

Variance Estimation for Survey-Weighted Data Using Bootstrap Resampling Methods: 2013 Methods-of-Payment Survey Questionnaire

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2017 STATA

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June 9, 2017

Outline

- Basic concepts
- Calibration estimator in survey: related to incomplete data estimator in econometrics
- Variance estimation of the calibration estimator
- Result

Basic Concepts

- Let U be a finite population of size N and the population total be $T_y = \sum_{i \in U} y_i$.
- A sample s is selected according to a sampling design $p(s)$ with inclusion probabilities, π_1, \dots, π_N .
- Calibration estimator:

$$\hat{T}_y = \sum_{i \in s} w_i y_i$$

where w_i is the calibrated weight.

We want to estimate $Var(\hat{T}_y)$.

Calibrated Weight

- Each unit i is assigned a **design weight**:

$$\{d_i = \pi_i^{-1}; i \in s\}.$$

- Calibration adjustment:** The weight d_i are adjusted to match known(imputed) population counts.

$$d_i \xrightarrow{\text{calibration}} w_i$$

where w_i from minimizing

$$\sum_{i \in s} d_i \left[\left(\frac{w_i}{d_i} \right) \log \left(\frac{w_i}{d_i} \right) - \left(\frac{w_i}{d_i} \right) + 1 \right]$$

subject to $\sum_{i \in s} w_i x_i = \mathbf{X}$.

Problem

Recall

$$\text{Var}(\hat{T}_y) = \text{Var} \left(\sum_{i \in s} w_i y_i \right)$$

where there are two sources of randomness in \hat{T}_y :

- 1 Sampling variation in the s ;
- 2 Sampling variation in the calibrated weight w_i .

However, people usually ignore the sampling variation in w_i , and this is problematic.

Problem (cont.)

- The estimated variance can be written as

$$\widehat{\text{Var}}(\widehat{T}_y) \approx \sum_{i,j \in s} \frac{d_{ij} - d_i d_j}{d_{ij}} \frac{y_i - \widehat{\beta} x_i}{d_i} \frac{y_j - \widehat{\beta} x_j}{d_j}$$

where $\widehat{\beta}$ is $[\sum_s x_i x_i']^{-1} [\sum_s y_i x_i']$.

- Compare to

$$\widehat{\text{Var}}^*(\widehat{T}_y) \approx \sum_{i,j \in s} \frac{w_{ij} - w_i w_j}{w_{ij}} \frac{y_i}{w_i} \frac{y_j}{w_j}$$

which ignores the sampling variation in w_i .

Bootstrap in STATA

- Bootstrap is easy to implement.
 - ① Use the *ipfraking* and *bsweights* commands in Stata (Kolenikov, 2010, 2014).
 - ② Generate replicate calibrated weights instead of recomputing the statistics for each resample.

weights

Replicate weights

Unit	y_1	y_2	...	y_p	ω	ω^{R1}	ω^{R2}	ω^{R3}
1	y_{11}	y_{21}	...	y_{p1}	ω_1	ω_1^{R1}	ω_1^{R2}	ω_1^{R3}
2	y_{12}	y_{22}	...	y_{p2}	ω_2	ω_2^{R1}	ω_2^{R2}	ω_2^{R3}
.
.
.
n	y_{1n}	y_{2n}	...	y_{pn}	ω_n	ω_n^{R1}	ω_n^{R2}	ω_n^{R3}

Implementation in Stata

- Step 1: Input the initial weight (d).
- Step 2: Generate the calibrated weight (w) using the *ipfraking* command.

Steps: $d_i \xrightarrow{\text{calibration}} w_i$

- Step 3: Generate the replicate raking weights using the *bsweights* command.
- Step 4: Declare the bootstrap survey environment in Stata:
`svyset [pw=], vce(bootstrap) bsrw()`

Result

	Ignoring w	Considering w
Cash on Hand		
Mean	1.03	1.03
Variance	1.51	0.85
Usage of CTC		
Mean	0.93	0.93
Variance	2.10	1.24

Note: The numbers in the second and third columns are divided by the numbers in the first column. CTC stands for the contactless feature of a credit card.