# ON THE ALLOCATION OF SAMPLE SIZE IN STRATIFIED SAMPLING

T. J. RAO

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## Summary

The problem of optimum allocation of sample size to strata is examined in the light of a priori distributions. In this context, we discuss with illustrations the justification for the assumption that the unknown proportionate values of  $\sigma_i^2$ 's can be replaced by the proportionate values of the known  $\alpha_i^2$ 's, which are estimates of  $\sigma_i^2$ 's. The sample size is allocated so as to minimize the expected variance of the strategy consisting of  $\pi PS$  sampling scheme and the Horvitz-Thompson estimator under a general super-population model. It is further shown that, in the sense of expected variance,  $\pi PS$  sampling for unstratified sampling is inferior to  $\pi PS$  stratified sampling with this type of allocation, unless the super-population parameter g attains the value 2, in which case both schemes are equivalent.

#### 1. Introduction

In the case of Neyman's optimum allocation ([12]) of sample size to strata, we have the allocation given by  $n_{i,\text{opt.}} = nN_i\sigma_i/\sum_{i=1}^{n}N_i\sigma_i$ , where  $N_i$  is the size of the *i*th stratum ( $i=1,2,\cdots,h$ ; n is the total sample size and  $\sigma_i^2$  is the within-variance for the *i*th stratum). Computation of  $n_{i,\text{opt.}}$ 's requires at least the proportionate values of  $\sigma_i^2$ 's, which are unknown. In practice, some estimates  $\alpha_i^2$ 's of  $\sigma_i^2$ 's are substituted. These estimates, usually, are the  $\sigma_i^2$ 's of some auxiliary information closely related to the characteristic under study. In most of the cases, the values of the same characteristic studied in a previous occasion are treated as the auxiliary information. The justification for the assumption that the unknown proportionate values of  $\sigma_i^2$ 's are usually not quite different from the proportionate values of the known  $\sigma_i^2$ 's is examined in the light of an  $\sigma_i^2$  distribution by Hanurav [5] and here it is further developed.

160 T. J. RAO

# 2. Optimum allocation of sample size and prior distributions

Let  $\Delta_g$  be the class of all prior distributions  $\partial$  for which we have

i) 
$$\mathcal{E}_{\partial}(Y_i | X_i) = aX_i$$
,

ii) 
$$CV_{\vartheta}(Y_i | X_i) = \sigma^2 X_i^{\sigma}$$
,  $1 \leq g \leq 2$ ,

iii) Cov<sub>$$\vartheta$$</sub> $(Y_i, Y_j | X_i, X_j) = 0$ .

Then,

$$\sigma_i^2 = \frac{1}{N_i} \left[ \sum_{j=1}^{N_i} Y_{ij}^2 - \frac{Y_i^2}{N_i} \right]$$

where  $Y_{ij}$  is the value of the characteristic Q for the *j*th unit of the *i*th stratum  $(j=1, 2, \dots, N_i; i=1, 2, \dots, h)$ . We then have

$$\begin{split} \mathcal{E}_{\delta}(\sigma_{i}^{2}) &= \frac{1}{N_{i}} \left[ \sum_{j} \left( \alpha^{2} X_{ij}^{2} + \sigma^{2} X_{ij}^{g} \right) - \frac{1}{N_{i}} \left( \alpha^{2} X_{i}^{2} + \sigma^{2} \sum_{j} X_{ij}^{g} \right) \right] \\ &= \frac{1}{N_{i}} \left[ \alpha^{2} \left( \sum_{j} X_{ij}^{2} - \frac{X_{i}^{2}}{N_{i}} \right) + \sigma^{2} \left( 1 - \frac{1}{N_{i}} \right) \sum_{j} X_{ij}^{g} \right] \\ &= \alpha^{2} \alpha_{i}^{2} + \sigma^{2} \frac{N_{i} - 1}{N_{i}} \left( \sum_{j} X_{ij}^{g} / N_{i} \right) \\ &\cong \alpha^{2} \alpha_{i}^{2} + \sigma^{2} \sum_{j} X_{ij}^{g} / N_{i} , \end{split}$$

