Strategies for dealing with Missing Data

Maarten L. Buis

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Maarten L. Buis Strategies for dealing with Missing Data

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- Simple example
 - We have a theory that working for cash is mainly "men's work" and collecting stuff from the forest is mainly "women's work".

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Simple example

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- We go to PEN and make a table

% women	% forest income
0–25	
25–50	
50–75	
75–100	

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 0-25
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 75-100
- Question \rightarrow Observe stuff \rightarrow Answer

Simple example

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- We go to PEN and make a table
 % women % forest income
 0-25
 25-50
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 75-100
- Question \rightarrow Observe stuff \rightarrow Answer
- analysis strategy is just there to summarize the observed stuff so we can see the answer

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What are missing data?

person or company or village	var1	var2	var3
1	2	3	5
2	3	7	3

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1	2	3	5
2	3	7	3
3	4	?	5

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you can quickly loose a frightening proportion of your data

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 - composite variable, like income

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- composite variable, like income
- regression with multiple variables

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- you can quickly loose a frightening proportion of your data
 - composite variable, like income
 - regression with multiple variables
- you may not measure what you want to measure (bias)
 - prices more often forgotten remembered when collected small amounts

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Imputation

We want to be able to use the observed part of a case without "adding" information on the missing data.

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Imputation

- We want to be able to use the observed part of a case without "adding" information on the missing data.
- Reproducing patterns in observed data on the missing data, so we can use the observed part of a case

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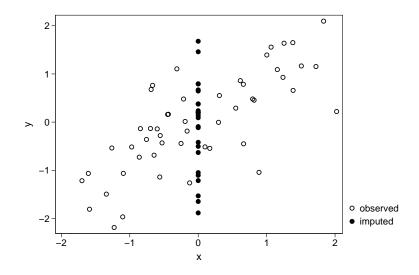
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Imputation

- We want to be able to use the observed part of a case without "adding" information on the missing data.
- Reproducing patterns in observed data on the missing data, so we can use the observed part of a case
- Not recovering the lost observation

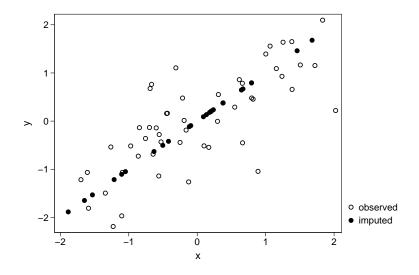
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mean median mode imputation



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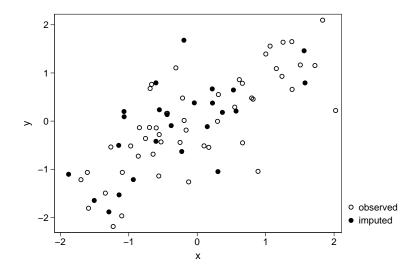
regression imputation



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regression imputation + uncertainty



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Mean imputation is simple, can't we safe it?

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- Mean imputation is simple, can't we safe it?
- What about adding a control variable that says that it is imputed?

$$y = b_0 + b_1 x + b_2 D$$

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What happens when x is missing:

$$y = b_0 + b_1 \bar{x} + b_2 \mathbf{1}$$
$$y = b_0^*$$

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What if the missingness happens in a control variable (z)

$$y = b_0 + b_1 x + b_2 z + b_3 D$$

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What if the missingness happens in a control variable (z)

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What happens when z is observed:

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 $y = b_0 + b_1 x + b_2 z$

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$$y = b_0 + b_1 x + b_2 \overline{z} + b_3 1$$

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b₁ is now a mixture of the effect of x while controlling for z, and the effect of x while not controlling for z.

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- Shouldn't the imputed observations count less than observed observations?
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- Yes
- This is where multiple imputation comes in

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 - point estimate is the average of the point estimates

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- Imputed values are random draws.
- Give each missing value multiple imputed values.
- Result: multiple imputed datasets.
- Do your analysis separately in each dataset.
- Summarize the results:
 - point estimate is the average of the point estimates
 - Uncertainty about the point estimate (standard error) is a combination of:
 - 1. the average uncertainty and
 - 2. the degree to which the results in the different datasets differ from one another

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multiple imputation (3)

what variables to include in imputation model?

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multiple imputation (3)

- what variables to include in imputation model?
- All variables included in your model of interest.

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multiple imputation (3)

- what variables to include in imputation model?
- All variables included in your model of interest.
- Including your dependent variable.

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Where we are going with PEN

Challenges

- we don't know how the dataset is going to be used
- not all users have access to the way of analyzing multiply imputed data

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Where we are going with PEN

Challenges

- we don't know how the dataset is going to be used
- not all users have access to the way of analyzing multiply imputed data
- Criteria
 - document what is real and what is imputed
 - useable for users
 - conservative (don't impute everything)

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- ► This would consist of e.g. price_{fire wood} × quantity_{fire wood}+ price_{brazil nuts} × quantity_{brazil nuts}

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- By default users will get no imputed values.
- Optionally, they can get single imputed data
- Only impute prices and quantities
- We expect most people to use income components from different sectors
- ► This would consist of e.g. price_{fire wood} × quantity_{fire wood}+ price_{brazil nuts} × quantity_{brazil nuts}
- So this is where there are 4 opportunities of getting a missing value, and each imputed value adds relatively little information.

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Recommendations

While in the field collect as much information as possible.

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- While in the field collect as much information as possible.
- If you do your own imputation go for multiple imputation.
 - ► Use all explanatory variables *and* the explained variable.
 - Look at the imputed and observed values (graphs).

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Recommendations

- While in the field collect as much information as possible.
- If you do your own imputation go for multiple imputation.
 - Use all explanatory variables and the explained variable.
 - Look at the imputed and observed values (graphs).
- If you impute for general use data
 - Document the imputed values.
 - Get an idea of how the data is going to be used.
 - Make a trade-off between 'correct' statistical procedure and what the users are comfortable with using.

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