ESA/STAT/AC.320/28

Expert Group Meeting on Data Disaggregation 27-29 June 2016
New York

The Philippine Statistics Authority (PSA) Small Area Poverty Estimation Project By Bernadette Balamban

# The Philippine Statistics Authority (PSA) Small Area Poverty Estimation Project

Expert Group Meeting on Data Disaggregation June 27-29, 2016, New York City

# **Outline of Presentation**

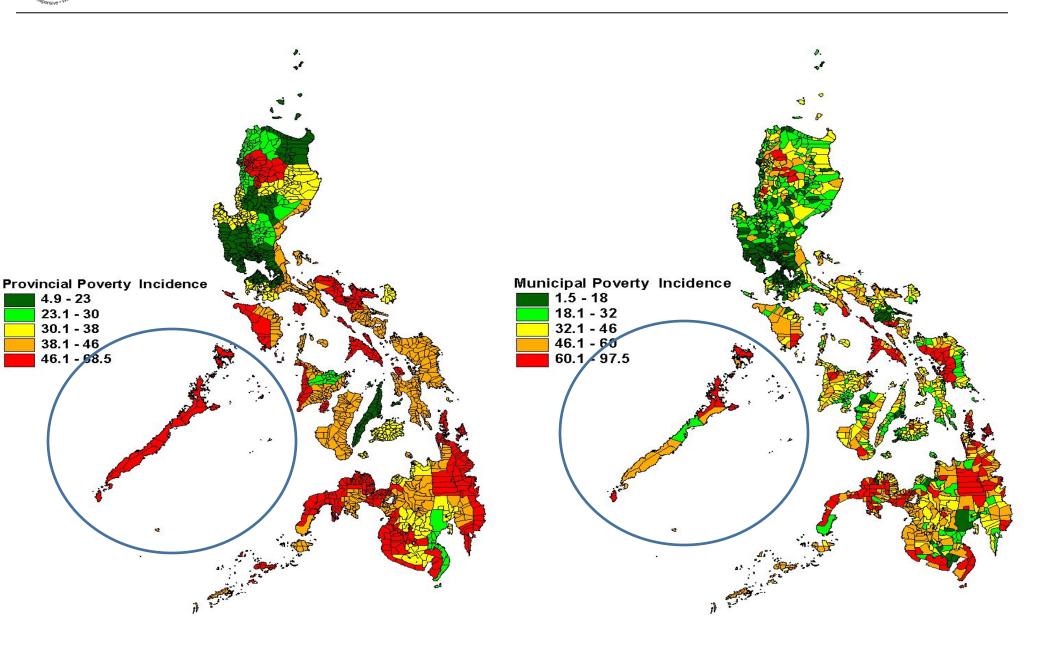
- I. Introduction
- II. Methodology
- III. Some Results
- IV. Actual Policy Uses
- V. Next Steps



### As of 31 December 2015

- > 17 Regions
- > 79 Provinces
- 145 Cities and1,489 Municipalities
- 42,036 Barangays

- Official poverty statistics in the Philippines are available at the national, regional and provincial levels. This are directly estimated from the Family Income and Expenditure Survey (FIES).
- However, more geographically disaggregated statistics are needed to make better decisions.
- Thus, there is a strong clamor from policymakers and program implementers for information on smaller domains like the cities and municipalities, specially for poverty statistics.



- Small Area Estimation could be used to target the municipalities and cities were most of the poor are found. The smaller the area, the better is the targeting.
- Given limited resources, local government officials, as well as those in the national government, would like to know how they can best allocate resources, which area needs to be prioritized.
- Programs could be better monitored if the estimates were made in smaller domains.

