

# ON EFFICIENCY OF THE SAMPLING WITH VARYING PROBABILITIES WITHOUT REPLACEMENT

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## 1. INTRODUCTION

It is generally felt that in a sub-sampling design the selection of primary units with varying probabilities without replacement leads to a more efficient estimate than sampling with varying probabilities with replacement. In this connection a paper by Narain (1951) deserves special mention. He compared the two systems of sampling in a two-stage sampling design where the primary units were selected with varying probabilities and the secondary units with equal probabilities without replacement in both cases. He showed the necessary condition, viz.,  $\frac{n}{n-1} \lambda_{ij} < 2\lambda_i \lambda_j$  for the former being more efficient than the latter, where  $\lambda_i$  is the probability of  $i$ th primary unit being included in a sample of size  $n$  in a system of sampling with varying probabilities without replacement and  $\lambda_{ij}$  is the probability of both  $i$ th and  $j$ th units being included in a sample of size  $n$ . Durbin (1953) has also stated (without proof) in a recent paper that it is not difficult to find out some situation in which sampling without replacement will be less efficient as compared to that with replacement. This paper will be a study in that direction.

## 2. EVALUATION OF PROBABILITIES

Before actually comparing the two systems we will evaluate probabilities and consider its behaviour in a sampling design without replacement. Let there be  $N$  primary units and  $p_i$  ( $i = 1, 2, \dots, N$ ) be the initial probability that the  $i$ th unit will be drawn in the first draw. If  $P_{ij, \dots, mn}$  denote the probability that  $i$ th,  $j$ th,  $\dots$  and  $n$ th units will be included in a sample of size  $n$ ,

$$P_{ij, \dots, mn} = \sum \frac{p_i p_j \dots p_m p_n}{(1-p_i)(1-p_i-p_j), \dots, (1-p_i-p_j-\dots-p_m)} \quad (1)$$

(Summation over all the  $n!$  terms which will be formed by exchanging the position of  $i, j, \dots, m$ , and  $n$ )

For example, for  $n = 2$ , the probability that  $i$ th and  $j$ th units will be included in the sample is given by

$$P_{ij} = P_{ji} = p_i p_j \left( \frac{1}{1 - p_i} + \frac{1}{1 - p_j} \right) \quad (2)$$

Similarly for  $n = 3$ , the probability that  $i$ th,  $j$ th, and  $k$ th units will be included in the sample is given by

$$\begin{aligned} P_{ijk} &= P_{ikj} = \dots = P_{kij} \\ &= \sum \frac{p_i p_j p_k}{(1 - p_i)(1 - p_i - p_j)} \\ &\quad \text{(Summation over all the } 3! \text{ terms)} \\ &= p_i p_j p_k \left\{ \frac{1}{1 - p_i - p_j} \left( \frac{1}{1 - p_i} + \frac{1}{1 - p_j} \right) + \frac{1}{1 - p_i - p_k} \right. \\ &\quad \left. \times \left( \frac{1}{1 - p_i} + \frac{1}{1 - p_k} \right) + \frac{1}{1 - p_j - p_k} \left( \frac{1}{1 - p_j} + \frac{1}{1 - p_k} \right) \right\} \quad (3) \end{aligned}$$

Thus the exact expression for  $P_{ij} \dots_n$  may be written for any value of  $n$  although the work involved will be quite laborious for a large value of  $n$  as there will be  $n!$  terms in all. Further it may be seen that

$$\sum_{i=1}^N \sum_{j=i+1}^N \dots \sum_{m=l+1}^N \sum_{n=m+1}^N P_{ij} \dots_{mn} = 1 \quad (4)$$

Now the probability of any unit being included in a sample of  $n$  may easily be calculated from the above expression (1). For example, the probability that  $i$ th unit will be selected in a sample of  $n$  is given by

$$P_i(n) = \sum_j^N \sum_k^N \dots \sum_n^N P_{ijk} \dots_n \quad (5)$$

$i \neq j \neq k \neq \dots \neq n$

Thus

$$P_i(2) = p_i + p_i \sum_{\substack{j=1 \\ j \neq i}}^N p_j / (1 - p_j);$$

$$P_i(3) = P_i(2) + p_i \sum_{\substack{j=1 \\ j \neq i}}^N \sum_{\substack{l=1 \\ l \neq j, i}}^N \left[ \frac{p_j p_l}{1 - p_j - p_l} \left( \frac{1}{1 - p_l} + \frac{1}{1 - p_j} \right) \right];$$

