Methodological Considerations for Disaggregation

Panel 2: Counting the Uncounted

Global Preparatory Seminar for the United Nations World Data Forum

September 7, 2016

My perspective

- Academic demographer
- Work on questions in demography and epidemiology in Africa
- Collaborate closely with academic statisticians developing new statistical methods for population and health data and research

Disaggregation

- Quick review of why disaggregation is hard
- The number of unique 'cells' in a dataset is the product of the number of effective categories defined by each variable:

 $cells = categories_{v1} \times categories_{v2} \times \cdots \times categories_{vn}$

- Adding either new variables or new categories for existing variables can *greatly* increases the number of cells
- To produce useful population-level measures, each cell must have a reasonable number of observations
- Consequently, the number of observations required is (very) large and must include a wide variety of subjects

Possible ways to disaggregate

There are at least four general approaches to address the challenges of disaggregation

- 1. Collect lots of data that includes many subjects in each cell
 - This is logistically difficult and complex
 - Very expensive
 - Produces measures rather than estimates
- 2. Smoothing and interpolation
 - Can incorporate data from many different sources to infer reasonable values for cells with missing data
 - Can operate on many dimensions of the data simultaneously and account for uncertainty well
 - Produces estimated values, not raw measures

Possible ways to disaggregate

- 3. Use models to generate/estimate values
 - These incorporate independent, external knowledge of the processes generating the data
 - The resulting estimates are a hybrid of data and our understanding of how the data are generated

4. Borrow data from a similar setting

- When direct measures are not possible, use what we know about a similar setting
- The result is a contextualized version of information from elsewhere
- 5. All of the above

Thoughts on the way forward

1. Short term: use what we have better

- Utilize all alternative approaches to disaggregation that do not require large amounts of new data
- Key requirements: more people trained in required methods and better availability of existing data
- 2. Medium term: selectively invest in collecting new data where it is most effective in improving estimates
 - Key requirements: understanding the importance/influence of different data and data sources and focused investment in the most useful of those
- 3. Long term: collect a lot more data

Example: child mortality in Tanzania

 Goal: national and small-area estimates of child mortality (U5MR) for past several decades using as much of available data as possible

Data:

- 1. All demographic and health surveys for Tanzania: DHS household sample surveys
- Two demographic and health surveillance system sites: HDSS - intensive surveillance of small, geographically limited populations
- These are very different data sources with completely different designs

Example: child mortality in Tanzania

Methods:

- Survival analysis of child mortality in all possible times and places using all data sources accounting for data design - sampling, etc.
- Space-time smoothing model to integrate/interpolate/smooth all estimates so they are consistent with one another

Results:

- Consistent time trends in child mortality at both national and subnational levels
- Consistent uncertainty/precision in all estimates
- Possible to disaggregate further, e.g. by age and socioeconomic status, within the same framework, possibly using models and limited information from similar populations