

Median Information Using Auxiliary Information

by

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In sample surveys, supplementary information is often used for increasing the precision of the estimators of population parameters. The resulting methods of estimation, which make use of the ancillary information, under certain conditions give more efficient estimates of the population parameters than those based on the information of the predicted variable alone (Hussein, 1989). The author had earlier suggested position and stratified median estimators for stratified samples (Hussein, 1998). The author had also suggested a separate and combined ratio estimators (Hussein, 1999). Most of the researches of median estimation deals exclusively with the survey variable of interest alone and do not use the auxiliary variables in the construction of the estimators (Kuk and Mak 1989, Rao and shao 1996). This paper deals with the estimation, under simple random sampling of a finite population median in the presence of more than one auxiliary variables. The two proposed estimator are used to estimate the median water per capita in water short countries.

I. Position Estimator in the presence of more than one auxiliary variables:

The estimator defined here attempts to utilize two auxiliary variables X_{11} , X_{12} to construct a p with smaller expected mean square error and consequently smaller variance for estimating the population median.

Consider the three-way classification:

	$X_1 \leq M_{X_1}$ $X_2 \leq M_{X_2}$	$X_1 \leq M_{X_1}$ $X_2 > M_{X_2}$	$X_1 > M_{X_1}$ $X_2 \leq M_{X_2}$	$X_1 > M_{X_1}$ $X_2 > M_{X_2}$
$Y \leq M_T$	P_{111}	P_{112}	P_{121}	P_{122}
$Y > M_T$	P_{211}	P_{212}	P_{221}	P_{222}
	$P_{.11}$	$P_{.12}$	$P_{.21}$	$P_{.22}$

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Let $Y_{(1)} \leq Y_{(2)} \leq \dots \leq Y_{(n)}$ be the ordered Y values in S_n . Let $n_{x_1 x_2}$ be the number of units in S_n with $X_1 \leq M_{X_1}$ and $X_2 \leq M_{X_2}$. Let n_{x_1} be the number of units in S_n with $X_1 \leq M_{X_1}$ and $X_2 > M_{X_2}$. Let n_{x_2} be the number of units in S_n with $X_1 > M_{X_1}$ and $X_2 \leq M_{X_2}$.

So, we can propose a position estimator of P as an extension of the estimation given by Kuk and Mak (1988) as follows:

$$\hat{P}_B = n^{-1} \left\{ n_{x_1 x_2} \frac{P_{111}}{P_{.11}} + n_{x_1} \frac{P_{112}}{P_{.12}} + n_{x_2} \frac{P_{121}}{P_{.21}} + (n - n_{x_1} - n_{x_2} - n_{x_1 x_2}) \frac{P_{122}}{P_{.22}} \right\}$$

Thus an estimator of M_Y is alternatively given by $M_{BY} = Q_Y(P_B)$ Where Q_Y is the commutative distribution function.

Since

$$\begin{aligned} P_{.11} &= \frac{n_{x_1 x_2}}{n}, \quad P_{.12} = \frac{n_{x_1}}{n} \\ P_{.21} &= \frac{n_{x_2}}{n}, \quad P_{.22} = \frac{n - n_{x_1} - n_{x_2} - n_{x_1 x_2}}{n} \\ \therefore V(\hat{P}_B) &= \frac{1-f}{n} \left\{ n_{x_1 x_2} P_{111}(1 - P_{111}) + n_{x_1} P_{112}(1 - P_{112}) \right. \\ &\quad \left. + n_{x_2} P_{121}(1 - P_{121}) + (n - n_{x_1} - n_{x_2} - n_{x_1 x_2}) P_{122}(1 - P_{122}) \right\} \end{aligned}$$

where $f = \frac{n}{N}$

Considering the first auxiliary variable alone

	$X_1 \leq M_{X_1}$	$X_1 > M_{X_1}$
$Y \leq M_Y$	P_{11}	P_{12}
$Y > M_Y$	P_{21}	P_{22}
	$P_{.1}$	$P_{.2}$

$$\begin{aligned} \hat{P}_{B_1} &= n^{-1} \left\{ n_X \frac{P_{11}}{P_{.1}} + (n - n_{X_1}) \frac{P_{12}}{P_{.2}} \right\} \\ V(\hat{P}_{B_1}) &= \frac{1-f}{n} \left\{ n_X P_{11}(1 - P_{11}) + (n - n_X) P_{12}(1 - P_{12}) \right\} \end{aligned}$$

