Basic Dataset Exploration & File Preparation

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You should get to know a dataset before you begin heavily analyzing it. Variable coding may be problematic, e.g. response categories may not be properly ordered; or, a variable may be coded 1/2 when you want it coded 0/1. Missing data may not be handled correctly, e.g. missing data codes may be being treated as legitimate values; or it may not be clear to you why data are missing in the first place.

Read the Dataset Documentation

At least skim through the study's documentation. It hopefully explains why it is a good dataset, and you'll want to note that if you use the data yourself. It may also tell you about any weaknesses in the data. There may be descriptions of additional variables that the collectors of the data tacked on at the end of the dataset; for example, a computed SES scale may be something you want to use yourself. The correct way to weight the data is often described. The questionnaire can help you be clear on exactly how questions were worded and how the skip patterns work; it can be frustrating if half the data are missing on a variable and you don't know why (e.g. maybe only married people were asked the question).

Document Your Own Work

- Be sure to document any manipulations you make to the dataset. I've had people come to me with scales they have computed, and they've lost the code that showed how the calculations were done. Or, they have a variable like gender that is coded 0/1, but they can't tell me if 1 = male or 1 = female. I often give my programs names like run01.do, run02.do, etc. so I know and can reproduce the order in which I did things.
- Related to this, you may want to create new variables rather than manipulate existing ones. You'll often want clearer variable names than are used in the data set anyway. Also, if you use the original name for a variable, but have substantially recoded it, others may get confused when they try to replicate your work.
- Including comments in your code may be very helpful. You may understand why you did something now, but six months from now it could be a total mystery.
- In any event, whether you create new variables or not, make sure you can get back to the values in the original dataset.
- Remember too that there is increasing emphasis on replicability of results. Someone may want your data some day, or want to know how you did something. If you can't show what you did it may undermine your credibility. I'm a big fan of Trenton Mize at Purdue partly because he does a great job of making it easy to replicate his results. https://www.trentonmize.com/research.

Use fre, not tab1, for frequencies

You'll no doubt want to run frequencies on many of your variables. For frequencies, I generally prefer Ben Jann's user-written fre, because it gives both the numeric code and the value label for the code (tabl only gives you one or the other). For example:

. webuse nhanes2f, clear . tabl race										
-> tabulation of	race									
Race	-	Percent	Cum.							
White Black	9,051 1,086 200	87.56 10.51 1.93	98.07 100.00							
	10,337									
. fre race										
race Race										
	Freq.	. Percent	Valid	Cum.						

		Freq.	Percent	Valid	Cum.
Valid	1 White 2 Black 3 Other Total	9051 1086 200 10337	87.56 10.51 1.93 100.00	87.56 10.51 1.93 100.00	87.56 98.07 100.00

To get fre, type

ssc install fre

More generally, to find user-written commands that may be of interest to you, you can use the findit command. With findit you can specify a specific program or package of programs (e.g. fre or spost13_ado) or a topic area you are interested in, e.g. user-written commands for dealing with collinearity.

findit fre
findit spost13_ado
findit collinearity

Include options with tab2

With two-way cross-tabulations, you can ask for additional output. In particular I like to get a chi-square statistic, which indicates how strongly the variables are related, and the row percentages. (Many other statistics are available.) In the following example the chi-square statistic shows that race and diabetes are related, i.e. the likelihood of having diabetes differs by race. More specifically, the row percentages show you that, while about 8% of the Black respondents have diabetes, only about 4.5% of the White and Other respondents do.

. webuse nhanes2f, clear . tab2 race diabetes, lrchi2 row									
-> tabulation of race by diabetes									
+	+ 								
	 frequency row percentage								
Race 1	Diabetes s Not diabe D		Total						
+ White 	8,645 95.54		9,049 100.00						
Black	1,000 92.08	86 7.92	1,086 100.00						
Other 	191 95.50	9 4.50	200 100.00						
 Total 	9,836 95.17	499 4.83	•						
Likelihood-r	atio chi2(2)	= 21.7902	Pr = 0.000						

Make sure missing data are being handled correctly

Before you do any extensive analysis with your data, you should make sure missing data is coded correctly. The Stata missing value codes are ., .a, .b, .c,..., .z (i.e. . and .a to .z). Even if you downloaded your data in Stata format, the missing data codes may not be correct. For example,

