#### Vol.3.Issue.2.2015



http://www.bomsr.com

**RESEARCH ARTICLE** 

# BULLETIN OF MATHEMATICS AND STATISTICS RESEARCH

A Peer Reviewed International Research Journal



## COMBINATION OF TWO RATIO TYPE ESTIMATOR FOR ESTIMATING POPULATION MEAN USING AUXILIARY VARIABLE WITH DOUBLE SAMPLING IN PRESENCE OF NON-RESPONSE

## **RAJKUMAR HAZRA**

M.Sc. (Indian Institute of Technology Kanpur, UP, India) Guest Lecturer, Department of Statistics, Midnapore College (Autonomus), Midnapore, WB, India



#### ABSTRACT

Combination of two ratio type estimator for estimating population mean using auxiliary variable with double sampling in presence of non-response has been proposed and study its properties like as mean square error (MSE) and bias. A numerical comparative study of the proposed estimator has been made with the relevant estimators for optimum value of  $\alpha$ .

Keywords: Auxiliary Variable, Bias, Double Sampling, MSE, Non-response and Study Variables.

**©KY PUBLICATIONS** 

## INTRODUCTION

In practical situation almost all surveys suffer from non-response. The problem of non-response often happens due to the refusal of the subject, absenteeism and sometimes due to the lack of information. The pioneering work of Hansen and Hurwitz (1946), assumed that a sub-sample of initial non-respondents is recontacted with a more expensive method, suggesting the first attempt by mail questionnaire and the second attempt by a personal interview. In estimating population parameters such as the mean, total or ratio, sample survey experts sometimes use auxiliary information to improve precision of the estimates. This information may be used at the design stage (leading, for instance, to stratification, systematic or probability proportional to size sampling designs), at the estimation stage or at both stages. The literature on sample survey describes a great variety of techniques for using auxiliary information by means of ratio, product and regression methods. Sodipo and Obisesan (2007) have considered the problem of estimating the population mean in the presence of non-response, in sample survey with full response of an auxiliary character x. Other authors such as Cochran (1977), Rao (1986, 1987), Khare and Srivastava (1993, 1995, 1997), Khare and Rehman (2014), Okafor and Lee (2000) and Tabasum and Khan (2004, 2006) and Singh and Kumar (2008a,b) have studied the problem of non-response under double (two-stage) sampling.

In the present paper, we have proposed combination of two ratio type estimator for estimating population mean using auxiliary variable with double sampling in presence of non-response. We have obtained the expressions for bias and mean square errors of the proposed estimators for the fixed first phase sample (n'), second phase sample (n) and also for the optimum values of the constants. A comparative study of the mean square errors (MSE) of the proposed estimators is made with the relevant estimators.

## SAMPLING SCHEME

The double sampling in presence of non-response sampling scheme is that; let a population have N units and the population is divided in two parts say N<sub>1</sub> of responding units and N<sub>2</sub> of non-responding units. In case when population mean of auxiliary information is not known, we first estimate population mean of auxiliary variable(x) is in the first phase sample of size n' (<N) from population of size N by using SRSWOR scheme and estimate population mean by , which is the mean of the values of x on the first phase sample. Further we select a smaller second phase sample of size n is selected from n' by SRSWOR scheme and observe that Non-response occurs on the second phase sample of size n in which n<sub>1</sub> units respond and n<sub>2</sub> units do not for study variable (y) and as well as auxiliary variable (x). From the n<sub>2</sub> non-respondents, by simple random sampling without replacement (SRSWOR) a sample of r = n<sub>2</sub>/k ; k > 1 units is selected where k is the inverse sampling rate at the second phase sample of size n. All the r units respond at this time round.



#### NOTATION AND TERMINOLOGY

In this paper we used sufficient number of notations and terms, they are defined as follows  $(\bar{y}_1, \bar{y}_2)$ be the sample mean based on n<sub>1</sub> and r units of study variable and  $(\bar{x}_1, \bar{x}_2', \bar{x}', \bar{x})$  be the sample mean of auxiliary variable based on  $n_1$ , r, n', n units respectively.  $\overline{Y}$  and  $\overline{X}$  be the population mean of study variable and auxiliary variable based on population size N=N<sub>1</sub>+N<sub>2</sub>. Also  $\overline{Y}_{2}$ ,  $\overline{X}_{2}$  be the population mean of study variable and auxiliary variable based on population size  $N_2$  (non-response part).

