# Some Order Relations Between Selection and Inclusion Probabilities for PPSWOR Sampling Scheme 

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#### Abstract

In this paper we study some order relations between the selection and the inclusion probabilities for PPSWOR Sampling Scheme. We also establish some interesting bounds on the inclusion probabilities in terms of the selection probabilities.


## 1 Introduction

Consider a finite population $U=\{1, \ldots, N\}$ of $N$ identifiable units and a positivevalued size measure $X$ taking value $X_{i}$ on unit $i$. We denote by $p_{i}$ the normed size measure $X_{i} / \sum_{i=1}^{N} x_{i}$ and by $p=\left(p_{1}, \ldots, p_{N}\right)$ the normed size vector. A sample of size $n$ is selected from the population using a probability proportional to size without replacement (PPSWOR) sampling design based on $p$. For an arbitrary subset $s$ of $U$ let $\pi(s)$ denote the probability of including $s$ in the sample. For $s=\{i\}, \pi(s)$ will be denoted by $\pi_{i}$. In many practical survey problems it is of interest to control the values of $\pi_{i}$ (and also of $\pi\{i, j\}$ ) in order to get stable estimates or to obtain certain preferred samples. It is thus important to find out how the values of $\pi_{i}$ and $\pi_{j}$ are related for a given relationship between $p_{i}$ and $p_{j}$. It is also possible that a transformation on the available size measure or an altogether different choice of a new size measure would lead to a more suitable choice of $\pi_{i}$ values, say $\pi_{i}^{\prime}$, under the PPSWOR scheme. In this paper we consider some questions relating to the behaviour of $\pi_{i}$ values for a given $p$ as well as the behaviour of $\pi_{i}^{\prime}$ and $\pi_{i}$ vis-a-vis the relationship between $p_{i}^{\prime}$ and $p_{i}$, where

[^0]$p_{i}^{\prime}=X_{i}^{\prime} / \sum_{i=1}^{N} X_{i}^{\prime}, X_{i}^{\prime}$ being the value on the unit $i$ for the alternative choice of the size measure. In the process we also establish some interesting bounds on $\pi_{i}$ 's in terms of $p_{i}$ 's.

## 2 Main Results

For PPSWOR sampling scheme based on an initial selection probability vector $p$, the probability of selecting an ordered sample $\left(i_{1}, \ldots, i_{n}\right)$, $1 \leq i_{1} \neq \ldots \neq i_{n} \leq N$, is given by

$$
\begin{equation*}
p\left(i_{1}, \ldots, i_{n}\right)=p_{i_{1}} \ldots p_{i_{n}}\left(1-p_{i_{1}}\right)^{-1} \ldots\left(1-p_{i_{1}}-\ldots-p_{i_{n-1}}\right)^{-1} . \tag{2.1}
\end{equation*}
$$

The probability of selecting an unordered sample (subset of size $n$ ) $s_{n}$ is obtained by summing the probabilities of selection of $n$ ! ordered samples given by the $n$ ! permutations of the elements of $s_{n}$. Andreatta and Kaufman (1986) in their Theorem 4.1 obtained an interesting and useful integral representation of $p\left(s_{n}\right)$ given by

$$
\begin{align*}
p\left(s_{n}\right) & =\left(\sum_{i=1}^{N} X_{i}-\sum_{i \in s_{n}} X_{i}\right) \int_{0}^{\infty} e^{-\lambda\left(\sum_{i=1}^{N} X_{i}-\sum_{i \in s_{n}} X_{i}\right)} \prod_{i \in s_{n}}\left(1-e^{-\lambda X_{i}}\right) d \lambda \\
& =\left(1-\sum_{i \in s_{n}} p_{i}\right) \int_{0}^{1} t^{-\sum_{i \in s_{n}} p_{i}} \prod_{i \in s_{n}}\left(1-t^{p_{i}}\right) d t \tag{2.2}
\end{align*}
$$

(on substituting $t=e^{\left.-\lambda \sum_{i=1}^{N} X_{i}\right) \text {. }}$
For a given $\boldsymbol{p}$, we first prove certain inequalities connecting $\pi_{i}$ 's and corresponding $p_{i}$ 's. We first have an order relation involving $\pi_{i}$ 's and $p_{i}$ 's. More generally we prove the following.

