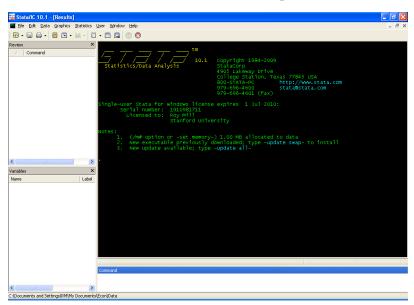
INTRODUCTION TO STATA

Stata - love at first sight?



Datasets

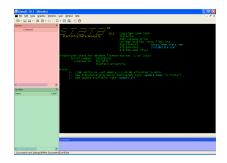
Datasets are the objects of statistical analysis. They contain a matrix of which rows represent different observations (draws of random variables) and the columns are the variables.

Each cell contains the value of the variable for the observation in question:

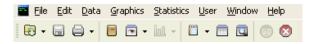
| Preserve Restore idcode idcode 2 1 3 1 | idcode year 70 71 | (() [1] = 1 birth_yr 51 | Hide | Dele | msp | nev_mar | grade | 114 | |
|--|----------------------------|-----------------------------------|------|------|-----|---------|-------|----------|------------|
| 1 1 2 1 | year 70 | birth_yr 51 | - | race | msp | nev mar | onade | 114 | |
| 1 1 2 1 | 70 | 51 | - | race | msp | nev mar | anade | 11 | |
| 2 1 | | | 18 | | | | graue | collgrad | not_smsa 🔨 |
| | 71 | | 10 | 2 | 0 | 1 | 12 | 0 | 0 🔳 |
| 2 1 | | 51 | 19 | 2 | 1 | 0 | 12 | 0 | 0 |
| , , | 72 | 51 | 20 | 2 | 1 | 0 | 12 | 0 | 0 |
| 4 1 | 73 | 51 | 21 | 2 | 1 | 0 | 12 | 0 | 0 |
| 5 1 | 75 | 51 | 23 | 2 | 1 | 0 | 12 | 0 | 0 |
| 6 1 | 77 | 51 | 25 | 2 | 0 | 0 | 12 | 0 | 0 |
| 7 1 | 78 | 51 | 26 | 2 | 0 | 0 | 12 | 0 | 0 |
| 8 1 | 80 | 51 | 28 | 2 | 0 | 0 | 12 | 0 | 0 |
| 9 1 | 83 | 51 | 31 | 2 | 0 | 0 | 12 | 0 | 0 |
| 10 1 | 85 | 51 | 33 | 2 | 0 | 0 | 12 | 0 | 0 |
| 11 1 | 87 | 51 | 35 | 2 | 0 | 0 | 12 | 0 | 0_ |
| 12 1 | 88 | 51 | 37 | 7 | 0 | 0 | 12 | 0 | > |

Main windows

- <u>Results</u> (black) text output of the commands you run
- <u>Command</u> (blue) allows to enter commands to run
- Review (red) shows previously run commands
- <u>Variables</u> (green) shows the variables in the loaded dataset



Menu and Upper bar



- • Open a data file
- 🗏 Save dataset in memory to a file on disk
- — Open data editor (left), or data browser (right), for dataset in memory
- □ Start a new do-file in a new do-file editor window
- Stop execution of a command

From the menu you can easily call forms that will run some commands for you. For example: "Data \rightarrow Describe Data \rightarrow Summary Statistics" will open a form and then run the summarize command accordingly.

Commands and Syntax Conventions

Commands in Stata usually take the following form:

```
<command name> [... something ...] [if] [in] [, options]
```

A few conventions:

- Angle brackets $\langle \rangle$ mean that you *must* put something in their place. In other words, they are *mandatory*
- Square brackets [] mean that you may put something in their place. In other words, they are optional
- General syntax will be in blue color, specific examples in green

For example:

```
use mydataset.dta, clear
drop if male==1
save mydataset_females.dta, replace
```

Use and Save

Stata works with a single dataset in memory. It can work with multiple files, but not at the same time. You will need to load the dataset to the internal memory from the disk and you can save it back to the disk after you are done.