$$\begin{split} S_{y}^{2} &= \sum_{i=1}^{N} (y_{i} - \bar{Y})^{2} / (N - 1) , S_{x}^{2} = \sum_{i=1}^{N} (x_{i} - \bar{X})^{2} / (N - 1) , S_{y2}^{2} = \sum_{i=1}^{N_{2}} (y_{i} - \bar{Y}_{2})^{2} / (N_{2} - 1) , \\ S_{x2}^{2} &= \sum_{i=1}^{N_{2}} (x_{i} - \bar{X}_{2})^{2} / (N_{2} - 1) , C_{y}^{2} = S_{y}^{2} / \bar{Y}^{2} , \qquad C_{y2}^{2} = S_{y2}^{2} / \bar{Y}^{2} , \qquad C_{x}^{2} = S_{x}^{2} / \bar{X}^{2} , \qquad C_{x2}^{2} = S_{x2}^{2} / \bar{X}^{2} \\ S_{yx} &= \frac{\sum_{i=1}^{N} (y_{i} - \bar{Y}) (x_{i} - \bar{X})}{N - 1} , S_{yx2} = \frac{\sum_{i=1}^{N_{2}} (y_{i} - \bar{Y}) (x_{i} - \bar{X})}{N_{2} - 1} , \rho_{yx} = \frac{S_{yx}}{S_{y}S_{x}} , \rho_{yx2} = \frac{S_{yx2}}{S_{y2}S_{x2}} , \beta_{yx} = \frac{S_{yx}}{S_{x}^{2}} , \\ \beta_{yx2} &= \frac{S_{yx2}}{S_{x2}^{2}} , K_{yx} = \frac{\rho_{yx}C_{y}}{C_{x}} , K_{yx2} = \frac{\rho_{yx2}C_{y2}}{C_{x2}} , b^{*} = \frac{\hat{S}_{yx}}{\hat{S}_{x}^{2}} , b^{**} = \frac{\hat{S}_{yx}}{s_{x}^{2}} , s_{x}^{2} = \sum_{i=1}^{n} \frac{(x_{i} - \bar{x})^{2}}{n - 1} , C_{yx} = \frac{S_{yx}}{\bar{X}\bar{Y}} , \end{split}$$

 $C_{yx2} = \frac{S_{yx2}}{RT}$ ,  $\hat{S}_{yx}$  and  $\hat{S}_x^2$  are estimates of  $S_{yx}$  and  $S_x^2$  respectively based on  $n_1$ +r units.  $\rho_{yx}$  and  $\rho_{yx2}$ are respectively the correlation coefficient of response and non-response group between study variable y and auxiliary variable x.

$$w_1 = n_1/n, \ w_2 = n_2/n, \ f = n/N, \ W_2 = N_2/N, \ \lambda = (1 - f)/n, \ \lambda' = (1 - f')/n', \ \lambda^* = \frac{W_2(k-1)}{n}, \ f' = n'/N.$$
  
We also use the constant a

We also use the constant a.

#### SOME WELL KNEW ESTIMATORS AND ITS MSE

A usual unbiased estimator for the population mean  $\overline{Y}$  of the study variable y, proposed by Hansen and Hurwitz (1946), is defined by

 $\bar{y}^* = w_1 \bar{y}_1 + w_2 \bar{y}_2'$ 

The variance of  $\overline{y}^*$  is given by

 $Var(\bar{y}^*) = \bar{Y}^2 \left( \lambda C_y^2 + \lambda^* C_{y2}^2 \right)$ 

It is well known that in estimating the population mean, sample survey experts sometimes use auxiliary information to improve the precision of the estimates. Let x denote an auxiliary variable with population mean  $\overline{X}$ . The (1946)estimator Hansen and Hurwitz is  $\bar{x}^* = w_1 \bar{x}_1 + w_2 \bar{x}_2'$ 

The variance of  $\vec{x}^*$  is given by

$$Var(\bar{x}^*) = \bar{X}^2 (\lambda C_x^2 + \lambda^* C_{x2}^2)$$

Khare and Srivastava (1993), Tabasum and Khan's (2004) defined ratio estimator in presence of nonresponse

$$T_{1} = \bar{y}^{*} \frac{\bar{x}'}{\bar{x}'} \text{ and its MSE is}$$
  
$$MSE(T_{1}) = \bar{Y}^{2} \left[ (\lambda - \lambda') \{ C_{y}^{2} + (1 - 2K_{yx})C_{x}^{2} \} + \lambda^{*} \{ C_{y2}^{2} + (1 - 2K_{yx2})C_{x2}^{2} \} + \lambda' C_{y}^{2} \right]$$

