

Confirmatory Factor Analysis III

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1 Preliminaries

In this section, the RStudio workspace and console panes are cleared of old output, variables, and other miscellaneous debris. Packages are loaded and any required data files are retrieved.

```
options(replace.assign = TRUE, width = 65, digits = 4, scipen = 4, fig.width = 4,
        fig.height = 4)
# Clear the workspace and console.
rm(list = ls(all = TRUE))
cat("\f")
```

```
# Turn off showing of significance asterisks.
options(show.signif.stars = F)
# Set the contrast option; important for ANOVAs.
options(contrasts = c("contr.sum", "contr.poly"))
how_long <- Sys.time()
set.seed(123)
library(knitr)
```

```
library(psych)
library(car)

## Loading required package: carData
##
## Attaching package: 'car'
## The following object is masked from 'package:psych':
##
##     logit

library(multcomp)

## Loading required package: mvtnorm
## Loading required package: survival
## Loading required package: TH.data
## Loading required package: MASS
##
## Attaching package: 'TH.data'
## The following object is masked from 'package:MASS':
##
##     geyser
```

```

library(ggplot2)

##
## Attaching package: 'ggplot2'
## The following objects are masked from 'package:psych':
##
##    %+%, alpha

library(MASS)
library(parallel)
library(ellipse)

##
## Attaching package: 'ellipse'
## The following object is masked from 'package:car':
##
##    ellipse
## The following object is masked from 'package:graphics':
##
##    pairs

library(FactoMineR)

## Warning: package 'FactoMineR' was built under R version 3.5.1

library(PerformanceAnalytics)

## Warning: package 'PerformanceAnalytics' was built under R version 3.5.1
## Loading required package: xts
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##    as.Date, as.Date.numeric
##
## Attaching package: 'PerformanceAnalytics'
## The following object is masked from 'package:graphics':
##
##    legend

library(plotpc)

## Loading required package: grid

library(sciplot)
library(GPArotation)
library(GGally)
library(MVN)

## sROC 0.1-2 loaded

library(qqplotr)
library(scatterplot3d)
library(rgl)

## Warning: package 'rgl' was built under R version 3.5.1

library(cowplot)

```

```
##
## Attaching package: 'cowplot'
## The following object is masked from 'package:ggplot2':
##
##    ggsave

library(lavaan)

## This is lavaan 0.6-1
## lavaan is BETA software! Please report any bugs.
##
## Attaching package: 'lavaan'
## The following object is masked from 'package:psych':
##
##    cor2cov

library(semPlot)
library(semTools)

##
## #####
## This is semTools 0.4-14
## All users of R (or SEM) are invited to submit functions or ideas for functions.
## #####
##
## Attaching package: 'semTools'
## The following object is masked from 'package:PerformanceAnalytics':
##
##    kurtosis
## The following object is masked from 'package:psych':
##
##    skew

library(MVN)
```

1.1 Data Files

We will use the mental abilities data sets (full and split), the need for cognition data set, the self-esteem data set, and the HolzingerSwineford data set for the example analyses.

```
# Get the drug use data from the working directory.
setwd("C:\\Courses\\Psychology 516\\PowerPoint\\2018")
Mental <- read.table("mental.csv", sep = ",", header = TRUE)
Mental <- as.data.frame(Mental)
Mental <- na.omit(Mental)
names(Mental) <- c("Grammar", "Paragraph_Comprehension", "Vocabulary",
  "Sentence_Completion", "Geometry", "Algebra", "Numerical_Puzzles",
  "Series_Completion", "Practical_Problem_Solving", "Symbol_Manipulation",
  "Analytical_Ability", "Formal_Logic")
Mental_1 <- read.table("mental_1.csv", sep = ",", header = TRUE)
Mental_1 <- as.data.frame(Mental_1)
Mental_1 <- na.omit(Mental_1)
names(Mental_1) <- c("Grammar", "Paragraph_Comprehension", "Vocabulary",
  "Sentence_Completion", "Geometry", "Algebra", "Numerical_Puzzles",
```

```

    "Series_Completion", "Practical_Problem_Solving", "Symbol_Manipulation",
    "Analytical_Ability", "Formal_Logic")
Mental_2 <- read.table("mental_2.csv", sep = ",", header = TRUE)
Mental_2 <- as.data.frame(Mental_2)
Mental_2 <- na.omit(Mental_2)
names(Mental_2) <- c("Grammar", "Paragraph_Comprehension", "Vocabulary",
    "Sentence_Completion", "Geometry", "Algebra", "Numerical_Puzzles",
    "Series_Completion", "Practical_Problem_Solving", "Symbol_Manipulation",
    "Analytical_Ability", "Formal_Logic")
Mental_G <- read.table("mental_groups.csv", sep = ",", header = TRUE)
Mental_G <- as.data.frame(Mental_G)
Mental_G <- na.omit(Mental_G)
names(Mental_G) <- c("Grammar", "Paragraph_Comprehension", "Vocabulary",
    "Sentence_Completion", "Geometry", "Algebra", "Numerical_Puzzles",
    "Series_Completion", "Practical_Problem_Solving", "Symbol_Manipulation",
    "Analytical_Ability", "Formal_Logic", "Group")
Mental_G$Group <- as.factor(Mental_G$Group)
NC <- read.table("need_for_cognition.csv", sep = ",", header = TRUE)
NC <- as.data.frame(NC[, 1:18])
NC <- na.omit(NC)
names(NC) <- c("nc1", "nc2", "nc3", "nc4", "nc5", "nc6", "nc7", "nc8",
    "nc9", "nc10", "nc11", "nc12", "nc13", "nc14", "nc15", "nc16",
    "nc17", "nc18")
SE <- read.table("Set_4.csv", sep = ",", header = TRUE)
SE <- as.data.frame(SE)
SE <- na.omit(SE[, 2:21])

HS <- as.data.frame(HolzingerSwineford1939)
HS <- HS[, c(5, 7:15)]
HS <- na.omit(HS)

```

2 Screen the Data

For each data set we do a quick screen to make sure there are no severe multivariate outliers or multivariate normality problems.

2.1 Mental Abilities

First we will do a quick screen of the data to make sure there are no severe multivariate outliers.

2.1.1 Mahalanobis Distance

```

CV <- cov(Mental)
D2 <- mahalanobis(Mental, center = colMeans(Mental), cov = CV)
D2 <- as.data.frame(D2)
describe(D2)

##      vars   n  mean   sd median trimmed  mad  min   max range skew
## X1      1 500 11.98 4.75   11.4   11.65 4.66 2.74 28.45  25.7 0.61
##      kurtosis   se
## X1          0.05 0.21

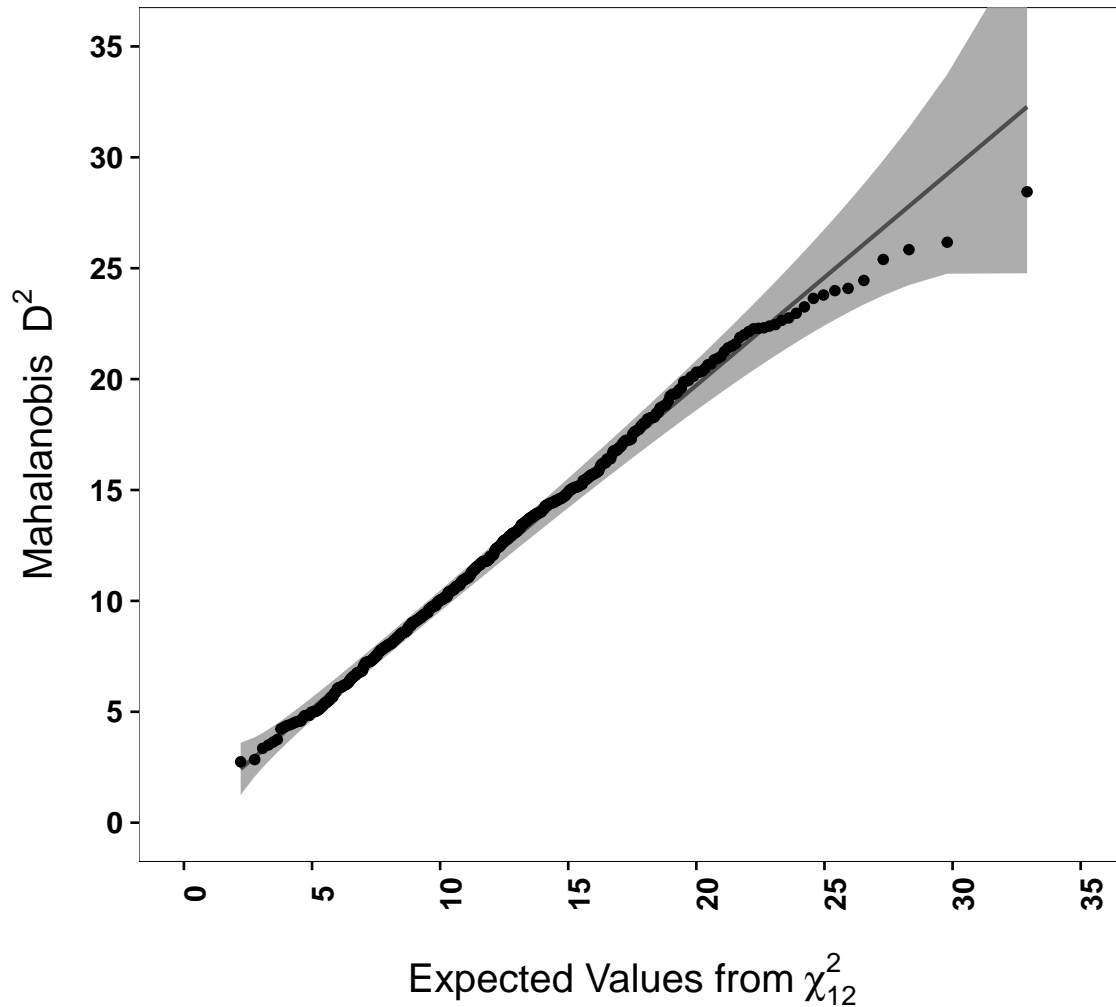
```

```

ggplot(D2, aes(sample = D2)) + stat_qq_band(distribution = "chisq",
  dparams = list(df = 12)) + stat_qq_line(distribution = "chisq",
  dparams = list(df = 12)) + stat_qq(distribution = "qchisq", dparams = list(df = 12)) +
  scale_y_continuous(breaks = seq(0, 35, 5)) + scale_x_continuous(breaks = seq(0,
  35, 5)) + coord_cartesian(xlim = c(0, 35), ylim = c(0, 35)) +
  xlab(expression("Expected Values from" *  $\sim \chi^2_{12}$ )) + ylab(expression("Mahalanobis " *
   $\sim D^2$ )) + theme(text = element_text(size = 14, family = "sans",
  color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
  size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
  size = 12, face = "bold", angle = 90), axis.title.x = element_text(margin = margin(15,
  0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
  15, 0, 0), size = 16), axis.line.x = element_blank(), axis.line.y = element_blank(),
  plot.title = element_text(size = 16, face = "bold", margin = margin(0,
  0, 20, 0), hjust = 0.5), panel.background = element_rect(fill = "white",
  linetype = 1, color = "black"), panel.grid.major = element_blank(),
  panel.grid.minor = element_blank(), plot.background = element_rect(fill = "white"),
  plot.margin = unit(c(1, 1, 1, 1), "cm"), legend.position = "bottom",
  legend.title = element_blank()) + ggtitle(expression("Q-Q Plot of Mahalanobis " *
   $\sim D^2$  * " vs. Quantiles of" *  $\sim \chi^2_{12}$ ))

