SAMPLE ALLOCATION AND SELECTION METHODS FOR OVERSAMPLING SUBPOPULATIONS

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1 Introduction

Generally, in sample surveys the estimates for certain small subpopulations whose members cannot be identified in advance of sampling are not precise because the number of sampled units belonging to these subpopulations is small. Often, there is a need to improve the precision of these estimates or sometimes there is a requirement for a predetermined expected number of units belonging to a subpopulation in the overall sample for purposes of data analysis. Several techniques are available for increasing the expected sample size belonging to a subpopulation or subpopulations. For example, a simple but expensive method of achieving an expected sample size for a subpopulation is to increase the size of the overall sample. Another method is to stratify the population according to the density of the subpopulation and then use nonproportional allocation. A third method is to use a two-phase design in which a screening interview is conducted on a large sample in the first phase to identify the members of the subpopulation and retain them in the sample, and then select a subsample of the sample from the group not belonging to this subpopulation. Strategies for improving subpopulation estimates with differential sampling rates when there are two strata in the population have been given by Waksberg (1973) and Weller, Huggins and Singh (1991). A discussion of some of the techniques for increasing the sample size in the subpopulations can be found in Massey, Judkins and Waksberg (1993).

Some of the techniques used to achieve the desired sample size from the subpopulations might make the overall estimates very inefficient due to allocation or selection methods that are very different from the optimum methods needed to obtain precise overall estimates. Therefore, there is a need to balance the need for improving the precision of the subpopulation estimates with the loss in efficiency of the estimates for the general population.

In this paper, some methods of sample allocation are proposed which attempt to keep the strata sample sizes close to the sample sizes which are considered optimum from the point of view of efficiency of the overall estimates. The stratification boundaries are the same as those created for maximizing the efficiency of the overall estimates. A method of revising the probabilities of selection of primary sampling units which maximizes the expected proportion of sampled units belonging to a subpopulation in the overall sample subject to certain constraints is also suggested.

2. Sample Allocation Methods

2.1 Allocation with an increase in sample size

Let a population of N units be stratified into L strata. Let Nh be the number of units in the population in the hth stratum. Let n be the number of units in the sample that is required to obtain an overall estimate of some characteristic of interest with a prespecified precision.

Let Nhfh be the number of units in the population in the hth stratum belonging to the subpopulation or domain of interest. Then, Nhfh/Nh is the proportion of the population in the hth stratum which belongs to the subpopulation or domain. Let nh be the sample number of units allocated to the hth stratum using proportional, optimum, Neyman or some other allocation for purposes of estimating the overall population parameters like totals, means, ratios or proportions efficiently. The expected number of units in the sample belonging to the domain is given by

\[ E(n_d) = \sum_{h=1}^{L} E(n_{dh}) \]

The expected number of units belonging to the domain in the overall sample is given by

\[ E(n_d) = \sum_{h=1}^{L} E(n_{dh}) \]
Suppose $E(n_d)$ resulting from an allocation like proportional or Neyman of the overall sample is too small for obtaining reasonably precise estimates for the domain of interest and there is a requirement that this be number be $n^*$. To meet this requirement, the sample size in each stratum may have to be increased resulting in an increase in the overall sample size. Since the proportions of the population belonging to the domain of interest could be very different in different strata, there are several ways of allocating the sample to different strata to get the required domain sample size. Allocations that strictly minimize the overall increase in sample size may make the overall estimates more imprecise than necessary.

One criterion in determining the new allocation is to make the differences between the new sample allocation and the old allocation as small as possible and at the same time achieve the desired expected sample size for the domain. Let $n^*_h$ be the number of units that are required to be selected from the $h$th stratum.

We want $n^*_h$ to be such that

$$
\sum_{h=1}^{L} (n^*_h - n_h)^2
$$

is a minimum subject to the constraint

$$
\sum_{h=1}^{L} n_h N_h \frac{N_{dh}}{N_h} = n_d^* .
$$

The allocation which satisfies the above criterion is as follows.

$$
n^*_h = n_h + [n^*_d - E(n_d)] \sum_{h=1}^{L} P_{dh} \frac{N_{dh}}{N_h}
$$

where $P_{dh} = \frac{N_{dh}}{N_h}$.