Khare and Srivastava (1993), Tabasum and Khan's (2006) defined a ratio type estimator in presence of non-response as  $T_2 = \overline{y}^* \frac{\overline{x}'}{\overline{x}}$  and its MSE is

$$MSE(T_2) = \overline{Y}^2 \left[ (\lambda - \lambda') \{ C_y^2 + (1 - 2K_{yx}) C_x^2 \} + \lambda^* C_{y2}^2 + \lambda' C_y^2 \right]$$

Singh and Kumar's (2008a) defined a ratio type estimator in presence of non-response as

$$T_3 = \bar{y}^* \left( \frac{\bar{x}'}{\bar{x}^*} \right) \left( \frac{\bar{x}'}{\bar{x}} \right)$$

and its MSE is

$$MSE(T_3) = \overline{Y}^2 \left[ (\lambda - \lambda') \{ C_y^2 + 4(1 - K_{yx})C_x^2 \} + \lambda^* \{ C_{y2}^2 + (1 - 2K_{yx2})C_{x2}^2 \} + \lambda' C_y^2 \right]$$

Singh and Kumar's (2008a) defined a product type estimator in presence of non-response as

$$T_4 = \bar{y}^* \left(\frac{\bar{x}^*}{\bar{x}'}\right) \left(\frac{\bar{x}}{\bar{x}'}\right)$$

and its MSE is

$$MSE(T_4) = \overline{Y}^2 \left[ (\lambda - \lambda') \{ C_y^2 + 4(1 - K_{yx})C_x^2 \} + \lambda^* \{ C_{y2}^2 + (1 + 2K_{yx2})C_{x2}^2 \} + \lambda' C_y^2 \right]$$

Singh and Ruiz Espejo (2007) defined an estimator in presence of nonresponse as  $T_5 = \bar{y}^* \left\{ b \frac{\bar{x}'}{\bar{x}^*} + (1-b) \frac{\bar{x}^*}{\bar{x}'} \right\}$ 

where b is any suitably chosen constant and its MSE is

$$D = \{\lambda' S_x^2 + \lambda^* S_{x2}^2\}, \qquad D^* = \{(\lambda - \lambda') K_{yx} S_x^2 + \lambda^* K_{yx2} S_{x2}^2\}$$
$$MSE(T_5) = \left[\lambda' C_y^2 + (\lambda - \lambda') \left\{S_y^2 + \frac{D^*}{D} \left(\frac{D^*}{D} - 2\beta_{yx}\right) S_x^2\right\} + \lambda^* \left\{S_{y2}^2 + \frac{D^*}{D} \left(\frac{D^*}{D} - 2\beta_{yx2}\right) S_{x2}^2\right\}\right]$$

Khare and Srivastava (1995), defined an regression type estimator in presence of non-response

$$\mathbf{T}_6 = \mathbf{\bar{y}}^* + \mathbf{b}^{**}(\mathbf{\bar{x}} - \mathbf{\bar{x}})$$

and its MSE is

$$MSE(T_6) = Var(\bar{y}^*) - \bar{Y}^2 \left(\frac{1}{n} - \frac{1}{n'}\right) \rho_{yx}^2 C_y^2$$

Khare and Rehman (2014) defined generalized ratio in regression type estimators of non-response as

$$T_7 = \overline{y}^* \left(\frac{\overline{x}'}{\overline{x}^*}\right)^a + b^* \left(\overline{x}' - \overline{x}^*\right)$$

and its MSE is

$$\begin{split} \mathsf{MSE}(\mathsf{T}_7) &= \mathsf{Var}(\overline{y}^*) + \left(\frac{1}{n} - \frac{1}{n'}\right) \{ \overline{Y}^2 \alpha^2 \mathsf{C}_x^2 + \overline{X}^2 \mathsf{B}^2 \mathsf{C}_x^2 - 2 \overline{Y}^2 \alpha \mathsf{C}_{yx} - 2 \overline{X} \overline{Y} \mathsf{B} \mathsf{C}_{yx} + 2 \overline{X} \overline{Y} \mathsf{B} \alpha \mathsf{C}_x^2 \} \\ &+ \lambda^* \{ \overline{Y}^2 \alpha^2 \mathsf{C}_{x2}^2 + \overline{X}^2 \mathsf{B}^2 \mathsf{C}_{x2}^2 - 2 \overline{Y}^2 \alpha \mathsf{C}_{yx2} - 2 \overline{X} \overline{Y} \mathsf{B} \mathsf{C}_{yx2} + 2 \overline{X} \overline{Y} \mathsf{B} \alpha \mathsf{C}_{x2}^2 \} \end{split}$$

