# Getting Started with Mokken Scale Analysis in R

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# Chapter 1

# Getting started

#### 1.1 Introduction

This report aims at researchers who have Windows installed on their computer and who wish to conduct Mokken scale analysis using the freeware Rpackage mokken (Van der Ark, 2007) but who do not know anything about R. It is a step by step guide from scratch to actually performing Mokken scale analysis. A more elaborate book on learning R for SPSS and SAS users is Muenchen (2008). The report is organized as follows. In this chapter (chap. 1), I discuss the preparations that are needed before mokken can be used. In section 1.2, I discuss the installation of R and all the necessary packages. In section 1.3, I discuss how SPSS, SAS, STATA, and Splus data sets should be converted to R. In section 1.4, I show a few R commands that come in handy for data manipulation (e.g., variable selection). In chapter 2, I explain mokken. In section 2.1, I give an overview of mokken's most important commands (known as functions in R) illustrated by examples. In section 3, I explain the use of mokken by showing the code for most analyses in the book Introduction to nonparametric item response theory (Sijtsma & Molenaar, 2003).

#### 1.2 Installation

#### 1.2.1 What is R?

R (R Development Core Team, 2006) is a language and environment for statistical computing and graphics. It is something in between a statistical package such as SPSS, STATA, or SAS and a programming language such as C++, PASCAL, or FORTRAN. It has two big advantages: It is for free and it has open source. Because it is for free it is accessible for anyone at any time, and because it has an open source researchers can add *packages* to R. Currently, over 2100 packages are added — the package mokken is one of

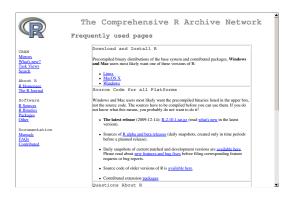


Figure 1.1: R website. Click on Windows.

them — allowing the user to conduct almost every possible statistical procedure ranging from common statistical procedures, which are also available in commercial software packages, to statistical techniques such as item response theory, spectral analysis, marginal modelling, Bayesian analysis, and latent class analysis. Every two months or so R releases a new version. At the time of writing version R-2.10.1 was the most recent version. Although, it is good to have a recent version of R, I only update a new version once a year or so. The major R website is <a href="http://cran.r-project.org/">http://cran.r-project.org/</a>. It contains an abundance of information. A special page is devoted to packages that are of interest to psychometricians

(http://cran.r-project.org/web/views/Psychometrics.html).

#### 1.2.2 Installing R

Installing R requires the following steps

- 1. Go to http://cran.r-project.org/.
- 2. Click on Windows (See Figure 1.1)
- 3. Click on base (See Figure 1.2)
- 4. Click on Download R 2.10.1 for Windows (or a more recent version; see Figure 1.3)
- 5. Save the file R-2.10.1-win32.exe on your computer (e.g., on C:/).
- 6. Run the file R-2.10.1-win32.exe from your computer. You can choose all the default values in the installation Wizard.



Figure 1.2: R website. Click on base.



Figure 1.3: R website. Click on Download R 2.10.1 for Windows.

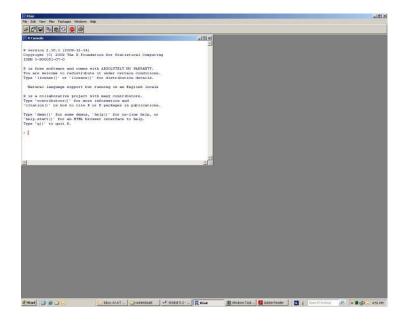


Figure 1.4: R console.

7. R will be available from the desk top icon, and from the programme's menu.

#### 1.2.3 Installing the package mokken

- 1. Open R (Figure 1.4 shows the R console).
- 2. In the pull down menu choose Packages, Install package(s), choose a location nearby you, and choose the package mokken (Figure 1.5).

The package mokken is now installed on your computer and need not be installed anymore. It may be noted that the same procedure applies to the installation of all packages.

#### 1.2.4 Working with R

Except for loading packages (section 1.2.3), almost everything in R is conducted by typing (or pasting) code in the R console (Figure 1.4) just after the prompt (>) and end with hard return (Enter). Code to be typed is printed in *slanted typewriter text* and preceded by a > (a prompt). The resulting output is printed exactly as it is printed on the screen in typewrite text. Note that R is case sensitive. Some examples.

> 6 - 3

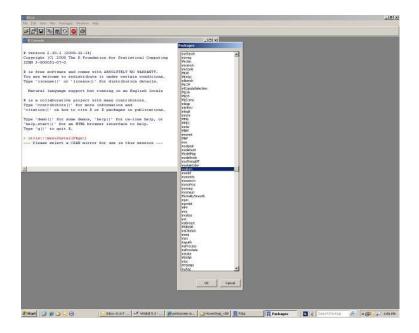


Figure 1.5: R console. Choose mokken.

```
[1] 3
> options(width = 60)
> x <- sqrt(2)
> x
[1] 1.414214
```

R assigns the value  $\sqrt{2}$  to variable x (line 1) and displays the value of x (line 2)

To load the procedures of mokken into the memory of R type

> library(mokken)

To quit R, type

> q()

Free of charge introductions to R are available on the Internet.

• R Development Core Team (2009). An Introduction to R. Retrieved from http://cran.r-project.org/doc/manuals/R-intro.html (html) or http://cran.r-project.org/doc/manuals/R-intro.pdf (pdf).

- Paradis, E. (2005). *R for beginners*. Retrieved from http://cran.r-project.org/doc/contrib/Paradis-rdebuts\_en.pdf
- Baron, J., & Li, Y. (2007). Notes on the use of R for psychology experiments and questionnaires. Retrieved from http://www.psych.upenn.edu/~baron/rpsych/rpsych.html

Many more sources are available from http://cran.r-project.org/ (Contributed Documentation).

### 1.3 Converting data to R and back again

Converting a data set from a commercial package to R is the Achilles Heel of Mokken scale analysis in R. Commercial packages have no interest in free software that can easily read their data sets and these companies put no effort making their data files compatible with R. As a result, small things that you may not be aware of (e.g., whether your computer uses a point or a comma as a decimal separator, whether or not the rows in your data set have labels) may affect the conversion. An elaborated manual for converting many types of files in to files that can be read by R is available from http://cran.r-project.org/doc/manuals/R-data.html. Here only conversions to and from SPSS, SAS, STATA, and Splus are briefly discussed. The fasted strategy is to read the SPSS, SAS, STATA, or Splus file directly in R. Direct reading may occasionally go wrong and an alternative option is to save the SPSS, SAS, STATA, or Splus file as a text-only file (ASCII file), and read the ASCII file into R. In the latter procedure, the variable names may get lost.

#### 1.3.1 SPSS files

#### Converting SPSS files directly

I assume that an SPSS data set named ExampleSPSS.sav has been saved on  $C:/^{1}$ .

- 1. Type the following code in the R console
  - > library(foreign)
  - > ExampleR <- data.frame(read.spss("C:/ExampleSPSS.sav"))</pre>
  - > fix(ExampleR)

<sup>&</sup>lt;sup>1</sup>ExampleSPSS.sav is a completely arbitrary name and your data set probably has a different name and may be located on another drive than C:/. Therefore, you should replace C:/ExampleSPSS.sav by your complete path and file name. The SPSS-file is not included in the Mokken package.

Note that data.frame() is an R function; it saves the data in a matrix-like manner, allowing different measurement levels for the scores in each column. Most data sets in R belong to the class data.frame. The data file is now stored in the memory of R under the name ExampleR<sup>2</sup>. The last command is not necessary. It opens the R data in a spread sheet in another window in R; the spreadsheet can be used to check whether the transformation went well. If necessary, the spread sheet may be modified. If the spreadsheet window is closed (by clicking the close button in the upper right-hand corner, see Figure 1.6) the changes are saved. Note that library(foreign) may be omitted, if it has been typed in before during the same R session.

2. If R is closed, ExampleR are lost. Therefore, the data should be saved in an R format that can be retrieved easily. To save the data (in the file C:/ExampleR.Rdata) type

```
> save(ExampleR, file = "C:/ExampleR.Rdata")
To get the data back into R type
> load("C:/ExampleR.Rdata")
```

#### Saving SPSS files as ASCII files and read the ASCII files

Save the data as a tab delimited ASCII file (.dat file) This format can be read easily by R. The SPSS syntax is

```
SAVE TRANSLATE OUTFILE='C:\ExampleSPSS2.dat'

/TYPE=TAB

/MAP

/REPLACE

/FIELDNAMES

/CELLS=LABELS.
```

#### Converting R data to SPSS

To convert R data sets to SPSS directly is impossible. R creates an SPSS syntax file and an ASCII data file. The SPSS syntax file should be run within SPSS. To create the syntax file "ExampleSPSS.SPS" and the data file "ExampleSSPS.txt" from the R data ExampleR, type

```
> library(foreign)
> write.foreign(ExampleR, datafile = "C:/ExampleSPSS.txt",
+ codefile = "C:/ExampleSPSS.SPS", package = "SPSS")
```

<sup>&</sup>lt;sup>2</sup>Again ExampleR is a completely arbitrary name and you may decide to name it differently, for example, NKSPdata2008.



Figure 1.6: Close the spreadsheet by clicking the button in the upper right-hand corner

#### 1.3.2 SAS XPORT files

#### Converting SAS XPORT files directly

I assume that the SAS data set ExampleSAS.xpt has been saved on C:/.

- 1. Type the following code in the R console
  - > library(foreign)
  - > ExampleR <- data.frame(read.xport("C:/ExampleSAS.xpt"))</pre>
  - > fix(ExampleR)

The data file is now stored in the memory of R under the name ExampleR. The last command is not necessary. It opens the R data in a spread sheet, which can be used to check whether the transformation went well. If necessary, the spread sheet may be modified. If the spread-sheet window is closed the changes are saved. Note that library(foreign) may be omitted, if it has been typed in before during the same R session.

