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Targeting the Poor in Vietnam using a Small Area Estimation Method

Nguyen Viet Cuong¹

Abstract

To estimate the poverty rate at the commune level, the Ministry of Labor, Invalids and Social Affairs firstly collects information on households' income per capita, and compares their income with a defined poverty line. A household is identified as a poor one if their income per capita is below the poverty line. Then, the poverty rate of a commune is simply the ratio of the defined poor households in that commune. This method can raise questions about the quality of the collected data on income. This paper examines the quality of expenditure data collected in this simple way. It is found that the expenditure data collected in the simple way are considerably lower than expenditure data collected using detailed questionnaires. Thus collection of expenditure or income data using simple questionnaires tends to underestimate the actual expenditure or income. For poverty targeting in small areas such as communes or districts, this paper suggests the application of this small areas estimation method.

Keywords: Poverty, the poor, poverty targeting, poverty mapping, small area estimation

JEL classification: I31, I32, O15

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1. Introduction

Poverty reduction is a major development policy in Vietnam. The Government of Vietnam has set up a target of reducing the poverty rate from 26% in 2005 to 15% in 2010 (MOLISA, 2005)². In order to support the poor, the first step is to locate where the poor live and to identify who the poor are. The Ministry of Finance will allocate more resources to the poorest communes, and then the Ministry of Labor, Invalid and Social Affairs (MOLISA) will identify who the poor are to provide supports. Officially, MOLISA collects information on households' income per capita, and then compares their income with a defined poverty line. A household then is identified as a poor one if their income per capita is below the poverty line. The poverty rate of a commune is simply the ratio of the defined poor households in that commune. This method can raise a question about the quality of the collected data on income. As is known, the income of a household can come from many sources and varies from one month to another. A quick interview using short questionnaires would result in income data with a large measurement error. However identifying the poor throughout the country using long interview with detailed questionnaires is not feasible, since it is extremely costly.

Estimation of poverty rates in small areas is proposed by Elbers et. al. (2003) and Hentchel et. al. (2000). They combine population censuses and household surveys to estimate poverty indices at the district and commune levels. Population censuses cover the whole population and collect information on basic household characteristics such as demography and housing, but not household income and expenditure. In contrast, household surveys often collect data on both household characteristics and expenditures (or also income) using very detailed questionnaires. Household surveys, however, are sample surveys and representative for national or regional levels. To estimate the poverty rate for small areas, an equation of expenditure (or income) is firstly estimated using a household survey, and then it is applied in a population census to predict expenditure (income) using information on household characteristics and the poverty rate. Once the poverty rate at the commune level can be estimated, the poor households in the commune

² This figure is estimated based on the new poverty line for the period 2006-2010. According to the poverty line, a household is classified as poor if their income per capita is below 200 thousand VND for the rural areas, or below 260 thousand VND for the urban areas.

can be identified by alternative methods such as through commune meetings or by using income data from MOLISA.

The most recent poverty map that used the small area estimation method was produced in 2003 by (Minot, Baulch, and Epprecht, 2003). They combine data from the Vietnam Living Standard Survey in 1998 (VLSS 1998) and data from the Population and Housing Census in 1999 to estimate the poverty rates at a small level, i.e. province, district and commune. In 2006 the Vietnam Census on Agriculture, Rural Areas and Aquaculture will be conducted, which allows for possibility to update the poverty map for the rural areas of Vietnam.

This paper has two main objectives. The first is to investigate whether collection of data on expenditure using quick and simple questionnaires can yield reliable expenditure data. The second is to test how the small area estimation works in the recent Vietnam Living Household Standard Survey (VHLSS) in 2004. If the method works well, there will be the possibility of updating the poverty map for Vietnam using the 2006 Agricultural Census and the 2006 VHLSS.

This paper is divided into 5 sections. The second section introduces data sources used in the paper. The third section examines the measurement error in expenditure data collected using the simple method. The fourth section presents results from the small area estimation. Finally, some conclusions are drawn in the final section.