where

$$a^{opt} = \Big[ \Big( \frac{1}{n} - \frac{1}{n'} \Big) \big( \overline{Y} C_{yx} - \overline{X} B C_x^2 \big) + \lambda^* \big( \overline{Y} C_{yx2} - \overline{X} B C_{x2}^2 \big) \Big] \Big/ \Big[ \Big( \frac{1}{n} - \frac{1}{n'} \Big) \overline{X} C_x^2 + \lambda^* \overline{Y} C_{x2}^2 \Big] \text{ and } B = \frac{\rho_{yx} S_y}{S_x}$$

#### PROPOSED ESTIMATOR WITH BIAS AND MSE

In the given sampling scheme we have proposed combination of two ratio type estimator for estimating population mean using auxiliary variables with double sampling in the presence of non-response, which is given as follows:

$$T_{\rm P} = \alpha T_1 + (1 - \alpha) T_2$$
  
$$T_{\rm P} = \bar{\mathbf{y}}^* \left\{ \alpha \frac{\bar{\mathbf{x}}'}{\bar{\mathbf{x}}^*} + (1 - \alpha) \frac{\bar{\mathbf{x}}'}{\bar{\mathbf{x}}} \right\}$$
(1)

where **a** is constant.

To obtain the bias and variance of the estimators  $T_p$ , we write

$$\overline{y}^* = \overline{Y}(1 + \varepsilon_0), \overline{x}^* = \overline{X}(1 + \varepsilon_1), \overline{x}' = \overline{X}(1 + \varepsilon_1'), \overline{x} = \overline{X}(1 + \varepsilon_2)$$

such that,

$$\begin{split} E(\varepsilon_0) &= E(\varepsilon_1) = E(\varepsilon_1') = E(\varepsilon_2) = 0 \text{ and} \\ E(\varepsilon_0^2) &= \lambda C_y^2 + \lambda^* C_{y2}^2, E(\varepsilon_1^2) = \lambda C_x^2 + \lambda^* C_{x2}^2, E(\varepsilon_1'^2) = \lambda' C_x^2, E(\varepsilon_2^2) = \lambda C_x^2, E(\varepsilon_0\varepsilon_1') = \lambda' \rho_{yx} C_y C_x, \\ E(\varepsilon_0\varepsilon_1) &= \lambda \rho_{yx} C_y C_x + \lambda^* \rho_{yx2} C_{y2} C_{x2}, E(\varepsilon_0\varepsilon_2) = \lambda \rho_{yx} C_y C_x, E(\varepsilon_1\varepsilon_1') = \lambda' C_x^2, E(\varepsilon_1\varepsilon_2) = \lambda C_x^2, \\ E(\varepsilon_2\varepsilon_1') &= \lambda' C_x^2 \end{split}$$

So the estimator  $T_P$  can be expressed in terms of  $\varepsilon$ 's as follows

$$T_p = \bar{Y}(1 + \varepsilon_0)[\alpha(1 + \varepsilon_1')(1 + \varepsilon_1)^{-1} + (1 - \alpha)(1 + \varepsilon_1')(1 + \varepsilon_2)^{-1}]$$
(2)

If we assume that  $|\varepsilon_0| < 1$ ,  $|\varepsilon_1'| < 1$ ,  $|\varepsilon_2| < 1$  then the right hand side of (2) is expandable. Now, expanding the right hand side of (2) to the second degree of approximation, we have

$$T_{p} - \bar{Y} = \bar{Y}(1 + \varepsilon_{0}) \left[ \alpha (1 + \varepsilon_{1}') \left( 1 - \varepsilon_{1} + \frac{\varepsilon_{1}^{2}}{2} \right) + (1 - \alpha) (1 + \varepsilon_{1}') \left( 1 - \varepsilon_{2} + \frac{\varepsilon_{1}^{2}}{2} \right) \right] - \bar{Y}$$

$$T_{p} - \bar{Y} = \bar{Y}(1 + \varepsilon_{0}) \left[ \alpha \left( 1 + \varepsilon_{1}' - \varepsilon_{1} - \varepsilon_{1}\varepsilon_{1}' + \frac{\varepsilon_{1}^{2}}{2} \right) + (1 - \alpha) \left( 1 + \varepsilon_{1}' - \varepsilon_{2} - \varepsilon_{1}'\varepsilon_{2} + \frac{\varepsilon_{1}^{2}}{2} \right) \right] - \bar{Y}$$

