# Estimation of Mean of Finite Population Using Double Sampling Scheme under Non-Response 

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#### Abstract

This paper presents a theoretical study on the estimation of finite population mean in simple random sampling using double sampling scheme under non-response. We have proposed the estimators of population mean utilizing the information on an auxiliary variable under the situation in which both study and auxiliary variables are suffered from non-response. The expressions for the biases and mean square errors of the proposed estimators up to the first order of approximation have been obtained. A theoretical study on the cost of the survey has been carried out. An empirical study has also been carried out to demonstrate the performances of the proposed estimators.


Keywords: Simple random sampling, double sampling scheme, population mean, auxiliary variable and non-response.

## 1 Introduction

The problem of non-response is a big issue in the mail surveys. In a selected sample of specified size, some of the units may not respond or these may not be contacted during the survey period. Due to this reason, the prescribed size of the sample may be reduced and hence the precision of the estimator may decrease in estimating the characteristics of the population. It is obvious that the error due to non-response is not so important if the non-responding units are similar in their characteristics to those of responding units. But, the non-similarity is inherent between the groups of responding and non-responding units in the population, it is imperative to consider the seriousness of non-response. Hansen and Hurwitz [1] were the first who tackled the seriousness of non-response in estimating the population mean. They introduced a technique of sub-sampling of non-respondents to deal with the problem of non-response and its adjustments.

The auxiliary information may be utilized in order to estimate the population characteristics if it is highly correlated with the study character. There is a lot of works which have been discussed in estimating the population parameters using auxiliary information in the presence of non-response. Khare [2] has suggested the estimation technique for estimating the population mean under optimum allocation in the presence of non-response. Khan et al. [3] have conferred the procedure of optimum allocation while conducting a mail survey in multivariate stratified random sampling under non-response. Chaudhary et al. [4] and Chaudhary and Singh [5] have suggested the families of estimators of population mean in various sampling schemes viz. stratified random sampling and two-stage sampling over non-response.

The auxiliary information may easily be utilized to improve the precision of the estimator for study character if the parametric values of auxiliary characters are known. The situations in which the parametric values of auxiliary characters are not known, one can utilize the procedure of double (two-phase) sampling scheme in estimating the parameters of study character. The two-phase sampling involves the method of selecting a larger sample for gathering the information on auxiliary character and then selecting a sub-sample from it for collecting the information on study character. Khare and Sinha [6] have proposed some estimators of population ratio in double sampling scheme under non-response. Singh and Kumar [7] have recommended a general class of estimators of population mean adopting two-phase sampling under non-response. Chaudhary et al. [8] have suggested a class of estimators for assessing the population mean using double sampling scheme in the presence of non-response. Recently, Chaudhary and Kumar [9] have proposed a class of estimators for estimating the mean of a stratified population in two-phase sampling scheme under non-response.

[^0]In most of the mail surveys, it is generally seen that the non-response occurs on both the study and auxiliary variables and leads to reduce the efficiency of the estimators. In such situations, it is very difficult to estimate the population parameters of study character. Thus, the present study aims to suggest the estimation procedures for estimating the finite population mean under the situation in which both study and auxiliary variables are suffered from non-response. The study about the cost of survey has also been carried out.

## 2 Proposed Estimators

Let us suppose that a population consists of $N$ units $\left(U_{1}, U_{2}, \ldots, U_{N}\right)$. Let $Y$ and $X$ be the characteristics under study and auxiliary information respectively. Let their respective population means be $\bar{Y}$ and $\bar{X}$. Let $y_{i}(i=1,2, \ldots N)$ and $x_{i}(i=1,2, \ldots N)$ be the observation on the $i^{t h}$ unit in the population for study and auxiliary variables respectively. Let us assume that both study and auxiliary variables are suffered from non-response and the information on $\bar{X}$ is not available. In such situation, we use the two-phase sampling scheme to estimate the population mean $\bar{Y}$. Let a sample of $n^{\prime}$ units be selected from N units by simple random sampling without replacement (SRSWOR) scheme at the first phase and then a smaller sub-sample of $n$ units be selected from $n^{\prime}\left(n<n^{\prime}\right)$ units by SRSWOR at the second phase. At the first phase, it is observed that there are $n_{1}^{\prime}$ units respond and $n_{2}^{\prime}$ units do not respond on auxiliary variable. Now, we select a sub-sample of $h_{2}^{\prime}$ units from the $n_{2}^{\prime}$ units by $\operatorname{SRSWOR}\left(h_{2}^{\prime}=\frac{n_{2}^{\prime}}{L^{\prime}}, L^{\prime}>1\right)$ where $L^{\prime}$ is the inverse sampling rate at the first phase and collect the information on all the $h_{2}^{\prime}$ units [see Hansen and Hurwitz [1]]. Thus, the estimate of $\bar{X}$ at the first phase is given by

$$
\begin{equation*}
\bar{x}^{\prime *}=\frac{n_{1}^{\prime} \bar{x}_{n 1}^{\prime}+n_{2}^{\prime} \bar{x}_{h 2}^{\prime}}{n^{\prime}} \tag{1}
\end{equation*}
$$

where $\bar{x}_{n 1}^{\prime}$ and $\bar{x}_{h 2}^{\prime}$ are respectively the means based on $n_{1}^{\prime}$ responding units and $h_{2}^{\prime}$ non-responding units.
The estimator $\bar{x}^{\prime *}$ gives unbiased estimate of $\bar{X}$ and the variance of $\bar{x}^{\prime *}$ is given as

$$
\begin{equation*}
V\left(\bar{x}^{\prime *}\right)=\left(\frac{1}{n^{\prime}}-\frac{1}{N}\right) S_{X}^{2}+\frac{\left(L^{\prime}-1\right)}{n^{\prime}} W_{2} S_{X 2}^{2} \tag{2}
\end{equation*}
$$

where $S_{X}^{2}$ and $S_{X 2}^{2}$ are the mean squares of entire group and non-response group respectively for the auxiliary variable. $W_{2}$ is the non-response rate in the population.

At the second phase, it is noted that there are $n_{1}$ respondent units and $n_{2}$ non-respondent units out of $n$ units for both study and auxiliary variables. Applying Hansen and Hurwitz [1] technique of sub-sampling of non-respondents, we select a sub-sample of $h_{2}$ units from the $n_{2}$ non-respondents by SRSWOR at the second phase ( $h_{2}=\frac{n_{2}}{L}, L>1$ ) and gather the information on all the $h_{2}$ units ( $L$ being inverse sampling rate at the second phase). Thus, the Hansen and Hurwitz [1] estimators of $\bar{Y}$ and $\bar{X}$ at the second phase are respectively given by

$$
\begin{equation*}
\bar{y}^{*}=\frac{n_{1} \bar{y}_{n 1}+n_{2} \bar{y}_{h 2}}{n} \tag{3}
\end{equation*}
$$

and

$$
\begin{equation*}
\bar{x}^{*}=\frac{n_{1} \bar{x}_{n 1}+n_{2} \bar{x}_{h 2}}{n} \tag{4}
\end{equation*}
$$

where $\bar{y}_{n 1}$ and $\bar{x}_{n 1}$ are the means based on $n_{1}$ respondent units for study and auxiliary variables respectively. $\bar{y}_{h 2}$ and $\bar{x}_{h 2}$ are respectively the means based on $h_{2}$ non-respondent units for study and auxiliary variables.

