

ISSN: 2456-799X, Vol.07, No.(2) 2022, Pg. 82-87

Oriental Journal of Physical Sciences

www.orientjphysicalsciences.org

Logarithmic-Product-Cum-Ratio Type Estimator for Estimating Finite Population Coefficient of Variation

MOJEED ABIODUN YUNUSA¹, AHMED AUDU¹, and AWWAL ADEJUMOBI^{2*}

¹Department of Statistics, Usmanu Danfodiyo University, Sokoto, Nigeria. ²Department of Mathematics, Kebbi State University of Science and Technology, Aliero, Nigeria.

Abstract

Estimation of population parameters have been a challenging aspect in sample survey for sometimes and many efforts have been made to enhance the exactness of the parameters of these estimators. We suggested the logarithmic-product-cum-ratio type estimator. Expression of the MSE of the intended estimator originated using the Taylor series technique. A numerical illustration was conducted and the results revealed that the modified work is ameliorated than sample means with other estimators observed.

Introduction

Statistically, the coefficient of variation is used to make a comparison between two or more things with different units or dimensions, for example, the weight and height of individuals, are in kilograms (kg) and centimeters (cm). Auxiliary variables are often used in sample surveys to obtain improved precision in the estimate of the population characteristics of the study variable. This information may be used at both the design and estimation stages. Ratio, regression, and product method of estimation are used in this setting.



Article History Received: 29 July 2022 Accepted: 23 November 2022

Keywords: Auxiliary Information; Coefficient of Variation; Mean Square Error; Study Variable; Simple Random Sampling.

For appraising population parameters of the study variable using the information on auxiliary variables, several researchers^{3,4,9,11,12,13,15,16 and17} have performed a considerable amount of work Those of them ceased to emphasize the difficulty of estimating the coefficient of variation for a long time. But some still do, like¹ first suggested the estimator for the coefficient of variation when samples were selected using SRSWOR while other work includes.^{2,6,7,8,10,18,19} and ²⁰

CONTACT Awwal Adejumobi awwaladejumobi@gmail.com Department of Mathematics, Kebbi State University of Science and Technology, Aliero, Nigeria.



© 2022 The Author(s). Published by Oriental Scientific Publishing Company

This is an **∂** Open Access article licensed under a Creative Commons license: Attribution 4.0 International (CC-BY). Doi: http://dx.doi.org/10.13005/OJPS07.02.05



In this current study, a class of logarithmic-productcum-ratio type estimatorsfor the estimation of the coefficient of variation have been proposed. The proposed class of estimators is expected to produce an estimate nearer to the accurate population coefficient of variation.

Methodology

Let us consider a simple random sample n drawn from the given population of N units. Let the value of the study variable Y and the auxiliary variable X for the ith units (i=1, 2,3, 4, ..., N) of the population be denoted by Y_i and X_i for the ith unit in methods: population means of the auxiliary variables, population variance of the study variable, population variance of the auxiliary variable, population covariance of the auxiliary and study variable, *p* denotes the correlation coefficient between X and Y.

$$\begin{split} \overline{y} &= \frac{1}{n} \sum_{i=1}^{n} y_{i} , \overline{x} = \frac{1}{n} \sum_{i=1}^{n} x_{i} , \ s_{y}^{2} = \frac{1}{n-1} \sum_{i=1}^{n} (y_{i} - \overline{y})^{2} \\ s_{x}^{2} &= \frac{1}{n-1} \sum_{i=1}^{n} (x_{i} - \overline{x})^{2} , \ C_{y} = \frac{S_{y}}{\overline{Y}} \\ s_{xy} &= \frac{1}{n-1} \sum_{i=1}^{n} (x_{i} - \overline{x})(y_{i} - \overline{y}) , \ \overline{Y} = \frac{1}{N} \sum_{i=1}^{N} Y_{i} , \ \overline{X} = \frac{1}{N} \sum_{i=1}^{N} X_{i} , \\ \gamma &= \frac{(1-f)}{n} , \ f = \frac{n}{N} , \\ S_{y}^{2} &= \frac{1}{N-1} \sum_{i=1}^{N} (Y_{i} - \overline{Y})^{2} , \ S_{x}^{2} = \frac{1}{N-1} \sum_{i=1}^{N} (X_{i} - \overline{X})^{2} , \\ S_{xy} &= \frac{1}{N-1} \sum_{i=1}^{N} (X_{i} - \overline{X})(Y_{i} - \overline{Y}) , \\ C_{x} &= \frac{S_{x}}{\overline{X}} , \ \lambda_{y_{x}} = \frac{\mu_{y_{x}}}{\mu_{20}^{r/2} \mu_{02}^{s/2}} \end{split}$$