assuming  $\frac{N_i-1}{N_i}$ 's to be nearly equal to 1,

$$= \alpha^{2} \alpha_{i}^{2} + \sigma^{2} \left[ \left( \sum_{j} X_{ij}^{g} / N_{i} \right) - \left( \sum_{j} X_{ij}^{2} / N_{i} \right) + \alpha_{i}^{2} + \bar{X}_{i}^{2} \right]$$

$$= (\alpha^{2} + \sigma^{2}) \alpha_{i}^{2} + \sigma^{2} \left[ \bar{X}_{i}^{2} + \frac{1}{N_{i}} \left( \sum_{j} X_{ij}^{g} - \sum_{j} X_{ij}^{2} \right) \right].$$

Thus  $\sigma_i^2$ 's can be expected to be in the same proportion as  $\alpha_i^2$ 's if  $\alpha_i^2$ 's are proportional to  $\bar{X}_i^2 + \frac{1}{N_i} (\sum_j X_{ij}^g - \sum_j X_{ij}^2)$  i.e., if the squares of the corrected coefficients of variation (c.c.v.) of  $\mathcal{X}$ -character, defined by  $\frac{\alpha_i^2}{\bar{X}_i^2 - \partial_i/N_i^2}$ , where  $\partial_i = N_i (\sum_j X_{ij}^2 - \sum_j X_{ij}^g)$  are equal in all the strata and when this condition is satisfied the allocation is given by

$$egin{aligned} n_i = & n rac{N_i \sqrt{ar{X}_i^2 - rac{{ar{\partial}_i}}{N_i^2}}}{\sum\limits_{i = 1}^h N_i \sqrt{ar{X}^2 - rac{{ar{\partial}_i}}{N_i^2}}} \ = & n rac{\sqrt{T_i^2 - {ar{\partial}_i}}}{\sum\limits_{i = 1}^h \sqrt{T_i^2 - {ar{\partial}_i}}} \;\; , \qquad ext{where} \quad T_i = \sum\limits_{j} X_{ij} \;. \end{aligned}$$

Thus we have

THEOREM 1. Under the super-population model  $\Delta_g$ , Neyman's optimum allocation reduces to allocation proportional to  $\sqrt{T_i^2-\partial_i}$ , where  $T_i=\sum\limits_{j=1}^{N_i}X_{ij}$  is the total of the auxiliary variate for the ith stratum and  $\partial_i=N_i(\sum\limits_{j=1}^{N_i}X_{ij}^2-\sum\limits_{j=1}^{N_i}X_{ij}^g)$ , when the corrected coefficients of variation of the  $\mathcal{X}$ -character are equal in all the strata.

From the above theorem follows immediately

THEOREM 2. (See [5].) Neyman's optimum allocation reduces to allocation proportional to the stratum totals of the auxiliary variate  $\mathcal{X}$ , under  $\mathcal{L}_2$ , when the coefficients of variation of  $\mathcal{X}$ -character are equal in all the strata.

PROOF. Under  $\Delta_2$ , we have  $\partial_i = 0$ ,

hence,

$$n=nT_i/\sum_{i=1}^n T_i=nT_i/T$$
 etc.

## 3. Illustrations and remarks

Based on theorem 2, optimum allocation can be approximated by allocation proportional to the stratum totals under a particular model ( $\mathcal{A}_2$ ) when the condition that the within-strata coefficients of variation are more or less equal is satisfied. Especially, when stratification is based on geographical contiguity or administrative and operational considerations this condition is mostly satisfied. Mahalanobis [10] proposed equalization of the stratum totals along with equal allocation as an approximation to optimum allocation. Hansen, Hurwitz and Madow [7] have demonstrated the use of this method for slight adjustments of strata sizes. Kitagawa [9] analyzed further and justified the principle of equipartition. Considering the allocation of a sample of 1000 saw mills, Hansen, Hurwitz and Madow [7] show that allocation proportional to stratum totals

serves as an approximation to optimum allocation. Murthy [11] considered live data on 128 villages of an Indian tehsil and Table 1 shows the results and allocation.

