# Introduction to Computing for Sociologists Neustadtl

## Stata "Gotchas"

Stata has a couple of issues that can bite you if you are not aware of them—the treatment of missing values and the precision of the data stored in variables and comparisons.

### **Missing Values**

So far we have treated a period (".") in a numeric variable as a missing value that leads that case to be excluded from calculations and statistical analyses. Stata actually has twenty-seven numeric missing values—the period is just one, but the one most people use. The missing values .a, .b, .c, ..., .z, are called "extended missing values". Extended missing values are used when it is necessary to differentiate between different types of missing data. For example survey questions sometimes allow for the following responses in addition to simply missing: "Not Applicable", "Don't Know", "Refused to Answer". These different reasons for missing data may be coded using .a, .b, and .c in case someone wanted to do an analysis of the missing responses.

Stata numeric missing values are represented by *large positive values*. The ordering is all nonmissing numbers < . < .a < .b < ... < .z. The expression *sociability* > 10 is true if variable *sociability* is greater than 10 or missing. To exclude missing data you must specify that the value is less than ".". For instance, "list if sociability > 10 & sociability < .".

Stata has one string missing value, which is denoted by "" (blank).

The following Stata code creates a small dataset to illustrate how missing values work. Case four is missing on the numeric variable *y* and case five is missing on the string variable *id*:



The following Stata code creates four new variables using different methods:

gen byte  $y_1=(y>=3)$  /\* Missing values are set equal to 1 \*/ gen byte  $y_2=(y>=3 \& y<.)$  /\* Missing values are set equal to 0 \*/ generate  $y_3=.$ replace  $y_3=0$  if y<3replace  $y_3=1$  if y>=3 & y<.generate byte  $y_4=cond(y<3, 0, cond(y>=3 \& y<., 1, .))$ list, noobs . gen byte  $y_1=(y>=3)$  /\* Missing values are set equal to 1 \*/ . gen byte  $y_2=(y>=3 \& y<.)$  /\* Missing values are set equal to 0 \*/

```
. gen byte y2=(y>=3 & y<.) /* Missing Values are set equal to (</li>
. generate y3=.
(5 missing values generated)
. replace y3=0 if y<3</li>
(2 real changes made)
. replace y3=1 if y>=3 & y<.</li>
(2 real changes made)
. generate byte y4=cond(y<3,0,cond(y>=3 & y<.,1,.))</li>
(1 missing value generated)
. list, noobs
```

i d	У	у1	у2	у3	у4
а	1	0	0	0	0
b	2	0	0	0	0
c	3	1	1	1	1
d		1	0		
	5	1	1	1	1

The variable yl demonstrates a significant issue with missing values. This was an attempt to create a dummy variable where all values greater than or equal to 3 are set equal to one; all other values are equal to 0. Stata dutifully did this task and converted the missing value to be equal to 1 since it too is greater than or equal to 3.

The second variable y2 takes this into account. Now, values greater than or equal to 3 and less than . (missing) are equal to 1. But, the value for case four is equal to 0! How did this happen? The Stata statement (y>=4 & y<.) is a Boolean or logical expression that evaluates to either false (0) or true (1). So, you can see that the missing value in case 4 becomes a 0 using this logic.

Look at the Stata code that created  $y_3$  to see one way to handle missing values and logical conditions. First a new variable  $y_3$  is created where every value is equal to missing (.). Then these values are selectively replaced to be equal to 0 (if y is less than 3) or 1 (if y is greater than or equal to three and y is less than missing).

Three lines of Stata code are inefficient and can be reduced to a single line of code by using the cond function (help cond) and logical expressions. The general form of this function is:

cond(x, a, b)

So, for example, cond(y<3, 0, 1) has the same result as method one (y1) discussed above since x=y, a=0, and b=1. The cond() function evaluates x and if x is true assigned the value a. If x is

false the value b is assigned. But, functions can be nested and here the third argument b is replaced with another cond function.

The following example creates a new variable *y*<sup>5</sup> that is missing for both cases 4 and 5. However, the missing values are differentiated with case 4 equal to the "normal" missing period (.) and case 5 missing is equal to .a.



Three different I i st commands are issued. The first simply lists all values of y5 including the missing values. The second lists all non-missing values of y5. This example uses the missing() function which returns 1 (true) if any of the function arguments are missing and 0 (false) if none of the arguments are missing. This function is combined with the "not" operator ("! "). The mi ssi ng() function [hel p mi ssi ng()] is used to limit the listing results note that the Stata code ! mi ssi ng(y5) is read as "not missing on y5" (hel p operator). The last example lists all non-missing values of y5 and values equal to missing (". ").

