



THE INTER-SECRETARIAT
**WORKING GROUP ON
HOUSEHOLD SURVEYS**

Small area estimation for SDGs: from experiment to production

United Nations Statistics Division

 unstats.un.org

Outline

- ❑ Briefly: ISWGHS and IAEG-SDGs
- ❑ Why the UN Small Area Estimation (SAE) Toolkit?
- ❑ From experiment to production: challenges
- ❑ Lessons learnt from countries that did it successfully
- ❑ What is next?

The ISWGHS: a primer

- ❑ Established in 2015 under the aegis of the UNSC
- ❑ Objectives:
 - ❑ Improve coordination of household surveys
 - ❑ Advance cross-cutting survey methodology
 - ❑ Enhance communication and advocacy
- ❑ Governance
 - Membership: 11 international agencies + 8 (rotating) member states
 - Secretariat: UN Statistics Division
 - Current co-chairs: WB and UNW
- ❑ Work through time-bound Task Forces, led by and with contribution from members and non-member experts.

Inter-agency and Expert Group on Sustainable Development Goal Indicators (IAEG-SDGs)

The 2030 Agenda for Sustainable Development

- A global blueprint for people, planet, prosperity, peace and partnerships, now and in the future
- 17 Goals, 169 targets and “Leaving no one behind” principle



The IAEG-SDGs :

- Composed of 28 Member States (and representatives of regional commissions, regional and international agencies and CSOs are observers)
- Developed the global indicator framework for SDGs (**231 indicators**)

IAEG-SDGs workstream on data disaggregation:

- Compilation of existing guidelines and methodologies on data disaggregation
- Preparation of Handbook on data disaggregation for SDGs
- Task Force on Small Area Estimation (joint with ISWGHS)



Small area estimation: why and what is it?

❑SDGs: Leave no-one behind

- Sustainable Development Goal indicators should be disaggregated, where relevant, by income, sex, age, race, ethnicity, migratory status, disability and geographic location, or other characteristics, in accordance with the Fundamental Principles of Official Statistics.

❑Limitations of household surveys

- Cross-classification (i.e., geographic, demographic) usually leads to small sample sizes even for very large surveys, hence acceptable accuracy cannot be obtained through direct estimates

❑Small area estimation

- Small area: domain where sample size is too small for reliable direct estimation
- Small area estimation: Modeling, “borrow strength” from other data sources

Why SAE Toolkit (<https://unstats.un.org/wiki/display/SAE4SDG/>)

- ❑ Many countries have experimented with SAE in the past but very few were able to transform from experiment to official production. The Toolkit:
 - Finds out why this is happening?
 - Establishes a close link of SAE to SDG monitoring: SAE is very resource intensive therefore needs to be focused on a smaller number of indicators. The SDG indicators are therefore a perfect framework to start with if resource is limited.
 - Provides hands-on exercise, including “semi-synthetic” data (national data + noises) and programming guide. The data sets are used to illustrate, for 3 SDG indicators, on how to work with a complete cycle of SAE estimation such as data preparation, selection of methods, analysis and adaptation and evaluation/benchmarking.
 - Incorporates [national examples and case studies](#) through two angles: (a) documenting the lessons learnt and challenges of countries in using SAE for official data production; and (b) illustrating SAE practices for indicators under different SDG goals.
 - Includes a [long discussion](#), based on our discussion with countries, on the challenges and enabling environment for countries to move from SAE experiment to official production. We hope this discussion can help countries build appropriate measures, in addition to technical capacity, in using SAE for official production.
 - Provides an up-to-date and comprehensive list of SAE software packages in major languages (R/Stata/SAS/Python).

Guiding through steps with practical example

8.5.2 Unemployment rate

R Code

- > User needs
- > Data availability
- > Specification
- > Analysis & Adaptation

Evaluation & Benchmarking

To evaluate the domain indicators, the model is fitted and the MSE and the CV as measure for the uncertainty of the estimates are estimated. The estimation of the MSE and CV is triggered by setting the parameter MSE to "TRUE". For the transformed area-level model with bias-corrected backtransformation, a bootstrap MSE is provided. The parameter B controls the number of bootstrap iterations. It is advisable to set B to a minimum value of 100 in order to obtain reliable MSE estimates.

Precision, accuracy and reliability

> Expand source

The estimated regional indicators (the unemployment rate in this example) with its MSE and CV can be obtained in the form of a table. Generally, the CV should be used with caution when the indicator of interest is a ratio since really low point estimates can also be the reason for large CVs. In these cases, it is recommendable to focus on the MSE.

