, however, for the sake of illustration, we present the *same* model in two different ways.

* nicely formatted table of results

```
estimates table M1 M2, b(%9.3f) star ///
      title("M1 and M2 are the Same Model Presented Differently")
```

M1 and M2 are the Same Model Presented Differently

Variable	M1	M2
height		
age_base	0.214***	0.214***
site		
2	-3.316***	-3.316***
3	-8.095***	-8.095***
4	-11.510***	-11.510***
5	-15.866***	-15.866***
_cons	8.233***	8.233***
lnsl_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

8 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
```

```

                                M1
-----
age_base                0.214 **
site
  2                   -3.316 **
  3                   -8.095 **
  4                  -11.510 **
  5                  -15.866 **
_cons                   8.233 **
var(_cons)              2.171
var(e)                  8.393
Number of observations    1200
-----
** p<.01, * p<.05
```

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

9 Add One More Set of Estimates for Illustration

9.1 Multiple Estimates With `estimates` table

```

mixed height age_base i.site i.location || tree_ID: // mixed model

estat sd, variance post // post results as variance scale rather than log scale

est store M3 // store the estimates (this would work with multiple stored
estimates)

est table M2 M3, b(%9.3f) star
```

Performing EM optimization ...

Performing gradient-based optimization:
Iteration 0: Log likelihood = -3047.8267
Iteration 1: Log likelihood = -3047.8267

Computing standard errors ...

Mixed-effects ML regression Number of obs = 1,200

Group variable: tree_ID

Number of groups = 107
 Obs per group:
 min = 5
 avg = 11.2
 max = 15

Log likelihood = -3047.8267

Wald chi2(11) = 8700.21
 Prob > chi2 = 0.0000

height	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
age_base	.2143854	.0023822	89.99	0.000	.2097163	.2190544
site						
2	-2.994348	.5335979	-5.61	0.000	-4.04018	-1.948515
3	-7.764809	.563856	-13.77	0.000	-8.869947	-6.659672
4	-10.84402	.6356708	-17.06	0.000	-12.08991	-9.59813
5	-15.17887	.7953014	-19.09	0.000	-16.73763	-13.6201
location						
2	-.3215123	1.246019	-0.26	0.796	-2.763665	2.120641
3	.4745482	.6385101	0.74	0.457	-.7769087	1.726005
4	.0598813	.7092946	0.08	0.933	-1.330311	1.450073
5	-.4502186	.5372169	-0.84	0.402	-1.503144	.6027071
6	-.2549412	.7074584	-0.36	0.719	-1.641534	1.131652
7	-1.453754	.7466009	-1.95	0.052	-2.917065	.0095567
_cons	8.180898	.5441571	15.03	0.000	7.11437	9.247426

Random-effects parameters	Estimate	Std. err.	[95% conf. interval]	
tree_ID: Identity				
var(_cons)	1.981234	.3765076	1.365137	2.875382
var(Residual)	8.396723	.3589345	7.721889	9.130533

LR test vs. linear model: chibar2(01) = 118.04 Prob >= chibar2 = 0.0000

	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
height						
age_base	.2143854	.0023822	89.99	0.000	.2097163	.2190544
site						
2	-2.994348	.5335979	-5.61	0.000	-4.04018	-1.948515

3		-7.764809	.563856	-13.77	0.000	-8.869947	-6.659672
4		-10.84402	.6356708	-17.06	0.000	-12.08991	-9.59813
5		-15.17887	.7953014	-19.09	0.000	-16.73763	-13.6201
location							
2		-.3215123	1.246019	-0.26	0.796	-2.763665	2.120641
3		.4745482	.6385101	0.74	0.457	-.7769087	1.726005
4		.0598813	.7092946	0.08	0.933	-1.330311	1.450073
5		-.4502186	.5372169	-0.84	0.402	-1.503144	.6027071
6		-.2549412	.7074584	-0.36	0.719	-1.641534	1.131652
7		-1.453754	.7466009	-1.95	0.052	-2.917065	.0095567
_cons		8.180898	.5441571	15.03	0.000	7.11437	9.247426

tree_ID							
var(_cons)		1.981234	.3765076			1.365137	2.875382

Residual							
var(e)		8.396723	.3589345			7.721889	9.130533

Variable		M2 M3

height		
age_base		0.214*** 0.214***
site		
2		-3.316*** -2.994***
3		-8.095*** -7.765***
4		-11.510*** -10.844***
5		-15.866*** -15.179***
location		
2		-0.322
3		0.475
4		0.060
5		-0.450
6		-0.255
7		-1.454
_cons		8.233*** 8.181***

tree_ID		
var(_cons)		2.171*** 1.981***

```

Residual      |
var(e) |      8.393***      8.397***
-----
Legend: * p<0.05; ** p<0.01; *** p<0.001

```

9.2 Multiple Estimates With etable

```

mixed height age_base i.site i.location || tree_ID: // mixed model

est store M4

etable, estimates(M1 M4) /// use these estimate(s)
novarlabel /// variable names only
cstat(_r_b) /// beta's only
showstars showstarsnote ///
column(estimate) // column is modelname

```

Performing EM optimization ...

Performing gradient-based optimization:
Iteration 0: Log likelihood = -3047.8267
Iteration 1: Log likelihood = -3047.8267

Computing standard errors ...

Mixed-effects ML regression	Number of obs	=	1,200
Group variable: tree_ID	Number of groups	=	107
	Obs per group:		
	min	=	5
	avg	=	11.2
	max	=	15
	Wald chi2(11)	=	8700.21
Log likelihood = -3047.8267	Prob > chi2	=	0.0000

height	Coefficient	Std. err.	z	P> z	[95% conf. interval]
age_base	.2143854	.0023822	89.99	0.000	.2097163 .2190544
site					
2	-2.994348	.5335979	-5.61	0.000	-4.04018 -1.948515
3	-7.764809	.563856	-13.77	0.000	-8.869947 -6.659672
4	-10.84402	.6356708	-17.06	0.000	-12.08991 -9.59813
5	-15.17887	.7953014	-19.09	0.000	-16.73763 -13.6201

```

location |
  2 | -.3215123  1.246019  -0.26  0.796  -2.763665  2.120641
  3 | .4745482  .6385101   0.74  0.457  -.7769087  1.726005
  4 | .0598813  .7092946   0.08  0.933  -1.330311  1.450073
  5 | -.4502186  .5372169  -0.84  0.402  -1.503144  .6027071
  6 | -.2549412  .7074584  -0.36  0.719  -1.641534  1.131652
  7 | -1.453754  .7466009  -1.95  0.052  -2.917065  .0095567
    |
    _cons |  8.180898  .5441571  15.03  0.000   7.11437  9.247426
-----

Random-effects parameters | Estimate Std. err. [95% conf. interval]
-----+-----
tree_ID: Identity         |
      var(_cons) |  1.981234  .3765076   1.365137  2.875382
-----+-----
      var(Residual) |  8.396723  .3589345   7.721889  9.130533
-----

LR test vs. linear model: chibar2(01) = 118.04      Prob >= chibar2 = 0.0000

-----
                                M1      M4
-----
age_base                0.214 **   0.214 **
site
  2                -3.316 **  -2.994 **
  3                -8.095 **  -7.765 **
  4               -11.510 ** -10.844 **
  5               -15.866 ** -15.179 **
location
  2                        -0.322
  3                        0.475
  4                        0.060
  5                       -0.450
  6                       -0.255
  7                       -1.454
_cons                8.233 **   8.181 **
var(_cons)           2.171     1.981
var(e)               8.393     8.397
Number of observations 1200     1200
-----
** p<.01, * p<.05

```