



## A Novel Class of Ratio Estimators Incorporating Auxiliary Variables in Simple Random Sampling for Estimating Population Mean: An Application to Real-Life Data

**Qamruz Zaman<sup>1</sup>, Bushra Haider<sup>2</sup>, Qasim Shah<sup>3</sup>, Mohammad Anwar<sup>4</sup>, Saima Kousar<sup>5</sup>, Faran Ali<sup>6</sup>, Gohar Ayub<sup>\*7</sup>**

<sup>1,2,3</sup>Department of Statistics, University of Peshawar, Pakistan

<sup>4</sup>Zhejiang Gongshang University (ZJGU)

<sup>5</sup>Saima Kousar<sup>4</sup>College of Home Economics, University of Peshawar, Pakistan

<sup>6</sup>Macquarie University Australia

<sup>7</sup>Department of Mathematics and Statistics, University of Swat, Swat, Pakistan

\*Corresponding author email: Gohar Ayub, Department of Mathematics and Statistics, University of Swat, Swat, Pakistan

Emails : cricsportsresearchgroup@gmail.com

goharayub@uswat.edu.pk

<sup>1</sup>Email: ayanqamar@gmail.com , <sup>2</sup>Email: hyderbushra47@gmail.com,

<sup>3</sup>Email: qasimshah707@gmail.com, <sup>4</sup>Email: anwarmohammad.uos@gmail.com

<sup>5</sup>Email: gcwussaimakousar@gmail.com , <sup>6</sup>Email: faran.ali@students.mq.edu.au

### Abstract

This paper presents a novel class of ratio estimators using non-conventional measures as auxiliary information to estimate the population mean of the study variable under simple random sampling without replacement. Three special cases are discussed, utilizing the coefficient of variation, median, and quartile deviation as auxiliary variables. The bias and mean square error (MSE) of the proposed estimators are derived up to the first order of approximation. Furthermore, theoretical conditions are established to compare the proposed class with existing estimators, supported by real-life data. Numerical results demonstrate that the proposed class of estimators is more efficient than traditional and other ratio-type estimators.

**Keywords:** Coefficient of variation, median, quartile deviation, bias, mean square error."



## Introduction

It is usual practice to use sampling techniques when the population is very large, or it is difficult to obtain information about each and every unit of the population. The sampling technique is the best alternative to study the behavior of the study variable. In sampling techniques, we take a small portion of a population and draw inferences about the whole population. It is very common to estimate the population parameter by its sample statistic, for example, to estimate the population mean, an appropriate estimator is a sample mean which is an unbiased estimator but has a large variance. Lots of researches have been done to derive such an estimator for the population mean which is more efficient than the existing ones. These research areas are concerned to make use of auxiliary information to improve the sample estimate of the population parameter. For example, Kadilar and Cingi [1, 2] produced a ratio estimator in stratified random sampling scheme, Subramani and Kumarapandiyani [3] worked on the modified ratio estimator by using the quartiles as auxiliary variables, Yadav et.al [4] improved the estimation of the population mean by using non-Conventional measures of dispersion, Subzar et.al [5] worked on the new class of efficient ratio estimators, Sisodia and Dwivedi [6] improved the classical ratio estimator by use of a coefficient of variation as an auxiliary variable. Jeelani et.al [7] presented a Modified ratio estimator by using the linear combination of the coefficient of skewness and quartile deviation, for other estimators, we refer to see [8-20]. The most recent contributions to this area of research are presented by Kumar et.al [21], Unal and Kadilar [22], Singhand Usman [23], and Priam [24].

The basic aim of the paper is to produce a new class of ratio estimators which improved the precision of estimators by using the auxiliary information. We obtain MSE of the class of proposed estimators. The class of proposed estimators and other existing estimators were compared at both theoretical conditions and numerical illustrations.

## Notation

The following notation would be of interest in the research work.

$N$  – Size of the population,  $n$  – Size of the sample,  $Y$  – Study variable

$X$  – Auxiliary variable,  $\bar{X}, \bar{Y}$  - Population means,  $\bar{x}, \bar{y}$  - Sample means

$S_x, S_y$  - Population standard deviation,  $C_x, C_y$  - Coefficient of variation

$\rho_{xy}$  - Population Correlation coefficient between  $X$  and  $Y$ ,  $\beta_{2(x)}$  - Kurtosis of auxiliary variable

$M_d$  - Median, Q.D – Quartile deviation, MSE – Mean square error

SRSWOR – Simple random sampling without replacement.