To load a dataset into the memory we run the use command:

```
use <file path> [, clear]
```

- file path required path to the dta file you want to load.
- <u>clear</u> Ignore existing dataset in memory, even if unsaved.

Use and Save

Now, after messing with the file, we might want to save it on file for later use.

```
save <file path> [, replace]
```

- file path required path to the dta file you want to save.
- replace If a filename by that name in this folder exists, replace it with the dataset in memory.

Notes:

- The clear and replace options will appear in the future and will have the same use: clear will overwrite the current dataset in memory and replace will overwrite the file on disk.
- If you will use the icons in the upper bar for loading and saving datasets, Stata will actually run the the use and save commands.

Use and Save

Examples:

```
use "C:\Documents and Settings\RM\Papers\RawData.dta", clear
// Alternatively...
cd "C:\Documents and Settings\RM\Papers"
use RawData, clear
// (... do some stuff ...)
// Now save it to the "data" subfolder
save data\GoodData, replace
```

Sniffing Around - What's in the Data

So we loaded a dataset and we want to learn more about what's there. Here are a few commands worthy of notice:

- describe Lists the variables names, labels and formats
- <u>list</u> Lists the observations' matrix: each observation with its values for each of the variables.
- <u>browse</u> Like list, but opens up a window and is more convenient
- <u>tabulate</u> Reports a histogram of a variable or joint histogram of two variables.
- <u>summarize</u> For each variable requested, reports the number of observations with non-missing values, the mean, standard deviation, and other summary statistics.

Examples and their Output

describe idcode year birth_yr

| variable name | storage type | display format | value label | variable label |
|---------------|-----------------|-------------------|----------------|----------------|
| idcode | int | %8.0g | | NLS ID |
| year | byte | %8.0g | | interview year |
| birth_yr | byte | %8.0g | | birth year |

tab mothers_brothers fathers_brothers

| mothers_br | | | fathers_ | brothers | | |
|------------|-------|-----|----------|----------|----|-------|
| others | 0 | 1 | 2 | 3 | 4 | Total |
| 0 | 132 | 138 | 74 | 24 | 7 | 375 |
| 1 | 113 | 131 | 69 | 26 | 9 | 348 |
| 2 | 72 | 87 | 37 | 7 | 2 | 205 |
| 3 | 21 | 18 | 12 | 4 | 1 | 56 |
| 4 | 7 | 5 | 2 | 2 | 0 | 16 |
| Total | l 345 | 379 | 194 | 63 | 19 | 1,000 |

Examples and their Output

su south race age

| Variable | • | Obs M | | Dev. M | Max |
|----------|------|------------|-----------|--------|-------|
| south | | 3526 .4095 | | 7605 | 0 1 |
| race | 1 28 | 3534 1.303 | 3392 .482 | 2773 | 1 3 |
| age | 1 28 | 3510 29.04 | 511 6.70 | 0584 | 14 46 |

su tenure, detail

job tenure, in years

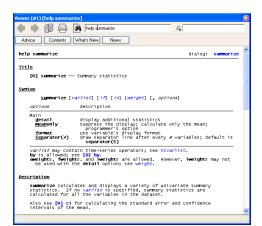
| | Percentiles | Smallest | | |
|-----|-------------|----------|-------------|----------|
| 1% | 0 | 0 | | |
| 5% | .0833333 | 0 | | |
| 10% | .1666667 | 0 | Obs | 28101 |
| 25% | .5 | 0 | Sum of Wgt. | 28101 |
| | | | | |
| 50% | 1.666667 | | Mean | 3.123836 |
| | | Largest | Std. Dev. | 3.751409 |
| 75% | 4.166667 | 23.33333 | | |
| 90% | 8.416667 | 24.5 | Variance | 14.07307 |
| 95% | 11.41667 | 24.75 | Skewness | 1.939685 |
| 99% | 16.91667 | 25.91667 | Kurtosis | 6.901501 |
| | | | | |

Note: su is short for summarize

Commands' Help Files

Each command should come with an accompanying help file. To learn more about additional options, other features, or to troubleshoot, the first place to look at is the help file:

help <command-name>



Commands' Help Files

Main parts of a usual help file (by the order I usually read them):

- Syntax How to specify your command
- Description Tells you what the command does generally
- Examples (at the bottom) Shows you specific examples of how to run the command. Sometimes with an explanation.
- Options The same command with different options can do totally different things. Skim through the options and look for the good ones.