$$T_{p} - \bar{Y} = \bar{Y} \left[ \begin{array}{c} \alpha (\varepsilon_{0} + \varepsilon_{1}' - \varepsilon_{1} + \varepsilon_{0}\varepsilon_{1}' - \varepsilon_{0}\varepsilon_{1} - \varepsilon_{1}\varepsilon_{1}') + \\ (1 - \alpha) (\varepsilon_{0} + \varepsilon_{1}' - \varepsilon_{2} + \varepsilon_{0}\varepsilon_{1}' - \varepsilon_{0}\varepsilon_{2} - \varepsilon_{1}'\varepsilon_{2}) + \frac{\varepsilon_{1}^{2}}{2} \right]$$

$$(3)$$

Taking expectation on both side of (3), we get the bias of  $T_p$  to the first degree of approximation is given by

$$B(T_p) = \bar{Y} \begin{bmatrix} \alpha \left( \lambda' \rho_{yx} C_y C_x - \lambda \rho_{yx} C_y C_x - \lambda^* \rho_{yx2} C_{y2} C_{x2} - \lambda' C_x^2 \right) + \\ (1 - \alpha) \left( \lambda' \rho_{yx} C_y S_x - \lambda \rho_{yx} C_y S_x - \lambda' C_x^2 \right) + \begin{pmatrix} \left( \lambda C_x^2 + \lambda^* C_{x2}^2 \right) \\ \left( \lambda' \rho_{yx} C_y C_x - \lambda \rho_{yx} C_y C_x - \lambda' C_x^2 + \left( \frac{\lambda^* C_{x2}^2}{2} \right) \right) - \alpha \left( \lambda^* \rho_{yx2} C_{y2} C_{x2} \right) \end{bmatrix}$$

$$B(T_p) = \bar{Y} \left[ \left( \lambda' \rho_{yx} C_y C_x - \lambda \rho_{yx} C_y C_x - \lambda' C_x^2 + \left( \frac{\lambda^* C_{x2}^2}{2} \right) \right) - \alpha \left( \lambda^* \rho_{yx2} C_{y2} C_{x2} \right) \right]$$

$$(4)$$

So our estimator  $T_{\mbox{\scriptsize P}}$  is approximately unbiased if the value of the constant is

$$\alpha = \left[ \left( \left\{ (\lambda' - \lambda) \rho_{yx} C_y C_x \right\} - \lambda' C_x^2 + \left( \frac{\lambda^* C_{x2}^2}{2} \right) \right) / \left( \lambda^* \rho_{yx2} C_{y2} C_{x2} \right) \right].$$

Rewrite (3) we have

$$T_p - \bar{Y} = \bar{Y} \begin{bmatrix} \alpha(-\varepsilon_1 - \varepsilon_0\varepsilon_1 - \varepsilon_1\varepsilon_1') + \\ (\varepsilon_0 + \varepsilon_1' - \varepsilon_2 + \varepsilon_0\varepsilon_1' - \varepsilon_0\varepsilon_2 - \varepsilon_1'\varepsilon_2) - \alpha(-\varepsilon_2 - \varepsilon_0\varepsilon_2 - \varepsilon_1'\varepsilon_2) + \frac{\varepsilon_1^2}{2} \end{bmatrix}$$
(5)

Squaring both sides of (5) and neglecting terms of  $\varepsilon$ 's involving power greater than two, we have

$$(T_p - \bar{Y})^2 = \bar{Y}^2 [\alpha(\varepsilon_2 - \varepsilon_1) + (\varepsilon_0 + \varepsilon_1' - \varepsilon_2)]^2 (T_p - \bar{Y})^2 = \bar{Y}^2 [\alpha^2(\varepsilon_2 - \varepsilon_1)^2 + (\varepsilon_0 + \varepsilon_1' - \varepsilon_2)^2 + 2\alpha(\varepsilon_2 - \varepsilon_1)(\varepsilon_0 + \varepsilon_1' - \varepsilon_2)] (T_p - \bar{Y})^2 = \bar{Y}^2 \begin{bmatrix} \alpha^2(\varepsilon_2^2 + \varepsilon_1^2 - 2\varepsilon_1\varepsilon_2) + (\varepsilon_0^2 + \varepsilon_1'^2 + \varepsilon_2^2 + 2\varepsilon_0\varepsilon_1' - 2\varepsilon_0\varepsilon_2 - 2\varepsilon_1'\varepsilon_2) + \\ 2\alpha(\varepsilon_0\varepsilon_2 - \varepsilon_0\varepsilon_1 + \varepsilon_1'\varepsilon_2 - \varepsilon_1\varepsilon_1' - \varepsilon_2^2 + \varepsilon_1\varepsilon_2) \end{bmatrix}$$
(6)