- Recognizing the need to be relevant and responsive, the former National Statistical Coordination Board (NSCB), which is now part of the PSA, with external funding and technical support from World Bank and AusAid and recently from the Philippine government, implemented projects on SAE to generate poverty incidences at the city/municipal levels.
- Adopted the Elbers, Lanjouw and Lanjouw (ELL) methodology of the World Bank
- The Project was made possible through technical assistance from the following:
  - 2000 Dr. Stephen Hasslett and Dr. Geoff Jones
  - 2003 Dr. Peter Lanjouw, Dr. Roy Vanderweide, Dr. Zita Albacea
  - 2006, 2009 and 2012 Dr. Zita Albacea

Project	Output	Year Released	Funding Source	Methodology/ Data Sets Used
Poverty Mapping in the Philippines	2000 city/ municipal level poverty estimates	2005	World Bank	ELL; National Model 2000 CPH, 2000 FIES/ Labor Force Survey (LFS)
Intercensal Updating of Small Area Estimates (SAE) on Poverty	2003 city/ municipal level poverty estimates	2008	World Bank	Modified ELL; Regional Model 2000 CPH, 2003 FIES/LFS Barangay Listing
Updating of SAE on Poverty	2006 city/ municipal level poverty estimates	2013	World Bank, AusAid, Gov't. of the Philippines (GOP)	Modified ELL; Regional Model 2000 CPH, 2006 FIES/LFS Barangay Listing
Updating of SAE on Poverty	2009 city/ municipal level poverty estimates	2012	World Bank, AusAid, GOP	Modified ELL; Regional Model 2007 CP, 2009 FIES/LFS Barangay Listing
Updating of SAE on Poverty	2012 city/ municipal level poverty	2014	GOP	Modified ELL; Regional Model 2010 CPH, 2012 FIES/LFS Barangay Listing

# II. Methodology

#### Main idea

- Merge information from different types of data sources to come up with small area poverty estimates
- "Borrow strength" from the much more detailed coverage of the census data to supplement the direct measurements of the survey

### **Basic procedure**

- Use the household survey data to estimate a model of per capita income (Y) as a function of variables that are common to both the household survey and the census (X's).
- Use the resulting estimated equation/model to predict per capita income for each household in the census.
- The estimated household-level per capita income are then compared with the threshold to identify poor and non-poor households.
- Proportion of poor individuals are then aggregated for small areas, such as cities and municipalities.

# II. Methodology

### **Regression Model**

$$\ln Y_{ij} = X_{ij}\beta + h_i + e_{ij}$$

where Y<sub>ij</sub> is the target variable (per capita income) is log-transformed to make the distribution more symmetrical;

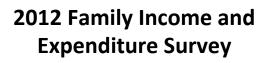
X<sub>ij</sub> are the household and community level characteristics;

h<sub>i</sub> is the error term held in common by the i<sup>th</sup> cluster; and

e<sub>ij</sub> is the household level error within the cluster.

# Philippine Statistics Authority

# II. Methodology



**2012 Labor Force Survey** 

**2010** Census of Population and Housing

Variable definition, values and labels were checked for consistency



Timeinvariant variables



Variables from Barangay listing



Averages at the municipal or city level

"Best" Predicting Model



**Indirect Estimation** of Poverty **Statistics** 

Model **Building by Region and** Model **Evaluation/** Selection

2012 City and

**Municipal Level** 

**Poverty Statistics** 

based on SAE



**Predictors** of the model



Validation and Dissemination of Estimates

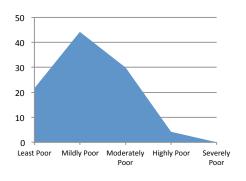
# II. Methodology

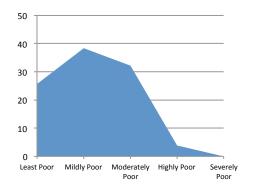
### Criteria in Choosing the "Best" Predicting Model

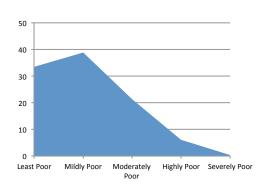
- The relationship of the variables, whether positive or negative, on Y is generally consistent with earlier researches on poverty (e.g. education should have a positive effect on income).
- The models should be robust, which means that small changes to the model do not greatly affect the significance or signs of the variables.
- Estimated regional poverty incidence does not largely differ from the official regional poverty estimates (within 2 standard error away from the official estimates). Preserve the ranking of the official provincial estimates within a region.
- 'Good' statistical properties of the model like acceptable model adequacy; significant regression coefficients; parsimonious model;

