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Modified Ratio Estimator in Survey Sampling using a linear combination

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Abstract

A new modified ratio estimator of population mean of the main variable by means of the linear combination of known values of Co-efficient of Kurtosis and Tri-Mean of the auxiliary variable has been worked out. Mean Square Error (MSE) and Bias of the proposed estimator is calculated and compared with Yan & Tian (2010) estimator. The comparison is demonstrated theoretically and illustrated practically and the conclusion is drawn that the proposed estimator is performing deftly with the above existing estimator.

Keywords: Estimator, Auxiliary variable, Mean square error, Bias, Co-efficient of Kurtosis, Tri-Mean

1. Introduction

The information on main variable \mathbf{Y} is studied with the information on secondary variable \mathbf{X} also known as auxiliary variable. This information on auxiliary variable \mathbf{X} , may be better employed to obtain a more efficient estimator of the population mean. The method is utilized under the condition only if the variables are correlated and do impact on one another. when the population parameters of the auxiliary variable X such as Population Mean, Tri-mean, Coefficient of variation, Kurtosis, Skewness, Correlation coefficient, Median etc., are known, a number of estimators such as ratio, product and linear regression estimators and their modifications are available in literature and are performing better than the simple random sample mean under certain conditions. Among these estimators the ratio estimator and its modifications have widely attracted researchers throughout the world for the estimation of the mean of the study variable (see for example (Kadilar and Cingi, 2004, 2006a, b; Cingi and Kadilar, 2009; Subramani, 2013, Subzar *et al* (2018a, b); Subzar *et al* 2019; Khare and Khare 2019; Priya and Tailor 2019; Neha and Pachori 2019; Priyaranjan and Sunani 2019)^[3, 1, 12, 13, 15, 6, 10, 9, 11]. Further, we know that the values of linear combinations are less affected by the extreme values or presence of outliers in the population values. These facts have motivated the authors to propose modified ratio estimators using the above linear combinations. The basic ratio estimator for population mean \mathbf{F} is defined as:

$$\hat{\bar{Y}}_{R} = \frac{y}{\bar{x}}\bar{X} = \hat{R}\bar{X}, \tag{1.1}$$

Where $\hat{R} = \frac{y}{\bar{x}} = \frac{y}{x}$ is the estimate of $R = \frac{y}{\bar{x}} = \frac{y}{\bar{x}}$ and it is assumed that the population mean \bar{X} of the auxiliary variate X is known. Hence \bar{y} is the sample mean of the variate of the interest and \bar{x} is the sample mean of the auxiliary variate. The bias and the mean square error of \hat{Y}_R to the first degree of approximation is given below:

$$Bias(\bar{\tilde{y}}_{R}) \cong \frac{1-f}{n}\bar{Y}\left(\mathcal{C}_{x}^{2} - 2\mathcal{C}_{x}\mathcal{C}_{y}\rho\right)$$

$$\tag{1.2}$$

$$MSE(\bar{y}_{R}) \cong \frac{1-f}{n} \bar{Y}^{2} \left(C_{y}^{2} + C_{x}^{2} - 2C_{x}C_{y}\rho \right)$$
(1.3)

Where, C_{x}, C_{y} are the co-efficient of variation and P is the co-efficient of correlation.

2. Yan and Tian (2010) ^[17] Estimator:

Yan and Tian (2010)^[17] defined two modified ratio estimators using the linear combinations of Co-efficient of skewness and Co-efficient of kurtosis as

$$\hat{\vec{Y}}_1 = \vec{y} \begin{bmatrix} \vec{x} \beta_2 + \beta_1 \\ \vec{x} \beta_2 + \beta_1 \end{bmatrix}$$
(2.1)

$$\hat{\vec{Y}}_2 = \vec{y} \left[\frac{\vec{x} \beta_1 + \beta_2}{\vec{x} \beta_1 + \beta_2} \right]$$
(2.2)

