

# Ratio estimation - the details

Week 4 (4.1)

Stat 260, St. Clair

# Ratio estimation: for a ratio parameter

\* SRS

- For each sampling unit, we measure two variables:  $x_i$  and  $y_i$
- **Ratio Parameter** Suppose our parameter of interest looks like

$$B = \frac{\sum_{i=1}^N y_i}{\sum_{i=1}^N x_i} = \frac{t_y}{t_x} = \frac{\bar{y}_U}{\bar{x}_U}$$

- **Estimator** with a SRS of units:

$$\hat{B} = \frac{\sum_{i=1}^n y_i}{\sum_{i=1}^n x_i} = \frac{\bar{y}}{\bar{x}} = \frac{\overbrace{\bar{y}}^{N \hat{y}}}{\overbrace{\bar{x}}^{n \hat{x}}}$$

- SE and bias: now!

# Ratio estimation: for a ratio parameter

- SE: when  $n$  is "large" (or bias about 0)

$$SE(\hat{B}) \approx \sqrt{\left(1 - \frac{n}{N}\right) \frac{s_e^2}{n\bar{x}^2}}$$

where  $s_e$  is the sample SD of the "errors"  $e_i = y_i - \hat{B}x_i$

- Computational "by hand" short cut:

$$s_e^2 = s_y^2 + \hat{B}^2 s_x^2 - 2\hat{B}\hat{R}s_y s_x$$

◦  $s_x, s_y$ : sample SD's of x/y measurements

◦  $\hat{R}$ : correlation coefficient of the sampled x/y measurements

intro stats

$$r = \hat{R} \leq 1$$

$$\begin{aligned} & \hat{R} \approx 1 \\ & y \approx c \cdot x \\ & \frac{y}{x} \approx c = B \\ & e_i \approx 0 \quad (\text{small}) \end{aligned}$$

# Example 1: revisit petition

$$N = 676 \quad y_i = \# \text{ voter sig.}$$
$$n = 50 \quad x_i = \# \text{ sig.}$$

Signatures to a petition were collected on 676 sheets of paper.

- You take a SRS of 50 sheets and find 1515 signatures, 771 of which are from registered voters.
- The estimated proportion signatures belonging to registered voters is

$$\hat{p} = \frac{771}{1515} = 0.5089 \quad = \widehat{B}$$

Compute the SE for this estimate.

$$P = B = \frac{\sum_{i=1}^{676} y_i}{\sum_{i=1}^{676} x_i}$$

$$\bar{X} = \frac{\# \text{ of sig}}{\# \text{ sheets}} = \# \text{ sig}/\text{sheet} = \frac{1515}{50} \quad \checkmark$$

# Example 1: revisit petition

$$\hat{\beta} = .5089$$

$s_x$

- The standard deviation for the number of signatures per sheet is 2.929 and the standard deviation for the number of registered voter signatures per sheet is 3.909.  $\rightarrow s_y$
- The sample correlation between the total number of signatures and number of registered voter signatures per sheet is 0.4497.

$\rightarrow \hat{R}$

$$s_e^2 = 3.909^2 + (.5089)^2 2.929^2 - \\ - 2(.5089)(.4497)(2.929)(3.909) \\ = 11.51$$

$$SE(\hat{\beta}) = \sqrt{\left(1 - \frac{50}{676}\right) \frac{11.51}{(50)(1515/50)}} \approx .015$$

est.  $\Rightarrow 50.9\%$  of sig. were from reg. voters  
 (SE of 1.5%)

# Ratio estimation for a mean or total

- Suppose we **know a population mean or total** for an **auxiliary variable**  $x$

$$t_x \text{ and/or } \bar{x}_U \text{ are known}$$

$$\begin{aligned} t_y &= \frac{t_y}{t_x} \cdot t_x \\ &= B \cdot t_x \end{aligned}$$

- Population total estimate** for response  $y$  can be written

$$\hat{t}_{y,r} = \hat{B}t_x$$

- Population mean estimate** for response  $y$  can be written

$$\hat{y}_r = \hat{B}\bar{x}_U$$

- SE and bias: Now!