Table 1

No of strata-2

Method of stratification	Relative variance for allocation proportional to		
Wethod of Stratification	T <sub>i</sub>	Nioi	
I	.005266	.005005	
II	.005052	.004773	
III	.004878	.004702	

Next, in allocating a sample of 1000 farms in the province of Söedermanland, Sweden, Dalenius [2] considered various types of allocations for four acreage groups. Allocation proportional to total acreage (termed 'value-allocation') is found to offer a good compromise to minimum variance allocation. Hansen, Hurwitz and Madow [7] discuss this value-allocation and minimum variance allocations based on data for various crops. In fact it is interesting to observe that the minimum variance allocation for autumn wheat and the value-allocation are as close as (260, 278, 311, 151) and (268, 263, 313, 156).

In most of the situations one assumes the model  $\Delta_2$ , but there are populations for which we have to consider a more general model  $\Delta_q$  and for completeness and general academic interest, theorem 1 is proved which is an extension of the above case and when one has an idea of the value of g then an allocation as suggested therein would be more apt. We illustrate this considering data on the 1920 inhabitants of 64 large cities in the United States, grouped into 2 strata (Cochran [1]). Table 2 gives the results when g=1.75.

Table 2

Stratum	Stratum size	Stratum totals $T_i = \sum X_{ij}$	$\sum\limits_{j}X_{ij}^{2}$	
1	16	8349	4,756,619	
2	48	7941	1,474,871	
			Allocation prop	

i		$\alpha_i$	$\partial_i$	$\sqrt{T_{i^2}-\partial_i}$	Allocation proportional to			
					αι	$T_i$	$\sqrt{T_{i^2}-\partial_i}$	c.c.v.
	1	173.8	60,655,664	3008.4	.5n	.51n	.47n	.93
	2	57.9	51,822,528	3352.1	.5 <i>n</i>	.49n	.53n	.84

# 4. Stratified $\pi PS$ sampling

Consider a population consisting of h strata with  $N_i$  as the ith stratum,  $i=1,2,\cdots,h$ . Let  $Y_{ij}$  and  $X_{ij}$  be the values of the  ${}^{Q}$  characteristic and the  ${}^{Q}$  characteristic (auxiliary information) respectively, for the jth unit of the ith stratum. Let a  $\pi PS$  ( $\pi_i$ , the probability of inclusion of the ith unit, being proportional to size) sample (see Hanurav [6]) of size  $n_i$  be taken from the ith stratum, such that  $\sum_{i=1}^{n} n_i = n$ . Let  $\pi_{ij}$  be the probability of inclusion of the jth unit in the ith stratum in the sample. As an estimator of the population total consider the Horvitz-Thompson estimator (Horvitz and Thompson [8])

(1) 
$$\hat{Y} = \sum_{i=1}^{h} \sum_{j=1}^{n_i} y_{ij} / \pi_{ij}.$$

THEOREM 3. Under the super-population model  $\Delta_q$ , allocation of the sample size to the strata which minimizes the expected variance of (1) is given by

$$n_i = rac{n\sqrt{T_i \sum\limits_{j=1}^{N_i} X_{ij}^{g-1}}}{\sum\limits_{i=1}^{h} \sqrt{T_i \sum\limits_{j=1}^{N_i} X_{ij}^{g-1}}}$$

where  $T_i$  is the total of X-values for the ith stratum.

