# Dataset Manipulation

LPO 9951 | Fall 2015

**PURPOSE** Learning to manipulate datasets is a key skill for statistical analysis. Today we'll work on four skills: subsetting data using preserve and restore commands, appending data, doing a simple one-to-one merge, and collapsing data.

#### Read in data

For today's lecture, we'll use the apipop dataset, which is available online.

```
. global urldata "http://www.ats.ucla.edu/stat/stata/library/apipop"
. global datadir "../data/"
. // read web data into memory
. use $urldata, clear
```

## Subsetting with preserve and restore

A feature and sometimes curse of Stata is that it can only hold one dataset in active memory at a time. As a feature, it helps you keep your work organized and, through numerous warning messages, tries to make sure you don't lose your work by accidentally forgetting to save or mindlessly overwritting your data. The feature feels more like a curse when you have multiple datasets that you would like to work with simultaneously or, as we will do below, split a single dataset into smaller parts.

To repeatedly subset a large dataset, there are two primary choices:

- 1. Reload the full dataset into memory after each subset and save
- 2. Use the preserve and restore commands

In the code below, notice how the preserve and restore commands bookend the keep command, which keeps only those observations that fulfill the if statement (in this case, the type of school). The steps are:

- 1. preserve dataset in memory
- 2. subset to keep only school type that we want
- 3. save new subset dataset
- 4. restore old dataset

```
. // elementary schools
. preserve

. keep if stype == 1
(1,773 observations deleted)

. save ${datadir}elem, replace
file ../data/elem.dta saved
. restore
```

- . // high schools
- . preserve

```
. keep if stype == 2
(5,439 observations deleted)
. save ${datadir}hs, replace
file ../data/hs.dta saved
. restore
. // middle schools (keep this one in memory so no preserve/restore needed)
. keep if stype == 3
(5,176 observations deleted)
. save ${datadir}middle, replace
file ../data/middle.dta saved
```

NB: Note the construction of the global macro to save the datasets in the proper subdirectory. Because we to paste the realization of the global (../data/) and the file name, we have to use the construction \${<global name>} followed with no spaces by the name we want to give it. As you can see, S{datadir}hs becomes ../data/hs.dta.

# **Appending Data**

Appending data is done when we want to add additional *observations* to an existing dataset, using a dataset that has exactly the same variable names but different observations. Suppose you have data on high schools, middle schools, and elementary schools on a variety of performance indicators and you'd like to merge them together. The syntax uses, appropriately enough, the append command, which takes the format append <new dataset> (the command assumes the first dataset is the one in memory; remember that the middle school subset data are still in memory):

```
. append using ${datadir}elem (label yr_rnd already defined) (label awards already defined) (label both already defined) (label comp_imp already defined) (label sch_wide already defined) (label stype already defined)

. append using ${datadir}hs (label stype already defined) (label sch_wide already defined) (label sch_wide already defined) (label comp_imp already defined) (label both already defined) (label awards already defined) (label yr_rnd already defined)
```

The append command will not copy over labels from the using dataset, so you'll need to make sure they're right in the master dataset. The most common error with an append command is to not have exactly matching variable names.

#### Merging Data

You can also use Stata's merge command to do an append operation in special cases. This happens when the merging variable doesn't have repeated *observations* in the two datasets, which in turn have exactly the same

variable structure. Think of a Venn diagram where the circles contain exactly the same types of information, but don't overlap; in combining them, we've really just grown them into one bigger circle. One of the virtues of using merge when append will suffice is that you have access to more information about where the data came from once you're done.

```
. merge 1:1 snum using ${datadir}hs, gen(_merge_a)
(label yr_rnd already defined)
(label awards already defined)
```

(label both already defined)

. use \${datadir}elem, clear

(label comp\_imp already defined)

(label sch\_wide already defined)

(label stype already defined)

Result	# of obs.	
not matched from master from using		(_merge_a==1) (_merge_a==2)
matched	0	(_merge_a==3)

```
. merge 1:1 snum using ${datadir}middle, gen(_merge_b)
```

(label stype already defined)

(label sch\_wide already defined)

(label comp\_imp already defined)

(label both already defined)

(label awards already defined)

(label yr\_rnd already defined)

Result	# of obs.	
not matched from master from using		(_merge_b==1) (_merge_b==2)
matched	0	(_merge_b==3)

Once you've completed the merge, you can take a look at the \_merge\_\* variables that were generated to see where the data came from.

## . tab \_merge\_a

Cum.	Percent	Freq.	_merge_a
85.41 100.00	85.41 14.59	4,421 755	master only (1)   using only (2)
	100.00	5,176	Total

. tab \_merge\_b

_merge_b	Freq.	Percent	Cum.
master only (1)   using only (2)	5,176 1,018	83.56 16.44	83.56 100.00
Total	6,194	100.00	

## QUICK EXERCISE

Create a dataset that has just middle and elementary schools. Do this using first the append command and then the merge command.

## One-to-one merges

. // view merge stats

A one-to-one merge is when you have exactly the same *observations* but new variables to add to the dataset. Say you have *observations* with variables split across datasets, e.g., School 1 has variables A, B, and C in dataset 1 and variables X, Y, and Z in dataset two. As long as School 1 has a unique identifier—a name, an id number, etc—you can merge these two datasets together so that you have access to all of the school's variables for your analysis.

First, we need to subset our data again, only this time by splitting along columns (variables) rather than rows (observations):

```
. use $urldata, clear
. preserve
                                   // variable set 1
. keep snum api00 api99 ell meals
. save ${datadir}api_1, replace
file ../data/api 1.dta saved
. restore
                                      // variable set 2
. keep snum full emer
. save ${datadir}api_2, replace
file ../data/api_2.dta saved
Now for the merge and view of merge stats:
. merge 1:1 snum using ${datadir}api_1
   Result
                                  # of obs.
   not matched
   matched
                                  6,194 (_merge==3)
   _____
```

. tab \_merge

_merge	Freq.	Percent	Cum.
matched (3)	6,194	100.00	100.00
Total	   6,194	100.00	

## QUICK EXERCISE

Create a dataset that has only mobility and percent tested. Next create another dataset that has only the year round and percent responding variables. Now merge these two datasets together using a one-to-one merge.

## Collapsing data

Collapsing data refers to summarizing data across a type and creating a new dataset as a result. Say we want to create a county-level dataset from our school data, using the average figures for the schools across a set of characteristics. The command would look like this:

- . // reload main dataset, since we didn't preserve it before  $% \left( 1\right) =\left( 1\right) \left( 1\right)$
- . use \$urldata, clear
- . // count of unique counties in dataset
- . unique cnum

Number of unique values of cnum is 57 Number of records is 6194

- . // mean of pcttest and mobility within countyr  $% \frac{1}{2}\left( \frac{1}{2}\right) =\frac{1}{2}\left( \frac{1}{2}\right)$
- . collapse (mean) pcttest mobility, by (cnum)
- . // give count of number of observations (should be number of unique counties)
- . count

57

NB: Notice the check using the unique command. If the number didn't align or gave some unexpected value, we should recheck our data and code.

## QUICK EXERCISE

Create a district level dataset that contains district level averages for the following variables:

- 1. apioo
- 2. api99
- 3. ell
- 4. meals

Do the same thing using just district medians.

Init: 15 August 2015; Updated: 15 September 2015