Existing Estimators

The sample coefficient of variationis given as,

$$t_0 = \hat{C}_y = \frac{s_y}{\overline{y}} \qquad \dots (1)$$

The variance is given by.

$$MSE(t_0) = C_y^2 \gamma \left(C_y^2 + \frac{1}{4} (\lambda_{40} - 1) - C_y \lambda_{30} \right) \qquad \dots (2)$$

²ratio estimator of population coefficient of variation of the study variable utilizing information on the population mean.

$$t_{AR1} = \hat{C}_{y} \left(\frac{\bar{X}}{\bar{x}} \right) \qquad \dots (3)$$

$$MSE(t_{AR1}) = C_y^2 \gamma \left[C_y^2 + \frac{1}{4} (\lambda_{40} - 1) + C_x^2 - C_x \lambda_{21} - C_y \lambda_{30} + 2\rho C_y C_x \right]$$

...(4)

They also proposed a ratio estimator of the population coefficient of variation of the study variable utilizing information on the population variance.

$$t_{AR2} = \hat{C}_{y} \left(\frac{S_x^2}{s_x^2} \right) \qquad \dots (5)$$

$$MSE(t_{AR2}) = C_y^2 \gamma \left[C_y^2 + \frac{1}{4} (\lambda_{40} - 1) + (\lambda_{04} - 1) - (\lambda_{22} - 1) - C_y \lambda_{50} + 2C_y \lambda_{42} \right]$$
...(6)

¹⁸the following estimators for the coefficient of variation based on information on a single auxiliary variable, the population mean, as

$$t_1 = \hat{C}_{y} \left(\frac{\overline{X}}{\overline{X}}\right)^{\alpha} \qquad \dots (7)$$

$$t_{2} = \hat{C}_{y} \exp\left\{\beta\left(\frac{\overline{X} - \overline{x}}{\overline{X} + \overline{x}}\right)\right\} \qquad \dots (8)$$
$$t_{3} = \hat{C}_{y} + d_{1}\left(\overline{X} - \overline{x}\right) \dots (9)$$

The MSE expressions of the estimators are given by, $MSE(t_1) = C_y^2 \gamma \left[C_y^2 + \frac{\lambda_{40} - 1}{4} + \alpha^2 C_x^2 - C_y \lambda_{30} + 2\rho C_y C_x - \alpha C_x \lambda_{21} \right]$

$$MSE(t_2) = C_y^2 \gamma \left[C_y^2 + \frac{\lambda_{40} - 1}{4} + \frac{\beta^2 C_x^2}{4} - C_y \lambda_{30} + \beta \rho C_y C_x - \frac{\beta}{2} C_x \lambda_{21} \right]$$
...(11)

$$MSE(t_{3}) = \gamma \left[C_{y}^{2} \left(C_{y}^{2} - C_{y} \lambda_{30} + \frac{\lambda_{40} - 1}{4} \right) + d_{1}^{2} \overline{X}^{2} C_{x}^{2} + 2 d_{1} \overline{X} \rho C_{y}^{2} C_{x} - d_{1} \overline{X} C_{y} C_{x} \lambda_{21} \right] \dots (12)$$

Where
$$\alpha = \frac{\lambda_{21} - 2\rho C_y}{2C_x}$$
, $\beta = \frac{\lambda_{21} - 2\rho C_y}{C_x}$, $d_1 = \frac{C_y \lambda_{21} - 2\rho C_y^2}{2\bar{X}C_x}$.