Taking expectation on both sides on (6), we get the MSE of the estimator  $T_p$  to the first degree of approximation, we get

$$MSE(T_p) = \bar{Y}^2 \begin{bmatrix} \alpha^2 (\lambda C_x^2 + \lambda C_x^2 + \lambda^2 C_{x2}^2 - 2\lambda C_x^2) + \\ (\lambda C_y^2 + \lambda^* C_{y2}^2 + \lambda' C_x^2 + \lambda C_x^2 + 2\lambda' \rho_{yx} C_y C_x - 2\lambda \rho_{yx} C_y C_x - 2\lambda' C_x^2) + \\ 2\alpha (\lambda \rho_{yx} C_y C_x - \lambda \rho_{yx} C_y C_x - \lambda^* \rho_{yx2} C_{y2} C_{x2} + \lambda' C_x^2 - \lambda' C_x^2 - \lambda C_x^2 + \lambda C_x^2) \end{bmatrix}$$

$$MSE(T_p) = \bar{Y}^2 \begin{bmatrix} \alpha^2 (\lambda^* C_{x2}^2) + (\lambda C_y^2 + \lambda^* C_{y2}^2 - (\lambda' - \lambda) C_x^2 + 2(\lambda' - \lambda) \rho_{yx} C_y C_x) - \\ 2\alpha (\lambda^* \rho_{yx2} C_{y2} C_{x2}) \end{bmatrix}$$
(7)

The MSE (7) is minimized for

$$\alpha = \frac{\left(\lambda^* \rho_{yx2} C_{y2} C_{x2}\right)}{\lambda^* C_{x2}^2}$$

Hence the optimal value of  $\pmb{\alpha}$  is

$$\alpha^{opt} = \frac{\left(\rho_{yx2} C_{y2}\right)}{C_{x2}}$$

The optimal variance is  $MSE(T_p)^{opt} = \bar{Y}^2 \Big[ \Big( \lambda C_y^2 + \lambda^* C_{y2}^2 - (\lambda' - \lambda) C_x^2 + 2(\lambda' - \lambda) \rho_{yx} C_y C_x \Big) - \lambda^* \rho_{yx2}^2 C_{y2}^2 \Big]$ 

## **EMPERICAL STUDY**

To illustrate the results we considered the data earlier consider by Khare and Sinha(2009), Khare and Rehman (2014). The description of the population is given below:

Here we study 96 village wise population of rural area under Police-station – Singur, District -Hooghly, West Bengal from the District Census Handbook 1981. The 25% villages (i.e. 24 villages) whose area is greater than 160 hectares have been considered as non-response group of the population. The number of agricultural labors in the village is taken as study character (y) while the area (in hectares) of the village, the number of cultivators in the village and the total population of the village are taken as auxiliary characters x and z respectively. The values of the parameters of the population under study are as follows:

Term	Value	Term	Value	Term	Value	Term	Value
Ŧ	137.9271	C <sub>y2</sub>	4.34246	Syx2	28362.05463	Í	0.729167
$\overline{Y}^2$	19023.88	C <sup>2</sup> <sub>x2</sub>	0.885105	C <sub>yx2</sub>	1.419396	λ	0.014583
X	144.872	$\boldsymbol{\rho}_{yx}$	0.773	K <sub>yx2</sub>	1.603649	a'	0.003869
X2	20987.9	$\rho_{yx}^2$	0.597529	N	96	λ – λ΄	0.010714
S <sub>y</sub> <sup>2</sup>	33306.69	S <sub>yx</sub>	16585.1	n	40	В	1.19998
C <sub>y</sub> <sup>2</sup>	1.750783	Cyx	0.830012	'n	70	B <sup>2</sup>	1.43994
S <sub>x</sub> <sup>2</sup>	13821.21	Kyx	1.260396	$\frac{1}{n} - \frac{1}{n}$	0.010714	$\beta_{yx}$	1.19997
C <sub>x</sub> <sup>2</sup>	0.658532	$\rho_{yx2}$	0.724	f	0.416667	$\beta_{yx2}$	1.52677
	k=2	k=3	k=4	k=5		Sy	182.5012
λ•	0.00625	0.0125	0.01875	0.025		Sy2	287.4202
D	169.5779727	285.6811	401.7841	517.8871711		С <sub>у2</sub> .	2.08386
D*	208.9057936	231.167	253.42746	275.6882877		C <sub>x</sub>	0.8115
$\frac{D^*}{D}$	1.231915857	0.80918	0.630756	0.532332722		S <sub>x</sub>	117.5636
а	0.146709	0.20556	0.237288	0.257133		C <sub>x2</sub>	0.9408
a²	0.021524	0.042255	0.056306	0.066117		S <sub>x2</sub>	136.2956

