

A Note on Unrelated Question Randomized Response Model

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SUMMARY

It is shown that the derivations of most of the results in the randomized response technique proposed by Singh *et al.* (2000) are incorrect. Corrections of these results are given.

Key words: Randomized response, Sampling designs, Relative efficiency.

1. INTRODUCTION

Warner (1965) introduced an ingenious technique known as randomized response technique (RR) for estimating π_x , the proportion of population possessing certain stigmatized character x (say) by protecting the privacy of respondents and preventing the unacceptable rate of non-response. Since then Warner's (1965) technique has been modified by several researchers. A comprehensive review is available in Chaudhuri and Mukherjee (1988). Following the Moor's (1971) technique Singh *et al.* (2000) proposed two alternative RR techniques described as follows.

Singh *et al.* (2000) — Method 1: Two independent samples S_1 and S_2 were selected by simple random sampling without replacement (SRSWOR) method. Each respondent in the S_1 sample was asked to perform the randomized device R_1 while the respondents belonging to both the samples S_1 and S_2 were asked to perform randomized device R_2 as described above. The respondents belonging to S_2 but not S_1 were directly asked whether or not they possess the neutral character y . The proposed estimator of π_x is given by

$$\hat{\pi}_p = w\hat{\pi}_1 + (1-w)\hat{\pi}_2 \quad (1)$$

$$\text{where } \hat{\pi}_1 = \frac{\hat{\theta}_1 - (1-p_1)\hat{\pi}_{2y}}{p_1}, \hat{\pi}_2 = \frac{\hat{\theta}_2 - (1-p_2)\hat{\pi}_{2y}}{p_2}$$

$\hat{\theta}_1$ = proportion of "yes" answers in S_i , $i = 1, 2$

$\hat{\pi}_{2y}$ = proportion of the respondents belong to sample S_2 but not belong to S_1 possess the character y and W is a suitable weight.

Method 2: At first, an initially sample \tilde{s} of size n was selected from the population U by SRSWOR method. The sample \tilde{s} was divided at random into two sub samples \tilde{s}_1 and \tilde{s}_2 of sizes n_1 (to be determined appropriately) and $n_2 (= n - n_1)$ respectively. Respondents belonging to the first sub-sample \tilde{s}_1 , were asked to perform randomized device R_1 while respondents belonging to the sub-sample \tilde{s}_2 were asked directly to answer the question (ii) relating to possession of the neutral character y . The proposed estimator for π_x is given by

$$\tilde{\pi}_x = \frac{\tilde{\theta}_1 - (1-p_1)\tilde{\pi}_{2y}}{p_1}$$

where $\tilde{\theta}_1$ and $\tilde{\pi}_{2y}$ are the proportions of yes answers in the first and second samples.

2. CORRECTIONS OF SINGH *et al.* (2000) RESULTS

In this section, we will show that the following results obtained by Singh *et al.* (2000) are incorrect and we present corrections.

For the Method 1, let S_{21} be the sample of size n_{21} consisting of units belonging to both the samples

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S_1 and S_2 , and S_{22} is a sample of size n_{22} ($= n_2 - n_{21}$) belonging to S_2 but disjoint to S_1 i.e. $S_2 = S_{21} \cup S_{22}$. Let $z_i(z'_i)$ be the RR obtained from the i th unit if it belongs to $S_1(S_{21})$. Let $x_i = 1$ if i th unit possess the character x and $x_i = 0$ otherwise. Similarly, $y_i = 1$ if i th unit possess the neutral character y and $y_i = 0$ otherwise. Denoting $E_R(E_p)$ and $V_R(V_p)$ respectively as expectation and variance with respect to randomized response (sampling design) we note the following.

$$\pi_x = \sum_{i=1}^N x_i / N, \quad \pi_y = \sum_{i=1}^N y_i / N$$

$$E_R(z_i) = p_1 x_i + (1 - p_1) y_i = w_i$$

$$E_R(z'_i) = p_2 x_i + (1 - p_2) y_i = \gamma_i$$

$$V_R(z_i) = w_i(1 - w_i) = \sigma_i^2$$

$$V_R(z'_i) = \gamma_i(1 - \gamma_i) = \sigma_i'^2$$

$$\hat{\theta}_1 = \frac{1}{n_1} \sum_{i \in S_1} z_i = \bar{z}(S_1)$$

$$\hat{\pi}_{2y} = \bar{y}(S_{22}) = \frac{1}{n_{22}} \sum_{i \in S_{22}} y_i$$

The incorrect results of Singh *et al.* (2000) paper are presented using notations of this paper as follows.