Tips:

- Don't be afraid to experiment. Your data is saved on file, so you can always load it back if you made a mistake.
- Error messages looks scary, but don't let them fail you. READ them and try to understand them.
 - Remember: Errors don't mean that you are stupid, they mean that Stata is stupid.

Basic Data Manipulation - generate

To add a new variable we use the gen (short for generate) command:

```
gen <new-variable-name> = <expression> [if] [in]
```

For example:

```
gen four = 4
```

will create a variable (=column) that will contain the number 4 for all observations (=rows).

```
gen age_sq = age^2
```

will create a variable that will contain the square of the value in the age variable for the same row.

Conditions in Stata

Sometimes we want to apply a command only to some observations, not all.

We need to tell Stata what distinguishes those observations, so we construct a logical condition:

- Stata evaluates the condition and turns it to 1, if the statement is true, or 0 if the statement is false.
- For example, since (4 > 60) is not true, Stata will treat the expression (4 > 60) as if it was 0.
- The other conditions involve variable names. They will be invoked as part of a command. Stata will apply the command only to observations for which the values inside the specified variables make the statement true.
 - For example, this is how we ask Stata to run summarize on females only:

Conditions in Stata

Note that some of the observations - those for which male == 1 - will not be processed by the summarize command.

We can combine multiple conditions with AND, OR and NOT operators:

```
(male == 1 & age >= 21) | (male == 0 & age >= 50)
```

will be true for males aged 21 and above or females aged 50 and above.

Adding the ! operator before a condition will negate it:

```
!(age >= 21)
```

will be true for people strictly younger than 21.

Conditions in Stata

Lastly, do not forget operators precedence:

$$6+5\times2 \neq (6+5)\times2$$

The same goes for | and & (like + and \times respectively):

will be true for males older than 21 and all children under 10.

will be true for males only that are either older than 21 or younger than 10.

Note: missing values (.) are bigger than any value:

For a complete list of logical as well as other operators, see help operator

Back to Data Manipulation

gen can also take a condition. Observations for which the condition is false will have a missing value in the new variable:

```
gen age_sq_males = age^2 if male == 1
```

age_sq_males will contain the square of age for males and a missing value for females.

Note the difference between the assignment = and the comparison ==

One can also use the evaluation of a condition as the value to put into the new variable:

```
gen really_old = age > 22
```

Remember, the expression age > 22 will be translated to either 1 or 0 according to whether age is bigger than 22 or not.

Back to Data Manipulation

Question: Which of the next three commands is best?

```
gen dropout = 1 if schooling < 12
gen dropout = schooling < 12
gen dropout = schooling < 12 if schooling != .</pre>
```

All commands will make those with schooling < 12 have the value 1 in dropout. But what about the other ones?

- First line will assign missing values to all other observations
- Second line will assign zeroes to all other observations
- Third line will assign missing values to all observations that have a missing value in schooling and zeroes to the rest

So you will probably want to use the third line rather than the first two.

Meet replace, gen's sister

Just like gen, but for existing variables instead of new ones, use replace (other statistical packages such as SAS don't even have this distinction between generating and replacing):

```
replace <variable-name> = <expression> [if] [in]
```

For example:

```
replace four = 5 will change the values in the variable four to now be 5.
```

```
gen actual_price = discount_price if discount == 1 replace actual_price = full_price if discount == 0 will first create a variable that will contain the value from discount_price for all observations in which discount contains 1, then replace puts the value of full_price into actual_price if discount contains 0.
```

Do-files

Until now we used the command window to type in commands and run them one-by-one. This was working interactively. What if you have many commands to run?

A .do file is a text file that contains a batch of commands each written as a separate line. This way you can save your commands for:

- later review, improvement and additional work
- collaborating with your colleagues they can continue what you started

Comments

In a do file you can also explain what you are doing by adding comments. Here are a few ways to write comments:

Long lines in do-files

As you noticed, each Stata command takes one line. Once you hit the return, or enter, key, Stata runs the command. This is also true for do-files.

