SOC 561: Programing for the Social Sciences
Community Resource – Append, Merge, and Collapse
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A QUICK STATA GUIDE:
APPEND, MERGE, AND COLLAPSE

Before we begin, it is highly recommended that when performing append, merge, or collapse procedures the user performs the `preserve` / `restore` command beforehand to ensure their data are maintained. If you need a quick review, see the appendix.

1. APPEND

Adding cases / observations

Command:
`append using dataset.dta`

The append command combines the dataset in memory, known as the *master* dataset, with a dataset on disk, known as the *using* dataset. Typically, a user would implement the append command when they would like to add observations to an existing dataset with the same or similar variables. Let's assume we are interested in combining the following datasets:

```
. use city_size1

. list
```

	city	popula~n	sq_miles
1.	Boston	685094	90
2.	Chicago	2716000	234
3.	New York	8623000	468
4.	Philadelphia	1581000	142

```
. use city_size2.dta

. list
```

	city	popula~n	sq_miles
1.	Los Angeles	4000000	503
2.	Kansas City	488943	319
3.	Denver	704621	155
4.	St. Louis	308626	66

Each dataset provides three variables: city, population, and square miles. All variables are named and formatted the same for each dataset. To append "city_size2.dta" to "city_size1.dta" we use the following:

```
. use city_size1, clear

. append using city_size2, generate (new_obs) nolabel nonotes

. list
```

	city	popula~n	sq_miles	new_obs
1.	Boston	685094	90	0
2.	Chicago	2716000	234	0
3.	New York	8623000	468	0
4.	Philadelphia	1581000	142	0
5.	Los Angeles	4000000	503	1
6.	Kansas City	488943	319	1
7.	Denver	704621	155	1
8.	St. Louis	308626	66	1

The command produced a new dataset that combined the observations from the master dataset, (city_size1.dta) with the using dataset (city_size2.dta). Notice no new variables with the exception of "new_obs" were created in the process; only additional observations. The additional commands after the coma are optional and are not required to execute the command. The "`, generate (new_obs)`" option above provided a "new_obs" variable that identifies which observations were appended to the master dataset. Additionally, the options "`nolabel nonotes`" prevented any labels or notes from the using dataset copying over to the appended master dataset. A different option allows the user to omit variables from the using dataset. For example:

```
. use city_size1, clear

. append using city_size2, generate (new_obs) nolabel nonotes keep (population)

. list
```

	city	popula~n	sq_miles	new_obs
1.	Boston	685094	90	0
2.	Chicago	2716000	234	0
3.	New York	8623000	468	0
4.	Philadelphia	1581000	142	0
5.		4000000	.	1
6.		488943	.	1
7.		704621	.	1
8.		308626	.	1

The "`keep (population)`" identified only one variable to copy over while leaving the other fields ("city" and "sq_miles for this example) as missing.

If you attempt to append a using dataset with variables that do not match with the master dataset, they will be added to the appended dataset as additional variables. For example:

```
. use city_size1, clear

. append using city_size2sqrm, generate (new_obs)

. list
```

	city	popula~n	sq_miles	new_obs	sqr_mi~s
1.	Boston	685094	90	0	.
2.	Chicago	2716000	234	0	.
3.	New York	8623000	468	0	.
4.	Philadelphia	1581000	142	0	.
5.	Los Angeles	4000000	.	1	503
6.	Kansas City	488943	.	1	319
7.	Denver	704621	.	1	155
8.	St. Louis	308626	.	1	66

Even though the variables "sq_miles" and "sqr_miles" provide the same measure, they remained separate and provided missing data in observations where the variable name did not match.

Another consideration is if you are attempting to append a using dataset with a variable by the same name as the master dataset, but in a different format. For example, if you attempt to append the using dataset with a string variable (sq_miles) to a master dataset with a numeric variable by the same name (sq_miles), you will receive the following error message:

```
. use city_size1, clear

. append using city_size2strg
variable sq_miles is double in master but str3 in using data
    You could specify append's force option to ignore this numeric/string mismatch.
    The using variable would then be treated as if it contained numeric missing
    value.
r(106);
```

If we use the "`force`" option in this situation, Stata will append the using dataset and inform you that the variable from the using dataset will assume the format of the master dataset. Additionally, the values form the using datasets will change to missing.

```
. append using city_size2strg, force
(note: variable sq_miles was str3 in the using data, but will be double now)

. list
```

	city	popula~n	sq_miles
1.	Boston	685094	90
2.	Chicago	2716000	234
3.	New York	8623000	468
4.	Philadelphia	1581000	142
5.	Los Angeles	4000000	.
6.	Kansas City	488943	.
7.	Denver	704621	.
8.	St. Louis	308626	.