2. Data Sources

The first data source is the Vietnam Household Living Standard Survey (VHLSS) that was conducted by the General Statistical Office of Vietnam (GSO) in the year 2004. The survey collected information on household characteristics including basic demography, employment and labor force participation, education, health, income, expenditure, housing, fixed assets and durable goods, the participation of households in the most important poverty alleviation programs. The full household sample of the VHLSS covered the 45000 households, of which 9000 households were asked for detailed information on consumption expenditure and income. Data on expenditure and income are collected using

very detailed questionnaires. Small and detailed items on expenditure and income are collected, and then aggregated into the expenditure and income per capita. The sample of 9000 households is used in this research. The sample is representative for the whole country and 8 geographic regions.

The second data source is from a pilot survey conducted by the Institute of Labor Science and Social Affairs (ILSSA) of MOLISA. This survey was conducted in early 2005. The survey collected data on household characteristics, expenditure and income of households in 9 communes in 4 provinces: Tuyen Quang, Thanh Hoa, Quang Nam, and An Giang. Expenditure and income are collected using simple questionnaires. It should be noted that all households in each commune are quickly surveyed by a “questionnaires part A1” to remove the rich households from the sample. Basically, the Part A1 uses short and quick questionnaires to find households who have high value assets such as cars or boats, or have stable sources of high income. These households can be regarded as rich ones and are removed from the sample. In addition, in some communes the interviewers do not collect information from households who are considered very poor by the “questionnaires part A2”. The remaining households which are considered are the poor and near-poor ones. These households are interviewed using “questionnaires part B”. This research uses data on these remaining households collected based on this “questionnaires part B”. The number of households in this sample is 4526.

3. Measurement Error in Expenditure Data of the Pilot Survey

3.1. Methodology

It is often argued that collection of expenditure data using simple survey methods can underestimate the actual expenditure. This section examines the accuracy of expenditure data of the pilot survey by comparing these data with expenditure data collected in the 2004 VHLSS. The pilot survey collects expenditure data using very simple questionnaires, while the 2004 VHLSS collects expenditure data using detailed questionnaires and thorough interviews. Expenditure data of the 2004 VHLSS can be regarded as a

benchmark, and expenditure data of the pilot survey will be considered accurate if they are close to the expenditure data of the 2004 VHLSS.

The aim of this test is to estimate the average difference between expenditure per capita of households in the 2004 VHLSS and expenditure per capita of households in the pilot survey after household characteristics variables are controlled for.

Assume that expenditure per capita, denoted by y_i is a function of observed variables X and unobserved variables ε :

$$y_i = \alpha + X_i\beta + \varepsilon_i. \quad (1)$$

We pool two samples of VHLSS and the pilot survey, and denote D as a binary variable indicating whether a household is from the pilot survey. Equation (1) can be expressed as follows:

$$y_i = \alpha + X_i\beta + D_i\theta + \varepsilon_i. \quad (2)$$

If there is no difference in expenditure data between the two surveys, the coefficient of variable D is not statistically significant. In this paper, we estimate equation (2) by OLS and quantile regression. We also estimate the coefficient D by a matching method with the distance metric of propensity score. The idea of the matching method is to find similar households between the 2004 VHLSS and the pilot survey. This can be regarded as a non-parametric method that is used to estimate the parameter θ .

3.2. Empirical Results

Before pooling the two samples, we need to make them comparable to avoid the problem of sample selection bias. It should be noted that the pilot survey did not cover some rich households in all communes nor some very poor households in 5 communes in two provinces. Thus to make the two samples more comparable, the procedure to remove the rich households using the questionnaires Part A1 is applied to the VHLSS 2004. As a result, 3644 households are removed, accounting for 40.4% of the 2004 VHLSS sample.

In addition, only households who are in 4 communes of two provinces “Tuyen Quang” and “An Giang” are kept for the analysis, since one did not apply the questionnaires Part A2 to exclude very poor households in these communes. The procedure is rather arbitrary; thus we cannot apply this method to the VHLSS 2004 to exclude the very poor households so that all samples of the pilot survey can be used for this expenditure comparison analysis.