2.2 Example

We use a slightly modified example from Cochran (1977) to illustrate the allocation method. The data are derived from a stratified sample of tire dealers. The dealers are assigned to strata according to the number of new tires held. The population means $m_h$ and variances $v_h$ of the number of new tires held are shown. If we are interested in estimating the overall average number of new tires held, then using Neyman allocation, the number of tire dealers to be sampled ($n_h$) from each stratum is shown. Suppose we are interested in tire dealers who belong to a specific domain, for example, those who hold a specific brand of tire and we want to estimate the total number of tires belonging to this brand. In this case, there may be a requirement for a certain number of sampled units belonging to this domain. The proportion of dealers belonging to the subpopulation or domain $P_{dh}$ in each stratum is also shown. The stratum boundaries are denoted by $B_h$.

<table>
<thead>
<tr>
<th>$S$</th>
<th>$B_h$</th>
<th>$N_h$</th>
<th>$P_{dh}$</th>
<th>$m_h$</th>
<th>$v_h$</th>
<th>$n_h$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>1-9</td>
<td>19,850</td>
<td>0.05</td>
<td>4.1</td>
<td>34.8</td>
<td>2832</td>
</tr>
<tr>
<td>2</td>
<td>10-19</td>
<td>3,250</td>
<td>0.10</td>
<td>13.0</td>
<td>92.2</td>
<td>755</td>
</tr>
<tr>
<td>3</td>
<td>20-29</td>
<td>1,007</td>
<td>0.15</td>
<td>25.0</td>
<td>174.2</td>
<td>321</td>
</tr>
<tr>
<td>4</td>
<td>30-39</td>
<td>606</td>
<td>0.20</td>
<td>38.2</td>
<td>320.4</td>
<td>262</td>
</tr>
<tr>
<td>T</td>
<td></td>
<td>24,713</td>
<td>0.064</td>
<td></td>
<td></td>
<td>4170</td>
</tr>
</tbody>
</table>

Using $n_h$ and $P_{dh}$ given above, we see that $E(n_d) = 318$. Suppose $n_d^* = 432$. We want the domain expected sample size to be 432 instead of 318. The new sample allocation and the expected domain sample size are shown below along with old sample size and the expected domain sample size.

<table>
<thead>
<tr>
<th>$S$</th>
<th>$n_h$</th>
<th>$E(n_{ah})$</th>
<th>$n_h^*$</th>
<th>$n_{dh}^*$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>2832</td>
<td>142</td>
<td>2908</td>
<td>145</td>
</tr>
<tr>
<td>2</td>
<td>755</td>
<td>76</td>
<td>907</td>
<td>91</td>
</tr>
<tr>
<td>3</td>
<td>321</td>
<td>48</td>
<td>549</td>
<td>83</td>
</tr>
<tr>
<td>4</td>
<td>262</td>
<td>52</td>
<td>566</td>
<td>113</td>
</tr>
<tr>
<td>T</td>
<td>4170</td>
<td>318</td>
<td>4930</td>
<td>432</td>
</tr>
</tbody>
</table>

The increase in sample size is 760 units. If we were drawing a simple random sample, then we would require a sample of 6750 units to get an expected domain sample size of 432.
To get an idea of the loss in efficiency due to this allocation, we compare $n_h^*$ with $n_{opt}$ using Neyman allocation. The two are shown below.

<table>
<thead>
<tr>
<th>Stratum</th>
<th>$n_h^*$</th>
<th>$n_{opt}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>2908</td>
<td>3347</td>
</tr>
<tr>
<td>2</td>
<td>907</td>
<td>892</td>
</tr>
<tr>
<td>3</td>
<td>549</td>
<td>380</td>
</tr>
<tr>
<td>4</td>
<td>566</td>
<td>311</td>
</tr>
<tr>
<td>Total</td>
<td>4930</td>
<td>4930</td>
</tr>
</tbody>
</table>

Comparing the variances of the sample mean $\bar{Y}_s$, for the two allocations, we see that the loss in efficiency due to using $n_h^*$ instead of $n_{opt}$ is around 6%. Of course, using $n_{opt}$ will not meet the requirement of getting a sample 432 belonging to the domain. The expected sample size will be 375.