## Review of existing Estimators

The concept of an auxiliary variable was suggested by Cochran [10] for the first time to estimate the population mean of the variable of interest. The classical estimator proposed by [10] is given by

$$\bar{y}_r = \frac{\bar{y}}{\bar{x}} \bar{X} = \hat{R} \bar{X} \quad (1)$$

The MSE of the above proposed estimator up to the first degree of approximation is

$$MSE(\bar{y}_r) \cong \frac{1-f}{n} (R^2 S_x^2 - 2RS_{xy} + S_y^2) \quad (2)$$

where  $f = n/N$ , and  $R = \bar{Y} / \bar{X}$  is the population ratio.

Sisodia and Dwivedi [6] presented a new modified ratio estimator. The estimator is defined by

$$\bar{y}_{SD} = \bar{y} \frac{\bar{X} + C_x}{\bar{x} + C_x} = \frac{\bar{y}}{\bar{x}_{SD}} \bar{X}_{SD} = \hat{R}_{SD} \bar{X}_{SD} \quad (3)$$

The MSE to the first degree of approximation is given as

$$MSE(\bar{y}_{SD}) \cong \frac{1-f}{n} \bar{Y}^2 [C_y^2 + C_x^2 \alpha(\alpha - 2k)] \quad (4)$$

Where  $C_x$  is the coefficient of variation of the auxiliary variable,  $C_y$  is the coefficient of variation of the study variable,  $k = \rho \frac{C_y}{C_x}$  ( $\rho$  is correlation coefficient between Y and X) and  $\alpha = \frac{\bar{x}}{\bar{x} + C_x}$ .

Singh and Kakran [8] defined modified ratio estimator by using the coefficient of Kurtosis as auxiliary variable. The modified estimator is expressed as

$$\bar{y}_{sk} = \bar{y} \frac{\bar{X} + \beta_{2(x)}}{\bar{x} + \beta_{2(x)}} = \frac{\bar{y}}{\bar{x}_{sk}} \bar{X}_{sk} = \hat{R}_{sk} \bar{X}_{sk} \quad (5)$$

MSE corresponding to the above expression up to the first degree of approximation is given by

$$MSE(\bar{y}_{sk}) \cong \frac{1-f}{n} \bar{Y}^2 [C_y^2 + C_x^2 \delta(\delta - 2k)] \quad (6)$$



where  $\delta = \frac{\bar{X}}{\bar{X} + \beta_2}$  and  $k = \rho \frac{C_y}{C_x}$  ( $\rho$  is the coefficient of correlation between Y and X)

The estimator suggested by Upadhyaya and Singh [11] is defined by

$$\bar{y}_{US1} = \bar{y} \frac{\bar{X} \beta_{2(x)} + C_x}{\bar{x} \beta_{2(x)} + C_x} = \frac{\bar{y}}{\bar{x}_{US1}} \bar{X}_{US1} = \hat{R}_{US1} \bar{X}_{US1} \quad (7)$$

$$\bar{y}_{US2} = \bar{y} \frac{\bar{X} C_x + \beta_{2(x)}}{\bar{x} C_x + \beta_{2(x)}} = \frac{\bar{y}}{\bar{x}_{US2}} \bar{X}_{US2} = \hat{R}_{US2} \bar{X}_{US2} \quad (8)$$

The related MSE up to the first degree of approximation is

$$MSE(\bar{y}_{US1}) \cong \frac{1-f}{n} \bar{Y}^2 [C_y^2 + C_x^2 \omega(\omega - 2k)] \quad (9)$$

$$MSE(\bar{y}_{US2}) \cong \frac{1-f}{n} \bar{Y}^2 [C_y^2 + \vartheta(\vartheta - 2k)C_x^2] \quad (10)$$

where  $w$  and  $\vartheta$  are the real numbers which is defined by

$w = \frac{\bar{X} \beta_{2(x)}}{\bar{x} \beta_{2(x)} + C_x}$ ,  $\vartheta = \frac{\bar{X} C_x}{\bar{x} C_x + \beta_{2(x)}}$  and  $k = \rho \frac{c_y}{c_x}$  ( $\rho$  is the coefficient of correlation between Y and X)

## Proposed estimators

We propose a class of efficient ratio estimator and presents three of its special cases by using coefficient of variation, median, and quartile deviation as auxiliary variables. The suggested class of ratio estimators is defined by

$$y_{bi} = \bar{y} \left[ \frac{\bar{X} + \beta_{2(x)}/G(x)}{\bar{x} + \beta_{2(x)}/G(x)} \right] \quad (11)$$

where  $G(x)$  is the general auxiliary variable which would be chosen by choice and  $\beta_{2(x)}$  is the kurtosis of the auxiliary variable.