II. Comparison of the Position Estimator in the Presence of Two Auxiliary Variables with that of One Auxiliary Variable.

$$\text{Since } V(\hat{P}_{B_1}) = \frac{1-f}{n} \{ n_x P_{11}(1-P_{11}) + (n-n_x)P_{12}(1-P_{12}) \}$$

$$\text{and } n_x = n_{x_1x_2} + n_{x_1}$$

$$n - n_x = n_{x_2} + (n - n_{x_1} - n_{x_2} - n_{x_1x_2})$$

$$\therefore V(\hat{P}_{B_1}) = \frac{1-f}{n} \{ n_{x_1x_2} + n_{x_1} \} (P_{111} + P_{112})(1 - P_{111} - P_{112})$$

$$+ (n_{x_2} + (n - n_{x_1} - n_{x_2} - n_{x_1x_2}))(P_{121} + P_{122})(1 - P_{121} - P_{122}) \}$$

$$= \frac{1-f}{n} \{ n_{x_1x_2} P_{111}(1 - P_{111}) + n_{x_1} P_{112}(1 - P_{112}) + n_{x_2} P_{121}(1 - P_{121})$$

$$+ n - n_{x_1} - n_{x_2} - n_{x_1x_2} P_{122}(1 - P_{122}) \}$$

$$+ n_{x_1x_2} P_{112}(1 - P_{112} - 2P_{111}) + n_{x_1} P_{111}(1 + P_{111} - 2P_{112})$$

$$+ n_{x_2} P_{122}(1 - P_{122} - 2P_{121}) + (n - n_{x_1} - n_{x_2} - n_{x_1x_2})$$

$$P_{121}(1 - P_{121} - 2P_{122})$$

$$\therefore V(\hat{P}_{B_1}) = V(P_B) n_{x_1x_2} P_{112}(1 - P_{112} - 2P_{111})$$

$$+ n_{x_1} P_{111}(1 - P_{111} - 2P_{112}) + n_{x_2} P_{122}(1 - P_{122} - 2P_{121})$$

$$+ (n - n_{x_1} - n_{x_2} - n_{x_1x_2}) P_{121}(1 - P_{121} - 2P_{122})$$

then $V(\hat{P}_{B_1}) > V(P_B)$ if

$$\begin{aligned} & P_{111} + 2P_{112} < 1 \quad \text{and} \quad P_{112} + 2P_{121} < 1 \\ \text{and} \quad & P_{112} + 2P_{111} < 1 \quad \text{and} \quad P_{121} + 2P_{122} < 1 \end{aligned}$$

These conditions can be satisfied if

$$3P_{111} + 3P_{112} + 3P_{121} + 3P_{122} < 4$$

$$P_{111} + P_{112} + P_{121} + P_{122} < \frac{4}{3}$$

Which is always satisfied

$$\text{Since } P_{111} + P_{112} + P_{121} + P_{122} < 1$$

III. Bivariate Ratio Type Median Estimator:

Let $(X_{11}, X_{12}, Y_1), \dots, (X_{n_1}, X_{n_2}, Y_n)$ be the associated values of the variables X_1, X_2, Y for the units in S_n . When the values of the

auxiliary variables X_{i1} , X_{i2} are available a natural modification of the ratio estimator proposed by Kuk and Mak (1988) can be expressed as follows:

$$\text{Let } M_{YR_1} = M_Y M_{X_1} / M_{X_1}$$

$$\text{and } M_{YR_2} = M_Y M_{X_2} / M_{X_2}$$

$$\text{Let } M_{BYR} = W_1 M_Y M_{X_1} / M_{X_1} + W_2 M_Y M_{X_2} / M_{X_2}$$

Where W_i are weights to be determined to maximize the precision of \hat{M}_{BYR} subject to $\sum w_i = 1$

The method is described for two X- varieties since this should be the most frequent application.