Result 1. (Lemma 3.3, page 247)

$$\text{Var}(\hat{\theta}_1) = \frac{\theta_1(1 - \theta_1)}{n_1} - \frac{n_1 - 1}{n_1(N - 1)} \pi_x(1 - \pi_x)$$

Result 2. (Lemma 3.4, page 247)

$$\text{Var}(\hat{\theta}_2) = \left(\theta_2(1 - \theta_2) - \frac{\pi_x(1 - \pi_x)}{N - 1} \right) E\left(\frac{1}{n_{21}} \right) - \frac{\pi_x(1 - \pi_x)}{N - 1}$$

Result 3. (Lemma 3.5, page 248)

For uncorrelated x and y

$$\text{Cov}(\hat{\theta}_1, \hat{\pi}_{2y}) = \frac{N - n_1}{n_1(N - 1)} (1 - p_1) \pi_y(1 - \pi_y)$$

Result 4. (Lemma 3.6, page 248)

For uncorrelated x and y

$$\text{Cov}(\hat{\theta}_2, \hat{\pi}_{2y}) = \frac{N - n_1}{n_1(N - 1)} (1 - p_2) \pi_y(1 - \pi_y)$$

Result 5. (Lemma 3.7, page 249)

$$\text{Var}(\tilde{\theta}_1) = \left(\frac{\theta_1(1 - \theta_1)}{n_1} - \frac{(n_1 - 1)\pi_x(1 - \pi_x)}{n_1(N - 1)} \right) \quad (\text{for Method 2})$$

2.1 Corrections of the above Results

Result 1.

$$\begin{aligned} \text{Var}(\hat{\theta}_1) &= \text{Var}[\bar{z}(S_1)] = E_p(V_R(\bar{z}(S_1))) \\ &\quad + V_p(E_R(\bar{z}(S_1))) \\ &= \frac{1}{n_1 N} \sum_{i=1}^N \sigma_i^2 + V_p\left(\frac{1}{n} \sum_{i \in S_1} w_i\right) \\ &= \frac{\theta_1(1 - \theta_1)}{n_1} - \frac{n_1 - 1}{n_1(N - 1)} (p_1^2 \pi_x(1 - \pi_x) \\ &\quad + (1 - p_1)^2 \pi_y(1 - \pi_y) + 2p_1(1 - p_1)\pi_{xy}^*) \end{aligned}$$

where $\pi_{xy}^* = \pi_{xy} - \pi_x \pi_y$, $\pi_{xy} = \sum_{i=1}^N x_i y_i / N$

In case x and y are independent $\pi_{xy}^* = 0$ and we get

$$\begin{aligned} \text{Var}(\hat{\theta}_1) &= \frac{\theta_1(1 - \theta_1)}{n_1} - \frac{n_1 - 1}{n_1(N - 1)} (p_1^2 \pi_x(1 - \pi_x) \\ &\quad + (1 - p_1)^2 \pi_y(1 - \pi_y)) \end{aligned}$$

which is quite different from Result 1 obtained by Singh *et al.* (2000). It should be noted that the expression $\text{Var}(\hat{\theta}_1)$, obtained by Singh *et al.* (2000), is independent of π_y which is incorrect and can be checked from the fact that $z_i = y_i = w_i$ for $p_1 = 0$.

Result 2.

$$\begin{aligned} \text{Var}(\hat{\theta}_2) &= \text{Var}(\bar{z}'(S_{21})) \\ &= E_p(V_R(\bar{z}'(S_{21}))) + V_p(E_R(\bar{z}'(S_{21}))) \end{aligned}$$

Now writing $E_{n_{21}}$ as the unconditional expectation over n_{21}

$$\sigma_i'^2 = V_R(z'_i) = p_2 x_i + (1 - p_2) y_i - \gamma_i^2$$

$$\gamma_i = E_R(z'_i) = p_2 x_i + (1 - p_2) y_i$$

$$\begin{aligned}
 E_p(V_R(\bar{z}'(S_{21}))) &= E_p\left(\frac{1}{n_{21}} \sum_{i \in S_{21}} \sigma_i'^2\right) \\
 &= E_{n_{21}}\left(E_p\left(\frac{1}{n_{21}} \sum_{i \in S_{21}} \sigma_i'^2 \mid n_{21}\right)\right) \\
 &= \frac{1}{N} \sum_{i \in U} \sigma_i'^2 E\left(\frac{1}{n_{21}}\right) \tag{2}
 \end{aligned}$$