The variances of $\bar{y}^{*}$ and $\bar{x}^{*}$ are respectively represented as

$$
\begin{align*}
& V\left(\bar{y}^{*}\right)=\left(\frac{1}{n}-\frac{1}{N}\right) S_{Y}^{2}+\frac{(L-1)}{n} W_{2} S_{Y 2}^{2}  \tag{5}\\
& V\left(\bar{x}^{*}\right)=\left(\frac{1}{n}-\frac{1}{N}\right) S_{X}^{2}+\frac{(L-1)}{n} W_{2} S_{X 2}^{2} \tag{6}
\end{align*}
$$

where $S_{Y}^{2}$ and $S_{Y 2}^{2}$ are respectively the mean squares of entire group and non-response group for the study variable.

We now propose some estimators of population mean $\bar{Y}$ using an auxiliary variable whenever the population mean $\bar{X}$ is not known under the condition that both study and auxiliary variables are suffered from non-response. Thus, the ratio, product and regression type estimators of population mean $\bar{Y}$ under the above circumstances are respectively given by

$$
\begin{align*}
& T_{1}^{\prime *}=\frac{\bar{y}^{*}}{\bar{x}^{*}} \bar{x}^{\prime *}  \tag{7}\\
& T_{2}^{\prime *}=\frac{\bar{y}^{*}}{\bar{x}^{\prime *}} \bar{x}^{*} \tag{8}
\end{align*}
$$

and

$$
\begin{equation*}
T_{3}^{\prime *}=\bar{y}^{*}+b^{*}\left(\bar{x}^{\prime}-\bar{x}^{*}\right) \tag{9}
\end{equation*}
$$

where $b^{*}=\frac{s_{x y}^{*}}{s_{x}^{2 *}}$ is an estimator of population regression coefficient of $Y$ on $X$ i.e. $\beta=\frac{S_{X Y}}{S_{X}^{2}}, s_{x y}^{*}$ and $s_{x}^{2 *}$ are respectively the unbiased estimators of $S_{X Y}$ and $S_{X}^{2}$ based on $\left(n_{1}+h_{2}\right)$ units.

In order to obtain the biases and mean square errors (MSE) of the estimators $T_{1}^{\prime *}, T_{2}^{\prime *}$ and $T_{3}^{\prime *}$, we use large sample approximation. Let us consider
$\bar{y}^{*}=\bar{Y}\left(1+e_{0}\right), \bar{x}^{*}=\bar{X}\left(1+e_{1}\right), \bar{x}^{*}=\bar{X}\left(1+e_{1}^{\prime}\right), s_{x y}^{*}=S_{X Y}\left(1+e_{2}\right)$ and $s_{x}^{2 *}=S_{X}^{2}\left(1+e_{3}\right)$
such that $E\left(e_{0}\right)=E\left(e_{1}\right)=E\left(e_{1}^{\prime}\right)=E\left(e_{2}\right)=E\left(e_{3}\right)=0$,

$$
\begin{gathered}
E\left(e_{0}^{2}\right)=\left(\frac{1}{n}-\frac{1}{N}\right) C_{Y}^{2}+\frac{(L-1)}{n} W_{2} \frac{S_{Y 2}^{2}}{\bar{Y}^{2}}, E\left(e_{1}^{2}\right)=\left(\frac{1}{n}-\frac{1}{N}\right) C_{X}^{2}+\frac{(L-1)}{n} W_{2} \frac{S_{X 2}^{2}}{\bar{X}^{2}} \\
E\left(e_{1}^{\prime 2}\right)=\left(\frac{1}{n^{\prime}}-\frac{1}{N}\right) C_{X}^{2}+\frac{\left(L^{\prime}-1\right)}{n^{\prime}} W_{2} \frac{S_{X 2}^{2}}{\bar{X}^{2}}, E\left(e_{0} e_{1}\right)=\left(\frac{1}{n}-\frac{1}{N}\right) \rho C_{X} C_{Y}+\frac{(L-1)}{n} W_{2} \rho_{2} \frac{S_{X 2}}{\bar{X}} \frac{S_{Y 2}}{\bar{Y}} \\
E\left(e_{0} e_{1}^{\prime}\right)=\left(\frac{1}{n^{\prime}}-\frac{1}{N}\right) \rho C_{X} C_{Y}+\frac{\left(L^{\prime}-1\right)}{n^{\prime}} W_{2} \rho_{2} \frac{S_{X 2}}{\bar{X}} \frac{S_{Y 2}}{\bar{Y}} \\
E\left(e_{1} e_{1}^{\prime}\right)=\left(\frac{1}{n^{\prime}}-\frac{1}{N}\right) C_{X}^{2}+\frac{\left(L^{\prime}-1\right)}{n^{\prime}} W_{2} \frac{S_{X 2}^{2}}{\bar{X}^{2}}, C_{X}=\frac{S_{X}}{\bar{X}}, C_{Y}=\frac{S_{Y}}{\bar{Y}}
\end{gathered}
$$

where $\rho$ and $\rho_{2}$ are respectively the correlation coefficients between $Y$ and $X$ for entire group and non-response group.
Now expressing the equation (7) in the terms of $e_{0}, e_{1}, e_{1}^{\prime}$ and neglecting the terms involving powers of $e_{0}, e_{1}$ and $e_{1}^{\prime}$ greater than two, we get

$$
\begin{equation*}
T_{1}^{\prime *}-\bar{Y}=\bar{Y}\left[e_{0}+e_{1}^{\prime}-e_{1}+e_{0} e_{1}^{\prime}-e_{0} e_{1}-e_{1} e_{1}^{\prime}+e_{1}^{2}\right] \tag{10}
\end{equation*}
$$

Taking the expectation on both sides of the equation (10), we get $E\left[T_{1}^{\prime *}-\bar{Y}\right]=\bar{Y}\left[E\left(e_{0} e_{1}^{\prime}\right)-E\left(e_{0} e_{1}\right)-E\left(e_{1} e_{1}^{\prime}\right)+E\left(e_{1}^{2}\right)\right]=\bar{Y}\left[\left(\frac{1}{n^{\prime}}-\frac{1}{N}\right) \rho C_{X} C_{Y}+\frac{\left(L^{\prime}-1\right)}{n^{\prime}} W_{2} \rho_{2} \frac{S_{X_{2}}}{X} \frac{S_{Y_{2}}}{\bar{Y}}-\left(\frac{1}{n}-\frac{1}{N}\right) \rho C_{X} C_{Y}-\frac{(L-1)}{n} W_{2} \rho_{2} \frac{S_{X_{2}}}{X} \frac{S_{Y_{2}}}{Y}\right.$

$$
\left.-\left(\frac{1}{n^{\prime}}-\frac{1}{N}\right) C_{X}^{2}-\frac{\left(L^{\prime}-1\right)}{n^{\prime}} W_{2} \frac{S_{X 2}^{2}}{\bar{X}^{2}}+\left(\frac{1}{n}-\frac{1}{N}\right) C_{X}^{2}+\frac{(L-1)}{n} W_{2} \frac{S_{X 2}^{2}}{\bar{X}^{2}}\right]
$$

