### DOMAIN ESTIMATION IN FINITE POPULATIONS<sup>1</sup>

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### Summary

This paper suggests unbiased estimators (UE's) for the size, mean and total of a domain, with specific features, in a given finite population on the basis of simple random sampling without replacement (SRSWOR) continued till a preassigned number of domain members is observed.

Key words: Domain parameters; finite population: inverse sampling; unbiased estimator.

#### 1. Introduction

Consider a finite population of known size N having an unknown number, M, of units of a given category forming a domain A. Denoting the variate values for the M domain members by  $X_1, \ldots, X_M$ , suppose UE's are required for M,  $T = \sum_{i=1}^{M} X_i$ ,  $\mu = T/M$ , the domain size, total and mean respectively. Haldane (1945) and Sampford (1962) considered inverse sampling with replacement to estimate F = M/N and Rao (1975) gave biased ratio estimators for  $\mu$  based on SRSWOR with a fixed number of draws. This paper gives UE's for the above three domain parameters based on inverse SRSWOR sampling and notes that SRSWOR sampling with a fixed number of draws yields a UE for  $\mu$  only under severely restrictive conditions.

# 2. Unbiased estimation with inverse SRSWOR

For inverse SRSWOR let u denote the random number of draws required to realize a preassigned number, m, of domain members. Denote the parametric space of M by  $\mathcal{M} = \{r, r+1, \ldots, N\}$  and, to avoid trivialities, assume that  $1 < m \le r (\le N)$  which holds in most practical situations. Then the probability distribution of u is given by

$$P(u=n) \approx \frac{\binom{M}{m-1}\binom{N-M}{n-m}}{\binom{N}{n-1}} \cdot \frac{M-m+1}{N-n+1} = g_{Mn} \quad (m \le n \le N-M+m)$$
(2.1)

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**Theorem 2.1.** Every parametric function f(M) admits a unique UE involving u.

**Proof.** For h(u) to be a UE of f(M) it is necessary and sufficient that for any M,  $r \le M \le N$ ,

$$f(M) = \sum_{n=m}^{N-M+m} h(n)g_{Mn},$$

which may be written  $\mathbf{f} = \mathbf{Gh}$ , where  $\mathbf{f} = \{f(r), \dots, f(N)\}'$ ,  $\mathbf{h} = \{h(m), \dots, h(N-r+m)\}'$  and

Since G is clearly nonsingular,  $h = G^{-1}f$  is the unique UE of f(M) among functions of u.

Corollary 2.1. The unique UE of M based on u is  $\hat{M}(u) = N(m-1)/(u-1) = \hat{M}$ .

Formulae for  $E(\hat{M}^2)$  and  $V(\hat{M})$  are available by (2.1), where, as usual, E and V stand for expectation and variance operators. Let  $S^2 = (M-1)^{-1} \sum_i^M (X_i - \mu)^2$ , q(u) and l(u) be UE's, by Theorem 2.1, for  $M^{-1}$  and  $M^2$  respectively,  $\sum_i^V$  denote summation over the A-units in the sample,  $\bar{x} = m^{-1} \sum_i^V X_i$ ,  $Z = m^{-1} \sum_i^V X_i^2$  and  $s^2 = (m-1)^{-1} \sum_i^V (X_i - \bar{x})^2$ . The following results are obtained by conditioning on u.

- (i) A UE for  $\mu$  is  $\bar{x}$  with  $V(\bar{x}) = S^2(1/m 1/M)$ ,
- (ii) A UE for T is  $\hat{T} = \hat{M}\bar{x}$  with

$$V(\hat{T}) = S^{2}(1/m - 1/M)E(\hat{M}^{2}) + \mu^{2}V(\hat{M}),$$

- (iii)  $v(\bar{x}) = s^2(m^{-1} q(u))$  is a UE for  $V(\bar{x})$  and
- (iv)  $v(\hat{T}) = \hat{T}^2 [l(u)(Z s^2) + \hat{M}s^2]$  is a UE for  $V(\hat{T})$ .

## 3. Estimation from SRSWOR with a fixed number of draws

Let SRSWOR be based on a fixed number, d, of draws and c denote the number of A-units in a sample so drawn. If  $d \ge N - r + 1$ , then P(c = 0) = 0,  $\forall M \in \mathcal{M}$  and  $\hat{\mu} = c^{-1} \sum^r X_i$  is a UE of  $\mu$ . On the other hand, unless  $d \ge N - r + 1$  it is possible to get a sample containing no members of the domain so that  $\mu$  cannot be unbiasedly estimated. Thus  $\mu$  admits a UE if and only if  $d \ge N - r + 1$ . The mathematical

details in this regard can be worked out (vide Chaudhuri & Mukerjee (1983)) but are omitted here to save space.

Since (N-r+1) is large in practice especially when r is small, SRSWOR with a fixed number of draws is clearly less convenient than inverse SRSWOR in unbiasedly estimating  $\mu$ . For the former scheme, as with the latter, it is not difficult to find and use UE's for M and T.

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