```

Q-Q Plot of Mahalanobis D^2 vs. Quantiles of χ^2_{12}



The squared Mahalanobis distances suggest the data are well behaved.

2.1.2 Multivariate Normality

We can also test for multivariate normality.

```
mvn(Mental, mvnTest = "mardia")
```

```
## $multivariateNormality
##      Test      Statistic      p value Result
## 1 Mardia Skewness  367.600071004817 0.437353917013108 YES
## 2 Mardia Kurtosis -0.866369530731211 0.386287547416061 YES
## 3      MVN      <NA>      <NA>      YES
##
## $univariateNormality
```

```

##          Test          Variable Statistic  p value
## 1 Shapiro-Wilk      Grammar      0.9955    0.1597
## 2 Shapiro-Wilk Paragraph_Comprehension  0.9980    0.8256
## 3 Shapiro-Wilk      Vocabulary      0.9951    0.1166
## 4 Shapiro-Wilk Sentence_Completion      0.9972    0.5408
## 5 Shapiro-Wilk      Geometry      0.9961    0.2666
## 6 Shapiro-Wilk      Algebra      0.9964    0.3168
## 7 Shapiro-Wilk Numerical_Puzzles      0.9966    0.3863
## 8 Shapiro-Wilk Series_Completion      0.9959    0.2195
## 9 Shapiro-Wilk Practical_Problem_Solving  0.9974    0.6374
## 10 Shapiro-Wilk Symbol_Manipulation      0.9962    0.2767
## 11 Shapiro-Wilk Analytical_Ability      0.9971    0.5221
## 12 Shapiro-Wilk Formal_Logic      0.9961    0.2500
## Normality
## 1 YES
## 2 YES
## 3 YES
## 4 YES
## 5 YES
## 6 YES
## 7 YES
## 8 YES
## 9 YES
## 10 YES
## 11 YES
## 12 YES
##
## $Descriptives
##          n      Mean Std.Dev   Median   Min
## Grammar      500 -0.007632  0.9765 -0.019820 -3.003
## Paragraph_Comprehension  500 -0.003750  1.0633  0.035079 -3.024
## Vocabulary      500 -0.014783  0.9989  0.023294 -3.492
## Sentence_Completion      500 -0.087844  1.0863 -0.087127 -3.092
## Geometry      500  0.010634  0.9789 -0.028235 -2.522
## Algebra      500 -0.028644  0.9615 -0.002082 -2.906
## Numerical_Puzzles      500 -0.015062  0.9890 -0.062496 -2.804
## Series_Completion      500 -0.006348  0.9638 -0.029178 -3.038
## Practical_Problem_Solving  500 -0.016828  0.9361 -0.040021 -2.636
## Symbol_Manipulation      500  0.000833  0.9718 -0.056943 -3.073
## Analytical_Ability      500 -0.004649  0.9834  0.009074 -3.751
## Formal_Logic      500 -0.062950  1.0165 -0.092532 -3.197
##          Max    25th   75th    Skew
## Grammar      2.235 -0.6782  0.6886 -0.05736
## Paragraph_Comprehension  2.952 -0.7485  0.7003 -0.09118
## Vocabulary      2.763 -0.6567  0.6660 -0.24969
## Sentence_Completion      3.005 -0.8542  0.6525  0.01663
## Geometry      2.779 -0.6281  0.6738  0.10710
## Algebra      2.893 -0.6320  0.5580 -0.07334
## Numerical_Puzzles      2.881 -0.5965  0.6013  0.01468
## Series_Completion      2.288 -0.6444  0.6381 -0.11470
## Practical_Problem_Solving  3.033 -0.6044  0.6685  0.12662
## Symbol_Manipulation      2.740 -0.6110  0.6397  0.08702
## Analytical_Ability      2.636 -0.6867  0.6939 -0.15307
## Formal_Logic      3.729 -0.6654  0.5589  0.05588

```

```
##           Kurtosis
## Grammar          -0.210884
## Paragraph_Comprehension -0.185360
## Vocabulary       -0.047192
## Sentence_Completion -0.238030
## Geometry         -0.174019
## Algebra           0.171153
## Numerical_Puzzles  0.046084
## Series_Completion -0.214526
## Practical_Problem_Solving -0.019846
## Symbol_Manipulation 0.120057
## Analytical_Ability 0.001184
## Formal_Logic      0.441237
```

The data are multivariate normal, which is of course not surprising given that they were generated from a multivariate normal distribution.

2.2 Need for Cognition

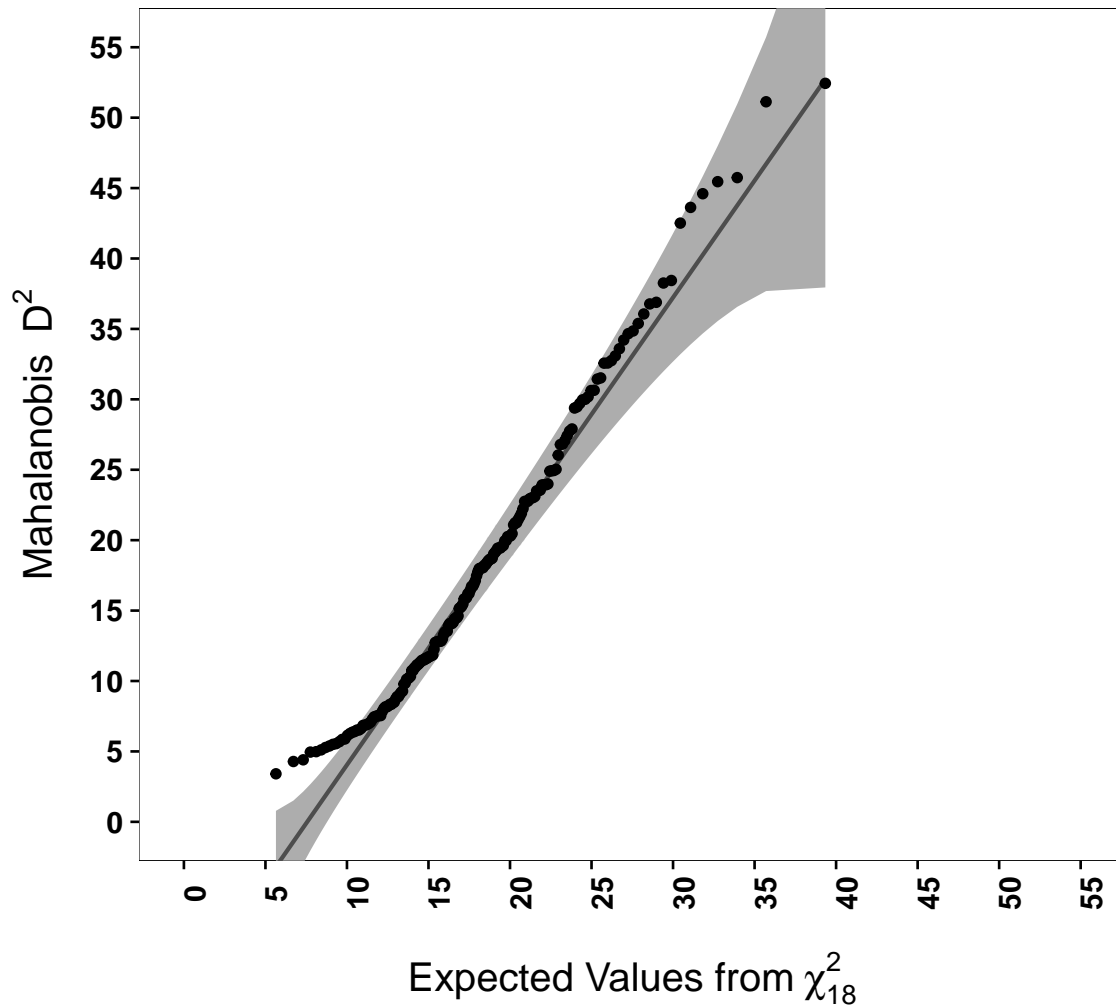
2.2.1 Mahalanobis Distance

```
CV <- cov(NC)
D2 <- mahalanobis(NC, center = colMeans(NC), cov = CV)
D2 <- as.data.frame(D2)
describe(D2)

##      vars   n mean    sd median trimmed   mad  min   max range
## X1      1 195 17.91 10.18  15.92    16.8 10.15  3.41 52.44 49.03
##      skew kurtosis   se
## X1 0.97      0.61 0.73

ggplot(D2, aes(sample = D2)) + stat_qq_band(distribution = "chisq",
  dparams = list(df = 18)) + stat_qq_line(distribution = "chisq",
  dparams = list(df = 18)) + stat_qq(distribution = "qchisq", dparams = list(df = 18)) +
  scale_y_continuous(breaks = seq(0, 55, 5)) + scale_x_continuous(breaks = seq(0,
  55, 5)) + coord_cartesian(xlim = c(0, 55), ylim = c(0, 55)) +
  xlab(expression("Expected Values from" * ~chi[18]^2)) + ylab(expression("Mahalanobis " *
  ~D^2)) + theme(text = element_text(size = 14, family = "sans",
  color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
  size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
  size = 12, face = "bold", angle = 90), axis.title.x = element_text(margin = margin(15,
  0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
  15, 0, 0), size = 16), axis.line.x = element_blank(), axis.line.y = element_blank(),
  plot.title = element_text(size = 16, face = "bold", margin = margin(0,
  0, 20, 0), hjust = 0.5), panel.background = element_rect(fill = "white",
  linetype = 1, color = "black"), panel.grid.major = element_blank(),
  panel.grid.minor = element_blank(), plot.background = element_rect(fill = "white"),
  plot.margin = unit(c(1, 1, 1, 1), "cm"), legend.position = "bottom",
  legend.title = element_blank()) + ggtitle(expression("Q-Q Plot of Mahalanobis " *
  ~D^2 * " vs. Quantiles of" * ~chi[18]^2))
```


