

Journal of Statistics Applications & Probability Letters An International Journal

http://dx.doi.org/10.12785/jsapl/020108

# Dual to Separate Product Type Exponential Estimator in Sample Surveys

Hilal A. Lone \* and Rajesh Tailor

School of Studies in Statistics, Vikram University, Ujjain-456010, M.P. India

Received: 18 Sep. 2014, Revised: 9 Oct. 2014 ., Accepted: 11 Oct. 2014 Published online: 1 Jan. 2015

**Abstract:** This paper addresses the problem of estimation of finite population mean in case of post-stratification. In this paper dual to separate product type exponential estimator is proposed using the same approach adopted by Srivenkataramana [15] and Bandhyopadhyaya [2]. Conditions under which the proposed estimator is more efficient than usual unbiased estimator, usual separate product type estimator, dual to separate product type estimator and separate product type exponential estimator are obtained. The bias and mean squared error expressions are obtained upto the first degree of approximation. An empirical study has been carried out to demonstrate the performance of the proposed estimator.

Keywords: Separate estimator; Post-stratification; Bias, Mean squared error

### **1** Introduction

Cochran [4] and Robson [11] envisaged classical ratio and product estimators which were studied in case of post stratification by Ige and Tripathi [6]. Tailor et al. [23] proposed dual to Ige and Trapthi [6] ratio and product estimators. Chouhan [3] studied the Bhal and Tuteja [1] product type exponential estimator in case of post-stratification. Srivenkataramana [15] and Bandhyopadhyaya [2] envisaged dual to classical ratio and product estimators using transformation on auxiliary variate. Tailor and Tailor [22] proposed dual to Bhal and Tuteja [1] product type exponential estimator. The problem of estimating the population parameters has been discussed by various researchers including Tailor and Lone [17,20,21], Lone and Tailor [9],Lone et al.[7], Lone et al.[10],Singh et al. [12] and Singh et al.[13].At the estimation stage separate type estimators have been discussed by few researchers including Vishwakarma and Singh [24], Yadav et al. [25],Tailor and Lone [19], Chouhan [3],Tailor and Lone [18], Lone and Tailor [8]. Consider a finite population  $U = (U_1, U_2, ..., U_N)$  of size N. A sample of size n is drawn from population U using simple random sampling without replacement. After selecting the sample, it is observed that which units belong to  $h^{th}$  stratum. Let  $n_h$  be the size of the sample falling in  $h^{th}$  stratum such

that  $\sum_{h=1}^{L} n_h = n$ . Here it is assumed that *n* is so large that possibility of  $n_h$  being zero is very small.

Let  $x_{hi}$  be the observation on  $i^{th}$  unit that fall in  $h^{th}$  stratum for auxiliary variate x and  $y_{hi}$  be the observation on  $i^{th}$  unit that fall in  $h^{th}$  stratum for study variate y, then

$$\overline{X}_{h} = \frac{1}{N_{h}} \sum_{i=1}^{N_{h}} x_{hi} : h^{th} \text{ stratum mean for auxiliary variate } x,$$

\* Corresponding author e-mail: Hilalstat@gmail.com

$$\overline{Y}_{h} = \frac{1}{N_{h}} \sum_{i=1}^{N_{h}} y_{hi} : h^{th} \text{ stratum mean for study variate } y,$$

$$\overline{X} = \frac{1}{N} \sum_{h=1}^{L} \sum_{i=1}^{N_h} x_{hi} = \sum_{h=1}^{L} W_h \overline{X}_h$$
: Population mean of auxiliary variate x,

$$\overline{Y} = \frac{1}{N} \sum_{h=1}^{L} \sum_{i=1}^{N_h} y_{hi} = \frac{1}{N} \sum_{h=1}^{L} N_h \overline{Y}_h = \sum_{h=1}^{L} W_h \overline{Y}_h$$
: Population mean of the study variate y

In case of post-stratification, usual unbiased estimator of population mean  $\overline{Y}$  is defined as

$$\overline{y}_{PS} = \sum_{h=1}^{L} W_h \overline{y}_h , \qquad (1.1)$$

where

 $W_h = \frac{N_h}{N}$  is the weight of the  $h^{th}$  stratum and  $\overline{y}_h = \frac{1}{n_h} \sum_{i=1}^{n_h} y_{hi}$  is sample mean of  $n_h$  sample units that fall in the  $h^{th}$ 

stratum.