The new sample allocation depends on the initial allocation. If the initial allocation has been different as shown in the following table, then the new allocation to meet the sample requirements follows closely the initial allocation. Both the allocations are shown below.

<table>
<thead>
<tr>
<th>Stratum</th>
<th>$n_h$</th>
<th>$\bar{Y}(n_{dh})$</th>
<th>$n_h^*$</th>
<th>$n_{dh}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>3000</td>
<td>150</td>
<td>3083</td>
<td>154</td>
</tr>
<tr>
<td>2</td>
<td>600</td>
<td>60</td>
<td>767</td>
<td>77</td>
</tr>
<tr>
<td>3</td>
<td>340</td>
<td>51</td>
<td>590</td>
<td>88</td>
</tr>
<tr>
<td>4</td>
<td>230</td>
<td>46</td>
<td>563</td>
<td>113</td>
</tr>
<tr>
<td>Total</td>
<td>4170</td>
<td>307</td>
<td>5003</td>
<td>432</td>
</tr>
</tbody>
</table>

2.3 Allocation with a fixed sample size

Sometimes it is possible to achieve the domain sample size without an increase in the overall sample size, if this is important because of a fixed budget. In this case, as before we want to minimize the quantity

$$\sum_{h=1}^{L} (n_h^* - n_h)^2$$

but subject to two constraints which are

$$\sum_{h=1}^{L} n_h^* \frac{N_{dh}}{N_h} = n_d^*$$

and

$$\sum_{h=1}^{L} n_h^* = n$$

The allocation that satisfies these conditions is $n_h^*$

$$n_h^* = n_h + \frac{[n_d^* - E(n_d)]}{(P_{dh} - \frac{\sum_{h=1}^{L} P_{dh}^2}{L})}$$

$$\frac{\sum_{h=1}^{L} P_{dh}^2}{L}$$

Turning to the previous example, and using the above expression and values of $n_h$ as those that were obtained under Neyman allocation we get the new allocation as follows.

<table>
<thead>
<tr>
<th>Stratum</th>
<th>$N_k$</th>
<th>$n_k$ Neyman</th>
<th>$n_h^*$</th>
<th>$n_h^*$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>19,850</td>
<td>2832</td>
<td>2148</td>
<td>1893</td>
</tr>
<tr>
<td>2</td>
<td>3,250</td>
<td>755</td>
<td>527</td>
<td>688</td>
</tr>
<tr>
<td>3</td>
<td>1,007</td>
<td>321</td>
<td>549</td>
<td>983</td>
</tr>
<tr>
<td>4</td>
<td>606</td>
<td>262</td>
<td>946</td>
<td>606</td>
</tr>
<tr>
<td>T</td>
<td>24,713</td>
<td>4170</td>
<td>4170</td>
<td>4170</td>
</tr>
</tbody>
</table>
The initial allocation yields a sample of 946 in stratum 4, but the population in this stratum is only 606. Therefore, all the units in stratum 4 are included in the sample. The remaining number of units to be sampled which is 3564 is reallocated using the same formula given, but now applied to three strata, which results in the final allocation. The values of $n_d^*$ resulting from the new allocation are given below.

<table>
<thead>
<tr>
<th>St.</th>
<th>$n_h^*$</th>
<th>$n_{dh}^*$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>1893</td>
<td>95</td>
</tr>
<tr>
<td>2</td>
<td>688</td>
<td>69</td>
</tr>
<tr>
<td>3</td>
<td>983</td>
<td>147</td>
</tr>
<tr>
<td>4</td>
<td>606</td>
<td>121</td>
</tr>
<tr>
<td>Total</td>
<td>4170</td>
<td>432</td>
</tr>
</tbody>
</table>

Generally, the initial overall sample size is not large enough to provide a sample of the desired size belonging to a domain. In such cases, the overall sample will have to be necessarily increased.

2.4 Allocation for two domains

Suppose there are two domains for which we need prespecified expected number of units in the overall sample. Then, one method of meeting this requirement is to do the allocation sequentially. First the overall sample is allocated for maximizing the precision of the overall estimate. Then this allocation is changed to accommodate the first requirement. The resulting allocation is again changed to satisfy the second requirement.

