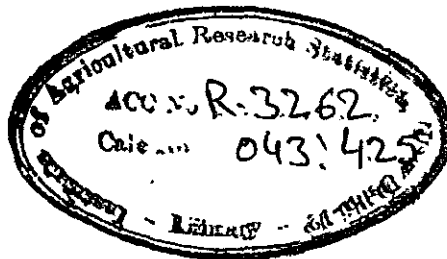


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**USE OF ANCILLARY INFORMATION IN ENSURING A  
REPRESENTATIVE SAMPLE**

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**Dissertation submitted in fulfillment of the  
requirements of Post-Graduate Diploma  
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## CERTIFICATE

This is to certify that the work incorporated in the dissertation entitled ' USE OF ANCILLARY INFORMATION IN ENSURING A REPRESENTATIVE SAMPLE ' by VIJAY KUMAR and submitted for the award of Post-Graduate Diploma in Agricultural Statistics of the Institute of Agricultural Research Statistics ( I.C.A.R. ) New Delhi was done under my guidance.

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## CHAPTER - 1

### INTRODUCTION

The importance of sampling in providing data with greater speed, greater accuracy and lesser cost is well recognised. The element or group of elements on which information can be taken is known as the 'sampling unit'. The group of units specified by the objectives of survey is referred to as 'population'. (A population having ~~finite~~ number of units is called 'finite' and a population having uncountable number of units is called 'infinite'. In sampling, we generally deal with finite populations and derive approximations for infinite populations.

The population to be sampled is called 'Sampled Population' and the population about which information is wanted is called 'Target Population'. Former is more restricted than the latter.

Part of population selected by some procedure is called a 'sample' and the process of selection is called 'sampling'. A 'representative sample' is that sample which has characteristics similar to that of the population. Sampling is said to be 'with replacement' if the unit once selected is replaced back into the population for further selection and 'without replacement' if the unit once selected is not given the chance of further selection in the sample. The procedure of selecting the sample unit by unit is known as 'sampling scheme'. The set of all possible samples is called 'sample space'. The sample

space together with the associated probability measure is called the 'sampling design'. A sampling scheme gives rise to a sampling design. The procedure of estimating population parameter together with the standard error on the basis of sample values is called 'Estimation Procedure'.

Let  $y$  be the character defined on the population which takes value  $y_i$  for  $i$ -th unit of the population. Any function of population values is termed as the 'parameter'. The corresponding function based on sample values is known as 'estimator'. The particular value which the estimator takes is called the 'estimate'. A sampling design together with an estimation procedure is termed as 'sampling strategy'.

A scientific and objective method of selecting the sample units and collecting information on selected units of the population is called a sample survey or a 'survey' and collecting information on each unit of the population is called complete enumeration or 'census'. A sample survey helps in deeper scientific investigation of a population with limited trained staff. It also provides the measure of error involved in estimating population parameter such as mean etc. The census is used when cost is no consideration, characters can be easily recorded, non sampling errors are not large and detailed break up for smallest administrative unit is required.

The standard error is generally taken as 'measure of error'.

The character in which we are interested is called the 'study character'. In most of the cases another character highly correlated with the study character is available for all the units of the population. Such a character is called 'auxiliary character' and the information based on it the 'auxiliary information'. Such information is used for obtaining efficient results in estimation of population mean or total. The auxiliary information can be made use of either for selection of sampling units from the population as in PPS sampling or for stratification of the units in the population or for ratio and regression methods of estimation.

When the units vary considerably in size, the selection of sampling units with simple random sampling is not advantageous since it does not take into account the importance to be attached to larger units in the population. In such situations it becomes utmost important to select the sampling units with probability proportional to size ( PPS ). The PPS sampling makes use of the available information on some auxiliary character. This scheme was initially proposed by Hansen and Horwitz ( 1949 ). Das ( 1951 ), Narain ( 1951 ), Horwitz and Thompson ( 1952 ), Lahiri ( 1951 ), Nidusuo ( 1952 ), Yates and Grundy ( 1953 ), Des Raj ( 1956 ), Hartley and Rao ( 1962 ), Rao, Hartley and

Cochran (1962), Hanurav (1967), Durbin (1967), Sampford (1962), Das and Mohanty (1973) and many others also developed various methods for selection of sampling units with varying probabilities.

