Exploring the multivariate structure of missing values using the R package VIM

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Motivation

Missing values

- Real data sets often contain missing values:

\[ X = \begin{pmatrix}
    x_{11} & \ldots & \ldots & x_{1p} \\
    \vdots & NA & \vdots & \vdots \\
    \vdots & NA & \vdots & \vdots \\
    x_{n1} & \ldots & \ldots & x_{np}
\end{pmatrix}, \]

with \( n \) observations, \( p \) variables, and some missing values. (NA)

- Examples: nonresponse in surveys, element concentration below detection limit in chemical analyses.
Most statistical methods can only be applied to complete data. In order to select an appropriate imputation method (especially for model-based imputation), it is necessary to know the multivariate structure of the missing values beforehand. Visualizing missing values may not only help to detect the missing value mechanisms, but also to gain insight into the quality and various other aspects of the data.
Missing value mechanisms

Three important cases (e.g., Little and Rubin 2002):

- **MCAR** (**M**issing **C**ompletely **A**t **R**andom):  
  \[ P(X_{miss}|X) = P(X_{miss}) \]

- **MAR** (**M**issing **A**t **R**andom):  
  \[ P(X_{miss}|X) = P(X_{miss}|X_{obs}) \]

- **MNAR** (**M**issing **N**ot **A**t **R**andom):  
  \[ P(X_{miss}|X) = P(X_{miss}|X_{obs}, X_{miss}) \]

where \( X = (X_{obs}, X_{miss}) \) denotes the complete data, and \( X_{obs} \) and \( X_{miss} \) are the observed and missing parts, respectively.
Famous books and almost all articles about missing values do not address visualization.

Visualization tools for missing values are rarely or not at all implemented in SAS, SPSS, STATA or even R.

Through linking, missing values can be highlighted in GGobi (Cook and Swayne 2007) and Mondrian (Theus 2002).

MANET (Unwin et al. 1996, Theus et al. 1997) is quite powerful, but only available for older Apple systems with PowerPC architecture and Mac OS.

Visualization tools for missing values need to be available for the R community so that visualization of missing values, imputation and analysis can all be done from within R, without the need of additional software.
Figure: Austrian EU-SILC data from 2004 with missings generated in variable age.
Figure: Austrian EU-SILC data from 2004.
Scatterplot matrix

Figure: Austrian EU-SILC data from 2004.
Visualization of missing values

Matrixplot

Figure: Austrian EU-SILC data from 2004.
Visualization of missing values

Parallel coordinate plot

Figure: Austrian EU-SILC data from 2004
Figure: Austrian EU-SILC data from 2004.
General Statements

- The detection of missing value mechanisms is quite complex when using models or tests.
- Statistical methods frequently lead to only vague statements about the missing value mechanisms.
- Non-robust methods lead to erroneous statements about missing value mechanisms for data containing outliers.
- Visualization tools are easier to handle and more powerful, but flexible, easy-to-use visualization software is required.
The R package VIM (Templ and Filzmoser 2008, Templ and Alfons 2009) has all previously shown plots implemented, along with some more. It is a tool for explorative data analysis of data with missing values. It makes it possible to analyze the multivariate structure of missing values. It comes with a graphical user interface (GUI). It contains interactive features. It allows producing high-quality graphics for publications. It is available on CRAN (http://cran.r-project.org/package=VIM).
Graphical user interface of the R Package VIM

Figure: VIM GUI
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