

A Generalized Mixture Estimator Of The Mean Of A Sensitive Variable In The Presence Of Non-Sensitive Auxiliary Information

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Abstract

Gupta et al. (2012) proposed a generalized regression-cum-ratio estimator and Koyuncu et al. (2014) proposed a generalized exponential estimator for the mean of the sensitive variable utilizing a non sensitive auxiliary variable. We propose a new generalized mixture estimator for estimating the population mean of a sensitive study variable. The expressions for Bias and Mean Square Error are derived up to the first order of approximation. Numerical examples show that the proposed generalized mixture estimator performs better than many of the existing estimators.

Keywords: Generalized regression-cum-ratio estimator, Generalized exponential estimator, Generalized mixture Estimator, Population mean, Auxiliary information

1 Introduction

Randomized response technique (RRT) can be used to estimate the mean of a sensitive variable Y where direct observation on Y is subject to bias. We assume a non sensitive auxiliary variable X is available and can be observed directly. Sousa et al. (2010) introduced a ratio type estimator and Gupta et al. (2012) proposed a regression and generalized regression-cum-ratio estimators based on RRT models to deal with this situation. Following Bahl & Tuteja (1991), Koyuncu et al. (2014) also proposed a generalized exponential type estimator to improve the efficiency of the mean estimator based on RRT models.

In this paper we propose an ordinary exponential ratio type estimator and two generalized mixture estimators where the RRT estimators of the mean of Y are further improved by using information from an auxiliary variable X . Expressions for the Bias and Mean Square Error are derived up to the first order of approximation. We will use the following notations.

Let Y be the sensitive study variable which cannot be observed directly. Let X be a non sensitive auxiliary variable which has a positive correlation with Y , and let S be a scrambling variable. Assume that S is independent of Y and X . The respondent is asked to report a scrambled response for Y given by $Z = Y + S$, but is asked to provide the true response for X . Let a random sample of size n be drawn without replacement from a finite population $U = (U_1, U_2, \dots, U_N)$. For i th population element, let y_i and x_i respectively be the values of the study variable Y and auxiliary variable X . Let $\bar{Y} = E(Y)$, $\bar{X} = E(X)$ and $\bar{Z} = E(Z)$ be the population means for Y, X and Z respectively. We assume that the population mean \bar{X} and the population variance S_x^2 of the auxiliary variable are known. Also, assume that population mean and the population variance for the scrambling variable S are known and given as $\bar{S} = E(S) = 0$

of the commonly known RRT estimators. For the proposed estimators all the percent relative efficiencies are greater 100 indicating that all these estimators are better than the RRT ordinary mean estimator. We also note that both of the proposed generalized mixture estimators are more efficient than the other estimators considered here. Furthermore, the choice $\alpha = 2$ works better than $\alpha = 1$. We may note that at a theoretical level, one may be tempted to optimize α . Our goal though was to have a general family of estimators where many of the existing estimators become special cases of the proposed estimator with specific choice of α . For example, with $\alpha = 0$ our generalized mixture estimator II becomes combination of the regression and exponential ratio type estimators. For $\alpha = 1$, it involves the ratio term also. For $\alpha = -1$, it involves the product term.

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