This procedure is illustrated by going back to the original example and assuming two domains of interest. The proportions of the two domains in each stratum are shown below. Let $P_{dh}$ denote the proportion of units in domain $d$ in stratum $h$ and $P_{d'h}$ denote the proportion of units in domain $d'$ in stratum $h$.

<table>
<thead>
<tr>
<th>St.</th>
<th>$N_h$</th>
<th>$n_h$ (Normal)</th>
<th>$P_{dh}$</th>
<th>$P_{d'h}$</th>
<th>$n_{dh}$</th>
<th>$n_{d'h}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>19,850</td>
<td>2832</td>
<td>0.05</td>
<td>0.02</td>
<td>142</td>
<td>56</td>
</tr>
<tr>
<td>2</td>
<td>3,250</td>
<td>755</td>
<td>0.10</td>
<td>0.05</td>
<td>76</td>
<td>38</td>
</tr>
<tr>
<td>3</td>
<td>1,007</td>
<td>321</td>
<td>0.15</td>
<td>0.10</td>
<td>48</td>
<td>32</td>
</tr>
<tr>
<td>4</td>
<td>606</td>
<td>262</td>
<td>0.20</td>
<td>0.15</td>
<td>52</td>
<td>39</td>
</tr>
<tr>
<td>T</td>
<td>24,713</td>
<td>4170</td>
<td>0.064</td>
<td>0.03</td>
<td>318</td>
<td>165</td>
</tr>
</tbody>
</table>

There are two domains of interest. 6.4% of the total units in the population belong to domain 1 and 3% of the total units belong to domain 2. We want say 432 units in the sample which belong to domain 1 and 215 units in the sample belonging to domain 2. That is we want $n_d^* = 432$ and $n_{d'}^* = 215$.

Using the allocation given earlier for one domain with an increase in sample size, we now get a new allocation to satisfy the requirement for domain 2. This is shown in the table below. We also compute the expected domain sample size for domain 1 with this new allocation.

<table>
<thead>
<tr>
<th>St.</th>
<th>$n_h^*$</th>
<th>$n_{d'h}^*$</th>
<th>$n_{dh}^*$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>2860</td>
<td>57</td>
<td>143</td>
</tr>
<tr>
<td>2</td>
<td>825</td>
<td>41</td>
<td>82</td>
</tr>
<tr>
<td>3</td>
<td>462</td>
<td>46</td>
<td>69</td>
</tr>
<tr>
<td>4</td>
<td>474</td>
<td>71</td>
<td>95</td>
</tr>
<tr>
<td>T</td>
<td>4621</td>
<td>215</td>
<td>389</td>
</tr>
</tbody>
</table>

We see from the above table that the requirement for domain 2 is satisfied but not domain 1. We need 432 but we only have 389. The allocation is now changed to meet the requirement for domain 2. This results in the final allocation as follows.

<table>
<thead>
<tr>
<th>St.</th>
<th>$n_h^*$ (final)</th>
<th>$n_{d'h}^*$</th>
<th>$n_{dh}^*$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>2889</td>
<td>144</td>
<td>58</td>
</tr>
<tr>
<td>2</td>
<td>882</td>
<td>88</td>
<td>44</td>
</tr>
<tr>
<td>3</td>
<td>548</td>
<td>82</td>
<td>55</td>
</tr>
<tr>
<td>4</td>
<td>589</td>
<td>118</td>
<td>88</td>
</tr>
<tr>
<td>T</td>
<td>4908</td>
<td>432</td>
<td>245</td>
</tr>
</tbody>
</table>

In the final allocation both the requirements are met.

An alternative is to minimize the quantity

$$\sum_{h=1}^{L} (n_{hi}^*-n_{h})^2$$
under the constraints
\[ \sum_{h=1}^{L} n_h \cdot N_{dh}^h / N_h = n_d^* \quad \text{and} \quad \sum_{h=1}^{L} n_h \cdot N_{dh}^h / N_h = n_d^*. \]

The allocation \( n_h^* \) which satisfies the above conditions is given below.

\[
n_h^* = n_h \cdot \frac{(n_d^* - E(n_d))}{\left( \sum_{h=1}^{L} P_{dh} \cdot d_{h} \right) - \sum_{h=1}^{L} P_{dh} \sum_{h=1}^{L} P_{dh}^2 \cdot d_{h}} + \frac{[n_h^* - n_d^*]}{\left( \sum_{h=1}^{L} P_{dh} \cdot d_{h} \right)^2 - \sum_{h=1}^{L} P_{dh}^2 \sum_{h=1}^{L} P_{dh}^2 \cdot d_{h}} \cdot \frac{1}{\sum_{h=1}^{L} P_{dh} \cdot d_{h}} \cdot \frac{1}{d_{h}}
\]

This allocation will give the exact expected sample sizes in the domains but may not be necessarily be efficient.

