

# Other statistical software for continuous longitudinal endpoints:

## SAS macros for multiple imputation

James Roger

London School of Hygiene & Tropical Medicine, London.

Joint EFSPI and BBS virtual event, 8 December 2022.

Addressing intercurrent events: Treatment policy and hypothetical strategies.

## Setting the scene

- Same background as previous slides by Marcel Wolbers & Alessandro Noci.

I will concentrate on:

- The history before using “jackknife with conditional mean imputation” was proposed.  
An open discussion is needed on suitability of classic type 1 error as measure of “confidence”.
- A series of SAS macros to carry out these Multiple Imputations available from the DIA working Party TAB on the LSHTM Missing Data web site.

<https://www.lshtm.ac.uk/research/centres-projects-groups/missing-data>

# Emergence of Reference-based imputation 1

- 200x** Importance of Missing not at Random (MNAR) in data from clinical trials.  
As evidenced by EMA Guideline on Missing Data in Confirmatory Clinical Trials (Agreed by Working Party, April 2010)
- 2008** Carpenter & Kenward: Monologue; Missing data in clinical trials — a practical guide.  
Sensitivity analysis to MAR using Delta.

# Emergence of Reference-based imputation 1

2008 Roger(PSI presentation):

Sensitivity analysis based on “ITT-like imputation”.

Start of Reference-based imputation using MI as a Sensitivity analysis.

- ▶ The **%MIWithD SAS macro** (Multiple Imputation to handle withdrawals).
- ▶ Based on MCMC statement in proc MI.

2012 The **GSK 5 macros** for Reference based Imputation (PSI conference).

Based on very early version of proc MCMC.

- ▶ A flexible system for Reference-based imputation.
- ▶ GSK willing to make code openly available and support it.  
[James Roger, Carly Donovan & Tom Drury]

## Early references

**2013** Carpenter, Roger & Kenward: Analysis of longitudinal trials with protocol deviation: a framework for relevant, accessible assumptions, and inference via multiple imputation.

Jump-to-Reference (J2R), Copy-increment-from-Reference (CIR) etc. along with a marginal RM model (unstructured covariance by treatment arm).

**2014** Ayele, Lipkovich, Molenberghs, Mallinckrodt: A multiple-imputation-based approach to sensitivity analyses and effectiveness assessments in longitudinal clinical trials.

Placebo Multiple Imputation (pMI) using MONOTONE statement in Proc MI for conditional RM model.

Named Copy-Reference (CR) by Carpenter et al based on marginal model.

# Two algorithms for MI with continuous outcome measured repeatedly

- Stepwise  
Visit-by-Visit, each univariate Outcome is regressed conditional upon previous data.
- Marginal Multivariate  
A full multivariate Normal Repeated Measures model is fitted.  
Allows more complex models.

## Conditional (Stepwise) algorithm

- The MONOTONE statement in proc MI is a classic example.
- If the baseline covariates remain completely unchanged from regression to regression this is equivalent to the marginal RM model (unstructured covariance) with these baseline covariates crossed with time.
- Regressing on previous residuals rather than previous observed values gives correct marginal model, as in J2R and CIR, despite covariates changing with time.
- The **%MIStep and %MIStepWrap SAS macros** allow sequential fitting based on previous residuals.

# %MIStep and %MIStepWrap SAS macros and Off-treatment modelling

- Written to confirm the equivalence of regressing on residuals to the marginal model.
- Generates imputed data set for reference-based, or off-treatment models.
- Roger 2017 PSI conference poster, “Joint modelling of On-treatment and Off-treatment data”.
- Polverejan & Dragalin (2019) use the macro to impute using off-treatment data.



# Markov Chain Monte Carlo (MCMC) and Multiple Imputation

- When fitting a Bayesian model, random missing outcomes can be treated as if they are Bayesian parameters.
- When the imputation model matches the estimation model, a sample from the posterior can be used for imputation.
- For J2R, CIR etc. (but not CR) or off-treatment models, imputed values can be generated directly.
- For reference-based imputation fix the design matrix for the missed values (e.g. switch to reference).
- For Off-Treatment data simply build the required design matrix for all subject/visit combinations.

## Conjugate prior based macros

- With a single covariance matrix within each subject, conjugate priors can be used for all parameters allowing direct sampling of linear model, covariance and missing-value parameters.
- The **%mymcmc** and **%RMConj** and **%RMCONJPlus** SAS macros take advantage of this delivering independent imputations with minimal thinning using proc MCMC.
- A second imputation stage in (**RMConjPlus**) handles separate imputation models such as CR and also Carpenter et al separate covariances for each arm.
- The **%RefBTVT** macro builds the required design matrix for Reference based imputation.
- The **%mymcmc** macro allows extension to models such as the Diggle-Kenward MNAR model.