The special cases of the above general class of estimators respectively then take the following form



$$y_{b1} = \bar{y} \left[ \frac{\bar{X} + \beta_{2(x)}/C_x}{\bar{x} + \beta_{2(x)}/C_x} \right] \quad (12)$$

where  $C_x$  is the coefficient of variation of the auxiliary variable X.

$$y_{b2} = \bar{y} \left[ \frac{\bar{X} + \beta_{2(x)}/M_d}{\bar{x} + \beta_{2(x)}/M_d} \right] \quad (13)$$

where  $M_d$  is the median of the auxiliary variable X.

$$y_{b3} = \bar{y} \left[ \frac{\bar{X} + \beta_{2(x)}/QD}{\bar{x} + \beta_{2(x)}/QD} \right] \quad (14)$$

where  $QD$  is the quartile deviation of the auxiliary variable X.

The bias and mean square error of the proposed estimators up to the first order of approximation is given by

$$Bias(\bar{y}_{bi}) \cong \bar{Y} \left( \frac{1-f}{n} \right) \left( \frac{\theta^2}{2!} C_x^2 - \theta_i \rho_{xy} C_x C_y \right) \quad (15)$$

$$MSE(\bar{y}_{bi}) = \bar{Y}^2 \left( \frac{1-f}{n} \right) \left( C_y^2 + \theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y \right) \quad (16)$$

where  $i = 1, 2, 3$

$$\text{and } \theta_1 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/C_x}, \theta_2 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/M_d}, \theta_3 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/QD}.$$

## Efficiency Comparison

In this section, the proposed class of estimators is compared at theoretical conditions with other existing estimators of the population mean under simple random sampling without replacement scheme.



## Comparison with usual ratio estimator

We compare the MSE of the proposed class of estimators given in Eq (16) with the MSE of the classical ratio estimator given in Eq (2). The proposed estimator performs better than the classical ratio estimator if,

$$\begin{aligned}
 &MSE(\bar{y}_{bi}) < MSE(\bar{y}_r) \\
 &\bar{Y}^2 \left( \frac{1-f}{n} \right) (C_y^2 + \theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y) < \left( \frac{1-f}{n} \right) (R^2 S_x^2 - 2RS_{xy} + S_y^2) \\
 &\bar{Y}^2 (C_y^2 + \theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y) < (R^2 S_x^2 - 2RS_{xy} + S_y^2) \\
 &\bar{Y}^2 (C_y^2 + \theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y) < (\bar{Y}^2 C_x^2 - 2\bar{Y}^2 \rho_{xy} C_x C_y + \bar{Y}^2 C_y^2) \\
 &\bar{Y}^2 (C_y^2 + \theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y) < \bar{Y}^2 (C_x^2 - 2\rho_{xy} C_x C_y + C_y^2) \\
 &\theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y < C_x^2 - 2\rho_{xy} C_x C_y \\
 &\theta_i^2 C_x^2 - C_x^2 < -2\rho_{xy} C_x C_y + 2\theta_i \rho_{xy} C_x C_y \\
 &C_x^2 (\theta_i^2 - 1) < 2\rho_{xy} C_x C_y (-1 + \theta_i) \\
 &C_x (\theta_i - 1)(\theta_i + 1) < 2\rho_{xy} C_y (\theta_i - 1) \\
 &2\rho_{xy} C_y > C_x (\theta_i + 1) \\
 &\rho_{xy} > \frac{C_x (\theta_i + 1)}{2C_y} \tag{17}
 \end{aligned}$$

where,  $i=1,2,3$  and  $\theta_1 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/C_x}$ ,  $\theta_2 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/M_d}$ ,  $\theta_3 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/QD}$ .