2. If R is closed, ExampleR are lost. Therefore, the data should be saved in an R format that can be retrieved easily. To save the data (in the file C:/ExampleR.Rdata) type

```
> save(ExampleR, file = "C:/ExampleR.Rdata")
To get the data back into R type
```

> load("C:/ExampleR.Rdata")

#### Converting R data to SAS

To convert R data sets to SAS directly is impossible. R creates a SAS syntax file and an ASCII data file. The SAS syntax file should be run within SAS. To create the syntax file "ExampleSAS.XXX" and the data file "ExampleSAS.txt" from the R data ExampleR, type

```
> library(foreign)
> write.foreign(ExampleR, datafile = "C:/ExampleSAS.txt",
+ codefile = "C:/ExampleSAS.XXX", package = "SAS")
```

#### 1.3.3 STATA files

#### Converting STATA files directly

I assume that the STATA data set ExampleSTATA.dta has been saved on C:/.

1. Type the following code in the R console

```
> library(foreign)
> ExampleR <- data.frame(read.dta("C:/ExampleSTATA.dta"))
> fix(ExampleR)
```

The data file is now stored in the memory of R under the name ExampleR. The last command is not necessary. It opens the R data in a spread sheet, which can be used to check whether the transformation went well. If necessary, the spread sheet may be modified. If the spread-sheet window is closed the changes are saved. Note that library(foreign) may be omitted, if it has been typed in before during the same R session.

2. If R is closed, ExampleR are lost. Therefore, the data should be saved in an R format that can be retrieved easily. To save the data (in the file C:/ExampleR.Rdata) type

```
> save(ExampleR, file = "C:/ExampleR.Rdata")
```

To get the data back into R type

> load("C:/ExampleR.Rdata")

#### Converting R data to STATA

To convert R data sets to STATA directly is impossible. R creates a STATA syntax file and an ASCII data file. The STATA syntax file should be run within STATA. To create the syntax file "ExampleSTATA.do" and the data file "ExampleSTATA.dat" from the R data ExampleR, type

```
> library(foreign)
> write.foreign(ExampleR, datafile = "C:/ExampleSTATA.dat",
+ codefile = "C:/ExampleSTATA.do", package = "Stata")
```

#### 1.3.4 Splus files

#### Converting Splus files directly

I assume that the Splus data set ExampleSplus.ssc has been saved on C:/.

1. Type the following code in the R console

```
> library(foreign)
> ExampleR <- data.frame(read.s("C:/ExampleSplus.ssc"))
> fix(ExampleR)
```

The data file is now stored in the memory of R under the name ExampleR. The last command is not necessary. It opens the R data in a spread sheet, which can be used to check whether the transformation went well. If necessary, the spread sheet may be modified. If the spread-sheet window is closed the changes are saved. Note that library(foreign) may be omitted, if it has been typed in before during the same R session.

2. If R is closed, ExampleR are lost. Therefore, the data should be saved in an R format that can be retrieved easily. To save the data (in the file C:/ExampleR.Rdata) type

```
> save(ExampleR, file = "C:/ExampleR.Rdata")
To get the data back into R type
> load("C:/ExampleR.Rdata")
```

#### Converting Splus objects to R objects

I assume that the you have an Splus object ExampleSplus in Splus, and that all data can be stored in C:/. Type in the Splus console

```
> dump(ExampleSplus, "C:/Example.dmp")
```

Next, type in the R console

> ExampleR <- dget("C:/Example.dmp")</pre>

## 1.4 R commands required for mokken

Rather than typing commands in the R console, I advice to type the commands in a plain text file, save the file, and paste a command or a series of commands into R. In this way the commands will not be lost.

- If mokken is used, then one should start each R session with
  - > library(mokken)
- If help is required at any stage use the command help(). For example,
  - > help(mokken)

The help file contains examples of mokken. It can be instructive to paste these examples into the R console.

• A hash (#) indicates that everything beyond it on the same line is a comment.

does not do anything.

- There are three data sets included in mokken: acl, cavalini, and transreas. Typing
  - > data(acl)
  - > data(cavalini)
  - > data(transreas)

makes them available in R. Note that without these data() commands, the data sets are not available.

> help(acl)

will give all the information on acl

> fix(cavalini)

will show cavalini in a spreadsheet.

• An arrow <- is used for assignment. Examples

```
> X <- acl
> Y <- 3
> Z <- c(1, 2, 3, 8:11)</pre>
```

The value of X is the data matric acl (X and acl are now equivalent). The value of Y is 3. The value of Z is the vector (1,2,3,8,9,10,11). It can be verified by typing

> Y

[1] 3

> Z

[1] 1 2 3 8 9 10 11

• To select columns and rows from the data matrix brackets are used.

X1 are the scores on the first item 'Reliable')

X2 are the scores on items 11 to 20 (i.e., only the scores on the 10 items of the scale 'Achievement')

X3 are the scores of the first 10 respondents items on all items

X4 is the score of respondent 232 on item 133

 $\mathtt{X5}$  are the scores of respondents 1-100 and 201-300, on items 1, 2, and 4

X6 are the scores of those respondents who had a score 2 on item 1.

Note that in data matrices X3 to X6, the cases (rows) not selected are thrown away, and case numbers are not available. Case numbers can be made through the following commands. If you want to identify the them, you can create case numbers for acl.

#### > dimnames(acl)[[1]] <- 1:nrow(acl)</pre>

If you repeat the analyses above, you may observe that the case numbers have been preserved.

# Chapter 2

# The R package mokken

## 2.1 An overview of the functions

The package mokken consists of the following functions

#### 2.1.1 aisp

Function aisp performs Mokken's (1971) automated item selection algorithm. In the example, the scores on the first ten items from ACL are used; these are the items of the scale Communality. Mokken's automated item selection algorithm is applied to the ten items. The output (in blue) shows that items unscrupulous\* and unintelligent\* are unscalable, that items reliable, honest, deceitful\*, and dependable are in scale 1, and items obnoxious\*, thankless\*, unfriendly\*, and cruel\* are in scale 2.

```
> data(acl)
> Communality <- acl[, 1:10]
> scale <- aisp(Communality, verbose = FALSE)
> scale
```

	Scale
reliable	1
honest	1
unscrupulous*	0
deceitful*	1
unintelligent*	0
obnoxious*	2
thankless*	2
unfriendly*	2
dependable	1
cruel*	2

Variations of aisp (output not shown) are the following

• Use a genetic algorithm (Straat, van der Ark, & Sijtsma, 2010) rather than Mokken's algorithm.

```
> scale <- aisp(Communality, search = "ga")</pre>
```

• Use different values for the lower bound (default lowerbound = .3) and or the nominal type I error rate (default alpha = .05)

```
> scale <- aisp(Communality, lowerbound = 0.2, alpha = 0.1)
```

• Output on the screen during the item selection (default verbose = TRUE)

```
> aisp(Communality, verbose = TRUE)
```

• For more information type

```
> help(aisp)
```

Note that search = "extend" has not yet been implemented.

#### 2.1.2 coefH

Computes scalability coefficients  $H_{ij}$ ,  $H_i$ , and H for a set of items. In the example, the scores on the first ten items from ACL are used; these are the items of the scale Communality. First, scalability coefficients  $H_{ij}$ ,  $H_i$ , and H are computed, plus their standard errors. The argument  $\mathbf{se} = \mathbf{FALSE}$  drops the standard errors. Second, only the item scalability coefficients are extracted. Third, the item scalability coefficients are extracted but rounded to two integers. Fourth, H is extracted. Fifth, the respondents are divided into two subgroups, those having a score less than 2 on item 11, and those having a score greater than or equal to 2 on item 11. Sixth, the scalability coefficients are computed for the entire sample, and for each of the two subsamples using the item-step ordering from the entire sample (this corresponds to the function CHECK=GROUPS in MSP). Seventh, the item-scalability coefficients are reported

```
unscrupulous*
                 0.242
                          (0.068)
                                   0.250
                                           (0.071)
deceitful*
                                   0.282
                 0.331
                          (0.055)
                                           (0.057)
unintelligent*
                 0.132
                                   0.139
                                           (0.055)
                          (0.057)
                 0.273
                                   0.162
obnoxious*
                          (0.059)
                                           (0.057)
thankless*
                 0.183
                          (0.056)
                                   0.160
                                           (0.058)
unfriendly*
                 0.247
                          (0.061)
                                   0.192
                                           (0.074)
dependable
                 0.717
                          (0.046)
                                   0.552
                                           (0.054)
                          (0.069)
cruel*
                                   0.141
                                           (0.080)
                 0.116
                unscrupulous* se
                                        deceitful* se
reliable
                 0.242
                               (0.068)
                                         0.331
                                                    (0.055)
honest
                 0.250
                               (0.071)
                                         0.282
                                                    (0.057)
unscrupulous*
                                         0.343
                                                    (0.059)
                               (0.059)
deceitful*
                 0.343
unintelligent*
                 0.145
                               (0.060)
                                         0.102
                                                    (0.053)
obnoxious*
                 0.239
                               (0.053)
                                         0.321
                                                    (0.050)
thankless*
                 0.121
                               (0.051)
                                         0.336
                                                    (0.054)
unfriendly*
                 0.241
                               (0.063)
                                         0.399
                                                    (0.052)
dependable
                 0.229
                               (0.066)
                                         0.317
                                                    (0.053)
cruel*
                 0.291
                               (0.061)
                                         0.423
                                                    (0.055)
                unintelligent* se
                                         obnoxious* se
reliable
                                          0.273
                 0.132
                                (0.057)
                                                     (0.059)
honest
                 0.139
                                (0.055)
                                          0.162
                                                     (0.057)
unscrupulous*
                 0.145
                                (0.060)
                                          0.239
                                                     (0.053)
deceitful*
                 0.102
                                (0.053)
                                          0.321
                                                     (0.050)
                                          0.201
unintelligent*
                                                     (0.052)
                 0.201
                                (0.052)
obnoxious*
                 0.133
                                          0.339
                                                     (0.052)
thankless*
                                (0.051)
unfriendly*
                 0.112
                                (0.056)
                                          0.407
                                                     (0.052)
                                          0.253
dependable
                 0.077
                                (0.059)
                                                     (0.058)
cruel*
                 0.005
                                (0.055)
                                          0.372
                                                     (0.061)
                thankless* se
                                     unfriendly* se
reliable
                 0.183
                            (0.056)
                                      0.247
                                                  (0.061)
                 0.160
                            (0.058)
                                      0.192
honest
                                                  (0.074)
unscrupulous*
                 0.121
                            (0.051)
                                      0.241
                                                  (0.063)
deceitful*
                 0.336
                            (0.054)
                                      0.399
                                                  (0.052)
unintelligent*
                 0.133
                            (0.051)
                                      0.112
                                                  (0.056)
obnoxious*
                 0.339
                                      0.407
                            (0.052)
                                                  (0.052)
thankless*
                                      0.430
                                                  (0.055)
unfriendly*
                 0.430
                            (0.055)
                            (0.060)
dependable
                 0.227
                                      0.254
                                                  (0.053)
cruel*
                 0.267
                            (0.066)
                                      0.430
                                                  (0.065)
                dependable se
                                     cruel*
                                             se
reliable
                 0.717
                            (0.046)
                                      0.116
                                              (0.069)
honest
                 0.552
                            (0.054)
                                      0.141
                                              (0.080)
```

```
unscrupulous*
               0.229
                         (0.066) 0.291 (0.061)
deceitful*
               0.317
                         (0.053)
                                  0.423 (0.055)
unintelligent*
               0.077
                         (0.059)
                                  0.005 (0.055)
               0.253
                         (0.058) 0.372 (0.061)
obnoxious*
thankless*
               0.227
                         (0.060)
                                 0.267 (0.066)
unfriendly*
               0.254
                         (0.053)
                                  0.430 (0.065)
dependable
                                  0.118 (0.077)
cruel*
               0.118
                         (0.077)
```