Q-Q Plot of Mahalanobis D^2 vs. Quantiles of χ^2_{18}



The squared Mahalanobis distances suggest no multivariate outliers but the data appear to violate multivariate normality.

2.2.2 Multivariate Normality

We can also test for multivariate normality.

```
mvn(NC, mvnTest = "mardia")
```

```
## $multivariateNormality
##      Test      Statistic      p value Result
## 1 Mardia Skewness 2396.78535052338 1.67732326269208e-91 NO
## 2 Mardia Kurtosis 17.7171063217399      0      NO
## 3      MVN      <NA>      <NA>      NO
##
```

```
## $univariateNormality
##           Test Variable Statistic   p value Normality
## 1  Shapiro-Wilk   nc1      0.8666 <0.001      NO
## 2  Shapiro-Wilk   nc2      0.8308 <0.001      NO
## 3  Shapiro-Wilk   nc3      0.7518 <0.001      NO
## 4  Shapiro-Wilk   nc4      0.7873 <0.001      NO
## 5  Shapiro-Wilk   nc5      0.7576 <0.001      NO
## 6  Shapiro-Wilk   nc6      0.8931 <0.001      NO
## 7  Shapiro-Wilk   nc7      0.8703 <0.001      NO
## 8  Shapiro-Wilk   nc8      0.8739 <0.001      NO
## 9  Shapiro-Wilk   nc9      0.8156 <0.001      NO
## 10 Shapiro-Wilk  nc10      0.7986 <0.001      NO
## 11 Shapiro-Wilk  nc11      0.7450 <0.001      NO
## 12 Shapiro-Wilk  nc12      0.7249 <0.001      NO
## 13 Shapiro-Wilk  nc13      0.9039 <0.001      NO
## 14 Shapiro-Wilk  nc14      0.8485 <0.001      NO
## 15 Shapiro-Wilk  nc15      0.8637 <0.001      NO
## 16 Shapiro-Wilk  nc16      0.8466 <0.001      NO
## 17 Shapiro-Wilk  nc17      0.7721 <0.001      NO
## 18 Shapiro-Wilk  nc18      0.8907 <0.001      NO
##
## $Descriptives
##           n Mean Std.Dev Median Min Max 25th 75th      Skew Kurtosis
## nc1  195 3.241  1.2387      4   1   5   2.0   4 -0.5261 -0.8380
## nc2  195 3.913  1.0242      4   1   5   3.5   5 -0.9719  0.4887
## nc3  195 1.821  1.0618      1   1   5   1.0   2  1.3361  1.0606
## nc4  195 1.974  1.1462      2   1   5   1.0   2  1.1327  0.3874
## nc5  195 1.831  1.0439      2   1   5   1.0   2  1.2589  0.7722
## nc6  195 3.287  1.2012      4   1   5   2.0   4 -0.4373 -0.7488
## nc7  195 2.487  1.3174      2   1   5   1.0   4  0.4179 -1.0842
## nc8  195 2.518  1.2156      2   1   5   2.0   4  0.5151 -0.8168
## nc9  195 2.267  1.2644      2   1   5   1.0   3  0.8479 -0.4672
## nc10 195 4.056  0.9481      4   1   5   4.0   5 -1.1948  1.4206
## nc11 195 4.205  0.9627      4   1   5   4.0   5 -1.5193  2.2967
## nc12 195 1.754  1.0605      1   1   5   1.0   2  1.4276  1.2688
## nc13 195 3.287  1.1485      3   1   5   3.0   4 -0.3693 -0.6502
## nc14 195 3.749  1.0952      4   1   5   3.0   5 -0.8062 -0.1036
## nc15 195 3.677  1.0713      4   1   5   3.0   4 -0.7868  0.1836
## nc16 195 2.292  1.1892      2   1   5   1.0   3  0.7793 -0.3685
## nc17 195 1.903  1.1287      2   1   5   1.0   2  1.1535  0.3776
## nc18 195 3.328  1.2078      4   1   5   2.0   4 -0.3654 -0.9283
```

The data are not univariate or multivariate normal.

2.3 Self-Esteem Sample 1

2.3.1 Mahalanobis Distance

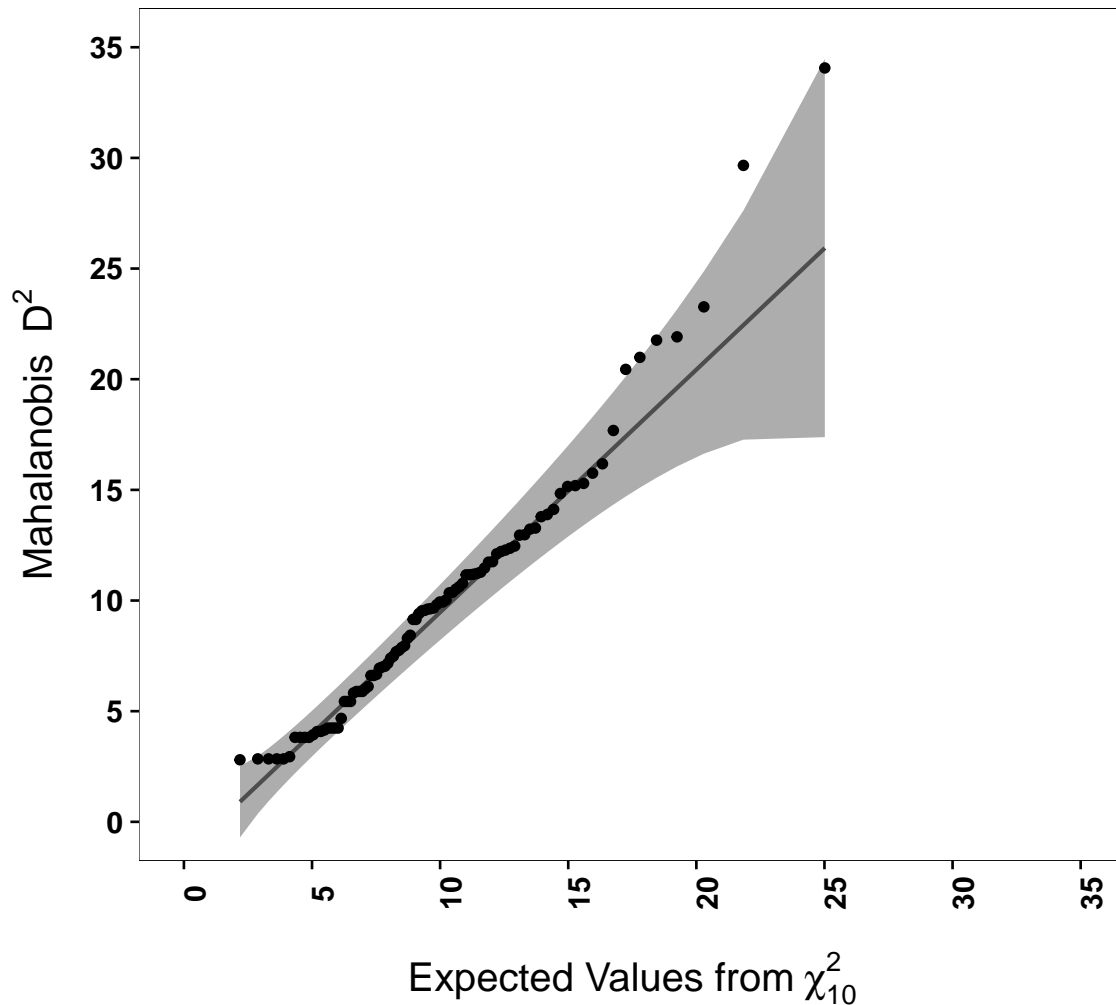
```
CV <- cov(SE[, 1:10])
D2 <- mahalanobis(SE[, 1:10], center = colMeans(SE[, 1:10]), cov = CV)
D2 <- as.data.frame(D2)
describe(D2)

##      vars  n mean    sd median trimmed  mad min    max range skew
```

```
## X1      1 94 9.89 5.72    9.54    9.15 5.08 2.8 34.06 31.26 1.51
##      kurtosis    se
## X1      3.33 0.59

ggplot(D2, aes(sample = D2)) + stat_qq_band(distribution = "chisq",
  dparams = list(df = 10)) + stat_qq_line(distribution = "chisq",
  dparams = list(df = 10)) + stat_qq(distribution = "qchisq", dparams = list(df = 10)) +
  scale_y_continuous(breaks = seq(0, 35, 5)) + scale_x_continuous(breaks = seq(0,
  35, 5)) + coord_cartesian(xlim = c(0, 35), ylim = c(0, 35)) +
  xlab(expression("Expected Values from" *  $\sim \chi[10]^2$ )) + ylab(expression("Mahalanobis " *
   $\sim D^2$ )) + theme(text = element_text(size = 14, family = "sans",
  color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
  size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
  size = 12, face = "bold", angle = 90), axis.title.x = element_text(margin = margin(15,
  0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
  15, 0, 0), size = 16), axis.line.x = element_blank(), axis.line.y = element_blank(),
  plot.title = element_text(size = 16, face = "bold", margin = margin(0,
  0, 20, 0), hjust = 0.5), panel.background = element_rect(fill = "white",
  linetype = 1, color = "black"), panel.grid.major = element_blank(),
  panel.grid.minor = element_blank(), plot.background = element_rect(fill = "white"),
  plot.margin = unit(c(1, 1, 1, 1), "cm"), legend.position = "bottom",
  legend.title = element_blank()) + ggtitle(expression("Q-Q Plot of Mahalanobis" *
   $\sim D^2$  * " vs. Quantiles of" *  $\sim \chi[10]^2$ ))
```

Q-Q Plot of Mahalanobis D^2 vs. Quantiles of χ^2_{10}



The squared Mahalanobis distances suggest no multivariate outliers but the data appear to violate multivariate normality.

