

CHAPTER II :

ON NECESSARY BEST ESTIMATOR

2.0 SUMMARY

In this chapter we consider the concept of NBE of various orders as introduced by Prabhu-Ajgaonkar [12] and defined in definition 1.2.5. We simplify the proof of Prabhu-Ajgaonkar for NBE of order 1 and also correct his assertion regarding NBE of order 2.

2.1 NBE OF ORDER 1

While introducing the concept of NBE, Prabhu-Ajgaonkar [12], proved that a NBE of order 1 exists and it coincides with the Horvitz-Thompson estimator given by

$$T(s, \tilde{y}) = \sum_{i \in s} [Y_i / \pi_i] \text{ where } \pi_i = \sum_{s \ni i} p(s) \text{ is the}$$

inclusion probability for unit i .

His original proof involves complicated notation. We give below a simpler proof for the same.

Let $t(s, \tilde{y})$ be a linear estimator of the population

total for the finite population $\Omega = \{1, 2, \dots, N\}$. Then $t(s, \underline{Y})$ has the form (1.2.6). Also $t(s, \underline{Y})$ is unbiased for \underline{Y} if, and only if

$$\sum_{s \ni i} b(s, i) p(s) = 1, \quad i=1, 2, \dots, N. \quad \dots(2.1.1)$$

Further the variance of $t(s, \underline{Y})$ is given by (1.2.10).

Then NBE of order 1 can be obtained by choosing the coefficients $b(s, i)$ in (1.2.6) by minimising (1.2.10), subject to conditions (2.1.1), at those values of \underline{Y} for which exactly one Y_i is non-zero. For this we consider

$$\begin{aligned} \phi &= \sum_{i=1}^N \sum_{s \ni i} b^2(s, i) Y_i^2 p(s) \\ &+ \sum_{i \neq j}^N \sum_{s \ni \{i, j\}} b(s, i) \cdot b(s, j) Y_i Y_j p(s) \\ &- 2 \lambda_i \sum_{s \ni i} b(s, i) p(s). \end{aligned} \quad \dots(2.1.2)$$

Here the λ_i are Lagrange's multipliers. Differentiating (2.1.2) with respect to $b(s, i)$ and equating the derivative to zero, we get

$$b(s, i) Y_i^2 p(s) = \lambda_i p(s). \quad \dots(2.1.3)$$

Since $p(s) > 0$,

$$b(s, i) = \lambda_i / Y_i^2. \quad \dots(2.1.4)$$

Thus it is clear from (2.1.4) that, for $s \in S$, the coefficient $b(s, i)$ depends only upon i and not on the sample.

Now from (2.1.1) one can see that $b(s, i) = \frac{1}{n_i}$. This completes the proof.

2.2 NBE OF ORDER 2

We now prove here that NBE of Order 2 and higher do not exist for a non-unicluster design. This contradicts a result of Prabhu-Ajgaonkar [12].

To get NBE of Order 2, the coefficients $b(s, i)$ in (1.2.6) have to be obtained by minimising (1.2.10), subject to conditions (2.1.1), at those \underline{y} for which exactly two Y_i 's are non zero. We show that this procedure leads to a $t(s, \underline{y})$ for which $b(s, i)$ depends on i and not on s . We next show that this condition leads to a contradiction.

Choose s_1, s_2 in S such that $s_1 \neq s_2$ and a unit $i \in s_1 \cap s_2$. This is possible because the design is non-unicluster. We may assume that some unit $j \in s_1$ but $j \notin s_2$. Let \underline{y} be such that only Y_i and Y_j are non-zero. To minimise (1.2.10) subject to conditions (2.1.1) we consider

$$\phi = \text{Var} [t(s, \underline{y})] - 2 \lambda_i \sum_{s \ni i} b(s, i) p(s) - 2 \lambda_j \sum_{s \ni j} b(s, j) p(s) \quad \dots(2.2.1)$$

where λ_i, λ_j are Lagrange's multipliers.

Differentiating (2.2.1) with respect to $b(s_1, i)$ and $b(s_2, i)$ and equating the results to zero, we get

$$Y_i^2 b(s_1, i) p(s_1) + Y_i Y_j b(s_1, j) p(s_1) = \lambda_i p(s_1) \quad \dots(2.2.2)$$

$$\text{and } Y_i^2 b(s_2, i) p(s_2) = \lambda_i p(s_2). \quad \dots(2.2.3)$$

Since $p(s_1)$ and $p(s_2)$ are positive, we get

$$b(s_1, i) Y_i^2 + b(s_1, j) Y_i Y_j = \lambda_i \quad \dots(2.2.4)$$

$$\text{and } b(s_2, i) Y_i^2 = \lambda_i. \quad \dots(2.2.5)$$

From (2.2.4) and (2.2.5), we get

$$[b(s_1, i) - b(s_2, i)] Y_i^2 + b(s_1, j) Y_i Y_j = 0. \quad \dots(2.2.6)$$

Since (2.2.6) holds for all Y_i, Y_j , we must have

$$b(s_1, i) = b(s_2, i) \text{ and } b(s_1, j) = 0.$$

Thus on the one hand $b(s, i)$ depends only on i ; but on other hand $b(s_1, j) = 0$. This shows that $b(s, j) = 0$ for all $s \ni j$.

This contradicts the unbiasedness of $t(s, \underline{y})$. This contradiction proves that a NBE of Order 2 does not exist.

Example : Suppose that we take simple random samples of size 2 from a population of size 3. Then the samples with positive probabilities are $s_1 = \{1, 2\}$; $s_2 = \{2, 3\}$ and $s_3 = \{1, 3\}$.

Let T_1 be the HT estimator. Then

$$\begin{aligned} E(T_1^2) &= \frac{3}{4} \{ (Y_1 + Y_2)^2 + (Y_2 + Y_3)^2 + (Y_3 + Y_1)^2 \} \\ &= \frac{3}{2} \{ Y_1^2 + Y_2^2 + Y_3^2 + Y_1 Y_2 + Y_2 Y_3 + Y_3 Y_1 \} \end{aligned}$$

Consider the estimator T_2 for which

$$\begin{aligned} T_2(s_1, \underline{y}) &= T_2(s_3, \underline{y}) = \frac{3}{2} Y_1 \text{ and} \\ T_2(s_2, \underline{y}) &= 3(Y_2 + Y_3). \end{aligned}$$

Then

$$\begin{aligned} E(T_2^2) &= \frac{3}{2} Y_1^2 + 3(Y_2 + Y_3)^2 \\ &= \frac{3}{2} Y_1^2 + 3Y_2^2 + 3Y_3^2 + 6Y_2 Y_3. \end{aligned}$$

Hence

$$\text{Var}(T_2) - \text{Var}(T_1) = \underline{y}' A \underline{y}$$

where $A = \begin{bmatrix} 0 & -3/4 & -3/4 \\ -3/4 & 3/2 & 9/4 \\ -3/4 & 9/4 & 3/2 \end{bmatrix}.$

The leading 2×2 principal minor of A is negative. Thus

T_1 is not a NBE of Order 2.