$$\begin{aligned}
 P_i(4) = P_i(3) &+ \sum_{\substack{j \\ i \neq j}}^N \sum_{\substack{k \\ j \neq k}} \sum_{\substack{l \\ k \neq l}} \left[ \frac{1}{1-p_j-p_k} \left( \frac{1}{1-p_j} + \frac{1}{1-p_k} \right) \right. \\
 &+ \frac{1}{1-p_j-p_l} \left( \frac{1}{1-p_j} + \frac{1}{1-p_l} \right) \\
 &+ \left. \frac{1}{1-p_k-p_l} \left( \frac{1}{1-p_k} + \frac{1}{1-p_l} \right) \right] \\
 &\times \frac{p_j p_k p_l}{1-p_j-p_k-p_l}.
 \end{aligned}$$

Similarly the probability of  $i$ th and  $j$ th units being included in the sample is given by

$$\begin{aligned}
 P_{ij}(n) = \sum_k^N \sum_l^N \dots \sum_m^N \sum_n^N P_{ijkl \dots mn} \\
 i \neq j \neq k \neq \dots \neq m \neq n
 \end{aligned}$$

Thus

$$P_{ij}(2) = p_i p_j \left( \frac{1}{1-p_i} + \frac{1}{1-p_j} \right) \quad (6)$$

$$\begin{aligned}
 P_{ij}(3) = P_{ij}(2) &+ p_i p_j \sum_{\substack{k \\ i \neq j \neq k}}^N \left[ \frac{1}{1-p_i-p_k} \left( \frac{1}{1-p_i} + \frac{1}{1-p_k} \right) \right. \\
 &+ \left. \frac{1}{1-p_k-p_j} \left( \frac{1}{1-p_j} + \frac{1}{1-p_k} \right) \right] p_k;
 \end{aligned}$$

$$\begin{aligned}
 P_{ij}(4) = P_{ij}(3) &+ \sum_{\substack{k \\ i \neq j \neq k}}^N \sum_{\substack{l \\ k \neq l}}^N \left[ \frac{1}{(1-p_i)(1-p_i-p_l-p_k)} \right. \\
 &\times \left( \frac{1}{1-p_i-p_k} + \frac{1}{1-p_i-p_l} \right) + \frac{1}{(1-p_j)(1-p_j-p_k-p_l)} \\
 &\times \left( \frac{1}{1-p_j-p_k} + \frac{1}{1-p_j-p_l} \right) + \frac{1}{1-p_i-p_k-p_l} \\
 &\times \left. \left\{ \frac{1}{(1-p_i)(1-p_i-p_l)} + \frac{1}{(1-p_k)(1-p_i-p_k)} \right\} \right]
 \end{aligned}$$

$$\begin{aligned}
 & + \frac{1}{1-p_j-p_k-p_l} \left\{ \frac{1}{(1-p_l)(1-p_l-p_j)} \right. \\
 & + \left. \frac{1}{(1-p_k)(1-p_k-p_j)} \right\} + \frac{1}{1-p_l-p_k} \left( \frac{1}{1-p_l} + \frac{1}{1-p_k} \right) \\
 & \times \left( \frac{1}{1-p_l-p_k-p_j} + \frac{1}{1-p_l-p_k-p_j} \right) \Big] p_l p_k.
 \end{aligned}$$

Now we will prove the following:—

$P_i(n) P_j(n) > P_{ij}(n)$  for all values of  $N$  and

$$p_i < 1 \quad (i = 1, 2, \dots, N), \text{ if } n = 2.$$

For simplicity of notation we will show that

$$P_1(2) P_2(2) > P_{12}(2),$$

and this result will follow for all  $i$  and  $j$  ( $i, j = 1, 2, \dots, N$ ).

