

# Small Area Estimation: Part I

**Partha Lahiri**

**JPSM, Univ. of Maryland,  
College Park, USA**

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**Definition:** A subpopulation of interest with meager or no survey data.

**Examples:**

- In a nationwide survey, cells obtained by finer classification of age-group, race, gender even at the national level (small domains).

## Examples of Small Areas or Domains: Contd.

- **US NCHS: Estimation of health variables using the NHANSE III - a majority of US states (small areas) do not have sample**
- **US Census Bureau: Poverty estimation for US counties and school districts using the American Community Survey**
- **NASS-USDA: Estimating crop acres, production and yields for counties**

- **A convenient way to display spatial variations of different socio-economic and health related estimates**
  - **Disease mapping**
  - **Poverty Mapping**
- **Reliable maps are useful to public policymakers in planning intervention and allocation of government resources.**

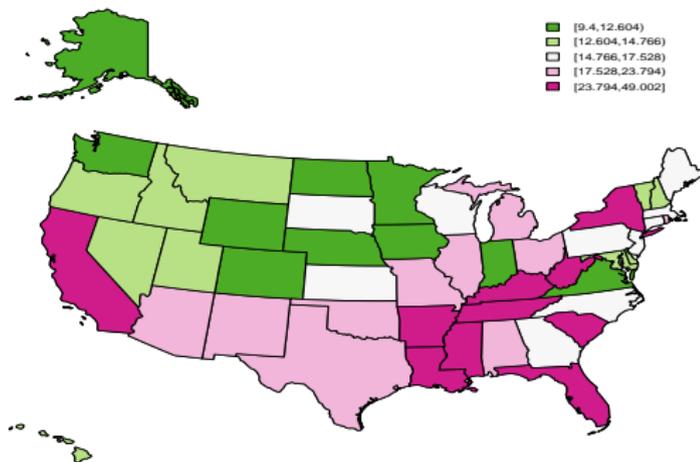
**Example:** Estimation of poverty rates for over 300 comunas in Chile is of great interest to Chilean government.

**Main data source:** CASEN, a multipurpose sample survey targeting the civilian non-institutionalized population that resides in housing units throughout the Chilean territory.

**Design-based direct estimates:** survey weighted proportions that gives differential weights to individuals depending of their inclusion probability into the sample.

**Caution:** The direct estimators are highly unreliable due to small sample sizes in the areas. They have high variability and could be highly biased, depending on the situation.

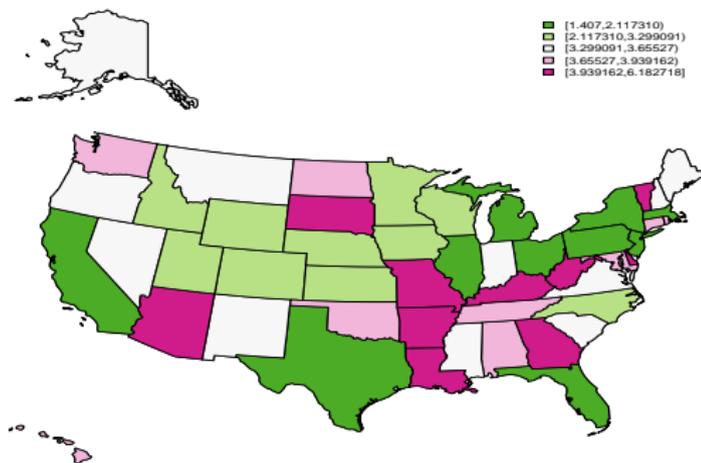
### SAIPE 93' Direct Estimate of Poverty



## Map of Direct Survey Estimates of Poverty

### Rates

### SAIPE 93' sqrt(Di) of Poverty



## Map of Standard Error Estimates of Direct Survey estimates of Poverty

- **11th century England and 17th century Canada - based on census or administrative records**
- **There is an increasing demand for small area statistics, due to growing use in formulating policies and programs in the allocation of government funds and in regional planning**

- **Stratification - Use a large number of smaller strata**
- **Degree of Clustering - Minimize clustering**
- **Sample Allocation - Reallocate sample from large planned domains to smaller planned domains**
- **Rolling samples (ACS), multiple frames**

**Reallocates sample from larger planned domains to smaller planned domains.**

- **Small reduction in sample size for large domains usually has little effect.**
- **Small increases in small domains may have a large effect on reliability.**

**Two-Step Allocation:** 42,000 Households for national and province level estimates, 17,000 for UIR (SAE) level estimates.

### Effects of Reallocation on Areas:

- **Canada:**  $E(CV)$  from 1.3 to 1.5
- **Ontario:**  $E(CV)$  from 2.8 to 3.2
- **UI region:**  $E(CV)$  from 17.7 to 9.4

**”..the client will always require more than specified at the design stage” (Fuller, 1999)**

### **Relevant Source of Information**

- **Census/Administrative information**
- **Related surveys**

### **Method of Combining Information**

- **Choices of good small area models**
- **Use of a good statistical methodology**

**Estimate the median number of radio stations heard during the day for over 500 counties of the USA (small areas).**

**Ref: Hansen et al. (1953)**

**Two different survey data used**

## Mail Survey

- large sample (1000 families/county) from an incomplete list frame
- response rate was low (about 20%)
- estimates  $x_i$  are biased due to non-response and incomplete coverage

## Personal Interview Survey

- stratified multi-stage area frame
- Nonresponse and coverage error properties were better than the mail survey
- reliable estimates  $y_i$  for the 85 sampled counties were available, but no estimate can be produced for the remaining 415 counties

- Using  $(y_i, x_i)$  for the 85 sampled counties, the following fitted line (synthetic estimator) was obtained:

$$\hat{Y}_i^{Syn} = 0.52 + 0.74x_i$$

- Use  $y_i$  for the 85 sampled counties and  $\hat{y}_i$  for the rest.