¹⁸the following estimators for coefficient of variation based on information on single auxiliary variable, population variance, as

$$t_4 = \hat{C}_y \left(\frac{S_x^2}{s_x^2}\right)^{\alpha} \qquad \dots (13)$$

$$t_{5} = \hat{C}_{y} \exp\left\{\beta\left(\frac{S_{x}^{2} - s_{x}^{2}}{S_{x}^{2} + s_{x}^{2}}\right)\right\} \qquad \dots (14)$$

...(10)

$$t_6 = \hat{C}_y + d_2 \left(S_x^2 - s_x^2 \right) \qquad \dots (15)$$

The mean square error (MSE) expressions of the estimators are given by.

$$MSE(t_4) = C_y^2 \gamma \left[C_y^2 + \frac{\lambda_{40} - 1}{4} + \alpha^2 \left(\lambda_{04} - 1 \right) - C_y \lambda_{50} + 2\alpha C_y \lambda_{12} - \alpha \left(\lambda_{22} - 1 \right) \right]$$
(16)

$$MSE(t_{5}) = C_{y}^{2} \gamma \left[C_{y}^{2} + \frac{\lambda_{40} - 1}{4} + \frac{\beta^{2} (\lambda_{04} - 1)}{4} - C_{y} \lambda_{50} + \beta C_{y} \lambda_{12} - \frac{\beta}{2} (\lambda_{22} - 1) \right] \dots (17)$$

$$MSE(t_{\delta}) = \gamma \begin{bmatrix} C_{y}^{2} \left(C_{y}^{2} - C_{y} \lambda_{30} + \frac{\lambda_{40} - 1}{4} \right) + 2C_{y}^{2} d_{2} S_{x}^{2} \lambda_{12} + d_{2}^{2} S_{x}^{4} \left(\lambda_{04} - 1 \right) \\ -C_{y} d_{2} S_{x}^{2} \left(\lambda_{22} - 1 \right) \end{bmatrix}$$
...(18)

Where
$$\alpha = \frac{\lambda_{22} - 1 - 2C_y \lambda_{12}}{2(\lambda_{04} - 1)}, \beta = \frac{\lambda_{22} - 1 - 2C_y \lambda_{12}}{(\lambda_{04} - 1)}, d_2 = \frac{C_y (\lambda_{22} - 1) - 2C_y^2 \lambda_{12}}{2S_x^2 (\lambda_{04} - 1)}.$$

²⁰logarithmic ratio type estimator for the estimation of population coefficient of variation when the natural logarithm of the auxiliary variable is known as,

$$T_{y} = \hat{C}_{y} \left(\frac{Ln(S_{x}^{2})}{Ln(s_{x}^{2})} \right) \qquad \dots (19)$$

The mean square error (MSE) of (19) is given by,

$$MSE(T_{y}) = C_{y}^{2} \gamma \left(C_{y}^{2} + \frac{(\lambda_{40} - 1)}{4} + \frac{(\lambda_{04} - 1)}{(Ln(S_{x}^{2}))^{2}} - \frac{(\lambda_{22} - 1)}{Ln(S_{x}^{2})} - C_{y} \lambda_{30} + \frac{2C_{y} \lambda_{12}}{Ln(S_{x}^{2})} \right) \dots (20)$$

Proposed Estimator

Motivated by the work of,²⁰ we suggested a logarithmic-product-cum-ratio type estimator for estimating the coefficient of variation as,

$$T_{am} = \hat{C}_{y} \left(\frac{Ln(\bar{x})}{Ln(\bar{X})} \right) \left(\frac{Ln(S_{x}^{2})}{Ln(s_{x}^{2})} \right) \qquad \dots (21)$$

The above estimator is defined under the assumptions that,

$$e_0 = \frac{\overline{y} - \overline{Y}}{\overline{Y}}, e_1 = \frac{\overline{x} - \overline{X}}{\overline{X}}, e_2 = \frac{s_y^2 - S_y^2}{S_y^2}, e_3 = \frac{s_x^2 - S_x^2}{S_x^2}$$