But what if you have a really long command?

```
su income mot_educ fat_educ school age agesq south bigcity tenure comm
```

You can break the line with ///:

```
su income mot_educ fat_educ school age agesq south bigcity ///
tenure commute kids_u5 kids_18 tot_kids siblings ///
if male & professional
```

Another way is to write #delimit; at the beginning of the do file and then end each command with a semicolon (;):

```
#delimit;
su income mot_educ fat_educ school age agesq south bigcity
tenure commute kids_u5 kids_18 tot_kids siblings
if male & professional;
tab age;
```

Log files

One last file you can save your work to is your log file. Unlike the do file, a log file will also save the text output resulted by your commands.

Whatever appeared in the results (big black) window, from when a log file was opened until it was closed, will be saved to the requested log file.

A log file is used to see what your program have done. Unlike the do file that will be edited and improved by you, the log file is automatically created by your program.

```
// Open a log file
log using <log-file-name> [, append replace text]
/* ... */
// Close current log file
log close
```

Log files

- replace overwrite the file on disk if it already exists
- <u>append</u> add the output to the end of the existing file if one exists otherwise open a new one
- <u>text</u> save the output in text format. In some cases Stata's default is to save it in a text-like format of its own called SMCL.

Here's a tip:

If a log is already open (usually after the last run ended tragically with an error), opening a log will create yet another error

To solve that, add the following line right before the log using line:

```
cap log close
```

File Types Summary

| | data files | do files | log files |
|----------------------|----------------------------|--|---|
| What's in it? | Observations and variables | A batch of com- mands | Text output from your commands |
| How do you open it? | use command mainly | Any text file editor | log commands |
| What is it good for? | Saving your data | Saving sequence of actions on the data | Recording commands you ran and their output |
| File extensions | .dta | .do | .log |

What is a Macro?

A Macro is a string (= a sequence of characters) that we name and can refer to.

One type of a macro is the local macro (local = can not be referred to outside the program):

```
// Define local
local <macro name> = <expression>
// Refer to local
[...] '<macro name>'
```

Note the back-quote and quote signs: ' is the character usually on the upper left corner of the main part of your keyboard (where $\tilde{\ }$ is). ' is the usual single-quote sign you're using.

Globals

```
// Define the global and assign an expression to it
global <macro name> = <expression>

// Refer to the global
[...] ${<macro name>}

Example:
global a = 4

// Refer to the global
di "Four is ${a}" // will print: Four is 4
```

Strings and Macros

In Stata, we put string expressions between double-quotes. For example:

```
gen girl = 1 if sex == "female"
replace girl = 0 if sex == "male"
```

If we don't put double quotes in a string expression, Stata will look for a variable with that name. We need to tell Stata the word is a value instead of part of the command. This is why we need the double quotes.

Same goes for macros:

First loop: forvalues

Loops are lines of code that can be run more than one time. Each time is called an **iteration** and in **for** loops there is also an index that is changing each iteration (the index is actually a local).

In the case of forvalues, the index is incremented each iteration:

```
// Define the loop
forvalues <index name> = <starting value>/<ending value> {
    // Commands to run each iteration
    // ... more commands ...
}
```

The loop will put <starting value> into <index name>, then run the commands until it reaches the closing }. Then it will go back, increase the value of <index name> by one and run the commands again, until it is done with the commands for the <ending value>.

First loop: forvalues

For example:

```
forvalues i = 1/3 {
   di "Iteration #'i'"
Will print:
Iteration #1
Iteration #2
Iteration #3
forvalues i = 7(7)21 {
  replace age = 0 in 'i'
```

Will set the value of the variable age to 0 for observations 7, 14 and 21.