Using the results from Stephen [16] the variance of  $\overline{y}_{PS}$  to the first degree of approximation is obtained as

$$Var(\overline{y}_{PS}) = \left(\frac{1}{n} - \frac{1}{N}\right) \sum_{h=1}^{L} W_h S_{yh}^2 + \frac{1}{n^2} \sum_{h=1}^{L} (1 - W_h) S_{yh}^2, \qquad (1.2)$$
  
where  $S_{yh}^2 = \frac{1}{N_h - 1} \sum_{i=1}^{N_h} (y_{hi} - \overline{Y}_h)^2.$ 

When the correlation between the study variate y and the auxiliary variate x is negative, Ige and Tripathi [6] proposed a product type estimator in case of post-stratification as

$$\hat{\overline{Y}}_{PPS} = \overline{y}_{PS} \left( \frac{\overline{x}_{PS}}{\overline{X}} \right).$$
(1.3)

The separate version of Ige and Tripathi [6] product type estimator can be written as

$$\hat{\overline{Y}}_{PPS}^{S} = \sum_{h=1}^{L} W_h \ \overline{y}_h \left( \frac{\overline{x}_h}{\overline{X}_h} \right).$$
(1.4)

Upto the first degree of approximation, mean squared error of  $\hat{T}^{S}_{PPS}$  is obtained as

$$MSE\left(\hat{\overline{Y}}_{PPS}^{S}\right) = \left(\frac{1}{n} - \frac{1}{N}\right) \left[\sum_{h=1}^{L} W_h S_{yh}^2 + \sum_{h=1}^{L} W_h R_h^2 S_{xh}^2 + 2\sum_{h=1}^{L} W_h R_h S_{yxh}\right]$$
(1.5)

Following Srivenkataramana [15] and Bandhyopadhyayh [2] transformation, we define dual to separate product type estimator in case of post-stratification as

$$\hat{\overline{Y}}_{PPS}^* = \sum_{h=1}^{L} W_h \overline{y}_h \left( \frac{\overline{X}_h}{\overline{x}_h^*} \right), \tag{1.6}$$

90

where 
$$\overline{x}_h^* = \frac{N_h \overline{X}_h - n_h \overline{x}_h}{N_h - n_h}$$
.

To the first degree of approximation, bias and mean squared error of dual to separate product type estimator are obtained as

$$B(\hat{\bar{Y}}_{PPS}^{*}) = \left(\frac{1}{n} - \frac{1}{N}\right) \sum_{h=1}^{L} \bar{Y}_{h} \left(C_{xh}^{2} a_{h}^{2} + a_{h} \rho_{xyh} C_{xh} C_{yh}\right),$$
(1.7)

and

$$MSE\left(\hat{\bar{Y}}_{PPS}^{*}\right) = \left(\frac{1}{n} - \frac{1}{N}\right) \left[\sum_{h=1}^{L} W_h S_{yh}^2 + \sum_{h=1}^{L} W_h R_h^2 a_h^2 S_{xh}^2 + 2\sum_{h=1}^{L} W_h a_h R_h S_{yxh}\right],$$
(1.8)

where 
$$R_h = \frac{Y_h}{\overline{X}_h}$$
 and  $a_h = \frac{n_h}{N_h - n_h}$ .

In case of negative correlation coefficient between the study variate y and the auxiliary variate x, Bahl and Tuteja [1] proposed product type exponential estimator for population mean as

$$\hat{\overline{Y}}_{1P} = \overline{y} \exp\left(\frac{\overline{x} - \overline{X}}{\overline{x} + \overline{X}}\right).$$
(1.9)

Motivated by Bahl and Tuteja [1], Singh et al. [14] proposed product type exponential estimator in stratified random sampling as

$$\hat{\overline{Y}}_{St}^{Pe} = \overline{y}_{st} \exp\left(\frac{\overline{x}_{st} - \overline{X}}{\overline{x}_{st} + \overline{X}}\right).$$
(1.10)

Chouhan [3] proposed product type exponential estimator in case of post-stratification as

$$\hat{\overline{Y}}_{PS}^{Pe} = \overline{y}_{PS} \exp\left(\frac{\overline{x}_{PS} - \overline{X}}{\overline{x}_{PS} + \overline{X}}\right).$$
(1.11)