Such that

$$\overline{y} = \overline{Y}(1+e_0), \overline{x} = \overline{X}(1+e_1), s_y = S_y(1+e_2)^{1/2}, s_x = S_x(1+e_3)^{1/2},$$

$$s_y^2 = S_y^2(1+e_2), s_x^2 = S_x^2(1+e_3)$$

$$E(e_0) = E(e_1) = E(e_2) = E(e_3) = 0$$

$$\begin{split} & E(e_0^2) = \gamma C_y^2, E(e_1^2) = \gamma C_x^2, E(e_2^2) = \gamma (\lambda_{40} - 1), E(e_3^2) = \gamma (\lambda_{04} - 1), \\ & E(e_0e_1) = \gamma P C_y C_x, E(e_0e_2) = \gamma C_y \lambda_{30}, E(e_0e_3) = \gamma C_y \lambda_{12}, \\ & E(e_1e_2) = \gamma C_x \lambda_{21}, E(e_1e_3) = \gamma C_x \lambda_{03}, E(e_2e_3) = \gamma (\lambda_{22} - 1) \end{split}$$

Derivation of the Proposed Estimator T_{am}

Expressing (21) in terms of error terms defined in section 1, we have,

$$T_{am} = \frac{S_{y}(1+e_{3})^{1/2}}{\overline{Y}(1+e_{0})} \left(\frac{Ln(\overline{X}(1+e_{1}))}{Ln(\overline{X})}\right) \left(\frac{Ln(S_{x}^{2})}{Ln(S_{x}^{2}(1+e_{3}))}\right) \qquad \dots (22)$$

Expand (22) using law of logarithm, we obtained (23)

$$T_{am} = \frac{S_{y}(1+e_{3})^{1/2}}{\overline{Y}(1+e_{0})} \left(\frac{Ln(\overline{X})+Ln(1+e_{1})}{Ln(\overline{X})}\right) \left(\frac{Ln(S_{x}^{2})}{Ln(S_{x}^{2})+Ln(1+e_{3})}\right) \dots (23)$$

Simplifying (23) by factorizing \overline{X} and $Ln(S_x^2)$ from the numerator and denominator of the expression in the last two brackets of (23), we have obtained (24)

$$T_{am} = C_{y} \left(1 + e_{2}\right)^{1/2} \left(1 + e_{0}\right)^{-1} \left(1 + \theta_{1} Ln(1 + e_{1})\right) \left(\frac{1}{1 + \theta_{2} Ln(1 + e_{3})}\right)$$
...(24)
$$T_{am} = C_{y} \left(1 + e_{2}\right)^{1/2} \left(1 + e_{0}\right)^{-1} \left(1 + \theta_{1} Ln(1 + e_{1})\right) \left(1 + \theta_{2} Ln(1 + e_{3})\right)^{-1}$$
...(25)

where,
$$\theta_1 = 1/Ln(\overline{X}), \theta_2 = 1/Ln(S_x^2)$$

By expanding $Ln(1+e_3)$, $(1+e_2)^{1/2}$, $Ln(1+e_1)$ and $(1+e_3)^{-1}$ up to first order of approximation we have,

$$T_{am} = C_{y} \left(1 + \frac{e_{2}}{2} - \frac{e_{2}^{2}}{8} \right) \left(1 - e_{0} + e_{0}^{2} \right) \left(1 + \theta_{1} \left(e_{1} - \frac{e_{1}^{2}}{2} \right) \right) \left(1 - \theta_{2} \left(e_{3} - \frac{e_{3}^{2}}{2} \right) + \theta_{2}^{2} e_{3}^{2} \right) \dots (26)$$

$$T_{am} = C_{y} \left(1 - e_{0} + e_{0}^{2} + \frac{e_{2}}{2} - \frac{e_{0}e_{2}}{2} - \frac{e_{2}^{2}}{8} \right) \left(1 + \theta_{1}e_{1} + \frac{\theta_{1}e_{1}^{2}}{2} \right) \left(1 - \theta_{2}e_{3} + \left(\theta_{2}^{2} + \frac{\theta_{2}}{2} \right)e_{3}^{2} \right)$$
...(27)

Simplify, Subtract from both sides and consider terms of degree one, we have

$$T_{am} - C_y = C_y \left(\frac{e_2}{2} - e_0 + \theta_1 e_1 - \theta_2 e_2 \right) \qquad \dots (28)$$