Table 1 presents the estimates of the difference in expenditure per capita between the VHLSS 2004 and the pilot survey controlling for some observed variables that are correlated with the expenditure. Explanatory variables include urban and regional dummy variables, commune characteristics, household composition, education of household members, education and employment of household head and head’s spouse, housing characteristics, and assets. Households included in the analysis are restricted to low expenditure samples.³ Estimation from the regressions shows that expenditure data in the pilot survey are statistically significantly lower than expenditure data in the 2004 VHLSS. The non-parametric methods also produce a similar trend, but some of the estimates are not statistically significantly at 5%.

Table 1: Average difference in expenditure per capita between the 2004 VHLSS and the pilot survey: regression and propensity score matching estimations

| Estimation | Difference (D) (VND thousand) | Std. Err. | [95% Conf. Interval] | |
|---|---|------------------|-----------------------------|----------|
| OLS Regression | -492.145 | 83.907 | -656.626 | -327.665 |
| Quantile regression | | | | |
| 0.25 | -370.212 | 56.117 | -480.216 | -260.207 |
| 0.50 | -457.487 | 70.556 | -595.796 | -319.178 |
| 0.75 | -493.455 | 96.903 | -683.411 | -303.498 |
| Matching | | | | |
| One nearest neighbor | -334.2 | 371.0 | -1079.6 | 411.3 |
| Three nearest neighbors | -302.9 | 301.6 | -909.1 | 303.2 |
| Five nearest neighbors | -463.8 | 288.5 | -1043.6 | 115.9 |
| Kernel matching – h = 0.01 | -452.8 | 291.3 | -1038.1 | 132.6 |
| Kernel matching – h = 0.05 | -492.2 | 220.4 | -935.1 | -49.3 |
| Kernel matching – h = 0.1 | -566.8 | 173.0 | -914.4 | -219.2 |
| <i>Standard error in the matching method is estimated using bootstrap with 200 replications</i> | | | | |
| <i>Source: Estimation form the 2004 VHLSS and pilot survey</i> | | | | |

³ Full results of estimation are not reported in this paper. Those who are interested in the results can contact the authors for this information.

Thus, the collection of expenditure data using simple questionnaires and quick interview is more likely to underestimate the actual expenditure. The consumption basket has a large number of small items, and detailed questionnaires help people remember these small items in their consumption. These results suggest that income data collected by MOLISA might not be reliable in estimating the poverty rate at the commune or district level. An alternative way to estimate poverty indexes in small areas is to combine a census and a household survey using the method of small area estimation.

4. Small Area Estimation Method

4.1. Methodology

Poverty is widely measured by three Foster-Greer-Thorbecke poverty indices, which can all be calculated using the following formula (Foster, Greer and Thorbecke, 1984):

$$P_{\alpha} = \frac{1}{n} \sum_{i=1}^q \left[\frac{z - y_i}{z} \right]^{\alpha}, \quad (3)$$

where y_i is a welfare indicator (consumption expenditure per capita in this paper) for person i , z is the poverty line, n is the number of people in the sample population, q is the number of poor people, and α can be interpreted as a measure of inequality aversion.

When $\alpha = 0$, we have the headcount index H which measures the proportion of people below the poverty line. When $\alpha = 1$ and $\alpha = 2$, we have the poverty gap PG which measures the depth of poverty, and the squared poverty gap P_2 which measures the severity of poverty, respectively.

Poverty indices are often estimated using data from household surveys. In Vietnam, during the period 2000-2010, the GSO conducted Vietnam Household Living Standard Surveys (VHLSS) once every two years. However samples of VHLSS are often small, and results cannot be representative in small areas such as communes or districts. In contrast, population censuses cover large sections of the population but do not collect information on expenditure or income. Thus, poverty indices cannot be estimated directly

from censuses either. The method of “small area estimation”, developed by Elbers et. al., (2003), and Hentschel et. al., (2000), combines a household survey and a population census to estimate a poverty headcount index (H) for small areas. The main idea is to estimate an expenditure equation from a household survey, and use this equation to predict expenditure for households in a census given the households’ characteristics. Once predicted expenditures are available, poverty rate can be estimated at small areas.