Up to the first degree of approximation, the bias and mean squared error of estimator  $\hat{\overline{Y}}_{PS}^{Pe}$  are obtained as

$$B\left(\hat{\bar{Y}}_{PS}^{Pe}\right) = \left(\frac{1}{n} - \frac{1}{N}\right) \sum_{h=1}^{L} \frac{1}{\bar{X}} \left(\frac{3}{8}RS_{xh}^{2} - \frac{1}{2}S_{yxh}\right) , \qquad (1.12)$$

and

$$MSE\left(\widehat{\overline{Y}}_{PS}^{Pe}\right) = \left(\frac{1}{n} - \frac{1}{N}\right) \sum_{h=1}^{L} W_h\left(S_{yh}^2 + \frac{1}{4}R^2S_{xh}^2 + RS_{yxh}\right) .$$

$$(1.13)$$
where  $R = \frac{\overline{Y}}{\overline{X}}.$ 

Here we define the separate version of Chouhan [3] product type exponential estimator as

$$\hat{\overline{Y}}_{PS}^{SPe} = \sum_{h=1}^{L} W_h \, \overline{y}_h \left( \frac{\overline{x}_h - \overline{X}_h}{\overline{x}_h + \overline{X}_h} \right). \tag{1.14}$$

Upto the first degree of approximation mean squared error of  $\hat{\overline{Y}}_{PS}^{SPe}$  is obtained as

$$MSE\left(\hat{Y}_{PS}^{SPe}\right) = \left(\frac{1}{n} - \frac{1}{N}\right) \sum_{h=1}^{L} W_h\left(S_{yh}^2 + \frac{1}{4}R_h^2 S_{xh}^2 + R_h S_{yxh}\right) .$$
(1.15)

where 
$$S_{xh}^2 = \frac{1}{N_h - 1} \sum_{i=1}^{N_h} (x_{hi} - \overline{X}_h)^2$$
 and  $S_{yxh} = \frac{1}{N_h - 1} \sum_{i=1}^{N_h} (y_{hi} - \overline{Y}_h) (x_{hi} - \overline{X}_h).$ 

### 2. Proposed Estimator

Using transformation  $x_i^* = \frac{N\overline{X} - nx_i}{N - n}$  on auxiliary variate x, Srivenkataramana [15] and Bandhyopadhyayh [2] defined dual to classical product estimator as

$$\hat{\overline{Y}}_{2P}^* = \overline{y} \left( \frac{\overline{X}}{\overline{x}^*} \right) .$$
(2.1)
$$= \frac{\overline{X}}{\overline{x}^*} - \frac{N\overline{X} - n\overline{x}}{\overline{x}} \quad \text{is unbiased estimator of population maps} \quad \overline{Y}$$

where  $\overline{x}^* = \frac{NX - n\overline{x}}{N - n}$  is unbiased estimator of population mean  $\overline{X}$ .

Motivated by Srivenkataramana [15] and Bondyopadhyayh [2], Tailor and Tailor [22] proposed dual to Bahl and Tuteja [1] product type exponential estimator as

$$\hat{\overline{Y}}_{3P} = \overline{y} \left( \frac{\overline{X} - \overline{x}^*}{\overline{X} + \overline{x}^*} \right).$$
(2.2)

By using the same transformation adopted by Srivenkataramana [15] and Bondyopadhyayh [2], we propose dual to separate product type exponential estimator for population mean  $\overline{Y}$  in case of post-stratification as

$$\hat{\overline{Y}}_{PS}^{*Pe} = \sum_{h=1}^{L} W_h \overline{y}_h \exp\left(\frac{\overline{X}_h - \overline{x}_h^*}{\overline{X}_h + \overline{x}_h^*}\right).$$
(2.3)

To obtain the bias and mean squared error of suggested estimator  $\hat{Y}_{PS}^{*Pe}$  we write

c 2015 NSP Natural Sciences Publishing Cor.

1 NS

$$\begin{split} \overline{y}_{h} &= \overline{Y}_{h} \left( 1 + e_{0h} \right), \ \overline{x}_{h} = \overline{X}_{h} \left( 1 + e_{1h} \right) \text{ such that} \\ E(e_{0h}) &= E(e_{1h}) = 0, \\ E(e_{0h}^{2}) &= \left( \frac{1}{nW_{h}} - \frac{1}{N_{h}} \right) C_{yh}^{2}, \\ E(e_{1h}^{2}) &= \left( \frac{1}{nW_{h}} - \frac{1}{N_{h}} \right) C_{xh}^{2}, \\ E(e_{0h}e_{1h}) &= \left( \frac{1}{nW_{h}} - \frac{1}{N_{h}} \right) \rho_{yxh} C_{yh} C_{xh} , \end{split}$$