Theorem 2.1: Let $\left.i, j \in U, i \neq j, s_{0} \subset U \backslash i, j\right\}$ with $0 \leq l=\left|s_{0}\right| \leq n-1, s_{1}=s_{0}+\{i\}$ and $s_{2}=s_{0}+\{j\}$. Then $\pi\left(s_{1}\right) \geq \pi\left(s_{2}\right)$ according to $p_{i} \geq p_{j}$.

Proof: Consider first $l=n-1$ and assume without loss of generality $i=1, j=2$. Then $s_{1}=\left\{i_{1}, i_{2}, \ldots, i_{n}\right\}, s_{2}=\left\{j_{1}, j_{2}, \ldots, j_{n}\right\}$ with $i_{1}=1, j_{1}=2, i_{v}=j_{v}, 2 \leq v \leq n$ and
$\left|s_{1}\right|=\left|s_{2}\right|=n$. We then have $\pi\left(s_{1}\right)-\pi\left(s_{2}\right)=\sum c_{\tau}\left[H_{\tau}\left(p_{1}\right)-H_{\tau}\left(p_{2}\right)\right]$, where $T_{n}$ is the set of all permutations $\tau$ of $\{1, \ldots, n\}, \tau \in T_{n}$

$$
\begin{aligned}
c_{\tau} & =\prod_{v=1}^{v_{0}}\left[1-\sum_{\mu=1}^{v-1} p_{\mu}^{*}\right]^{-1} \prod_{\substack{j=1 \\
j \neq v_{0}}}^{n} p_{j}^{*}, \\
H_{\tau}(x) & =x \cdot \prod_{v=v_{0}+1}^{n}\left[d_{v}(\tau)-x\right]^{-1}, \quad d_{v}(\tau)=1-\sum_{\substack{\mu=1 \\
\mu \neq v_{0}}}^{v-1} p_{\mu}^{*},
\end{aligned}
$$

$v_{0}=\tau^{-1}(1)$ and $p_{v}^{*}=p_{i_{\tau(v)}}$ for $v \neq v_{0}$.
Since $c_{\tau}>0$ and $H_{\tau}(x)$ is strictly increasing for $0<x<\min _{v_{0}+1 \leq v \leq n} d_{v}(\tau)$ the assertion follows for $l=n-1$.

For general $l(0 \leq l \leq n-1)$ the assertion follows from the first case by noting that $\pi\left(s_{1}\right)-\pi\left(s_{2}\right)=\sum_{s \in S^{*}}\left[\pi\left(s_{1} U s\right)-\pi\left(s_{2} U s\right)\right]$, where $S^{*}$ consists of all subsets $s$ from $U \backslash\left(s_{0} U\{i, j)\right.$ ) containing exactly $n-l-1$ elements.

As corollary we obtain

## Corollary 2.1: If for any $i$ and $j, p_{i} \geq p_{j}$ then $\pi_{i} \geq \pi_{j}$ and conversely.

Remark 2.1: Such order relation between $\pi_{i}$ 's and $p_{i}$ 's is trivially true for all $\pi P S$ sampling schemes, where $\pi_{i}=n p_{i}$. This is also obvious for Midzuno-Sen (Midzuno (1952), Sen (1952)) sampling scheme, for in this case

$$
\pi_{i}=\frac{n-1}{N-1}+\frac{N-n}{N-1} p_{i} .
$$

Rao (1961) and Seth (1966) considered the sampling scheme, where the first two units are drawn by PPSWOR and the remaining ( $n-2$ ) units of the sample by simple random sampling without replacement (SRSWOR). For this scheme, we have

$$
\pi_{i}=\pi_{i}(\mathbf{2}) \frac{N-n}{N-2}+\frac{n-2}{N-2}
$$

where $\pi_{i}(2)$ is the inclusion probability of unit $i$ in the first two draws. From Corollary 2.1, it follows that if $p_{i} \geq p_{j}$ then $\pi_{i}(2) \geq \pi_{j}(2)$ and hence $\pi_{i} \geq \pi_{j}$.

For the case of SRSWOR, we have $p_{i}=1 / N, \pi_{i}=n / N \forall i$. When we deviate from the SRSWOR probability and consider a general $p$, we shall find out how the corresponding $\pi_{i}$ 's behave. We first establish the following theorem which provides some simple bounds on $\pi_{i}$ 's in terms of $\min \left(p_{i}\right)$ and max $\left(p_{i}\right)$.