PROOF. We have

(2) 
$$\mathcal{E}_{d_g}V(\hat{Y}) = \sum_{i=1}^h \sum_{j=1}^{N_i} \left(\frac{1}{\pi_{ij}} - 1\right) \sigma_{ij}^2,$$

where  $\sigma_{ij}^2 = \sigma^2 X_{ij}^g$  (see Godambe [3], Hanurav [4]). We now, minimize (2) subject to the condition  $\sum_{i=1}^{n} n_i = n$ . Introducing the Lagrange multiplier  $\lambda$ , consider

$$\textstyle \mathcal{E}_{{\scriptscriptstyle A_g}} V(\hat{Y}) + \lambda \Big( \sum\limits_{i=1}^{h} n_i - n \Big) = \sum\limits_{i=1}^{h} \sum\limits_{j=1}^{N_t} \Big( \frac{1}{n_i P_{i,i}} - 1 \Big) \sigma_{ij}^2 + \lambda \Big( \sum\limits_{i=1}^{h} n_i - n \Big) \text{ ,}$$

where

$$P_{ij} = X_{ij}/T_i$$
 .

Differentiating with respect to  $n_i$  and equating to zero, we have

$$-\frac{1}{n_i^2}\sum_{j=1}^{N_i}\frac{\sigma_{ij}^2}{P_{ij}}+\lambda=0$$
 ,

which gives

$$n_i = \frac{1}{\sqrt{\lambda}} \sqrt{\sum_{j=1}^{N_i} \frac{\sigma_{ij}^2}{P_{ij}}}$$

and thus

$$\begin{split} n_i &= n \sqrt{\sum_{j=1}^{N_i} \frac{\sigma_{ij}^2}{P_{ij}}} / \sum_{i=1}^h \sqrt{\sum_{j=1}^{N_i} \frac{\sigma_{ij}^2}{P_{ij}}} \\ &= n \sqrt{T_i \sum_{j=1}^{N_i} X_{ij}^{g-1}} / \sum_{i=1}^h \sqrt{T_i \sum_{j=1}^{N_i} X_{ij}^{g-1}} , \end{split}$$

hence, we have the theorem.

DEFINITION 1. The allocation of sample size to the strata given above is called  $\Delta_g$ -optimum allocation.

THEOREM 4. Under the super-population model  $\Delta_2$ , allocation of the sample size to the strata proportional to the stratum totals of the X-variate minimizes the expected variance of (1).

PROOF. The proof follows directly from theorem 3, by putting g=2.

Let  $n_i$  be the  $\Delta_2$ -optimum sample size for the *i*th stratum and select the sample by a  $\pi PS$  sampling scheme within each stratum. Then as an estimator of the population total consider the estimator (1)

$$\hat{Y} = \sum_{i=1}^{h} \sum_{j=1}^{n_i} \frac{y_{ij}}{\pi_{ij}} = \sum_{i=1}^{h} \sum_{j=1}^{n_i} \frac{y_{ij}}{(n_i x_{ij} | T_i)}$$
.

We now have the following theorem.

Theorem 5. In the sense of expected variance, under  $\Delta_g$ ,  $\pi PS$  unstratified sampling is inferior to  $\pi PS$  stratified sampling with  $\Delta_g$ -optimum allocation and for g=2, both the schemes are equivalent.

Proof. We have

$$\mathcal{C}_{\delta}V(\hat{Y}_{\mathrm{st.}})\!=\!\sum\limits_{i=1}^{h}\sum\limits_{j=1}^{N_{i}}{[\pi_{ij}^{-1}\!-\!1]\sigma^{2}X_{ij}^{g}}$$
 ,

where

$$\pi_{ij} = \frac{n\mu_i X_{ij}}{X_i}$$
 and  $\mu_i = \frac{\sqrt{X_i \sum_j X_{ij}^{g-1}}}{\sum\limits_{i=1}^h \sqrt{X_i \sum_j X_{ij}^{g-1}}}$ ,

and

$$V(\hat{Y}_{ ext{unst.}}) = \sum\limits_{i=1}^h \sum\limits_{j=1}^{N_i} [\pi'_{ij}^{-1} - 1] \sigma^2 X_{ij}^g$$
 ,

where

$$\pi'_{ij} = nX_{ij}/X$$
.

