Data validation

LPO 9951 | Fall 2020

Data validation refers to the process of ensuring that the characteristics of your data match the known characteristics of the population as measured by other analysts. If you have large discrepancies between your estimates and the estimates compiled by others, this is a clear "red flag" that something has gone wrong. Usually this is a problem that can be solved by going back to cleaning the data, but sometimes your sample may diverge in important ways from the samples collected by others. You will ne ed to state why this is the case in your write-up of the data.

Data validation can be done in several ways:

- You can compare the estimates from your dataset with the estimates from another analysis of the same dataset. This is what we will do with the datasets used in this class.
- Sometimes you will be the first one to analyze your dataset. In this case, you need to look for others who have collected similar samples and compare with them.
- Sometimes you won't have any other samples to work with. In this case, you'll need to see if there are population data that might be useful. Many people use the Census as a "check" on the data they have collected.
- Last, you need to use common sense. If you have data on private elite institutions of higher education, and you calculate an average tuition of \$2,000, you can rest assured that you have not found a hidden bargain but rather a flaw in your data.

```
. capture log close // closes any logs, should they be open

. set linesize 90

. log using "validation.log", replace // open new log

name: <unnamed>
log: /Users/doylewr/lpo_prac/lessons/s1-08-validation/validation.log
log type: text
opened on: 21 Oct 2020, 11:20:03

// clear memory
```

Calculating estimates and comparing them with known results

Today, we'll use the plans dataset. We're going to compare our results with several tables published by NCES. Let's start with educational expectations of high school sophomores. We start by survey setting the data:

```
. use ${ddir}plans.dta

. svyset psu [pw = bystuwt], str(strat_id) singleunit(scaled)

    pweight: bystuwt
         VCE: linearized
Single unit: scaled
    Strata 1: strat_id
        SU 1: psu
        FPC 1: <zero>
```

Account for missing data

The next step is to account for missing data properly:

```
. local allvar bystexp bysex byrace byses1 f1psepln
. mvdecode `allvar', mv(-9/-2)
    bystexp: 924 missing values generated
      bysex: 819 missing values generated
     byrace: 924 missing values generated
     byses1: 924 missing values generated
   f1psepln: 1958 missing values generated
. recode bystexp (-1=8)
(bystexp: 1450 changes made)
. label define expect 1 "Less than HS" ///
. label values bystexp expect
. label define race 1 "American Indian/AK Native" ///
                                   2 "Asian/PI" ///
                                   3 "African American/Black" ///
                                   4 "Hispanic No Race Specified" ///
                                   5 "Hispanic, Race Specified" ///
                                   6 "Multiracial, non Hispanic" ///
                                   7 "White"
```

2 '

. label values byrace race

Get estimates

Next, we tabulate expectations for college and compare it to a known estimate.

. tab bystexp

how far in |
school |
student |
thinks will |
get-composit |

е	1	Freq.	Percent	Cum.
Less than HS	-+- 	128	0.84	0.84
HS/GED	1	983	6.45	7.29
2 Yr		879	5.77	13.06
Attend 4	1	561	3.68	16.74
BA Degree		5,416	35.55	52.29
Master's		3,153	20.69	72.99
PhD		2,666	17.50	90.48
Don't Know'	1	1,450	9.52	100.00
Total		15,236	100.00	

. svy: proportion bystexp

(running proportion on estimation sample)

Survey: Proportion estimation

```
      Number of strata =
      361
      Number of obs =
      16,160

      Number of PSUs =
      751
      Population size =
      3,408,319

      Design df =
      390
```

```
_prop_1: bystexp = Less than HS
_prop_2: bystexp = HS/GED
_prop_3: bystexp = 2 Yr
_prop_4: bystexp = Attend 4
_prop_5: bystexp = BA Degree
_prop_6: bystexp = Master's
_prop_8: bystexp = Don't Know'
```

	 Proportion	Logit [95% Conf. Interval]	
bystexp			
_prop_1	.0094831	.00098	.007738 .0116172
_prop_2 _prop_3 _prop_4	.0724693	.0030538	.0666899 .0787074
	.0643949	.0028925	.0589365 .0703211
	.0389852	.0018459	.0355139 .0427808
_prop_5	.3578959	.0046507	.3488048 .3670902
_prop_6	.1971035	.004424	.1885502 .2059464
PhD	.1608805	.0039873	.1531947 .1688749
_prop_8	.0987875	.0030196	.0930076 .1048851