#### > coefficients\$Hi

	Item H	se
reliable	0.304	(0.029)
honest	0.265	(0.034)
unscrupulous*	0.236	(0.034)
deceitful*	0.319	(0.028)
unintelligent*	0.116	(0.033)
obnoxious*	0.288	(0.028)
thankless*	0.245	(0.030)
unfriendly*	0.309	(0.032)
dependable	0.299	(0.032)
cruel*	0.252	(0.036)

#### > coefficients\$H

Scale H se 0.264 (0.020)

- > subgroup <- ifelse(acl[, 11] < 2, 1, 2)</pre>
- > coefficients <- coefH(Communality, group.var = subgroup)</pre>
- > coefficients\$Groups[[1]]\$Hi

	Item H	se
reliable	0.269	(0.068)
honest	0.107	(0.075)
unscrupulous*	0.117	(0.070)
deceitful*	0.229	(0.072)
$\verb"unintelligent""$	0.130	(0.064)
obnoxious*	0.240	(0.056)
thankless*	0.170	(0.061)
unfriendly*	0.217	(0.064)
dependable	0.149	(0.078)
cruel*	0.208	(0.077)

#### > coefficients\$Groups[[2]]\$Hi

	Item H	se
reliable	0.313	(0.032)
honest	0.300	(0.037)
unscrupulous*	0.258	(0.038)
deceitful*	0.334	(0.030)
$\verb"unintelligent""$	0.108	(0.036)
obnoxious*	0.291	(0.032)
thankless*	0.255	(0.034)
unfriendly*	0.323	(0.036)
dependable	0.338	(0.032)
cruel*	0.258	(0.040)

#### 2.1.3 check.iio

Investigates invariant item ordering (IIO) using method  $Manifest\ IIO$  (MIIO; Ligtvoet, Van der Ark, Te Marvelde, & Sijtsma, 2010) and methods  $Manifest\ Scale$  -  $Cumulative\ Probability\ Model$  (MS-CPM) and  $Increasingness\ in\ Transposition$  (IT) (Ligtvoet, Van der Ark, Bergsma, & Sijtsma, 2010). Method Manifest IIO is the default. First, all result with respect to IIO are saved in iio.results. In the example, the scores on the first ten items from ACL are used; these are the items of the scale Communality. Simply typing iio.results produces a list with lots of output for each item. summary() reduces this output by giving a summary of the results. The output shows the method used (i.e., Manifest IIO), the violations of manifest IIO, the items selected using the backward selection algorithm, and scalability coefficient  $H^T$  for the final scale (items unfriendly\* and deceitful\* excluded).

```
> data(acl)
> Communality <- acl[, 1:10]
> iio.results <- check.iio(Communality)
> summary(iio.results)
$method
[1] "MIIO"
```

#### \$item.summary

·	mean	#ac	#vi	#vi/#ac	maxvi	SIIM	sum/#ac	tmax
cruel*	3.48	27	0	0.00	0.00	0.00	0.00	0.00
$\verb"unintelligent"*$	3.32	26	2	0.08	0.15	0.29	0.01	2.17
unscrupulous*	3.32	26	1	0.04	0.14	0.14	0.01	1.21
unfriendly*	3.30	27	1	0.04	0.15	0.15	0.01	2.17
thankless*	3.26	27	1	0.04	0.12	0.12	0.00	1.50
dependable	3.25	27	0	0.00	0.00	0.00	0.00	0.00
obnoxious*	3.25	27	1	0.04	0.12	0.12	0.00	1.50

```
reliable
               3.09
                     27
                          0
                                0.00 0.00 0.00
                                                   0.00 0.00
                          2
honest
               3.02
                     27
                                0.07 0.18 0.31
                                                   0.01 2.08
deceitful*
               2.94
                     25
                          2
                                0.08 0.18 0.31
                                                   0.01 2.08
               #tsig
cruel*
                   0
unintelligent*
                   1
unscrupulous*
                   0
unfriendly*
thankless*
                   0
dependable
                   0
obnoxious*
                   0
reliable
                   0
honest
deceitful*
```

#### \$backward.selection

	step	1	step	2	step 3
cruel*		0		0	0
$\verb"unintelligent""$		1		1	0
unscrupulous*		0		0	0
unfriendly*		1		1	NA
thankless*		0		0	0
dependable		0		0	0
obnoxious*		0		0	0
reliable		0		0	0
honest		1		0	0
deceitful*		1	1	NΑ	NA

#### \$HT

#### [1] 0.05807174

Variations of check.iio (output not shown) are the following.

- Other values for minvi and minsize (Molenaar & Sijtsma, 2000, pp. 45-46) .
  - > check.iio(Communality, minvi = 0, minsize = 50)
- $\bullet\,$  Using methods MS-CPM and IT (Ligtvoet, Van der Ark, Bergsma, & Sijtsma, 2010)
  - > summary(check.iio(Communality, method = "MS-CPM"))
    > summary(check.iio(Communality, method = "IT"))
- Different nominal Type I error rate for t-test (method MIIO), z-test (Method MS-CPM), and McNemar test (method IT).

```
> summary(check.iio(Communality, alpha = 0.01))
```

• Without backward selection algorithm, and with information screen

```
> summary(check.iio(Communality, item.selection = FALSE))
> summary(check.iio(Communality, verbose = TRUE))
```

- For more information type
  - > help(check.iio)

#### 2.1.4 check.monotonicity (a.k.a. check.single)

Investigates the monotonicity assumption using the observable property manifest monotonicity (Molenaar & Sijtsma, 2000, pp. 70-77). In the example, the scores on the first ten items from ACL are used; these are the items of the scale Communality. First, all result with respect to manifest monotonicity are saved in monotonicity.results. Simply typing monotonicity.results produces a list with lots of output for each item. summary() and plot() reduce this output by giving a summary of the results and graphically displaying the estimated item (step) response functions, respectively. For interpretation of the output see Molenaar and Sijtsma (2000, chap. 6, chap. 7). Without further specifications plot() displays 10 graphs (1 for each item) in a separate R Window, and requires a hard return to go to the next graph. Figure 2.1.4 (p. 22) shows the 10 graphs.

```
> data(acl)
> Communality <- acl[, 1:10]
> monotonicity.results <- check.monotonicity(Communality)
> summary(monotonicity.results)
> plot(monotonicity.results, items = c(1, 2))
```

	${\tt ItemH}$	#ac	#vi	#vi/#ac	${\tt maxvi}$	sum	sum/#ac
reliable	0.30	15	0	0.00	0.00	0.00	0
honest	0.27	16	0	0.00	0.00	0.00	0
unscrupulous*	0.24	20	0	0.00	0.00	0.00	0
deceitful*	0.32	22	0	0.00	0.00	0.00	0
unintelligent*	0.12	20	1	0.05	0.07	0.07	0
obnoxious*	0.29	19	0	0.00	0.00	0.00	0
thankless*	0.25	17	0	0.00	0.00	0.00	0
unfriendly*	0.31	19	0	0.00	0.00	0.00	0
dependable	0.30	17	0	0.00	0.00	0.00	0
cruel*	0.25	19	0	0.00	0.00	0.00	0
	zmax ‡	‡zsię	5				
reliable	0.00	(	)				

```
honest
                0.00
                          0
                0.00
unscrupulous*
                          0
deceitful*
                0.00
                          0
unintelligent* 0.85
                          0
obnoxious*
                0.00
                          0
thankless*
                0.00
unfriendly*
                0.00
                          0
dependable
                0.00
                          0
cruel*
                0.00
                          0
```

Variations of check.monotonicity (output not shown) are the following.

• Other values for minvi and minsize (Molenaar & Sijtsma, 2000, pp. 45-46).

```
> check.monotonicity(Communality, minvi = 0, minsize = 50)
```

• Plot the results for items 1 and 2 only

```
> plot(check.monotonicity(Communality), item = c(1,
+ 2))
```

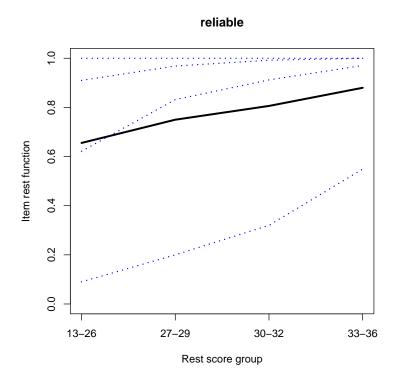
• Save graphs in a pdf file. ask=FALSE assures that no hard return is required between subsequent graphs. The functions pdf() and dev.off() are not part of mokken.