2.3.2 Multivariate Normality

We can also test for multivariate normality.

```
mvn(SE[, 1:10], mvnTest = "mardia")

## $multivariateNormality
##      Test      Statistic      p value Result
## 1 Mardia Skewness 500.450121412264 6.81205353279162e-24 NO
## 2 Mardia Kurtosis 4.10286690060031 0.000040806202145971 NO
## 3      MVN          <NA>          <NA>      NO
##
```

```
## $univariateNormality
##           Test Variable Statistic   p value Normality
## 1 Shapiro-Wilk r_1_1      0.5919 <0.001      NO
## 2 Shapiro-Wilk r_1_2      0.5753 <0.001      NO
## 3 Shapiro-Wilk r_1_3      0.6562 <0.001      NO
## 4 Shapiro-Wilk r_1_4      0.7373 <0.001      NO
## 5 Shapiro-Wilk r_1_5      0.6739 <0.001      NO
## 6 Shapiro-Wilk r_1_6      0.7866 <0.001      NO
## 7 Shapiro-Wilk r_1_7      0.7878 <0.001      NO
## 8 Shapiro-Wilk r_1_8      0.8678 <0.001      NO
## 9 Shapiro-Wilk r_1_9      0.8704 <0.001      NO
## 10 Shapiro-Wilk r_1_10     0.8231 <0.001      NO
##
## $Descriptives
##           n Mean Std.Dev Median Min Max 25th 75th      Skew
## r_1_1  94 3.670  0.5937      4  1  4  3  4 -1.89147
## r_1_2  94 3.723  0.4730      4  2  4  3  4 -1.28100
## r_1_3  94 3.596  0.6103      4  2  4  3  4 -1.20663
## r_1_4  94 3.372  0.6219      3  1  4  3  4 -0.70899
## r_1_5  94 3.457  0.8252      4  1  4  3  4 -1.56957
## r_1_6  94 3.202  0.6968      3  1  4  3  4 -0.66526
## r_1_7  94 3.202  0.6653      3  1  4  3  4 -0.46013
## r_1_8  94 2.755  0.9912      3  1  4  2  4 -0.22450
## r_1_9  94 2.638  0.8902      3  1  4  2  3  0.03653
## r_1_10 94 3.064  0.9482      3  1  4  2  4 -0.57445
##           Kurtosis
## r_1_1  3.80261
## r_1_2  0.35707
## r_1_3  0.36166
## r_1_4  0.77997
## r_1_5  1.84237
## r_1_6  0.55887
## r_1_7  0.09848
## r_1_8 -1.06337
## r_1_9 -0.86502
## r_1_10 -0.81931
```

The data are not univariate or multivariate normal.

2.4 Self-Esteem Sample 2

2.4.1 Mahalanobis Distance

```
CV <- cov(SE[, 11:20])
D2 <- mahalanobis(SE[, 11:20], center = colMeans(SE[, 11:20]), cov = CV)
D2 <- as.data.frame(D2)
describe(D2)

##      vars  n mean   sd median trimmed  mad  min   max range skew
## X1      1 94 9.89 6.82   7.86    8.81 5.31 2.52 31.82 29.3 1.37
##      kurtosis se
## X1      1.22 0.7

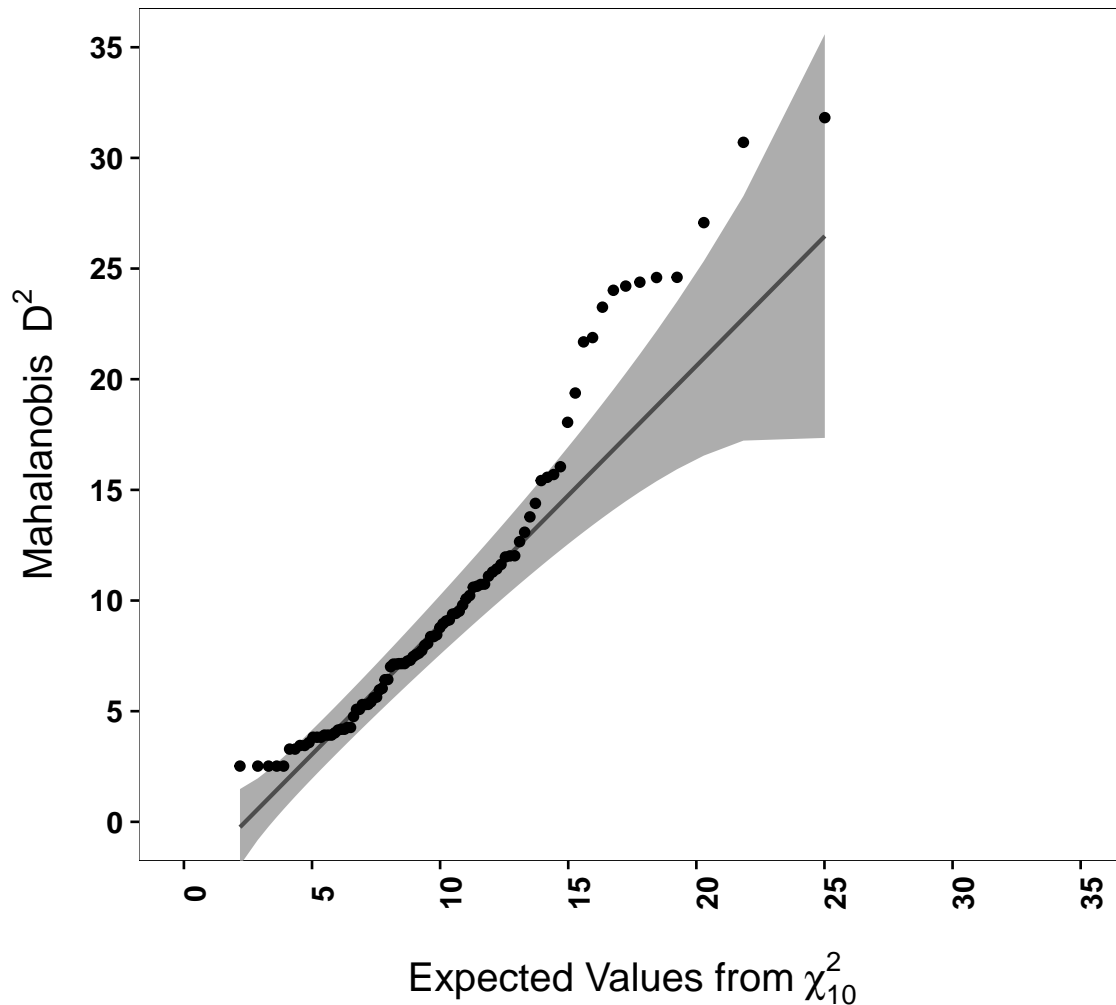
ggplot(D2, aes(sample = D2)) + stat_qq_band(distribution = "chisq",
```

```

dparams = list(df = 10)) + stat_qq_line(distribution = "chisq",
dparams = list(df = 10)) + stat_qq(distribution = "qchisq", dparams = list(df = 10)) +
scale_y_continuous(breaks = seq(0, 35, 5)) + scale_x_continuous(breaks = seq(0,
35, 5)) + coord_cartesian(xlim = c(0, 35), ylim = c(0, 35)) +
xlab(expression("Expected Values from" * ~chi[10]^2)) + ylab(expression("Mahalanobis " *
~D^2)) + theme(text = element_text(size = 14, family = "sans",
color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
size = 12, face = "bold", angle = 90), axis.title.x = element_text(margin = margin(15,
0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
15, 0, 0), size = 16), axis.line.x = element_blank(), axis.line.y = element_blank(),
plot.title = element_text(size = 16, face = "bold", margin = margin(0,
0, 20, 0), hjust = 0.5), panel.background = element_rect(fill = "white",
linetype = 1, color = "black"), panel.grid.major = element_blank(),
panel.grid.minor = element_blank(), plot.background = element_rect(fill = "white"),
plot.margin = unit(c(1, 1, 1, 1), "cm"), legend.position = "bottom",
legend.title = element_blank()) + ggtitle(expression("Q-Q Plot of Mahalanobis " *
~D^2 * " vs. Quantiles of" * ~chi[10]^2))

```

Q-Q Plot of Mahalanobis D^2 vs. Quantiles of χ^2_{10}



The squared Mahalanobis distances suggest no multivariate outliers but the data appear to violate multivariate normality.