What is it good for? Part 1

Imagine you have three different specifications:

```
y_t = \beta_0 + \beta_1 x_t + \varepsilon_t
y_t = \beta_0 + \beta_1 x_t + \beta_2 age_t + \beta_3 agesq_t + \beta_4 educ_t + \eta_t
y_t = \beta_0 + \beta_1 x_t + \beta_2 age_t + \beta_3 agesq_t + \beta_4 educ_t + \beta_5 mo\_educ_t + \beta_6 fa\_educ_t + u_t
local spec1
local spec2 "age agesq educ"
local spec3 "'spec2' mo_educ fa_educ"
forvalues i = 1/3 {
    reg v x 'spec'i'
```

Still not convinced? You're right. This example, as it is now, is longer than just writing three lines of regressions. But hold on...

foreach

When you want to iterate on other lists - not just an arithmetic sequence of numbers - you will want to use foreach.

The simplest form to use foreach is:

```
// Commands to run each iteration
}
For example:
foreach i in 3 15 17 39 {
    di "I am number 'i'"
}
foreach dep_var in income consumption health_score {
    reg 'dep_var' educ age agesq
}
```

foreach <index name> in <list separated by space> {

Even though we didn't put double-quotes on the values, since they are inside a foreach loop with the in word, Stata knows to treat them as values

foreach

If you want to loop over values that have space in them, use the double-quotes:

```
foreach fullname in "Roy Mill" "John Doe" Elvis Presley Madonna {
   di "Hello 'fullname'"
}
```

Will print:

```
Hello Roy Mill
Hello John Doe
Hello Elvis
Hello Presley
Hello Madonna
```

foreach and variables lists

When you iterate over variables' names it's better to put of varlist instead of in:

```
foreach <index name> of varlist <varlist> {
    // Commands to run each iteration
}
```

This way:

- Stata will check that each element of the variables list is actually a variable (avoid typos)
- You will be able to use wildcards

```
foreach mother_vars of varlist mother_* {
    // This loop will go over all variables that begin with mother_
}
foreach setvar of varlist set?_score {
    // This loop will go over all variables that have one character
    // where the ? is. For example set1_score, set2_score, ...
    // (but not set14_score)
}
```

What is it good for? Part 2

Remember our three specifications? Now imagine we want to run them on three different samples: males, females and both. Here is one way to do that:

```
local spec1 ""
local spec2 "age agesq educ"
local spec3 "'spec2' mo_educ fa_educ"

foreach sampleCond in "if male == 1" "if male == 0" "" {
   forvalues i = 1/3 {
     reg y x 'spec'i'' 'sampleCond'
   }
}
```

What is it good for? Part 2

This loop is equivalent to running:

```
reg y x if male == 1
reg y x age agesq educ if male == 1
reg y x age agesq educ mo_educ fa_educ if male == 1
reg y x if male == 0
reg y x age agesq educ if male == 0
reg y x age agesq educ mo_educ fa_educ if male == 0
reg y x
reg y x age agesq educ
reg y x age agesq educ mo_educ fa_educ
```

Now imagine you want to change the standard errors to robust, or add another control variable to the second specification. How much work will you need for the loops version and how much for the this version? And wait until you will need to post the results to a table.

Getting values returned by commands

su union

| Variable | 0bs | Mean | Std. Dev. | Min | Max |
|----------|-------|----------|-----------|-----|-----|
| | | | | | |
| union | 19238 | .2344319 | .4236542 | 0 | 1 |

return list

scalars:

```
r(N) = 19238

r(sum_w) = 19238

r(mean) = .2344318536230377

r(Var) = .179482889232214

r(sd) = .4236542095060711

r(min) = 0

r(max) = 1

r(sum) = 4510
```

Getting values returned by commands

```
reg union age south c_city
 ereturn list
scalars:
                  e(N) = 19226
               e(df_m) = 3
               e(df_r) = 19222
                  e(F) = 159.8965961359796
[...]
               e(11.0) = -10764.76183026799
macros:
            e(cmdline) : "regress union age south c_city"
              e(title): "Linear regression"
                e(vce) : "ols"
[...]
          e(estat_cmd) : "regress_estat"
matrices:
                  e(b) : 1 \times 4
                  e(V) : 4 \times 4
```