Theorem 2.2: Let $i_{1}, i_{2} \in U$ with $p_{i_{1}}=\min _{1 \leq i \leq N} p_{i}$ and $p_{i_{2}}=\max _{1 \leq i \leq N} p_{i}$ and let $p_{i} T_{r}(i)$ denote the probability of getting unit $i$ in the $r$ th draw, $1 \leq r \leq n, 1 \leq i \leq N$. Then
(a) $\quad T_{r}\left(i_{1}\right) \leq T_{r+1}\left(i_{1}\right)$
(b) $\quad T_{r}\left(i_{2}\right) \geq T_{r+1}\left(i_{2}\right), \quad 1 \leq r \leq n$
(c) $n p_{i_{1}} \leq \pi_{i_{1}} \leq n / N$ and $n / N \leq \pi_{i_{2}} \leq n p_{i_{2}}$
(d) $n p_{i_{1}} \leq \pi_{i} \leq n p_{i_{2}}, \quad i \in U \backslash\left\{i_{1}, i_{2}\right\}$.

Proof: We can write

$$
T_{r+1}\left(i_{1}\right)=\sum_{s_{r} \in S_{r}} p\left(s_{r}\right)\left(1-\sum_{i \in s_{r}} p_{i}\right)^{-1},
$$

where $\boldsymbol{p}\left(s_{r}\right)$ denotes the probability of obtaining a PPSWOR $(N, r)$ subset $s_{r}$ and $S_{r}$ denotes the set of all subsets of $U \backslash\left\{i_{1}\right\}$ containing exactly $r$ units. Now

$$
\begin{aligned}
T_{r+1}\left(i_{1}\right)= & \sum_{s_{r} \in S_{r}}\left(1-\sum_{i \in s_{r}} p_{i}\right)^{-1} \sum_{s_{r-1} \subset s_{r}} p\left(s_{r-1}\right)\left(1-\sum_{i \in s_{r-1}} p_{i}\right)^{-1} \sum_{i \in s_{r} \backslash s_{r-1}} p_{i} \\
= & \sum_{s_{r-1} \in S_{r-1}} p\left(s_{r-1}\right)\left(1-\sum_{i \in s_{r-1}} p_{i}\right)^{-1} \sum_{s_{r} \supset s_{r-1}}\left(1-\sum_{i \in s_{r}} p_{i}\right)^{-1} \sum_{i \in s_{r} \backslash s_{r-1}} p_{i} \\
\geq & \sum_{s_{r-1} \in S_{r-1}} p\left(s_{r-1}\right)\left(1-\sum_{i \in s_{r-1}} p_{i}\right)^{-1}\left(1-p_{i_{1}}-\sum_{i \in s_{r-1}} p_{i}\right)^{-1} \\
& \times \sum_{\substack{s_{r} \supset s_{r-1} \\
s_{r} \in S_{r}}} \sum_{i \in S_{r} \backslash S_{r-1}} p_{i} \\
= & \sum_{s_{r-1} \in S_{r-1}} p\left(s_{r-1}\right)\left(1-\sum_{i \in s_{r-1}} p_{i}\right)^{-1}=T_{r}\left(i_{1}\right)
\end{aligned}
$$

which proves (a). The part (b) follows by similar arguments.

The rart of the inequalities in (c) with $n / N$ as a bound is easily established by conradiction. Let, if possible, $\pi_{i_{1}}>n / N$. Then, by Corollary 2.1, $\pi_{i}>n / N \quad \forall i$, which contradicts the relation that $\sum_{1}^{N} \pi_{i}=n$. Hence, we must have $\pi_{i_{4}} \leq n / N$. Similarly, it can be shown that $\pi_{i_{2}} \geq n / N$.

To prove the other part of the inequalities in (c) we note that $\pi_{i}=$ $p_{i} \sum_{r=1}^{n} T_{r}(i), T_{1}(i)=1$, whence the assertions follow by (a) and (b). The part (d) follows from (c) and Corollary 2.1.