In stratification, the population is to be divided into strata such that within strata variation is as small as possible. Dalenius (1950) attempted this problem of determining optimum points of stratification on the basis of study variable. The method used by him was iterative in nature and required the knowledge of certain parameters which are functions of boundary points. Dalenius and Gurney (1951) suggested the use of information on auxiliary character for the purpose of stratification. Taga (1967) and Singh (1968) also obtained points of stratification for the character under study based on auxiliary character.

Further, the suitable use of auxiliary character at the estimation stage also results in considerable reduction in variance of the estimate. The ratio and regression methods of estimation make use of auxiliary character for obtaining estimates of population parameters. The ratio method of estimation was first introduced by Cochran (1940) whereas theoretical basis of regression method of estimation was also first discussed by



Cochran (1942). Product method of estimation was first proposed by Murthy (1964). Sukhatme (1944), Quenouille (1956), Robson (1957), Olkin (1958), Goodman and Hartley (1958), Nansama, Murthy and Sethi (1959), Singh (1965), T. L. S. (1965) and others made further contributions to the theory of ratio, regression and product methods of estimation.

Of all the sampling procedures considered above, none provides any control on sampling error which is an important aspect of sampling. In the procedure under investigation an attempt has been made to use the ancillary information for selecting a representative sample i.e. a sample for which relative error of sample estimate from population parameter is within specified margin. It is of importance to study the probability that sample estimate for study character will also differ from population parameter by specified margin of error. This has been investigated in chapter - II and its behaviour has been studied with sample size, correlation coefficient, coefficient of variation and margins of error. For the procedure suggested in chapter - II, the inclusion probabilities for individual <sup>and</sup> pairwise units have been worked out in chapter - III to get Horvitz Thompson estimator. Further, the selection procedure has been suitably modified providing non-zero inclusion probabilities for individual and pairwise units.

{ This definition is somewhat different from the one normally used.

A sampling technique is defined as introducing control into the selection of  $n$  out of  $N$  sampling units when it increases the probabilities of selection for preferred samples and thus decreases the probabilities for non preferred samples. The objective of controlled selection is to reduce the variances of estimates for most of sampling techniques at a given cost. Controlled selection in probability sampling was first introduced by Goodman and Kish (1950). Further, Avadhani and Sukhatne (1965) and (1966) proposed controlled simple random sampling and its use in ratio and regression method of estimation.

This concept of controlled selection has been used in the suggested procedure to provide larger probabilities of selection to the representative samples as compared to the remaining samples. The calculation of inclusion probabilities and the estimation procedure have also been discussed for the controlled selection. The empirical investigations have been carried out to study the relative efficiencies of the suggested procedures with some existing procedures.

## CHAPTER - II

### USE OF ANCILLARY INFORMATION IN ENSURING A REPRESENTATIVE SAMPLE

**2.1 Introduction :** The main aim of sampling in general consists in selecting a sample and then building an estimator of the population parameter. In practice it is desirable that the estimator so formed should be within some preassigned margin of error from the corresponding parameter. Many times the information on all the units of the population is available for the auxiliary character, highly correlated with the character under study. The use of this ancillary information in selecting sample whose mean is not different from the population mean for the auxiliary character beyond a specified margin of error has been suggested in this chapter. The procedure of selecting such sample has been explained with an example. The probability of estimating the population mean with desired margin of error has also been worked out. The properties of this probability have been examined in respect of sample size, correlation coefficient between study and auxiliary variables, <sup>and</sup> margins of error. Some numerical illustrations have also been given with different values of these parameters.

**2.2 The Suggested Procedure :** Suppose that the population under study consists of  $N$  distinct and identifiable units. Let  $y_i$  and  $x_i$

be the values of the study and auxiliary characters for the  $i$ -th unit of the population. Further, let  $x_i$  be known for all  $i$ , ( $i = 1, 2, \dots, N$ ). Then the suggested procedure consists of the following steps :

Step - I: Select a simple random sample without replacement of size  $n$  from the population of size  $N$ .

Step - II: Calculate  $(\bar{x} - \mu_x) / \mu_x$  for the selected sample and test whether  $|(\bar{x} - \mu_x) / \mu_x| \leq e_x$  for some preassigned  $e_x$ , where  $\bar{x}$  and  $\mu_x$  are respectively the sample mean and population mean for the character  $x$ .

Step - III: If  $|(\bar{x} - \mu_x) / \mu_x| \leq e_x$ , retain the sample, otherwise proceed to step - I for selection of another sample till a sample satisfying  $|(\bar{x} - \mu_x) / \mu_x| \leq e_x$  is obtained.

