Panel Data and Multilevel Models for Categorical Outcomes: Setting up the Data

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We often have data where variables have been measured for the same subjects (or countries, or companies, or whatever) at multiple points in time. These are typically referred to as Panel Data or as Cross-Sectional Time Series Data. We need special techniques for analyzing such data, e.g. it would be a mistake to treat 200 individuals measured at 5 points in time as though they were 1,000 independent observations. Therefore, Stata has an entire manual and suite of XT commands devoted to panel data, e.g. xtreg, xtlogit, xtpoisson, etc. Some other commands, like clogit, can also sometimes be used. (Conversely, the xt commands can sometimes be used when you don't have panel data, e.g. you have data from students within a school. In such situations you might also use the me, mixed-effects, commands.)

In order to use these commands, though, the data set needs to be properly structured. This will sometimes require that the data be restructured from wide to long. In wide format, a data set has one record for each subject. This record has several variables, e.g. income1, income2, income3, where each of the income variables gives the value of income at a different time point. In long format, the data are restructured to have one record for each subject for each time point. I am going to give some examples of how to do this, but if in doubt be sure to read the Stata documentation for help on setting up your data.

Here is an example from Allison's 2009 book *Fixed Effects Regression Models*. Data are from the National Longitudinal Study of Youth (NLSY). The data set has 1151 teenage girls who were interviewed annually for 5 years beginning in 1979. Here is a listing of the values for the first three cases:

- . version 13.1
- . use https://academicweb.nd.edu/~rwilliam/statafiles/teenpov, clear
- . rename inschool* school*
- . list in 1/3

mother4

1.	id 22	pov1	mother1	spouse1	school1	hours1 21	pov2	mother2	spouse2	school2	hours2 15	pov3 0	mother3	spouse3		hours3	pov4 0
	mc	other4 0	spous	se4 0	school4	hours4		age 16	black 0	pov5 0	mo	ther5	spou	se5 0	school5	hou	rs5 0
2.	id 75	pov1 0	mother1	spouse1	school1	hours1	pov2	mother2	spouse2	school2	hours2	pov3	mother3	spouse3		hours3	pov4 0
	mc	other4 0	spous	se4 0	school4	hours	- 1	age 17	black 0	pov5 1	mo	ther5	spous	se5 0	school5	hou	rs5 0
3.	id	pov1	mother1	anauaa1	school1	hours1	pov2	mother2	spouse2	school2	hours2	pov3	mother3	spouse3	school3	hours3	nou4
٥.	92	0 povi	motheri 0	spouse1	1	30	0	mother2	o spousez	1	noursz 27	0	mothers 0	spouses		nours3 24	pov4 1

The numbers at the ends of some variable names reflect the time period the variable refers to (1 =

mother5

1979, 2 = 1980, etc.) Variables without numbers in the names do not vary across time.
id is the subject id number and is the same across each wave of the survey

- pov_t is coded 1 if the subject was in poverty during that time period, 0 otherwise.
- age is the age at the first interview.
- black is coded 1 if the respondent is a Black person, 0 otherwise.
- mother, is coded 1 if the respondent currently has at least 1 child, 0 otherwise.
- spouse_t is coded 1 if the respondent is currently living with a spouse, 0 otherwise.
- school_t is coded 1 if the respondent is currently in school, 0 otherwise.
- hours_t is the hours worked during the week of the survey.

The data are currently in wide format. There is one record per case with multiple variables representing values at different points in time. We need to get the data into long format instead. In Stata, we can do this with the reshape command.

```
. reshape long pov mother spouse school hours, i(id) j(year)
(note: j = 1 2 3 4 5)
Dat.a
                               wide -> long
                              1151 ->
                                          5755
Number of obs.
                               28 -> 9
-> year
Number of variables
j variable (5 values)
xij variables:
                  pov1 pov2 ... pov5 -> pov
          mother1 mother2 ... mother5 -> mother
          spouse1 spouse2 ... spouse5 -> spouse
          school1 school2 ... school5 ->
                                          school
             hours1 hours2 ... hours5 -> hours
```

The reshape long part of the command told Stata we wanted to reshape the data from wide to long. (There is also a reshape wide command for going from long to wide.) The variable list that followed was the list of variables (actually the stubnames of the variables) that varied across time (you should use a consistent naming convention, e.g. pov1, mother1, etc. pov79, mother79, pov80, mother80, would have also been ok. Be careful about doing something like inc2, inc79, inc80, inc81, where inc2 = income squared; Stata will think inc2 is another of the time-varying variables.) The variables not listed are those that do not vary across time; their values will be copied on to each of the new records for the case. i (varlist) specifies the variables whose unique values denote a logical observation. i() is required. In this case only i (id) was needed but in other cases multiple variables might define a case. j (varname) specifies the variable whose unique values denote a subobservation. Here is what the reshaped data for the first 3 (now 15) cases looks like.

	id	year	age	black	pov	mother	spouse	school	hours
1.	22	1	16	0	1	0	0	1	21
2.	22	2	16	0	0	0	0	1	15
3.	22	3	16	0	0	0	0	1	3
4.	22	4	16	0	0	0	0	1	0
5.	22	5	16	0	0	0	0	1	0
6.	75	1	17	0	0	0	0	1	8
7.	75	2	17	0	0	0	0	1	0
8.	75	3	17	0	0	0	0	1	0
9.	75	4	17	0	0	0	0	1	4
10.	75	5	17	0	1	0	0	1	0
11.	92	1	16	0	0	0	0	1	30
12.	92	2	16	0	0	0	0	1	27
13.	92	3	16	0	0	0	0	1	24
14.	92	4	16	0	1	1	0	0	31
15.	92	5	16	0	1	1	0	0	0

Each of the original cases now has 5 records, one for each year of the study. The value of year varies from 1 to 5. The values of age (age at first interview) and black have been duplicated on each of the 5 records. Instead of 5 poverty variables, we have 1, whose value can differ across the five records (e.g. the original value of pov2 for id 22 is now the value of pov for id 22 year 2). The same is true for the other time-varying variables.