Square both sides of equation (28) and expand right hand side relation to first order of approximation we have,

$$\left(T_{am} - C_{y}\right)^{2} = C_{y}^{2} \begin{pmatrix} \frac{e_{2}^{2}}{4} + e_{0}^{2} + \theta_{1}^{2}e_{1}^{2} + \theta_{2}^{2}e_{3}^{2} - e_{0}e_{2} + \theta_{1}e_{1}e_{2} - \theta_{2}e_{2}e_{3} - 2\theta_{1}e_{0}e_{1} \\ + 2\theta_{2}e_{0}e_{3} - 2\theta_{1}\theta_{2}e_{1}e_{3} & \dots (29) \end{pmatrix}$$

Take expectation on both sides of equation (29) to obtain the MSE of the proposed estimator as:

$$MSE(T_{ann}) = C_y^2 \gamma \left(\frac{(\lambda_{40} - 1)}{4} + C_y^2 + \theta_1^2 C_x^2 + \theta_2^2 (\lambda_{04} - 1) - C_y \lambda_{30} + \theta_1 C_x \lambda_{21} - \theta_2 (\lambda_{22} - 1)) - 2\theta_1 \rho C_y C_x + 2\theta_2 C_y \lambda_{12} - 2\theta_1 \theta_2 C_x \lambda_{03} \dots (30) \right)$$

Numerical Analysis

A numerical analysis was carried out to clarify the accomplishment of proposed estimator. Data used are from the book⁵ and.¹⁴

Dataset 1

X: Area under wheat in 1963, Y: Area under wheat in 1964

N=34, n=15, \overline{X} =208.88, \overline{Y} =199.44, C_x=0.72, C_y=0.75,

 $p=0.98, \lambda_{21}=1.0045, \lambda_{12}=0.9406, \lambda_{40}=3.6161, \lambda_{04}=2.8266, \lambda_{30}=1.1128, \lambda_{03}=0.9206, \lambda_{22}=3.0133$

Dataset 2

X: Number of fish caught in year 1993, Y: Number of fish caught in year 1995

N=69, n=40, \overline{X} =4591.07, \overline{Y} =4514.89, C_x=1.38, C_y=1.35, *p*=0.96, λ_{21} =2.19, λ_{12} =2.30, λ_{40} =7.66, λ_{04} =9.84, λ_{30} = 1.11, λ_{03} =2.52, λ_{22} =8.19

Table1 above revealed that mean square error of the suggested estimator is minimal compared to those estimators considered in this study. This implies that the suggested estimator is ameliorated, more so, this is evidence, as the proposed estimator has a higher percentage of relative efficiency.

Table 1	: MSE an	nd Percentage	Relative	Efficiencies	of proposed	estimator
		and the	existing	estimators.		

ESTIMATORS	DATASET 1		DATASET 2	DATASET 2		
	MSE	PRE	MSE	PRE		
t _o	0.0080036	100.00	0.038088	100.00		
t _{AR1}	0.0258907	30.91	0.0851798	44.71		
t_{AB2}	0.0336578	23.78	0.18860299	20.19		
t_1	0.006868341	116.53	0.03731461	102.07		
t_2	0.006868341	116.53	0.03731461	102.07		
t_3	0.006868341	116.53	0.03731461	102.07		
$\check{t_{A}}$	0.006962763	114.95	0.037568156	101.38		
t_5	0.006962763	114.95	0.037568156	101.38		
$\check{t_6}$	0.006962763	114.95	0.037568156	101.38		
\check{T}_{v}	0.0071255	112.32	0.0375686	101.38		
, T _{am}	0.005672477	141.10	0.03591946	106.04		

Results and Discussion

An efficient product-cum-ratio of finite population coefficient of variation is proposed. The attribute of the proposed estimator was secured. The results hows the mean square errors and percentage relative efficiencies of the suggested estimator and some existing work using the two data sets. The results revealed that the proposed estimator is more efficient than other existing estimators considered in the study.

Conclusion

Currently, we suggested logarithmic-product-cumratio type estimator for estimating the coefficient of variation of the study variable, and this estimator employed information on the natural logarithm of the sample and population mean as well as the sample and population variance of the auxiliary variable. From numerical analysis, the results show that the proposed estimator is more efficient than other existing estimators considered in the study. Hence, the ameliorated proposed estimatoris recommended for practical usage.