Hence,

$$V(M_{BYR}) = W_1^2 V(M_{YR_1}) + 2W_1 W_2 \text{Cov}(M_{YR_1}, M_{YR_2}) + W_2^2 V(M_{YR_2})$$

$$\begin{aligned} \text{Let } V_{11} &= V(M_{YR_1}) \\ V_{12} &= \text{Cov}(M_{YR_1}, M_{YR_2}) \\ V_{22} &= V(M_{YR_2}) \end{aligned}$$

$$\text{Then } V(\hat{M}_{BYR}) = W_1^2 V_{11} + 2W_1 W_2 V_{12} + W_2^2 V_{22}$$

Where the values of W_1 , W_2 can be determined by minimizing $V(\hat{M}_{BYR})$ under the condition

$$W_1 + W_2 = 1$$

It follows that

$$W_1 = \frac{V_{22} - V_{12}}{V_{11} + V_{22} - 2V_{12}} \quad W_2 = \frac{V_{11} - V_{12}}{V_{11} + V_{22} - 2V_{12}}$$

then

$$M_{\min}(M_{BYR}) = \frac{V_{11}V_{22} - V_{12}^2}{V_{11} + V_{22} - 2V_{12}}$$

where

$$V_{11} = n^{-1}(1-f) \left[\frac{1}{4} \{f_T(M_Y)\}^{-2} + \frac{1}{4} \left(\frac{M_Y}{M_{X_1}} \right)^2 \{f_{X_1}(M_{X_1})\}^{-2} \right]$$

$$\begin{aligned}
& -2\left(\frac{M_Y}{M_{X1}}\right)\{f_Y(M_Y)f_{X1}(M_{X1})\}^{-1}\left(P_{11}-\frac{1}{4}\right) \\
V_{22} = n^{-1}(1-f) & \left[\frac{1}{4}\{f_Y(M_Y)\}^{-2} + \frac{1}{4}\left(\frac{M_T}{M_{X2}}\right)^2\{f_{X2}(M_{X2})\}^{-2}\right. \\
& \left.-2\left(\frac{M_T}{M_{X2}}\right)\{f_Y(M_Y)f_{X2}(M_{X1})\}^{-1}\left(P_{11}-\frac{1}{4}\right)\right]
\end{aligned}$$

$$V_{12} = \text{Cov}(M_{YR1}, M_{YR2})$$

$$M_{YR1} - M_Y = (M_{X1}M_Y - M_YM_{X1})/M_{X1}$$

Since $M_{X1}/M_{X1} \rightarrow 1$ in probability

$M_{YR1} - M_Y$ has the same asymptotic distribution as

$$(M_{X1}M_Y - M_YM_{X1})/M_{X1} = (M_Y - M_Y) - \left(\frac{M_Y}{M_{X1}}\right)(M_{X1} - M_{X1})$$

In the same way

$M_{YR2} - M_Y$ has the same asymptotic distribution as

$$(M_{X2}M_Y - M_YM_{X2})/M_{X2} = (M_Y - M_Y) - \left(\frac{M_Y}{M_{X2}}\right)(M_{X2} - M_{X2})$$

then

$$\begin{aligned}
& \text{Cov}(M_{YR1}, M_{YR2}) \\
& = \text{Cov}\left[\left\{(M_Y - M_Y) - \frac{M_Y}{M_{X1}}(M_{X1} - M_{X1})\right\}, \right. \\
& \quad \left.\left\{(M_Y - M_Y) - \frac{M_Y}{M_{X2}}(M_{X2} - M_{X2})\right\}\right] \\
& = V(M_Y) - \frac{M_Y}{M_{X2}}\text{Cov}(M_Y, M_{X2}) - \frac{M_Y}{M_{X1}}\text{Cov}(M_Y, M_{X1}) \\
& \quad + \frac{M_Y^2}{M_{X1}M_{X2}}\text{Cov}(M_{X1}, M_{X2})
\end{aligned}$$

IV. The Mean Square Error of the Bivariate Ratio Type Median Estimator Compared to that of the Ratio Type Median Estimator

$$V(M_{YR}) - M_{\min}(M_{BYR}) =$$

$$V_{11} - \frac{V_{11}V_{22} - V_{12}^2}{V_{11} + V_{22} - 2V_{12}} = \frac{(V_{11} - V_{12})^2}{V_{11} + V_{22} - 2V_{12}}$$

$$M_{\min}(M_{BYR}) < M(M_{YR}) \quad \text{if}$$

$$V_{11} + V_{22} - 2V_{12} > 0$$

$$\frac{1}{2}(V_{11} + V_{22}) > V_{12}$$

This inequality is always satisfied since the equality holds only if the two auxiliary variables are identical, that is:

$$E(M_{YR1} - M_{YR1} - M_{YR2} + M_{YR2})^2 \geq 0$$

$$E[(M_{YR1} - M_{YR1}) - (M_{YR2} - M_{YR2})]^2 \geq 0$$

$$E(M_{YR1} - M_{YR1})^2 + E(M_{YR2} - M_{YR2})^2$$

$$- 2E(M_{YR1} - M_{YR1})(M_{YR2} - M_{YR2}) \geq 0$$

$$V(M_{YR1}) + V(M_{YR2}) - 2\text{Cov}(M_{YR1}, M_{YR2}) \geq 0$$

$$V_{11} + V_{22} \geq 2V_{12}$$

$$\frac{1}{2}(V_{11} + V_{22}) \geq V_{12}$$

V. Numerical Example:

Consider data for three variables where the dependent variable is water per capita with cubic meters per year in 1996. The first auxiliary variable is the population in millions and the second auxiliary variable is the population growth rate. A simple random sample of 24 countries is chosen from 48 water short countries, table (2) illustrates the three way classification of the sample according to the dependent variable and the two auxiliary variables. Table (3) and Table (4) illustrate the two way classification according to the dependent variable and one of the auxiliary variable at a time. The distribution of the dependent variable is skewed since the skewness measure is .413. So, the median estimator is more appropriate than the

mean. Some statistical measures are illustrated in table (1) for the population and the sample. The statistics in the table are used to estimate the median water per capita using the two proposed estimators. The estimates and their relative efficiencies are illustrated in tables (6) and (7). Using the ratio type median estimator is not good in this example since the correlation coefficients between the dependent variable and the auxiliary variables are very low. These correlations are illustrated in table (2).

Statistic	Population	Sample	Population	Sample
Mean	1227.21	1227.21	1227.21	1227.21
Std dev	100.00	100.00	100.00	100.00
Kurtosis	3.00	3.00	3.00	3.00
Skewness	0.00	0.00	0.00	0.00
Range	1000	1000	1000	1000
Maximum	2227	2227	2227	2227
Minimum	1227	1227	1227	1227
Percentile				
25	1227.21	1227.21	1227.21	1227.21
50	1227.21	1227.21	1227.21	1227.21
75	1227.21	1227.21	1227.21	1227.21

Table (2)
The Correlation Between the Dependent
and the Auxiliary Variables

Gr	Wpc	Gr
1	1018	1018
2	1018	1018
3	1018	1018
4	1018	1018
5	1018	1018
6	1018	1018
7	1018	1018
8	1018	1018
9	1018	1018
10	1018	1018

Table (3)

Water per capita by population

(in thousands)

Method of Estimation	Estimate	Standard Error	Relative Efficiency
1. Simple random sampling	1227.21	100.00	100.00
2. Systematic sampling	1227.21	100.00	100.00
3. Stratified sampling	1227.21	100.00	100.00
4. Cluster sampling	1227.21	100.00	100.00
5. Ratio type median estimator	1227.21	100.00	100.00
6. Product type median estimator	1227.21	100.00	100.00
7. Ratio type median estimator	1227.21	100.00	100.00
8. Product type median estimator	1227.21	100.00	100.00
9. Ratio type median estimator	1227.21	100.00	100.00
10. Product type median estimator	1227.21	100.00	100.00

Table (1)
Population and Sample Characteristics of
The Dependent and the Auxiliary Variables

	Population			Sample		
	Wpc	Pop	Gr	Wpc	Pop	Gr
Mean	1400.85	36.80	2.10	1371.59	19.24	1.84
Std dev	983.11	133.46	.95	806.77	21.10	.988
Kurtosis	-.787	45.09	-.29	-.47	.25	-.853
Skewness	.413	6.63	-.37	0.43	1.17	-.452
Range	3470	928.70	3.80	2882.0	68.0	3.20
Maximum	3552	929.0	3.90	2964.0	68.40	3.30
Minimum	82.0	.30	.10	82.0	.40	.10
Percentile						
25	457.25	2	1.60	400.25	2.25	1.53
50	1228.0	9.35	2.15	1215	16.15	2.05
75	1965	27.03	2.85	1820.25	38.15	2.65

Table (2)
The Correlation Between the Dependent
and the Auxiliary Variables

	Wpc	Pop	Gr
Wpc	1	.1618	.2522
Pop		1	-.0395
Gr			1

Table (3)
Water per capita by population
and growth rate

	$X_1 \leq MX_1$	$X_1 \leq MX_1$	$X_1 > MX_1$	$X_1 > MX_1$
	$X_1 \leq MX_2$	$X_1 > MX_2$	$X_1 \leq MX_2$	$X_1 > MX_2$
$Y \leq M_Y$	4	3	2	2
	(.1666)	(.125)	(.083)	(.083)
$Y > M_Y$	1	1	5	6
	(.042)	(.0416)	(.208)	(.25)
	5	4	7	8
	(.20833)	(.0416)	(.208)	(.25)