Expressing (2.3) in terms of  $e_{ih}$  's, we have

$$\begin{split} \hat{\bar{Y}}_{PS}^{*Pe} &= \sum_{h=1}^{L} W_{h} \overline{\bar{Y}}_{h} \left(1+e_{0h}\right) \exp\left(\frac{a_{h} \overline{\bar{X}}_{h} e_{1h}}{2 \overline{\bar{X}}_{h} - a_{h} \overline{\bar{X}}_{h} e_{1h}}\right) \\ \hat{\bar{Y}}_{PS}^{*Pe} &= \sum_{h=1}^{L} W_{h} \overline{\bar{Y}}_{h} \left(1+e_{0h}\right) \exp\left\{\frac{a_{h} e_{1h}}{2} \left(1-\frac{a_{h} e_{1h}}{2}\right)^{-1}\right\} \\ \hat{\bar{Y}}_{PS}^{*Pe} &= \sum_{h=1}^{L} W_{h} \overline{\bar{Y}}_{h} \left(1+e_{0h}\right) \left[1+\frac{1}{2} a_{h} e_{1h} \left(1-\frac{a_{h} e_{1h}}{2}\right)^{-1}+\frac{1}{2} \left\{\frac{a_{h}^{2} e_{1h}^{2}}{4} \left(1-\frac{a_{h} e_{1h}}{2}\right)^{-2}\right\} + \dots \right] \\ \hat{\bar{Y}}_{PS}^{*Pe} &= \sum_{h=1}^{L} W_{h} \overline{\bar{Y}}_{h} \left(1+e_{0h}\right) \left[1+\frac{a_{h} e_{1h}}{2}+\frac{a_{h}^{2} e_{1h}^{2}}{8}+\dots\right] \\ \left(\hat{\bar{Y}}_{PS}^{*Pe} - \overline{Y}\right) &= \sum_{h=1}^{L} W_{h} \overline{\bar{Y}}_{h} \left[e_{0h} + \frac{a_{h} e_{1h}}{2}+\frac{3a_{h}^{2} e_{1h}^{2}}{8}+\frac{a_{h} e_{0h} e_{1h}}{2}\right] \end{split}$$

$$(2.4)$$

Taking expectation both sides to (2.4), we get the bias of the proposed estimator  $\hat{\vec{Y}}_{PS}^{*Pe}$  upto the first degree of approximation is obtained as

$$B(\hat{\overline{Y}}_{PS}^{*Pe}) = \left(\frac{1}{n} - \frac{1}{N}\right) \sum_{h=1}^{L} \frac{a_h}{4\,\overline{X}_h} \left[\frac{3}{2} R_h \, a_h S_{xh}^2 + 2S_{yxh}\right].$$
(2.5)

Squaring both sides of (2.4) and then taking expectation, we get the mean squared error of the proposed estimator  $\hat{\overline{Y}}_{PS}^{*Pe}$  upto the first degree of approximation as

$$MSE\left(\hat{\bar{Y}}_{PS}^{*Pe}\right) = \left(\frac{1}{n} - \frac{1}{N}\right) \left[\sum_{h=1}^{L} W_h S_{yh}^2 + \frac{1}{4} \sum_{h=1}^{L} W_h a_h^2 R_h^2 S_{xh}^2 + \sum_{h=1}^{L} W_h a_h R_h S_{yxh}\right].$$
(2.6)



## **3.** Efficiency Comparisons of the Proposed Estimator $\hat{\overline{Y}}_{PS}^{*Pe}$ with $\overline{y}_{PS}$ , $\hat{\overline{Y}}_{PPS}^{s}$ , $\hat{\overline{Y}}_{PPS}^{*}$ and $\hat{\overline{Y}}_{PS}^{SPe}$

Comparing (1.2) and (2.6), it is observed that the proposed estimator  $\hat{\overline{Y}}_{PS}^{*Pe}$  would be more efficient than the usual unbiased estimator  $\overline{y}_{PS}$  i.e.

$$MSE\left(\hat{\bar{Y}}_{PS}^{*p_{e}}\right) < V(\bar{y}_{PS}) \text{ if}$$

$$\sum_{h=1}^{L} W_{h} a_{h}^{2} R_{h}^{2} S_{xh}^{2} + 4 \sum_{h=1}^{L} W_{h} a_{h} R_{h} S_{yxh} < 0$$
(3.1)