Major Options:

`generate(newvar)` newvar marks source of resulting observations
`keep(varlist)` keep specified variables from appending dataset(s)
`nolabel` do not copy value-label definitions from dataset(s) on disk
`nonotes` do not copy notes from dataset(s) on disk
`force` append string to numeric or numeric to string without error

For additional information and examples, you can view the online `append` manual here.

2. MERGE

Adding variables

Command:
```
merge using dataset.dta
```

The merge command combines the dataset in memory, known as the *master* dataset, with a dataset on disk, known as the *using* dataset. While `append` added observations to a master dataset, the general purpose of `merge` is to add variables to existing observations. In its simplest form from past Stata versions (the command above), datasets are merged based on their observation (or row) order (e.g., the first observation is paired with the first outcomes for each variable). This older syntax is not recommended as it can be potentially dangerous if the two datasets are sorted differently or possess more or less id variables (see merge warning in the appendix). You can produce the same results using the following command recognized by newer Stata versions:

```
merge 1:1 _n using filename
```

However, the new Stata commands in versions 10 or later mitigate chances for mismatched variables and observations. We will now focus on the two primary types: 'one-to-one' and a 'one-to-many' (or 'many-to-one').

One-to-one merging:

Command:
```
merge 1:1 varlist using filename
```

For this command "1:1" specifies that there is one id variable in each dataset that needs to be merged. For example, imagine you had a master dataset, "city_size.dta" that possessed the id variable "city" with multiple size variables (e.g., population, total square miles), and a separate using dataset, "city_market.dta" that possessed the same id variables with corresponding market variables (e.g., number of grocery retailers and total GDP). Given that you wanted to merge these two datasets and the presence of one identifier, "city" in this case, you would perform a one-to-one merge.

```
. use city_size

. list
```

	city	popula~n	sq_miles
1.	Boston	685094	90
2.	New York	8623000	468
3.	Chicago	2716000	234
4.	Philadelphia	1581000	142

```
. use city_market

. list
```

	city	grocer~l	GDP
1.	Boston	84	293
2.	New York	303	1550
3.	Chicago	262	525
4.	Philadelphia	180	347
5.	Kansas City	62	161

```
. use city_size.dta

. merge 1:1 city using city_market.dta

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

    matched                                 4  (_merge==3)
    -----------------------------------------

. list
```

	city	popula~n	sq_miles	grocer~l	GDP	_merge
1.	Boston	685094	90	84	293	matched (3)
2.	Chicago	2716000	234	262	525	matched (3)
3.	New York	8623000	468	303	1550	matched (3)
4.	Philadelphia	1581000	142	180	347	matched (3)
5.	Kansas City	.	.	62	161	using only (2)

Stata merged the using dataset, “city_market.dta” variables to the corresponding observations in the “city” variable within the master dataset, “city_size.dta.” Also, the merge occurs based on the id variable regardless of sort order. The command will also create an additional variable “merge” that identifies if an observation was matched in the merge. It provides three indicators:

1 = observation found only in the master dataset
2 = observation found only in the using dataset
3 = observation found in both master and using dataset (complete match)

The “Kansas City” observation received a “2” identifier because this observation was only provided in the using dataset. Additionally, it resulted in missing values for the variables in the master dataset. The merge can occur based on other id variables if desired.

Many-to-one & one-to-many merge:

Command:

```
merge 1:m varlist using filename

merge m:1 varlist using filename
```

You can also merge datasets that have similar id variables with observations at different levels of analysis. For example, let's suppose in addition to the "city_size.dta" and "city_market.dta" files, you have a "city_person.dta" dataset with variables that capture a person's city of residence and yearly income.

```
. use city_person,clear

. list
```

	city	per_id	income~r
1.	Boston	1	36000
2.	New York	2	80000
3.	Chicago	3	54000
4.	Philadelphia	4	130000
5.	Kansas City	5	70000
6.	Boston	6	34000
7.	Philadelphia	7	81000
8.	New York	8	65000
9.	New York	9	94000
10.	Chicago	10	49000

