## To Laura

**To My Parents and Grandparents** 

To All of My Teachers

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#### List of Abbreviations

ACASI Audio Computer Assisted Self Interviewing

ANES American National Election Study
BGPS British Gambling Prevalence Survey
CAPI Computer Assisted Personal Interviewing

CHI Contact History Instrument

CI Confidence Interval

CIA Central Intelligence Agency

EBLUP Empirical Best Linear Unbiased Predictor

EHCS English House Condition Survey

EM Expectation-Maximization ESS European Social Survey

FR Field Researcher

FRQ Field Researcher Questionnaire

GDR Gross Difference Rate

HH Household

HLM Hierarchical Linear Modeling HSE Health Survey for England IPT Interpersonal Perception Task IRB Institutional Review Board

IWER Interviewer

LRT Likelihood Ratio Test MAR Missing at Random

MCAR Missing Completely at Random

MC-SIMEX Misclassification Simulation-Extrapolation

MI Multiple Imputation

NHANES National Health and Nutrition Examination Survey

NHEFS NHANES Epidemiologic Follow-Up Study

NHIS National Health Interview Survey

NSDUH National Survey of Drug Use and Health NSFG National Survey of Family Growth

NSR Non-Self-Representing

PASS Labor Market and Social Security (German)

PMM Pattern Mixture Model

PPMA Proxy Pattern Mixture Analysis
PQL Penalized Quasi Likelihood
PSU Primary Sampling Unit
RDD Random Digit Dialing
RMSE Root Mean Squared Error
SCF Survey of Consumer Finances

Self-Representing Total Survey Error United Kingdom United States Department of Agriculture Zip Code Tabulation Area SR TSE UK

USDA

**ZCTA** 

#### **ABSTRACT**

Interviewer observations are an important source of auxiliary information in survey research. Interviewers can record observations for all units in a sample, and selected observations may be associated with both key survey variables and response propensity. Survey statisticians use auxiliary variables with these properties to compute post-survey nonresponse adjustments to survey estimates that reduce both bias and variance in the estimates engendered by nonresponse. Unfortunately, interviewer observations are typically judgments and estimates, making them prone to error. To date, no studies have considered the implications of these errors for the effectiveness of nonresponse adjustments, effective observational strategies leading to reduced error rates, predictors of observation accuracy in face-to-face surveys, or alternative estimation methods for mitigating the effects of the errors on estimates. This dissertation presents results from three research studies designed to fill these important gaps in the existing literature.

The first study 1) analyzes the error properties of two interviewer observations collected in the National Survey of Family Growth (NSFG), finding accuracy rates ranging from 72-78% and evidence of systematic errors; 2) examines the effectiveness of nonresponse adjustments based in part on the observations, finding evidence of associations with key NSFG variables and response propensity but only slight shifts in estimates; and 3)

simulates the implications of errors in the observations for the effectiveness of weighting class adjustments for nonresponse, finding that adjustments based on the error-prone observations attenuate possible reductions in bias. The second study uses multilevel modeling techniques to identify several respondent- and interviewer-level predictors of accuracy in the two NSFG observations, including those supported by social psychological theories of what leads to improved judgment accuracy. The third study develops pattern-mixture model (PMM) estimators of means for the case when an auxiliary variable is error-prone, true values for the variable are collected from survey respondents, and the true values are predictive of unit nonresponse under a non-ignorable missing data mechanism. Simulation studies show that the PMM estimators have several favorable properties in these situations relative to other popular estimators, and R code is provided implementing the PMM approaches.