From (1.5) and (2.6), it is concluded that the proposed estimator  $\hat{\vec{Y}}_{PS}^{*Pe}$  would be more efficient than  $\hat{\vec{Y}}_{PPS}^{S}$  i.e.

$$MSE\left(\hat{\bar{Y}}_{PS}^{*Pe}\right) < MSE\left(\hat{\bar{Y}}_{PPS}^{S}\right) \text{ if }$$

$$\sum_{h=1}^{L} W_{h}R_{h}^{2}S_{xh}^{2}\left(a_{h}^{2}-4\right) + 4\sum_{h=1}^{L} W_{h}R_{h}S_{yxh}\left(a_{h}-2\right) < 0$$
(3.2)

From (1.8) and (2.6), it is observed that the proposed estimator  $\hat{\vec{Y}}_{PS}^{*Pe}$  would be more efficient than  $\hat{\vec{Y}}_{PPS}^{*}$  i.e.

$$MSE\left(\hat{\bar{Y}}_{PS}^{*Pe}\right) < MSE\left(\hat{\bar{Y}}_{PPS}^{*}\right) \text{ if }$$
  
-3\sum\_{h=1}^{L} W\_{h}a\_{h}^{2}R\_{h}^{2}S\_{xh}^{2} - 4\sum\_{h=1}^{L} W\_{h}a\_{h}R\_{h}S\_{yxh} < 0. (3.3)

From expression (1.15) and (2.6), it is concluded that the proposed estimator  $\hat{\overline{Y}}_{PS}^{*Pe}$  would be more efficient than  $\hat{\overline{Y}}_{PS}^{SPe}$  i.e.

$$MSE\left(\hat{Y}_{PS}^{*Pe}\right) < MSE\left(\hat{Y}_{PS}^{SPe}\right) \text{ if }$$

$$\sum_{h=1}^{L} W_{h}R_{h}^{2}S_{xh}^{2}\left(a_{h}^{2}-1\right) + 4\sum_{h=1}^{L} W_{h}R_{h}S_{yxh}\left(a_{h}-1\right) < 0$$
(3.4)

### 4. Empirical Study

To exhibit the performance of the proposed estimator in comparison to other considered estimators, a population data set is being considered. In this data set x:Total annual sunshine hours and y:Snowy days. The description of population is given below:

Constants	$n_h$	$N_h$	$\overline{X}_h$	$\overline{Y}_h$	$S_{x_1}^{2}$	$S_{y_1}^{2}$	$ ho_{yxh}$	S <sub>yxh</sub>
Stratum I	4	10	1629.99	142.8	10438.71	37.31	-0.38	-239.25
Stratum II	4	10	2035.96	91.0	10662.63	43.16	-0.35	-240.45

Table 4.1 -	Population-	I [Source:	[5]]
-------------	-------------	------------	------

c 2015 NSP Natural Sciences Publishing Cor.

J. Stat. Appl. Pro. Lett. 2, No. 1, 89-96 (2015) / www.naturalspublishing.com/Journals.asp



PRE	100.00	67.89	95.17	108.58	113.99
Estimators	$\overline{y}_{PS}$	$\hat{\overline{Y}}^{S}_{PPS}$	$\hat{\overline{Y}}_{PPS}^{*}$	$\hat{\overline{Y}}_{PS}^{SPe}$	$\hat{Y}_{PS}^{*Pe}$
able 4.2 Percent r	elative Efficie	ncy of $y_{PS}$ , $Y$	$Y_{PPS}^{S}, Y_{PPS}^{PPS}, 1$	$Y_{PS}^{SIT}$ and $Y_{PS}^{T}$	with respect to $y_{PS}$

- CPe <u>^</u>\* <u>^</u>\_\_\_ <u>^</u>\* n Т

### 5. Conclusion

Table 4.2 exhibits that there is a significant gain in efficiency by using the proposed estimator  $\hat{Y}_{PS}^{*Pe}$  over usual unbiased estimator  $\overline{y}_{PS}$ , usual separate product type estimator  $\hat{\overline{Y}}_{PPS}^{S}$ , dual to separate product type estimator  $\hat{\overline{Y}}_{PPS}^{*}$  and separate product type exponential estimator  $\hat{\bar{Y}}_{PS}^{SPe}$ . Section 3 provides the conditions under which the proposed estimator has less mean squared error in comparison to other considered estimators. Thus the proposed estimator  $\hat{Y}_{PS}^{*Pe}$  is recommended for use in practice if the conditions obtained in section 3 are satisfied.

