A Class Of Separate-Type Estimators For Population
Mean In Stratified Sampling Using Known Parameters
Under Non-Response

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ABSTRACT

The objective of the present paper is to propose a family of separate-type
estimators of population mean in stratified random sampling in presence of non-
response based on the family of estimators proposed by Khoshnevisan et al.
(2007). Under simple random sampling without replacement (SRSWOR) the
expressions of bias and mean square error (MSE) up to the first order of
approximation are derived. The comparative study of the family with respect to
usual estimator has been discussed. The expressions for optimum sample sizes
of the strata in respect to cost of the survey have also been derived. An
empirical study is carried out to shoe the properties of the estimators.

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Key words and phrases: Stratified sampling, separate-type estimator, non-
response, auxiliary information, mean square error.
1. INTRODUCTION

Upadhyaya and Singh (1999) have suggested the class of estimators in simple random sampling using some known population parameter(s) of an auxiliary variable. These estimators have been extended by Kadilar and Cingi (2003) for stratified random sampling. In an attempt to improve the estimators, Kadilar and Cingi (2005), Shabbir and Gupta (2005, 2006) and Singh and Vishwakarma (2008) have suggested new ratio estimators in stratified random sampling. Using power transformation Singh et al. (2008) have suggested a class of estimators adapting the estimators developed by Kadilar and Cingi (2003). Koyuncu and Kadilar (2008, 2009) have proposed a family of combined-type estimators in stratified random sampling based on the family of estimators proposed by Khoshnevisan et al. (2007). Singh et al. (2008) suggested some exponential ratio type estimators in stratified random sampling. Recently Koyuncu and Kadilar (2010) have suggested a family of estimators in stratified random sampling following Diana (1993) and Kadilar and Cingi (2003).

Let \( Y \) and \( X \) be the study and auxiliary variables respectively, with respective population means \( \bar{Y} \) and \( \bar{X} \). Khoshnevisan et al. (2007) have proposed a family of estimators for population mean using known values of some population parameters in simple random sampling (SRS) given by

\[
t = y \left[ \frac{a\bar{X} + b}{\alpha(ax + b) + (1 - \alpha)(a\bar{X} + b)} \right]^g
\]

(1.1)

where \( a \neq 0 \), \( b \) are either real numbers or functions of known parameters of the auxiliary variable \( X \).

In this paper, we have proposed a family of separate-type estimators of population mean in stratified random sampling in presence of non-response on study variable adapting the above family of estimators. The properties of the proposed family of estimators in comparison with usual estimators have been discussed. The expressions for optimum sample sizes of the strata with respect to cost of the survey have been obtained.
Let us consider a finite population of size, $N$, is divided into $k$ strata. Let $N_i$ be the size of $i^{th}$ stratum ($i = 1, 2, ..., k$) and a sample of size $n_i$ is drawn from the $i^{th}$ stratum using SRSWOR scheme such that $\sum_{i=1}^{k} n_i = n$. It is assumed that the non-response is detected on study variable $Y$ only and auxiliary variable $X$ is free from non-response.

Let $\bar{y}_{i1}$ and $\bar{x}_i$ are the unbiased estimators of population means $\bar{Y}_i$ and $\bar{X}_i$ respectively, for the $i^{th}$ stratum, given as

$$
\bar{y}_i^* = \frac{n_{i1} \bar{y}_{i1} + n_{i2} \bar{y}_{i2}}{n_i}
$$

(2.1)

where $\bar{y}_{i1}$ and $\bar{y}_{i2}$ are the means based on $n_{i1}$ response units and $u_{i2}$ non-response units of sub sample selected from $n_{i2}$ non-response units respectively. $\bar{x}_i$ be the sample mean based on $n_i$ units.

Therefore an unbiased estimator of population mean $\bar{Y}$ is given by

$$
\bar{y}_{si}^* = \sum_{i=1}^{k} p_i \bar{y}_i^*
$$

(2.2)

and variance of the estimator is expressed as

$$
V(\bar{y}_{si}^*) = \sum_{i=1}^{k} \left( \frac{1}{n} - \frac{1}{N} \right) p_i^2 S_{yi}^2 + \sum_{i=1}^{k} (k_i - 1) W_{i2} n_i p_i^2 S_{yi2}^2
$$

(2.3)

where $S_{yi}^2$ and $S_{yi2}^2$ are respectively the mean-squares of entire group and non-response group of study variable in the population for the $i^{th}$ stratum. $k_i = \frac{n_{i2}}{u_{i2}}$,

$$
p_i = \frac{N_i}{N} \quad \text{and} \quad W_{i2} = \text{non-response rate of the } i^{th} \text{ stratum in the population} = \frac{N_{i2}}{N_i}.
$$
2.1 SUGGESTED FAMILY OF ESTIMATORS