In this example, it can be seen that the CV of the model-based estimate (FH) is generally lower than for the direct estimate. However, there are also cases where the CV is slightly larger. One reason could be that the number of bootstrap iterations is too low.

MSE and CV per domain

> Expand source

The model-based estimates are commonly compared with the results of direct estimates. The function `compare_plot` in `emdi` provides some plots for this comparison.

Comparison with direct estimation

> Expand source

Comparing direct with model-based estimates helps to evaluate if the model-based estimates are more reliable than the direct estimates measured in terms of the MSE or the CV. The boxplots confirm that the model-based estimates have lower CVs overall. Approximately, 75% of the model-based domain estimates show a CV below 20%. It is also apparent that the increase in efficiency is not huge. Furthermore, the second plot shows that there are also domains where the CV of the model-based estimates is larger than the one of the direct counterpart.

The left plot is a boxplot titled 'FH' comparing the Coefficient of Variation (CV) for two methods: 'Direct' and 'Model-based'. The y-axis lists the methods, and the x-axis represents CV from 0.1 to 0.5. The 'Direct' method shows a higher median CV (around 0.25) and a wider interquartile range compared to the 'Model-based' method, which has a lower median CV (around 0.15) and a narrower interquartile range. The right plot is a scatter plot titled 'FH' showing the CV of 'Model-based' estimates on the y-axis against the CV of 'Direct' estimates on the x-axis. The x-axis is labeled 'Domain (ordered by increasing CV of Direct)'. The plot shows a strong positive correlation, with most points falling below the diagonal line (y=x), indicating that model-based estimates generally have lower CVs than direct estimates. A legend for both plots identifies 'Direct' as blue squares and 'Model-based' as grey circles.

When comparing the direct and model-based point estimates, it can be seen that these do not differ strongly from each other.

Case studies covering different SDG goals/indicators

Goal 1. End poverty in all its forms everywhere

> Case studies

Goal 2. End hunger, achieve food security and improved nutrition and promote sustainable agriculture

> Case studies

Goal 3. Ensure healthy lives and promote well-being for all at all ages

> Case studies

Goal 4. Ensure inclusive and equitable quality education and promote lifelong learning opportunities for all

No case studies yet

Goal 5. Achieve gender equality and empower all women and girls

No case studies yet

Goal 6. Ensure availability and sustainable management of water and sanitation for all

No case studies yet

Goal 7. Ensure access to affordable, reliable, sustainable and modern energy for all

> Case studies

Goal 8. Promote sustained, inclusive and sustainable economic growth, full and productive employment and decent work for all

SAE methodologies used by countries and international agencies

Dashboard / SAE4SDG   38 views

SAE practices

Created by Haoyi Chen, last modified on May 04, 2021

Asian Development Bank

FAO

UNICEF

US Census Bureau

Asian Development Bank

Created by Haoyi Chen, last modified by Arturo Jr M. Martinez on May 04, 2021

Brief introduction of the organisation

ADB is committed to achieving a prosperous, inclusive, resilient, and sustainable Asia and the Pacific, while sustaining its efforts to eradicate extreme poverty. Established in 1966, it is owned by 68 members—49 from the region and 19 from other parts of the world.

A description of the SAE work within the organisation

In 2017, the Asian Development Bank (ADB) launched the Data for Development project which aims to support the statistical capacity of national statistics offices (NSOs) in Asia and the Pacific, helping them comply with the Sustainable Development Goals (SDGs). This component focuses on strengthening the capacity of NSOs to generate fine-grained data for policies and programs targeted to vulnerable sectors of society.

One of the outputs of this component is a guide on disaggregation of official statistics, which includes an inventory of various small area estimation (SAE) methodologies to yield granular data for official statistics compilation. The guide explains SAE techniques with examples of how the easily accessible R analytical platform can be used to implement them, particularly to estimate indicators on poverty, employment, and health outcomes.


Reference:

- Asian Development Bank. Introduction to Small Area Estimation Techniques: A Practical Guide for National Statistics Offices

Future work on SAE

The guide compiles various SAE techniques and worked examples on how to implement the methodology, which were covered in a series of country training workshops provided to the staff of several national statistics offices. Furthermore, since its publication in May 2020, several researchers and academics have reported the usefulness of the guide in their work.

Moving forward, the team will continue exploring potential areas of collaboration with national statistical systems who may need technical assistance in building capacity on the application of SAE methods.