#### References

[1] Bahl, S. and Tuteja, R. K. (1991). Ratio and product type exponential estimators. Infor. & Optimiz. Science, 12, 159-163.

[2]Bandyopadhyay (1980). Improved ratio and product estimators. Sankhya, 42, C, 45-49.

[3] Chouhan, S. (2012). Improved estimation of parameters using auxiliary information in sample surveys. Ph.D.Thesis, Vikram University, Ujjain, M.P., India.

[4] Cochran, W. G. (1940). The estimation of the yields of the cereal experiments by sampling for the ratio of grain to total produce. J. Agri.Sci., 30, 262-275.

[5] http://www.data.jma.gr.jb/obd/stats/data/en/index.html (Official website of Japan Meteorological society).

[6] Ige, A. F. and Tripathi. T. P. (1989). Estimation of population mean using post-stratification and auxiliary information. Abacus, 18, 2,265-276.

[7] Lone, H. A., Tailor, R. and Singh, H. P. (2014). Generalized Ratio-Cum-Product Type Exponential Estimator in Stratified Random Sampling.Comm. statist. Theo. & Meth., 10.1080/03610926.2014.901379.

[8] Lone, H. A. and Tailor, R. (2014). Dual to Separate Ratio Type Exponential Estimator in Post-Stratification. J. Stat. Appl. Pro. 3, 3, 1-6..doi.org/10.12785/jsap/paper

[9] Lone, H. A. and Tailor, R. (2014). On The Family of Estimators of Population Variance in Simple Random Sampling. International J. of Advanced Scientific & Technical Research. 4, 5,103-127.

[10] Lone, H. A., Tailor, R., Singh, H. P. and Verma, M. R. (2014). New alternatives to ratio estimators of population variance in sample surveys. Appl. Math. & Comput. 247, 255-265.

[13].Singh, H. P., Sharma, N. and Tarry, T.A. (2014). An efficient class of two-phase exponential ratio and product type estimators for estimating the finite population mean. J. Statist. Appl. Prob. (in Press)

[14] Singh, R., Kumar, M., Singh, R. D., and Chaudhary, M.K. (2008). Exponential ratio type estimators in stratified random sampling. Presented in International Symposium on Optimisation and Statistics (I.S.O.S) at A.M.U., Aligarh, India, during 29-31 Dec 2008.

[15] Srivenkataramana, T. (1980). A dual of ratio estimator in sample surveys. Biom. J., 67, 1, 199-204.

[16] Stephan, F. (1945). The expected value and variance of the reciprocal and other negative powers of a positive Bernoulli an variate. Ann. Math. Statist. 16, 50-61.



[17] Tailor, R. and Lone, H. A. (2013) Ratio-cum-product type estimator for population variance in sample surveys, Octagon Math. J. 21 (2), 649–654.

[18] Tailor R. and Lone, H. A. (2014a). Ratio –cum- product estimator of finite population mean in double sampling for stratification. J. Rel. & Statist. Studies 7, 1, 93-101.

[19] Tailor, R. and Lone, H.A. (2012). Separate ratio-cum- product estimators of finite population mean using auxiliary information. J. of Rajasthan Statist. Assoc., 1, 2, 94-102.

[20] Tailor, R. and Lone, H. A. (2014b). Separate ratio type estimators of population mean in stratified random sampling. J. Mod. Appl. Statist. Meth. 13, 1, 223-233

[21] Tailor, R. and Lone, H. A. (2014c).Improved ratio-cum-product type exponential estimators for ratio of two population means in sample surveys. Modl. Assist. Statist. Appl.,10.3233/MAS-140300

[22] Tailor, R. and Tailor, R. (2012). Dual to ratio and product type exponential estimators of finite population mean. Presented in National conference on Statistical Inference, at A.M.U., Aligarh, India, Aligarh-202002, Feb 11-12,2012.

[23] Tailor, R. Jatwa, N. K. and Lone, H. A. (2014). Dual to ratio and product type estimators in case of post –stratification. J. Mod. Appl. Statist. Meth. (in press).

[24] Vishwakarma, G. K. and Singh, H.P. (2011). Separate ratio-product estimators for estimating population mean using auxiliary information. J. of statist. Theo. and Appli.10 (4),653-664

[25] Yadev, R., Upadhyaya, L. N., Singh, H.P. and Chatterjee, S. (2011).Improved separate ratio exponential estimator for population mean using auxiliary information. Statist. Trans. - new series, 12, 2, 401-412.