Acknowledgement

The author would like to thank, Department of Mathematics, Kebbi State University of Science and Technology, Aliero, Nigeria. for their guidance and support to complete this article.

Funding

The authors received no financial support for the research, authorship and publication of this article.

Conflict of Interest

The authors declare no conflict of interest.

References

- Das A.K. and Tripathi T.P., A class of Estimators for co-efficient of Variation using knowledge on co-efficient of variation of an auxiliary character. *In annual conference of Ind. Soc. Agricultural Statistics*, Held at New Delhi, India. (1981).
- Archana V. and Rao A., Some improved Estimators of co-efficient of variation from bivariate normal distribution. A Monte Carlo comparison. *Pakistan Journal of Statistics* and Operation Research, 10(1), (2014).
- Adichwal N.K., Sharma P. and Singh R., Generalized class of estimators for population variance using information on two auxiliary variables. *International journal* of applied computational mathematics, 3(2), 651-66, (2017).
- Kumar N. andAdichwal R., Estimation of finite population mean using Auxiliary Attribute in sample surveys. (2016).
- 5. Murthy M.N., Sampling theory and methods. *Sampling theory and methods.* (1967).
- 6. Patel P.A. and Rina S., A Monte Carlo comparison of some suggested estimators of coefficient of variation in finite population. *Journal of statistics science*, 1(2), 137-147, (2009).
- Rajyaguru A. and Gupta P., On the estimation of the coefficient of variation from finite population-I, *Model Assisted Statistics and Application*, 36(2), 145-156, (2002).
- Rajyaguru A. and Gupta P., On the estimation of the coefficient of variation from finite population –II, *Model Assisted Statistics and Application*, 1(1), 57-66, (2006).
- Sahai A. and Ray S.K., An efficient estimator using auxiliary information, *Metrika*, 27(4), 271-275, (1980).

- Shabbir J. and Gupta S., Estimation of Population coefficient of variation in sample and stratified Random Sampling under Two-phase sampling scheme when using Two Auxiliary variables. *Communications in Statistics Theory and Methods,* (just accept), 00-00, (2016).
- 11. Singh H.P. and Singh R., A class of chain ratio type estimators for the coefficient of variation of finite population in two-phase sampling, Algarh. *Journal of statistics,* vol. 22, 1-9, (2002).
- 12. Singh H.P. and Solanki R.S., An efficient class of estimators for the population mean using auxiliary information in systematic sampling. *Journal of Statistical Theory and Practice*, 6(2), 274-285, (2012).
- Singh H.P. and Tailor R., Estimation of finite population mean with known coefficient of variation of an auxiliary character. *Statistical*, 65(3), 301-313, (2005).
- 14. Singh S., *Advanced Sampling Theory with Applications. How Michael "Selected" Amy* (vol. 2). Springer science and media, (2003).
- Sisodia B.V.S. andDwivedi V.K., Modified ratio estimator using coefficient of variation of auxiliary variable. *Journal-Indian society* of Agricultural statistics, (1981).
- Solanki R.S. and Singh H.P., And in proved class of estimators for the general population parameters using Auxiliary information. *Communications in Statistics-Theory and Methods*, 44(20), 4241-4262, (2015).
- Srivastava S.K. andJhajj H. A., A class of estimators of the population mean in survey sampling using auxiliary information. *Biometrika*, 68(1), 341-343, (1981).
- 18. Singh R., Mishra M., Singh B., Singh P.

and Adichwal N. K., Improved estimators for population coefficient of variation using auxiliary variables. *Journal of Statistics and Management.* 21(7), 1335-1355, (2018).

 Audu A., Yunusa M. A., Ishaq O. O., Lawal M. K., Rashida A., Muhammed A. H., Bello A. B., Hairullahi M. U.and Muili J. O., Difference-Cum-Ratio type estimators for estimating finite population coefficient of variation in Simple random sampling. *Asian Journal of Probability and Statistics.* 13(3): 13-29, (2021).

 Yunusa M. A., Audu A., Musa N., Beki D. O., Rashida A., Bello A. B.and Hairullahi M. U., Logarithmic ratio type estimator of population of variation. *Asian Journal of Probability and Statistics.* 14(2): 13-22, (2021).