Adapting the idea of Khoshnevisan et al. (2007), we propose a family of separate-type estimators of population mean $\bar{Y}$, given by

$$T_s = \sum_{i=1}^{k} p_i T_i^*$$  \hspace{1cm} (2.4)

where

$$T_i^* = \bar{y}_i \left[ \frac{a\bar{X}_i + b}{\alpha(ax_i + b) + (1-\alpha)(a\bar{X}_i + b)} \right]^g$$  \hspace{1cm} (2.5)

Obviously, $T_s$ is biased for $\bar{Y}$. Therefore, bias and MSE of $T_s$ can be obtained on using large sample approximations. Let

$$\bar{y}_i = \bar{Y}_i (1 + e_0) \; ; \; \bar{x}_i = \bar{X}_i (1 + e_1)$$

such that $\text{E}(e_0) = \text{E}(e_1) = 0$ and

$$E(e_0^2) = \frac{V(\bar{y}_i)}{\bar{Y}_i^2} = f_i C_{\bar{y}_i}^2 + \frac{(k_i - 1)W_i S_{\bar{y}_i}^2}{n_i \bar{Y}_i^2},$$

$$E(e_1^2) = \frac{V(\bar{x}_i)}{\bar{X}_i^2} = f_i C_{\bar{x}_i}^2,$$

$$E(e_0 e_1) = \frac{\text{Cov}(\bar{y}_i, \bar{x}_i)}{\bar{Y}_i \bar{X}_i} = f_i \rho_i C_{\bar{y}_i} C_{\bar{x}_i},$$

where

$$f_i = \frac{N_i - n_i}{N_i n_i}, \quad C_{\bar{y}_i} = \frac{S_{\bar{y}_i}^2}{\bar{Y}_i^2}, \quad C_{\bar{x}_i} = \frac{S_{\bar{x}_i}^2}{\bar{X}_i^2}, \quad S_{\bar{y}_i}^2, S_{\bar{x}_i}^2$$

be the mean-square of entire group of auxiliary variable in the population for the $i^{th}$ stratum and $\rho_i$ is the correlation coefficient between $Y$ and $X$ in the $i^{th}$ stratum.

Expressing the estimator $T_s$ in terms of $e_0$ and $e_1$, we get

$$T_s = \sum_{i=1}^{k} p_i \bar{Y}_i (1 + e_0) [1 + \alpha \lambda_i e_1]^g$$  \hspace{1cm} (2.6)

where

$$\lambda_i = \frac{a\bar{X}_i}{a\bar{X}_i + b}.$$  

Suppose $|\alpha \lambda_i e_1| < 1$ so that $[1 + \alpha \lambda_i e_1]^g$ is expandable. Expanding the right hand side of the equation (2.6) up to the first order of approximation, we obtain
\[
(T_s - \bar{Y}) = \sum_{i=1}^{k} \bar{Y}_i \left[ e_0 - g \alpha \lambda_i e_i + \frac{g(g + 1)}{2} \alpha^2 \lambda_i^2 e_i^2 - g \alpha \lambda_i e_0 e_i \right] \tag{2.7}
\]

Taking expectations of both sides of (2.7), we get the bias of \( T_s \) up to the first order of approximation, as

\[
B(T_s) = \sum_{i=1}^{k} p_i f_i \bar{Y}_i \left[ \frac{g(g + 1)}{2} \alpha^2 \lambda_i^2 C_{X_i}^2 - \alpha \lambda_i g \rho_{Y_i} C_{Y_i} C_{X_i} \right] \tag{2.8}
\]

Squaring both side of equation (2.7) and taking expectations on both sides of this equation, we get the MSE(\( T_s \)) to the first order of approximation as given below:

\[
MSE(T_s) = \sum_{i=1}^{k} p_i^2 \left[ f_i \bar{Y}_i^2 \left( C_{Y_i}^2 + \alpha^2 \lambda_i^2 g^2 C_{X_i}^2 - 2 \alpha \lambda_i g \rho_{Y_i} C_{Y_i} C_{X_i} \right) + \frac{(k_i - 1)}{n_i} \frac{W_i}{n_i} S_{Y_i}^2 \right] \tag{2.9}
\]

### 2.2 SOME SPECIAL CASES

**Case 1:** If we put \( \alpha = 1, \ a = 1, \ b = 0 \) and \( g = 1 \) in equation (2.4), we get

\[
T_s = \sum_{i=1}^{k} p_i \bar{y}_i \bar{X}_i \tag{2.10}
\]

which is separate ratio estimator of population mean \( \bar{Y} \) under non-response.