Thus the bias of $T_{1}^{*}$ to the first order of approximation is given by

$$
\begin{equation*}
B\left(T_{1}^{\prime *}\right)=\bar{Y}\left[\left(\frac{1}{n}-\frac{1}{n^{\prime}}\right)\left(C_{X}^{2}-\rho C_{X} C_{Y}\right)+W_{2}\left(\frac{S_{X 2}^{2}}{\bar{X}^{2}}-\rho_{2} \frac{S_{X 2}}{\bar{X}} \frac{S_{Y 2}}{\bar{Y}}\right)\left(\frac{(L-1)}{n}-\frac{\left(L^{\prime}-1\right)}{n^{\prime}}\right)\right] \tag{11}
\end{equation*}
$$

Squaring both the sides of the equation (10) and then taking expectation on neglecting the terms involving powers in $e_{0}, e_{1}$ and $e_{1}^{\prime}$ higher than two, we get

$$
E\left[T_{1}^{\prime *}-\bar{Y}\right]^{2}=\bar{Y}^{2}\left[E\left(e_{0}^{2}\right)+E\left(e_{1}^{2}\right)+E\left(e_{1}^{\prime 2}\right)+2 E\left(e_{0} e_{1}^{\prime}\right)-2 E\left(e_{0} e_{1}\right)-2 E\left(e_{1} e_{1}^{\prime}\right)\right]
$$

$$
\begin{gathered}
=\bar{Y}^{2}\left[\left(\frac{1}{n}-\frac{1}{N}\right) C_{Y}^{2}+\frac{(L-1)}{n} W_{2} \frac{S_{Y 2}^{2}}{\bar{Y}^{2}}+\left(\frac{1}{n}-\frac{1}{N}\right) C_{X}^{2}+\frac{(L-1)}{n} W_{2} \frac{S_{X 2}^{2}}{\bar{X}^{2}}+\right. \\
\left(\frac{1}{n^{\prime}}-\frac{1}{N}\right) C_{X}^{2}+\frac{\left(L^{\prime}-1\right)}{n^{\prime}} W_{2} \frac{S_{X 2}^{2}}{\bar{X}^{2}}+2\left(\frac{1}{n^{\prime}}-\frac{1}{N}\right) \rho C_{X} C_{Y}+2 \frac{\left(L^{\prime}-1\right)}{n^{\prime}} W_{2} \rho_{2} \frac{S_{X 2}}{\bar{X}} \frac{S_{Y 2}}{\bar{Y}}- \\
\left.2\left(\frac{1}{n}-\frac{1}{N}\right) \rho C_{X} C_{Y}-2 \frac{(L-1)}{n} W_{2} \rho_{2} \frac{S_{X 2}}{\bar{X}} \frac{S_{Y 2}}{\bar{Y}}-2\left(\frac{1}{n^{\prime}}-\frac{1}{N}\right) C_{X}^{2}-2 \frac{\left(L^{\prime}-1\right)}{n^{\prime}} W_{2} \frac{S_{X 2}^{2}}{\bar{X}^{2}}\right] .
\end{gathered}
$$

Thus the MSE of $T_{1}^{\prime *}$ to the first degree of approximation is given by

$$
\begin{align*}
\operatorname{MSE}\left(T_{1}^{\prime *}\right)=\left(\frac{1}{n^{\prime}}-\frac{1}{N}\right) S_{Y}^{2}+ & \left(\frac{1}{n}-\frac{1}{n^{\prime}}\right)\left(S_{Y}^{2}+R^{2} S_{X}^{2}-2 \rho R S_{X} S_{Y}\right)+\frac{(L-1)}{n} W_{2} S_{Y 2}^{2} \\
& +W_{2}\left(R^{2} S_{X 2}^{2}-2 \rho_{2} R S_{X 2} S_{Y 2}\right)\left[\frac{(L-1)}{n}-\frac{\left(L^{\prime}-1\right)}{n^{\prime}}\right] . \tag{12}
\end{align*}
$$

We now express the equation (8) in terms of $e_{0}, e_{1}$ and $e_{1}^{\prime}$ on neglecting the contribution of terms involving powers ine $e_{0}, e_{1}$ and $e_{1}^{\prime}$ greater than two as

$$
\begin{equation*}
T_{2}^{\prime *}-\bar{Y}=\bar{Y}\left[e_{0}+e_{1}-e_{1}^{\prime}+e_{0} e_{1}-e_{0} e_{1}^{\prime}-e_{1} e_{1}^{\prime}+e_{1}^{\prime 2}\right] . \tag{13}
\end{equation*}
$$

Thus the bias of $T_{2}^{*}$ * to the first order of approximation is given by

$$
\begin{align*}
& B\left(T_{2}^{\prime *}\right)=E\left[T_{2}^{\prime *}-\bar{Y}\right]=\bar{Y}\left[E\left(e_{0} e_{1}\right)-E\left(e_{0} e_{1}^{\prime}\right)-E\left(e_{1} e_{1}^{\prime}\right)+E\left(e_{1}^{\prime 2}\right)\right] \\
& \quad=\bar{Y}\left[\left(\frac{1}{n}-\frac{1}{n^{\prime}}\right) \rho C_{X} C_{Y}+W_{2} \rho_{2} \frac{S_{X 2}}{\bar{X}} \frac{S_{Y 2}}{\bar{Y}}\left(\frac{(L-1)}{n}-\frac{\left(L^{\prime}-1\right)}{n^{\prime}}\right)\right] \tag{14}
\end{align*}
$$

Squaring both the sides of the equation (13) and taking expectation, we get

$$
E\left[T_{2}^{\prime *}-\bar{Y}\right]^{2}=\bar{Y}^{2}\left[E\left(e_{0}^{2}\right)+E\left(e_{1}^{2}\right)+E\left(e_{1}^{\prime 2}\right)+2 E\left(e_{0} e_{1}\right)-2 E\left(e_{0} e_{1}^{\prime}\right)-2 E\left(e_{1} e_{1}^{\prime}\right)\right] .
$$