2.2.1 Illustrative Example: Suppose that we have a population of size 6 with  $x$  values given in the table below,

Sr. No.	1	2	3	4	5	6
$x$	1	2	2	3	3	4

Suppose a sample of size 4 is desired to be drawn from the 15 possible samples of size 4 listed ~~below~~. If  $e_x$  is .05 then the five samples which can be selected are at Sr. Nos. 5, 6, 8, 9 and 11.

Sample No.	Sample units	$\bar{x}$	$ (\bar{x} - \mu_x) / \mu_x $
1	1, 2, 3, 4	2.00	.20
2	1, 2, 3, 5	2.00	.20
3	1, 2, 3, 6	2.25	.10
4	1, 2, 4, 5	2.25	.10
5	1, 2, 4, 6	2.50	.00
6	1, 2, 5, 6	2.50	.00
7	1, 3, 4, 5	2.25	.10
8	1, 3, 4, 6	2.50	.00
9	1, 3, 5, 6	2.50	.00
10	1, 4, 5, 6	2.75	.10
11	2, 3, 4, 5	2.50	.00
12	2, 3, 4, 6	2.75	.10
13	2, 3, 5, 6	2.75	.10
14	2, 4, 5, 6	3.00	.20
15	3, 4, 5, 6	3.00	.20

2.3. Probability of estimation : Under this procedure

it is of interest to find the probability of estimating the population mean for  $y$  with the relative margin of error  $e_y$ . i.e.

$$P \left[ |(\bar{y} - \mu_y) / \mu_y| \leq e_y \mid |(\bar{x} - \mu_x) / \mu_x| \leq e_x \right] = P(y | x)$$

where  $\bar{y}$  and  $\mu_y$  are respectively the sample mean and the population mean for character  $y$ . To calculate this probability let us assume that  $x$  and  $y$  follow a bivariate normal distribution with coefficient of correlation  $\rho$ . This joint probability density function of  $(\bar{x}, \bar{y})$  is given by :

$$f(\bar{x}, \bar{y}) = \frac{n}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \exp\left[\frac{-1}{2(1-\rho^2)} \left\{ \left(\frac{\bar{x}-\mu_x}{\sigma_x/\sqrt{n}}\right)^2 + 2\rho\left(\frac{\bar{x}-\mu_x}{\sigma_x/\sqrt{n}}\right)\left(\frac{\bar{y}-\mu_y}{\sigma_y/\sqrt{n}}\right) + \left(\frac{\bar{y}-\mu_y}{\sigma_y/\sqrt{n}}\right)^2 \right\}\right]$$

where  $\sigma_x$  and  $\sigma_y$  are the standard deviations for  $x$  and  $y$  respectively. The required probability is then given by:

$$P(y|x) = \frac{\int_{\bar{x}}^{\infty} \int_{-\infty}^{\infty} f(\bar{x}, \bar{y}) d\bar{x} d\bar{y}}{\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\bar{x}, \bar{y}) d\bar{x} d\bar{y}}$$

with the limits for  $\bar{x}$  and  $\bar{y}$  as under,

$$\mu_x(1-\epsilon_x) \leq \bar{x} \leq \mu_x(1+\epsilon_x) \quad \text{and}$$

$$\mu_y(1-\epsilon_y) \leq \bar{y} \leq \mu_y(1+\epsilon_y).$$

For evaluating  $P(y|x)$  we put

$$\frac{\bar{y}-\mu_y}{\sigma_y/\sqrt{n}} = u \quad \text{with} \quad d\bar{y} = \frac{\sigma_y}{\sqrt{n}} du \quad \text{and}$$

$$\frac{\bar{x}-\mu_x}{\sigma_x/\sqrt{n}} = v \quad \text{with} \quad d\bar{x} = \frac{\sigma_x}{\sqrt{n}} dv. \quad \text{Thus we get .}$$

$$P(y|x) = \frac{\int_{u_0}^{\infty} \int_{v_0}^{\infty} f(u, v) dv du}{\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(u, v) dv du}$$

$$\text{where } f(u, v) = \frac{1}{2\pi\sqrt{1-\rho^2}} \exp. \left[ \frac{-1}{2(1-\rho^2)} (u^2 - 2\rho uv + v^2) \right]$$

The limits for  $u$  and  $v$  in the integral will be as under,

$$-\frac{c_x \mu_x}{\sigma_x / \sqrt{n}} \leq v \leq \frac{c_x \mu_x}{\sigma_x / \sqrt{n}}$$

$$-\frac{c_y \mu_y}{\sigma_y / \sqrt{n}} \leq u \leq \frac{c_y \mu_y}{\sigma_y / \sqrt{n}}$$

The above probability can be calculated if  $\rho$ ,  $c_x$ ,  $c_y$ ,

$\sigma_x / \mu_x = c_x$ ,  $\sigma_y / \mu_y = c_y$  and  $n$  are known. It may be

mentioned here that most of the parameters such as  $\rho$  and  $c_y$

are generally unknown. But some guess value of these parameters

can always be known from the previous data on the same characteristics

or from the sample selected.

