Supplemental Online Material

A Comparison of Multilevel Imputation Schemes for Random Coefficient Models:

Fully Conditional Specification and Joint Model Imputation with Random Covariance Matrices

Craig K. Enders¹, Timothy Hayes², and Han Du¹

¹University of California – Los Angeles

²Florida International University

- A. Supplement A: jomo Syntax for Joint Model Imputation
- B. Supplement B: Blimp Syntax for Fully Conditional Specification
- C. Supplement C: R Syntax for Analysis and Pooling Steps
- D. Supplement D: Trellis Plots of Relative Bias

Uncorrelated Random Effects, Auxiliary as the Cause of Missingness

- E. Supplement E: Trellis Plots of Relative BiasCorrelated Random Effects, Auxiliary as the Cause of Missingness
- F. Supplement F: Trellis Plots of Relative BiasUncorrelated Random Effects, Outcome as the Cause of Missingness
- G. Supplement G: Trellis Plots of Relative BiasCorrelated Random Effects, Outcome as the Cause of Missingness
- H. Supplement H: Trellis Plots of Mean Squared Error Ratios
- I. Supplement I: Blimp Syntax for Substantive Model-Compatible Imputation
- J. Supplement J: Trellis Plots of Relative Bias for Substantive Model-Compatible Imputation

Supplement A: jomo Syntax for Joint Model Imputation

```
# load packages
library(jomo)
# read raw data
rawdata <- read.table("~/desktop/example/math.csv", sep = ",")</pre>
names(rawdata) = c("school", "stanmath", "pctminor", "math", "mathse",
   "ability", "teachexp", "condition")
rawdata[rawdata == 999] <- NA</pre>
rawdata$icept <- 1</pre>
# imputation
set.seed(90291)
llmiss <- c("math", "mathse")</pre>
l2miss <- c("teachexp")</pre>
l2comp <- c("icept", "condition")</pre>
impdata <- jomo(rawdata[l1miss], Y2 = rawdata[l2miss],</pre>
   X2 = rawdata[l2comp], clus = rawdata$school, nburn = 2000,
   nbetween = 2000, nimp = 20, meth = "random")
# write data
impdata <- impdata[impdata$Imputation != 0,]</pre>
keepvars <- c("Imputation", "clus", "math", "mathse", "teachexp",</pre>
   "condition")
write.table(impdata[keepvars],"~/desktop/example/jomoimputations.csv",
   sep = ",", col.names = F, row.names = F)
```

Supplement B: Blimp Syntax for Fully Conditional Specification

```
DATA: ~/desktop/example/math.csv;
VARIABLES: school stanmath pctminor math mathse ability
    teachexp condition;
ORDINAL: ;
NOMINAL: ;
MISSING: 999;
MODEL: school ~ math:mathse teachexp condition;
NIMPS: 20;
THIN: 2000;
BURN: 2000;
SEED: 90291;
OUTFILE: ~/desktop/example/blimpimputations.csv;
OPTIONS: stacked;
```

Supplement C: R Syntax for Analysis and Pooling Steps

```
# Required packages
library(mitml)
library(lme4)
```

Read blimp data
filepath <- "~/desktop/example/blimpimputations.csv"
impdata <- read.csv(filepath, header = F)
names(impdata) <- c("imputation", "school", "stanmath", "pctminor",
"math", "mathse", "ability", "teachexp", "condition")</pre>

```
# Analyze data and pool estimates
model1 <- "math ~ mathse + teachexp + condition + (mathse|school)"
implist <- as.mitml.list(split(impdata, impdata$imputation))
randomslope <- with(implist, lmer(model1, REML = F))
estimates <- testEstimates(randomslope, var.comp = T, df.com = NULL)
estimates
```

```
# Compare models with likelihood ratio test
model0 <- "math ~ mathse + teachexp + condition + (1|school)"
ranintercept <- with(implist, lmer(model0, REML = F))
testModels(randomslope, ranintercept, method = "D3")</pre>
```

```
# Read jomo data
filepath <- "~/desktop/example/jomoimputations.csv"
impdata <- read.csv(filepath, header = F)
names(impdata) <- c("imputation", "school", "math", "mathse",
"teachexp", "condition")</pre>
```

```
# Analyze data and pool estimates
model1 <- "math ~ mathse + teachexp + condition + (mathse|school)"
implist <- as.mitml.list(split(impdata, impdata$imputation))
randomslope <- with(implist, lmer(model1, REML = F))</pre>
```

estimates <- testEstimates(randomslope, var.comp = T, df.com = NULL)
estimates</pre>

Compare models with likelihood ratio test
model0 <- "math ~ mathse + teachexp + condition + (1|school)"
ranintercept <- with(implist, lmer(model0, REML = F))
testModels(randomslope, ranintercept, method = "D3")</pre>

Supplement D: Trellis Plots of Relative Bias

Uncorrelated Random Effects, Auxiliary as the Cause of Missingness













Supplement E: Trellis Plots of Relative Bias

Correlated Random Effects, Auxiliary as the Cause of Missingness













Supplement F: Trellis Plots of Relative Bias

Uncorrelated Random Effects, Outcome as the Cause of Missingness













Supplement G: Trellis Plots of Relative Bias

Correlated Random Effects, Outcome as the Cause of Missingness













Supplement H: Trellis Plots of Mean Squared Error Ratios





IVISSING Data Rate \bullet 10% IVISSING \rightharpoonup 20% IVISSING \mp 30% IVISSING



IVISSING Data Rate \bullet 10% IVISSING \rightharpoonup 20% IVISSING \mp 30% IVISSING



Supplement I: Blimp Syntax for Substantive Model-Compatible Imputation

The syntax below applies substantive model-compatible imputation from Enders et al. (2018) to an artificial data set from the simulations reported in the body of the paper. Additional information about Blimp code is available in the user's guide (Keller & Enders, 2018).

DATA: ~/desktop/data.csv;

VARNAMES: id a1 a2 y x w;

! specify random slope with y:x;

MODEL: id ~ a1 a2 y:x w;

! specify outcome variable from the substantive analysis;

OUTCOME: y;

BURN: 2500;

THIN: 2500;

NIMPS: 20;

MISSING: 999;

SEED: 90291;

OUTFILE: ~/desktop/imp*.csv;

OPTIONS: separate;

Supplement J: Simulation Results Evaluating Substantive Model-Compatible Imputation

To provide a preliminary evaluation of substantive model-compatible fully conditional specification (SMC-FCS), we applied the procedure to the subset of simulation conditions with an ICC = .50 and correlated random effects. The trellis plots display relative bias values for this new procedure as well as the conventional imputation approaches evaluated in the manuscript. Additional details on SMC-FCS can be found in Enders et al. (2018).

Supplemental Figure 1. Average relative bias values for design cells with ICC = .50 and 15 clusters. Relative bias is defined as the difference between an average estimate and the true value expressed as a proportion of the true value. The dashed lines represent bias values of ± 0.10 .



• Complete \Box Joint Model + FCS × SMC-FCS

Supplemental Figure 2. Average relative bias values for design cells with ICC = .50 and 30 clusters. Relative bias is defined as the difference between an average estimate and the true value expressed as a proportion of the true value. The dashed lines represent bias values of ± 0.10 .



• Complete \Box Joint Model + FCS \times SMC-FCS

Supplemental Figure 3. Average relative bias values for design cells with ICC = .50 and 100 clusters. Relative bias is defined as the difference between an average estimate and the true value expressed as a proportion of the true value. The dashed lines represent bias values of ± 0.10 .



• Complete \Box Joint Model + FCS \times SMC-FCS