Thus the MSE of $T_{2}{ }^{*}$ to the first degree of approximation is represented as

$$
\begin{align*}
M S E\left(T_{2}^{*}\right)=\left(\frac{1}{n^{\prime}}-\frac{1}{N}\right) S_{Y}^{2}+ & \left(\frac{1}{n}-\frac{1}{n^{\prime}}\right)\left(S_{Y}^{2}+R^{2} S_{X}^{2}+2 \rho R S_{X} S_{Y}\right)+\frac{(L-1)}{n} W_{2} S_{Y 2}^{2} \\
& +W_{2}\left(R^{2} S_{X 2}^{2}+2 \rho_{2} R S_{X 2} S_{Y 2}\right)\left[\frac{(L-1)}{n}-\frac{\left(L^{\prime}-1\right)}{n^{\prime}}\right] \tag{15}
\end{align*}
$$

Expressing the equation (9) in terms of $e_{0}, e_{1}, e_{1}^{\prime}, e_{2}, e_{3}$ and neglecting the contribution of terms involving powers in $e_{0}, e_{1}, e_{1}^{\prime}, e_{2}, e_{3}$ higher than two, we get

$$
\begin{equation*}
T_{3}^{\prime *}-\bar{Y}=\bar{Y} e_{0}+\beta \bar{X}\left[e_{1}^{\prime}-e_{1}+e_{1}^{\prime} e_{2}-e_{1}^{\prime} e_{3}+e_{1} e_{3}-e_{1} e_{2}\right] \tag{16}
\end{equation*}
$$

Taking expectation on both the sides of the equation (16), we get

$$
E\left[T_{3}^{\prime *}-\bar{Y}\right]=\beta \bar{X}\left[E\left(e_{1}^{\prime} e_{2}\right)-E\left(e_{1}^{\prime} e_{3}\right)+E\left(e_{1} e_{3}\right)-E\left(e_{1} e_{2}\right)\right]
$$

Thus the bias of $T_{3}^{\prime *}$ to the first order of approximation is given by

$$
\begin{equation*}
B\left(T_{3}^{\prime *}\right)=\beta\left[\frac{N^{2}}{(N-1)(N-2)}\left(\frac{1}{n}-\frac{1}{n^{\prime}}\right)\left(\frac{\mu_{30}}{S_{X}^{2}}-\frac{\mu_{21}}{S_{X Y}}\right)+W_{2}\left(\frac{\mu_{30(2)}}{S_{X}^{2}}-\frac{\mu_{21(2)}}{S_{X Y}}\right)\left(\frac{(L-1)}{n}-\frac{\left(L^{\prime}-1\right)}{n^{\prime}}\right)\right] \tag{17}
\end{equation*}
$$

where $\mu_{r s}=\frac{1}{N} \sum_{i=1}^{N}\left(x_{i}-\bar{X}\right)^{r}\left(y_{i}-\bar{Y}\right)^{s}, \mu_{r s(2)}=\frac{1}{N_{2}} \sum_{i}^{N_{2}}\left(x_{i}-\bar{X}_{2}\right)^{r}\left(y_{i}-\bar{Y}_{2}\right)^{s}$,

$$
\bar{X}_{2}=\frac{1}{N_{2}} \sum_{i}^{N_{2}} x_{i} \text { and } \bar{Y}_{2}=\frac{1}{N_{2}} \sum_{i}^{N_{2}} y_{i}
$$

Squaring both the sides of the equation (16) and taking expectation on neglecting the terms involving powers of $e_{0}, e_{1}, e_{1}^{\prime}, e_{2}, e_{3}$ greater than two, we get

$$
E\left[T_{3}^{\prime *}-\bar{Y}\right]^{2}=\bar{Y} E\left(e_{0}^{2}\right)+\beta^{2} \bar{X}^{2}\left[E\left(e_{1}^{\prime 2}\right)+E\left(e_{1}^{2}\right)-2 E\left(e_{1} e_{1}^{\prime}\right)\right]+2 \beta \overline{X Y}\left[E\left(e_{0} e_{1}^{\prime}\right)-E\left(e_{0} e_{1}\right)\right]
$$

Thus the MSE of $T_{3}^{\prime *}$ to the first degree of approximation is expressed as

$$
\begin{align*}
& \operatorname{MSE}\left(T_{3}^{\prime *}\right)=\left[\left(\frac{1}{n^{\prime}}-\frac{1}{N}\right)+\left(\frac{1}{n}-\frac{1}{n^{\prime}}\right)\left(1-\rho^{2}\right)\right] S_{Y}^{2}+\frac{(L-1)}{n} W_{2} S_{Y 2}^{2} \\
&+W_{2}\left(\beta^{2} S_{X 2}^{2}-2 \beta \rho_{2} S_{X 2} S_{Y 2}\right)\left[\frac{(L-1)}{n}-\frac{\left(L^{\prime}-1\right)}{n^{\prime}}\right] \tag{18}
\end{align*}
$$

## 3 Cost of the Survey and Optimum Values of $n^{\prime}, n$, $L^{\prime}$ and $L$

Let $c^{\prime}$ be the unit cost associated with the first phase sample of size $n^{\prime}$ on first attempt. Let $c_{1}^{\prime}$ and $c_{2}^{\prime}$ be the costs per unit of enumerating $n_{1}^{\prime}$ respondent units and $h_{2}^{\prime}$ non-respondent units respectively. Let $c$ be the cost per unit on first attempt associated with the second phase sample of sizen. Now, let $c_{1}$ and $c_{2}$ respectively be the costs per unit of enumerating $n_{1}$ respondent units and $h_{2}$ non-respondent units. Thus the total cost is given by

$$
C=c^{\prime} n^{\prime}+c n+c_{1}^{\prime} n_{1}^{\prime}+c_{1} n_{1}+c_{2}^{\prime} h_{2}^{\prime}+c_{2} h_{2}
$$

Now, the total expected average cost is given as

$$
\begin{align*}
& C_{0}=E(C)=c^{\prime} n^{\prime}+c n+c_{1}^{\prime} n_{1}^{\prime}+c_{1} n_{1}+c_{2}^{\prime} h_{2}^{\prime}+c_{2} h_{2} \\
& =n^{\prime}\left(c^{\prime}+c_{1}^{\prime} W_{1}+c_{2}^{\prime} \frac{W_{2}}{L^{\prime}}\right)+n\left(c+c_{1} W_{1}+c_{2} \frac{W_{2}}{L}\right) \tag{19}
\end{align*}
$$

where $W_{1}$ is the response rate in the population.
Let us define the Lagrange functions

$$
\begin{align*}
& \phi_{1}=M S E\left(T_{1}^{\prime *}\right)+\mu C_{0}  \tag{20}\\
& \phi_{2}=M S E\left(T_{2}^{\prime *}\right)+\mu C_{0}  \tag{21}\\
& \phi_{3}=M S E\left(T_{3}^{\prime *}\right)+\mu C_{0} \tag{22}
\end{align*}
$$

where $\mu$ is Lagrange's multiplier.