Consider now a simple deviation from the SRSWOR given by the initial selection probability vector

$$
p=\left(p^{(1)}, \ldots, p^{(1)}, 1 / N, \ldots, 1 / N, p^{(2)}, \ldots, p^{(2)}\right)
$$

where $p^{(1)}<1 / N<p^{(2)}$. This situation may occur in practice when one has three different types of units homogeneous within each type and one wishes to select them with three different types of probabilities say $p^{(1)}, 1 / N$ and $p^{(2)}$. If $\pi^{(i)}$ denotes the inclusion probability of $i$ th type, $i=1,2,3$, it follows from (c) of Theorem 2.2 that $\pi^{(1)} \geq n p^{(1)}$ and $\pi^{(3)} \leq n p^{(2)}$. However, it does not follow that $\pi^{(2)} \geq$ or $\leq n / N$. It may thus be of interest to compare the values of $\pi^{(2)}$ with $n / N$ and more generally, the value of $\pi_{i}$ with $n p_{i}$ for a general $p$.

Towards this, we establish the following theorem giving lower bounds on $\pi_{i}$ 's in terms of corresponding $p_{i}$ 's.

Theorem 2.3: For a given $p=\left(p_{1}, \ldots, p_{N}\right)$,

$$
\pi_{i} \geq p_{i} c\left(p_{i}\right), \quad 1 \leq i \leq N,
$$

where $c\left(p_{i}\right)$ is the value of $\pi_{i} / p_{i}$ based on

$$
p^{(i)}=\left(\frac{1-p_{i}}{N-1}, \ldots, \frac{1-p_{i}}{N-1}, p_{i}, \frac{1-p_{i}}{N-1}, \ldots, \frac{1-p_{i}}{N-1}\right)
$$

Proof: Without loss of generality, take $i=1$. We then have,

$$
\begin{equation*}
\pi_{1}=p_{1}\left(1+\sum_{r=2}^{n} T_{r}(1)\right)=p_{1}\left(1+g\left(p_{2}, \ldots, p_{N}\right)\right), \quad \text { say } \tag{2.3}
\end{equation*}
$$

where $T_{r}(1)$ is defined as in Theorem 2.2. Now, by Lemma A. 2 of the Appendix, $g\left(p_{2}, \ldots, p_{N}\right)$ and hence, $\pi_{1} / p_{1}$ is minimum subject to $\left.\sum^{N} p_{i}=\right\}-p_{1}$ 2 when
$p_{2}=\ldots=p_{N}=\left(1-p_{1}\right) / N-1$. This completes the proof of the theorem.
It is easy to verify that $c\left(p_{i}\right)$ is decreasing in $p_{i}$ and that for $p_{i}=1 / N$, $c\left(p_{i}\right)=n$. From these, we immediately have the following.

Corollary 2.2: If $p_{i} \leq p_{i 0}$, then $\pi_{i} \geq p_{i} c\left(p_{i 0}\right)$.

Corollary 2.3: If $p_{i}=1 / N$, then $\pi_{i} \geq n / N$.

Corollary 2.4: If $p_{i} \leq 1 / N$, then $\pi_{i} \geq n p_{i}$

Remark 2.2: For $p_{i}>1 / N$, it is not, however, necessarily true that $\pi_{i}<n p_{i}$. The following is a counter-example.

Example 2.1: $N=3 . p=(0.01,0.34,0.65)$. Let $n=2$. Here $p_{2}>1 / N=\mathbf{0} 3333$. But $\pi_{2}=0.9748>2 p_{2}=0.68$.

Motivated by these results, we shall now ask the question whether for two initial selection probability vectors $\boldsymbol{p}$ and $\boldsymbol{p}^{\prime}$, the corresponding vector of inclusion probabilities $\pi=\left(\pi_{1}, \ldots, \pi_{N}\right)$ and $\pi^{\prime}=\left(\pi_{1}^{\prime}, \ldots, \pi_{N}^{\prime}\right)$ are related in the same way as $\boldsymbol{p}$ and $\boldsymbol{p}^{\prime}$ are.

The following example shows that $p_{i}<p_{i}^{\prime}$ does not necessarily imply that $\pi_{i}<\pi_{i}^{\prime}$ even in the case where $p_{i}^{\prime}$ is obtained as a simple transformation from $p_{i}$, namely $p_{i}^{\prime}=\alpha+\beta p_{i}$.

Example 2.2: Let $N=3, n=2$, and $p_{i}^{\prime}=0.1 / 3+0.9 p_{i}$. For $p=(0.1,0.2,0.7)$, $p^{\prime}=(0.37 / 3,0.64 / 3,1.99 / 3)$. Here $p_{2}<p_{2}^{\prime}$ but $\pi_{2}=0.6889>\pi_{2}^{\prime}=0.6637$.