2.4.2 Multivariate Normality

We can also test for multivariate normality.

```
mvn(SE[, 11:20], mvnTest = "mardia")

## $multivariateNormality
##      Test      Statistic      p value Result
## 1 Mardia Skewness 579.546878261195 3.64214052345009e-34 NO
## 2 Mardia Kurtosis 8.45722209724254      0      NO
## 3      MVN          <NA>          <NA>      NO
##
```

```
## $univariateNormality
##           Test Variable Statistic   p value Normality
## 1 Shapiro-Wilk r_2_1      0.5926 <0.001      NO
## 2 Shapiro-Wilk r_2_2      0.5513 <0.001      NO
## 3 Shapiro-Wilk r_2_3      0.6786 <0.001      NO
## 4 Shapiro-Wilk r_2_4      0.7554 <0.001      NO
## 5 Shapiro-Wilk r_2_5      0.6884 <0.001      NO
## 6 Shapiro-Wilk r_2_6      0.7947 <0.001      NO
## 7 Shapiro-Wilk r_2_7      0.7767 <0.001      NO
## 8 Shapiro-Wilk r_2_8      0.8491 <0.001      NO
## 9 Shapiro-Wilk r_2_9      0.8298 <0.001      NO
## 10 Shapiro-Wilk r_2_10     0.7418 <0.001      NO
##
## $Descriptives
##           n Mean Std.Dev Median Min Max 25th 75th Skew
## r_2_1  94 3.702  0.5044      4  2  4  3  4 -1.3574
## r_2_2  94 3.734  0.4442      4  3  4  3  4 -1.0425
## r_2_3  94 3.564  0.6145      4  2  4  3  4 -1.0659
## r_2_4  94 3.351  0.6988      3  1  4  3  4 -0.9632
## r_2_5  94 3.479  0.7722      4  1  4  3  4 -1.4545
## r_2_6  94 3.181  0.6873      3  2  4  3  4 -0.2421
## r_2_7  94 3.191  0.6439      3  1  4  3  4 -0.4308
## r_2_8  94 2.904  0.9624      3  1  4  2  4 -0.3124
## r_2_9  94 2.979  0.9501      3  1  4  2  4 -0.3304
## r_2_10 94 3.266  0.9636      4  1  4  3  4 -0.9710
##           Kurtosis
## r_2_1  0.80618
## r_2_2 -0.92260
## r_2_3  0.05539
## r_2_4  0.98720
## r_2_5  1.55834
## r_2_6 -0.91969
## r_2_7  0.32236
## r_2_8 -1.06265
## r_2_9 -1.15290
## r_2_10 -0.34993
```

The data are not univariate or multivariate normal.

2.5 Holzinger and Swineford (1939)

2.5.1 Mahalanobis Distance

```
CV <- cov(HS[, 2:10])
D2 <- mahalanobis(HS[, 2:10], center = colMeans(HS[, 2:10]), cov = CV)
D2 <- as.data.frame(D2)
describe(D2)

##      vars   n mean   sd median trimmed  mad  min   max range skew
## X1      1 301 8.97 4.75   8.09    8.38 3.97 1.97 28.67 26.69 1.41
##      kurtosis   se
## X1      2.66 0.27

ggplot(D2, aes(sample = D2)) + stat_qq_band(distribution = "chisq",
```

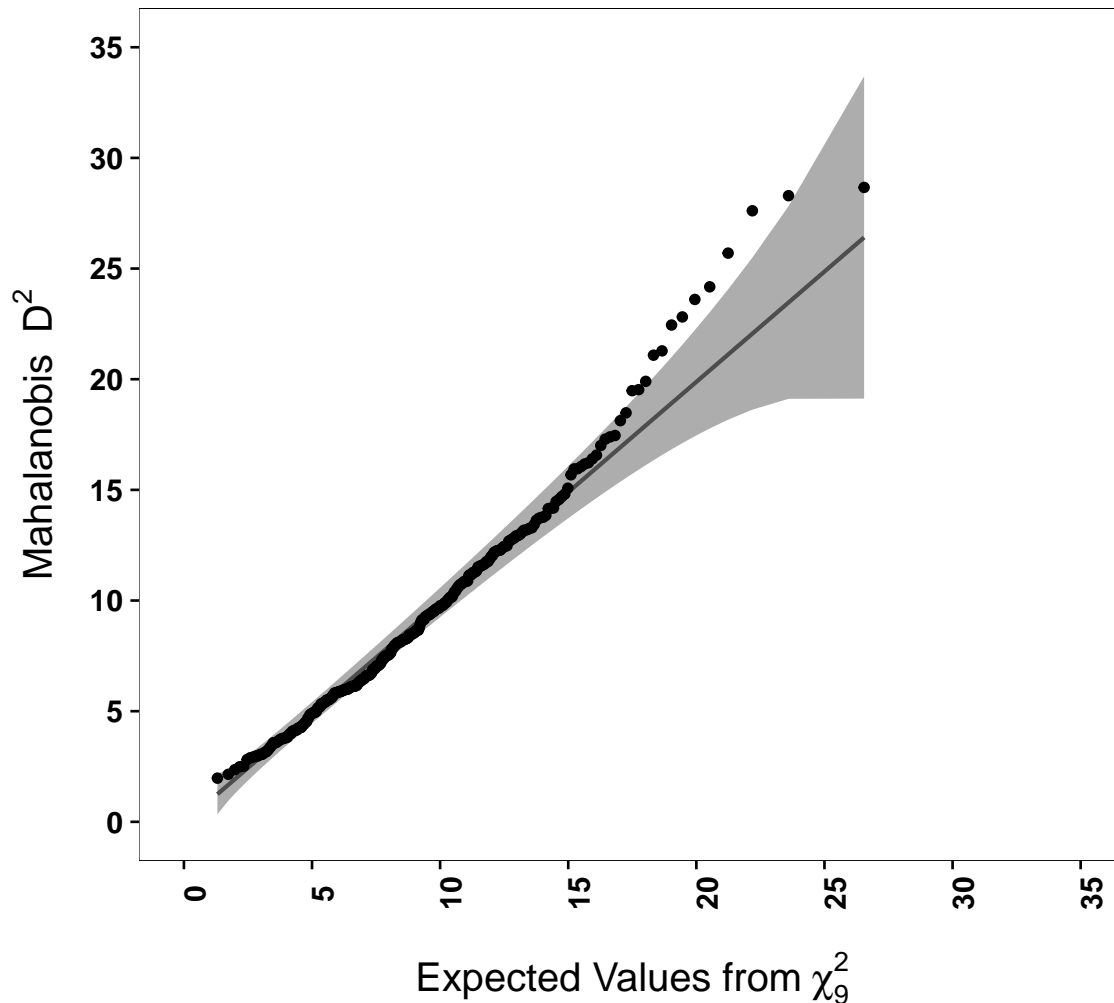


```

dparams = list(df = 9)) + stat_qq_line(distribution = "chisq",
dparams = list(df = 9)) + stat_qq(distribution = "qchisq", dparams = list(df = 9)) +
scale_y_continuous(breaks = seq(0, 35, 5)) + scale_x_continuous(breaks = seq(0,
35, 5)) + coord_cartesian(xlim = c(0, 35), ylim = c(0, 35)) +
xlab(expression("Expected Values from" * ~chi[9]^2)) + ylab(expression("Mahalanobis " *
~D^2)) + theme(text = element_text(size = 14, family = "sans",
color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
size = 12, face = "bold", angle = 90), axis.title.x = element_text(margin = margin(15,
0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
15, 0, 0), size = 16), axis.line.x = element_blank(), axis.line.y = element_blank(),
plot.title = element_text(size = 16, face = "bold", margin = margin(0,
0, 20, 0), hjust = 0.5), panel.background = element_rect(fill = "white",
linetype = 1, color = "black"), panel.grid.major = element_blank(),
panel.grid.minor = element_blank(), plot.background = element_rect(fill = "white"),
plot.margin = unit(c(1, 1, 1, 1), "cm"), legend.position = "bottom",
legend.title = element_blank()) + ggtitle(expression("Q-Q Plot of Mahalanobis " *
~D^2 * " vs. Quantiles of" * ~chi[9]^2))

```

Q-Q Plot of Mahalanobis D^2 vs. Quantiles of χ_9^2



The squared Mahalanobis distances suggest no multivariate outliers but the data appear to violate multivariate normality.

2.5.2 Multivariate Normality

We can also test for multivariate normality.

```
mvn(HS[, 2:10], mvnTest = "mardia")
```

```
## $multivariateNormality
##      Test      Statistic      p value Result
## 1 Mardia Skewness 344.905276936947 8.86473544030093e-15 NO
## 2 Mardia Kurtosis 2.83021589156651 0.00465166039260279 NO
## 3      MVN          <NA>          <NA>      NO
##
```

```
## $univariateNormality
##           Test Variable Statistic    p value Normality
## 1 Shapiro-Wilk    x1          0.9928 0.1582      YES
## 2 Shapiro-Wilk    x2          0.9697 <0.001      NO
## 3 Shapiro-Wilk    x3          0.9523 <0.001      NO
## 4 Shapiro-Wilk    x4          0.9827 0.0011      NO
## 5 Shapiro-Wilk    x5          0.9769 0.0001      NO
## 6 Shapiro-Wilk    x6          0.9538 <0.001      NO
## 7 Shapiro-Wilk    x7          0.9908 0.056       YES
## 8 Shapiro-Wilk    x8          0.9807 0.0004      NO
## 9 Shapiro-Wilk    x9          0.9942 0.307       YES
##
## $Descriptives
##      n Mean Std.Dev Median    Min    Max 25th 75th  Skew
## x1 301 4.936  1.167  5.000 0.6667  8.500 4.167 5.667 -0.2543
## x2 301 6.088  1.177  6.000 2.2500  9.250 5.250 6.750  0.4701
## x3 301 2.250  1.131  2.125 0.2500  4.500 1.375 3.125  0.3834
## x4 301 3.061  1.164  3.000 0.0000  6.333 2.333 3.667  0.2675
## x5 301 4.341  1.290  4.500 1.0000  7.000 3.500 5.250 -0.3498
## x6 301 2.186  1.096  2.000 0.1429  6.143 1.429 2.714  0.8579
## x7 301 4.186  1.090  4.087 1.3043  7.435 3.478 4.913  0.2491
## x8 301 5.527  1.013  5.500 3.0500 10.000 4.850 6.100  0.5253
## x9 301 5.374  1.009  5.417 2.7778  9.250 4.750 6.083  0.2039
##      Kurtosis
## x1  0.30753
## x2  0.33239
## x3 -0.90753
## x4  0.08013
## x5 -0.55254
## x6  0.81656
## x7 -0.30740
## x8  1.17156
## x9  0.28991
```

The data are not univariate or multivariate normal.

