Generating missing values with ampute

Rianne Schouten

Department of Methodology and Statistics, University Utrecht
 DPA Professionals Data Science Excellence Program

May 18, 2017

Overview

- ▶ What is amputation?
- Characteristics of missing data problems
- Stepwise Univariate Amputation
- Multivariate Amputation

require(mice)

?ampute

What is amputation?

Amputation is the generation of missing values in complete data

Evaluation of a missing data methodology:

- Simulate complete data set
- Generate missing values
- Deal with missing data with new method
- Compare statistical inferences between the original, complete data set and after dealing with the missing values

Characteristics of missing data problems

MCAR : Missingness is not related to X or Y at all

MAR: Missingness is related to X but not to Y

MNAR : Missingness is related to Y

MCAR : Pr(Y = missing) = 0.5

MAR : logit(Pr(Y = missing)) = X

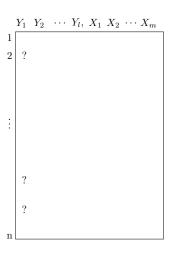
MNAR : logit(Pr(Y = missing)) = Y

Characteristics of missing data problems

- Mechanism
- Proportion
- Severity
- Which variables
- Influence of observed data
- Combinations

Stepwise Univariate Amputation

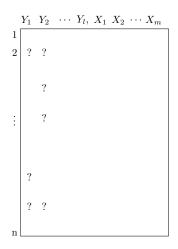
1. *Y*₁



Stepwise Univariate Amputation

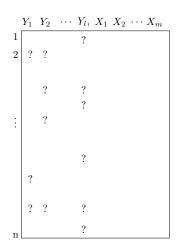


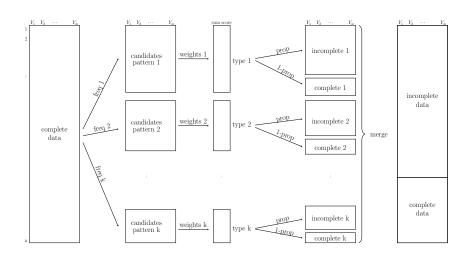
2. *Y*₂



Stepwise Univariate Amputation

- 1. Y₁
- 2. *Y*₂
- 3. ...
- 4. Y₁





A missing data pattern is a specific combination of variables with missing values and variables without missing values.

0: incomplete variable

1: complete variable

```
## y1 y2 x

## 1 4.587067 5.364096 10.333693

## 2 3.070874 5.139578 9.588529

## 3 4.557591 5.771920 9.834172

## 4 4.709714 3.481258 9.890152

## 5 5.080492 6.236113 10.939586

## 6 4.255989 4.980705 9.431149
```

```
inc_data[1:12, ]
```

```
##
            у1
                     y2
                                X
## 1
            NΑ
                     NA 10.333693
## 2
            NA 5.139578 9.588529
## 3
            NΑ
                     NA 9.834172
## 4
            NΑ
                     NA 9.890152
## 5
            NA 6.236113 10.939586
## 6
     4.255989 4.980705 9.431149
## 7
            NA 5.711130 9.607115
## 8
            NΑ
                     NA 11.388759
##
  9
            NΑ
                     NA 10.708886
  10 5.046014 5.876691 10.369283
   11 4.143609 4.770560 9.499410
   12 5.193741 4.729886 8.890983
```

Multivariate Amputation: Proportion and Frequency

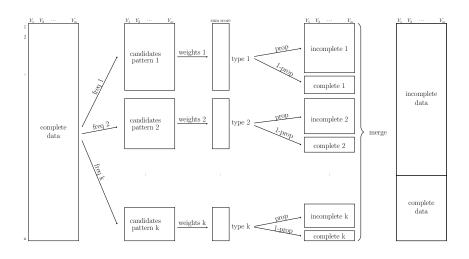
```
amp$prop

## [1] 0.5

amp$freq
```

[1] 0.5 0.5

Multivariate Amputation: Proportion and Frequency



Multivariate Amputation: Proportion and Frequency

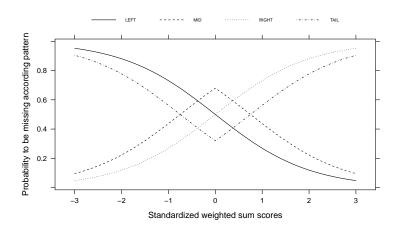
```
## x y2 y1
## 197 1 1 1 0
## 156 1 1 0 1
## 647 1 0 0 2
## 0 647 803 1450
```

```
ampute(data, prop = 0.5, patterns = NULL, freq =
NULL, mech = "MAR", weights = NULL, cont = TRUE, type
= NULL, odds = NULL, bycases = TRUE, run = TRUE)
```

amp\$patterns

```
## y1 y2 x
## 1 0 0 1
## 2 0 1 1
```

Multivariate Amputation: Distribution functions



```
amp$type
```

```
## [1] "RIGHT" "RIGHT"
```