## Synthetic Estimation: Implicit Modeling

- $N_{ig}$  = Female population size for the  $g$ th race x age-group for the  $i$ th state. Data source: hospital registration system.
- $p_{.g}$  = national level direct estimate of the proportion of jaundiced infants whose mother is in the  $g$ th group. Data source: 1980 National Natality Survey.

# Synthetic Estimation: Implicit Model

| Subgroup  |          | $N_{i.g}$ | $p.g$  | $N_{i.g}p.g$ |
|-----------|----------|-----------|--------|--------------|
| White     | Under 20 | 16382     | 0.216  | 3539         |
|           | 20-24    | 44100     | 0.214  | 9437         |
|           | 25-29    | 46421     | 0.222  | 10305        |
|           | 30-34    | 22400     | 0.224  | 5018         |
|           | 35+      | 5896      | 0.244  | 1439         |
| All Other | Under 20 | 5493      | 0.173  | 950          |
|           | 20-24    | 7657      | 0.167  | 1279         |
|           | 25-29    | 5063      | 0.19   | 962          |
|           | 30+      | 3387      | 0.266  | 901          |
|           |          |           | 156799 |              |

- A synthetic estimate of the percentage of jaundiced infants in Pennsylvania:  $p_i^s = \frac{33830}{156799} * 100 = 21.6\%$ .
- Estimate of total number of jaundiced infants in Pennsylvania =  $N_i.p_i^s = 156,799 \times 0.216 = 33,869$ .

- **Flexible**
- **Borrows strength from different relevant sources**
- **Uses appropriate multi-level model that captures different sources of variations**
- **Improves on both direct and synthetic methods**

# Example: U.S. Small Area Income and Poverty Estimates (SAIPE) Program

**Parameters of interest:** true proportions of 5-17 year old children in poverty for the fifty states and the District of Columbia.

**Direct estimator:** The survey-weighted proportions are obtained using the American Community Survey (ACS) data.

## Auxiliary Variables

- **proportion of child exemptions reported by families in poverty on tax returns**
- **proportion of people under age 65 not included in an income tax return**
- **proportion of people receiving food stamps**

## Two-Level Model

- **Level 1:** Describes the sampling distribution of ACS survey-weighted poverty rates for the states
- **Level 2:** Links the true state poverty rates to state level auxiliary variables

**Estimation Method:** Empirical/Hierarchical

Bayes

## Example: County Level Estimation of Crop Yield

- **Yield=production/harvested acreage**
- **Data from multiple surveys are pooled for county estimation**
- **Survey weights are not available. Direct estimates are based on county specific data using a simple county specific regression model.**

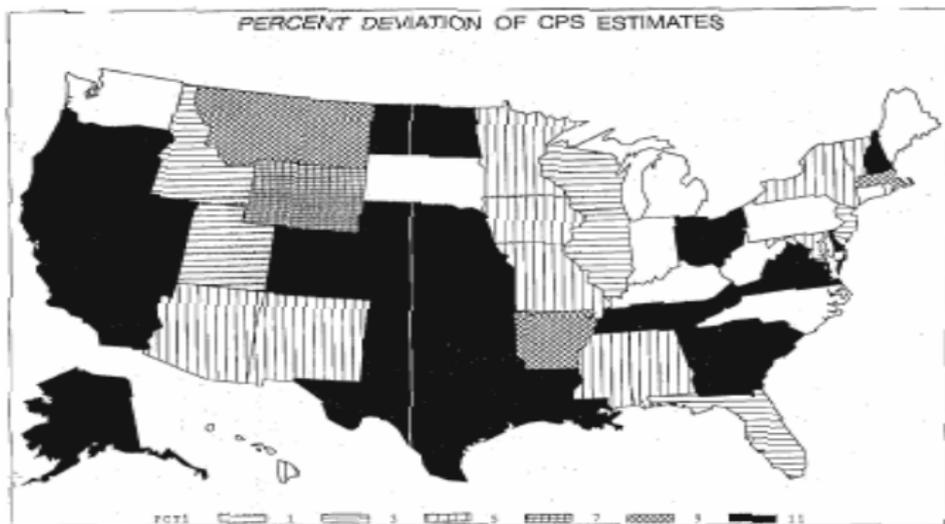
## Example: County Level Estimation of Crop Yield

- **Empirical Bayes estimates use pooled survey data plus county level administrative and satellite data**
- **Evaluation criteria (AAD, etc.) are computed by measuring different distances from the census yield and then averaging over all the counties in the state. Smaller the better.**

# Example: Estimation of Median Income of Four-Person Families for the states

- **Primary Data: Current Population Survey (CPS)**
- **Auxiliary data: administrative/census data**
- **Model: Cross-sectional and time series multi-level model**
- **Evaluation: Map of relative errors:**

$$|est - census|/census.$$



Map of Relative Errors of Direct Survey Estimates of Median

Income of 4-Person Families