3. Revision of selection probabilities.

In large scale household surveys, the sample is usually selected in several stages. The primary sampling units at the first stage of selection are usually selected with probability proportional to the total population in each primary sampling unit in order to obtain efficient estimates of population totals and means. If the primary sampling units have very different proportions of the total population which belong to a subpopulation of interest and if it is desired to maximize the expected proportion of the sample which belong to this population, then it is possible to revise the original probabilities of selection to achieve this goal and at the same time keep the revisions to a minimum so as not to affect the efficiency of the overall estimates. In this section, a simple procedure of revising the probabilities is suggested.

Suppose we are selecting a sample of \( n \) primary sampling units (PSU) from \( N \) units. Let the number of elementary units in the \( i \)-th PSU be \( N_i \) and the number of elementary units belonging to the \( j \)-th domain in the \( i \)-th PSU be \( N_{ij} \). Let \( N_{ij} \) be the proportion of the \( j \)-th domain in the \( i \)-th PSU.

The size of the domain in the population is

\[ N_j = \sum_{i=1}^{M} N_{ij}. \]

Let \( n_j \) be the sample size from domain \( j \) in the overall sample. Let \( \pi_i \) be the probability of inclusion of \( i \)-th PSU in the sample. Generally, these inclusion probabilities are proportional to some measure of size to maximize the efficiency of overall estimates.

Procedures for revising the measures of size or deriving composite measures of size to reflect domain sizes and also achieve desired domain sample sizes can be found in Folsom, Potter and Williams (1987) and Fahimi and Judkins (1991).

In this section, the idea is to revise the optimum probabilities of selection so as to maximize the expected proportion of the sample which belongs to a specific domain. The objective is to keep the revised probabilities of selection as close to the original as possible and at the same time maximize the expected proportion of domain units in the sample.

Let \( \pi_i^* \) be the revised probability of selection of the \( i \)-th PSU. We want to maximize the quantity

\[ \sum_{i=1}^{M} \pi_i^* \cdot P_i - \sum_{i=1}^{M} (\pi_i^* - \pi_i)^2 \]

subject to the constraint that
Using this criterion, the revised probabilities for the selection of psus when we are interested in one domain is

\[ \pi_i^* = \pi_i + \frac{1}{2} \left[ \frac{\sum_{i=1}^{M} P_{ij}}{M} \right] \]

Consider the following example in which we want to select 3 psus out of 7.

The initial probabilities of inclusion and also the revised probabilities are shown.

<table>
<thead>
<tr>
<th>PSU</th>
<th>( N_i )</th>
<th>( P_i )</th>
<th>( \pi_i )</th>
<th>( \pi_i^* )</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>150</td>
<td>0.2</td>
<td>0.375</td>
<td>0.32</td>
</tr>
<tr>
<td>2</td>
<td>50</td>
<td>0.4</td>
<td>0.125</td>
<td>0.17</td>
</tr>
<tr>
<td>3</td>
<td>300</td>
<td>0.10</td>
<td>0.750</td>
<td>0.65</td>
</tr>
<tr>
<td>4</td>
<td>250</td>
<td>0.50</td>
<td>0.625</td>
<td>0.72</td>
</tr>
<tr>
<td>5</td>
<td>150</td>
<td>0.15</td>
<td>0.375</td>
<td>0.30</td>
</tr>
<tr>
<td>6</td>
<td>100</td>
<td>0.60</td>
<td>0.250</td>
<td>0.40</td>
</tr>
<tr>
<td>7</td>
<td>200</td>
<td>0.20</td>
<td>0.500</td>
<td>0.44</td>
</tr>
<tr>
<td>Total</td>
<td>1200</td>
<td></td>
<td>3</td>
<td>3</td>
</tr>
</tbody>
</table>

References