**Case 2:** If \( \alpha = 1, \ a = 1, \ b = 0 \) and \( g = -1 \), the equation (2.4) gives

\[
T_s = \sum_{i=1}^{k} p_i \bar{y}_i \frac{-\bar{X}_i}{X_i} \tag{2.11}
\]

which is separate product estimator of population mean \( \bar{Y} \) under non-response.

**Case 3:** If we take \( \alpha = 0, \ a = 0, \ b = 0 \) and \( g = 0 \), the equation (2.4) provides

\[
T_s = \sum_{i=1}^{k} p_i \bar{y}_i \tag{2.12}
\]

which is the usual estimator of population mean \( \bar{Y} \) under non-response.

Similarly, we can obtain the various existing estimators of the family under non-response on different choices of \( \alpha, \ a, \ b \) and \( g \) [ See Khoshnevisan et al. (2007)].
2.3 OPTIMUM CHOICE OF $\alpha$

In order to obtain the optimum $\alpha$ we minimize $MSE(T_s)$ with respect to $\alpha$. Differentiating $MSE(T_s)$ with respect to $\alpha$ and equating the derivative to zero, we get the normal equation

$$\frac{\partial MSE(T_s)}{\partial \alpha} = \sum_{i=1}^{k} p_i^2 f_i \overline{Y_i}^2 \left[ 2\alpha_l^2 \lambda_i C_{X_i}^2 - 2\alpha\lambda_i g \rho_i C_{X_i} C_{Y_i} \right] = 0 \quad (2.13)$$

$$\Rightarrow \alpha_{(opt)} = \frac{\sum_{i=1}^{k} p_i^2 f_i \overline{Y_i}^2 \rho_i C_{X_i} C_{Y_i}}{\lambda_i g \sum_{i=1}^{k} p_i^2 f_i \overline{Y_i}^2 C_{X_i}^2} \quad (2.14)$$

Thus the equation (2.14) provides the value of $\alpha$ at which $MSE(T_s)$ would be minimum.

2.4 OPTIMUM $n_i$ WITH RESPECT TO COST OF THE SURVEY

Let $C_{i0}$ be the cost per unit of selecting $n_i$ units, $C_{i1}$ be the cost per unit in enumerating $n_i$ units and $C_{i2}$ be the cost per unit of enumerating $u_{i2}$ units. Then the total cost for the $i^{th}$ stratum is given by

$$C_i = C_{i0} n_i + C_{i1} n_i + C_{i2} u_{i2} \quad \forall \ i = 1,2,\ldots,k \quad (2.15)$$

Now, we consider the average cost per stratum

$$E(C_i) = n_i \left[ C_{i0} + C_{i1} W_{i1} + C_{i2} \frac{W_{i2}}{k_i} \right] \quad (2.16)$$

Thus the total cost over all the strata is given by

$$C_0 = \sum_{i=1}^{k} E(C_i)$$

$$= \sum_{i=1}^{k} n_i \left[ C_{i0} + C_{i1} W_{i1} + C_{i2} \frac{W_{i2}}{k_i} \right] \quad (2.17)$$

Let us consider the function

$$\phi = MSE(T_s) + \mu C_0 \quad (2.18)$$
where \( \mu \) is Lagrangian multiplier. Differentiating the equation (2.18) with respect to \( n_i \) and \( k_i \) respectively and equating the derivatives to zero, we get the following normal equations:

\[
\frac{\partial \phi}{\partial n_i} = -\frac{p_i^2}{n_i^2} \left[ \left( C_i^2 + \alpha_i \rho \zeta C_{x_i}^2 \right) - 2\alpha_i \rho \zeta \left( C_i W_{n_i} C_{x_i} \right) + (k_i - 1) W_{i^2} S_{n_i}^2 \right] + \mu \left( C_{i0} + C_i W_{n_i} + C_{i2} \frac{W_{i^2}}{k_i} \right) = 0 \tag{2.19}
\]

\[
\frac{\partial \phi}{\partial k_i} = \frac{p_i^2 W_{i^2} S_{n_i}^2}{n_i} - \mu n_i C_{i2} \frac{W_{i^2}}{k_i} = 0 \tag{2.20}
\]