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US Census Bureau

Created by Haoyi Chen, last modified on May 04, 2021

Introduction

One of the most famous programmes on small area estimation for official statistics is the Small Area Income and Poverty Estimates (SAIPE) Program led by the US Census Bureau. SAIPE provides annual estimates of income and poverty for small areas across the United States. This discussion with the SAIPE team at the US Census Bureau as well as other reference materials.

How to motivate SAE - how did you convince the government to use small area estimates?

Answer: Prior to SAIPE, all local level income and poverty information can only be produced from the decennial census long-form. This means that small area estimates on poverty is only available every 10 years, based largely on "the number of children aged 5 to 17, inclusive, from families below the poverty level on the basis of the most recent satisfactory data, ..., available from the Department of Commerce." This law is the Department of Commerce, unless the Secretaries of Education and Commerce determine that the use of updated population data would be "inappropriate or unreliable." It also directs the Secretary of Education to determine if the use of updated population data would be "inappropriate or unreliable."

From the description above, three distinct features stand out:

1. A legal act is in place that requires that the Secretary of Education distribute Federal funds based on data produced at county and school district level, unless data are "inappropriate or unreliable".
2. The legal act also specifies that such data should be produced by the Department of Commerce that houses the US Census Bureau
3. Funding of an external expert panel to provide quality check

Therefore this is really a "top-down" approach where the law requires that quality data are to be used for policymaking, distributing Federal fund in this case. The program is well-funded because of the legislative

Input data

Surveys that provide poverty data: Current Population Survey (CPS) through 2004 and American Community Survey starting in 2005.

Administrative data:

- US Federal income tax data
- Supplemental Nutrition Assistance Program (SNAP) participants data
- Supplemental Security Income (SSI) reciprocity rate

Data from the Census Bureau Population Estimates Program are used to construct denominators of several of the regression covariates.

Source: An Overview of the US Census Bureau's Small Area Income and Poverty Estimates (SAIPE Program), Bell, Basel and Maples, 2015

Input data quality reflection

The quality of the input data is important. One administrative data that was considered but not used is the Free and Reduced-Price Lunch Data. Studies showed such data are not sufficiently precise for formal use in SAE. The reflection is on how household surveys could be better designed to allow good small area estimation. For example, CPS sample that collected poverty data are relatively small and for some small geographic areas.

Adjustment made on the model and estimates

Improvements of small area estimates are made over time by refining models and incorporating new or updated data sources. Since its inception SAIPE program has made many changes in its models and estimates.



Challenges in using SAE for official statistics

- Lack of interest and support from the top management, hence lack of resource
- Lack of dedicated resources for SAE research and implementation.
- While one household survey can produce a large number of indicators, great efforts are necessary for SAE to derive just one indicator for small domains
- Lack of in-house technical capacity
- Lack of proper input data (access to/poor quality of admin data source)
- Reluctance about the use of model-based estimates (vs. survey estimates that are design-based/model-assisted)
- Difficulties in communicating the technical aspects to users

Challenges in using SAE for official statistics (cont.)

- *"We did an experiment using small area estimation method for poverty but the results were not consistent with our own estimates so we did not pursue it again."*
- *"We do not have good input data source for SAE - census data are outdated, and administrative data sources do not have good coverage and lack proper auxiliary variables."*
- *"SAE method is complicated and we are not comfortable with independently developing the method."*
- *"It is very difficult to convince the managers to use model-based estimates."*
- *"Producing SAE requires a lengthy period of looking for input data, finding the right auxiliary variables, testing different models and their assumptions and validating the estimates."*

Source: UNSD conversations with NSOs

Lessons learnt: driven by needs for key policies and funding decisions

- ❑ *Colombian National Development Plan 2018-22 made it mandatory to redesign the national monetary transfer programs (Jóvenes en Acción and Familias en Acción), for population in poverty and in extreme poverty. This needs poverty data at municipal level. (Colombia)*
- ❑ *In 2009, the law of the Fondo Común Municipal (FCM) required the Ministry to provide poverty rate estimates every 2 years for all comunas in the country. Funding to all comunas will be allocated based on such data. (Chile)*
- ❑ *The 2005-2009 BPS Strategic Plan for Statistical Development defined “the development of an efficient and low-cost methodology, which allows for the creation of small area and local specific statistics data” as one of the main activities to support government decentralization (Indonesia)*
- ❑ *The Cabinet of the Government of Jamaica made a request for the Statistical Institute of Jamaica to use small-area estimation for poverty mapping, to produce poverty data for smaller geographical areas within the country. (Jamaica)*
- ❑ *Improving America’s Schools Act: “the number of children aged 5 to 17, inclusive, from families below the poverty level on the basis of the most recent satisfactory data, ..., available from the Department of Commerce” (US)*

Lessons learnt: access to good quality input data

- ❑ Access to auxiliary data sources (e.g., administrative data), regularly
- ❑ Input data are of good quality:
 - Coverage, accuracy and timeliness
 - Availability of auxiliary variables that have good prediction power for the outcome indicator

Table 20.5 Initial set of auxiliary variables reviewed for their possible inclusion as comuna level auxiliary variables in the area level model.