# **III. Some Results**

Poverty Classification	Poverty Incidence Among Population (%)	2006	2009	2012
Level 1	At most 20.0	357	419	545
Level 2	21.0 to 40.0	717	628	635
Level 3	41.0 to 60.0	484	524	349
Level 4	61.0 to 80.0	70	63	99
Level 5	Greater than 80.0	0	0	5
		1,628	1,634	1,633



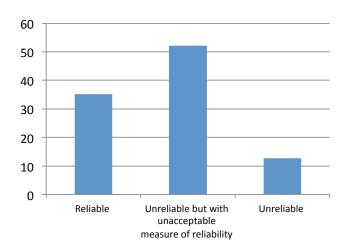




2012

# **III. Some Results**

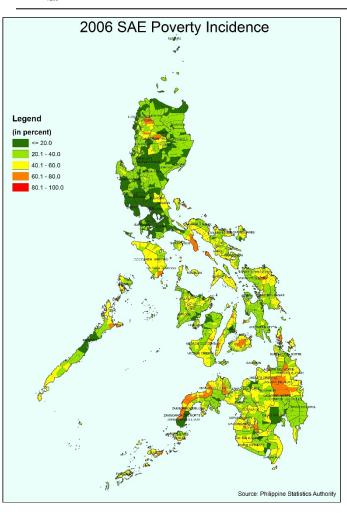
Type of Estimates	Coefficient of Variation (%)	Count	%	< RCF
Reliable	At most 10.0	574	35.2	35
Unreliable but with acceptable measure of reliability	10.1 to 20.0	851	52.1	87
Unreliable	Greater than 20.0	208	12.7	100

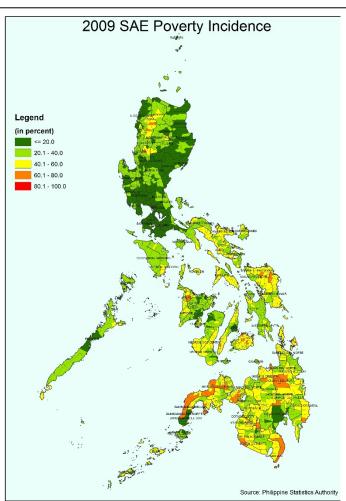


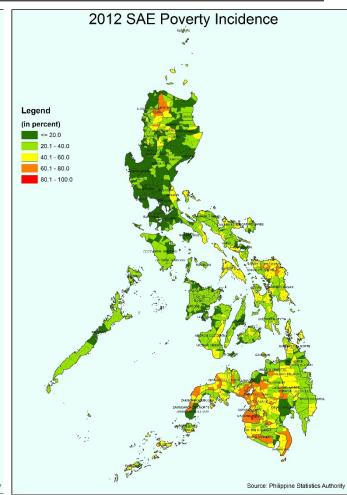
- Almost 87% of the resulting estimates are with acceptable measures of reliability. The rest are unreliable and should be used with much caution.
- Most reliable estimate is for the Municipality of Katipunan, Zamboanga del Norte with coefficient of variation of 3.3% and most unreliable is for the Municipality of Cainta in the Province of Rizal with coefficient as high as 84.8%.

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# **III. Some Results**







# IV. Actual Policy Uses

### A. In targeting beneficiaries of programs/projects

 Used the 2006 and 2009 SAE to identify the beneficiaries of Kalahi-CIDSS in Agusan del Norte in its implementation in 2013

### B. In policy formulation and planning

 Used as input in the BLISTT master planning activity (BLISTT stands for Baguio, La Trinidad, Itogon, Sablan, Tuba and Tublay in the Cordillera Autonomous Region);

### C. In poverty monitoring

 Used by the Pangasinan and La Union Provincial Government in the assessment of the progress of municipalities in their implementation of poverty reduction programs

- Adoption of an official methodology
- Study the use of other SAE techniques
- Production of infographic materials
- Generation of 2015 city and municipal level poverty statistics in 2017-2018
- Explore the use of the SAE technique in other variables
- Adoption of the 2013 Master Sample in our household surveys, with province as domain

- Explore the possibility of using "big data" as possible input in the monitoring of SDG indicators, addressing some of the data gaps, including the challenge of disaggregation
- Pilot on Batanes-Palawan Statistical Development Project 2016-2017

# Thank you! Maraming salamat po!

URL: http://www.psa.gov.ph