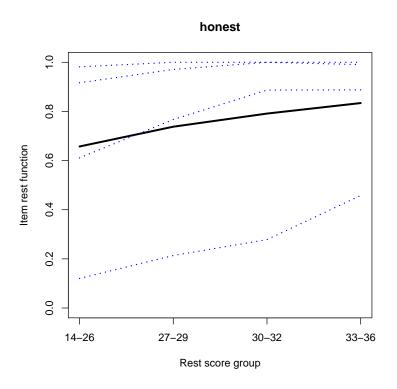
```
> pdf("monotonicity.pdf")
> plot(monotonicity.results, ask = FALSE)
> dev.off()
```

- For more information type
  - > help(check.monotonicity)

#### 2.1.5 check.pmatrix

Investigates the assumption of nonintersecting item step response functions using the P++ and P-- matrix (Molenaar & Sijtsma, 2000, pp. 80-85). In the example, the scores on the first ten items from ACL are used; these are the items of the scale Communality. First, all result with respect to the P++ (indicted by ppp) and P-- (indicated by pmm) matrix are saved in pmatrix.results. Simply typing pmatrix.results produces a list with lots of output for each item. summary() and plot() reduce this output by giving a summary of the results and graphically displaying the estimated item (step) response functions, respectively. For interpretation of the output





see Molenaar and Sijtsma (2000, pp. 80-85). Without further specifications plot() (no output shown) displays 20 graphs (2 for each item) in a separate R Window, and requires a hard return to go to the next graph.

```
> data(acl)
> Communality <- acl[, 1:10]</pre>
> pmatrix.results <- check.pmatrix(Communality)
> summary(pmatrix.results)
> plot(pmatrix.results)
> pmatrix.results <- check.pmatrix(Communality)</pre>
> summary(pmatrix.results)
                      #ac #vi #vi/#ac maxvi
                                               sum sum/#ac
                  Ηi
reliable
                0.30 4608
                           22
                                  0.00
                                        0.05 0.85
                                                     2e-04
honest
                0.27 4608
                             3
                                  0.00
                                        0.04 0.10
                                                     0e+00
                                  0.00
unscrupulous*
                0.24 4608
                             6
                                        0.05 0.23
                                                     1e-04
deceitful*
                0.32 4608
                             8
                                  0.00
                                        0.07 0.33
                                                     1e-04
unintelligent* 0.12 4608
                                  0.00
                                        0.08 0.50
                                                     1e-04
                           12
obnoxious*
                0.29 4608
                             7
                                  0.00
                                        0.08 0.31
                                                     1e-04
thankless*
                0.25 4608
                             7
                                  0.00
                                        0.08 0.33
                                                     1e-04
unfriendly*
                0.31 4608
                             9
                                  0.00
                                        0.08 0.39
                                                     1e-04
                                                     2e-04
dependable
                0.30 4608
                           24
                                  0.01
                                        0.08 1.14
cruel*
                0.25 4608
                           10
                                  0.00 0.06 0.37
                                                     1e-04
                zmax #zsig crit
reliable
                4.27
                        19
                              84
                2.75
honest
                         3
                              41
unscrupulous*
                4.06
                         6
                              60
                3.24
                         7
                              57
deceitful*
unintelligent* 4.55
                        10
                              82
obnoxious*
                5.12
                         7
                              69
thankless*
                4.75
                         6
                              67
unfriendly*
                4.45
                         8
                              67
dependable
                5.12
                        23
                              98
cruel*
                4.40
                         9
                              69
```

Variations of check.pmatrix are the following.

- Other values for minvi (Molenaar & Sijtsma, 2000, pp. 45-46).
  - > check.pmatrix(Communality, minvi = 0)
- Plot the results for P++, for items 1 and 2 only, and plot the results for P-- for item 5.

```
> plot(check.pmatrix(Communality), pmatrix = "ppp",
+    items = c(1, 2))
> plot(check.pmatrix(Communality), pmatrix = "pmm",
+    items = 5)
```

• Save graphs in a pdf file. ask=FALSE assures that no hard return is required between subsequent graphs. The functions pdf() and dev.off() are not part of mokken.

```
> pdf("pmatrix.pdf")
> plot(pmatrix.results, ask = FALSE)
> dev.off()
```

- For more information type
  - > help(check.pmatrix)

#### 2.1.6 check.reliability

Computes reliability coefficients  $\rho$  (a.k.a., the MS statistic; Molenaar & Sijtsma, 1984, 1988; Sijtsma & Molenaar, 1987; Van der Ark, 2010), Cronbach's (1951) alpha, lambda-2 (Guttman, 1945), and on request (not default) the latent class reliability coefficient (LCRC, Van der Ark, Van der Palm, & Sijtsma, 2011). In the example, the scores on the first ten items from ACL are used; these are the items of the scale Communality.

```
> data(acl)
> Communality <- acl[, 1:10]
> check.reliability(Communality)

$MS
[1] 0.75766

$alpha
[1] 0.7465871

$lambda.2
[1] 0.7568063
```

#### 2.1.7 check.restscore

Investigates the assumption of nonintersecting item step response functions using method restscore (Molenaar & Sijtsma, 2000, pp. 77-80). In the example, the scores on the first ten items from ACL are used; these are the items of the scale Communality. First, all result with respect to method restscore

are saved in restscore.results. Simply typing restscore.results produces a list with lots of output for each item pair. summary() and plot() reduce this output by giving a summary of the results and plotting the estimated item (step) response functions, respectively. For interpretation of the output see Molenaar and Sijtsma (2000, pp. 77-80). Without further specifications plot() displays  $\frac{1}{2} \times 10 \times 9 = 45$  graphs (1 for each item pair) in a separate R Window, and requires a hard return to go to the next graph. Figure 2.1.7 (p. 26) shows the rest score plots for the first two item pairs.

```
> data(acl)
> Communality <- acl[, 1:10]</pre>
> restscore.results <- check.restscore(Communality)
> summary(restscore.results)
> plot(restscore.results, item.pairs = c(1, 2))
                ItemH #ac #vi #vi/#ac maxvi sum sum/#ac
reliable
                 0.30 432
                             7
                                  0.02
                                        0.09 0.31
                 0.27 432
                                  0.01
                                        0.07 0.25
                                                          0
honest
                             5
unscrupulous*
                 0.24 416
                             6
                                  0.01
                                        0.11 0.42
                                                          0
deceitful*
                                  0.02
                                        0.09 0.40
                                                          0
                 0.32 400
                             8
                                        0.11 0.86
unintelligent*
                 0.12 416
                                  0.03
                                                          0
                           14
obnoxious*
                 0.29 432
                                  0.02
                                        0.11 0.52
                                                          0
                             7
thankless*
                 0.25 432
                             7
                                  0.02
                                        0.08 0.38
                                                          0
unfriendly*
                 0.31 432
                             8
                                  0.02
                                        0.11 0.48
                                                          0
dependable
                 0.30 432
                             9
                                  0.02
                                        0.09 0.48
                                                          0
                 0.25 432
                             3
                                  0.01
                                        0.04 0.12
                                                          0
cruel*
                zmax #zsig crit
reliable
                1.43
                              26
honest
                1.25
                         0
                              23
unscrupulous*
                1.46
                              33
deceitful*
                1.22
                         0
                              26
                         2
unintelligent* 1.99
                              62
obnoxious*
                1.83
                              44
                         1
thankless*
                1.14
                         0
                              27
unfriendly*
                1.99
                              43
                         1
dependable
                1.22
                         0
                              28
```

Variations of check.restscore are the following.

0

16

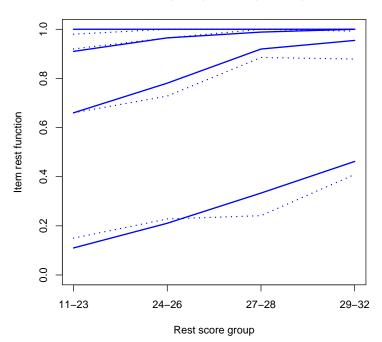
0.67

cruel\*

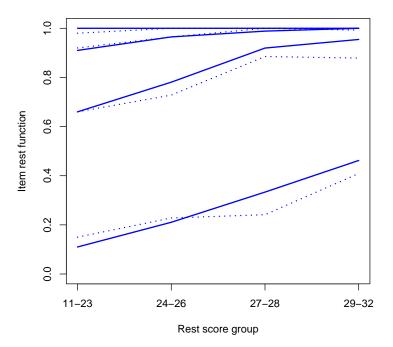
• Other values for minvi and minsize (Molenaar & Sijtsma, 2000, pp. 45-46).

```
> check.restscore(Communality, minvi = 0, minsize = 50)
```

## reliable (solid) honest (dashed)



## reliable (solid) honest (dashed)



- Plot the results for all item pairs.
  - > plot(check.restscore(Communality))
- Save graphs in a pdf file. ask=FALSE assures that no hard return is required between subsequent graphs. The functions pdf() and dev.off() are not part of mokken.

```
> pdf("restscore.pdf")
> plot(restscore.results, ask = FALSE)
> dev.off()
```

- For more information type
  - > help(check.restscore)

#### 2.1.8 check.groups

The package mokken does not have a function check.groups, which—in analogy to the function CHECK=GROUPS in the software program MSP (Molenaar & Sijtsma, 2000, pp. 85-88)—may have been expected.