Getting values returned by commands

```
// Show coefficients after reg
matrix list e(b)
e(b)[1,4]
              south c_city _cons
         age
v1 .00242889 -.11547906 .06972916 .18224179
// Show coefficients' variance-covariance matrix
matrix covmat = e(V)
matrix list covmat
symmetric covmat[4,4]
             age
                      south c_city
                                           _cons
  age 2.399e-07
south -9.321e-08 .00003756
c_city 2.739e-07 -6.147e-07 .00004085
cons -7.578e-06 -.00001244 -.00002226 .00025953
```

Extensions 1 - _variables

Besides the _b[] and _se[] we have other special variables that start with _:

• _n refers to the observation number:

- _cons refers to the constant term in a regression in the
 _b[] or _se[] context.
- _N contains the total number of observations in the dataset The relevant help file is help _variables.

Extensions 3 - while and if

 $\underline{\mathbf{if}}$ - up until now we used \mathbf{if} as an argument of commands, to let them know which observations to work on. \mathbf{if} can also be used to control the flow of the program - especially inside loops.

```
if <condition> {
     // Commands
}
[else if <condition> {
     // Commands
}]
[else {
     // Commands
}]
```

while - in addition to foreach and forvalues sometimes we don't know in advance how many iterations we will need. We just need to loop as long as some condition holds (for example, as long as we haven't reached convergence).

Extensions 3 - while and if

But be careful with whiles, because if the condition will not be satisfied, you will enter a never-ending loop.

Usually, it's preferable to use some maximum number of iterations, in case there is some probability the usual condition will not work:

```
local converged = 0
local iter = 0
local max_iter = 800
while (!'converged' & 'iter' < 'max_iter') {
    // Commands that do something and check whether convergence
    // was achieved. If convergence was achieved it does
    // local converged = 1
    local iter = 'iter' + 1
}</pre>
```

Extensions 3 - while and if

But then, if we're already counting iterations, we might as well do it all with forvalues:

```
local converged = 0
local max_iter = 800
forvalues iter = 1/'max_iter' {
    // Commands that do something and check whether convergence
    // was achieved. If convergence was achieved it does
    // local converged = 1
    if 'converged' {
        continue, break
    }
}
```

continue, without the break option, stops the execution of the *current* iteration and goes on to the next iteration. With the break option it exits the loop altogether.

egen

egen is a "super-command". It generates new variables and serves as an extension to the generate command.

Very useful in panel data and in any other hierarichal data:

- Student-level data with class-, school- and/or city-level variables.
- Any other individual level data with some observations grouped by some identifier.
 - But uses extend to non-group-related tasks too.

egen - Syntax

The syntax is generally pretty simple:

```
egen <new varname> = <function>(<expression>) [, ... by (<varlist>)]
```

Another way to do the same thing:

```
bysort <variables>: egen <new variable> = <function>(<expression>) [, ... ]
```

The function we specify in <function> will determine what egen will do.
 Each function is like a different command even though they all begin with egen

We will now go over main functions.

"Vertical" egen functions - mean()

To create a variable containing the mean of another variable we can do:

```
summarize gpa
gen meangpa = r(mean)
```

But if you want to create a variable containing the mean of another within the group of each observation, it will be much harder without egen.

```
egen meangpainyear = mean(gpa), by(year)
```

Example: your dataset is such that you have both year and cohort and you want to get the GPAs demeaned of the cohort-year mean GPA (for the class of 2012 in year 2010):

```
egen mean_gpa_in_cohort_year = mean(gpa), by(year cohort)
gen gpa_demeaned = gpa - mean_gpa_in_cohort_year
drop mean_gpa_in_cohort_year
```

```
"Vertical" egen functions - sum(), min(), max()
```

A function that works the same way but gives you the sum instead of the mean is called...

```
egen total_tax_in_county_year = sum(tax), by(state county year)
```

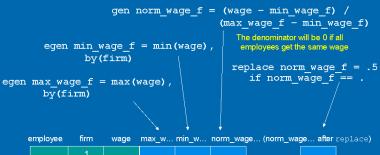
And when you want the minimum or maximum within a group:

```
// For a dataset with children that can be grouped to families.
egen youngest_sibling_age = min(age), by(familyid)

// For a dataset of basketball statistics per team, player, game.
egen highest_score_player = max(points), by(playerid)
egen highest_score_team = max(points), by(teamid)
egen highest_score_player_team = max(points), by(playerid teamid)
```