Multivariate Amputation: Weighted sum scores

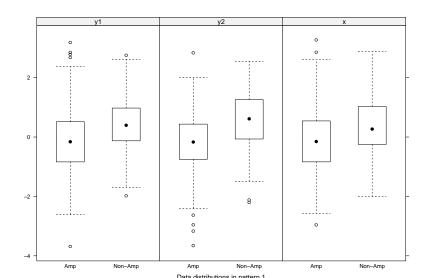
Missing values in multiple variables

$$\begin{array}{ccc} & Y_1 & Y_2 & X \\ P_1 & \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix} \end{array}$$

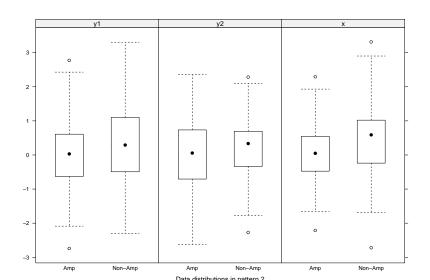
▶ Based on multiple variables $wss_i = w_{1,1} \cdot y_{1i} + w_{1,2} \cdot y_{2i} + w_{1,3} \cdot x_i$ if case i is in pattern 1

Multivariate Amputation: Weighted sum scores

\$`Boxplot pattern 1`



\$`Boxplot pattern 2`



- Mechanism
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```
ampute(data, prop = 0.5, patterns = NULL, freq =
NULL, mech = "MAR", weights = NULL, cont = TRUE, type
= NULL, odds = NULL, bycases = TRUE, run = TRUE)
```

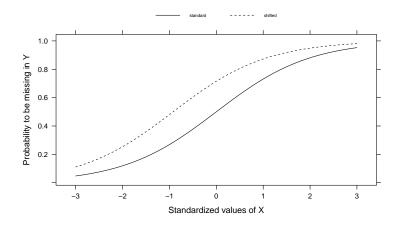
Documentation

Rianne Schouten, Peter Lugtig, Jaap Brand, Gerko Vink (2017) Generate missing values with ampute, available from: https://github.com/RianneSchouten/Amputation_with_ Ampute/tree/master/Manuscript%20article

Rianne Schouten, Peter Lugtig, Gerko Vink (2017) Generating missing values for simulation purposes: A multivariate amputation procedure. Under review. Available from: https://github.com/RianneSchouten/mice/tree/master/vignettes

```
require(mice)
?ampute
```

Additional slides

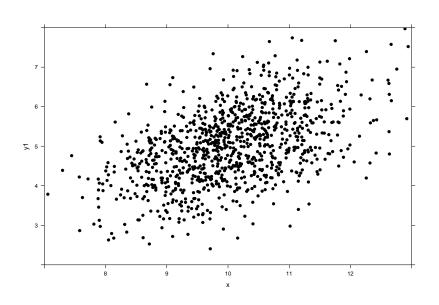


Additional slides

Table 1: Generation of MAR missingness on 2 variables with standard and shifted stepwise univariate amputation (SUA) and multivariate amputation (MA)

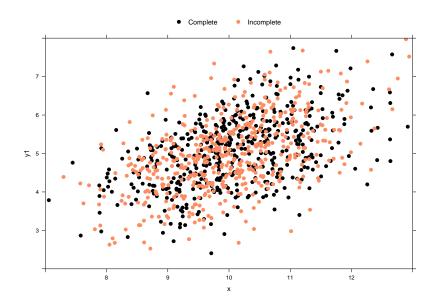
		%mis		complet	complete case analysis			multiple imputation		
cor	condition	int	obt	bias	ciw	cov	bias	ciw	cov	
	standard SUA	50	29	-0.146	0.144	0.028	-0.002	0.156	0.940	
0.5	shifted SUA	50	50	-0.233	0.172	0.000	-0.007	0.204	0.917	
	MA with ampute	50	50	-0.207	0.172	0.002	-0.005	0.193	0.936	

Characteristics of missing data problems



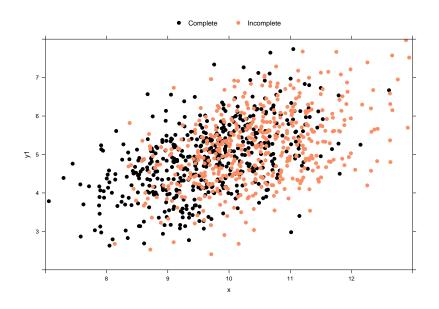
Characteristics of missing data problems: MCAR

Characteristics of missing data problems: MCAR



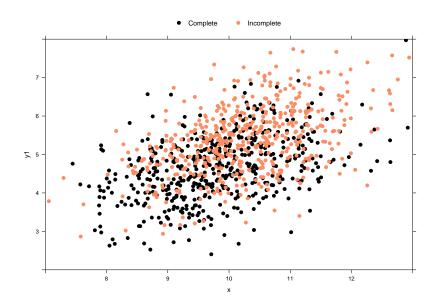
Characteristics of missing data problems: MAR

Characteristics of missing data problems: MAR



Characteristics of missing data problems: MNAR

Characteristics of missing data problems: MNAR



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