From the equations (2.19) and (2.20), we have

\[
n_i = \frac{p_i \sqrt{Y_i \left( C_{i0} + \alpha_i \rho \zeta C_{x_i}^2 \right) - 2\alpha_i \rho \zeta \left( C_i W_{n_i} C_{x_i} \right) + (k_i - 1) W_{i^2} S_{n_i}^2}}{\sqrt{\mu} \sqrt{C_{i0} + C_i W_{n_i} + C_{i2} \frac{W_{i^2}}{k_i}}} \tag{2.21}
\]

and \( \sqrt{\mu} = \frac{k_i p_i S_{n_i}^2}{n_i \sqrt{C_{i2}}} \) \tag{2.22}

Putting the value of the \( \sqrt{\mu} \) from equation (2.22) into the equation (2.21), we get

\[
k_{i_{\text{opt}}} = \frac{\sqrt{C_{i2} B_i}}{S_{n_i^2} A_i} \tag{2.23}
\]

where \( A_i = \sqrt{C_{i0} + C_i W_{n_i}} \)

and \( B_i = \sqrt{Y_i \left( C_{i0} + \alpha_i \rho \zeta C_{x_i}^2 \right) - 2\alpha_i \rho \zeta \left( C_i W_{n_i} C_{x_i} \right) - W_{i^2} S_{n_i}^2} \)

On substituting \( k_{i_{\text{opt}}} \) into equation (2.21), \( n_i \) can be expressed as
\[ n_i = \frac{B_i \sqrt{A_i} \left( \frac{C_{i_2} B_i W_{i_2} S_{i_2}}{A_i} \right)}{\sqrt{A_i} \sqrt{A_i} + \sqrt{C_{i_2} A_i W_{i_2} S_{i_2}} \frac{B_i}{B_i}} \]  

(2.24)

The \( \sqrt{\mu} \) in terms of total cost \( C_0 \) can be obtained by putting the values of \( k_{(opt)} \) and \( n_i \) from equations (2.23) and (2.24) respectively into equation (2.17) as

\[ \sqrt{\mu} = \frac{1}{C_0} \sum_{i=1}^{k} p_i \left( A_i \frac{B_i}{A_i} + \sqrt{C_{i_2} W_{i_2} S_{i_2}} \right) \]  

(2.25)

Thus the \( n_i \) can be expressed in terms of the total cost \( C_0 \) as

\[ n_{i,(opt)} = \frac{C_0}{\sum_{i=1}^{k} p_i \left( A_i \frac{B_i}{A_i} + \sqrt{C_{i_2} W_{i_2} S_{i_2}} \right) \sqrt{A_i} \sqrt{A_i} + \sqrt{C_{i_2} A_i W_{i_2} S_{i_2}} \frac{B_i}{B_i}} \]  

(2.26)

The optimum values of \( n_i \) and \( k_i \) can be obtained by the expressions (2.26) and (2.23) respectively.

3. **EMPIRICAL STUDY**

In this section, we use the data set in Koyuncu and Kadilar (2009). The data concerning the number of teachers as study variable and the number of students as auxiliary variable in both primary and secondary school for 923 districts at 6 regions (as 1: Marmara, 2: Agean, 3: Mediterranean, 4: Central Anatolia, 5: Black Sea, 6: East and Southeast Anatolia) in Turkey in 2007 (Source: Ministry of Education Republic of Turkey).
Table 1: Stratum means, Mean Squares and Correlation Coefficients

<table>
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<th>Stratum No.</th>
<th>$N_i$</th>
<th>$n_i$</th>
<th>$\bar{Y}_i$</th>
<th>$\bar{X}_i$</th>
<th>$S_{Y_i}$</th>
<th>$S_{X_i}$</th>
<th>$S_{XY_i}$</th>
<th>$\rho_i$</th>
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Table 2: Percent Relative Efficiency (P.R.E.) of $T_s$ with respect to $\bar{y}^*_{st}$ at

$\alpha_{(opt)} = 0.9317$, $a = 1$ and $b = 1$

<table>
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<th>$W_{i2}$</th>
<th>$k_i$</th>
<th>$P.R.E.(T_s)$</th>
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4. CONCLUSION
In this paper, a class of separate-type estimators for estimating the population mean in stratified random sampling under non-response has been proposed and method of finding the optimum estimator of the family has also been discussed. We have derived the expressions for optimum sample sizes in respect to cost of the survey. From the Table 2, it is easily observed that the optimum estimator of the proposed class $T_s$ provides better estimate than usual estimator $\bar{y}_n$ under non-response. It is also observed that the relative efficiency of $T_s$ decreases with increase in the non-response rate $W_i$ and $k_i$.

REFERENCES