Example: Transferring a variable to the [0,1] range

Using a within-group maximum and minimum



| | 1 | | | | |
|----|---|--------|--------|--------|------|
| 5 | 2 | 10,000 | 15,000 | 7,000 | .375 |
| 6 | 2 | 7,000 | 15,000 | 7,000 | 0 |
| 7 | 2 | 15,000 | 15,000 | 7,000 | 1 |
| 8 | 3 | 10,000 | 10,000 | 10,000 | 100 |
| 9 | 3 | 10,000 | 10,000 | 10,000 | 100 |
| 10 | 3 | 10,000 | 10,000 | 10,000 | 100 |
| | 4 | | | | |

| .375 | | |
|------|--|--|
| 0 | | |
| 1 | | |
| .5 | | |
| .5 | | |
| .5 | | |
| | | |

"Vertical" egen functions - count()

count() will put the number of **nonmissing** values in the variable.

```
egen studentsinclass = count(studentid), by(school grade class)
```

If you're interested in counting the number of observations, regardless of missing values, try to count _cons or _n. Every observation has _cons==1 and _n is the obs' number.

```
egen studentsinclass = count(_n), by(school grade class)
```

"Horizontal" egen functions

We sometimes want to do the sum, mean, count, min and max across variables for each observation, rather than across observations for each variable.

```
egen hours = rowtotal(hoursday hoursnight)

// suppose each is a judge score
egen disagreement = rowsd(evaluation1 evaluation2 evaluation3)

// suppose each is dummy for attendance at day
egen full_attendance = rowmin(mon tue wed thu fri)

// suppose each has gas quality or missing value
egen sampled_pumps = rownonmiss(leaded unleaded premium)
```

Two reasons for preferring egen rowtotal() and egen rowmean() over the simple gen with the respective formula:

- egen ignores missing values. If you specify two or more variables and some
 of them are missing, the sum or mean will be calculated only for the
 nonmissing values.
- egen can get varlists for example: evaluation_* or mon-fri.

reshape

Suppose you have observations in a two-dimensional dataset. For example, "panel" data with state and year. Alternatively, think about a survey per household with recurring questions for each of the household members.

Some of the variables $-X_i$ are common to all observations of the same group i (state area in the panel data, household income in the survey). Others $-X_{ij}$ – are changing with members j within the group i.

Two ways to structure a dataset matrix:

| Form | Each obs is | Member-level variables (X_{ij}) |
|------|-----------------------|-----------------------------------|
| Wide | Group (i) | Appear $\max(j)$ times |
| Long | Group-member (i, j) | Appear just once |

reshape

Wide form:

| i | X_{ij} | | | |
|--------|------------------------|------------------------|------------------------|------------------------|
| fam_id | ${ m kid}_{ m educ}1$ | ${ m kid}_{ m educ}2$ | ${ m kid}_{ m educ}3$ | ${ m kid}_{ m educ}4$ |
| A | 8 | 6 | | |
| В | 3 | | | |
| С | 14 | 10 | 8 | 6 |

| i | j | X_{ij} |
|--------|--------|----------|
| fam_id | kid_id | kid_educ |
| A | 1 | 8 |
| A | 2 | 6 |
| В | 1 | 3 |
| С | 1 | 14 |
| С | 2 | 10 |
| С | 3 | 8 |
| С | 4 | 6 |

Long form:

reshape

Panel commands usually work with long forms. Wide forms are ugly and inefficient. However, you sometimes get your data in wide form. Especially if it's a questionnaire dataset

reshape allows you to go from wide to long form or the other way around. The simple syntax:

```
reshape <long|wide> <stubnames>, i(<group-identifying-vars>) [j(<member-identif
```

Where stubname is the part of the variable that is not changing between members. In our case: kid_educ.

Examples:

```
// From wide to long
reshape long kideduc, i(famid) j(kidid)

// From long to wide
reshape wide kideduc, i(famid) j(kidid)
```