## Comparison with Sisodia and Dwivedi 1981 estimator

We compare the MSE of the proposed class of estimators given in Eq (16) with the MSE given in Eq (4). The proposed estimator performs better than the Sisodia and Dwivedi estimator if,



$$\begin{aligned}
 &MSE(\bar{y}_{bi}) < MSE(\bar{y}_{SD}) \\
 &\bar{Y}^2 \left( \frac{1-f}{n} \right) (C_y^2 + \theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y) < \left( \frac{1-f}{n} \right) \bar{Y}^2 [C_y^2 + C_x^2 \alpha(\alpha - 2k)] \\
 &\theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y < C_x^2 \alpha(\alpha - 2k) \\
 &\theta_i^2 C_x - 2\theta_i \rho_{xy} C_y < C_x \alpha(\alpha - 2k) \\
 &\theta_i^2 C_x - 2\theta_i \rho_{xy} C_y < C_x \alpha^2 - 2C_x \alpha k
 \end{aligned}$$

By putting  $k = \rho \frac{C_y}{C_x}$

$$\begin{aligned}
 &\theta_i^2 C_x - 2\theta_i \rho_{xy} C_y < C_x \alpha^2 - 2\alpha \rho_{xy} C_y \\
 &\theta_i^2 C_x - C_x \alpha^2 < -2\alpha \rho_{xy} C_y + 2\theta_i \rho_{xy} C_y \\
 &C_x (\theta_i^2 - \alpha^2) < 2\rho_{xy} C_y (-\alpha + \theta_i) \\
 &C_x (\theta_i^2 - \alpha^2) < 2\rho_{xy} C_y (\theta_i - \alpha) \\
 &C_x (\theta_i - \alpha)(\theta_i + \alpha) < 2\rho_{xy} C_y (\theta_i - \alpha) \\
 &C_x (\theta_i + \alpha) < 2\rho_{xy} C_y \\
 &2\rho_{xy} C_y > C_x (\theta_i + \alpha) \\
 &\rho_{xy} > \frac{C_x}{2C_y} (\theta_i + \alpha) \tag{18}
 \end{aligned}$$

where,  $i=1,2,3$  and  $\theta_1 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/C_x}$ ,  $\theta_2 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/M_d}$ ,  $\theta_3 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/QD}$ ,  $\alpha = \frac{\bar{X}}{\bar{X} + C_x}$ .

### Comparison with Singh and Kakran 1993 estimator

We compare the MSE of the proposed class of estimators given in Eq (16) with the MSE given in Eq (6). The proposed estimator performs better than the Singh and Kakran estimator if,



$$\begin{aligned}
 &MSE(\bar{y}_{bi}) < MSE(\bar{y}_{SK}) \\
 &\bar{Y}^2 \left( \frac{1-f}{n} \right) (C_y^2 + \theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y) < \left( \frac{1-f}{n} \right) \bar{Y}^2 [C_y^2 + C_x^2 \delta(\delta - 2k)] \\
 &\theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y < C_y^2 + C_x^2 \delta(\delta - 2k) \\
 &\theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y < C_x^2 \delta^2 - 2C_x^2 \delta k \\
 &\theta_i^2 C_x - 2\theta_i \rho_{xy} C_y < C_x \delta^2 - 2C_x \delta k
 \end{aligned}$$

By putting  $k = \rho \frac{C_y}{C_x}$

$$\begin{aligned}
 &\theta_i^2 C_x - 2\theta_i \rho_{xy} C_y < C_x \delta^2 - 2C_y \rho_{xy} \delta \\
 &\theta_i^2 C_x - C_x \delta^2 < -2C_y \rho_{xy} \delta + 2\theta_i \rho_{xy} C_y \\
 &C_x (\theta_i^2 - \delta^2) < 2C_y \rho_{xy} (-\delta + \theta_i) \\
 &C_x (\theta_i^2 - \delta^2) < 2C_y \rho_{xy} (\theta_i - \delta) \\
 &C_x (\theta_i - \delta)(\theta_i + \delta) < 2C_y \rho_{xy} (\theta_i - \delta) \\
 &C_x (\theta_i + \delta) < 2C_y \rho_{xy} \\
 &2C_y \rho_{xy} > C_x (\theta_i + \delta) \\
 &\rho_{xy} > \frac{C_x}{2C_y} (\theta_i + \delta) \tag{19}
 \end{aligned}$$

where,  $i=1,2,3$  and  $\theta_1 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/C_x}$ ,  $\theta_2 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/M_d}$ ,  $\theta_3 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/QD}$ ,  $\delta = \frac{\bar{X}}{\bar{X} + \beta_2}$