## Case (i): (Under Estimator $T_{1}^{\prime *}$ )

To get the normal equations, we differentiate the equation (20) with respect to $n^{\prime}, n, L^{\prime}$ and $L$ respectively, and equate the derivatives to zero. Thus, we have

$$
\begin{equation*}
\frac{\partial \phi_{1}}{\partial n^{\prime}}=\frac{1}{n^{\prime 2}}\left[R^{2} S_{X}^{2}-2 \rho R S_{X} S_{Y}+\left(L^{\prime}-1\right) W_{2}\left(R^{2} S_{X 2}^{2}-2 \rho_{2} R S_{X 2} S_{Y 2}\right)\right]+\mu\left(c^{\prime}+c_{1}^{\prime} W_{1}+c_{2}^{\prime} \frac{W_{2}}{L^{\prime}}\right)=0 \tag{23}
\end{equation*}
$$

$\frac{\partial \phi_{1}}{\partial n}=-\frac{1}{n^{2}}\left[\left(S_{Y}^{2}+R^{2} S_{X}^{2}-2 \rho R S_{X} S_{Y}\right)+(L-1) W_{2}\left(S_{Y 2}^{2}+R^{2} S_{X 2}^{2}-2 \rho_{2} R S_{X 2} S_{Y 2}\right)\right]+\mu\left(c+c_{1} W_{1}+c_{2} \frac{W_{2}}{L}\right)=0$

$$
\begin{equation*}
\frac{\partial \phi_{1}}{\partial L^{\prime}}=\frac{W_{2}}{n^{\prime}}\left(R^{2} S_{X 2}^{2}-2 \rho_{2} R S_{X 2} S_{Y 2}\right)+\mu n^{\prime} c_{2}^{\prime} \frac{W_{2}}{L^{\prime 2}}=0 \tag{24}
\end{equation*}
$$

and

$$
\frac{\partial \phi_{1}}{\partial L}=\frac{W_{2}}{n} S_{Y 2}^{2}+\frac{W_{2}}{n}\left(R^{2} S_{X 2}^{2}-2 \rho_{2} R S_{X 2} S_{Y 2}\right)-\mu n c_{2} \frac{W_{2}}{L^{2}}=0
$$

From equations (23), (24), (25) and (26), we respectively get

$$
\begin{gather*}
n^{\prime}=\frac{\sqrt{2 \rho R S_{X} S_{Y}-R^{2} S_{X}^{2}+\left(L^{\prime}-1\right) W_{2}\left(2 \rho_{2} R S_{X 2} S_{Y 2}-R^{2} S_{X 2}^{2}\right)}}{\sqrt{\mu\left(c^{\prime}+c_{1}^{\prime} W_{1}+c_{2}^{\prime} \frac{W_{2}}{L^{\prime}}\right)}}  \tag{27}\\
n=\frac{\sqrt{\left(S_{Y}^{2}+R^{2} S_{X}^{2}-2 \rho R S_{X} S_{Y}\right)+(L-1) W_{2}\left(S_{Y 2}^{2}+R^{2} S_{X 2}^{2}-2 \rho_{2} R S_{X 2} S_{Y 2}\right)}}{\sqrt{\mu\left(c+c_{1} W_{1}+c_{2} \frac{W_{2}}{L}\right)}}  \tag{28}\\
\sqrt{\mu}=\frac{L^{\prime} \sqrt{\left(2 \rho_{2} R S_{X 2} S_{Y 2}-R^{2} S_{X 2}^{2}\right)}}{n^{\prime} \sqrt{c_{2}^{\prime}}} \tag{29}
\end{gather*}
$$

and

$$
\begin{equation*}
\sqrt{\mu}=\frac{L \sqrt{\left(S_{Y 2}^{2}+R^{2} S_{X 2}^{2}-2 \rho_{2} R S_{X 2} S_{Y 2}\right)}}{n \sqrt{c_{2}}} \tag{30}
\end{equation*}
$$

Substituting $\sqrt{\mu}$ from equation (29) into the equation (27), we get optimum value of $L^{\prime}$ as

$$
\begin{equation*}
L_{o p t}^{\prime}=\frac{D \sqrt{c_{2}^{\prime}}}{A B} \tag{31}
\end{equation*}
$$

where $A=\sqrt{c^{\prime}+c_{1}^{\prime} W_{1}}, B=\sqrt{\left(2 \rho_{2} R S_{X 2} S_{Y 2}-R^{2} S_{X 2}^{2}\right)}$ and

$$
D=\sqrt{\left(2 \rho R S_{X} S_{Y}-R^{2} S_{X}^{2}\right)+W_{2}\left(R^{2} S_{X 2}^{2}-2 \rho_{2} R S_{X 2} S_{Y 2}\right)}
$$

Substituting the value of $L_{\text {opt }}^{\prime}$ from equation (31) into equation (27), we get

$$
\begin{equation*}
n^{\prime}=\frac{\sqrt{D^{2}+\frac{D \sqrt{c_{2}^{\prime}} B W_{2}}{A}}}{\sqrt{\mu \sqrt{A^{2}+\frac{\sqrt{c_{2}^{\prime}} W_{2} A B}{D}}}} . \tag{32}
\end{equation*}
$$

Putting $\sqrt{\mu}$ from equation (30) into the equation (28), we have

$$
\begin{equation*}
L_{o p t}=\frac{\sqrt{c_{2}} D^{\prime}}{A^{\prime} B^{\prime}} \tag{33}
\end{equation*}
$$

where $A^{\prime}=\sqrt{c+c_{1} W_{1}}, B^{\prime}=\sqrt{\left(S_{Y 2}^{2}+R^{2} S_{X 2}^{2}-2 \rho_{2} R S_{X 2} S_{Y 2}\right)}$ and

$$
D^{\prime}=\sqrt{\left(S_{Y}^{2}+R^{2} S_{X}^{2}-2 \rho R S_{X} S_{Y}\right)-W_{2}\left(S_{Y 2}^{2}+R^{2} S_{X 2}^{2}-2 \rho_{2} R S_{X 2} S_{Y 2}\right)}
$$

On substituting the value of $L_{\text {opt }}$ from equation (33) into the equation (28), we get

$$
\begin{equation*}
n=\frac{\sqrt{D^{\prime 2}+\frac{\sqrt{c_{2} D^{\prime} B^{\prime} W_{2}}}{A^{\prime}}}}{\sqrt{\mu} \sqrt{A^{\prime 2}+\frac{\sqrt{c_{2} W_{2} A^{\prime} B^{\prime}}}{D^{\prime}}}} . \tag{34}
\end{equation*}
$$

To obtain the value of $\sqrt{\mu}$ in terms of total $\operatorname{cost} C_{0}$, we put the values of $n^{\prime}, L^{\prime}, n$ and $L$ respectively from the equations (32), (31), (34) and (33) into equation (19), we get

$$
\begin{equation*}
\sqrt{\mu}=\frac{1}{C_{0}}\left[\sqrt{D^{2}+\frac{D \sqrt{c_{2}^{\prime}} B W_{2}}{A}} \sqrt{A^{2}+\frac{A \sqrt{c_{2}^{\prime}} B W_{2}}{D}}+\sqrt{D^{\prime 2}+\frac{\sqrt{c_{2} D^{\prime} B^{\prime} W_{2}}}{A^{\prime}}} \sqrt{A^{\prime 2}+\frac{\sqrt{c_{2}} A^{\prime} B^{\prime} W_{2}}{D^{\prime}}}\right] \tag{35}
\end{equation*}
$$

