Relevance or Irrelevance of Weights for Confidentiality and Statistical Analyses

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Outline

- What are weights for and where do they come from?
- To weight or not to weight: That is the question?
 - Issue for analysis of released microdata.
- What hazards do weights pose for confidentiality protection?
 - How can problems be averted or mitigated?
- Prescriptions for statisticians?

Survey Analysis

- The core of official statistics activities in many countries.
- Frequentist motivation and analyses.
 - "Weighted" analyses are central to methodology for population estimates.
 - Debate over design- vs. model-based estimation.
- Statistical models are now commonplace in statistical agencies today.

Sampling Weights

• Weights arise in sampling settings when units are selected with unequal probabilities from a finite population: $w_i = \frac{1}{w_i}$.

 p_i

- But in typical official statistics setting weight are typically product of 3 components:
- $w_i = \frac{1}{p_i} \times (\text{nonresponse}) \times (\text{poststratification})$ $p_i \text{ Model based: MAR Population controls}$

To Weight or Not To Weight?

- No debate about weight, at least for the probability of selection, about estimating population aggregates.
- Issue comes when we introduce statistical models:
 - e.g., weighted vs. unweighted regression models.
 - How do we combine models and sampling theory?
 - Naïve weighting
 - Estimating equations

Arguments for Weighting

- Almost always justified only from frequentist, design-based perspective:
 - Correct frequentist properties under finite population setup for aggregates.
 - Pays attention to sample space in evaluating contributions to estimation of model parameters.
 - Robustness.

Hansen, et al (1983); Kalton; Pfeffermann (1988, 1996), Pfefferman et al. (1998).

Arguments Against Use of Weights

- Stratification and clustering belong in model.
 - Weight components are insufficient for this purpose.
- Weights have non-sampling components, due to adjustment for *nonresponse*, and *post-stratification*.
 - Why are we incorporating these components into this way in likelihood?
- Estimating equation ideas collapse for complex hierarchically-structured models with latent variables.

Disability Among Elderly

• National Long Term Care Survey of US Medicare eligible population (aged 65+).

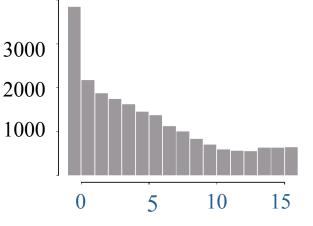
- 6 waves: 1982, 1984, 1989, 1994, 2004.

- Models for disability based on 16 binary measures called activities of daily living (ADL) and instrumental activities of daily living (IADL):
 - eating, getting in/out of bed, getting around inside, dressing, bathing, using a toilet, doing heavy house work, doing light house work, doing laundry, cooking, grocery shopping, getting about outside, traveling, managing money, taking medicine, telephoning.

Disability Among Elderly—II

• Data on *N*=21,574 individuals:

- 65,536 cells (3,152 non-zero):
 - 82% of cell counts < 5.
 - -4% of cell counts > 20.
 - 18% with no disabilities.
 - 3% with all 16.



- Bayesian hierarchical mixed membership models with informative latent structure:
 - Smoothing is crucial feature. Levels of hierarchy matter, as do choices of Dirichlet prior parameters.
 - Erosheva, Fienberg, Joutard (2007, AOAS)

Grade of Membership Model

- N individuals; K extreme profiles; J=16 items.
- Membership scores $\lambda = (\lambda_1, ..., \lambda_K)$ define how close individual is to each extreme profile.
- Probability distribution of response *j*, given full membership in extreme profile *k*, is

$$f(\mathbf{x}_{ij} \mid \boldsymbol{\lambda}_{ik} = \mathbf{1}; \ \boldsymbol{\theta}_{kj}) = Binomial(\boldsymbol{\theta}_{kj}),$$
$$Pr(\mathbf{x}_{j} \mid \boldsymbol{\lambda}) = \sum_{k} \boldsymbol{\lambda}_{k} \cdot f(\mathbf{x}_{j} \mid \boldsymbol{\theta}_{k}).$$

- Membership scores are random, i.e., $\lambda \sim D_{\alpha}$. - Weighted analyses fail here!

O'Bayes Rome 2007

Usual Privacy Questions

- How do we protect the privacy of the responses represented by 2¹⁶ contingency table?
- What about the sampling variability in the sample design?
- What about the model variability? And the variability associated with model selection?
 - Cynthia and Adam's issue but not addressed here.

Policy Issue and Full Question

- Has disability been declining over time?
 - Implications for Social Security and Medicare planning.
 - How should we examine this question given the longitudinal nature of the data?
 - Role of sampling weights?
 - Privacy questions now relate to
 - Protecting privacy of $(2^{16})^6$ cell contingency table
 - Protecting privacy of weights

Confidentiality and Weights

- At last!
- Question 1: What information do weights provide to intruder intent on identifying individuals in the sample?
- **Question 2:** Do they increase the probability of disclosures? How?
- Ton de Waal and Leon Willenborg (1997) J. Off. Statistics, 13.

de Waal and Willenborg

- Sampling weights can provide indirect identifying information regarding membership in substrata defined by sets of post-stratification variables.
- Idea is that intruder will have accurate information on post-stratification population counts, and can use numbers of sample people with given weights to *match* individuals with post-strata.
- How big is the problem? Is it different in Netherlands compared with the U.S.?

De Wall/Willenborg Prescription

- Subsampling
 - To reduce probability of correct matches
- Noise addition to weights
 - Ditto
 - Also messes up original notation of population controls!

Rubin on Multiple Imputation

- Compute relevant posterior distribution and generate multiple samples from it.
- These are synthetic samples and thus (he argues) they automatically solve the confidentiality problem.
- You might use weights to construct the posterior (although many of us wouldn't) but there is no requirement that the multiply imputed data be weighted!

My Prescription

- Get rid of population controls and thus the biggest part of the confidentiality concerns from sampling weights.
- Stop insisting that model-based analyses incorporate weights.
- Think about new approaches to survey design that deal *de novo* with confidentiality concerns as well as analytical goals; not just traditional sample efficiency goals.
 - Share real design information.
 - Address disclosure problems with cluster sampling at design stage.

Summary

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- Prescriptions for statisticians
 - De Waal and Willenborg
 - Rubin
 - Fienberg