Some analyses can be conducted using the command coefH.

```
> data(acl)
> Communality <- acl[, 1:10]
> subgroup <- ifelse(acl[, 11] < 2, 1, 2)
> coefH(Communality, group.var = subgroup)
```

#### \$Hij

. 5					
	reliable	se	honest	se se	
reliable			0.528	(0.052)	
honest	0.528	(0.052)			
unscrupulous*	0.242	(0.068)	0.250	(0.071)	
deceitful*	0.331	(0.055)	0.282	(0.057)	
$\verb"unintelligent""$	0.132	(0.057)	0.139	(0.055)	
obnoxious*	0.273	(0.059)	0.162	(0.057)	
thankless*	0.183	(0.056)	0.160	(0.058)	
unfriendly*	0.247	(0.061)	0.192	(0.074)	
dependable	0.717	(0.046)	0.552	(0.054)	
cruel*	0.116	(0.069)	0.141	(0.080)	
	unscrupul	lous* se	Ċ	leceitful*	se
reliable	0.242	(0	.068)	0.331	(0.055)
honest	0.250	(0	.071)	0.282	(0.057)
unscrupulous*				0.343	(0.059)
deceitful*	0.343	(0	.059)		

unintelligent*	0.145	(0.06	30) 0.1	02 (0.0	53)
obnoxious*	0.239	(0.05	53) 0.3	21 (0.0	50)
thankless*	0.121	(0.05	51) 0.3	36 (0.0	)54)
unfriendly*	0.241	(0.06	33) 0.3	99 (0.0	53)
dependable	0.229	(0.06	6) 0.3	17 (0.0	)53)
cruel*	0.291	(0.06	31) 0.4	23 (0.0	)55)
	unintellige			oxious* se	
reliable	0.132		057) 0.	273 (0.	059)
honest	0.139	(0.0	055) 0.	162 (0.	057)
unscrupulous*	0.145	(0.0	060) 0.	239 (0.	053)
deceitful*	0.102	(0.0	053) 0.	321 (0.	050)
unintelligent*			0.	201 (0.	052)
obnoxious*	0.201	(0.0	)52)		
thankless*	0.133	(0.0	051) 0.	339 (0.	052)
unfriendly*	0.112	(0.0	056) 0.	407 (0.	052)
dependable	0.077	(0.0	059) 0.	253 (0.	058)
cruel*	0.005	(0.0	055) 0.	372 (0.	061)
	thankless*	se	unfrien	dly* se	
reliable	0.183	(0.056)	0.247	(0.061	.)
honest	0.160	(0.058)	0.192	(0.074	<u>.</u> )
unscrupulous*	0.121	(0.051)	0.241	(0.063	3)
deceitful*	0.336	(0.054)	0.399	(0.053	3)
unintelligent*	0.133	(0.051)	0.112	(0.056	3)
obnoxious*	0.339	(0.052)	0.407	(0.052	2)
thankless*			0.430	(0.055	5)
unfriendly*	0.430	(0.055)			
dependable	0.227	(0.060)	0.254	(0.053	3)
cruel*	0.267	(0.066)	0.430	(0.065	5)
	dependable	se	cruel*	se	
reliable	0.717	(0.046)	0.116	(0.069)	
honest	0.552	(0.054)	0.141	(0.080)	
unscrupulous*	0.229	(0.066)	0.291	(0.061)	
deceitful*	0.317	(0.053)	0.423	(0.055)	
unintelligent*	0.077	(0.059)	0.005	(0.055)	
obnoxious*	0.253	(0.058)	0.372	(0.061)	
thankless*	0.227	(0.060)	0.267	(0.066)	
unfriendly*	0.254	(0.053)	0.430	(0.065)	
dependable			0.118	(0.077)	
cruel*	0.118	(0.077)			
\$Нi					

\$Hi

Item H se reliable 0.304 (0.029) honest 0.265 (0.034)

```
      unscrupulous*
      0.236 (0.034)

      deceitful*
      0.319 (0.028)

      unintelligent*
      0.116 (0.033)

      obnoxious*
      0.288 (0.028)

      thankless*
      0.245 (0.030)

      unfriendly*
      0.309 (0.032)

      dependable
      0.299 (0.032)

      cruel*
      0.252 (0.036)
```

\$Н

Scale H se 0.264 (0.020)

\$Groups
\$Groups\$`1`
\$Groups\$`1`\$Hij

	reliable	se	honest	se se	
reliable			0.51	0.121	.)
honest	0.510	(0.121)			
unscrupulous*	0.367	(0.153)	0.22	20 (0.130	))
deceitful*	0.275	(0.139)	0.23	34 (0.151	.)
$\verb"unintelligent""$	0.079	(0.129)	-0.06	39 (0.125	5)
obnoxious*	0.335	(0.121)	-0.04	18 (0.101	.)
thankless*	0.099	(0.133)	-0.05	64 (0.115	5)
unfriendly*	0.191	(0.123)	-0.09	95 (0.133	3)
dependable	0.583	(0.147)	0.38	31 (0.134	:)
cruel*	0.079	(0.186)	-0.03	39 (0.175	5)
	unscrupul	Lous* se	Ċ	leceitful*	se
reliable	0.367	(0	.153)	0.275	(0.139)
honest	0.220	(0	.130)	0.234	(0.151)
unscrupulous*				0.024	(0.133)
deceitful*	0.024	(0	.133)		
$\verb"unintelligent""$	-0.130	(0	.156)	0.192	(0.121)
obnoxious*	0.131	(0	.153)	0.233	(0.127)
thankless*	0.104	(0	.098)	0.169	(0.136)
unfriendly*	0.079	(0	.116)	0.214	(0.111)
dependable	0.194	(0	.156)	0.282	(0.158)
cruel*	0.087	(0	.122)	0.422	(0.126)
	unintelli	igent* s	е	obnoxious	* se
reliable	0.079	((	0.129)	0.335	(0.121)
honest	-0.069	((	0.125)	-0.048	(0.101)
unscrupulous*	-0.130	((	0.156)	0.131	(0.153)
deceitful*	0.192	((	0.121)	0.233	(0.127)
$\verb"unintelligent""$				0.254	(0.127)

obnoxious*	0.254	(0.	127)	
thankless*	0.205	(0.	137) 0.	278 (0.142)
unfriendly*	0.143	(0.	141) 0.	475 (0.115)
dependable	0.100	(0.	110) 0.	008 (0.124)
cruel*	0.323	(0.	118) 0.	492 (0.124)
	thankless*	se	unfriend	ly* se
reliable	0.099	(0.133)	0.191	(0.123)
honest	-0.054	(0.115)	-0.095	(0.133)
unscrupulous*	0.104	(0.098)	0.079	(0.116)
deceitful*	0.169	(0.136)	0.214	(0.111)
unintelligent*	0.205	(0.137)	0.143	(0.141)
obnoxious*	0.278	(0.142)	0.475	(0.115)
thankless*			0.491	(0.123)
unfriendly*	0.491	(0.123)		
dependable	0.057	(0.132)	0.051	(0.099)
cruel*	0.224	(0.171)	0.366	(0.150)
	${\tt dependable}$	se	cruel*	se
reliable	0.583	(0.147)	0.079	(0.186)
honest	0.381	(0.134)	-0.039	(0.175)
unscrupulous*	0.194	(0.156)	0.087	(0.122)
deceitful*	0.282	(0.158)	0.422	(0.126)
$\verb"unintelligent""$	0.100	(0.110)	0.323	(0.118)
obnoxious*	0.008	(0.124)	0.492	(0.124)
thankless*	0.057	(0.132)	0.224	(0.171)
unfriendly*	0.051	(0.099)	0.366	(0.150)
dependable			-0.192	(0.144)
cruel*	-0.192	(0.144)		

## \$Groups\$`1`\$Hi

	Item H	se
reliable	0.269	(0.068)
honest	0.107	(0.075)
unscrupulous*	0.117	(0.070)
deceitful*	0.229	(0.072)
unintelligent*	0.129	(0.064)
obnoxious*	0.239	(0.056)
thankless*	0.170	(0.061)
unfriendly*	0.215	(0.063)
dependable	0.149	(0.078)
cruel*	0.208	(0.077)

\$Groups\$`1`\$H Scale H se 0.185 (0.044)

\$Groups\$`2` \$Groups\$`2`\$Hij

Ψαισαροφ z ψπι	J					
	reliable	se	honest	se		
reliable			0.532	(0.057	')	
honest	0.532	(0.057)				
unscrupulous*	0.216	(0.075)	0.255	(0.080	)	
deceitful*	0.345	(0.059)	0.293	(0.061	.)	
$\verb"unintelligent""$	0.144	(0.062)	0.179	(0.060	)	
obnoxious*	0.259	(0.065)	0.211	(0.065	5)	
thankless*	0.208	(0.061)	0.221	(0.065	5)	
unfriendly*	0.266	(0.067)	0.270	(0.078	3)	
dependable	0.744	(0.043)	0.586	(0.057	·)	
cruel*	0.126	(0.072)	0.182	(0.088	3)	
	unscrupul	Lous* se	d	leceitful	* se	
reliable	0.216	(0.	.075)	0.345	(0.059)	
honest	0.255	(0.	.080)	0.293	(0.061)	
unscrupulous*				0.404	(0.064)	
deceitful*	0.404	(0.	.064)			
unintelligent*	0.182	(0.	.065)	0.077	(0.058)	
obnoxious*	0.256	(0.	.057)	0.330	(0.055)	
thankless*	0.122	(0.	.059)	0.366	(0.059)	
unfriendly*	0.272	(0.	.072)	0.434	(0.061)	
dependable	0.238	(0.	.073)	0.328	(0.055)	
cruel*	0.331	(0.	.067)	0.419	(0.061)	
unintelligent* se obnoxious* se						
reliable	0.144	((	0.062)	0.259	(0.065)	
honest	0.179	((	0.060)	0.211	(0.065)	
unscrupulous*	0.182	((	0.065)	0.256	(0.057)	
deceitful*	0.077	((	0.058)	0.330	(0.055)	
unintelligent*				0.182	(0.056)	
obnoxious*	0.182	((	0.056)			
thankless*	0.105	((	0.053)	0.332	(0.057)	
unfriendly*	0.091	((	0.060)	0.378	(0.059)	
dependable	0.077	((	0.067)	0.319	(0.062)	
cruel*	-0.061	((	0.057)	0.340	(0.067)	
	thankless	s* se	unfr	eiendly*	se	
reliable	0.208	(0.061	0.2	266	(0.067)	
honest	0.221	(0.065	5) 0.2	270	(0.078)	
unscrupulous*	0.122	(0.059	9) 0.2	.72	(0.072)	
deceitful*	0.366	(0.059	9) 0.4	:34	(0.061)	
unintelligent*	0.105	(0.053		91	(0.060)	
obnoxious*	0.332	(0.057			(0.059)	

thankless*			0.397	(0.063)
unfriendly*	0.397	(0.063)		
dependable	0.282	(0.065)	0.317	(0.058)
cruel*	0.267	(0.070)	0.439	(0.071)
	dependable	se	cruel*	se
reliable	0.744	(0.043)	0.126	(0.072)
honest	0.586	(0.057)	0.182	(0.088)
unscrupulous*	0.238	(0.073)	0.331	(0.067)
deceitful*	0.328	(0.055)	0.419	(0.061)
$\verb"unintelligent""$	0.077	(0.067)	-0.061	(0.057)
obnoxious*	0.319	(0.062)	0.340	(0.067)
thankless*	0.282	(0.065)	0.267	(0.070)
unfriendly*	0.317	(0.058)	0.439	(0.071)
dependable			0.196	(0.081)
cruel*	0.196	(0.081)		

## \$Groups\$`2`\$Hi

	Item H	se
reliable	0.313	(0.032)
honest	0.300	(0.037)
unscrupulous*	0.258	(0.038)
deceitful*	0.334	(0.030)
$\verb"unintelligent""$	0.108	(0.037)
obnoxious*	0.291	(0.032)
thankless*	0.255	(0.034)
unfriendly*	0.323	(0.036)
dependable	0.338	(0.032)
cruel*	0.258	(0.040)

\$Groups\$`2`\$H Scale H se 0.277 (0.022)

# Chapter 3

# Examples of Mokken scale analysis in R

This chapter shows the code for producing the tables in Sijtsma and Molenaar (2003).