On substituting the value of $\sqrt{\mu}$ from equation (35) into the equations (32) and (34), we respectively get the optimum values of $n^{\prime}$ and $n$ as

$$
\begin{equation*}
n_{o p t}^{\prime}=\frac{C_{0} \sqrt{D^{2}+\frac{D \sqrt{c_{2}^{\prime}} B W_{2}}{A}}}{\sqrt{A^{2}+\frac{\sqrt{c_{2}^{\prime}} W_{2} A B}{D}}\left[\sqrt{\left(D^{2}+\frac{D \sqrt{c_{2}^{\prime}} B W_{2}}{A}\right)\left(A^{2}+\frac{\sqrt{c_{2}^{\prime} W_{2} A B}}{D}\right)}+\sqrt{\left(D^{\prime 2}+\frac{\sqrt{c_{2} D^{\prime} B^{\prime} W_{2}}}{A^{\prime}}\right)\left(A^{\prime 2}+\frac{\sqrt{c_{2} A^{\prime} B^{\prime} W_{2}}}{D^{\prime}}\right)}\right]} \tag{36}
\end{equation*}
$$

and

$$
\begin{equation*}
n_{o p t}=\frac{C_{0} \sqrt{D^{\prime 2}+\frac{\sqrt{c_{2} D^{\prime} B^{\prime} W_{2}}}{A^{\prime}}}}{\sqrt{A^{\prime 2}+\frac{\sqrt{c_{2} W_{2} A^{\prime} B^{\prime}}}{D^{\prime}}}\left[\sqrt{\left(D^{2}+\frac{D \sqrt{c_{2}^{\prime}} B W_{2}}{A}\right)\left(A^{2}+\frac{\sqrt{c_{2}^{\prime} W_{2} A B}}{D}\right)}+\sqrt{\left(D^{\prime 2}+\frac{\sqrt{c_{2} D^{\prime} B^{\prime} W_{2}}}{A^{\prime}}\right)\left(A^{\prime 2}+\frac{\sqrt{c_{2}} A^{\prime} B^{\prime} W_{2}}{D^{\prime}}\right)}\right]} \tag{37}
\end{equation*}
$$

## Case (ii): (Under Estimator ${ }_{2}^{\prime *}$ )

Now, we differentiate the equation (21) with respect to $n^{\prime}, n, L^{\prime}$ and $L$ respectively, and equate the derivatives to zero. Thus, we get

$$
\begin{gather*}
\frac{\partial \phi_{2}}{\partial n^{\prime}}=\frac{1}{n^{\prime 2}}\left[R^{2} S_{X}^{2}+2 \rho R S_{X} S_{Y}+\left(L^{\prime}-1\right) W_{2}\left(R^{2} S_{X 2}^{2}+2 \rho_{2} R S_{X 2} S_{Y 2}\right)\right]+\mu\left(c^{\prime}+c_{1}^{\prime} W_{1}+c_{2}^{\prime} \frac{W_{2}}{L^{\prime}}\right)=0  \tag{38}\\
\frac{\partial \phi_{2}}{\partial n}=-\frac{1}{n^{2}}\left[\left(S_{Y}^{2}+R^{2} S_{X}^{2}+2 \rho R S_{X} S_{Y}\right)+(L-1) W_{2}\left(S_{Y 2}^{2}+R^{2} S_{X 2}^{2}+2 \rho_{2} R S_{X 2} S_{Y 2}\right)\right]+\mu\left(c+c_{1} W_{1}+c_{2} \frac{W_{2}}{L}\right)=0  \tag{39}\\
\frac{\partial \phi_{2}}{\partial L^{\prime}}=\frac{W_{2}}{n^{\prime}}\left(R^{2} S_{X 2}^{2}+2 \rho_{2} R S_{X 2} S_{Y 2}\right)+\mu n^{\prime} c_{2}^{\prime} \frac{W_{2}}{L^{\prime 2}}=0 \tag{40}
\end{gather*}
$$

and

$$
\begin{equation*}
\frac{\partial \phi_{2}}{\partial L}=\frac{W_{2}}{n} S_{Y 2}^{2}+\frac{W_{2}}{n}\left(R^{2} S_{X 2}^{2}+2 \rho_{2} R S_{X 2} S_{Y 2}\right)-\mu n c_{2} \frac{W_{2}}{L^{2}}=0 \tag{41}
\end{equation*}
$$

Solving the above equations in the similar manner as in case (i), we respectively get the optimum values of $L^{\prime}, L, n^{\prime}$ and $n$

$$
\begin{align*}
L_{o p t}^{\prime} & =\frac{D^{*} \sqrt{c_{2}^{\prime}}}{A B^{*}}  \tag{42}\\
L_{o p t} & =\frac{\sqrt{c_{2}} D^{\prime *}}{A^{\prime} B^{\prime *}} \tag{43}
\end{align*}
$$

$n_{o p t}^{\prime}=\frac{C_{0} \sqrt{-D^{* 2}-\frac{D^{*} \sqrt{c_{2}^{\prime}} B^{*} W_{2}}{A}}}{\sqrt{A^{2}+\frac{\sqrt{c_{2}^{\prime}} W_{2} A B^{*}}{D^{*}}}\left[\sqrt{\left(-D^{* 2}-\frac{D^{*} \sqrt{c_{2}^{\prime}} B^{*} W_{2}}{A}\right)\left(A^{2}+\frac{\sqrt{c_{2}^{\prime}} W_{2} A B^{*}}{D^{*}}\right)}+\sqrt{\left(D^{\prime * 2}+\frac{\sqrt{c_{2} D^{\prime} * B^{\prime} * W_{2}}}{A^{\prime}}\right)\left(A^{\prime 2}+\frac{\sqrt{c_{2} A^{\prime} B^{\prime} * W_{2}}}{D^{D^{\prime}}}\right)}\right]}$
and
$n_{\text {opt }}=\frac{C_{0} \sqrt{D^{\prime} * 2+\frac{\sqrt{c_{2} D^{\prime} * B^{\prime} * W_{2}}}{A^{\prime}}}}{\sqrt{A^{\prime 2}+\frac{\sqrt{c_{2}} W_{2} A^{\prime} B^{\prime} *}{D^{\prime} *}}\left[\sqrt{\left(-D^{* 2}-\frac{D^{*} \sqrt{c_{2}^{\prime}} B^{*} W_{2}}{A}\right)\left(A^{2}+\frac{\sqrt{c_{2}^{\prime}} W_{2} A B^{*}}{D^{*}}\right)}+\sqrt{\left(D^{\prime} * 2+\frac{\sqrt{c_{2} D^{\prime} * B^{\prime} * W_{2}}}{A^{\prime}}\right)\left(A^{\prime 2}+\frac{\sqrt{c_{2} A^{\prime} B^{\prime} * W_{2}}}{D^{\prime} *}\right)}\right]}$
where