MSE					Relative efficiency(%)					
Estimator	k=2	k=3	k=4	k=5	Estimator	k=2	k=3	k=4	k=5	
<i>y</i> *	1002.0	1518.4	2034.7	2551.0	<i>y</i> *	100	100	100	100	
T <sub>1</sub>	565.6	849.6	1133.7	1417.7	T <sub>1</sub>	177	179	179	180	
T <sub>2</sub>	797.9	1314.2	1830.5	2346.9	T <sub>2</sub>	126	116	111	109	
T <sub>3</sub>	629.9	914.0	1198.0	1482.0	<b>T</b> <sub>3</sub>	159	166	170	172	
<b>T</b> <sub>4</sub>	1305.0	2264.1	3223.2	4182.3	<b>T</b> <sub>4</sub>	77	67	63	61	
T <sub>5</sub>	399.6	777.2	1208.3	1651.6	<b>T</b> 5	251	195	168	154	
Т <sub>6</sub>	788.8	1305.1	1821.4	2337.8	<b>T</b> <sub>6</sub>	127	116	112	109	
T <sub>7</sub>	568.3	824.9	1069.4	1308.7	T <sub>7</sub>	176	184	190	195	
T <sub>p</sub>	527.3	772.9	1018.6	1264.3	T <sub>p</sub>	190	196	200	202	

Table-2: Relative efficiency of the estimators (in %) and MSE with respect to  $\overline{y}^*$  for fixed values of n, n and different values of k (N =96, n = 70 and n =40).

The figure-1 and table-2 shows that

- a) The relative efficiency (%) of the estimators  $T_2$ ,  $T_4$ ,  $T_5$ ,  $T_6$  decrease as the increase of the value of k.
- b) The relative efficiency (%) of the estimators  $T_1$ ,  $T_3$ ,  $T_7$ ,  $T_p$  increase as the increase of the value of k.
- c) Based on the empirical study proposed estimator is more efficient than the other estimator.
- d) Also we see that at k=2 the estimator  $T_5$  is more efficient than the proposed estimator  $T_p$ , but after k>2 i.e., increasing of k the proposed estimator goes to more efficient than the  $T_5$ .

Therefore the proposed estimator should be preferred for the estimation of population mean using auxiliary variable with double sampling in presence of non-response.



Figure-1: Relative efficiency (%) with respect to different estimator.

