

A NOTE ON ESTIMATION IN DOUBLE SAMPLING

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Let \bar{X}_t be the finite population means for variates x_t ($t=0, 1, 2$). Let \bar{x}'_2 be a first-phase sample (SRSWOR, of size n') mean on x_2 and \bar{x}_t be a second-phase sub-sample (SRSWOR, of size $n < n'$) means on x_t ($t=0, 1, 2$). For $R = \bar{X}_0/\bar{X}_1$ and for $P = \bar{X}_0\bar{X}_1$, Singh [1] gave double-sample estimators $\hat{R}_1 = \hat{R}\bar{x}_2/\bar{x}'_2$, $\hat{R}_2 = \hat{R}\bar{x}'_2/\bar{x}_2$ and $\hat{P}_1 = \hat{P}\bar{x}_2/\bar{x}'_2$, $\hat{P}_2 = \hat{P}\bar{x}'_2/\bar{x}_2$ respectively (where $\hat{R} = \bar{x}_0/\bar{x}_1$, $\hat{P} = \bar{x}_0\bar{x}_1$). Their asymptotic mean square errors (MSE), to the first order of approximation, derived using delta method, are vide Singh [1]:

$$MSE(\hat{R}_t) = M + fR^2 c_2^2 [1 - 2(-1)^t c_t], \quad t=1, 2$$

$$MS(E\hat{P}_t) = M_1 + fP^2 c_2^2 [1 - 2(-1)^t c^*], \quad t=1, 2$$

where $M = MSE(\hat{R})$, $M_1 = MSE(\hat{P})$, $f = \left(\frac{1}{n} - \frac{1}{n'} \right)$,

c_t = coefficient of variation of x_t ($t=0, 1, 2$),

$\rho_{tt'}$ = correlation coefficient between x_t and $x_{t'}$,

($t, t'=0, 1, 2$), $c = \rho_{02}(c_0/c_2) - \rho_{12}(c_1/c_2)$

and $c^* = \rho_{02}(c_0/c_2) + \rho_{12}(c_1/c_2)$.

Writing $s_{tt'}$ for the sum-sample (of size n)—covariance between x_t and $x_{t'}$ (for $t, t'=0, 1, 2$) we propose better alternative estimators respectively for R and P as

$$\hat{R}^* = \hat{R} + a(\bar{x}_2 - \bar{x}'_2),$$

where $a = -\frac{\hat{R}}{s_{22}}(s_{02}/\bar{x}_0 - s_{12}/\bar{x}_1)$

and
$$\hat{p}^* = \hat{p} + b(\bar{x}_2 - \bar{x}_2'),$$

where
$$b = - \frac{\hat{p}}{s_{22}} (s_{02}/\bar{x}_0 + s_{12}/\bar{x}_1).$$

Their asymptotic (to first order of approximation) *MSE's* are using delta methods again, respectively,

$$M(\hat{R}^*) = M - fR^2 c_2^2 c_2 < MSE(\hat{R}_t)$$

and $M(\hat{p}^*) = M_1 - fP^2 c_2^2 c_2 < MSE(\hat{p}_t)$ for both $t=1, 2$

AN ILLUSTRATION

For the example considered by Singh (1965), we have

$$MSE(\hat{R}) = k' (0.5318)$$

$$MSE(\hat{R}_1) = k' (0.5318) - k^* (1.42020)$$

$$MSE(\hat{R}_2) = k' (0.5318) - k^* (.2728)$$

$$MSE(\hat{R}^*) = k' (0.5318) - k^* (.3123)$$

where
$$k' = \frac{(N-n)}{nN} kR_2,$$

$$k^* = \left(\frac{1}{n} - \frac{1}{n'} \right) kR^2,$$

$$k = \frac{N}{(N-1)}$$

and N is the number of blocks in the population.

From above, we see that \hat{R}^* is the most efficient.

REFERENCE

- [1] Singh, M.P. (1965) : On the estimation of ratio and product of the population parameters. *Sankhya*, 27 (B), 32-128.