Data Science Hackathon

Missing Values Treatment

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Welcome

Introduction

- ► Rianne Schouten
- ► DPA Professionals
- ► Missing Data Specialist





Welcome

Introduction

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What do you expect of today?

Welcome

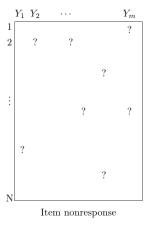
Introduction

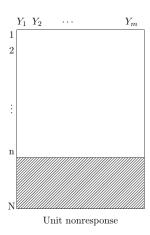
- ► Rianne Schouten
- ▶ DPA Professionals
- Missing Data Specialist

What do you expect of today?

In this presentation:

- ▶ What is missing data?
- ▶ How to deal with missing data?
- ► Today's challenge





head(inc_data)

```
## outcome feature
## 1 5.364150 NA
## 2 5.618218 NA
## 3 5.210518 10.73564
## 4 4.519554 NA
## 5 3.610414 NA
## 6 5.945725 NA
```

```
require(mice)
md.pattern(inc_data)
```

- ▶ MCAR: Missingness is fixed, not related to any variable
- ▶ MAR: Missingness is related to an observed variable
- MNAR: Missingness is related to the missingness itself or to an unobserved variable

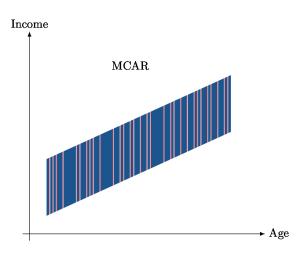
Example:

Consider outcome variable 'income' and feature 'age'

- ▶ MCAR: Some age values are missing, both older and younger ages
- ▶ MAR: Age values are missing, especially for people with a high income
- ▶ MNAR: Age values are missing, especially for older people

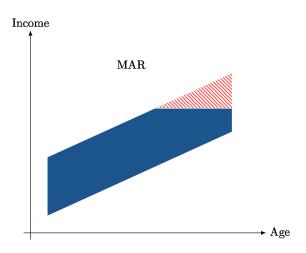
What is missing data: MCAR

Independent of value size, values on feature 'Age' are missing



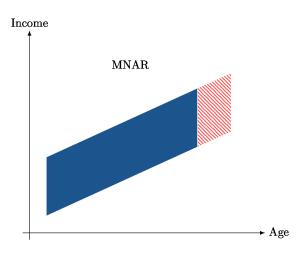
What is missing data: MAR

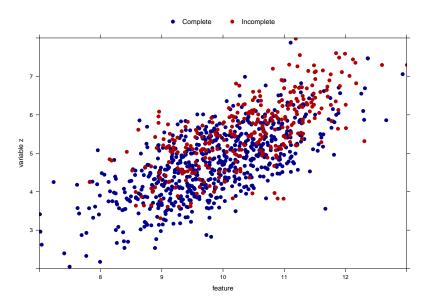
Records with a large value on 'Income' are missing on feature 'Age'



What is missing data: MNAR

Records with a large value on feature 'Age' are missing on feature 'Age'





How to deal with missing data?

- 1. Drop incomplete rows/columns
- 2. Imputation
- random imputation
- mean/median imputation
- regression imputation
- random forest imputation
- multiple imputation
- ▶ and more...
- 3. Other methods such as
- weighting procedures
- ▶ likelihood based methods
- and more...

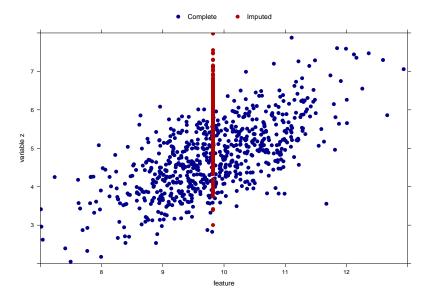
How to deal with missing data: mean imputation

head(inc_data)

```
## outcome feature
## 1 5.364150 NA
## 2 5.618218 NA
## 3 5.210518 NA
## 4 4.519554 9.939668
## 5 3.610414 NA
## 6 5.945725 NA
```

```
com_data <- inc_data
com_data[is.na(inc_data$feature), 'feature'] <-
mean(inc_data$feature, na.rm = TRUE)</pre>
```

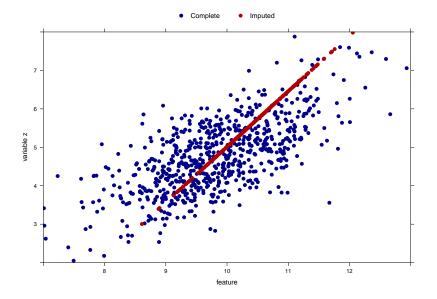
How to deal with missing data: mean imputation



How to deal with missing data: regression imputation

```
fit <- lm(feature ~ z, data = inc_data)
pred <- predict(fit, newdata = ic(inc_data))
com_data <- inc_data
com_data[is.na(inc_data$feature), 'feature'] <- pred</pre>
```

How to deal with missing data: regression imputation



Today's challenge

You receive:

- an incomplete training dataset
- outcome variable (dummy) is complete
- missingness in training and testset is comparable
- combination of MCAR, MAR and MNAR

Make sure:

- you fit your imputation method on your trainingset
- and transform/apply on testset

Question:

▶ How would you evaluate whether your imputation method is okay?

Contact information

Ask me anything, always:

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Follow DPA Data Science: dpadatascience.nl Follow my work: rianneschouten.github.io





Work in progress

Simulation with real dataset slump_test

