

Stat 576
11-14-17
(1)

A generalization of the idea behind
PPS sampling

Let π_j be the probability that Y_j is in
the sample.

Let π_{ij} be the joint probability that Y_i and Y_j
are both in the sample

$$\text{Let } \hat{t}_{HT} = \sum_{i=1}^n \frac{y_i}{\pi_i} \quad \text{Horvitz-Thompson estimator}$$

Horvitz-Thompson Theorem:

(2)

$$E(\hat{t}_{HT}) = t$$

$$\text{Proof: } E[\hat{t}_{HT}] = E\left[\sum_{i=1}^n \frac{y_i}{\pi_i}\right]$$

$$= E\left[\sum_{j=1}^N \frac{y_j}{\pi_j} u_j\right]$$

$$\text{Let } u_j = \begin{cases} 1 & \text{if } Y_j \text{ is in sample} \\ 0 & \end{cases}$$

$$= \sum_{j=1}^N \frac{y_j}{\pi_j} E(u_j)$$

$$u_j \sim \text{Bernoulli}(\pi_j)$$

$$= \sum_{j=1}^N \frac{y_j}{\pi_j} \pi_j = t$$

③

What about the variance?

$$\begin{aligned}
 V[\hat{t}_{HT}] &= V\left[\sum_{i=1}^N \frac{y_i}{\pi_i}\right] \\
 &= V\left[\sum_{j=1}^N \frac{y_j}{\pi_j} u_j\right] \\
 &= \sum_{j=1}^N \frac{y_j^2}{\pi_j^2} V[u_j] + \sum_{j=1}^N \sum_{k \neq j} \frac{y_j}{\pi_j} \frac{y_k}{\pi_k} \text{Cov}(u_j, u_k) \\
 &= \sum_{j=1}^N \frac{y_j^2}{\pi_j^2} \pi_j(1-\pi_j) + \sum_{j=1}^N \sum_{k \neq j} \frac{y_j y_k}{\pi_j \pi_k} [E(u_j u_k) - E(u_j)E(u_k)]
 \end{aligned}$$

$$= \sum_{j=1}^N \frac{y_j^2}{\pi_j^2} \pi_j(1-\pi_j) + \sum_{j=1}^N \sum_{k \neq j} \frac{y_j y_k}{\pi_j \pi_k} (\pi_{jk} - \pi_j \pi_k) \quad (4)$$

This is the true variance of \hat{t}_{HT} .

H & T showed 2 unbiased estimators of the variance

$$\hat{V}_1(\hat{t}_{HT}) = \sum_{i=1}^N (1-\pi_i) \frac{y_i^2}{\pi_i^2} + \sum_{i=1}^N \sum_{j \neq i} \frac{\pi_{ij} - \pi_i \pi_j}{\pi_{ij}} \frac{y_i y_j}{\pi_i \pi_j}$$

$$\hat{V}_2(\hat{t}_{HT}) = \sum_{i=1}^A \sum_{j \neq i} \frac{\pi_i \pi_j - \pi_{ij}}{\pi_{ij}} \left(\frac{y_i}{\pi_i} - \frac{y_j}{\pi_j} \right)^2 \quad (5)$$