. fre varl										
var1										
			F1	req.	Percent	Valid	Cum.			
Valid	1 Strongly 2 Disagree 3 Agree 4 Strongly 97 Don't Kr 98 Refused 99 Not App Total	y Agree		54 75 29 42 8 5 11 224	33.48 12.95 18.75 3.57 2.23 4.91	33.48 12.95	57.59 70.54 89.29 92.86 95.09			
. sum	var1									
	riable 			Sto	d. dev.	Min	Max			
	var1	224			.72513	1	99			

. use https://www3.nd.edu/~rwilliam/statafiles/fixcoding, clear

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The values 97, 98, and 99 are missing data codes. That might be correct coding for a program like SPSS, but in Stata those codes are treated as legitimate values, which totally distorts statistics involving the variable, e.g. the mean and standard deviation are wrong here. OLS or logistic regression results could also be way off if you don't fix the MD coding.

The mvdecode command is one of the many ways to solve the problem (the recode command is another) :

```
. mvdecode var1, mv(97=.a \setminus 98 = .b \setminus 99=.c)
        var1: 24 missing values generated
. fre var1
var1
_____
                                   | Freq. Percent Valid
                                                                                         Cum.
------

      Valid
      1
      Strongly Disagree
      54
      24.11
      27.00
      27.00

      2
      Disagree
      75
      33.48
      37.50
      64.50

      3
      Agree
      29
      12.95
      14.50
      79.00

      4
      Strongly Agree
      42
      18.75
      21.00
      100.00

      Total
      200
      89.29
      100.00
      100.00

      Missing
      .a
      8
      3.57
      .b
      .c
      11
      4.91

      Total
      24
      10.71
      100.00
      100.00
      100.00
      100.00

_____
 . sum var1
    Variable | Obs Mean Std. dev. Min
                                                                                       Max
______
          var1 | 200 2.295 1.083441 1
                                                                                         4
```

Much better! Further, suppose var1 thru var20 are consecutive variables in the data set and are all coded the same way. We might then be able to say

mvdecode var1-var20, $mv(97=.a \ 98 = .b \ 99=.c)$

Or, better yet, suppose all variables in the data set use the same missing value codes. You could then say

mvdecode _all, $mv(97=.a \ 98 = .b \ 99=.c)$

If we want, we can also tidy up the value labels a bit. var1 uses a value label called agreement (using the same value label for several variables that share the same values is often convenient). We can get rid of the old labels and add the new with the commands

	de	fine agreement 97 ' fine agreement .a '				.c "Not Appl	icable", add
var1							
				-	Percent	Valid	Cum.
Valid	1	Strongly Disagree		54	24.11	27.00	27.00
	2	Disagree	1	75	33.48	37.50	64.50
	3	Agree		29	12.95	14.50	79.00
	4	Strongly Agree		42	18.75	21.00	100.00
	То	tal		200	89.29	100.00	
Missing	.a	Don't Know		8	3.57		
	.b	Refused		5	2.23		
	.c	Not Applicable		11	4.91		
	То	tal	1	24	10.71		
Total				224	100.00		

Other notes:

- Never just assume you did things right! Check things out before and after like I did.
- The missing data codes were pretty obvious in this case. Other times they won't be. Try to check the dataset documentation if you can.
- It is nice when every variable uses the same MD codes, but that doesn't have to be the case. For example, 99 may be a missing value for one variable and a valid value for another.
- Sometimes all missing data are just coded ., the system missing value. That is often fine, but at other times it is helpful to know why data are missing. If you use Stata's multiple imputation commands it is very important that you use different MD codes for different types of MD. Eventually, you may decide that different types of missing data will be treated differently in your analysis.
- See help mvdecode for more information and examples.
- Chuck Huber has a nice 2-minute video on "How to convert missing value codes to missing values". I prefer to directly write out code when I can, but sometimes the menudriven approach he shows is better or easier. See https://www.youtube.com/watch?v=6HV2773-dVM.

Convert String Variables to Numeric Variables

Sometimes variables are in string format rather than numeric format. (Further complicating things, when you are viewing the data it isn't always obvious whether a variable is string or numeric.) For example, a variable called sex might be coded 1 = female, 0 = male, which is a numeric coding. Or, it might be coded "female" and "male" which makes it a string variable, i.e. words are used instead of numbers to code values. Stata usually wants numeric variables in analyses, so string variables must be converted to numeric variables.