## Other SAS MCMC based routes

- The SAS BGLIMM procedure fits Bayesian generalized linear models and handles missing data automatically.
- The **%BGI SAS macro** extracts sampled missed values and builds a multiple imputed data set.
- Useful for non-Normal data.

## Analyzing imputed data set

- The **%MIAnalyze SAS macro** fits a univariate linear model to a multiple imputed data set and combines results using Rubin's rules like proc MI.
- BY= allows for analysis at repeated visits (not MMRM).
- Adds an estimate of Monte Carlo error for each treatment difference.
- Not needed for GSK 5 macros as a complete system.

## Now ...

- Availability of formally validated Stata and R code for RBI is timely.
- My personal choice for research ...  
The **%RMConjPlus SAS macro** as it is fast and flexible.

# References

Ayele, B. T., Lipkovich, I., Molenberghs, G., & Mallinckrodt, C. H. (2014). A multiple-imputation-based approach to sensitivity analyses and effectiveness assessments in longitudinal clinical trials. *Journal of Biopharmaceutical Statistics*, 24(2), 211-228. <https://doi.org/10.1080/10543406.2013.859148>

Carpenter JR, Roger JH & Kenward MG. (2013) Analysis of longitudinal trials with protocol deviation: a framework for relevant, accessible assumptions, and inference via multiple imputation. *J Biopharm Stat.* 23(6):1352-1371.

Elena Polverejan & Vladimir Dragalin (2019): Aligning Treatment Policy Estimands and Estimators A Simulation Study in Alzheimers Disease, *Statistics in Biopharmaceutical Research*, DOI: 10.1080/19466315.2019.1689845)

Ian White, Royes Joseph & Nicky Best (2020) A causal modelling framework for reference-based imputation and tipping point analysis in clinical trials with quantitative outcome, *Journal of Biopharmaceutical Statistics*, 30:2, 334-350, DOI: 10.1080/10543406.2019.1684308

The supporting slides.

## ... and the future perhaps ...

- Rather than use multiple imputation, fit joint Bayesian modelling of the outcome and the missing-data/withdrawal process and predict estimand value based on sample from posterior. Repeat over draws from posterior and summarize.
- Easy with lots of computer power, a suitable model and some well defined priors.



## ITT and collection of off-treatment data

- Leads into importance of Estimand and ICH E9 (R1).
- The term “sensitivity analysis” changes to mean within the same estimand unlike in the EMA guidance.
- Reference-based MI becomes more than simply about sensitivity.
- Start to collect off-treatment data.  
Model off-treatment data to impute under “Treatment-policy” (ITT).
- Both require a linear model for mean and a multivariate Normal model for residuals.  
Often about half off-treatment data is “missing” so perhaps impute.

## The legal stuff

- Apart from GSK 5 macros all were written to research and demonstrate methodology. Extensively tested but not formally validated.
- Open source licence, inbuilt parameter checking, diagnostics and ongoing support from James Roger  
james.roger@lshtm.ac.uk OR james@livedata.co.uk .
- All macros available at the DIA working Party TAB on the LSHTM Missing Data web site.  
<https://www.lshtm.ac.uk/research/centres-projects-groups/missing-data>
- Also here, ...
  - ▶ “Sequential imputation with tipping point and delta adjustment” based on proc MI and MIANALZE.
  - ▶ Imputation for recurrent event and also time-to-event data.

# What is now available

- MIWithD  
RBI: Still used by some. Based on MCMC statment in proc MI.
- GSK 5 macros  
RBI: proc MCMC based. Slow but allows several methods based on a single Bayesian fit. System includes summary of Univariate analysis. No Treatment\*Covariate interactions.
- MIStep  
RBI & OT: Sequential algorithm. Monotone missing only. Uses DATA step only.
- mymcmc  
Most flexible Bayesian conjugate prior fitting macro.
- RMConj and RMConjPlus  
RBI & OT: General conjugate RM model with ability to add separate imputation stage using same model but differeng covariate values for subject\*visit. RefBTVT macro builds RBI model.
- MIAnalyze  
Bundles univariate analysis with summary using Rubin's rules.
- BGI  
Allows extraction of imputed data sets from proc BGLIMM output.