#### 3.1 Table 3.1

Get the transitive reasoning data, and split them into the grades (first column of the data matrix), and the items scores (the remaining columns in the data matrix).

```
> library(mokken)
```

- > data(transreas)
- > grades <- transreas[, 1]</pre>
- > item.scores <- transreas[, -1]</pre>

Obtaining the overall mean scores, and the mean scores per grade

> apply(item.scores, 2, mean)

```
T09L T12P T10W T11P T04W T05W 0.3011765 0.4752941 0.5200000 0.6423529 0.7835294 0.8023529 T02L T07L T03W T01L T08W T06A 0.8094118 0.8447059 0.8847059 0.9411765 0.9670588 0.9741176
```

> apply(item.scores[grades == 2, ], 2, mean)

```
T09L T12P T10W T11P T04W T05W 0.2093023 0.6627907 0.4069767 0.4883721 0.8372093 0.7093023 T02L T07L T03W T01L T08W T06A 0.8139535 0.7209302 0.8023256 0.8488372 0.9302326 0.9534884
```

```
> apply(item.scores[grades == 3, ], 2, mean)
```

```
T09L T12P T10W T11P T04W T05W
0.3294118 0.5058824 0.5764706 0.6000000 0.6470588 0.8000000
T02L T07L T03W T01L T08W T06A
0.7294118 0.8352941 0.9294118 0.9882353 0.9647059 0.9529412
```

#### > apply(item.scores[grades == 4, ], 2, mean)

```
      T09L
      T12P
      T10W
      T11P
      T04W
      T05W

      0.3048780
      0.3902439
      0.3902439
      0.6707317
      0.8170732
      0.7804878

      T02L
      T07L
      T03W
      T01L
      T08W
      T06A

      0.8902439
      0.8658537
      0.9146341
      0.9756098
      0.9756098
      0.9878049
```

#### > apply(item.scores[grades == 5, ], 2, mean)

```
    T09L
    T12P
    T10W
    T11P
    T04W
    T05W

    0.2588235
    0.5176471
    0.5529412
    0.7058824
    0.8235294
    0.8235294

    T02L
    T07L
    T03W
    T01L
    T08W
    T06A

    0.7882353
    0.8705882
    0.8470588
    0.9294118
    0.9764706
    0.9764706
```

#### > apply(item.scores[grades == 6, ], 2, mean)

```
T09L T12P T10W T11P T04W T05W
0.4022989 0.2988506 0.6666667 0.7471264 0.7931034 0.8965517
T02L T07L T03W T01L T08W T06A
0.8275862 0.9310345 0.9310345 0.9655172 0.9885057 1.0000000
```

Construction of Table 3.1 (advanced R code).

```
> Total.group <- round(apply(item.scores, 2, mean),
+ 2)
> for (i in 2:6) assign(paste("Grade.", i, sep = ""),
+ round(apply(item.scores[grades == i, ], 2,
+ mean), 2))
> Task <- c(9, 12, 10, 11, 4, 5, 2, 7, 3, 1, 8,
+ 6)
> Property <- attributes(transreas)$property
> Format <- attributes(transreas)$format
> Objects <- attributes(transreas)$measures
> Measures <- attributes(transreas)$measures
> Table.3.1 <- data.frame(Task, Property, Format,
+ Objects, Measures, Total.group, Grade.2, Grade.3,
+ Grade.4, Grade.5, Grade.6)
> Table.3.1
```

```
Task Property
                                Format Objects
T09L
             length YA = YB < YC = YD</pre>
                                         sticks
       12
T12P
             pseudo
       10
             weight YA = YB < YC = YD
T10W
                                          balls
T11P
       11
             pseudo
TO4W
        4
             weight YA = YB = YC = YD
                                           cubes
T05W
                          YA < YB < YC
        5
             weight
                                          balls
T02L
        2
             length YA = YB = YC = YD
                                           tubes
T07L
        7
             length
                          YA > YB = YC
                                         sticks
TO3W
                          YA > YB > YC
        3
             weight
                                          tubes
T01L
        1
             length
                          YA > YB > YC
                                          sticks
W80T
        8
                          YA > YB = YC
                                          balls
             weight
T06A
                          YA > YB > YC
        6
               area
                                           discs
                         Measures Total.group Grade.2 Grade.3
        12.5, 12.5, 13, 13 (cm)
T09L
                                          0.30
                                                   0.21
                                                            0.33
T12P
                                          0.48
                                                   0.66
                                                            0.51
            60, 60, 100, 100 (g)
                                                   0.41
T10W
                                          0.52
                                                            0.58
T11P
                                          0.64
                                                   0.49
                                                            0.60
TO4W
                           65 (g)
                                          0.78
                                                   0.84
                                                            0.65
T05W
                 40, 50, 70 (cm)
                                                   0.71
                                          0.80
                                                            0.80
T02L
                          12 (cm)
                                                            0.73
                                          0.81
                                                   0.81
T07L
           28.5, 27.5, 27.5 (cm)
                                          0.84
                                                   0.72
                                                            0.84
TO3W
                  45, 25, 18 (g)
                                                   0.80
                                                            0.93
                                          0.88
T01L
               12, 11.5, 11 (cm)
                                          0.94
                                                   0.85
                                                            0.99
W80T
                  65, 40, 40 (g)
                                                   0.93
                                                            0.96
                                          0.97
T06A 7.5, 7, 6.5 (diameter; cm)
                                                   0.95
                                                            0.95
                                          0.97
     Grade.4 Grade.5 Grade.6
T09L
        0.30
                 0.26
                          0.40
T12P
        0.39
                 0.52
                          0.30
T10W
        0.39
                 0.55
                          0.67
        0.67
                 0.71
                          0.75
T11P
TO4W
        0.82
                 0.82
                          0.79
T05W
        0.78
                 0.82
                          0.90
T02L
        0.89
                 0.79
                          0.83
T07L
        0.87
                 0.87
                          0.93
T03W
        0.91
                 0.85
                          0.93
T01L
        0.98
                 0.93
                          0.97
T08W
        0.98
                 0.98
                          0.99
T06A
        0.99
                 0.98
                          1.00
```

#### 3.2 Table 3.2

To get the data, see Table 3.1.

Obtain scalability coefficients and Z coefficients for items and total scale.

> coefH(item.scores, FALSE)\$Hi

```
T09L
                  T12P
                              T10W
                                           T11P
                                                       T04W
0.17075112 -0.13516424
                        0.13762914 -0.02534203
                                                 0.04791651
      T05W
                  T02L
                              T07L
                                           TO3W
                                                       T01L
0.09166970 0.08126425
                        0.17766194 0.18791305
                                                0.20624790
      W80T
                  T06A
0.28245200 0.39559408
```

- > coefH(item.scores, FALSE)\$H
- [1] 0.090514
- > coefZ(item.scores)\$Zi

```
T09L
              T12P
                        T10W
                                             TO4W
                                                       T05W
                                   T11P
4.357153 -4.650737
                    4.949901 -0.967124
                                         1.948629
                                                   3.746486
              T07L
    T02L
                        T03W
                                   T01L
                                             W80T
                                                       T06A
3.307606
         6.868759
                    6.582933
                              5.684119
                                        6.326821
                                                   8.019603
```

- > coefZ(item.scores)\$Z
- [1] 7.395366

Obtain scalability coefficients and Z coefficients for items and total scale, when the pseudo items (2 and 4) are deleted

> coefH(item.scores[, -c(2, 4)], FALSE)\$Hi

 T09L
 T10W
 T04W
 T05W
 T02L

 0.28926218
 0.28083167
 0.03625403
 0.13900577
 0.07160711

 T07L
 T03W
 T01L
 T08W
 T06A

 0.24977935
 0.29285894
 0.28657074
 0.38746871
 0.48272634

- > coefH(item.scores[, -c(2, 4)], FALSE)\$H
- [1] 0.2048305
- > coefZ(item.scores[, -c(2, 4)])\$Zi

T09L T10W T04W T05W T02L T07L 5.548097 7.506324 1.347390 5.285802 2.725229 9.192316 T03W T01L T08W T06A 9.922119 7.821882 8.732480 9.881951

```
[1] 13.54854
Construction of Table 3.2 (advanced R code).
    > Task <- c("9", "12", "10", "11", "4", "5", "2",
          "7", "3", "1", "8", "6", "Total item set")
    > Property <- c(attributes(transreas)$property,
          "")
    > Format <- c(attributes(transreas)$format, "")
    > Table.3.2 <- data.frame(Task, Property, Format,
          matrix(NA, 13, 8))
    > analysis <- list(c(1:12), c(1, 3, 5:12), c(1,
          3, 6, 8:12), c(1, 3, 8:12))
    > k <- 3
    > for (i in 1:4) for (j in 1:2) {
          k < -k + 1
          Table.3.2[c(analysis[[i]], 13), k] <- c(round(coefH(item.scores[,</pre>
              analysis[[i]]], FALSE)$Hi, 2), round(coefH(item.scores[,
              analysis[[i]]], FALSE)$H, 2))
    + }
    > dimnames(Table.3.2)[[2]][4:11] <- paste(c("k=12",</pre>
          "k=12", "k=10", "k=10", "k=8", "k=8", "k=7",
          "k=7"), c("Hi", "Zi"))
    > Table.3.2
                 Task Property
                                           Format k=12 Hi
    1
                         length YA = YB < YC = YD
                                                      0.17
    2
                   12
                                                     -0.14
                         pseudo
    3
                   10
                        weight YA = YB < YC = YD
                                                      0.14
    4
                                                     -0.03
                   11
                        pseudo
    5
                    4
                        weight YA = YB = YC = YD
                                                      0.05
    6
                    5
                        weight
                                     YA < YB < YC
                                                      0.09
    7
                    2
                        length YA = YB = YC = YD
                                                      0.08
                    7
                                     YA > YB = YC
    8
                        length
                                                      0.18
                                     YA > YB > YC
                                                      0.19
    9
                    3
                        weight
    10
                    1
                         length
                                     YA > YB > YC
                                                      0.21
                    8
                                     YA > YB = YC
    11
                         weight
                                                      0.28
                                     YA > YB > YC
    12
                    6
                           area
                                                      0.40
    13 Total item set
                                                      0.09
      k=12 Zi k=10 Hi k=10 Zi k=8 Hi k=8 Zi k=7 Hi k=7 Zi
```