This dataset includes a personal id variable. Using "city" as the id variable, you can merge "city_size.dta" as the using dataset with "city_person.dta" as the master dataset. Performing a ***many-to-one merge*** produces the following output:

```
. merge m:1 city using city_size

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

    matched                                 9  (_merge==3)
    -----------------------------------------

. sort city per_id

. list
```

	city	per_id	income~r	popula~n	sq_miles	_merge
1.	Boston	1	36000	685094	90	matched (3)
2.	Boston	6	34000	685094	90	matched (3)
3.	Chicago	3	54000	2716000	234	matched (3)
4.	Chicago	10	49000	2716000	234	matched (3)
5.	Kansas City	5	70000	.	.	master only (1)
6.	New York	2	80000	8623000	468	matched (3)
7.	New York	8	65000	8623000	468	matched (3)
8.	New York	9	94000	8623000	468	matched (3)
9.	Philadelphia	4	130000	1581000	142	matched (3)
10.	Philadelphia	7	81000	1581000	142	matched (3)

Now, all city level measures are assigned to each person depending on which city they reside in. Notice that values for "population" and "sq_miles" are missing since "Kansas City" is an identifier only provided in the "city_person.dta" dataset, thus the `_merge==1` result. If we attempt a one-to-many command with the same using and master dataset arrangement, Stata will present an error:

```
. merge 1:m city using city_size
variable city does not uniquely identify observations in the master data
r(459);
```

The "city_person.dta" dataset fails to provide an id variable that Stata can recognize as a unique identifier (e.g., "Boston" is assigned to more than one observation). If we switch the datasets where "city_size.dta" is the master dataset and "city_person.dta" is the using dataset, a ***one-to-many merge*** is possible.

```
. use city_size, clear

. merge 1:m city using city_person

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

    matched                                 9  (_merge==3)
    -----------------------------------------

. sort city per_id

. list
```

	city	popula~n	sq_miles	per_id	income~r	_merge
1.	Boston	685094	90	1	36000	matched (3)
2.	Boston	685094	90	6	34000	matched (3)
3.	Chicago	2716000	234	3	54000	matched (3)
4.	Chicago	2716000	234	10	49000	matched (3)
5.	Kansas City	.	.	5	70000	using only (2)
6.	New York	8623000	468	2	80000	matched (3)
7.	New York	8623000	468	8	65000	matched (3)
8.	New York	8623000	468	9	94000	matched (3)
9.	Philadelphia	1581000	142	4	130000	matched (3)
10.	Philadelphia	1581000	142	7	81000	matched (3)

Notice the one-to-many merged dataset sorted on "city" and "per_id" produces the same output as the many-to-one merged dataset.

A ***many-to-many merge*** can occur when you are unaware of how many of the same identifiers exist between two datasets, but believe there is at least one pair. A many-to-many command is not recommended. As stated in the Stata Data Management Reference Manual (Release 15):

> Because m:m merges are such a bad idea, we are not going to show you an example. If you think that you need an m:m merge, then you probably need to work with your data so that you can use a 1:m or m:1 merge.

It is recommended that the user is familiar enough with the datasets they desire to merge that a many-to-one or one-to-many is used for the desired outcomes.

Major options:

`keepusing (varlist)`	allows you to merge only select variables from the using dataset.
`generate (newvar)`	changes "_merge" variable name to one of your choosing
`nogenerate`	"_merge" variable not created after a merge
`nolabel`	prevents value/label definitions copying over from the using dataset
`nonotes`	prevents notes copying over form the using dataset
`noreport`	prevents the match results from showing after the merge

For additional information and examples, you can view the online `merge` manual here.

3. Collapse

Command:
`collapse (statistic) var1, by (var2)`

This command takes an open (or master) dataset and creates a new dataset by summarizing statistics on a selected variable. Let's use the "city_person_cp.dta" dataset to go through some examples. For these examples we add three new variables: "female" (1 = female, 0 = male), "like_live" (rating of how much a person likes the city they live in (1 = completely dislike to 6 = completely like), and "willing_move" (rating of how willing a person is to move to a different city (1 = strongly unwilling to 6 = strongly willing).

```
. use city_person_cp.dta

. list
```

	city	per_id	income~r	female	like_l~e	willing_move
1.	Boston	1	36000	1	4	somewhat willing
2.	New York	2	80000	1	3	strongly willing
3.	Chicago	3	54000	0	2	strongly unwilling
4.	Philadelphia	4	130000	1	6	willing
5.	Kansas City	5	70000	0	4	somewhat unwilling
6.	Boston	6	34000	0	2	strongly willing
7.	Philadelphia	7	81000	1	3	somewhat unwilling
8.	New York	8	65000	0	1	somewhat unwilling
9.	New York	9	94000	1	6	unwilling
10.	Chicago	10	49000	1	3	strongly willing

