## **Outline for Discussion at NCES Roundtable on Imputation**

- 1. Brief discussion of Bamberg (2011) presentation
  - Types of applications for multiple imputation
    - Traditional (will come back to NHANES DXA imputation later)
    - Bridging (and other types of combining information)
      - Note uncongeniality issues reported in Rubin and Schenker (1987, JOS)
    - o Measurement error

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- Topics for future research
  - o Flexible models and methods
  - Diagnostics for imputation models
  - "Portability" of bridging models when the two surveys have different contexts 0
  - o "Uncongeniality" between imputation model and analysis model
  - o Methods for reflecting complex sample designs in imputation models
- 2. "Hot-deck imputation versus multiple imputation"
  - Not really the issue, because multiple hot-deck imputation possible
    - To reflect variability more fully, draw bootstrap sample from complete data before 0 creating each set of imputations
      - Rubin and Schenker (1986, JASA; 1991, Statistics in Medicine))
  - Two big issues
    - Single imputation versus multiple imputation
    - o Hot-deck versus explicit-model-based imputation
  - Hot-deck versus explicit-model-based imputation
    - Hot deck

- Imputes values that have actually occurred
- Less parametric flavor => possible robustness
  - See Schenker and Taylor (1996, Computational Statistics and Data Analysis) •
- Explicit-model-based 0
  - Easier to explain the model
  - Handles general patterns of missing data better
    - Can include more variables as predictors (e.g., by omitting high-order interactions)
      - Can improve prediction and make missingness at random more plausible
- 3. Some issues of interest for NHES imputation
  - Single imputation versus multiple imputation -
    - So far, differences in variance estimates not major (note low item nonresponse rates)
    - See if there are classes of analyses for which differences are larger
  - Possible advantages of explicit-model-based imputation over hot-deck imputation
    - Handles general patterns of missing data better 0
      - Predictors (analogous to "boundary variables") can have missingness
      - Note that "random imputation" (used for "boundary variables") probably okay for marginal distributions, but may attenuate multivariate analyses
    - Can include more variables as predictors 0
      - Could reduce bias and decrease variance
      - No need to worry about number of donors in cells
      - Note that there is a bias/variance trade-off associated with number of donors, collapsing cells etc. (see Schenker and Taylor 1996 for some relevant work)
  - Effects of manual imputation and post-imputation edits
    - Any attenuation of the positive effects of the prior imputation?