Thus,

$$\begin{split} \mathcal{C}_{\delta} \bigg[ \frac{V(\hat{Y}_{\text{unst.}}) - V(\hat{Y}_{\text{st.}})}{\sigma^{2}} \bigg] &= \sum_{i=1}^{h} \sum_{j=1}^{N_{i}} \bigg[ \frac{X_{ij}^{g}}{\pi_{ij}^{f}} - \frac{X_{ij}^{g}}{\pi_{ij}} \bigg] \\ &= \sum_{i} \sum_{j} \bigg[ \frac{X_{ij}^{g}}{(nX_{ij}/X)} - \frac{X_{ij}^{g}}{(n\mu_{i}X_{ij}/X_{i})} \bigg] \\ &= \frac{X}{n} \sum_{i} \sum_{j} X_{ij}^{g-1} - \sum_{i} \frac{X_{i}}{n\mu_{i}} \sum_{j} X_{ij}^{g-1} \\ &= \frac{X}{n} \sum_{i} \sum_{j} X_{ij}^{g-1} - \bigg( \sum_{i} \sqrt{X_{i}} \sum_{j} \overline{X_{ij}^{g-1}} \bigg)^{2} \bigg/ n \\ &= \frac{1}{n} \bigg[ \sum_{i=1}^{h} X_{i} \sum_{i=1}^{h} \bigg( \sum_{j=1}^{N_{i}} X_{ij}^{g-1} \bigg) - \bigg( \sum_{i=1}^{h} \sqrt{X_{i}} \sqrt{\sum_{j=1}^{N_{i}} X_{ij}^{g-1}} \bigg)^{2} \bigg] \\ &\geq 0 \end{split},$$

by Cauchy-Shwartz's inequality.

It is easily seen that when g=2, equality occurs. Hence we have the theorem.

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INDIAN STATISTICAL INSTITUTE, CALCUTTA

#### REFERENCES

- [1] W. G. Cochran, Sampling Techniques, John Wiley, New York, 1953.
- [2] T. Dalenius, "Anvaendmingen av Stickprovsmetodik vid 1950 Ars Arealinventering och Kreatursraeknig," *Jordbruksekonomiska Meddelanden*, (1951), 59-75.
- [3] V. P. Godambe, "A unified theory of sampling from finite populations," Jour. Roy. Stat. Soc., 17B (1955), 269-278.
- [4] T. V. Hanurav, "On Horvitz and Thompson's estimator," Sankhyā, A 24 (1962), 429-436.
- [5] T. V. Hanurav, "Optimum Sampling Strategies and Related Problems," Ph. D. thesis, Indian Statistical Institute, 1965.
- [6] T. V. Hanurav, "Optimum utilization of auxiliary information: πPS sampling of two units from a stratum," J. R. Statist. Soc., 29B (1967), 374-391.

- [7] M. H. Hansen, W. N. Hurwitz and W. G. Madow, Sample Survey Methods and Theory, John Wiley, New York, 1953.
- [8] D. G. Horvitz and D. J. Thompson, "A generalization of sampling without replacement from a finite universe," J. Amer. Statist. Ass., 47 (1952), 663-685.
- [9] T. Kitagawa, "Some contributions to the design of sample surveys," Sankhyā, 17 (1956), 1-36.
- [10] P. C. Mahalanobis, "Some aspects of the design of sample surveys," Sankhyā, 12 (1952), 1-7.
- [11] M. N. Murthy, Sampling Theory and Methods, Statistical Publishing Society, Calcutta, 1967.
- [12] J. Neyman, "On two different aspects of the representative method: the method of stratified sampling and the method of purposive selection," J. R. Statist. Soc. 97 (1934), 558-606.