#### 3.4 Numerical Illustration :

To study the behaviour of above

probability with sample size, correlation coefficient, the coefficients

of variation and the margin of error, some empirical results have

been given below in tables 2.1 to 2.5. To calculate this probability,

'Tables for Statisticians and Biometricians', Part 2 by Pearson.

H. (Tables VIII and IX) have been used. These tables provide

the value of  $\iint_{uv} f(u, v) dv du$  for positive  $\rho$  as well as negative  $\rho$ ,

Making use of these tables,

$$P(y|x) = \frac{\int_{-h}^h \int_{-h}^h f(u, v) dv du}{\int_{-\infty}^{\infty} \int_{-h}^h f(u, v) dv du} \quad \text{with } h = \frac{\sigma_y \mu_y}{\sigma_y / \sqrt{n}} \quad \text{and}$$

$$k = \frac{\sigma_x \mu_x}{\sigma_x / \sqrt{n}} \quad \text{has been computed as described below:}$$

The numerator of  $P(y|x)$  is given by

$$2 \left[ \int_0^{\infty} \int_0^{\infty} f(u, v) dv du + \int_h^{\infty} \int_h^{\infty} f(u, v) dv du \right. \\ \left. - \int_h^{\infty} \int_0^{\infty} f(u, v) dv du - \int_0^{\infty} \int_h^{\infty} f(u, v) dv du \right] \\ + 2 \text{ ( same four integrals with } \rho \text{ negative ) .}$$

Similarly the denominator of the required probability  $P(y|x)$  is

$$\text{given by } 2 \left[ \int_0^{\infty} \int_0^{\infty} f(u, v) dv du + \int_{-\infty}^{\infty} \int_h^{\infty} f(u, v) dv du \right. \\ \left. - \int_{-\infty}^{\infty} \int_0^{\infty} f(u, v) dv du - \int_{-\infty}^{\infty} \int_h^{\infty} f(u, v) dv du \right] \\ + 2 \text{ ( same four integrals with } \rho \text{ negative ) .}$$

As tables provide values of bivariate integral for the range of

$h$  and  $k$  from 0 to 2.6 only, the values for  $h$  and  $k$  ( $> 2.6$ ) have been taken as 2.6 in the integration as an approximation.

All probabilities have been rounded upto 4 digits for simplicity.



Tables showing the value of the probability for different values of  $n$  and  $\rho$

Table 2.1  $\epsilon_x = \epsilon_y = .01, c_x = c_y = .10$

$n/\rho$	.5	.6	.7	.8	.9
16	.3536	.3784	.4145	.4765	.7884
36	.5047	.5330	.5750	.6365	.7986
64	.6313	.6580	.6949	.7464	.8372
100	.7319	.7544	.7832	.8215	.8778

Table 2.2  $\epsilon_x = \epsilon_y = .01, c_x = c_y = .05$

$n/\rho$	.5	.6	.7	.8	.9
16	.6313	.6580	.6949	.7464	.8372
36	.8102	.8185	.8476	.8742	.9116
64	.9135	.9205	.9228	.9401	.9569
100	.9679	.9699	.9721	.9761	.9815

Table 2.3  $\epsilon_x = \epsilon_y = .05, c_x = c_y = .10$

$n/\rho$	.5	.6	.7	.8	.9
16	.9679	.9699	.9721	.9761	.9815
36	1.0000	1.0000	1.0000	1.0000	1.0000
64	1.0000	1.0000	1.0000	1.0000	1.0000
100	1.0000	1.0000	1.0000	1.0000	1.0000

Table : 2.4  $\epsilon_x = .01, \epsilon_y = .05, c_x = c_y = .10$

n/p	.5	.6	.7	.8	.9	1.0
16	.9804	.9886	.9936	.9986	1.0000	1.0000
36	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
64	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
100	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000