Once you create estimates from a command like proportion you can save them for later, using the estimates store command. These can be replayed using replay and can be brough back into memory using restore

```
. estimates store expect_tab
```

. estimates replay expect_tab

Model expect_tab ______

```
Survey: Proportion estimation
```

```
      Number of strata =
      361
      Number of obs =
      16,160

      Number of PSUs =
      751
      Population size =
      3,408,319

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```

```
_prop_1: bystexp = Less than HS
_prop_2: bystexp = HS/GED
_prop_3: bystexp = 2 Yr
_prop_4: bystexp = Attend 4
```

_prop_5: bystexp = BA Degree
_prop_6: bystexp = Master's

_prop_8: bystexp = Don't Know'

I		Linearized	Logi	Logit		
	Proportion	Std. Err.	[95% Conf.	<pre>Interval]</pre>		
bystexp						
_prop_1	.0094831	.00098	.007738	.0116172		

```
.0030538
                                      .0666899
                                                  .0787074
_prop_2 |
            .0724693
_prop_3 |
            .0643949
                       .0028925
                                      .0589365
                                                  .0703211
_prop_4 |
            .0389852
                       .0018459
                                      .0355139
                                                  .0427808
_prop_5 |
            .3578959
                       .0046507
                                      .3488048
                                                  .3670902
_prop_6 |
            .1971035
                        .004424
                                      .1885502
                                                  .2059464
            .1608805
                       .0039873
                                      .1531947
   PhD |
                                                  .1688749
            .0987875
                       .0030196
                                      .0930076
_prop_8 |
                                                  .1048851
```

. estimates restore expect_tab
(results expect_tab are active now)

Estimates can be stored using a simplified approach, using eststo and then the name of the estimates to be stored.

. eststo expect_tab: svy: tabulate bystexp
(running tabulate on estimation sample)

Number	of	strata	=	361	Number of obs	=	15,236
Number	of	PSUs	=	751	Population size	=	3,408,319
					Design df	=	390

how far | in school | student thinks will get-compo | site | proportion Less tha | .0095 HS/GED | .0725 .0644 2 Yr | Attend 4 | .039 BA Degre | .3579 Master's | .1971 PhD | .1609 Don't Kn | .0988 Total | 1

Key: proportion = cell proportion

Nicer tables

We get output in the console, but let's use the eststo and esttab commands to store our estimates and produce nicer tables. Using esttab alone, we'll get a nicely formatted table in the console. By adding ... us ing <file> we save an .rtf version of the same table. We can easily paste this table in a paper.

. estpost svy: tabulate bystexp

```
      (running tabulate on estimation sample)

      Number of strata = 361
      Number of obs = 15,236

      Number of PSUs = 751
      Population size = 3,408,319

      Design df = 390

      how far |
```

in school | student thinks will get-compo | site | proportion -----Less tha | .0095 HS/GED | .0725 2 Yr | .0644 Attend 4 | .039 .3579 BA Degre | Master's | .1971 PhD | .1609 Don't Kn | .0988 Total |

Key: proportion = cell proportion

saved vectors:

```
e(b) = cell proportions
e(se) = standard errors of cell proportions
e(lb) = lower 95% confidence bounds for cell proportions
e(ub) = upper 95% confidence bounds for cell proportions
e(deff) = deff for variances of cell proportions
e(deft) = deft for variances of cell proportions
e(cell) = cell proportions
e(count) = weighted counts
e(obs) = number of observations
```

```
. eststo expect_tab
. esttab expect_tab using expect_tab.rtf, /// b(3) /// /* 3 decimal position
(output written to `"expect_tab.rtf"')
```

Validate with published data

Now that we have a clean table to look at, is this the same as Table 2 on page 22 of the report? Yes. Checking the standard errors on page B-3 reveals that these were also correctly done. Now we need to check this for all of the other variables in our dataset.

Not-so-quick Exercise

I want you to replicate Table 34 on page 128 of NCES 2005-338. We'll split this up, but I want the class to come up with a single table that has exactly the same results as the NCES document.

```
. log close
    name: <unnamed>
    log: /Users/doylewr/lpo_prac/lessons/s1-08-validation/validation.log
log type: text
closed on: 21 Oct 2020, 11:20:03
```

. exit