

Topics to be included in this tutorial are proportion estimator and ratio estimator for simple random sampling.

1. Proportion estimator

The sample proportion is basically a special case of sample mean when the variable of interest $y_i \in \{0, 1\}$. Therefore, all the theory and results of sample mean under SRS still hold for the sample proportion. Besides proportion $\bar{y} = \frac{1}{n} \sum_{i=1}^n y_i$, sometimes we may also be interested in the number of units in the population with certain property, i.e., $T = \sum_{i=1}^N y_i$. The estimators and standard errors are given in the table below.

- parameter of interest:	$\bar{y} = \frac{1}{n} \sum_{i=1}^n y_i$	$T = \sum_{i=1}^N y_i$
- estimator:	$\hat{p} = \frac{1}{n} \sum_{i=1}^n y_i$	$\hat{T} = \frac{N}{n} \sum_{i=1}^n y_i = N\hat{p}$
- variance of the estimator:	$\text{Var}(\hat{p}) = (1 - \frac{n}{N}) \frac{S^2}{n}$, $S^2 = \frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2 = \frac{N}{n-1} \bar{y}(1-\bar{y})$	$\text{Var}(\hat{T}) = N^2 (1 - \frac{n}{N}) \frac{S^2}{n}$, $S^2 = \frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2 = \frac{N}{n-1} \bar{y}(1-\bar{y})$
- standard error of the estimator:	$\text{se}(\hat{p}) = \sqrt{(1 - \frac{n}{N}) \frac{\hat{p}(1-\hat{p})}{n-1}}$	$\text{se}(\hat{T}) = N \sqrt{(1 - \frac{n}{N}) \frac{\hat{p}(1-\hat{p})}{n-1}}$
- $(1-\alpha)$ - CI:	$\hat{p} \pm z_{\alpha/2} \cdot \text{se}(\hat{p})$	$\hat{T} \pm z_{\alpha/2} \cdot \text{se}(\hat{T})$

Example. Suppose that the governor wants to estimate the total number of people infected by a certain disease in a certain area. To this end, he takes an SRS of size 1000 and finds that 120 among them are infected. Try to give a point estimate as well as a 95% CI. for the total number of infected people in this area. (It's known that the population in this area is 10^5).

Solution. Let $y_i = \begin{cases} 1, & \text{if unit } i \text{ is infected,} \\ 0, & \text{otherwise} \end{cases}$ ($i=1, \dots, N=10^5$), then we

estimate $T = \sum_{i=1}^N y_i$ by $\hat{T} = N\hat{p} = N \cdot \frac{1}{n} \sum_{i=1}^n y_i = 10^5 \times \frac{120}{1000} = 1.2 \times 10^4$. The 95%

C.I. is given by $\hat{T} \pm z_{0.025} \times N \sqrt{(1 - \frac{n}{N}) \frac{\hat{p}(1-\hat{p})}{n-1}} = 12000 \pm 1.96 \times 10^5 \sqrt{0.99 \times \frac{0.12 \times 0.88}{999}}$
 $\approx [9995, 14005]$.

2. Ratio estimator

In class, we have learned that the ratio estimator can be used to improve our estimation for \bar{Y} if the variable of interest y_i and the auxiliary variable x_i are highly correlated. In this

case, $\hat{\bar{Y}}_r = \bar{X} \cdot \frac{\bar{y}}{\bar{x}}$ with $se(\hat{\bar{Y}}_r) = \sqrt{(1 - \frac{n}{N}) \frac{S_y^2}{n}}$, $S_y^2 = \frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2$. Sometimes in practice,

it is the ratio itself $R = \frac{\sum_{i=1}^N y_i}{\sum_{i=1}^N x_i}$ of interest and we can use the ratio estimator

$\hat{R} = \bar{y} / \bar{x}$ to estimate R . The standard error of \hat{R} is given by $se(\hat{R}) = \frac{1}{\bar{x}} \sqrt{(1 - \frac{n}{N}) \frac{S_y^2}{n}}$.

As usual, a $(1-\alpha)$ -C.I. for R is $\hat{R} \pm z_{\alpha/2} \cdot se(\hat{R})$.

Example. Suppose the population consists of ¹⁰⁰ agricultural fields of different sizes. Let
 y_i = bushels of grain harvested in field i

x_i = acreage of field i

We want to estimate the average yield per acre, i.e. $R = \bar{Y} / \bar{X}$. Suppose we

have selected an SRS of size 6 and the numbers are recorded as

below.

y_i	251	150	387	166	480	308
x_i	3	2	5	2	6	4

Then we can estimate the average yield per acre by $\hat{R} = \bar{y} / \bar{x} = \frac{\sum_{i=1}^n y_i}{\sum_{i=1}^n x_i} = 79.18$

and a 95% CI is given by $\hat{R} \pm z_{\alpha/2} \cdot se(\hat{R}) = 79.18 \pm 1.96 \times \frac{1}{\bar{x}} \sqrt{(1 - 0.06) \cdot \frac{S_y^2}{6}}$,
 $= [73.90, 84.46]$.