You can often use the encode command to solve this.

. use https://www3.nd.edu/~rwilliam/statafiles/fixcoding, clear

. fre genderstr

genderstr

			Freq.	Percent	Valid	Cum.
Valid	female male		116	51.79 48.21	51.79 48.21	51.79 100.00
	male Total		224	48.21	48.21	100.00
. des o	enderst					
Variabl		Storage	Displa	v Value		
nam	-	type	-	<u> </u>	Variak	ole label
genders	str	str6	%9s			

In the above, genderstr is a string variable with two values, "male" and "female." Using encode, we can easily create a numeric variable with the values correctly labeled (the value label will have the same name as the generated variable, in this case gender):

. encode genderstr, gen(gender)

. fre gender*

genders	str					
			-	Percent		
	female	: :	116	51.79	51.79	51.79
				48.21 100.00		100.00
gender						
				Percent		Cum.
	1 fema	ile	116	51.79	51.79	
				48.21 100.00		100.00
. des g	gender*					
Variabl nam				y Value Label	Variab.	le label
genders gender			%9s %8.0g	gender		

The variable gender is now numeric, with value label gender, and 1 =female, 2 =male.

Numeric codes are assigned alphabetically, and female comes before male so it got coded 1. But suppose what you really want is a variable coded 0 = male, 1 = female. You can do this by defining the value label before you generate the numerical variable:

. label define female 0 "male" 1 "female"

- . encode genderstr, gen(female)
- . fre gender female

gender					
		Freq.	Percent	Valid	Cum.
 Valid	1 female 2 male Total	116 108 224	51.79 48.21 100.00	51.79 48.21 100.00	51.79 100.00
female					
		Freq.	Percent	Valid	Cum.
Valid	0 male 1 female Total	108 116 224	48.21 51.79 100.00	48.21 51.79 100.00	48.21 100.00

Stata has some good resources describing how to do this:

<u>https://www.youtube.com/watch?v=Js_i3wI2-jY</u> <u>https://www.youtube.com/watch?v=ZRWHjdIZyxo</u> <u>https://www.stata.com/support/faqs/data-management/numeric-variables-input-as-string/</u>

You can also just read the help for the encode command. The destring command is also helpful in some cases.

Create Binary or Ordinal Variables from Continuous Measures

Sometimes you have continuous measures that you want to convert to binary or ordinal variables. For example, birthweight (a continuous variable) might be used to create a variable called lbw, where 1 = 1 ow birth weight, 0 = n ot low birth weight. Or, income (measured in thousands of dollars) might be used to create a variable where 0 = No Income, 1 =\$1 to \$4999, 2 =\$5000-\$14,999, ... 9 =\$100,000 and above. Or, you might just want to take a variable that is coded 1/2 and create a new variable that is coded 0/1. Chuck Huber shows how to do this at

https://www.youtube.com/watch?v=XWVaXN2KwmA

Huber illustrates how to do this via menus, but once you know how, it is probably easier to use the recode command directly.

Handling More Complicated Data Structures

There are several types of datasets that raise their own special issues. Panel data involve the same cases measured at multiple points in time, e.g. children interviewed at ages 8, 10, 12, 14, and 16. Multilevel data involves subjects nested within other structures, e.g. you might have a

sample of schools which in turn has a sample of students from within each school. Complex survey designs have cases that did not all have equal probability of selection. You might get a data set where multiple imputation of missing data has already been done.

To get these data ready to use, you might eventually have to use commands like xtset, svyset, miset, mi import, reshape, and others.

I won't try now to summarize all the issues involved with these sorts of data. The course has entire handouts dealing with these topics, and you may want to read ahead a bit to find out how to deal with them.

Other Useful Resources

The above are situations I have frequently encountered with students. There are many other common issues you might encounter when getting your data ready. Stata lists several resources for Data Management (as well as statistical analyses) at

https://www.stata.com/links/video-tutorials/

https://www.stata.com/support/faqs/

Whenever you use a Stata video or FAQ, I encourage you to read the help files for the commands shown, as they may have additional useful capabilities not otherwise shown in the Video or FAQ.