### Comparison with Upadhyaya and Singh 1999 estimator

We compare the MSE of the proposed class of estimators given in Eq (16) with the MSE given in Eq (8) and (10). The proposed estimator performs better than the Upadhyaya and Singh estimator if,



$$MSE(\bar{y}_{bi}) < MSE(\bar{y}_{US1})$$

$$\bar{Y}^2 \left( \frac{1-f}{n} \right) (C_y^2 + \theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y) < \left( \frac{1-f}{n} \right) \bar{Y}^2 [C_y^2 + C_x^2 \omega(\omega - 2k)]$$

$$(C_y^2 + \theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y) < [C_y^2 + C_x^2 \omega(\omega - 2k)]$$

$$\theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y < C_x^2 \omega(\omega - 2k)$$

$$\theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y < C_x^2 \omega^2 - 2\omega C_x^2 k$$

$$\theta_i^2 C_x - 2\theta_i \rho_{xy} C_y < C_x \omega^2 - 2C_x \omega k$$

By putting  $k = \rho \frac{C_y}{C_x}$

$$\theta_i^2 C_x - 2\theta_i \rho_{xy} C_y < C_x \omega^2 - 2C_y \rho_{xy} \omega$$

$$\theta_i^2 C_x - C_x \omega^2 < -2C_y \rho_{xy} \omega + 2\theta_i \rho_{xy} C_y$$

$$C_x (\theta_i^2 - \omega^2) < 2C_y \rho_{xy} (-\omega + \theta_i)$$

$$C_x (\theta_i - \omega)(\theta_i + \omega) < 2C_y \rho_{xy} (\theta_i - \omega)$$

$$C_x (\theta_i + \omega) < 2C_y \rho_{xy}$$

$$2C_y \rho_{xy} > C_x (\theta_i + \omega)$$

$$\rho_{xy} > \frac{C_x}{2C_y} (\theta_i + \omega) \tag{20}$$

where,  $i=1,2,3$  and  $\theta_1 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/C_x}$ ,  $\theta_2 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/M_d}$ ,  $\theta_3 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/QD}$ ,  $w = \frac{\bar{X} \beta_{2(x)}}{\bar{x} \beta_{2(x)} + C_x}$

### Comparison with Upadhyaya and Singh 1999 estimator

$$MSE(\bar{y}_{bi}) < MSE(\bar{y}_{US2})$$

$$\bar{Y}^2 \left( \frac{1-f}{n} \right) (C_y^2 + \theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y) < \left( \frac{1-f}{n} \right) \bar{Y}^2 [C_y^2 + C_x^2 \vartheta(\vartheta - 2k)]$$

$$(C_y^2 + \theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y) < [C_y^2 + C_x^2 \vartheta(\vartheta - 2k)]$$

$$\theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y < C_x^2 \vartheta(\vartheta - 2k)$$

$$\theta_i^2 C_x^2 - 2\theta_i \rho_{xy} C_x C_y < C_x^2 \vartheta^2 - 2\vartheta C_x^2 k$$

$$\theta_i^2 C_x - 2\theta_i \rho_{xy} C_y < C_x \vartheta^2 - 2C_x \vartheta k$$



By putting  $k = \rho \frac{C_y}{C_x}$

$$\theta_i^2 C_x - 2\theta_i \rho_{xy} C_y < C_x \vartheta^2 - 2C_y \rho_{xy} \vartheta$$

$$\theta_i^2 C_x - C_x \vartheta^2 < -2C_y \rho_{xy} \vartheta + 2\theta_i \rho_{xy} C_y$$

$$C_x (\theta_i^2 - \vartheta^2) < 2C_y \rho_{xy} (-\vartheta + \theta_i)$$

$$C_x (\theta_i - \vartheta)(\theta_i + \vartheta) < 2C_y \rho_{xy} (\theta_i - \vartheta)$$

$$C_x (\theta_i + \vartheta) < 2C_y \rho_{xy}$$

$$2C_y \rho_{xy} > C_x (\theta_i + \vartheta)$$

$$\rho_{xy} > \frac{C_x}{2C_y} (\theta_i + \vartheta) \tag{21}$$

where,  $i=1,2,3$  and  $\theta_1 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/C_x}$ ,  $\theta_2 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/M_d}$ ,  $\theta_3 = \frac{\bar{X}}{\bar{X} + \beta_{2(x)}/QD}$ ,  $\vartheta = \frac{\bar{X}C_x}{\bar{x}C_x + \beta_{2(x)}}$