$$
\begin{gather*}
B^{*}=\sqrt{\left(R^{2} S_{X 2}^{2}+2 \rho_{2} R S_{X 2} S_{Y 2}\right)}  \tag{45}\\
D^{*}=\sqrt{\left(R^{2} S_{X}^{2}+2 \rho R S_{X} S_{Y}\right)-W_{2}\left(R^{2} S_{X 2}^{2}+2 \rho_{2} R S_{X 2} S_{Y 2}\right)} \\
B^{*}=\sqrt{\left(S_{Y 2}^{2}+R^{2} S_{X 2}^{2}+2 \rho_{2} R S_{X 2} S_{Y 2}\right)}
\end{gather*}
$$

and

$$
D^{\prime *}=\sqrt{\left(S_{Y}^{2}+R^{2} S_{X}^{2}+2 \rho R S_{X} S_{Y}\right)-W_{2}\left(S_{Y 2}^{2}+R^{2} S_{X 2}^{2}+2 \rho_{2} R S_{X 2} S_{Y 2}\right)}
$$

## Case (iii): (Under Estimator $T_{3}{ }^{*}$ )

Using equation (22) and following the procedure adopted in the cases (i) and (ii), we can respectively obtain the optimum values of $L^{\prime}, L, n^{\prime}$ and $n$ for the estimator $T_{3}^{\prime *}$ as

$$
\begin{gather*}
L_{o p t}^{\prime}=\frac{E \sqrt{c_{2}^{\prime}}}{A F}  \tag{46}\\
L_{o p t}=\frac{\sqrt{c_{2}} E^{*}}{A^{\prime} F^{*}}  \tag{47}\\
n_{o p t}^{\prime}=\frac{C_{0} \sqrt{E^{2}+\frac{E \sqrt{c_{2}^{\prime} F W_{2}}}{A}}}{\sqrt{A^{2}+\frac{\sqrt{c_{2}^{\prime}} W_{2} A F}{E}}\left[\sqrt{\left(E^{2}+\frac{E \sqrt{c_{2}^{\prime}} F W_{2}}{A}\right)\left(A^{2}+\frac{\sqrt{c_{2}^{\prime}} W_{2} A F}{E}\right)}+\sqrt{\left(E^{* 2}+\frac{\sqrt{c_{2} E^{*} F^{*} W_{2}} A^{\prime}}{E}\right)\left(A^{\prime 2}+\frac{\sqrt{c_{2} A^{\prime} F^{*} W_{2}}}{E^{*}}\right)}\right]} \tag{8}
\end{gather*}
$$

and
$n_{\text {opt }}=\frac{C_{0} \sqrt{E^{* 2}+\frac{\sqrt{c_{2}} E^{*} F^{*} W_{2}}{A^{\prime}}}}{\sqrt{A^{\prime 2}+\frac{\sqrt{c_{2}} W_{2} A^{\prime} F^{*}}{E^{*}}}\left[\sqrt{\left(E^{2}+\frac{E \sqrt{c_{2}^{\prime}} F W_{2}}{A}\right)\left(A^{2}+\frac{\sqrt{c_{2}^{\prime}} W_{2} A F}{E}\right)}+\sqrt{\left(E^{* 2}+\frac{\sqrt{c_{2}} E^{*} F^{*} W_{2}}{A^{\prime}}\right)\left(A^{\prime 2}+\frac{\sqrt{c_{2}} A^{\prime} F^{*} W_{2}}{E^{*}}\right)}\right]}$
where $E=\sqrt{\rho^{2} S_{Y}^{2}+W_{2}\left(\beta^{2} S_{X 2}^{2}-2 \beta \rho_{2} S_{X 2} S_{Y 2}\right)}, \quad F=\sqrt{-\beta^{2} S_{X 2}^{2}+2 \beta \rho_{2} S_{X 2} S_{Y 2}}$,
$E^{*}=\sqrt{\left(1-\rho^{2}\right) S_{Y}^{2}-W_{2}\left(S_{Y 2}^{2}+\beta^{2} S_{X 2}^{2}-2 \beta \rho_{2} S_{X 2} S_{Y 2}\right)}$ and $F^{*}=\sqrt{\left(S_{Y 2}^{2}+\beta^{2} S_{X}^{2}-2 \beta \rho_{2} S_{X 2} S_{Y 2}\right)}$ $E^{*}=\sqrt{\left(1-\rho^{2}\right) S_{Y}^{2}-W_{2}\left(S_{Y 2}^{2}+\beta^{2} S_{X 2}^{2}-2 \beta \rho_{2} S_{X 2} S_{Y 2}\right)}$ and $F^{*}=\sqrt{\left(S_{Y 2}^{2}+\beta^{2} S_{X 2}^{2}-2 \beta \rho_{2} S_{X 2} S_{Y 2}\right)}$.

## 4 Numerical Study

To demonstrate the theoretical study, we have considered the two different data sets. One is used for showing the performances of the estimators $T_{1}^{\prime *}$ and $T_{3}^{\prime *}$ where the study and auxiliary variables are positively correlated. To show the performances of the estimators $T_{2}{ }^{*}$ and $T_{3}{ }^{*}$, another data set is considered where the study and auxiliary variables are negatively correlated.

## Data Set 1:

Here we have considered the data used by Srivastava [10]. The data consist of the population of seventy villages in a Tehsil of India along with their cultivated area (in acres) in 1981. The cultivated area (in acres) and the population are respectively considered as study and auxiliary variables. The particulars of the population are given below:
$N=70, n^{\prime}=40, n=25, \bar{Y}=981.29, \bar{X}=1755.53, S_{Y}=613.66, S_{X}=1406.13, S_{Y 2}=244.11, S_{X 2}=\frac{4}{5} S_{X}=$ $1124.905, \rho=0.778$, and $\rho_{2}=\frac{4}{5} \rho=0.622$.

The table 1 represents the variance of $\bar{y}^{*}, \operatorname{MSE}$ of $T_{1}^{\prime *}, T_{3}^{\prime *}$ and percentage relative efficiency (PRE) of $T_{1}^{\prime *}$ and $T_{3}^{\prime *}$ with respect to the sample mean estimator $\bar{y}^{*}$ for different choices of non-response rate $\left(W_{2}=0.1,0.2,0.3,0.4\right)$ and inverse sampling rates $\left(L^{\prime}=L=1.5,2.0,2.5,3.0\right)$.