Where, $\beta_1 = \frac{N \sum_{i=1}^{N} (X_i - \bar{X})^3}{(N-1)(N-2)S^3}$ is Co-efficient of Skewness of auxiliary variable and $\beta_2 = \frac{N(N+1)\sum_{i=1}^{N} (X_i - \bar{X})^4}{(N-1)(N-2)S^3} - \frac{3(N-1)^2}{(N-2)(N-3)}$ is Co-efficient of kurtosis. To the first degree of approximation the bias and mean square error are given below:

$$Bias\left(\bar{\tilde{Y}}_{1}\right) = \frac{(1-f)}{n}\bar{Y}\left(\theta_{1}^{2}C_{x}^{2} - \theta_{1}C_{x}C_{y}\rho\right)$$

$$\tag{2.3}$$

$$MSE\left(\bar{\tilde{Y}}_{1}\right) = \frac{(1-f)}{n}\bar{Y}^{2}(C_{y}^{2} + \theta_{1}^{2}C_{x}^{2} - 2\theta_{1}C_{x}C_{y}\rho)$$
(2.4)

Where,
$$\theta_1 = \frac{\bar{x}\beta_2}{\bar{x}\beta_2 + \beta_1}$$

and

$$Bias\left(\overline{\hat{Y}}_{2}\right) = \frac{(1-f)}{n}\overline{Y}(\theta_{2}^{2}C_{x}^{2} - \theta_{2}C_{x}C_{y}\rho)$$

$$(2.5)$$

$$MSE\left(\bar{\bar{Y}}_{2}\right) = \frac{(1-f)}{n}\bar{Y}^{2}(C_{y}^{2} + \theta_{2}^{2}C_{x}^{2} - 2\theta_{2}C_{x}C_{y}\rho)$$
(2.6)

Where,
$$\theta_2 = \frac{\bar{x}\beta_1}{\bar{x}\beta_1 + \beta_2}$$

3. The suggested adapted estimator:

We propose the new adopted or modified ratio estimator using the linear combination of Co-efficient of Kurtosis and Tri-Mean (TM) of the auxiliary variable. This measure is the weighted average of the population median and two quartiles. For more detailed properties of tri-mean see Wang *et al* (2007)^[16].

$$\hat{\vec{Y}}_{p1} = \vec{y} \left[\frac{\vec{x} \beta_2 + TM}{\vec{x} \beta_2 + TM} \right], \tag{3.1}$$

Where, $\beta_2 = \frac{N(N+1)\sum_{i=1}^{N}(X_i - \bar{X}_i)^*}{(N-1)(N-2)S^3} - \frac{3(N-1)^2}{(N-2)(N-3)}$, Co-efficient of kurtosis and (TM) which is the weighted average of the population median and two quartiles (Q1 and Q3) and is defined as:

$$TM = (Q_1 + Q_2 + Q_3)/3,$$

The bias of $\overline{\bar{Y}}_{p1}$ upto the first degree of approximation is given by;

$$B\left(\hat{Y}_{p1}\right) = \frac{(1-f)}{n} \bar{Y}(\theta_{p1}^2 \mathcal{L}_x^2 - \theta_{p1} \mathcal{L}_x \mathcal{L}_y \rho), \tag{3.2}$$

MSE of the above estimator can be found using Taylor series method as:

$$MSE\left(\hat{\bar{Y}}_{p1}\right) = \frac{(1-f)}{n}\bar{Y}^{2}\left(C_{y}^{2} + \theta_{p1}^{2}C_{x}^{2} - 2\theta_{p1}C_{x}C_{y}\rho\right);$$
(3.3)

Where, $\theta_{p1} = \frac{\bar{x}\beta_2}{\bar{x}\beta_2 + TM}$

4. Comparison of Efficiency

We compare the efficiencies for which the estimator given in (3.1) \bar{Y}_{p1} is more efficient than the existing estimator;

$$MSE\left(\hat{\bar{Y}}_{p1}\right) < MSE\left(\hat{\bar{Y}}_{i}\right)_{if} \rho < \frac{(\theta_{p1}+\theta_{i})}{2}\frac{c_{x}}{c_{y}}; 1,2 (Yan and Tian (2010)$$

$$(4.1)$$

5. Numerical illustration

Data from reference book Murthy, M N (1967)^[8] page 228 is utilized in which fixed capital is denoted by X (auxiliary variable) and output of 80 factories are denoted by Y (study variable) have been utilized for the purpose. For the assessment of the performance of the proposed estimator with that of the Yan and Tian (2010)^[17] estimators. The population parameters are given in Table 1 and Constants, Bias and MSE's of estimators are given in Table 2 below:

Ν	n	\overline{Y}	X	ρ	S_y	Cy	S_x	Cx	β_2	β_1	т ^M
80	20	51.8264	11.2646	0.9413	18.3569	0.3542	8.4542	0.750	2.866	1.05	9.318