# Ratio estimation for a mean or total

- SE: when  $n$  is "large" (or bias of  $\hat{B}$  about 0)

$$SE(\hat{t}_{y,r}) \approx \sqrt{\left(1 - \frac{n}{N}\right) \left(\frac{t_x}{\bar{x}}\right)^2 \frac{s_e^2}{n}}$$

$$SE(\hat{y}_r) \approx \sqrt{\left(1 - \frac{n}{N}\right) \left(\frac{\bar{x}_U}{\bar{x}}\right)^2 \frac{s_e^2}{n}}$$

where  $s_e$  is the sample SD of the "errors"  $e_i = y_i - \hat{B}x_i$

- Computational "by hand" short cut:

$$s_e^2 = s_y^2 + \hat{B}^2 s_x^2 - 2\hat{B}\hat{R}s_y s_x$$

$$SE(\hat{t}_{y,r}) = SE(\hat{B}t_x) = t_x \underline{SE(\hat{B})}$$

$$N=196 \quad n=49$$

## Example 4: a classic

SRS of cities

$\Rightarrow$  pop. size

$$\Rightarrow t_x = \sum_{i=1}^{196} x_i = 22,919$$

$$\hat{t}_{SRS,x} = 196 \left( \frac{5054}{49} \right)$$

$$= 20,216 \quad \text{population total}$$

$$\hat{t}_{SRS,y} = 196 \left( \frac{6262}{49} \right)$$

$$= 25,048 \quad \text{sample variance}$$

$$\text{sample correlation } \hat{R} = 0.9817$$

1920 ( $x$ )

1925 ( $y$ )

$$t_x = 22,919$$

$$t_y = ??$$

$$\sum_{i=1}^{49} x_i = 5054 \quad \sum_{i=1}^{49} y_i = 6262$$

$$s_x^2 = 10,900.4 \quad s_y^2 = 15,158.8$$

$$\hat{B}$$

$$\hat{t}_{y,ratio} = 28,397$$

$$SE(\hat{t}_{y,ratio}) = ??$$

$\rightarrow 4b$

$$\hat{t}_{y,r} = \left( \frac{\hat{t}_{y,SRS}}{\hat{t}_{x,SRS}} \right) t_x = 25,048 \left( \frac{22,919}{20,216} \right) = 25,048 \left( 1.134 \right)$$

$$S_e^2 = 621.85 \quad (\text{Alg. here})$$

$$SE(\hat{t}_{y,r}) = \sqrt{\left( 1 - \frac{49}{196} \right) \left( \frac{22,919}{5054/49} \right)^2 \frac{621.85}{49}}$$

$$= 685.5$$

Smaller than  $SE(\hat{t}_{y,SRS})$

# Ratio estimation for a mean or total vs. SRS estimation

When is

$$V(\hat{y}_r) \stackrel{?}{<} V(\bar{y}_{srs})$$
$$(1 - \frac{n}{N}) \left( \frac{\bar{x}_n}{\bar{x}} \right)^2 \frac{s_e^2}{n} \stackrel{?}{<} (1 - \frac{n}{N}) \frac{s_y^2}{n}$$

\* Assume

$$\frac{\bar{x}_n}{\bar{x}} \approx 1$$

$$\Rightarrow \boxed{s_e^2 \stackrel{?}{<} s_y^2}$$

$$e_i = y_i - \hat{\beta}x_i \approx 0$$

$y_i \approx c x_i \uparrow \hat{\beta} \Rightarrow$  positive correlation!

general idea:  
"strong" corr.  
between means  $s_e^2 < s_y^2$

# Theory: Bias approximation

$\text{Cov}$  = how do two variables "covary" together

$$\text{corr} = \frac{\text{cov}}{\text{SD} \cdot \text{SD}}$$

$$B = \frac{t_y}{t_x} \quad \hat{B} = \frac{\hat{t}_y}{\hat{t}_x}$$

$$\text{Bias}(\hat{B}) = E(\hat{B}) - B = -\frac{\text{Cov}(\hat{B}, \bar{x})}{\bar{x}_u}$$

