## **Domain Estimation**

Survey Sampling
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Often we want separate estimates for subpopulations, or **domains**.

Example: Want to estimate average salary for female lawyers, and male lawyers.

But sampling frame is just a list of lawyers.

Sample will include men and women, of course, but the resulting sample sizes are *random variables*, not fixed by the study design.

The appropriate estimate is fairly obvious — it's just what you think it should be. But standard error requires special care.

This is called **domain estimation**. It is a special case of **ratio estimation**, as we will demonstrate.

## **Domain estimation**

The set-up:

Suppose the population  $\mathcal{U}$  of N units has D domains:

$$\mathcal{U}_1 \cup \mathcal{U}_2 \cup \cdots \cup \mathcal{U}_D = \mathcal{U}$$

with  $N_d$  units in domain d, satisfying

$$N_1 + N_2 + \cdots + N_D = N$$

Any particular random sample  ${\mathcal S}$  from  ${\mathcal U}$ , of n units, can be similarly partitioned

$$S_1 \cup S_2 \cup \cdots \cup S_D = S$$

$$n_1 + n_2 + \dots + n_D = n$$

## **Estimator**

Estimate the domain d population mean

$$\bar{y}_{U_d} = \frac{1}{N_d} \sum_{i \in \mathcal{U}_d} y_i$$

by the domain d sample mean

$$\bar{y}_d = \frac{1}{n_d} \sum_{i \in \mathcal{S}_d} y_i$$

Note! Here  $n_d$  is a random variable.

That's what makes it ratio estimation.

Let

$$x_i = \begin{cases} 1 & i \in \mathcal{U}_d \\ 0 & i \notin \mathcal{U}_d \end{cases}$$

and

$$u_i = x_i y_i = \begin{cases} y_i & i \in \mathcal{U}_d \\ 0 & i \notin \mathcal{U}_d \end{cases}$$

Then

$$t_x = \sum_{i=1}^{N} x_i = N_d$$

and

$$\bar{x}_U = \frac{t_x}{N} = \frac{N_d}{N}$$

$$t_u = \sum_{i=1}^{N} u_i = t_{yd}$$

and

$$\bar{y}_{U_d} = \frac{t_u}{N_d} = \frac{t_u}{t_x} = B$$

and

$$\bar{x} = \frac{n_d}{n}$$

$$\bar{y}_d = \frac{\bar{u}}{\bar{x}} = \hat{B}$$

Thus

$$SE(\bar{y}_d) = SE(\hat{B}) = \frac{n}{n_d} \frac{s_e}{\sqrt{n}} \sqrt{1 - \frac{n}{N}}$$

where

$$s_e^2 = \frac{1}{n-1} \sum_{i \in \mathcal{S}} \left( u_i - \hat{B}x_i \right)^2 = \frac{1}{n-1} \sum_{i \in \mathcal{S}_d} \left( y_i - \hat{B} \right)^2 = \frac{(n_d - 1)s_{yd}^2}{n-1}$$

Thus we have

SE 
$$(\bar{y}_d) = \left\{ \frac{n}{n_d^2} \frac{(n_d - 1)s_{yd}^2}{n - 1} \left( 1 - \frac{n}{N} \right) \right\}^{1/2}$$

Interesting to note that

$$\mathsf{SE}\left(\bar{y}_d\right) = \left[rac{n(n_d-1)}{n_d(n-1)}
ight]^{1/2} imes (\mathsf{SE} \ \mathsf{for} \ \bar{y}_d \ \mathsf{based} \ \mathsf{on} \ \mathsf{SRS} \ \mathsf{of} \ n_d)$$

## **Estimating the domain total**

Wish to estimate  $t_u = t_{yd}$ .

Case 1: If  $N_d$  is known, estimate  $t_{yd}$  by  $N_d \bar{y}_d$ .

Standard error is  $N_d \times SE(\bar{y}_d)$ .

Case 2: If  $N_d$  is unknown, estimate  $t_{yd} = t_u$  by

$$\hat{t}_{yd} = \hat{t}_u = N\bar{u}$$

$$SE(\hat{t}_{yd}) = NSE(\bar{u}) = N \frac{s_u}{\sqrt{n}} \sqrt{1 - \frac{n}{N}}$$