> coefZ(item.scores[, -c(2, 4)])\$Z

0.44

NA

0.44

NA

0.50

NA

0.50

NA

0.29

NA

0.17

-0.14

1

0.29

NA

```
3
      0.14
                0.28
                         0.28
                                 0.46
                                         0.46
                                                 0.52
                                                         0.52
4
     -0.03
                  NA
                           NA
                                   NA
                                           NA
                                                   NA
                                                            NA
5
      0.05
                                                   NA
                0.04
                         0.04
                                   NA
                                           NA
                                                            NA
6
      0.09
                                 0.20
                                         0.20
                                                   NA
                                                            NA
                0.14
                         0.14
7
      0.08
                0.07
                         0.07
                                   NA
                                           NA
                                                   NA
                                                            NA
8
      0.18
                0.25
                         0.25
                                 0.39
                                         0.39
                                                 0.51
                                                         0.51
9
      0.19
                0.29
                         0.29
                                 0.45
                                         0.45
                                                 0.53
                                                         0.53
      0.21
                0.29
                         0.29
                                 0.42
                                         0.42
                                                 0.46
                                                         0.46
10
11
      0.28
                0.39
                         0.39
                                 0.51
                                         0.51
                                                 0.55
                                                         0.55
      0.40
                0.48
                         0.48
                                 0.57
12
                                         0.57
                                                 0.59
                                                         0.59
13
      0.09
                0.20
                         0.20
                                 0.40
                                         0.40
                                                 0.52
                                                         0.52
```

# 3.3 Table 5.1

To get the data, see Table 3.1.

Automated item selection algorithm

```
> options(width = 60)
> scale <- aisp(item.scores, verbose = FALSE)</pre>
```

Construction of Table 5.1 (advanced R code).

```
> options(width = 60)
> scale.1 <- c(12, 8, 1, 11, 9, 3, 10)
> scale.2 <- c(7, 5)
> Hi.top <- matrix(NA, 8, 6)
> for (i in 1:6) Hi.top[1:(i + 1), i] <- round(coefH(item.scores[, + scale.1[1:(i + 1)]], FALSE)$Hi, 2)
> for (i in 1:6) Hi.top[8, i] <- round(coefH(item.scores[, + scale.1[1:(i + 1)]], FALSE)$H, 2)
> dimnames(Hi.top)[[2]] <- paste("Step", 1:6)
> Table.5.1.top <- data.frame(Task = c(Task[scale.1], + "Total H"), Property = c(Property[scale.1], + ""), Format = c(Format[scale.1], ""), Pi = c(round(apply(item.scores[, + scale.1], 2, mean), 2), NA))
> Table.5.1.top <- cbind(Table.5.1.top, Hi.top)
> Table.5.1.top
```

	Task Property						Format Pi				Step 1	Step 2
T06A	6	area			${\tt YA}$	>	YΒ	>	${\tt YC}$	0.97	0.78	0.76
T07L	7	length			${\tt YA}$	>	YΒ	=	${\tt YC}$	0.84	0.78	0.59
T09L	9	length	$\mathtt{Y}\mathtt{A}$	=	YΒ	<	YC	=	YD	0.30	NA	0.53
W8OT	8	weight			YA	>	YΒ	=	YC	0.97	NA	NA

```
T03W
            3
                weight
                             YA > YB > YC 0.88
                                                      NA
                                                             NA
                weight YA = YB < YC = YD 0.52
T10W
           10
                                                      NA
                                                             NA
T01L
            1
                length
                             YA > YB > YC 0.94
                                                             NA
                                                      NA
     Total H
                                                   0.78
                                                           0.60
                                              NA
     Step 3 Step 4 Step 5 Step 6
T06A
       0.66
               0.64
                       0.64
                               0.59
       0.56
               0.50
                       0.51
T07L
                               0.51
                       0.51
T09L
       0.60
               0.57
                               0.50
W80T
       0.59
               0.62
                       0.61
                              0.55
T03W
         NA
               0.51
                       0.53
                              0.53
T10W
         NA
                 NA
                       0.52
                              0.52
T01L
         NA
                 NA
                         NA
                               0.46
       0.59
                               0.52
               0.55
                       0.53
```

#### 3.4 Table 5.2

Get the data, and dichotomize the scores, compute the P-values

- > data(cavalini)
- > X <- cavalini
- > X[cavalini < 2] <- 0
- > X[cavalini > 1] <- 1</pre>
- > apply(X, 2, mean)

```
Item1
                Item2
                            Item3
                                       Item4
                                                   Item5
0.60024155 0.43478261 0.59057971 0.42632850 0.20289855
                Item7
                            Item8
                                        Item9
                                                  Item10
0.08937198 0.04710145 0.16062802 0.06280193 0.51932367
                           Item13
    Item11
               Item12
                                      Item14
                                                  Item15
0.21618357 0.28623188 0.13647343 0.22342995 0.16304348
    Item16
               Item17
0.16666667 0.68599034
```

Make the table (advanced R code)

```
> Table.5.2 <- data.frame(1:17, attributes(X)$labels,
+ round(apply(X, 2, mean), 2))
> dimnames(Table.5.2)[[2]] <- c("Item.number", "Item.text",
+ "Pi")
> rownames(Table.5.2) <- NULL
> Table.5.2
```

```
2
                           no laundry outside 0.43
                   search source of malodour 0.59
3
             3
4
             4
                          no blankets outside 0.43
5
             5
                   try to find out solutions 0.20
6
             6
                   go elsewhere to fresh air 0.09
7
             7
                     call environment agency 0.05
             8
8
                     think of something else 0.16
             9
                  file complaint at producer 0.06
9
10
            10
                acquiesce in odour annoyance 0.52
            11 do something to get rid of it 0.22
11
12
            12 say: it might have been worse 0.29
13
                            experience unrest 0.14
                  talk to friends and family 0.22
14
            14
15
            15
                               seek diversion 0.16
                         avoid nose breathing 0.17
16
            16
17
            17
                   try to adapt to situation 0.69
```

## 3.5 Table 5.3

Get the data, and dichotomize the scores, see previous example

Automated item selection algorithm with different values for the lower bound.

```
> aisp(X, lowerbound = 0, verbose = FALSE)
> aisp(X, lowerbound = 0.05, verbose = FALSE)
> aisp(X, lowerbound = 0.1, verbose = FALSE)
```

Make the table (advanced R code)

```
> lower.bound <- seq(0, 0.6, by = 0.05)
> scaling.results <- matrix(NA, length(lower.bound),
+ ncol(X))
> for (i in 1:length(lower.bound)) scaling.results[i,
+ ] <- aisp(X, lowerbound = lower.bound[i],
+ verbose = FALSE)
> equal <- function(x, n) which(x == n)
> scale.1 <- sapply(apply(scaling.results, 1, "equal",
+ 1), paste, collapse = " ")
> scale.2 <- sapply(apply(scaling.results, 1, "equal",
+ 2), paste, collapse = " ")
> scale.3 <- sapply(apply(scaling.results, 1, "equal",
+ 3), paste, collapse = " ")
> scale.4 <- sapply(apply(scaling.results, 1, "equal",</pre>
```

```
4), paste, collapse = " ")
> scale.5 <- sapply(apply(scaling.results, 1, "equal",
      5), paste, collapse = " ")
> Table.5.3 <- data.frame(lower.bound, scale.1,
      scale.2, scale.3, scale.4, scale.5)
> Table.5.3
   lower.bound
                                    scale.1
                                                  scale.2
          0.00 1 2 3 4 5 6 7 8 9 11 13 14
1
                                              10 12 15 17
2
          0.05 1 2 3 4 5 6 7 8 9 11 13 14
                                              10 12 15 17
3
          0.10
                  1 2 3 4 5 6 7 9 11 13 14 8 10 12 15 17
4
          0.15
                 1 2 3 4 5 6 7 9 11 13 14 8 10 12 15 17
          0.20
                  1 2 3 4 5 6 7 9 11 13 14 8 10 12 15 17
5
          0.25
                      1 3 5 6 7 9 11 13 14
6
7
          0.30
                              3 5 6 7 9 11
                                                 1 2 4 13
          0.35
                                3 5 7 9 11
                                                     1 2 4
8
          0.40
                                3 5 7 9 11
                                                     1 2 4
9
10
          0.45
                                3 5 7 9 11
                                                     1 2 4
          0.50
                                3 5 7 9 11
                                                     1 2 4
11
12
          0.55
                                     7 9 11
                                                       2 4
                                       9 11
                                                     3 5 7
13
          0.60
         scale.3 scale.4 scale.5
1
2
3
4
5
   8 10 12 15 17
6
7
         8 15 17
                    10 12
         8 15 17
8
                    13 14
                            10 12
9
            8 17
                    13 14
                            10 12
10
            8 17
                    13 14
                            10 12
            8 17
                    13 14
11
12
             3 5
                      1 6
13
             2 4
```