The `collapse` command will allow us to find a statistic by a specific variable. For example, if we wanted to find the mean yearly income for each city based on the individual dataset, we would use the following:

```
. collapse income_yr, by (city)

. list
```

	city	income_yr
1.	Boston	35000
2.	Chicago	51500
3.	Kansas City	70000
4.	New York	79666.667
5.	Philadelphia	105500

The command 'collapsed' all individual yearly incomes in the dataset and produced a new dataset presenting the mean for each city. By default, `collapse` will provide the mean for each numeric variable listed. The collapse output can be changed to a variety of statistics. For example:

```
. use city_person_cp.dta, clear

. collapse (max) income_yr (mean) female like_live, by (city)

. list
```

	city	income~r	female	like_live
1.	Boston	36000	.5	3
2.	Chicago	54000	.5	2.5
3.	Kansas City	70000	0	4
4.	New York	94000	.66666667	3.3333333
5.	Philadelphia	130000	1	4.5

Here we ask Stata to find the maximum yearly income and the mean for female and like_live for each city. The output shows the highest yearly income for those in the sample who live in New York being $94,000 with roughly 67% of the New York respondents identifying as female and a mean score of how much they like living in New York being 3.33 (somewhat dislike). If we were to include the "willing_move" variable, Stata posts an error message identifying a "type mismatch" and require you to change this variable from a string to numeric format to perform the collapse.

```
. collapse (max) income_yr (mean) female like_live willing_move, by (city)
type mismatch
r(109);
```

Using multiple statistical outcomes from one collapse can make keeping track of statistic output somewhat difficult by looking at the variable name alone. Fortunately, Stata develops labels for each variable providing details on which statistic occurred from a `collapse`.

```
. describe

Contains data
  obs:             5
 vars:             4
 size:           180
```

variable name	storage type	display format	value label	variable label
city	str12	%12s		
income_yr	double	%10.0g		(max) income_yr
female	double	%10.0g		(mean) female
like_live	double	%10.0g		(mean) like_live

```
Sorted by: city
     Note: Dataset has changed since last saved.
```

If you wanted to perform more than one statistic for the same variable, you will need to tell Stata the new variable name. For example, if we wanted the max and the mean for yearly income, we would perform the following command:

```
. use city_person_cp, clear

. collapse (max) income_yr (mean) female like_live income_mean=income_yr, by (city)

. list
```

	city	income~r	female	like_live	income_~n
1.	Boston	36000	.5	3	35000
2.	Chicago	54000	.5	2.5	51500
3.	Kansas City	70000	0	4	70000
4.	New York	94000	.66666667	3.3333333	79666.667
5.	Philadelphia	130000	1	4.5	105500

```
. describe

Contains data
  obs:             5
 vars:             5
 size:           220
```

variable name	storage type	display format	value label	variable label
city	str12	%12s		
income_yr	double	%10.0g		(max) income_yr
female	double	%10.0g		(mean) female
like_live	double	%10.0g		(mean) like_live
income_mean	double	%10.0g		(mean) income_yr

```
Sorted by: city
     Note: Dataset has changed since last saved.
```

Stata will also allow a collapse to condition on combinations of variables. For example, we could perform the same collapse as the previous command, however instead of collapsing on just the city, we could collapse on the city and whether or not you identify as a female. For this next example, we will ask Stata to provide a count for the number of observations that occur for the specified collapse categories. For this to work, you must use a variable that has no missing values; “per_id” is a good fit for this case.

```
. use city_person_cp.dta, clear

. collapse (max) income_yr (mean) like_live income_mean=income_yr (count) count=per_id, by (city female)

. list
```

	city	female	income~r	like_l~e	income~n	count
1.	Boston	0	34000	2	34000	1
2.	Boston	1	36000	4	36000	1
3.	Chicago	0	54000	2	54000	1
4.	Chicago	1	49000	3	49000	1
5.	Kansas City	0	70000	4	70000	1
6.	New York	0	65000	1	65000	1
7.	New York	1	94000	4.5	87000	2
8.	Philadelphia	1	130000	4.5	105500	2

The output shows that the statistics based on the city and stratified on whether you identify as a female or not. For example, the max yearly income for those who live in New York and identify

as female is $94,000, with the mean yearly income being $87,000 and an average “like_live” rating of 4.5 (like). This can be compared to those who live in New York who do not identify as a female (1 person in this simple dataset) with a yearly income of $65,000 and a “like_live” rating of 1 (strongly dislike). Additionally, Philadelphia does not have a “0” category because the dataset only captured two female observations.

