# A MODIFICATION OF POISSON SAMPLING

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## 1. Introduction

The list selection probability in a sampling scheme is the probability that a unit will be selected at some stage in the sampling regardless of whether it appears in the final sample. The inclusion probability is the probability that an individual unit will be in the final sample.

In Poisson sampling a list of population units is generated and successive units are independently included in the sample, the ith unit being chosen with probability  $p_i$ . This sampling scheme ensures that the list selection probabilities and inclusion probabilities are identical. Moreover, the joint inclusion probabilities are products of the individual inclusion probabilities since the selections are independent. These are the great advantages of Poisson sampling. The main shortcoming of Poisson sampling is that the sampling scheme may yield any sample size for the final sample from 0 to N, the population size.

In this paper we present a modification of Poisson sampling such that a fixed sample size n is achieved. This modification, however, destroys the equality of the list selection probabilities and the inclusion probabilities, and disrupts the simple multiplicative relationships of the joint inclusion probabilities. Nevertheless, given inclusion probabilities, list selection probabilities can be computed that will yield these inclusion probabilities. Also joint inclusion probabilities can be obtained by a similar computation. See Ghosh and Vogt (1998) for other related sampling schemes.

Poisson sampling and the modification both utilize the Horvitz-Thompson estimator for the population total. It can be shown that for the modification the usual Yates-Grundy-Sen estimator of the sampling variance (of the Horvitz-Thompson estimator) is always non-negative. For some sampling methods non-negativity has only been established empirically and/or for small sample sizes. See Cochran (1977), pp. 259-270.

# 2. The Modification

Our modification consists in applying list selection probabilities  $p_i$  to draw an initial sample, but rejecting the sample if its size differs from n, and drawing another sample using the same list selection probabilities. We continue this process until we obtain a sample of size n. The inclusion probability  $\pi_i$  of the *i*-th unit is a rational function of the list selection probabilities but in general is distinct from  $p_i$ . The novelty of this procedure is that list selection probabilities exist and can be found approximately which, through this rejection mechanism, yield any desired inclusion probabilities.

The equation below demonstrates the connection between the list selection probabilities and the inclusion probabilities for the modified procedure.

 $\pi_k =$ 

$$\frac{l_k \sum \{l_{i_1} l_{i_2} \dots l_{i_{n-1}} : k \notin \{i_1, i_2, \dots, i_{n-1}\} \subset \{1, \dots, N\}\}}{\sum \{l_{i_1} l_{i_2} \dots l_{i_n} : \{i_1, i_2, \dots, i_n\} \subset \{1, \dots, N\}\}}$$

$$= f_k(p_1, p_2, ..., p_N)$$

for k = 1, 2, ..., N. Here  $l_i = \frac{p_i}{q_i}$  for i = 1, 2, ..., Nwhere  $p_i$  is the list selection probability of the *i*-th unit in the population,  $q_i = 1 - p_i$ , and  $\pi_k$  is the inclusion probability for the *k*-th unit in the population. Note that  $l_i$  is the odds ratio for list selection of the *i*-th unit.

For example, when the population size N is 6 and the desired sample size n is 3, then

$$\pi_1 = \frac{l_1 \sum \{l_i l_j : \{i, j\} \subset \{2, 3, 4, 5, 6\}\}}{\sum \{l_i l_j l_k : \{i, j, k\} \subset \{1, 2, 3, 4, 5, 6\}\}},$$

with similar formulas for  $\pi_2, ..., \pi_6$ .

To achieve desired inclusion probabilities  $\pi_k$ , an iterative process can be used to find the corresponding list selection probabilities  $p_k$ . We start with  $p_k^{(1)} = \pi_k$ . Then

$$\pi_k^{(1)} = f_k(p_1^{(1)}, ..., p_N^{(1)}).$$

We compare the  $\pi_k^{(1)}$ 's with the  $\pi_k$ 's and make adjustments to the  $p_k^{(1)}$ 's to obtain  $p_k^{(2)}$ 's. According as  $\pi_k^{(1)}$  is greater than or less than  $\pi_k$ , choose  $p_k^{(2)}$  smaller than or larger than  $p_k^{(1)}$ . Do this for each k. Calculate the new  $\pi_k^{(2)}$ 's that result and repeat the process. In general the  $\pi_k^{(j)}$ 's converge to the  $\pi_k$ 's as j increases and the  $p_k^{(j)}$ 's converge to the required list selection probabilities  $\pi_k$ . The iterative process can be performed by a computer program developed by the authors.

A mathematical proof that  $p_k$ 's exist for each set of  $\pi_k$ 's runs as follows. The function  $f = (f_1, ..., f_N)$ defined above has three important properties:

- i) f takes the set  $S = \{p \in \mathbb{R}^N : \sum_{k=1}^N p_k = n, 0 \le p_k \le 1 \text{ for each } k\}$ into itself;
- ii) f is continuous;
- iii)  $f_k(p) = p_k$  whenever  $p_k = 0$  or 1.

Assume that  $1 \leq n \leq N$  and without loss of generality that  $\pi^*$  is a member of S with no component equal to 0 or 1. Consider the map  $p \mapsto p + \lambda(\pi^* - f(p))$ . Here  $\lambda$  is the smaller of the numbers:

$$\inf\{\frac{p_i}{f_i(p) - \pi_i^*} : p \in S, f_i(p) > \pi_i^*, i = 1, 2, ..., N\}$$

and

$$\inf\{\frac{1-p_i}{\pi_i^* - f_i(p)} : p \in S, \pi_i^* > f_i(p), i = 1, 2, ..., N\}.$$

It is easily seen from compactness of S and ii) and iii) that these two numbers are positive and thus that  $\lambda$  is likewise. Furthermore, the definition of  $\lambda$ guarantees that the map just defined takes S into S. Since this map is continuous, the Brouwer Fixed Point Theorem applies, and the map has a fixed point  $p^*$  in S. Accordingly,  $f(p^*) = \pi^*$ .

#### 3. Some Properties

Under the modification the joint inclusion probability  $\pi_{jk}$  for the j-th and k-th population units,  $j \neq k$ , is:

$$\pi_{jk} =$$

$$\frac{l_j l_k \sum \{l_{i_1} l_{i_2} \dots l_{i_{n-2}} : j, k \notin \{i_1, i_2, \dots, i_{n-2}\} \subset \{1, \dots, N\}\}}{\sum \{l_{i_1} l_{i_2} \dots l_{i_n} : \{i_1, i_2, \dots, i_n\} \subset \{1, \dots, N\}\}}$$

It can be shown by algebraic manipulation that this expression is strictly smaller than  $\pi_i \pi_k$ .

The Horvitz-Thompson estimator for the population total is:

$$\sum_{k=1}^{N} \frac{y_k}{\pi_k} I_k$$

where  $I_k$  is the variable taking the value 1 if the kth unit is in the final sample and 0 otherwise, and  $y_k$  is the value of the variable of interest for the kth unit. Then, as in Cochran (1977), p. 261, the Yates-Grundy-Sen estimator of the variance of the Horvitz-Thompson estimator is:

$$\sum_{1 \le j < k \le N} \frac{\pi_j \pi_k - \pi_{jk}}{\pi_j \pi_k} (\frac{y_j}{\pi_j} - \frac{y_k}{\pi_k})^2 I_j I_k.$$

It follows from the above that this is nonnegative.

## References

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