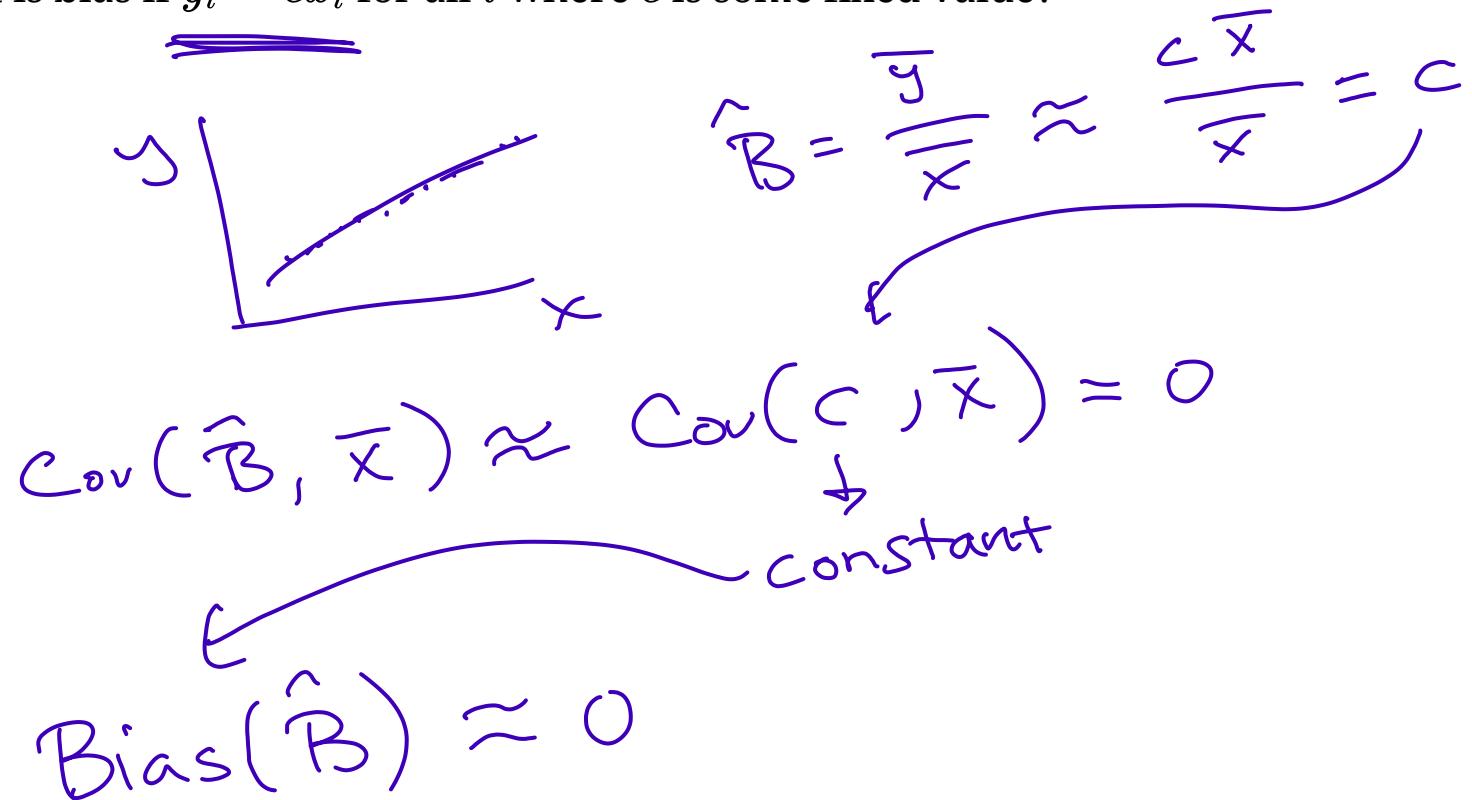
Proof A.2

$$\begin{aligned} \text{Cov}(\hat{B}, \bar{x}) &= E[\hat{B} \bar{x}] - E(\hat{B}) \bar{x}_u \\ &= E[\bar{y} \cdot \bar{x}] - E(\hat{B}) \bar{x}_u \\ &= \bar{y}_u - E(\hat{B}) \bar{x}_u \quad \Rightarrow \text{solve for } E(\hat{B}) \end{aligned}$$

$$E[\hat{B}] = \frac{\bar{y}_u - \text{Cov}(\hat{B}, \bar{x})}{\bar{x}_u} = B - \frac{\text{Cov}(\hat{B}, \bar{x})}{\bar{x}_u}$$

# Theory: Bias approximation

What is bias if  $y_i \approx cx_i$  for all  $i$  where  $c$  is some fixed value?



# Theory: Bias approximation

What is the maximum value of bias (approximately)?

$$\begin{aligned} |\text{Bias}(\hat{\beta})| &= \left| \frac{-\text{Cov}(\hat{\beta}, \bar{x})}{\bar{x}_n} \right| \\ &= \left| \frac{\text{Cov}_{rr}(\hat{\beta}, \bar{x}) \text{SE}(\hat{\beta}) \text{SE}(\bar{x})}{\bar{x}_n} \right| \\ &\leq \frac{\text{SE}(\hat{\beta}) \text{SE}(\bar{x}) \rightarrow \text{SRS}}{|\bar{x}_n|} \\ &\quad * \text{prob1(c)} \end{aligned}$$

(large  $n$ !  
 $x \& y$  positive  
corr.)