Remark 2.3: In this connection we may note that such relation between $p$ and $\pi$ is true if the scheme used is a $\pi P S$ one or the scheme due to Midzuno-Sen (1952) (see Remark 2.1).

Finally, we investigate whether the vector of inclusion probabilities $\pi$ is isotonic in $\boldsymbol{p}$. The specific question can be posed as follows. Let $\boldsymbol{p}$ and $\boldsymbol{p}^{\prime}$ be two normed size vectors satisfying $p<p^{\prime}$ and let $\pi$ and $\pi^{\prime}$ be the corresponding vectors of inclusion probabilities underlying the PPSWOR desings each of size $n$. Is it then true that $\pi<\pi^{\prime}$ ? Curiously enough, in general, this does not hold. A counter-example is obtained by taking

Example 2.3: $N=3, n=2, p=(0.1,0.3,0.6)$ and $p^{\prime}=(0.32 / 3,1.3 / 3,1.38 / 3)$, resulting in $\pi=(0.2929,0.7833,0.9238)$ and $\pi^{\prime}=(0.2791,0.8542,0.8667)$. It is readily seen that $p<\boldsymbol{p}^{\prime}$ while $\pi>\pi^{\prime}$.

However, the following theorem shows that if we compare any initial probability vector $p$ and the corresponding $\pi$ with the SRSWOR probabilities, namely $p_{0}=(1 / N, \ldots, 1 / N)$ and $\pi_{0}=(n / N, \ldots, n / N)$, we always have $p<p_{0}$ and $\pi<\pi_{0}$.

Theorem 2.4: For every normed size vector $\boldsymbol{p}=\left(p_{1}, \ldots, p_{N}\right), p_{1} \leq p_{2} \leq \ldots \leq p_{N}$ we have
(i) $\sum_{j=1}^{t} \pi_{j} \leq t n / N \quad \forall t \geq 1$ and
(ii) $\sum_{j=1}^{t} p_{j} \leq t / N \quad \forall t \geq 1$.

Proof: For $t=1$, (i) follows by (c) of Theorem 2.2. Suppose now (i) holds for $t=t$, but does not hold for $t=t+1$. Then we must have $\pi_{t+1}>n / N$ implying by Corollary $2.1, \pi_{j}>n / N \quad \forall j=t+2, \ldots, N$. This gives $\sum_{j=1}^{N} \pi_{j}=\sum_{j=1}^{t+!} \pi_{j}$, $+\sum_{j=t+2}^{N} \pi_{j}>\frac{(t+1) n}{N}+\frac{(N-t-1) n}{N}=n$, which contradicts the relation ${ }^{j=1}$ that $\sum_{j=1}^{N} \pi_{j}=n$. Hence, if (i) holds for $t$ then it also holds for $t+1$. As (i) holds for $t=1$, it holds $\forall t \geq 1$.

It can be established by similar arguments (replacing $\pi_{j}$ by $p_{j}$ and $t n / N$ etc. by $t / N$ etc.) that (ii) holds.

## Appendix

Lemma A.I: Let $a>1, c>0$ and integers $r, k$ with $1 \leq r \leq k$ be given. Define $X=\left\{\boldsymbol{x}=\left(x_{1}, \ldots, x_{k}\right) \in R^{k}: x_{i} \geq 0, \sum_{i=1}^{k} x_{i}=c\right\}$ and $f_{p}(a, x)=\sum_{l=1}^{r} \sum_{s \in S_{l}} \prod_{i \in s}\left(a^{x_{i}}-1\right)$ for $x \in X$, where $S_{l}$ denotes the set of all subsets of $\{1, \ldots, k\}$ containing exactly $l$ elements. Then for $x^{*} \in X$ it holds that $f\left(x^{*}\right)=\min _{x \in X} f(x)$ iff $x_{i}^{*}=c / k, 1 \leq i \leq k$.