$$P_1(2) = \sum_{i=2}^N P_{1i}$$

and

$$P_2(2) = \sum_{\substack{i=1 \\ i \neq 2}}^N P_{2i}$$

$$\therefore P_1(2) P_2(2) = \left( P_{12} + \sum_{i=3}^N P_{1i} \right) \left( P_{12} + \sum_{i=3}^N P_{2i} \right)$$

$$= P_{12} \left( P_{12} + \sum_{i=3}^N P_{1i} + \sum_{i=3}^N P_{2i} \right)$$

$$+ \left( \sum_{i=3}^N P_{1i} \right) \left( \sum_{i=3}^N P_{2i} \right)$$

$$= P_{12} \left[ 1 - \sum_{i=3}^N \sum_{j=i+1}^N P_{ij} \right] + \left( \sum_{i=3}^N P_{1i} \right) \left( \sum_{j=3}^N P_{2i} \right)$$

$$\therefore P_1(2) P_2(2) - P_{12}(2) = \left( \sum_{i=3}^N P_{1i} \right) \left( \sum_{i=3}^N P_{2i} \right) - P_{12} \sum_{i=3}^N \sum_{j=i+1}^N P_{ij}$$

$$[\text{since } P_{12}(2) = P_{12}]$$

Now using the relation (6),

$$P_{ij} = p_i p_j \left( \frac{1}{1-p_i} + \frac{1}{1-p_j} \right)$$

we find that

$$P_1(2) P_2(2) > P_{12}(2) \quad (7)$$

The inequality  $P_1(n) P_2(n) > P_{12}(n)$  will generally hold good even for  $n > 2$  if the  $p_i$ 's are not very heterogeneous. But if some of the values of  $p_i$ 's are so large as to dominate the entire population the above inequality may not hold good. In such cases it will generally

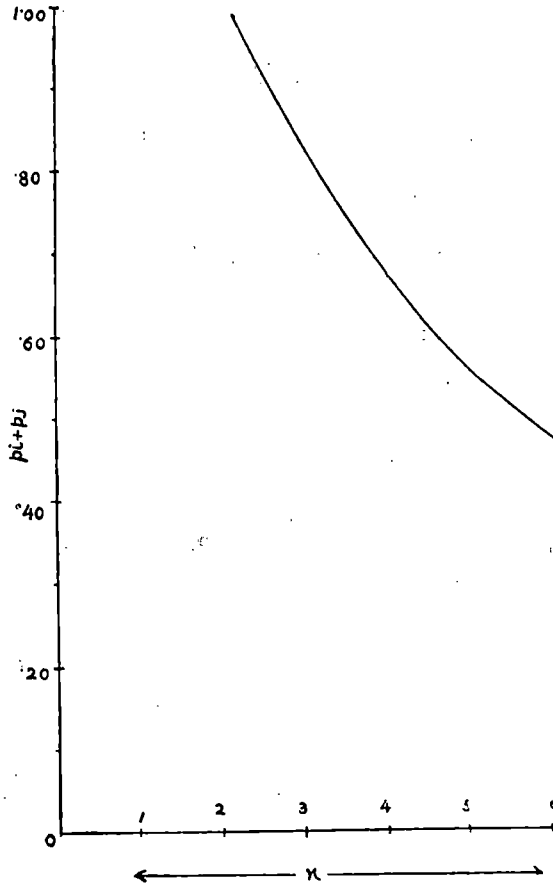


DIAGRAM 1.

$$P_i(n) P_j(n) - P_{ij}(n) = 0$$

- (i)  $P_i(n) P_j(n) > P_{ij}(n)$  below the curve
- (ii)  $P_i(n) P_j(n) < P_{ij}(n)$  above the curve