### Numerical illustration to applied data

For numerical illustration, the data set is taken from Kadilar and Cingi [2]. The data set represents the amount of apple production (as a variable of interest) and the number of apple trees (as an auxiliary variable) in 106 villages of the Aegean Region in 1999. The statistics about the population parameters is given below in table 1. It is noted that the correlation between study and the auxiliary variable is 86% and we take a sample of size  $n=20$ . It is of keen interest that the efficiency of the estimators is independent of a sample size  $n$ .

MSE of the proposed class of estimators as well as other existing estimators are computed and is given in table 2. As a general rule, the estimator with the fewer value of MSE leads to a more precise result. It has been noted that the proposed estimators have fewer MSE than others and hence we obtain satisfactory results.

**Table 1.** Auxiliary information

N=106	n=20	$\bar{Y}=2212.59$	$\bar{X}=27421.70$	$S_{yx}=568176176.10$	$S_y=11551.53$
$C_y=5.22$	$M_d=7297.50$	$QD=12156.25$	$S_x=57460.61$	$C_x=2.10$	$\beta_{2(x)}=34.57$



$\alpha=0.9999$	$\delta=0.9987$	$k=2.1327$	$\omega=0.9999$	$\rho=0.9994$	$\rho = 0.86$
$R=0.0807$	$R_{SD}=0.0807$	$R_{SK}=0.0806$	$R_{US1}=0.0023$	$R_{us2}=0.0385$	

**Table 2.** constant and MSE of the proposed and other estimators.

Estimator	Constant	MSE	PRE
$\bar{y}_r$	-----	2565510.488	100
$\bar{y}_{SD}$	$\alpha=0.9999$ $k=2.1327$	2551970.93	100.53
$\bar{y}_{Sk}$	$\alpha=0.9999$ $\delta=0.9987$	2554353.47	100.437
$\bar{y}_{US1}$	$\omega=0.9999$ $k=2.1327$	2551970.931	100.531
$\bar{y}_{US2}$	$\rho=0.9994$ $k=2.1327$	2552963.345	100.491
$y_{b1}$	$\theta_1=0.999400032$	<b>2544185.217</b>	<b>100.838</b>
$y_{b2}$	$\theta_2=0.9999998$	<b>2542989.056</b>	<b>100.886</b>
$y_{b3}$	$\theta_3=0.9999999$	<b>2542988.851</b>	<b>100.886</b>

**Table 3.** Efficiency Comparison of equation (17),(18),(19),(20),(21)

Equation(17)	$0.86 > 0.4022$	$0.86 > 0.4023$	$0.86 > 0.402298$
Equation(18)	$0.86 > 0.40216$	$0.86 > 0.4023$	$0.86 > 0.402279$
Equation(19)	$0.86 > 0.401917$	$0.86 > 0.40204$	$0.86 > 0.402037$
Equation(20)	$0.86 > 0.4022$	$0.86 > 0.4023$	$0.86 > 0.402298$
Equation(21)	$0.86 > 0.40206$	$0.86 > 0.40218$	$0.86 > 0.40218$

**Table 4:** PRE of the Proposed Estimators with the Estimators in Literature for this population

Estimators	$\bar{y}_r$	$\bar{y}_{SD}$	$\bar{y}_{Sk}$	$\bar{y}_{US1}$	$\bar{y}_{US2}$
$y_{b1}$	100.838	100.306	100.399	100.306	100.345
$y_{b2}$	100.886	100.353	100.447	100.353	100.392



$y_{b3}$	100.886	100.353	100.447	100.353	100.392
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## Result and Conclusion

The paper presents an improved class of ratio estimators for estimating the population mean by having auxiliary information. The class of estimators has been proposed under simple random sampling without replacement scheme. The equations for bias and mean square error have been obtained up to the first order of approximation. For illustration, a real data set has been used and the results reveal that the proposed class of estimators improves the efficiency of estimation than other existing estimators. Moreover, among the proposed estimators, the estimator  $y_{b3}$  performs better than rest of the two estimators.

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