In Vietnam, the most recent poverty map to use the small area estimation method was produced in 2003 by Minot et. al., 2003. They estimate poverty indices at the province, district and commune levels using the Vietnam Living Standard Survey of 1998 and the Population and Housing Census of 1999. In this paper, we examine how the method works using data from the 2004 VHLSS.

The method of small area estimation involves the following three steps:

Step 1: Select common variables from the household survey and a population census. These variables will be used in a regression of expenditure, and therefore should be correlated with expenditure.

Step 2: Run regression of expenditure on selected variables using data from the household survey:

$$\ln(y_i) = X'_i \beta + \varepsilon_i, \quad (4)$$

where:

- y_i is the real per capita consumption expenditure of household i
- X_i' is a $1 \times k$ vector of household characteristics of household i
- β is a $k \times 1$ vector of estimated coefficients
- ε_i is a random disturbance term distributed as $N(0, \sigma)$

The list of variables may be revised for the final equation.

Step 3: Apply this equation to the population census to predict the expected probability that household i is poor:

$$E[P_i | X_i, \beta, \sigma^2] = \Phi \left[\frac{\ln z - X_i^C \beta}{\sigma} \right], \quad (5)$$

where:

- P_i is a variable taking a value of 1 if the household is poor and 0 otherwise
- z is the poverty line
- Φ is the cumulative standard normal function
- X_i^C household characteristics for household i from the census

Then the poverty rate for an area can be estimated according to Elbers *et al* (2003):

$$E[P | X^C, \beta, \sigma^2] = \frac{m_i}{M} \sum_{i=1}^N \Phi \left[\frac{\ln z - X_i^C \beta}{\sigma} \right] \quad (6)$$

where:

- m_i is the size of household i
- M is the total population of the area in question
- N is the number of households in the area

To estimate the poverty gap index PG, and the poverty severity index P_2 , we employ the method proposed by Minot, et. al., 2003 to estimate the cumulative distribution of the expenditure per capita in the absence of the VBSP credit. This is done by changing the poverty line from the lowest expenditure per capita to the highest expenditure per capita in the sample. The estimated cumulative distribution is then used to estimate the poverty indexes PG and P_2 (in the state of no-credit from the program).

In this paper, we will estimate the poverty indices in two ways. In the first, poverty indices are estimated directly from expenditure data of the 2004 VHLSS. These estimates can be regarded as a benchmark, since they use the observed data on expenditure. The second way applies the method of small area estimation to obtain the poverty indices. If the small area estimation functions well in the 2004 VHLSS, it is expected that the two ways of estimation will produce close estimated poverty indexes.

4.2. Empirical Results

The poverty indices can be easily estimated from expenditure data using equation (3). The left panel of Table 1 presents poverty indices of 8 regions in Vietnam that are estimated using observed expenditure data. It is shown that there is large variation in the poverty indices between the regions. The poverty rate is lowest in the North East South, at around 5.4%. The region with the highest level poverty is the North West, in which 58.6% of the people are below the poverty line.

Table 2: Poverty indexes estimated by the small area estimation method in VHLSS 2004