Table : 2.5  $\epsilon_x = \epsilon_y = .05, c_x = c_y = .05$

n/p	.5	.6	.7	.8	.9	1.0
16	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
36	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
64	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
100	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000

It is clear from the above table, that the probability  $P(y|x)$  increases monotonically with increase of  $n$  the sample size,  $\rho$  the correlation coefficient,  $\epsilon_y$  margin of error for  $y$  and decreases with increase of coefficient of variation and  $\epsilon_x$  the margin of error for  $x$ . Further, it can be seen that for  $\epsilon_x = \epsilon_y = .05$  and  $\epsilon_x = .01, \epsilon_y = .05$  the probability is of the order 1 but for  $\epsilon_x = \epsilon_y = .01$  it is of high order *only* for high correlation or large  $n$ . Thus in practical situations with  $\epsilon_x = \epsilon_y = .05$  or  $\epsilon_x = .01, \epsilon_y = .05$  and for known  $c_y, c_x, n$  and  $\rho$  the suggested procedure can be used satisfactorily as the probability of estimation of the character under

study within given margin of error is of high order.

2.5 Summary : In this chapter , a method of utilising socially<sup>alt</sup> information, in ensuring a representative sample, has been suggested. The probability of estimating the mean of the character under study within the specified margin of error has been obtained for the suggested procedure. It has been seen that this probability increases monotonically with sample size , the correlation coefficient between study and auxiliary variables and the margin of error for the study variable. It has also been observed that generally the probability of estimating the mean of study character is of high order and thus the suggested procedure can be used satisfactorily.

MODIFIED PROCEDURES FOR UNBIASED ESTIMATION

3.1 Introduction: In chapter - II, the probability of estimating the mean of the character under study with desired margin of error had been worked out for the selection procedure which makes use of auxiliary character. It has been observed that this probability increases with the increase of sample size, correlation coefficient and margin of error for the study variable. However, the sample mean is not unbiased for the population mean under the suggested procedure. It is therefore, important to suggest the estimation procedure which provides unbiased estimate of population mean along with unbiased variance estimator under the selection procedure of chapter - II. This has been attempted in this chapter. For obtaining Horwitz-Thompson's estimates of population mean and its variance the knowledge of  $w_i$ 's for all  $i$  and  $w_{ij}$ 's for all  $i \neq j$  is required, the calculation of which has also been discussed. It may be mentioned that some  $w_i$ 's or  $w_{ij}$ 's can be zero for certain population which imposes a restriction on the suggested estimation procedure. Therefore, the procedure for the selection of the sample has been modified ensuring non zero  $w_i$ 's and  $w_{ij}$ 's. Further the concept of controlled selection has also been used in modifying the procedure to obtain non zero  $w_i$ 's and  $w_{ij}$ 's. The relative efficiency of the suggested procedures has also been compared with a number of known procedures empirically.

3.2 Estimation Procedure: Let  $M$  be the number of samples out of  $N_{c_n}$  possible samples which satisfy the condition

$$|(\bar{x} - \mu_x) / \mu_x| \leq e_x \text{ of the procedure suggested in chapter - II.}$$

Then the probability of selecting a sample out of these  $M$  samples is given by  $P_s = 1/M$ . Under this scheme it is easy to see that

$$v_i = \sum_{s \in S_i} P_s = \frac{K_i}{M}$$

and  $v_{ij} = \sum_{s \in S_{ij}} P_s = \frac{K_{ij}}{M} \quad (i \neq j)$

where  $K_i$  is the number of samples out of  $M$  having  $i$ -th unit and  $K_{ij}$  is the number of samples containing both  $i$ -th and  $j$ -th units. Knowing the values of the units and their  $v_i$ 's and  $v_{ij}$ 's the Horwitz-Thompson estimate can be used for estimating the population mean. If  $v_i > 0$  for all  $i$ , we have

$$\hat{Y}_{HT} = \frac{1}{N} \sum_{i=1}^n \frac{y_i}{v_i}$$

The Yates - Grundy form of variance of above estimator is given by

$$V(\hat{Y}_{HT}) = \frac{1}{N^2} \sum_{i < j}^n \sum_{i < j}^n (v_i v_j - v_{ij}) \left( \frac{y_i}{v_i} - \frac{y_j}{v_j} \right)^2$$

Also if  $v_{ij} > 0$  for all  $i \neq j$ , the Yates-Grundy form of estimate of variance is given by

$$V(\hat{Y}_{HT}) = \frac{1}{N^2} \sum_{i < j}^n \sum_{i < j}^n \left( \frac{v_i v_j - v_{ij}}{v_{ij}} \right) \left( \frac{y_i}{v_i} - \frac{y_j}{v_j} \right)^2$$

If a selection procedure provides larger inclusion probability for the dissimilar units and if  $w_i$ 's are approximately of the same order, it is expected that the procedure would result in considerable gain in efficiency as compared to simple random sampling. In the suggested procedure the inclusion probabilities for dissimilar units are expected to be large ~~as compared~~ as compared to similar units in respect of auxiliary character which would result in gain in efficiency.