$$\hat{P}_B = .5243804$$

$$\therefore M_Y = 1425.6$$

$$V(\hat{P}_B) = .04736$$

Table (4)
Water Per Capita by Population

	$X_1 \leq MX_1$	$X_1 > MX_1$
$Y \leq M_T$	7 (.29166)	4 (.1666)
$Y > M_T$	2 (.08333)	11 (.458333)
	9 (.375)	15 (.625)
$\hat{P}_{B_1} = .4582562$		$M_Y = 1227.46$
$V(\hat{P}_{B_1}) = .0821305$		

Table (5)

	$X_2 \leq MX_2$	$X_2 > MX_2$
$Y \leq M_T$	6 (.25)	5 (.20833)
$Y > M_T$	6 (.25)	7 (.29166)
$\hat{P}_{B_2} = .458333$		$M_Y = 1227.49$
$V(\hat{P}_{B_2}) = .0881057$		

Table (6)
**Relative Efficiency of The Position
Median Estimator**

Method of Estimation	Estimate	Stand error	Relative efficiency
Position (first variable)	1227.46	.28652	35.95
Position (second variable)	1337.49	.29683	33.51
Position (Two auxiliary)	1425.6	.21762	6234
Median estimator	1228	1.171829	1.00

Table(7)
Relation Efficiency of the Bivariate
Ratio Median estimator

Method of Estimation	Estimates	Relative efficiency
Ratio Median(first variable)	703	1.006
Ratio Median (second variable)	1158	0.740
Ratio Median (two auxiliary variables)	714	1.007
Median estimator	1228	1.00

IV Conclusion:

From table (5) note that using the position median estimator in presence of auxiliary information increased the precision of the estimates than using the dependent variable alone. Using two auxiliary information increased the precision 62 times that using the dependent variable alone.

From table (6) we note also that using the ratio median estimator in presence of auxiliary information increased the precision of the estimates except using the second auxiliary variable alone. Adding auxiliary variables increased the precision although this is not good example for ratio type median estimator. In this example the position estimator works better the comparison between the two approaches is very difficult since the theory behind the two approaches are radically different.

References:

1. *Bandyopadhyay, S.* (1980), Improved Ratio and Product Estimators, *Sankhya*, 2, series c. pt.1 and 2, pp. 45 – 49.
2. *Cochran, W.G.*, (1977), Sampling techniques, 3rd edition, John Wiley New York.
3. *Chang D.S., S.*(1986), The Asymptotic Distribution of multivariate Producer Estimator, *Chinese Journal of mathematics*, Vol. 14, No.3.
4. *Hussein, M.A.* (1988), On Increasing the Precision of Finite Population Estimators Using Auxiliary Variables, unpublished ph. D. Thesis University of Iowa USA.
5. *Hussein, M.A.* (1992), Alternative Estimators for Stratified Random Sampling The Egyptian Statistical Journal, ISSR Cairo University, Vol 35 No. 2, 1992.
6. *Hussein M.A.* (1995), An Estimator in Cluster Sample Combined with Stratification. The Egyptian Statistical Journal, ISSR, Cairo University, Vol 46 No.2,.
7. *Hussein M.A.* (1998), Estimator Finite Population Using Stratified Sample in the Presence of the Auxiliary Information. The Egyptian Statistical Journal, ISSR, Cairo University, Vol 15 No.
8. *Hussein M.A.* (1999), Separate Combined Ratio Estimator for the Median. The Egyptian Statistical Journal, ISSR, Cairo University, Vol 19 No.
9. *Ouyang, Z., Srivastaya J.N. and Schreuder, H.T.* (1992), IKA General Ratio Estimator and Its Application in Model Based Inference, *Ann. Inst. Statist. Math.* Vol. 45, No. 1, 113 - 127.
10. Population Reports, Volume xxv1, Number 1, September 1998.
11. *Sekkappan R.M.* (1996), Estimation in Sampling From Finite Populations Under the General Linerar Regression Model. *Journal of Indian Statistical association*, Vol. 24, 91 – 98.
12. *Scheaffer, R.L., Mendenhall, W. and Ott, L.* (1993), Elementary Survey Sampling. Wadsworth, Inc.
13. *Srivenkataramana, T.* (1980), Dual to the Ratio Estimator in Sample Surveys, *Biometrika*, 67, 1. Pp. 193 – 204.