Let’s now use a fictitious longitudinal dataset “city_hsgrad.dta” with variables that identify the city, city population, and percentage of high school graduates for the years 2017 and 2018 respectively. If we were interested in capturing the mean percentage of high school grads for the two different years, we could perform the following commands:

```
. use city_hsgrad, clear

. list
```

	city	year	popula~n	percnt~d
1.	Boston	2017	685094	72.4
2.	Boston	2018	670000	73.1
3.	Chicago	2017	2716000	73.5
4.	Chicago	2018	2591000	72.9
5.	New York	2017	8623000	71.1
6.	New York	2018	8498000	70.3
7.	Philadelphia	2017	1581000	81.2
8.	Philadelphia	2018	1526000	79.6

```
. collapse (mean) percnt_hsgrad, by (year)

. list
```

	year	percnt~d
1.	2017	74.55
2.	2018	73.975

However, our results assume equal weighting of the graduate percentages regardless of city population and provide inaccurate outputs. A more precise approach involves using a weight command that accounts for each percentage on their respective population size.

```
. collapse (mean) percnt_hsgrad [fw=population], by (year)

. list
```

	year	percnt_~d
1.	2017	72.818262
2.	2018	72.016553

Major Options:

`by(varlist)`	groups over which stat is to be calculated
`cw`	casewise deletion instead of all possible observations

For additional information and examples, you can view the online `collapse` manual here.

Appendix

A quick word on preserve / restore:

If you are writing a Stata program that temporarily changes the dataset by the following append, merge, or collapse commands, it is highly recommended that you preserve your original dataset by using the `preserve` command. After preserve is entered, you can experiment with different dataset manipulation commands without worry of permanently changing your original dataset. Simply use the command `restore` to go back to the preserved dataset.

```
preserve

[commands that alter the dataset]

restore
```

Additionally, if you are writing a Stata program that temporarily changes the order of the data and you want the data to be sorted in its original order at the end of execution, you can save a bit of programming by including `sortpreserve` on your program statement.

```
Program statdatasetfile, sortpreserve
```

Stata will automatically sort into its original order at the end of execution.

Merge Warning

Suppose we wanted to merge the following two files: autoexpense.dta and autosize_sortdif.dta.

```
. use autoexpense, clear
(1978 Automobile Data)

.       list
```

	make	price	mpg
1.	Toyota Celica	5,899	18
2.	BMW 320i	9,735	25
3.	Cad. Seville	15,906	21
4.	Pont. Grand Prix	5,222	19
5.	Datsun 210	4,589	35

```
. use autosize_sortdif, clear
(1978 Automobile Data)

.       list
```

	make	weight	length
1.	Datsun 210	2,020	165
2.	Toyota Celica	2,410	174
3.	BMW 320i	2,650	177
4.	Pont. Grand Prix	3,210	201
5.	Plym. Arrow	3,260	170
6.	Cad. Seville	4,290	204

Notice how the autosize_sortdif.dta file has the same car makes with the exception of the "Cadillac Seville," but they are in a different order. If a simple merge command is used, Stata will do so based on the order within each dataset (e.g., the first observation is paired with the first outcomes for each variable). Here's the result:

```
. merge using autosize_sortdif
(note: you are using old merge syntax; see [D] merge for new syntax)

. list
```

	make	price	mpg	weight	length	_merge
1.	Toyota Celica	5,899	18	2,020	165	3
2.	BMW 320i	9,735	25	2,410	174	3
3.	Cad. Seville	15,906	21	2,650	177	3
4.	Pont. Grand Prix	5,222	19	3,210	201	3
5.	Datsun 210	4,589	35	3,260	170	3
6.	Cad. Seville	.	.	4,290	204	2

The master dataset is now merged with the using dataset with a new "_merge" variable that indicates observations found in both master and using dataset (3) with the exception of the "Cadillac Seville" where observations were found only from the using dataset. There are two assumptions Stata was operating on: (1) the first variable was the id variable, and (2) the id variables were sorted in the same order. In this case Stata did not account for the different ordering of the "make" variable, and the datasets were merged incorrectly (e.g., the weight and length outcomes for the "Datsum 210" were assigned to the "Toyota Celica" in the merged dataset).