It has been explained for symmetrical distribution

3.2.1 Illustration: For the example considered in chapter - II, with  $n = 4$  and  $N = 6$  there are in all 5 samples out of 15 samples satisfying  $|(\bar{x} - \mu_x) / \mu_x| \leq c_x$ . The  $w_i$ 's and  $w_{ij}$ 's for different units are given by

$$w_1 = 4/5, w_2 = 3/5, w_3 = 3/5, w_4 = 3/5, w_5 = 3/5, w_6 = 4/5$$

$$w_{ij} = 1/5 \begin{bmatrix} & 2 & 3 & 4 & 5 & 6 \\ & 2 & 2 & 2 & 2 & 4 \\ & & 1 & 2 & 2 & 2 \\ & & & 2 & 2 & 2 \\ & & & & 1 & 2 \\ & & & & & 2 \end{bmatrix} \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix} \quad (i < j)$$

It may be remarked that if the distribution of  $x$  is symmetrical  $w_i$ 's will be more or less equal for all the units. Also the  $w_{ij}$  for the pair of units whose mean is closer to the population mean will be more than the other pair whose mean is different from the population mean, in respect of auxiliary character.

**3.3 Modified Procedure :** For the sampling procedure described in chapter - II,  $v_i$  or  $v_{ij}$  might be zero for some unit or pair of units which imposes a serious limitation on the use of Horwitz-Thompson estimates. To overcome this difficulty the procedure has been modified for the selection of the sample. The modified selection procedure consists of the following steps :

**Step - I :** Draw a simple random sample without replacement (SRSWOR) of size  $(n - 2)$  from the population of size  $N$  by the selection procedure given in chapter - II.

**Step - II :** Supplement the above sample by 2 units drawn by SRSWOR from the remaining  $(N - n + 2)$  units of the population.

Now if  $M'$  is the number of samples of size  $(n - 2)$  out of  $N C_{(n-2)}$  possible samples which satisfy  $|(\bar{x}' - \mu_x) / \mu_x| \leq \epsilon_x$ , where  $\bar{x}'$  is the mean based on  $(n - 2)$  units then  $v_i$ 's and  $v_{ij}$ 's will be given by

$$v_i = \frac{K_i'}{M'} + \left(1 - \frac{K_i'}{M'}\right) \frac{2}{N - n + 2} \quad \text{and}$$

$$v_{ij} = \frac{K_{ij}'}{M'} + \frac{K_i' + K_j' - 2K_{ij}'}{M'} \cdot \frac{2}{N - n + 2} + \left(1 - \frac{K_i' + K_j' - K_{ij}'}{M'}\right) \frac{2}{(N - n + 1)(N - n + 2)}$$

where  $K_i'$  is the number of samples out of  $M'$  which contain  $i$ -th unit and  $K_{ij}'$  is the number of samples containing both  $i$ -th and  $j$ -th units. Using these  $v_i$ 's and  $v_{ij}$ 's which are non zero, estimates of the population mean and its variance can be

obtained by the formulae given in section 3.2. In some cases ~~the~~ ~~under~~ this procedure may not hold good for  $|(X-\bar{X})/\mu_x| < \epsilon_x$ .

3.3.1 Illustration: Again for the example considered in chapter-II, the  $v_i$ 's and  $v_{ij}$ 's for  $n = 2$  are as under,

$$v_1 = 1/5, v_2 = 2/5, v_3 = 2/5, v_4 = 2/5, v_5 = 2/5, v_6 = 1/5$$

$$v_{ij} = 1/5 \begin{matrix} & \begin{matrix} i/j & 2 & 3 & 4 & 5 & 6 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix} & \begin{bmatrix} 0 & 0 & 0 & 0 & 1 \\ & 0 & 1 & 1 & 0 \\ & & 1 & 1 & 0 \\ & & & 1 & 0 \\ & & & & 0 \end{bmatrix} \end{matrix} \quad (i < j)$$