Proof: Since $X$ is a compact subset of $R^{k}$ and $f$ is continuous on $X$ there exists at least one $x^{*} \in X$ with $f\left(x^{*}\right)=\min _{x \in X} f(x)$. Suppose now that one component of
$x^{*}$ is not equal to $c / k$. Then there are $i_{1}, i_{2} \in\{1, \ldots, k\}$ such that $x_{i_{1}}^{*}<c / k<x_{i_{2}}^{*}$. Let now $\varepsilon>0$ with $x_{i_{1}}^{*}+\varepsilon \leq c / k \leq x_{i_{2}}^{*}-\varepsilon$ and define

$$
\hat{x}_{i}= \begin{cases}x_{i}^{*}, & \text { if } i \neq i_{1}, i_{2} \\ x_{i_{1}}^{*}+\varepsilon, & \text { if } i=i_{1} \\ x_{i_{2}}^{*}-\varepsilon, & \text { if } i=i_{2}\end{cases}
$$

Of course $x \in X$. Now it is easy to show

$$
f\left(x^{*}\right)-f(\hat{x})=b \cdot a^{-\varepsilon}\left(a^{\varepsilon}-1\right) \cdot\left(a^{\left.x_{2}^{*}-a^{x_{i_{1}}^{*}+\varepsilon}\right), ~}\right.
$$

where $b=\sum_{s \in S^{*}} \prod_{i \in s}\left(a^{x_{i}^{*}}-1\right)$ and $S^{*}$ is the set of all subsets of $\left\{1, \ldots, k \backslash \backslash i_{1}, i_{2}\right\}$ containing exactly $r-1$ elements. Because of $b>0, \varepsilon>0, a>1$ and $x_{i_{2}}^{*}>x_{i_{1}}^{*}+\varepsilon$ one would get $f\left(x^{*}\right)>f(\hat{x})$ in contradiction to $f\left(x^{*}\right)=\min _{x \in X} f(x)$. This completes the proof of the lemma.

Lemma A.2: For PPSWOR $(N, n)$ sampling scheme with an initial selection probability vector $p=\left(p_{1}, \ldots, p_{N}\right)$, let $g\left(p_{2}, \ldots, p_{N}\right)$ be defined as in (2.3). Then $g$ is minimum subject to $\sum_{i=2}^{N} p_{i}=1-p_{1}$ when $p_{i}=\left(1-p_{1}\right) /(N-1) \quad \forall i=2, \ldots, N$.

Proof: Observe that

$$
g\left(p_{2}, \ldots, p_{N}\right)=\sum_{r=1}^{n-1} \sum_{s_{r} \in S_{r}} p\left(s_{r}\right)\left(1-\sum_{i \in s_{r}} p_{i}\right)^{-1}
$$

where $p\left(S_{r}\right)$ denotes the probability of a PPSWOR $(N, r)$ subset $S_{r}$ and $S_{r}$ denotes the set of all subsets of $U \backslash\{1\}$ containing exactly $r$ units. Using the integral representation (2.2), we have

$$
g\left(p_{2}, \ldots, p_{N}\right\}=\sum_{r=1}^{n-1} \sum_{s_{r} \in S_{r}} \int_{0}^{1} \prod_{i \in s_{r}}\left(t^{-p_{i}}-1\right) d t=\int_{0}^{1} f_{n-1}\left(1 / t, p_{2}, \ldots, p_{N}\right) d t
$$

in the sotation of the function $f$ defined in Lemma A.1. By Lemma A.1, it now follows that $f_{n-1}\left(1 / t, p_{2}, \ldots, p_{n}\right)$ is minimum when $p_{i}=\frac{1-p_{1}}{N-1} \quad \forall 2 \leq i \leq N$, for every $t \in(0,1)$. Hence, the lemma follows.

Acknowledgement: The authors like to record their sincere thanks to the referee for suggesting the present proof of Lemma A. 1 and for many other suggestions which considerably improved the presentation of the results.

The authors also thank Dr. S. Das Gupta and Dr. J.C. Gupta for their comments which led to Example 2.3 and Theorem 2.4.

## References

1. Andreatta G, Kaufman GM (1986) Estimation of finite population properties when sampling is without replacement and proportional to magnitude. J Amer Statist Assoc 81:657-666
2. Midzuno $H$ (1952) On the sampling system with probability proportional to the sum of sizes. Ann Inst Stat Math 3:99-108
3. Rao JNK (1961) On the estimate of variance in unequal probability sampling. Ann Inst Stat Math 15:67-72
4. Sen AR (1952) Present status of probability sampling and use in estimation of farm characteristics (abstract). Econometrica 20:103
5. Seth GR (1966) On estimators of variance of estimate of population total in varying probabilities. J Ind Soc Agr Stat 18:52-56

Received 29 November 1989
Revised version 7 September 1990


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