#### REFERENCES

- Chand, L. (1975). Some ratio-type estimators based on two or more auxiliary variables. Ph.D.
   Thesis submitted to Iowa State University, Ames, IOWA.
- [2] Cohran, W. G. (1977). Sampling Techniques, 3rd Ed. New York: Wiley.
- [3] Das, P.R. and Mishra, G. (2011). An improved class of estimators in two phases sampling using two auxiliary variables. Commun. Stats.-Theo. Meth. 40(24), pp. 4347-4352.
- [4] Hansen, M. H. and Hurwitz, W. N. (1946). The problem of non-response in sample surveys. Jour. Amer. Assoc, 41, pp. 517-529.
- [5] Khare, B. B. and Srivastava, S. (1993). Estimation of population mean using auxiliary character in presence of non-response. Nat. Acad. Sci. Letters India, 16(3), pp. 111-114.
- [6] Khare, B. B. and Srivastava, S. (1995). Study of conventional and alternative two phase sampling ratio, product and regression estimators in presence of non-response. Proc. Nat. Acad. Sci. India, 65(A) II, pp. 195-203.
- [7] Khare, B. B. and Srivastava, S. (1996). Transformed product type estimators for population mean in presence of softcore observations. Proc. Math. Soc. B.H.U., 12, pp. 29-34.
- [8] Khare, B. B. and Srivastava, S. (1997). Transformed ratio type estimators for population mean in the presence of nonresponse. Commun. Statist. Theory Meth. 26(7), pp. 1779-1791.
- [9] Khare, B. B., Kumar, Anupam, Sinha, R. R. and Pandey, S. K. (2008). Two phase sampling estimators for population mean using auxiliary character in presence of non-response in sample surveys. Jour. Sci. Res. BHU, 52, pp. 271-281.
- [10] Khare, B. B. and Kumar, S. (2009). Transformed two phase sampling ratio and product type estimators for population mean in the presence of non-response. Aligarh J. Stats., 29, pp. 91-106.
- [11] Khare, B. B. and Sinha, R.R.(2009). On class of estimators for population means using multiauxiliary characters in presence of non-response. Stats. in Transition-new series.(Poland),10(1), pp. 3-14.
- [12] Khare, B. B. and Srivastava, S. (2010). Generalized two phase sampling estimators for the population mean in the presence of non-response. Aligarh. J. Stats., 30, pp. 39-54.
- [13] Khare, B. B. and Kumar, S. (2010). Chain regression type estimators using additional auxiliary variable in two phase sampling in the presence of non-response. Nat. Acad. Sci. Lett. India, 33(11&12), pp. 369-375.
- [14] Khare, B. B., Srivastava, U. and Kumar, Kamlesh (2011). Generalized chain estimators for the population mean in the presence of non-response. Proc. Nat. Acad. Sci. India 81(A), pt III, pp. 231-238.
- [15] Khare, B. B. and Rehman, H. U.(2014). A generalized ratio in regression type estimator for population mean using auxiliary variables in the presence of non-response. Int. Jour. Stat. & Eco., 15(3), pp. 64-68.
- [16] Kiregyera, B.(1980). A chain ratio type estimator in finite population double sampling using two auxiliary variables. Metrika, 27, pp. 217-223.
- [17] Kiregyera, B.(1984). Regression type estimators using two auxiliary variables and model of double sampling from finite populations. Metrika, 31, pp. 215-226.

- [18] Lewis, P. A., Jones, P. W., Polak, J. W. and Tillotson, H. T. (1991). The problem of conversion in method comparison studies. Applied Statistics, 40, 105-112.
- [19] Okafor, F. C. and Lee, H. (2000). Double sampling for ratio and regression estimation with sub sampling the non-respondent. Survey Methodology, 26, 183-188.
- [20] Rao, P. S. R. S. (1986). Ratio estimation with sub-sampling the non-respondents. Survey Methodology, 12(2), pp. 217-230.
- [21] Rao, P. S. R. S. (1987). Ratio and regression Estimates with sub-sampling the nonrespondents. Paper presented at a special contributed session of the International Statistical Association Meeting, September, 2-16, Tokyo, Japan.
- [22] Rao, P. S. R. S. (1990). Regression estimators with sub sampling of non-respondents. In: Gumer E. Liepins and V. R. R. Uppuluri, Marcel Dekker, eds. Data quality control theory and Pragmatics. New York, pp. 191-208.
- [23] Sahoo, J., Sahoo, L. N. and Mohanty, S. (1993). A regression approach to estimation in twophase sampling using two auxiliary variables. Current Science, 65(1), pp. 73-75.
- [24] Singh, A. K., Singh, H. P. and Upadhyaya, L. N. (2001). A generalized chain estimator for finite population mean in two phase sampling. Jour. Ind. Soc. Ag. Statist., 54(3), pp. 370-375.
- [25] Singh, H. P. and Kumar, S. (2008a). Estimation of mean in presence of non-response using two phase sampling scheme. Statistical Papers, DOI10.1007/s00362-008-0140-5.
- [26] Singh, H. P. and Kumar, S. (2008b). A regression approach to the estimation of finite population mean in presence of non-response. Aust. N. Z. J. Stat., 50, 395-408.
- [27] Sodipo, A. A. and Obisesan, K. O. (2007). Estimation of the population mean using difference cum ratio estimator with full response on the auxiliary character. Res. J. Applied Sci., 2, 769-772.
- [28] Srivastava, S. R., Khare, B. B. and Srivastava, S. R. (1990). A generalized chain ratio.
- [29] Srivastava, S. (1993). Some problems on the estimation of population mean using auxiliary character in presence of non-response in sample surveys. Thesis submitted to Banaras Hindu University, Varanasi, India.
- [30] Tabasum, R. and Khan, I. A. (2004). Double sampling for ratio estimation with non-response.J. Ind. Soc. Agril. Statist., 58, 300-306.
- [31] Tabasum, R. and Khan, I. A. (2006). Double sampling ratio estimator for the population mean in presence of non-response. Assam Statist. Review, 20, 73-83.