Table 1: Parameters and various constants of population are

Estimator	Constants	Bias	MSE
$ ilde{Y}_1$ Yan and Tian (2010) ^[17]	0.968	0.554	881.97
\tilde{Y}_{2} Yan and Tian (2010) ^[17]	0.804	0.316	539.35
$\overline{\tilde{Y}}_{p1}$ (Proposed Estimator)	0.787	0.305	415.27

Table 2: Constants, Bias and MSE's of estimators

% Relative Efficiency= MSE (Existing)/MSE (Proposed) * 100= 129.88%

It is evident from the Table.2 that the proposed estimator has smallest bias among the three as well as MSE value is less as compared to the Yan and Tian $(2010)^{[17]}$ estimators. We also calculate the value of the condition given in (4.1) $\rho < \frac{(\theta_{p_1} + \theta_l)}{2} \frac{c_x}{c_y}$; 1,2 Therefore the above condition is very much satisfied in case of modified estimator.

5. Conclusions

A new linear combination of Co-efficient of Kurtosis and Quartile deviation of the auxiliary variable has been adopted and a modified estimator has been proposed. Utilizing the numerical illustration it has also been demonstrated that the estimator produces smallest bias as well as MSE value as compared to the Yan and Tian (2010)^[17] estimators.

6. References

- 1. Cingi H, Kadilar C: Advances in Sampling Theory- Ratio Method of Estimation, Bentham Science Publishers, 2009.
- Cochran WG. Sampling Techniques, 3rdEdition, John Wiley & Sons, Inc., New York, 1977. 2.
- Kadilar C, Cingi H. Ratio estimators in simple random sampling, Applied Mathematics and Computation. 2004; 151:893-3. 902.
- Kadilar C, Cingi H. An improvement in estimating the population mean by using the correlation co-efficient, Hacettepe 4. Journal of Mathematics and Statistics. 2006a; 35(1):103-109.
- Kadilar C, Cingi H. Improvement in estimating the population mean in simple random sampling, Applied Mathematics 5. Letters. 2006b; 19:75-79.
- 6. Khare BB, Supriya Khare. On the utilization of known coefficient of variation and preliminary test of significance in the estimation of population mean. Int. J. Agricult. Stat. Sci. 2019; 15(1):35-37.
- Maqbool STA Raja, Shakeel Javid. Generalized Modified Ratio Estimator using non-conventional Location Parameter. Int 7. J. Agricult. Stat. Sci. 2016; 12(1):95-97.
- Murthy MN. Sampling theory and methods. Calcutta Statistical Publishing House, India, 1967. 8.
- Neha Garg, Pachori, Menakshi. Calibration estimation in population mean in stratified sampling Using coefficient of 9. Skewness. Int. J. Agricult. Stat. Sci. 2019; 15(1):211-219.
- 10. Priva Mehta, Rajesh Tailor. Estimation of ratio of two population means in stratified random sampling. Int. J. Agricult. Stat. Sci, 2019, 199-203.
- 11. Privaranjan Dash, Sunani Kalvani. Estimation of population proportion by using Calibration Estimator. Int. J. Agricult. Stat. Sci. 2019: 15(1):249-254.
- 12. Subramani J. Generalized modified ratio estimator for estimation of finite population mean, Journal of Modern Applied Statistical Methods. 2013; 12(2):121-155.
- 13. Subzar M, Maqbool S Raja TA, M Abid. Ratio Estimators for Estimating Population Mean in Simple Random Sampling Using Auxiliary Information. Appl. Math. Inf. Sci. Lett. 2018a; 6(3):123-130.
- 14. Subzar M, Maqbool, Showkat, Raja, Tariq Ahmad, Surya Kant Pal, Prayas Sharma. Efficient Estimators of Population Mean Using Auxiliary Information Under Simple Random Sampling. Statistics in Transition New Series. 2018b; 19(2):219-238.
- 15. Subzar M, Bouza CN, Maqbool S Raja TA, Para BA. Robust Ratio Type Estimators in Simple Random Sampling Using Huber M Estimation. Revista Investigacion Operacional. 2019; 40(2):201-209.
- 16. Wang T, Li Y, Cui H. On weighted randomly trimmed means. Journal of systems science and complexity. 2007; 20(1):47-65.
- 17. Yan Z, Tian B. Ratio Method to the Mean Estimation Using Co-efficient of Skewness of Auxiliary Variable. ICICA2010, PartII, CCIS106, 2010, 103-110.