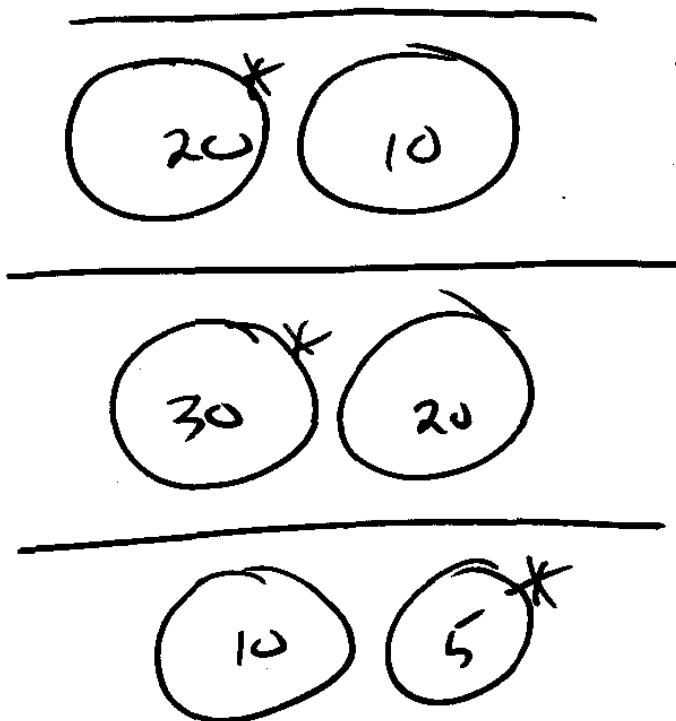
Example 6: Population has 3 strata

Each stratum has 2 clusters

We pick 1 cluster from each stratum

Select 2 people from each of the sampled clusters.

$$n=6$$



Item	π_i
y_1	$\frac{1}{2} \cdot \frac{2}{20} = \frac{1}{20}$
y_2	$\frac{1}{2} \cdot \frac{2}{20} = \frac{1}{20}$
y_3	$\frac{1}{2} \cdot \frac{2}{30} = \frac{1}{30}$
y_4	$\frac{1}{2} \cdot \frac{2}{30} = \frac{1}{30}$
y_5	$\frac{1}{2} \cdot \frac{2}{5} = \frac{1}{5}$
y_6	$\frac{1}{2} \cdot \frac{2}{5} = \frac{1}{5}$

(6)

$$\hat{t}_{HT} = \sum_{i=1}^4 \frac{y_i}{\pi_i} = 20(y_1 + y_2) + 30(y_3 + y_4) + 5(y_5 + y_6) \quad (7)$$

For $\hat{V}[\hat{t}_{HT}]$, we also need π_{ij} for every $i \neq j$

If $y_i \neq y_j$ are from different strata,

$$\text{then } \pi_{ij} = \pi_i \pi_j$$

$$\pi_{12} = \frac{1}{2} \frac{1}{\binom{20}{2}} = \frac{1}{2} \frac{2}{20 \cdot 19} = \frac{1}{20(19)}$$

$$\pi_{34} = \frac{1}{30(29)}$$

$$\pi_{56} = \frac{1}{5(4)}$$

Now, plug these into either $\hat{V}_1[\hat{t}_{HT}]$ or $\hat{V}_2[\hat{t}_{HT}]$

Design effect

$$d.eff. = \frac{V(\text{estimator} | \text{Complex design})}{V(\text{estimator} | \text{SRS of same size})}$$

Example: Stratified sampling with proportional allocation

$$V[\bar{y}_{STR}] = \frac{1}{N^2} \sum_{h=1}^H N_h^2 \frac{S_h^2}{n_h} \left(1 - \frac{n_h}{N_h}\right)$$

$$\text{And } n_h = \frac{N_h}{N} n$$

$$V[\bar{y}_{str}] = \frac{1}{N^2} \sum_{h=1}^H N_h^2 \frac{S_h^2}{\left(\frac{N_h}{N}\right)} \left(1 - \frac{\frac{N_h}{N}}{\frac{N}{N}}\right) \quad (9)$$

$$= \sum_{h=1}^H W_h \frac{S_h^2}{n} \left(1 - \frac{n}{N}\right)$$

$$\text{Also } V[\bar{y}] = \frac{S^2}{n} \left(1 - \frac{n}{N}\right)$$

$$\Rightarrow d.\text{eff.} = \frac{\sum_{h=1}^H W_h S_h^2}{S^2}$$

$$\hat{d.\text{eff.}} = \frac{\sum_{h=1}^H W_h s_h^2}{s^2}$$