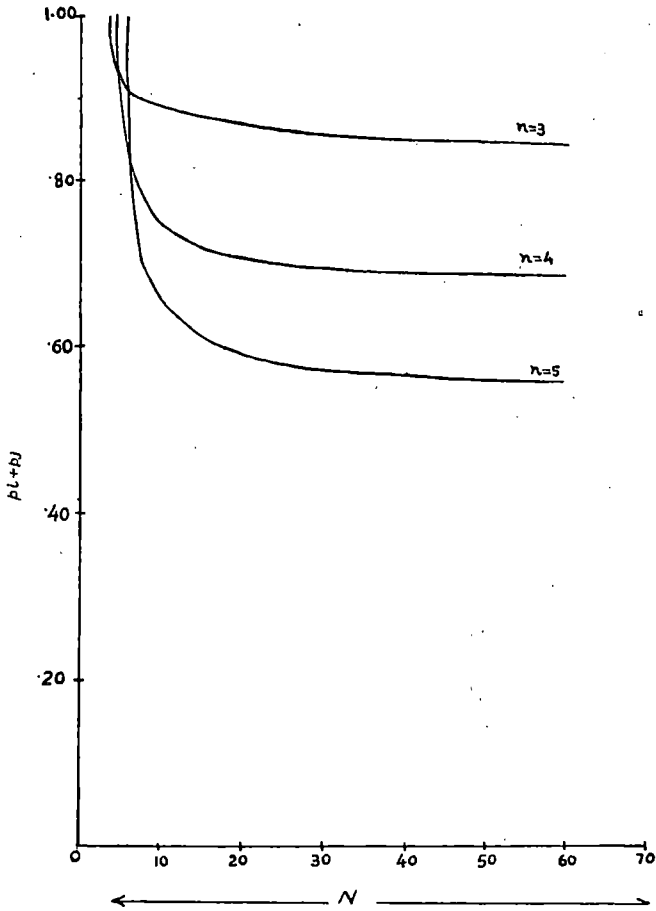


DIAGRAM 2.

$$P_i(n) P_j(n) - P_{ij}(n) = 0$$

- (i)  $P_i(n) P_j(n) > P_{ij}(n)$  below the curve
- (ii)  $P_i(n) P_j(n) < P_{ij}(n)$  above the curve

depend on  $p_i$  and  $p_j$ ,  $N$  and  $n$ . For example, it may be verified that for  $n = 3$ ,  $N = 22$ ,  $p_1 + p_2 = .90$  and  $p_k = .05$  ( $k = 3, 4, \dots, N$ )  $P_1(n) P_2(n) < P_{12}(n)$ . A complete picture as to how the value of  $P_1(n) \cdot P_2(n) - P_{12}(n)$  will depend on  $p_i$  and  $p_j$ ,  $N$  and  $n$  is shown in the diagrams 1 and 2. It may be seen from diagram 1 that the entire space  $P_1(n) P_2(n) - P_{12}(n)$  may be divided into two parts, in one  $P_1(n) P_2(n)$  is always greater than  $P_{12}(n)$  whereas in other it is otherwise. The diagram is indicative of the minimum value  $\phi(n)$

of  $P_1(n) P_2(n) - P_{12}(n)$  for given  $p_1 + p_2$ . For the actual values of  $P_1(n) P_2(n) - P_{12}(n)$  the dividing curve may move slightly up. The minimum value  $\phi(n)$  is calculated by assuming

$$p_k = \frac{1 - p_1 - p_2}{N - 2} \quad (k = 3, 4, \dots, N).$$

#### EFFICIENCY

3. We shall consider a sub-sampling system where a given probability for being included in the sample is assigned to each primary unit in a stratum. The sub-sampling of secondary units within a primary unit shall be always considered with equal probabilities without replacement. For simplicity of notation we shall consider sampling within a single stratum, the results being easily capable of generalization for any number of strata. Let

$N$  = number of p.s.u.'s in the population;

$M_i$  = number of secondary units in the  $i$ th p.s.u.;

$$M = \sum_{i=1}^N M_i,$$

$Y_{ij}$  = the value of  $j$ th secondary unit in the  $i$ th p.s.u.;

$$\bar{Y}_i = \frac{1}{M_i} \sum_{j=1}^{M_i} Y_{ij} = \text{population mean per secondary unit in the } i\text{th p.s.u.};$$

$$\bar{Y} = \frac{\sum_{i=1}^N \sum_{j=1}^{M_i} Y_{ij}}{M} = \text{population mean to be estimated};$$

$$T_i = \sum_{j=1}^{M_i} Y_{ij} = M_i \bar{Y}_i = \text{total for the } i\text{th p.s.u.};$$

$$T = \sum_{i=1}^N T_i = \text{total for the population};$$

$n$  = number of p.s.u. to be selected in the sample (in case of sampling with replacement all  $n$ , p.s.u. need not be distinct);

$m$  = number of secondary units to be sampled from a selected p.s.u.;

$$\sigma_i^2 = \frac{\sum_{j=1}^{M_i} (Y_{ij} - \bar{Y}_i)^2}{M_i - 1} = \text{variance within the } i\text{th p.s.u.};$$

and  $p_i$  has got the same meaning as given in Section 2.