| Region | Number of Obs. | Using expenditure data | | | Small area estimation | | |
|---------------------|----------------|------------------------|--------------------|--------------------|-----------------------|--------------------|--------------------|
| | | H | PG | P2 | H | PG | P2 |
| Red River Delta | 1944 | 0.1214 [0.0079] | 0.0211 [0.0017] | 0.0056 [0.0006] | 0.1146 [0.0062] | 0.0176 [0.0016] | 0.0038 [0.0005] |
| North East | 1317 | 0.2938 [0.0149] | 0.0701 [0.0043] | 0.0237 [0.0018] | 0.2856 [0.0144] | 0.0683 [0.0062] | 0.0226 [0.0030] |
| North West | 429 | 0.5857 [0.0259] | 0.1911 [0.0113] | 0.0803 [0.0061] | 0.4826 [0.0193] | 0.1305 [0.0092] | 0.0453 [0.0045] |
| North Central Coast | 1014 | 0.3190 [0.0165] | 0.0809 [0.0053] | 0.0292 [0.0026] | 0.3089 [0.0141] | 0.0740 [0.0056] | 0.0248 [0.0027] |
| South Central Coast | 852 | 0.1901 [0.0150] | 0.0510 [0.0059] | 0.0211 [0.0034] | 0.1928 [0.0120] | 0.0421 [0.0040] | 0.0131 [0.0017] |
| Central Highlands | 581 | 0.3315 [0.0232] | 0.1064 [0.0098] | 0.0451 [0.0053] | 0.3047 [0.0178] | 0.0787 [0.0066] | 0.0274 [0.0030] |
| North East South | 1188 | 0.0537 [0.0071] | 0.0120 [0.0019] | 0.0044 [0.0009] | 0.0389 [0.0040] | 0.0053 [0.0008] | 0.0010 [0.0002] |
| Mekong River Delta | 1863 | 0.1582 [0.0096] | 0.0299 [0.0024] | 0.0090 [0.0010] | 0.1428 [0.0072] | 0.0219 [0.0020] | 0.0047 [0.0006] |
| All Vietnam | 9169 | 0.1948 [0.0051] | 0.0472 [0.0015] | 0.0170 [0.0007] | 0.1952 [0.0043] | 0.0448 [0.0017] | 0.0147 [0.0008] |

Standard error (in bracket) is estimated using bootstrap with 200 replications

Source: Estimation from the 2004 VHLLS and pilot survey

The estimates of the poverty indices using the small area estimation method are presented in the right of Table 2. In this method, the first step is to construct the expenditure equation using the VHLSS 2004. In expenditure models, the dependent variables are often in logarithmic form, since this enables them to better fit the data. The main priority in constructing the expenditure model is to select appropriate explanatory variables taking into account that the number of explanatory variables should not be too large. In reality, the method often uses data from a census which cannot provide as

detailed information on households as household surveys. Explanatory variables include urban and regional dummy variables, commune characteristics, household composition, education of household members, education and employment of household heads and head's spouses, housing characteristics, and assets. The selection of explanatory variables is similar to the procedure of backward selection in stepwise regression. Variables are dropped when they have a P-value higher than 0.1.⁴

It is found that with the exclusion of the North West region, this method of small area estimation yields estimates of headcount ratios which are very close to those based on the collected data for all regions and the whole country. One reason for the large difference in poverty estimates in the North West is the small number of observations: there are 429 household observations in the 2004 VHLSS sample. Thus, a caution should be exercised when estimating poverty in areas with small numbers of observations. Another reason for the difference might be that the standard error tends to be higher for observations where values of the characteristic variables X are far from the sample mean values. In this paper we estimate a single equation of expenditure for the whole sample. The North West is the poorest region in the sample data, and predicted expenditures for this region can have large standard errors. When applying this method in a new census, different expenditure equations can be estimated for different regions or large provinces.

5. Conclusions

To support the poor, it is necessary to identify who the poor are and where they live. To classify the poor we need the information on welfare indicators such as income or expenditure. Collection of accurate data on expenditure (or income) using detailed questionnaires is unrealistic for a large sample. A traditional method is to collect expenditure (or income) using short and simple questionnaires. However, it is found that the expenditure data collected by the simple questionnaires tends to be lower than the actual expenditure.

⁴ Results from regressions and tests are not reported, but those who are interested in can contact the authors to obtain these results.

To estimate the poverty rate in small areas such as districts or communes, the method of small area estimation can be used. For reliable estimation, the number of observations in small areas should not be too small, and different models of expenditure should be employed in order to reduce standard errors of poverty estimates.

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