

Module 11: Three-Level Multilevel Models

Stata Practical

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Pre-requisites

- Modules 1-5

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Module 11 (Stata Practical): Three-Level Multilevel Models

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Introduction to the Television School and Family Smoking Prevention and Cessation Project

We will analyse data from the Television School and Family Smoking Prevention and Cessation Project (TVSFP) (Flay *et al.*, 1989). The project was designed to test the effect of two different school-based interventions on student tobacco and health knowledge: (1) A social-resistance classroom curriculum (CC); and (2) A television-based programme.

The study sample involved schools with seventh-grade students (age 12 to 13 years) in Los Angeles and San Diego, California. Schools were randomized to one of the four study conditions formed by crossing the two interventions in a 2×2 design.

		Television-based programme (TV)	
		No	Yes
Classroom Curriculum (CC)	No	Neither intervention	TV only
	Yes	CC only	CC and TV

The two interventions were delivered to the seventh-grade students in these schools in spring 1986. Students were baselined in January 1986, completed an immediate postintervention questionnaire in April 1986, a one-year follow-up questionnaire in April 1987, and a two-year follow-up questionnaire in April 1988. At each time point, students' knowledge was assessed using a tobacco and health knowledge scale (THKS), constructed as the number of correct answers to seven binary questionnaire items.

The data were restudied by Hedeker *et al.* (1994) who used them to illustrate the importance of clustering in clinical and public health research and how multilevel models could be used to account for two-level and three-level hierarchical clustering structures. They concentrated on the sub sample of students who studied at 28 Los Angeles schools and only analysed data from the baseline and postintervention time points. Students who missed data at either time point were listwise deleted.

In this Module, we will explore the three-level hierarchical structure of the data: students (level 1) in classrooms (level 2) in schools (level 3). We will fit three-level multilevel models to examine the relative importance of schools and classrooms as influences on student tobacco and health knowledge and we will pay particular attention to assessing the possible causal effects of the CC and TV interventions.

There is good reason to expect both school and classroom effects on students' THKS scores. While schools were randomly assigned to the four study conditions, implementation of the CC and TV interventions were carried out at the classroom level. It seems very likely that some schools and teachers would have been more enthused about the interventions than others and this is likely to have had a direct effect on the success of the interventions. We therefore expect to see both between-school and within-school-between-classroom variation in students' THKS

scores, even after accounting for baseline differences in their tobacco and health knowledge.

We use the Hedeker *et al.* sub sample of the original data. The data consist of 1,600 students (level 1) nested within 135 classrooms (level 2) nested within 28 schools (level 3).

The response variable is students' postintervention THKS. We shall treat this score as a continuous response variable in our multilevel models, though we note that we could equally treat this response as ordinal and therefore fit ordinal response multilevel models (see Module 9). The predictor variables of key interest are the school level binary indicators of whether each school was randomly assigned to the CC or TV interventions. The predictor variables also include students' baseline THKS scores. We will include this predictor variable in our models to adjust for baseline variation in students' tobacco and health knowledge.

The dataset contains the following variables

Variable name	Description and codes
schoolid	School ID
classid	Class ID
studentid	Student ID
postthks	Postintervention THKS score. Scores range from 0 to 7, with a higher score indicating a higher tobacco and health knowledge
prethks	Baseline THKS score. Scores are measured on the same scale as postthks .
cc	Classroom curriculum (CC) (0 = no CC, 1 = CC)
tv	Television (TV) (0 = no TV, 1 = TV)
ccXtv	CC × TV, the interaction between CC and TV. The variable is constructed by multiplying the variables cc and tv . Note that ccXtv is also binary and 1 = both CC and TV and 0 otherwise.

P11.1 Examining and Describing the Data

Load '11.1.dta' into memory and open the do-file for this lesson

From within the LEMMA learning environment

- Go to **Module 11: Three-Level Multilevel Models**, and scroll down to **Stata files**
- Click '**11.1.dta**' to open the dataset

and use the `describe` command to produce a summary of the dataset

```
. describe

Contains data from 11.1.dta
  obs:      1,600
  vars:      8                25 Jul 2012 14:43
  size:     20,800
-----
storage  display      value
variable name  type   format      label      variable label
-----
schoolid      int    %8.0g                School ID
classid       float %12.0g              Class ID
studentid     int    %9.0g                Student ID
postthks      byte  %8.0g                Postintervention THKS
prethks       byte  %8.0g                Baseline THKS
cc            byte  %8.0g                Classroom curriculum (CC)
tv            byte  %8.0g                Television (TV)
ccXtv        byte  %8.0g                Interaction (CC*TV)
-----
Sorted by:  schoolid  classid
```

The data consist of 1,600 observations on 8 variables and each variable has been given a variable label.

Standard univariate summary statistics can be requested using the `summarize` command

```
. summarize

Variable |      Obs      Mean   Std. Dev.   Min      Max
-----+-----
schoolid |     1600   421.9388   112.6662     193     515
classid  |     1600  422042.3  112667.2   193101  515113
studentid |     1600    800.5   462.0245      1    1600
postthks |     1600   2.661875   1.382929      0      7
prethks  |     1600   2.069375   1.26018      0      6
-----+-----
cc       |     1600   .476875   .4996211      0      1
tv      |     1600   .499375   .5001559      0      1
ccXtv   |     1600   .239375   .4268354      0      1
```

We see, for example, that the response variable `postthks` ranges from 0 to 7. We shall describe a range of summary statistics for the response and predictor variables in P11.1.2.

P11.1.1 Exploring the three-level data structure

We start by using the `list` command to list the data for the first 10 students in the data. We use the `compress` option to abbreviate the variable names and to therefore compress the width of the columns in the output. Doing this prevents each line of the output from being wrapped on to a second line and so ensures readability.

```
. list in 1/10, compress
```

	sch~d	clas~d	stu~d	pos~s	pre~s	cc	tv	ccXtv
1.	193	193101	1	2	1	0	0	0
2.	193	193101	2	2	3	0	0	0
3.	193	193101	3	3	0	0	0	0
4.	193	193101	4	2	3	0	0	0
5.	193	193101	5	1	1	0	0	0
6.	193	193101	6	2	2	0	0	0
7.	193	193101	7	4	3	0	0	0
8.	193	193101	8	2	3	0	0	0
9.	193	193101	9	3	3	0	0	0
10.	193	193101	10	3	1	0	0	0

We see, for example, that student 1 was taught in class 193101 within school 193. The student scored 1 out of 7 on the THKS at baseline (**pre~s**) and 2 out of 7 at postintervention (**pos~s**). The variables `cc` and `tv` (and therefore `ccXtv`) are both zero and so school 193 received neither intervention.

Next, we use the `codebook` command to confirm that the number of schools and classrooms in the data are 28 and 135, respectively.

```
. codebook schoolid classid
```

```
-----
```

schoolid						School ID

type:	numeric (int)					
range:	[193,515]		units:		1	
unique values:	28		missing .:		0/1600	
mean:	421.939					
std. dev:	112.666					
percentiles:	10%	25%	50%	75%	90%	
	197	405	415	509	514	

classid						Class ID

type:	numeric (float)					
range:	[193101,515113]		units:		1	
unique values:	135		missing .:		0/1600	
mean:	422042					
std. dev:	112667					

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