## 3.6 Table 5.4

Get the data, and dichotomize the scores, see previous example

Automated item selection algorithm with two different values for the lower bound.

```
> scale.3 <- aisp(X, lowerbound = 0.3)
```

```
Make the table (advanced R code)
    > scale.30 <- aisp(X, lowerbound = 0.3, verbose = F)
    > max.scale <- max(scale.30)</pre>
    > Table.5.4.left <- data.frame()</pre>
    > for (i in 1:max.scale) {
          max.item <- max(length(scale.30[scale.30 ==</pre>
               i]))
          Scale \leftarrow c(i, rep("", max.item - 1))
          Item.30 <- which(scale.30 == i)
          Hi.30 <- round(coefH(X[, scale.30 == i], FALSE)$Hi,</pre>
          H.30 \leftarrow c(rep("", max.item - 1), round(coefH(X[,
              scale.30 == i], FALSE)$H, 2))
          Table.5.4.left <- rbind(Table.5.4.left, data.frame(Scale = Scale,
               Item = Item.30, Hi = Hi.30, H = H.30),
              c("", "", "", ""))
    + }
   > rownames(Table.5.4.left) <- NULL
    > Table.5.4.left
       Scale Item Hi
                           Η
   1
           1
                3 0.57
   2
                5 0.51
                 6 0.3
   3
                7 0.53
    4
   5
                9 0.5
                11 0.45 0.47
    6
    7
           2
                1 0.52
   8
                2 0.61
   9
                4 0.61
   10
    11
                13 0.3 0.55
   12
    13
                8 0.41
               15 0.38
    14
               17 0.49 0.42
    15
    16
    17
               10 0.47
               12 0.47 0.47
    18
```

> scale.35 <- aisp(X, lowerbound = 0.35)

19

```
> scale.35 <- aisp(X, lowerbound = 0.35, verbose = F)</pre>
> max.scale <- max(scale.35)</pre>
> Table.5.4.right <- data.frame()</pre>
> for (i in 1:max.scale) {
      max.item <- max(length(scale.35[scale.35 ==</pre>
           i]))
      Scale <- c(i, rep("", max.item - 1))</pre>
      Item.35 <- which(scale.35 == i)
      Hi.35 <- round(coefH(X[, scale.35 == i], FALSE)$Hi,</pre>
           2)
      H.35 \leftarrow c(rep("", max.item - 1), round(coefH(X[,
           scale.35 == i], FALSE)$H, 2))
      Table.5.4.right <- rbind(Table.5.4.right,</pre>
           data.frame(Scale = Scale, Item = Item.35,
               Hi = Hi.35, H = H.35), c("", "", "",
               ""))
+ }
> rownames(Table.5.4.right) <- NULL</pre>
> Table.5.4.right
   Scale Item
                        Η
                 Ηi
             3 0.6
1
       1
2
             5 0.53
3
             7 0.6
4
             9 0.63
5
            11 0.49 0.55
6
7
       2
             1 0.54
             2 0.68
8
             4 0.67 0.64
9
10
11
            8 0.41
12
            15 0.38
            17 0.49 0.42
13
14
           13 0.53
15
       4
            14 0.53 0.53
16
17
18
            10 0.47
            12 0.47 0.47
19
20
```

#### 3.7 Table 6.1

Total

Get the data. The two pseudo task. Item 2 (column 3) and item 4 (column5) were not considered. Also, the first column (Group) is removed from the data. Tasks 3 and 4 (items 5 and 9) were investigated in detail. This is the item pair number 21.

```
> library(mokken)
> data(transreas)
> X <- transreas[, -c(1, 3, 5)]
> check.restscore(X, minsize = 2)$results[[21]]
[[1]]
[[1]]$FIRST.ITEM
[1] "TO4W"
[[1]] $SECOND.ITEM
[1] "TO3W"
$SUMMARY.MATRIX
                         E(X3)
                                    E(X7) P(X3>=1)
     Group Lo Hi
                   N
[1,]
               1
                   2 0.5000000 0.0000000 0.5000000
[2,]
               2
                   4 0.5000000 0.0000000 0.5000000
[3,]
         3
            3
               3
                   8 0.7500000 0.3750000 0.7500000
[4,]
         4
            4
                  27 0.7407407 0.6296296 0.7407407
           5 5 75 0.7600000 0.8266667 0.7600000
[5,]
         5
[6,]
         6
              6 127 0.8110236 0.9370079 0.8110236
[7,]
         7
            7 7 117 0.7863248 0.9487179 0.7863248
            8 8 65 0.8000000 0.9846154 0.8000000
[8,]
      P(X7>=1)
[1,] 0.0000000
[2,] 0.0000000
[3,] 0.3750000
[4,] 0.6296296
[5,] 0.8266667
[6,] 0.9370079
[7,] 0.9487179
[8,] 0.9846154
$VIOLATION.MATRIX
                             #vi/#ac maxvi
                  #ac #vi
P(X3>=1) P(X7>=1)
                        4 0.5714286
                                       0.5 1.486111
                    7
```

4 0.5714286

0.5 1.486111

7

```
sum/#ac    zmax #zsig
P(X3>=1) P(X7>=1) 0.2123016 0.8913385 0
Total
               0.2123016 0.8913385
$Z.SCORES
                Rest-score 1-1 Rest-score 2-2
P(X3>=1) P(X7>=1) 0 0.69194
                Rest-score 3-3 Rest-score 4-4
P(X3>=1) P(X7>=1) 0.8913385 0.5142975
                Rest-score 5-5 Rest-score 6-6
P(X3>=1) P(X7>=1)
                          0
                Rest-score 7-7 Rest-score 8-8
P(X3>=1) P(X7>=1)
                            0
                                         0
> check.restscore(X, minsize = 40)$results[[21]]
[[1]]
[[1]]$FIRST.ITEM
[1] "TO4W"
[[1]]$SECOND.ITEM
[1] "TO3W"
$SUMMARY.MATRIX
    Group Lo Hi N E(X3)
                              E(X7) P(X3>=1)
[1,] 1 1 4 41 0.7073171 0.4878049 0.7073171
[2,]
      2 5 5 75 0.7600000 0.8266667 0.7600000
[3,]
      3 6 6 127 0.8110236 0.9370079 0.8110236
[4,]
      4 7 7 117 0.7863248 0.9487179 0.7863248
      5 8 8 65 0.8000000 0.9846154 0.8000000
[5,]
    P(X7 >= 1)
[1,] 0.4878049
[2,] 0.8266667
[3,] 0.9370079
[4,] 0.9487179
[5,] 0.9846154
$VIOLATION.MATRIX
                #ac #vi #vi/#ac maxvi
                4 1 0.25 0.2195122 0.2195122
P(X3>=1) P(X7>=1)
                  4 1 0.25 0.2195122 0.2195122
Total
                   sum/#ac
                             zmax #zsig
P(X3>=1) P(X7>=1) 0.05487805 1.679342 1
```

```
Total 0.05487805 1.679342 1
```

#### \$Z.SCORES

Rest-score 1-4 Rest-score 5-5 P(X3>=1) P(X7>=1) 1.679342 0

Rest-score 6-6 Rest-score 7-7

P(X3>=1) P(X7>=1) 0 0

Rest-score 8-8

P(X3>=1) P(X7>=1) 0

> plot(check.restscore(X, minsize = 2), item.pairs = 21)

> plot(check.restscore(X, minsize = 40), item.pairs = 21)

 $> R \leftarrow apply(X[, -c(3, 7)], 1, sum)$ 

> table(X[, 3], X[, 7], R)

, , R = 1

0 1

0 1 0

1 1 0

, , R = 2

0 1

0 2 0

1 2 0

, , R = 3

0 1

0 1 1

1 4 2

, R = 4

0 1

0 1 6

1 9 11

, , R = 5

```
0 1
  0
   0 18
  1 13 44
, , R = 6
    0
      1
 0
    1 23
    7 96
, , R = 7
    0 1
    2 23
    4 88
, R = 8
    0 1
  0
    0 13
  1
    1 51
> as.numeric(table(X[, 3][R < 5], X[, 7][R < 5]))
[1] 5 16 7 13
```

# 3.8 Table 6.2

Get the data. The two pseudo task. Item 2 (column 3) and item 4 (column5) were not considered. Also, the first column (Group) is removed from the data.

```
> library(mokken)
> data(transreas)
> X <- transreas[, -c(1, 3, 5)]
> Task <- c(9, 10, 4, 5, 2, 7, 3, 1, 8, 6)
> ppp <- check.pmatrix(X)$results$Ppp
> dimnames(ppp) <- list(Task, Task)
> round(ppp, 2)
```

```
10
                    5
                         2
                              7
                                   3
                                        1
9
    NA 0.22 0.23 0.25 0.24 0.28 0.28 0.29 0.30 0.30
10 0.22
         NA 0.39 0.44 0.42 0.48 0.49 0.51 0.51 0.52
  0.23 0.39
              NA 0.64 0.69 0.67 0.69 0.73 0.76 0.76
5
  0.25 0.44 0.64
                   NA 0.66 0.69 0.73 0.77 0.78 0.79
  0.24 0.42 0.69 0.66
                        NA 0.68 0.72 0.76 0.78 0.80
  0.28 0.48 0.67 0.69 0.68
                             NA 0.79 0.82 0.83 0.84
  0.28 0.49 0.69 0.73 0.72 0.79
                                  NA 0.86 0.88 0.88
  0.29 0.51 0.73 0.77 0.76 0.82 0.86
                                       NA 0.92 0.93
  0.30 0.51 0.76 0.78 0.78 0.83 0.88 0.92
                                            NA 0.96
  0.30 0.52 0.76 0.79 0.80 0.84 0.88 0.93 0.96
> library(mokken)
> data(transreas)
> X <- transreas[, -c(1, 3, 5)]
> Task <- c(9, 10, 4, 5, 2, 7, 3, 1, 8, 6)
> pmm <- check.pmatrix(X)$results$Pmm
> dimnames(pmm) <- list(Task, Task)</pre>
> round(pmm, 2)
     9
               4
                    5
                         2
                              7
                                   3
          10
                                        1
    NA 0.40 0.14 0.15 0.13 0.13 0.10 0.05 0.03 0.02
10 0.40
         NA 0.08 0.12 0.09 0.12 0.09 0.04 0.03 0.02
  0.14 0.08
              NA 0.05 0.09 0.04 0.02 0.01 0.01 0.01
  0.15 0.12 0.05
                   NA 0.04 0.04 0.04 0.02 0.01 0.01
  0.13 0.09 0.09 0.04
                        NA 0.03 0.02 0.01 0.01 0.01
7
  0.13 0.12 0.04 0.04 0.03
                             NA 0.06 0.03 0.02 0.02
  0.10 0.09 0.02 0.04 0.02 0.06
                                  NA 0.04 0.02 0.02
  0.05 0.04 0.01 0.02 0.01 0.03 0.04
                                       NA 0.01 0.01
  0.03 0.03 0.01 0.01 0.01 0.02 0.02 0.01
```

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