In case of sampling with replacement it is not very difficult to see that (Cochran, 1953)

$$\begin{aligned} \bar{y} &= \frac{1}{nM} \sum_{i=1}^n \frac{M_i}{p_i} \left( \sum_{j=1}^m \frac{y_{ij}}{m} \right) \\ &= \frac{1}{nM} \sum_{i=1}^n \frac{M_i}{p_i} \bar{y}_i \end{aligned}$$

is an unbiased estimate of  $\bar{Y}$ , and

$$V_1 = V(\bar{y}) = \frac{1}{M^2 n} \left[ \sum_{i=1}^N \frac{M_i}{p_i} \frac{M_i - m}{m} \sigma_i^2 + \sum_{i=1}^N \frac{T_i^2}{p_i} - T^2 \right] \quad (8)$$

In case of sampling without replacement it is easy to see that (Horvitz and Thompson, 1952)

$$\begin{aligned} \bar{y}' &= \frac{1}{M} \sum_{i=1}^n \frac{M_i}{P_i(n)} \left( \sum_{j=1}^m \frac{y_{ij}}{m} \right) \\ &= \frac{1}{M} \sum_{i=1}^n \frac{M_i}{P_i(n)} \bar{y}_i \end{aligned}$$

is an unbiased estimate of  $\bar{Y}$ .

and

$$\begin{aligned} V_2 = V(\bar{y}') &= \frac{1}{M^2} \left[ \sum_{i=1}^N \frac{M_i (M_i - m)}{P_i(n) \cdot m} \sigma_i^2 + \sum_{i=1}^N \frac{T_i^2}{P_i(n)} \right. \\ &\quad \left. + \sum_{\substack{i,j \\ i \neq j}}^N \frac{P_{ij}(n) \cdot T_i \cdot T_j}{P_i(n) P_j(n)} - T^2 \right] \quad (9) \end{aligned}$$



Now

$$V_1 - V_2 = \frac{1}{M^2} \left[ \left\{ \sum_{i=1}^N \left( \frac{M_i(M_i - m)}{m} \sigma_i^2 + T_i^2 \right) \right. \right. \\ \left. \left. \times \left( \frac{1}{np_i} - \frac{1}{P_i(n)} \right) \right\} + \left\{ T^2 - \frac{T^2}{n} - \sum_{\substack{i,j \\ i \neq j}}^N \frac{P_{ij}(n) T_i T_j}{P_i(n) P_j(n)} \right\} \right] \quad (10)$$

The first term in the above expression

$$\sum_{i=1}^N \left\{ \frac{M_i(M_i - m)}{m} \sigma_i^2 + T_i^2 \right\} \left\{ \frac{1}{np_i} - \frac{1}{P_i(n)} \right\}$$

may be negligible in view of the fact that

$$\sum_{i=1}^N np_i = \sum_{i=1}^N P_i(n)$$

and  $P_i(n)$  approximates to  $np_i$ . Yates (1953) has stated that bias is negligible if  $np_i$  is taken for  $P_i(n)$  and also  $P_i(n) \geq np_i$  according as  $p_i \leq 1/N$ . Consider the second term. In order that the second term may always be non-negative irrespective of the character under study it is necessary and sufficient that all the principal minors of the discriminant of the quadratic

$$T^2 - \frac{T^2}{n} - \sum_{\substack{i,j \\ i \neq j}} \frac{P_{ij}(n)}{P_i(n) P_j(n)} T_i T_j \quad (11)$$

should be non-negative.

Considering the second order minors we have the necessary condition

$$\frac{n}{n-1} P_{ij}(n) \leq 2 P_i(n) P_j(n) \quad (12)$$

The inequality (12) will always be satisfied for all values of  $n$  if

$$P_{ij}(n) \leq P_i(n) P_j(n)$$

But we have seen in the previous section that this inequality is always satisfied for  $n = 2$ ; for  $n > 2$ , cases may arise when  $P_i(n) P_j(n) < P_{ij}(n)$  which has already been observed in Section 2. In such cases

the discriminant (11) may even be negative and sampling without replacement may lead to a less efficient estimate than the one obtained if the sampling with replacement is adopted. One thing appears very clear from the diagrams that as long as  $np_i \leq 1$  ( $i = 1, 2, \dots, N$ )

$$P_{ij}(n) \leq P_i(n) P_j(n).$$

#### SUMMARY

In a sub-sampling design the selection of primary units with varying probabilities without replacement often leads to a more efficient estimate than the one where the sampling of primary units is carried out with varying probabilities with replacement if  $n$  is 2 (the number of primary units in the sample). But if  $n > 2$ , the former system of sampling need not always lead to a more efficient estimate than the one under the latter system and in such cases the efficiency will generally depend, besides  $n$ , on the values of  $p_i$  (probability of the  $i$ th p.s.u. being selected) and  $N$ .

At the end I gratefully thank Dr. G. R. Seth who helped me in preparing this paper.

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