and $V_p(E_R(\bar{z}'(S_{21}))) = V_p(\bar{\gamma}(S_{21}))$

$$\begin{aligned}
 &= E_{n_{21}}(V_p(\bar{\gamma}(S_{21}) \mid n_{21})) \\
 &\quad + v(E_p(\bar{\gamma}(S_{21}) \mid n_{21})) \\
 &\doteq E_{n_{21}}\left(\frac{1}{n_{21}} - \frac{1}{N}\right) S_{\bar{\gamma}}^2 \\
 &= \left(E\left(\frac{1}{n_{21}}\right) - \frac{1}{N}\right) \frac{N}{(N-1)} \Pi_{xy}(p_2) \tag{3}
 \end{aligned}$$

where

$$\begin{aligned}
 \bar{\gamma}(S_1) &= \sum_{i \in S_{21}} \gamma_i / n_{21} \\
 (N-1)S_{\bar{\gamma}}^2 &= \sum_{i \in U} (\gamma_i - \bar{\gamma})^2 \\
 \bar{\gamma} &= \sum_{i \in U} \gamma_i / N
 \end{aligned}$$

and

$$\begin{aligned}
 \Pi_{xy}(p) &= \\
 &(p^2 \pi_x (1 - \pi_x) + (1-p)^2 \pi_y (1 - \pi_y) + 2p(1-p) \pi_{xy}^*) \\
 &\quad p = p_1, p_2 \tag{4}
 \end{aligned}$$

From (2) and (3), we get

$$\begin{aligned}
 \text{Var}(\hat{\theta}_2) &= \frac{1}{N} \sum_{i \in U} \sigma_i'^2 E\left(\frac{1}{n_{21}}\right) \\
 &\quad + \left(E\left(\frac{1}{n_{21}}\right) - \frac{1}{N}\right) \frac{N}{(N-1)} \Pi_{xy}(p_2) \\
 &= \left(\theta_2(1-\theta_2) + \frac{\Pi_{xy}(p_2)}{N-1}\right) E\left(\frac{1}{n_{21}}\right) - \frac{\Pi_{xy}(p_2)}{N-1}
 \end{aligned}$$

Result 3.

$$\begin{aligned}
 \text{Cov}(\hat{\theta}_1, \hat{\pi}_{2y}) &= \text{Cov}(\bar{z}(S_1), \bar{y}(S_{22})) \\
 &= E_{n_{21}}(\text{Cov}(\bar{w}(S_1), \bar{y}(S_{22}) \mid n_{21})) \\
 &\quad + \text{Cov}_{n_{21}}(E(\bar{w}(S_1 \mid n_{21})), E(\bar{y}(S_{22} \mid n_{21})))
 \end{aligned}$$

$$\begin{aligned}
 &= E_{n_{21}}(\text{Cov}[\bar{w}(S_1), \bar{y}(U - S_1) \mid n_{21}]) \\
 &(\text{since } \text{Cov}(E(\bar{w}(S_1 \mid n_{21})), E(\bar{y}(S_{22} \mid n_{21}))) = 0) \\
 &= -\frac{n_1}{N - n_1} E_{n_{21}} \text{Cov}(\bar{w}(S_1), \bar{y}(S_1)) \\
 &= -\frac{1}{N} S_{wy} - \frac{1}{N-1} (p_1 \pi_{xy}^* + (1-p_1) \pi_y (1 - \pi_y)) \\
 &\text{If } x \text{ and } y \text{ are uncorrelated, we get} \\
 \text{Cov}(\hat{\theta}_1, \hat{\pi}_{2y}) &= -\frac{(1-p_1) \pi_y (1 - \pi_y)}{N-1} \tag{5}
 \end{aligned}$$

Result 4.

$$\text{Cov}(\hat{\theta}_2, \hat{\pi}_{2y}) = -\frac{(1-p_2) \pi_y (1 - \pi_y)}{N-1} \text{ when } x \text{ and } y$$

are independent.

(Proof of the Result 4 follows from (5))

Result 5.

It can be easily checked that

$$\text{Var}(\tilde{\theta}_1) = \frac{\theta_1(1-\theta_1)}{n_1} - \frac{n_1-1}{n_1(N-1)} \Pi_{xy}(p_1) \tag{6}$$

For uncorrelated x and y, (6) reduces to

$$\begin{aligned}
 \text{Var}(\tilde{\theta}_1) &= \frac{\theta_1(1-\theta_1)}{n_1} - \frac{n_1-1}{n_1(N-1)} (p_1^2 \pi_x (1 - \pi_x) \\
 &\quad + (1-p_1)^2 \pi_y (1 - \pi_y)) \tag{7}
 \end{aligned}$$

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