Table 1: Variance of $\bar{y}^{*}, \operatorname{MSE}$ of $T_{1}{ }^{*}, T_{3}^{\prime *}$ and PRE of $T_{1}{ }^{*}$ and $T_{3}{ }^{*}{ }^{*}$ with respect to $\bar{y}^{*}$

| $W_{2}$ | $L^{\prime}=L$ | Var./MSE |  |  | $P R E$ |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\bar{y}^{*}$ | $T_{1}^{\prime}{ }^{*}$ | $T_{3}^{\prime}{ }^{*}$ | $T_{1}{ }^{*}$ | $T_{3}{ }^{*}$ |
| 0.1 | 1.5 | 9802.63 | 7964.93 | 6405.94 | 123.07 | 153.02 |
|  | 2.0 | 9921.81 | 8084.10 | 6525.12 | 122.73 | 152.06 |
|  | 2.5 | 10040.99 | 8203.28 | 6644.30 | 122.40 | 151.12 |
|  | 3.0 | 10160.17 | 8322.46 | 6763.48 | 122.08 | 150.22 |
| 0.2 | 1.5 | 9921.81 | 8237.33 | 6547.49 | 120.45 | 151.54 |
|  | 2.0 | 10160.17 | 8475.69 | 6785.84 | 119.87 | 149.73 |
|  | 2.5 | 10398.53 | 8714.05 | 7024.20 | 119.33 | 148.04 |
|  | 3.0 | 10636.88 | 8952.41 | 7262.55 | 118.82 | 146.46 |
| 0.3 | 1.5 | 10040.99 | 8509.74 | 6689.02 | 117.99 | 150.11 |
|  | 2.0 | 10398.53 | 8867.28 | 7046.56 | 117.27 | 147.57 |
|  | 2.5 | 10756.06 | 9224.82 | 7404.10 | 116.60 | 145.27 |
|  | 3.0 | 11113.60 | 9582.36 | 7761.64 | 115.98 | 143.19 |
| 0.4 | 1.5 | 10160.17 | 8782.15 | 6830.57 | 115.69 | 148.75 |
|  | 2.0 | 10636.88 | 9258.87 | 7307.29 | 114.88 | 145.57 |
|  | 2.5 | 11113.60 | 9735.59 | 7784.00 | 114.15 | 142.77 |
|  | 3.0 | 11590.32 | 10212.31 | 8260.72 | 113.49 | 140.31 |

## Data Set 2:

In this data set, we have used the data considered by Chaudhary and Shukla [11]. The data relate to the scores elicited by a diagnostic technique, called AMDN and the forced expiratory volume ( $\mathrm{FEV}_{1}$ ) scores for twenty two patients. Here, AMDN and $\mathrm{FEV}_{1}$ are respectively considered as study and auxiliary variables. The population parameters are given below:
$N=22, n^{\prime}=15, n=9, \bar{Y}=1.76, \bar{X}=78.18, S_{Y}^{2}=0.1212, S_{X}^{2}=631.83, S_{Y 2}^{2}=0.1041, S_{X 2}^{2}=\frac{4}{5} S_{X}^{2}=505.46$, $\rho=-0.877$, and $\rho_{2}=\frac{4}{5} \rho=-0.705$.

The table 2 depicts the variance of $\bar{y}^{*}, \operatorname{MSE}$ of $T_{2}{ }^{*}, T_{3}{ }^{*}$ and PRE of the estimators $T_{2}{ }^{*}$ and $T_{3}{ }^{*}{ }^{*}$ with respect to the sample mean estimator $\bar{y}^{*}$ for different choices of non-response rate ( $W_{2}=0.1,0.2,0.3,0.4$ ) and inverse sampling $\operatorname{rates}\left(L^{\prime}=L=1.5,2.0,2.5,3.0\right)$.

Table 2: Variance of $\bar{y}^{*}$, MSE of $T_{2}{ }^{*}, T_{3}^{\prime *}$ and PRE of $T_{2}{ }^{*}$ and $T_{3}{ }^{*}$ with respect to $\bar{y}^{*}$

| $W_{2}$ | $L^{\prime}=L$ | Var./MSE |  |  | PRE |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\bar{y}^{*}$ | $T_{2}^{\prime}{ }^{*}$ | $T_{3}{ }^{*}$ | $T_{2}{ }^{*}$ | $T_{3}{ }^{*}$ |
| 0.1 | 1.5 | 0.0085 | 0.0075 | 0.0043 | 114.16 | 199.26 |
|  | 2.0 | 0.0091 | 0.0081 | 0.0049 | 113.14 | 187.45 |
|  | 2.5 | 0.0097 | 0.0086 | 0.0054 | 112.26 | 178.16 |
|  | 3.0 | 0.0103 | 0.0092 | 0.0060 | 111.49 | 170.65 |
| 0.2 | 1.5 | 0.0091 | 0.0081 | 0.0048 | 112.30 | 191.75 |
|  | 2.0 | 0.0103 | 0.0093 | 0.0059 | 110.77 | 173.79 |
|  | 2.5 | 0.0114 | 0.0104 | 0.0071 | 109.57 | 161.72 |
|  | 3.0 | 0.0126 | 0.0116 | 0.0082 | 108.62 | 153.03 |
| 0.3 | 1.5 | 0.0097 | 0.0088 | 0.0052 | 110.71 | 185.59 |
|  | 2.0 | 0.0114 | 0.0105 | 0.0070 | 108.94 | 164.25 |
|  | 2.5 | 0.0132 | 0.0122 | 0.0087 | 107.67 | 151.43 |
|  | 3.0 | 0.0149 | 0.0140 | 0.0104 | 106.72 | 142.87 |
| 0.4 | 1.5 | 0.0103 | 0.0094 | 0.0057 | 109.34 | 180.45 |
|  | 2.0 | 0.0126 | 0.0117 | 0.0080 | 107.50 | 157.20 |
|  | 2.5 | 0.0149 | 0.0140 | 0.0103 | 106.26 | 144.38 |
|  | 3.0 | 0.0172 | 0.0163 | 0.0126 | 105.37 | 136.25 |

## 5 Conclusion

In the present paper, we have proposed the ratio, product and regression type estimators utilizing the information on an auxiliary variable under the situations in which both study and auxiliary variables are suffered from non-response. The biases and mean square errors of the proposed estimators have been obtained. The theoretical study on the cost of the survey has been presented. The optimum values of the first phase sample sizen ${ }^{\prime}$, second phase sample size $n$ and inverse sampling rates at first phase and second phase i. e. $L^{\prime}$ and $L$ have been obtained under each estimator. The table 1 represents PRE of the proposed estimators $T_{1}^{\prime *}$ and $T_{3}^{\prime *}$ with respect to sample mean estimator $\bar{y}^{*}$ for the different choices of non-response rate $W_{2}$ and inverse sampling rates $L^{\prime}$ and $L$ where study and auxiliary variables are positively correlated. Similarly, table 2 depicts PRE of the proposed estimators $T_{2}^{\prime *}$ and $T_{3}^{\prime *}$ with respect to sample mean estimator $\bar{y}^{*}$ for the different choices of non-response rate $W_{2}$ and inverse sampling rates $L^{\prime}$ and $L$ where study and auxiliary variables are negatively correlated. In both the tables, it is seen that the MSE of the suggested estimators increases with the increase in the non-response rate as well as to increase in the inverse sampling rates. The results are also intuitively expected.

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