



UiO : **Centre for Educational Measurement**
Det utdanningsvitenskapelige fakultet

Missing Data Treatment

A hand-on illustration using  package [mice](#)

Tony C. A. Tan

28 February 2022

Structure

1 Background

- Summary
- Data Missing Mechanism

2 The **mice** Package

- **mice** Process

Summary

■ Complete-case analyses:

- ✗ Wasteful
- ✗ Biased

■ Two approaches:


- ① Joint modelling (JM, Schafer, 1997)
 - ② Fully conditional specification (FCS)
- 👉 FCS aka multivariate imputation by chained equations (MICE, van Buuren & Groothuis-Oudshoorn, 2011)

■ Existing packages:

- [Amelia](#), [Hmisc](#), [jomo](#), [mi](#), [mice](#), [norm](#), [norm2](#), [pan](#)
- 📖 See Table 5.1, Kleinke et al. (2020) (p. 134) for popularity contest across various MI packages
- 📖 See Table 6, Grund et al. (2018) (pp. 134–135) for missing data treatment for multilevel models

Data Missing Mechanism (Rubin, 1976)


■ Missing completely at random (**MCAR**)

-  missingness of variables is independent of the variables considered in the study
- ✓ no treatment required, complete-case analyses valid and unbiased

■ Missing at random (**MAR**)

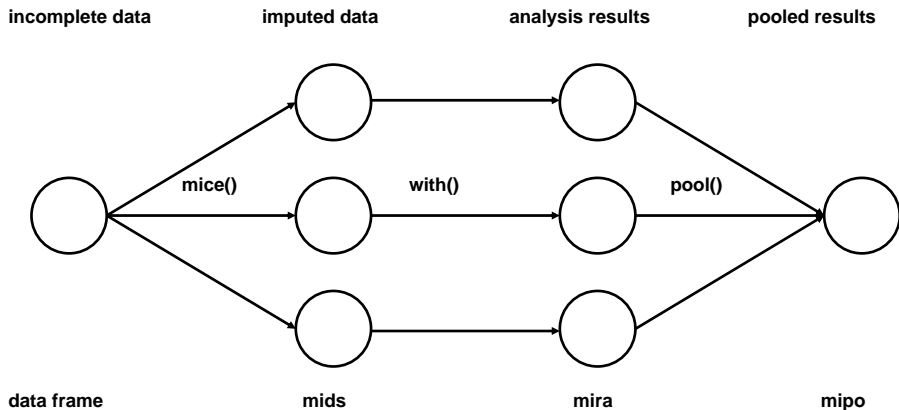
-  missingness depends exclusively on observable variables
- ✓ the assumption behind most MI procedures, including [mice](#)

■ Missing not at random (**MNAR**)

-  missingness depends on unobservable but important variables of interest in the study
- ✓ exact treatment rather complicated (Rose, 2013)
- ✓ in practice: introduce lots of covariates and hope $MNAR \cong MAR$

● Ignorable = { MCAR, MAR }; Nonignorable = { MNAR }

mice Process (van Buuren & Groothuis-Oudshoorn, 2011)



References I

- van Buuren, S., & Groothuis-Oudshoorn, K. (2011). mice: Multivariate imputation by chained equations in R. *Journal of Statistical Software*, 45(3), 1–67. <https://doi.org/10.18637/jss.v045.i03>
- Grund, S., Lüdtke, O., & Robitzsch, A. (2018). Multiple imputation of missing data for multilevel models: Simulations and recommendations. *Organizational Research Methods*, 21(1), 111–149. <https://doi.org/10.1177/1094428117703686>
- Kleinke, K., Reinecke, J., Salfrán, D., & Spiess, M. (2020). *Applied multiple imputation: Advantages, pitfalls, new developments and applications in R*. Springer.
<https://doi.org/10.1007/978-3-030-38164-6>

References II

- Rose, N. (2013). *Item nonresponses in educational and psychological measurement* [PhD Thesis, Friedrich-Schiller-Universität Jena]. Open Access Thesis and Dissertations.
https://www.db-thueringen.de/servlets/MCRFileNodeServlet/dbt_derivate_00027809/Diss/NormanRose.pdf
- Rubin, D. B. (1976). Inference and missing data. *Biometrika*, 63(3), 581–592. <https://doi.org/10.1093/biomet/63.3.581>
- Schafer, J. L. (1997). *Analysis of incomplete multivariate data*. Chapman & Hall; CRC.

UiO : Centre for Educational Measurement

Det utdanningsvitenskapelige fakultet



Tony C. A. Tan



Missing Data Treatment

A hand-on illustration using  package [mice](#)