3 CFA With Two Approaches to Scaling

The scale for the latent variables can either be set by using one of the manifest variables or by specifying the variances of the latent variables directly (usually by standardizing them). The two approaches will produce identical fits to the data but will differ in the particular parameters that get estimated.

```
scaling.model.1 <- "
# Latent variable definitions.
# Scale of the latent variables is set by the first listed manifest variable.
Verbal =~ Grammar+Paragraph_Comprehension+Vocabulary+Sentence_Completion
Math =~ Geometry+Algebra+Numerical_Puzzles+Series_Completion
Reasoning =~ Practical_Problem_Solving+Symbol_Manipulation+Analytical_Ability+Formal_Logic
# Latent variable covariances.
Verbal ~~ Math
Verbal ~~ Reasoning
Math ~~ Reasoning
```

```

"

scaling.model.2 <- "
# Latent variable definitions.
# Scale of the latent variables is set directly.
Verbal =~ NA*Grammar+Paragraph_Comprehension+Vocabulary+Sentence_Completion
Math =~ NA*Geometry+Algebra+Numerical_Puzzles+Series_Completion
Reasoning =~ NA*Practical_Problem_Solving+Symbol_Manipulation+Analytical_Ability+Formal_Logic
# Latent variable covariances.
Verbal ~~ Math
Verbal ~~ Reasoning
Math ~~ Reasoning
# Latent variable variances.
Verbal ~~ 1*Verbal
Math ~~ 1*Math
Reasoning ~~ 1*Reasoning
"

```

```

CFA_Fit_1 <- cfa(scaling.model.1, data = Mental, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(CFA_Fit_1, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

```

```

## lavaan (0.6-1) converged normally after 29 iterations
##
##   Number of observations              500
##   Number of missing patterns          1
##
##   Estimator              ML          Robust
##   Model Fit Test Statistic      55.498      56.546
##   Degrees of freedom           51          51
##   P-value (Chi-square)         0.309        0.276
##   Scaling correction factor              0.981
##   for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##   Minimum Function Test Statistic      1962.116      1980.927
##   Degrees of freedom                   66          66
##   P-value                             0.000        0.000
##
## User model versus baseline model:
##
##   Comparative Fit Index (CFI)           0.998        0.997
##   Tucker-Lewis Index (TLI)             0.997        0.996
##
##   Robust Comparative Fit Index (CFI)              NA
##   Robust Tucker-Lewis Index (TLI)                 NA
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)      -7510.369      -7510.369
##   Loglikelihood unrestricted model (H1) -7482.565      -7482.565
##

```

```

##      Number of free parameters              39          39
##      Akaike (AIC)                        15098.739    15098.739
##      Bayesian (BIC)                      15263.108    15263.108
##      Sample-size adjusted Bayesian (BIC)    15139.320    15139.320
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                                0.013        0.015
##      90 Percent Confidence Interval        0.000  0.032        0.000  0.033
##      P-value RMSEA <= 0.05                1.000        1.000
##
##      Robust RMSEA                                NA
##      90 Percent Confidence Interval        0.000        NA
##
## Standardized Root Mean Square Residual:
##
##      SRMR                                0.026        0.026
##
## Parameter Estimates:
##
##      Information                                Observed
##      Observed information based on              Hessian
##      Standard Errors                          Robust.huber.white
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv
## Verbal =~
##   Grammar      1.000
##   Prgrph_Cmprhns  1.065    0.080   13.256    0.000    0.762
##   Vocabulary    0.960    0.070   13.772    0.000    0.687
##   Sentenc_Cmpltn  1.108    0.079   13.940    0.000    0.793
## Math =~
##   Geometry      1.000
##   Algebra        1.002    0.082   12.225    0.000    0.670
##   Numericl_Pzzls  0.994    0.080   12.357    0.000    0.664
##   Series_Cmpltn   1.031    0.079   13.081    0.000    0.689
## Reasoning =~
##   Prctcl_Prblm_S  1.000
##   Symbol_Manpltn  1.030    0.081   12.652    0.000    0.670
##   Analytcl_Ablty  1.097    0.085   12.899    0.000    0.713
##   Formal_Logic    1.143    0.090   12.647    0.000    0.744
## Std.all
##
##      0.734
##      0.718
##      0.689
##      0.731
##
##      0.684
##      0.698
##      0.672
##      0.716
##
##      0.695

```

```

##      0.690
##      0.726
##      0.732
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
## Verbal ~~
##      Math      0.199   0.029   6.782   0.000   0.417
##      Reasoning  0.215   0.029   7.442   0.000   0.461
## Math ~~
##      Reasoning  0.176   0.026   6.759   0.000   0.405
## Std.all
##
##      0.417
##      0.461
##
##      0.405
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .Grammar      -0.008   0.044  -0.175   0.861  -0.008
##      .Prgrph_Cmprhns -0.004   0.048  -0.079   0.937  -0.004
##      .Vocabulary     -0.015   0.045  -0.331   0.741  -0.015
##      .Sentenc_Cmpltn -0.088   0.049  -1.808   0.071  -0.088
##      .Geometry       0.011   0.044   0.243   0.808   0.011
##      .Algebra        -0.029   0.043  -0.666   0.505  -0.029
##      .Numericl_Pzzls -0.015   0.044  -0.341   0.733  -0.015
##      .Series_Cmpltn  -0.006   0.043  -0.147   0.883  -0.006
##      .Prctcl_Prblm_S -0.017   0.042  -0.402   0.688  -0.017
##      .Symbol_Manpltn  0.001   0.043   0.019   0.985   0.001
##      .Analytcl_Ablty -0.005   0.044  -0.106   0.916  -0.005
##      .Formal_Logic   -0.063   0.045  -1.385   0.166  -0.063
##      Verbal          0.000           0.000
##      Math            0.000           0.000
##      Reasoning       0.000           0.000
## Std.all
##      -0.008
##      -0.004
##      -0.015
##      -0.081
##      0.011
##      -0.030
##      -0.015
##      -0.007
##      -0.018
##      0.001
##      -0.005
##      -0.062
##      0.000
##      0.000
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv

```

```

##      .Grammar      0.439    0.036   12.355    0.000    0.439
##      .Prgrph_Cmprhns 0.547    0.044   12.390    0.000    0.547
##      .Vocabulary     0.524    0.041   12.747    0.000    0.524
##      .Sentenc_Cmpltn 0.548    0.044   12.378    0.000    0.548
##      .Geometry       0.509    0.046   11.188    0.000    0.509
##      .Algebra        0.474    0.041   11.462    0.000    0.474
##      .Numericl_Pzzls 0.535    0.045   11.762    0.000    0.535
##      .Series_Cmpltn  0.452    0.040   11.349    0.000    0.452
##      .Prctcl_Prblm_S 0.452    0.039   11.562    0.000    0.452
##      .Symbol_Manpltn 0.494    0.042   11.846    0.000    0.494
##      .Analytcl_Ablty 0.456    0.042   10.850    0.000    0.456
##      .Formal_Logic   0.478    0.041   11.787    0.000    0.478
##      Verbal          0.513    0.060    8.520    0.000    1.000
##      Math            0.447    0.056    7.997    0.000    1.000
##      Reasoning       0.423    0.049    8.639    0.000    1.000
## Std.all
##      0.461
##      0.485
##      0.526
##      0.466
##      0.533
##      0.513
##      0.548
##      0.488
##      0.516
##      0.524
##      0.473
##      0.464
##      1.000
##      1.000
##      1.000
##
## R-Square:
##      Estimate
##      Grammar      0.539
##      Prgrph_Cmprhns 0.515
##      Vocabulary    0.474
##      Sentenc_Cmpltn 0.534
##      Geometry      0.467
##      Algebra       0.487
##      Numericl_Pzzls 0.452
##      Series_Cmpltn 0.512
##      Prctcl_Prblm_S 0.484
##      Symbol_Manpltn 0.476
##      Analytcl_Ablty 0.527
##      Formal_Logic  0.536

CFA_Fit_2 <- cfa(scaling.model.2, data = Mental, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(CFA_Fit_2, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 17 iterations
##
##      Number of observations      500

```

```

##      Number of missing patterns                1
##
##      Estimator                                ML      Robust
##      Model Fit Test Statistic                55.498    56.546
##      Degrees of freedom                      51        51
##      P-value (Chi-square)                    0.309      0.276
##      Scaling correction factor                0.981
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##      Minimum Function Test Statistic        1962.116    1980.927
##      Degrees of freedom                      66        66
##      P-value                                0.000      0.000
##
## User model versus baseline model:
##
##      Comparative Fit Index (CFI)             0.998      0.997
##      Tucker-Lewis Index (TLI)               0.997      0.996
##
##      Robust Comparative Fit Index (CFI)      NA
##      Robust Tucker-Lewis Index (TLI)        NA
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)           -7510.369    -7510.369
##      Loglikelihood unrestricted model (H1)    -7482.565    -7482.565
##
##      Number of free parameters              39        39
##      Akaike (AIC)                          15098.739    15098.739
##      Bayesian (BIC)                        15263.108    15263.108
##      Sample-size adjusted Bayesian (BIC)     15139.320    15139.320
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                                0.013      0.015
##      90 Percent Confidence Interval          0.000  0.032    0.000  0.033
##      P-value RMSEA <= 0.05                 1.000      1.000
##
##      Robust RMSEA                          NA
##      90 Percent Confidence Interval          0.000      NA
##
## Standardized Root Mean Square Residual:
##
##      SRMR                                0.026      0.026
##
## Parameter Estimates:
##
##      Information                          Observed
##      Observed information based on          Hessian
##      Standard Errors                      Robust.huber.white
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv

```



```

## Verbal =~
## Grammar 0.716 0.042 17.040 0.000 0.716
## Prgrph_Cmprhns 0.762 0.044 17.213 0.000 0.762
## Vocabulary 0.687 0.041 16.598 0.000 0.687
## Sentenc_Cmpltn 0.793 0.046 17.357 0.000 0.793
## Math =~
## Geometry 0.669 0.042 15.995 0.000 0.669
## Algebra 0.670 0.045 14.906 0.000 0.670
## Numericl_Pzzls 0.664 0.044 15.098 0.000 0.664
## Series_Cmpltn 0.689 0.041 16.783 0.000 0.689
## Reasoning =~
## Prctcl_Prblm_S 0.650 0.038 17.277 0.000 0.650
## Symbol_Manpltn 0.670 0.045 14.787 0.000 0.670
## Analytcl_Ablty 0.713 0.044 16.360 0.000 0.713
## Formal_Logic 0.744 0.046 16.102 0.000 0.744
## Std.all
##
## 0.734
## 0.718
## 0.689
## 0.731
##
## 0.684
## 0.698
## 0.672
## 0.716
##
## 0.695
## 0.690
## 0.726
## 0.732
##
## Covariances:
## Estimate Std.Err z-value P(>|z|) Std.lv
## Verbal ~~
## Math 0.417 0.046 9.108 0.000 0.417
## Reasoning 0.461 0.048 9.528 0.000 0.461
## Math ~~
## Reasoning 0.405 0.048 8.412 0.000 0.405
## Std.all
##
## 0.417
## 0.461
##
## 0.405
##
## Intercepts:
## Estimate Std.Err z-value P(>|z|) Std.lv
## .Grammar -0.008 0.044 -0.175 0.861 -0.008
## .Prgrph_Cmprhns -0.004 0.048 -0.079 0.937 -0.004
## .Vocabulary -0.015 0.045 -0.331 0.741 -0.015
## .Sentenc_Cmpltn -0.088 0.049 -1.808 0.071 -0.088
## .Geometry 0.011 0.044 0.243 0.808 0.011
## .Algebra -0.029 0.043 -0.666 0.505 -0.029

```

```

##      .Numericl_Pzzls  -0.015    0.044   -0.341    0.733   -0.015
##      .Series_Compltn  -0.006    0.043   -0.147    0.883   -0.006
##      .Prctcl_Prblm_S  -0.017    0.042   -0.402    0.688   -0.017
##      .Symbol_Manpltn    0.001    0.043    0.019    0.985    0.001
##      .Analytcl_Ablty  -0.005    0.044   -0.106    0.916   -0.005
##      .Formal_Logic    -0.063    0.045   -1.385    0.166   -0.063
##      Verbal            0.000                                0.000
##      Math              0.000                                0.000
##      Reasoning         0.000                                0.000
## Std.all
##      -0.008
##      -0.004
##      -0.015
##      -0.081
##      0.011
##      -0.030
##      -0.015
##      -0.007
##      -0.018
##      0.001
##      -0.005
##      -0.062
##      0.000
##      0.000
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      Verbal      1.000
##      Math        1.000
##      Reasoning    1.000
##      .Grammar     0.439    0.036   12.355    0.000    0.439
##      .Prgrph_Cmprhns 0.547    0.044   12.390    0.000    0.547
##      .Vocabulary    0.524    0.041   12.747    0.000    0.524
##      .Sentenc_Cmpltn 0.548    0.044   12.378    0.000    0.548
##      .Geometry     0.509    0.046   11.188    0.000    0.509
##      .Algebra       0.474    0.041   11.462    0.000    0.474
##      .Numericl_Pzzls 0.535    0.045   11.762    0.000    0.535
##      .Series_Compltn 0.452    0.040   11.349    0.000    0.452
##      .Prctcl_Prblm_S 0.452    0.039   11.562    0.000    0.452
##      .Symbol_Manpltn 0.494    0.042   11.846    0.000    0.494
##      .Analytcl_Ablty 0.456    0.042   10.850    0.000    0.456
##      .Formal_Logic   0.478    0.041   11.787    0.000    0.478
## Std.all
##      1.000
##      1.000
##      1.000
##      0.461
##      0.485
##      0.526
##      0.466
##      0.533
##      0.513
##      0.548

```

```
##      0.488
##      0.516
##      0.524
##      0.473
##      0.464
##
## R-Square:
##              Estimate
## Grammar      0.539
## Prgrph_Cmprhns 0.515
## Vocabulary    0.474
## Sentenc_Cmpltn 0.534
## Geometry      0.467
## Algebra       0.487
## Numericl_Pzzls 0.452
## Series_Cmpltn 0.512
## Prctcl_Prblm_S 0.484
## Symbol_Manpltn 0.476
## Analytcl_Ablty 0.527
## Formal_Logic  0.536
```

4 CFA With Covariance versus Correlation Matrices

Confirmatory factor analysis models are usually based on the decomposition of covariance matrices, not correlation matrices. The solutions hold, strictly speaking, for the analysis of covariance matrices. To the extent that the solution depends on the scale of the variables, analyses based on covariance matrices and correlation matrices can differ. The differences are a function of the differences in the variances across the variables.

```
NC.model.1 <- "
# Latent variable definitions.
NC =~ nc1 + nc2 + nc3 + nc4 + nc5 + nc6 + nc7 + nc8 + nc9 + nc10 +
nc11 + nc12 + nc13 + nc14 + nc15 + nc16 + nc17 + nc18
"
```

```
CFA_Fit_3 <- cfa(NC.model.1, data = NC, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(CFA_Fit_3, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 29 iterations
##
##      Number of observations              195
##      Number of missing patterns          1
##
##      Estimator              ML          Robust
##      Model Fit Test Statistic    267.315    228.948
##      Degrees of freedom          135         135
##      P-value (Chi-square)         0.000         0.000
##      Scaling correction factor              1.168
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
```

```

##
## Minimum Function Test Statistic          1061.451      874.878
## Degrees of freedom                      153          153
## P-value                                0.000          0.000
##
## User model versus baseline model:
##
## Comparative Fit Index (CFI)              0.854          0.870
## Tucker-Lewis Index (TLI)                0.835          0.853
##
## Robust Comparative Fit Index (CFI)              NA
## Robust Tucker-Lewis Index (TLI)              NA
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0)             -4984.773      -4984.773
## Loglikelihood unrestricted model (H1)       -4850.426      -4850.426
##
## Number of free parameters                 54           54
## Akaike (AIC)                             10077.545      10077.545
## Bayesian (BIC)                           10254.287      10254.287
## Sample-size adjusted Bayesian (BIC)        10083.223      10083.223
##
## Root Mean Square Error of Approximation:
##
## RMSEA                                    0.071          0.060
## 90 Percent Confidence Interval            0.059  0.084      0.047  0.072
## P-value RMSEA <= 0.05                    0.004          0.095
##
## Robust RMSEA                              NA
## 90 Percent Confidence Interval            NA          NA
##
## Standardized Root Mean Square Residual:
##
## SRMR                                    0.060          0.060
##
## Parameter Estimates:
##
## Information                               Observed
## Observed information based on              Hessian
## Standard Errors                          Robust.huber.white
##
## Latent Variables:
##
## Estimate Std.Err z-value P(>|z|) Std.lv
## NC =~
## nc1      1.000
## nc2      1.072  0.155  6.922  0.000  0.741
## nc3     -0.884  0.152 -5.818  0.000 -0.611
## nc4     -1.038  0.189 -5.496  0.000 -0.717
## nc5     -1.018  0.184 -5.543  0.000 -0.704
## nc6      0.682  0.155  4.395  0.000  0.472
## nc7     -0.957  0.164 -5.826  0.000 -0.662
## nc8     -0.787  0.143 -5.519  0.000 -0.544
## nc9     -0.951  0.182 -5.210  0.000 -0.657

```

##	nc10	0.801	0.134	5.980	0.000	0.554
##	nc11	0.906	0.149	6.095	0.000	0.626
##	nc12	-0.910	0.162	-5.602	0.000	-0.629
##	nc13	0.865	0.122	7.095	0.000	0.598
##	nc14	0.800	0.129	6.218	0.000	0.553
##	nc15	0.567	0.148	3.819	0.000	0.392
##	nc16	-0.604	0.153	-3.956	0.000	-0.418
##	nc17	-0.836	0.170	-4.912	0.000	-0.578
##	nc18	0.319	0.140	2.277	0.023	0.221
##	Std.all					
##						
##	0.560					
##	0.725					
##	-0.577					
##	-0.627					
##	-0.676					
##	0.394					
##	-0.504					
##	-0.448					
##	-0.521					
##	0.586					
##	0.652					
##	-0.594					
##	0.522					
##	0.506					
##	0.367					
##	-0.352					
##	-0.513					
##	0.183					
##						
##	Intercepts:					
##		Estimate	Std.Err	z-value	P(> z)	Std.lv
##	.nc1	3.241	0.089	36.537	0.000	3.241
##	.nc2	3.913	0.073	53.347	0.000	3.913
##	.nc3	1.821	0.076	23.943	0.000	1.821
##	.nc4	1.974	0.082	24.054	0.000	1.974
##	.nc5	1.831	0.075	24.490	0.000	1.831
##	.nc6	3.287	0.086	38.214	0.000	3.287
##	.nc7	2.487	0.094	26.363	0.000	2.487
##	.nc8	2.518	0.087	28.924	0.000	2.518
##	.nc9	2.267	0.091	25.034	0.000	2.267
##	.nc10	4.056	0.068	59.746	0.000	4.056
##	.nc11	4.205	0.069	60.997	0.000	4.205
##	.nc12	1.754	0.076	23.094	0.000	1.754
##	.nc13	3.287	0.082	39.966	0.000	3.287
##	.nc14	3.749	0.078	47.799	0.000	3.749
##	.nc15	3.677	0.077	47.929	0.000	3.677
##	.nc16	2.292	0.085	26.918	0.000	2.292
##	.nc17	1.903	0.081	23.538	0.000	1.903
##	.nc18	3.328	0.086	38.480	0.000	3.328
##	NC	0.000				0.000
##	Std.all					
##	2.623					
##	3.830					

```

##      1.719
##      1.727
##      1.758
##      2.744
##      1.893
##      2.077
##      1.797
##      4.290
##      4.379
##      1.658
##      2.869
##      3.432
##      3.441
##      1.933
##      1.690
##      2.763
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      1.049   0.131   7.997   0.000   1.049
##      .nc2      0.495   0.077   6.450   0.000   0.495
##      .nc3      0.748   0.116   6.440   0.000   0.748
##      .nc4      0.793   0.129   6.153   0.000   0.793
##      .nc5      0.588   0.085   6.945   0.000   0.588
##      .nc6      1.213   0.127   9.541   0.000   1.213
##      .nc7      1.289   0.129   9.990   0.000   1.289
##      .nc8      1.175   0.111  10.615   0.000   1.175
##      .nc9      1.159   0.136   8.526   0.000   1.159
##      .nc10     0.588   0.093   6.305   0.000   0.588
##      .nc11     0.530   0.084   6.311   0.000   0.530
##      .nc12     0.724   0.125   5.766   0.000   0.724
##      .nc13     0.955   0.092  10.336   0.000   0.955
##      .nc14     0.887   0.103   8.645   0.000   0.887
##      .nc15     0.988   0.132   7.515   0.000   0.988
##      .nc16     1.232   0.145   8.477   0.000   1.232
##      .nc17     0.934   0.113   8.255   0.000   0.934
##      .nc18     1.402   0.112  12.569   0.000   1.402
##      NC        0.478   0.125   3.823   0.000   1.000
## Std.all
##      0.687
##      0.474
##      0.667
##      0.606
##      0.543
##      0.845
##      0.746
##      0.799
##      0.728
##      0.657
##      0.574
##      0.647
##      0.728
##      0.744