- 1. Recode or replace the missing values for the individual sociability variables *socrel, socommun, socfrend,* and *socbar* respectively to the missing values . a, . b, . C, and . d. List out the values for these variables for the records 23,414 to 23,425 and describe the pattern of missing values.
- 2. Describe the missing value pattern of the sociability variables *socrel, socommun, socfrend,* and *socbar*. This is a little complicated but follows these general steps:
  - a. Recode each of the individual sociability variables to be equal to 0 if the case has a valid value (1 through 7) and 1 if it is missing. These variables should be called *socre1l*, *socommun1*, *socfrend1*, and *socbar1*.
  - b. Create a variable equal to the sum of these four component measures called *socmiss*. Values equal to 0 mean no missing values on the four component measures; values equal to ` mean missing on one and only one variables, and so on.
  - c. Use tabul ate and graph hist for all values great than 0 to describe the pattern of missing values.

### Precision

Programs like Stata store results in binary, and to the right of the decimal point, there is often not an exact equivalent between decimal and binary given a finite number of digits. For 0.5 there is an exact equivalent: 0.1 base 2. For 0.25 there is an exact equivalent: 0.01 base 2. For 0.125 there is an exact equivalent: 0.001 base 2. This means to the right of the binary point the powers are  $2^{-1}$ ,  $2^{-2}$ , and so on.

0.1 base 2 is 
$$1 \cdot 2^{-1} = \frac{1}{2}$$
. 0.01 base 2 is  $1 \cdot 2^{-2} = \frac{1}{4}$ . 0.11 base 2 is  $\frac{1}{2} + \frac{1}{4} = \frac{3}{4}$ .

The difference in representing numbers in decimal (what we use) and binary (what computers use) can create some problems. The following program creates a dataset with 10 records and one variable, x, where every value is equal to 7.3. Note that the default storage type Stata used for this variable is float.

[	. clear				
clear set obs 10 gen x=7.3 desc list	. set obs 10 obs was 0, now 10 . gen x=7.3 . desc				
	Contains data           obs:         10           vars:         1           size:         80 (99.9% of memory free)				
	storage display value variable name type format label variable label				
	x float %9.0g				
	Sorted by: Note: dataset has changed since last saved				
	. list				
	x       1.     7.3       2.     7.3       3.     7.3       4.     7.3       5.     7.3       6.     7.3       7.     7.3       8.     7.3       9.     7.3       10.     7.3				

Data stored as floats have about 7 digits of accuracy. Thus, 1234567 can be stored perfectly as a float, as can 1234567e+20. The number 123456789, however, would be rounded to 123456792.

Now, let's try to list some specific examples to demonstrate the problem:



When the command Iist if x=7.3 was issued no values of x satisfied the equality! The float() function rounds its argument (7.3 in this example) to float accuracy. Now, the two values are equal (within rounding) and the I ist command does as we expect.

The following Stata code displays three values to 14 decimal places: 1) the contents of variable x for the first record, 2) the number of 7.3 (as a double), and 3) the number of 7.3 represented in float accuracy.

	. di %16.14f x <b>7.30000019073486</b>
di %16.14f x di %16.14f 7.3 di %16.14f float(7.3)	. di %16.14f 7.3 <b>7.3000000000000</b>
	. di %16.14f float(7.3) <b>7.30000019073486</b>

Note that the first and second numbers are not equal (so the l i st command fails). But, the first and third numbers are identical (so the l i st command works). Another solution is to store your data as double (more precision than float), but this typically just wastes memory by increasing the storage requirements of your data. In a simple test a dataset that stored numbers as double was 1.8 times larger than the same numbers stored as floats.

3. There are no problems for you to complete, just be careful with precision issues.

## **Big** Numbers

The precision issue can also bite you when you have large numbers. Large numbers are often used as case identifiers (ID's). Consider the following:

. list, noobs			. format id %9.0f
clear inputid 123456789	id		. list, noobs
123456790	1. 23e+08	format id %9.0f list, noobs	id
123456791	1.23e+08 1.23e+08		123456792 123456792
123456793 123456794	1. 23e+08 1. 23e+08		123456792 123456792
123456795 123456796	1.23e+08		123456792
end List poobs	1. 23e+08		123456792 123456792
			123456800

First we read in some data and display them. The values are difficult to read because they are displayed in scientific notation. We can use the format command to tell Stata that we would like it to display the values with 9 values before the decimal place and with no values after the decimal place. Now we can see that the values were not stored they way we anticipated. The problem is that *id* is stored as float which can only store an integer value with up to 7 digits of accuracy (the *id* values are 9 digits).

There are two solutions 1) change the data storage type to long or double, or 2) store your identification numbers as strings (if they have no numerical importance). If identification numbers are integers and take 9 digits or less, store them as longs; otherwise, store them as doubles (doubles have 16 digits of accuracy). The following Stata code demonstrates both of these solutions:

. list, noobs		. list, noobs	
123456789		input str9 id	
123456790	id	123456789 123456790	id
123456791 123456792 123456793 123456794 123456795 123456796 end format id %9.0f list, noobs	123456789 123456790 123456791 123456792 123456793	123456791 123456792 123456793 123456794 123456795 123456796	123456789 123456790 123456791 123456792 123456793
	123456794 123456795 123456796	end list, noobs	123456794 123456795 123456796

4. There are no problems for you to complete, just be careful with precision issues and large numbers.