Name of the auxiliary variable	Institution responsible for data collection	Frequency of publication of the data
1. Subsidio Familiar	Unidad de Prestaciones Monetarias, Ministerio de Desarrollo Social	Monthly and yearly
2. Subsidio al Pago del Consumo de Agua Potable y Servicio de Alcantarillado de Aguas Servidas	Unidad de Prestaciones Monetarias, Ministerio de Desarrollo Social	Monthly and yearly
3. Bono Chile Solidario	Unidad de Prestaciones Monetarias, Ministerio de Desarrollo Social	Monthly and yearly
4. Subsidio de Discapacidad Mental	Unidad de Prestaciones Monetarias, Ministerio de Desarrollo Social	Monthly and yearly
5. Pensión Básica Solidaria (vejez e invalidez)	Unidad de Prestaciones Monetarias, Ministerio de Desarrollo Social	December
6. Aporte Previsional Solidario (vejez e invalidez)	Unidad de Prestaciones Monetarias, Ministerio de Desarrollo Social	December
7. Bonificación al Ingreso Ético	Unidad de Prestaciones Monetarias,	Monthly and yearly

Source: Example from Chile, Casas-Cordero, Encina and Lahiri (2016)

Lessons learnt: Input data in countries

☐ Chile:

- CASEN survey (cross-sectional multipurpose household survey)
- Comuna level administrative data

☐ Colombia:

- Integrated household survey (GEIH)
- Population census

☐ Indonesia:

- Indonesian National Socioeconomic Survey
- Village Potential Statistics (PODES)

☐ Jamaica:

- Jamaica Survey for Living Condition
- 2011 census

☐ US:

- American Community survey, Current Population Survey (Annual)
- Administrative data: Income tax; Supplemental Nutrition Assistance Program participants data; Supplemental Security Income Reciprocity rate

Lessons learnt: maintaining a high and fit-for-purpose quality standard

❑ Internal assessment to evaluate the models, the estimation procedure and corresponding results

- Compare with direct estimates, at national, urban/rural, principal cities and state level (Colombia)
- Coefficient of variation requirement (CV):
 - Colombia: CV < 30% for publishing
 - ISTAT: CV ≤15% for domains; CV ≤18% for small domains
 - Statistics Canada: CV ≤16.5% no release restriction; 16.5%<CV ≤33.3% add warnings; >33.3% not recommended for release

❑ External/independent evaluation:

- Public consultation: consultation is carried out with local government (Indonesia)
- Review by experts:
 - A National Academy of Sciences panel was funded to provide advise on the suitability of the Census Bureau estimates for use in allocating funds (United States)

Lessons learnt: effective capacity building

- ❑ Step 1: Organising broad training on SAE methods and why SAE outcomes are important to inform policy.
- ❑ Step 2: Providing technical training for staff working on SAE: covering basic foundations and introducing different methods and models.
- ❑ Step 3: customized hands-on training specific to the outcome indicator
 - A good understanding of data needs: outcome indicator and the level of disaggregation
 - Assessing input data availability/quality/timeliness/auxiliary variables
 - All exercises should be carried out by national staff, ideally also using country data

Consultations

- ❑ Key SAE experts: consultation meeting organized by JPSM Technical Group on SAE, May 2021
- ❑ Emails and focus-group discussions
 - Australia, Canada, Chile, Colombia, Indonesia, Italy, Jamaica, Republic of Moldova, Philippines, South Africa, US, UK, Viet Nam
- ❑ Next steps:
 - Continue to add case studies and national experiences
 - Provide training to countries through an eLearning course currently being developed by ECLAC-UNSD-UNFPA
 - Organise small technical group discussion (countries + academic) to address specific questions from countries
 - Explore potential of using non-traditional data sources such as remote sensing and mobile phone data for SAE

Thank you

Haoyi Chen
chen9@un.org