Also for the modified selection procedure for  $n = 4$ ,  $v_i$ 's and  $v_{ij}$ 's are given by

$$v_1 = 6/10, v_2 = 7/10, v_3 = 7/10, v_4 = 7/10, v_5 = 7/10, v_6 = 6/10$$

$$v_{ij} = 1/30 \begin{matrix} & \begin{matrix} i/j & 2 & 3 & 4 & 5 & 6 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix} & \begin{bmatrix} 11 & 11 & 11 & 11 & 10 \\ & 13 & 14 & 14 & 11 \\ & & 14 & 14 & 11 \\ & & & 13 & 11 \\ & & & & 11 \end{bmatrix} \end{matrix} \quad (i < j)$$



**3.4 Controlled Selection Procedure :** In the selection procedure suggested in chapter - II , a sample is selected if it satisfies  $|(\bar{x} - \mu_x) / \mu_x| \leq \epsilon_x$  otherwise it is rejected. This essentially implies that the sample space of SRSWOR is divided into two parts, the first of consisting of preferred samples for which  $|(\bar{x} - \mu_x) / \mu_x| \leq \epsilon_x$  and the second consisting of non preferred samples for which  $|(\bar{x} - \mu_x) / \mu_x| > \epsilon_x$  . The non preferred samples have a zero probability of selection under the procedure suggested in chapter-II which results sometimes in  $v_i = 0$  or  $v_{ij} = 0$  . The non-preferred samples can be assigned non-zero probability of selection ( however small ) which will provide non zero inclusion probabilities for all units and pairs of units. The controlled selection procedure for desired control say a consists of the following steps :

**Step I :** Draw a sample of size  $n$  by SRSWOR from the population of size  $N$ .

**Step II :** Test whether  $|(\bar{x} - \mu_x) / \mu_x| \leq \epsilon_x$  for given  $\epsilon_x$  .

**Step III :** (i) If  $|(\bar{x} - \mu_x) / \mu_x| \leq \epsilon_x$  . perform a Bernoulli trial with probability of success  $P_1$  . for selecting the sample and  
(ii) If  $|(\bar{x} - \mu_x) / \mu_x| > \epsilon_x$  perform a Bernoulli trial with probability of success  $P_2$  for selecting the sample, <sup>otherwise go to Step I.</sup> The values of  $P_1$  and  $P_2$  are determined from the equations,

$P_1 = c/M$  and  $P_2 = (1 - c) / ({}^N C_n - M)$ . Generally  $P_2$  is assigned a very small value and  $P_1$  is determined by the relation  $M P_1 + ({}^N C_n - M) P_2 = 1$ .

Under the scheme of controlled selection  $v_i$ 's and  $v_{ij}$ 's can be obtained by the formulae

$$v_i = P_1 K_i + P_2 ({}^{N-1} C_{n-1} - K_i) \quad \text{and}$$

$$v_{ij} = P_1 K_{ij} + P_2 ({}^{N-2} C_{n-2} - K_{ij})$$

Knowing  $v_i$ 's and  $v_{ij}$ 's for all individual units and pair of units, Horwitz-Thompson estimates of mean and corresponding variance can be obtained by the formulae given in section 3.2.

**3.4.1 Illustration:** For the example considered in chapter-II

for  $N = 6$ ,  $n = 4$ , and prefixing  $P_2 = .001$  we get

$P_1 = .198$ . Thus  $v_i$ 's and  $v_{ij}$ 's are given by

$$v_1 = .798, v_2 = .601, v_3 = .601, v_4 = .601, v_5 = .601, v_6 = .798.$$

$$v_{ij} = \begin{matrix} & \begin{matrix} 2 & 3 & 4 & 5 & 6 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix} & \begin{bmatrix} .400 & .400 & .400 & .400 & .794 \\ & .203 & .400 & .400 & .400 \\ & & .400 & .400 & .400 \\ & & & .203 & .400 \\ & & & & .400 \end{bmatrix} \end{matrix} \quad \left. \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix} \right\} (i < j)$$

**3.5 Efficiency:** In an attempt to compare the suggested procedure with some existing procedures five hypothetical populations have been considered. The populations have correlation coefficient ranging from .64 to .95 and are presented in table 3.1.