The next thing we want to do is xtset the data. The xtset command tells Stata that these are Panel data. The usual format is

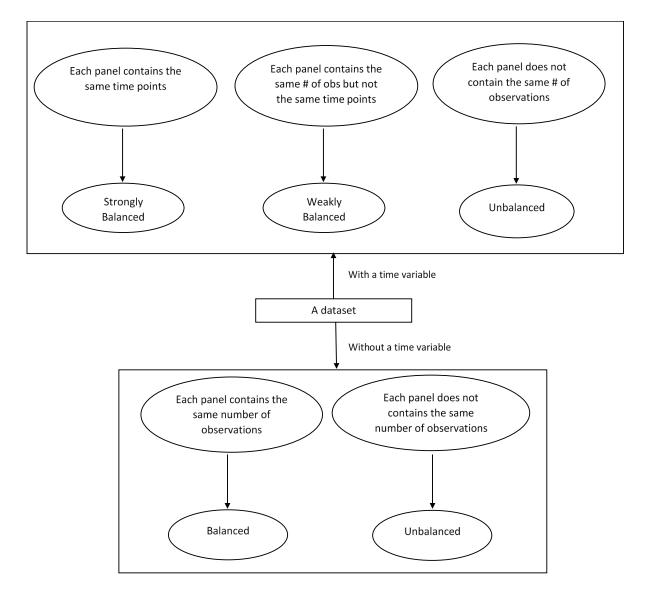
```
xtset panelvar xtset panelvar timevar
```

That is, we must tell Stata what the panelvar is; in this case it is id. The timevar is optional and may or may not be necessary depending on our analysis. In the current case the timevar is year. xtset typed with no parameters tells us how the data are xtset.

NOTE (adapted from the Stata 17 Manual): The terms balanced and unbalanced are often used to describe whether a panel dataset is missing some observations.

- If a dataset does not contain a time variable, then panels are considered balanced if each panel contains the same number of observations; otherwise, the panels are unbalanced.
- When the dataset contains a time variable, panels are said to be strongly balanced if each panel contains the same time points, weakly balanced if each panel contains the same number of observations but not the same time points, and unbalanced otherwise.

The following diagram prepared by Spring 2022 Teaching Assistant Junrong Sheng illustrates the above:



A data set might be unbalanced because data are missing for some years. If you were, say, analyzing countries, it might even be that the country did not exist during some time periods. Strongly balanced data are best but my understanding is that Stata can generally do a good job with unbalanced data.

Once the data are xtset, several commands are available to us; see help xt. For example, you can use the xtsum command, which is similar to the summarize command but contains some additional information.

. xtsum

Variable	2	Mean	Std. Dev.	Min	Max	Observations
id	overall between within	6016.672	3298.064 3299.211 0	22 22 6016.672	12539 12539 6016.672	N = 5755 n = 1151 T = 5
year	overall between within	3 	1.414336 0 1.414336	1 3 1	5 3 5	N = 5755 n = 1151 T = 5
age	overall between within	15.64639 	1.04682 1.047184 0	14 14 15.64639	17 17 15.64639	N = 5755 n = 1151 T = 5
black	overall between within	.5742832 	.4944942 .4946661	0 0 .5742832	1 1 .5742832	N = 5755 n = 1151 T = 5
pov	overall between within	.3768897 	.484649 .3100424 .3725925	0 0 4231103	1 1 1.17689	N = 5755 n = 1151 T = 5
mother	overall between within	.1986099 	.3989883 .3253864 .2310605	0 0 6013901	1 1 .9986099	N = 5755 n = 1151 T = 5
spouse	overall between within	.0992181 	.2989806 .2206498 .2018338	0 0 7007819	1 1 .8992181	N = 5755 n = 1151 T = 5
school	overall between within	.6304083	.4827361 .32013 .3614169	0 0 1695917	1 1 1.430408	N = 5755 n = 1151 T = 5
hours	overall between within	8.671764 	14.54341 9.363817 11.13062	0 0 -43.72824	90 52.4 72.07176	N = 5755 n = 1151 T = 5

The different values for the standard deviations can sometimes be useful. For id, age and black, the within subject standard deviation is 0. This is because, within each subject, the value of these variables does not vary, i.e. for each of the five records the case has, the values of these variables are the same. For year, the between subjects standard deviation is 0. This is because all subjects have the same set of values on year. For poverty, the between and within standard deviations are nearly the same. This tells us that the variation in poverty across women is nearly equal to that observed within a woman over time. That is, if you were to draw two women randomly from the data, the difference in poverty is expected to be nearly equal to the difference for the same woman in two randomly selected years.

As shown elsewhere, the amount of within-subject variability will impact which types of models work best for a particular problem.