**Table 3.1 Table showing five hypothetical populations with different values of  $\rho$ .** .....

Unit No.	Population									
	$\rho = .9869$		.8703		.8451		.7441		.6441	
	I		II		III		IV		V	
	X	Y	X	Y	X	Y	X	Y	X	Y
1	1	1	1	1	1	1	1	1	1	1
2	2	1	2	1	2	2	2	1	2	2
3	3	2	2	1	3	2	2	1	3	3
4	4	2	3	2	4	2	3	1	4	3
5	5	3	3	2	5	2	3	2	5	3
6	6	3	4	2	6	3	4	2	6	3

A natural population presented in table 3.2 has also been considered for comparison. It has been taken from 'Annals of Mathematical Statistics', 13, 179-206 and is based on estimation of volume of timber stands by strip sampling.  $x$  is here in 10 chain units and represents blockwise sum of strip lengths while  $y$  is in 1000 ft. board measurement units and represents blockwise sum of timber b volumes.

**Table 3.2 Table showing natural population ( $\rho = .917$ )** .....

Unit No.	X	Y	Unit No.	X	Y
1	180	8781	6	140	5622
2	142	6756	7	81	3677
3	76	1786	8	91	4368
4	104	2425	9	191	7919
5	109	4655	10	117	3861

For each population the variance for the estimate based on sample size 4 have been worked out for various procedures and results are presented in Table 3.3. The procedures considered for comparison are the following :

1. Horvitz-Thompson estimate in the suggested procedure ( $e_x = .05$ )
2. Horvitz-Thompson estimate in modified procedure ( $e_x = .05$ )
3. HT estimate in controlled selection procedure ( $e_x = .05$ ,  $P_2 = .001$ )
4. Simple mean in SRSWOR
5. Usual estimate in PPSWR
6. Regression estimate in SRSWOR

Table 3.3 Table showing the variances of different procedures

Proce- dure	POPULATION					
	I	II	III	IV	V	Natural
(1)	.0000	.0069	.0000	.0069	.0417	274316.828
(2)	.0556	.0007	.0278	.0003	.0556	1364083.647
(3)	.0007	.0037	.0003	.0074	.0194	329950.150
(4)	.0667	.0250	.0333	.0231	.0583	780200.000
(5)	.0400	.0278	.1130	.0440	.1220	184118.750
(6)	.0057	.0061	.0095	.0103	.0341	124100.000

Nothing should be concluded with certainty from these results as these are empirical comparisons only. However, it can be seen that H. T. estimate under the suggested procedure of chapter-II provides generally more efficient results than usual estimates of SRSWOR and PPSWR sampling procedures. The suggested procedures are also competitive with regression estimate in many situations.

**3.6 Summary :** In this chapter , a estimation procedure has been suggested which provides unbiased estimate of population mean along with unbiased variance estimator, under the selection procedure of chapter-II. But under this procedure  $v_i$  or  $v_{ij}$  may be zero for some  $i$  or  $i \neq j$  , therefore the procedure has been modified to provide non zero probabilities of inclusion for all the units and pair of units. Further the concept of controlled selection has been used to provide non zero probabilities of inclusion. All these procedures have been compared with some of the existing procedures empirically and the results have been found to be highly satisfactory.

## S U M M A R Y

It is well known that ancillary information helps in obtaining efficient results in estimation of population parameter as mean etc. A number of procedures based on ancillary information have been developed but none provides control on sampling error which is an very important aspect in sampling. In this dissertation, a selection procedure has been suggested which makes use of ancillary information to ensure the selection of a representative sample i. e. , a sample for which relative margin of error of sample estimate from population parameter is within specified margin. Further, the probability that sample estimate for study character will also differ from population parameter by a specified margin of error has been calculated, for different ranges of sample size, correlation coefficient between study and auxiliary variables, margin of errors and coefficients of variation. This probability is generally of high order and thus the suggested procedure can be used satisfactorily in most of practical situations.

To obtain Horvitz-Thompson's estimate of mean and variance, method of calculating  $v_i$ 's and  $w_{ij}$ 's for the suggested procedure has also been given. In case  $v_i = 0$  or  $w_{ij} = 0$  for some unit or pair of units, the selection procedure and method of computing  $v_i$ 's and  $w_{ij}$ 's for all the units have been modified.

The controlled selection procedure as modification of the suggested procedure has been given and the method of obtaining  $v_i$ 's and  $v_{ij}$ 's based on controlled selection has also been discussed. Some theoretical and natural populations have also been considered to compare the suggested schemes with existing procedures empirically. The suggested schemes have been found to be highly satisfactory in all the situations.

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