```

```

##      0.866
##      0.876
##      0.737
##      0.966
##      1.000
##
## R-Square:
##              Estimate
##      nc1          0.313
##      nc2          0.526
##      nc3          0.333
##      nc4          0.394
##      nc5          0.457
##      nc6          0.155
##      nc7          0.254
##      nc8          0.201
##      nc9          0.272
##      nc10         0.343
##      nc11         0.426
##      nc12         0.353
##      nc13         0.272
##      nc14         0.256
##      nc15         0.134
##      nc16         0.124
##      nc17         0.263
##      nc18         0.034

CFA_Fit_4 <- cfa(NC.model.1, data = as.data.frame(scale(NC)), missing = "ML",
  estimator = "MLR", likelihood = "wishart", representation = "LISREL")
summary(CFA_Fit_4, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 26 iterations
##
##      Number of observations          195
##      Number of missing patterns      1
##
##      Estimator          ML          Robust
##      Model Fit Test Statistic    267.315    228.948
##      Degrees of freedom          135         135
##      P-value (Chi-square)        0.000         0.000
##      Scaling correction factor          1.168
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##      Minimum Function Test Statistic    1061.451    874.878
##      Degrees of freedom          153         153
##      P-value          0.000         0.000
##
## User model versus baseline model:
##
##      Comparative Fit Index (CFI)          0.854         0.870
##      Tucker-Lewis Index (TLI)          0.835         0.853
##

```

```

## Robust Comparative Fit Index (CFI) NA
## Robust Tucker-Lewis Index (TLI) NA
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -4572.337 -4572.337
## Loglikelihood unrestricted model (H1) -4437.990 -4437.990
##
## Number of free parameters 54 54
## Akaike (AIC) 9252.673 9252.673
## Bayesian (BIC) 9429.415 9429.415
## Sample-size adjusted Bayesian (BIC) 9258.351 9258.351
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.071 0.060
## 90 Percent Confidence Interval 0.059 0.084 0.047 0.072
## P-value RMSEA <= 0.05 0.004 0.095
##
## Robust RMSEA NA
## 90 Percent Confidence Interval NA NA
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.060 0.060
##
## Parameter Estimates:
##
## Information Observed
## Observed information based on Hessian
## Standard Errors Robust.huber.white
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv
## NC =~
## nc1 1.000 0.558
## nc2 1.296 0.187 6.922 0.000 0.724
## nc3 -1.031 0.177 -5.818 0.000 -0.575
## nc4 -1.121 0.204 -5.496 0.000 -0.626
## nc5 -1.208 0.218 -5.543 0.000 -0.674
## nc6 0.704 0.160 4.395 0.000 0.393
## nc7 -0.900 0.155 -5.826 0.000 -0.502
## nc8 -0.801 0.145 -5.519 0.000 -0.447
## nc9 -0.931 0.179 -5.210 0.000 -0.520
## nc10 1.046 0.175 5.980 0.000 0.584
## nc11 1.166 0.191 6.095 0.000 0.651
## nc12 -1.062 0.190 -5.602 0.000 -0.593
## nc13 0.933 0.131 7.095 0.000 0.521
## nc14 0.905 0.146 6.218 0.000 0.505
## nc15 0.655 0.172 3.819 0.000 0.366
## nc16 -0.630 0.159 -3.956 0.000 -0.351
## nc17 -0.917 0.187 -4.912 0.000 -0.512
## nc18 0.328 0.144 2.277 0.023 0.183

```



```

## Std.all
##
## 0.560
## 0.725
## -0.577
## -0.627
## -0.676
## 0.394
## -0.504
## -0.448
## -0.521
## 0.586
## 0.652
## -0.594
## 0.522
## 0.506
## 0.367
## -0.352
## -0.513
## 0.183
##
## Intercepts:
## Estimate Std.Err z-value P(>|z|) Std.lv
## .nc1 -0.000 0.072 -0.000 1.000 -0.000
## .nc2 0.000 0.072 0.000 1.000 0.000
## .nc3 0.000 0.072 0.000 1.000 0.000
## .nc4 -0.000 0.072 -0.000 1.000 -0.000
## .nc5 0.000 0.072 0.000 1.000 0.000
## .nc6 0.000 0.072 0.000 1.000 0.000
## .nc7 -0.000 0.072 -0.000 1.000 -0.000
## .nc8 0.000 0.072 0.000 1.000 0.000
## .nc9 0.000 0.072 0.000 1.000 0.000
## .nc10 0.000 0.072 0.000 1.000 0.000
## .nc11 -0.000 0.072 -0.000 1.000 -0.000
## .nc12 0.000 0.072 0.000 1.000 0.000
## .nc13 0.000 0.072 0.000 1.000 0.000
## .nc14 0.000 0.072 0.000 1.000 0.000
## .nc15 0.000 0.072 0.000 1.000 0.000
## .nc16 0.000 0.072 0.000 1.000 0.000
## .nc17 0.000 0.072 0.000 1.000 0.000
## .nc18 -0.000 0.072 -0.000 1.000 -0.000
## NC 0.000 0.000
## Std.all
## -0.000
## 0.000
## 0.000
## -0.000
## 0.000
## 0.000
## -0.000
## 0.000
## 0.000
## 0.000
## -0.000

```

```

##      0.000
##      0.000
##      0.000
##      0.000
##      0.000
##      0.000
##     -0.000
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      0.683   0.085   7.997   0.000   0.683
##      .nc2      0.471   0.073   6.450   0.000   0.471
##      .nc3      0.664   0.103   6.440   0.000   0.664
##      .nc4      0.603   0.098   6.153   0.000   0.603
##      .nc5      0.540   0.078   6.945   0.000   0.540
##      .nc6      0.841   0.088   9.541   0.000   0.841
##      .nc7      0.742   0.074   9.990   0.000   0.742
##      .nc8      0.795   0.075  10.615   0.000   0.795
##      .nc9      0.725   0.085   8.526   0.000   0.725
##      .nc10     0.654   0.104   6.305   0.000   0.654
##      .nc11     0.571   0.091   6.311   0.000   0.571
##      .nc12     0.643   0.112   5.766   0.000   0.643
##      .nc13     0.724   0.070  10.336   0.000   0.724
##      .nc14     0.740   0.086   8.645   0.000   0.740
##      .nc15     0.861   0.115   7.515   0.000   0.861
##      .nc16     0.871   0.103   8.477   0.000   0.871
##      .nc17     0.733   0.089   8.255   0.000   0.733
##      .nc18     0.961   0.076  12.569   0.000   0.961
##      NC        0.311   0.081   3.823   0.000   1.000
## Std.all
##      0.687
##      0.474
##      0.667
##      0.606
##      0.543
##      0.845
##      0.746
##      0.799
##      0.728
##      0.657
##      0.574
##      0.647
##      0.728
##      0.744
##      0.866
##      0.876
##      0.737
##      0.966
##      1.000
##
## R-Square:
##      Estimate
##      nc1      0.313

```

```
##      nc2      0.526
##      nc3      0.333
##      nc4      0.394
##      nc5      0.457
##      nc6      0.155
##      nc7      0.254
##      nc8      0.201
##      nc9      0.272
##      nc10     0.343
##      nc11     0.426
##      nc12     0.353
##      nc13     0.272
##      nc14     0.256
##      nc15     0.134
##      nc16     0.124
##      nc17     0.263
##      nc18     0.034
```

`cov(NC)`

```
##      nc1      nc2      nc3      nc4      nc5      nc6      nc7
## nc1  1.5344  0.5778 -0.3122 -0.3855 -0.3920  0.3634 -0.46854
## nc2  0.5778  1.0491 -0.5054 -0.4146 -0.5303  0.3138 -0.46246
## nc3 -0.3122 -0.5054  1.1274  0.4541  0.4952 -0.2420  0.44872
## nc4 -0.3855 -0.4146  0.4541  1.3138  0.6812 -0.2297  0.52802
## nc5 -0.3920 -0.5303  0.4952  0.6812  1.0898 -0.2810  0.53132
## nc6  0.3634  0.3138 -0.2420 -0.2297 -0.2810  1.4429 -0.38805
## nc7 -0.4685 -0.4625  0.4487  0.5280  0.5313 -0.3881  1.73566
## nc8 -0.3574 -0.4443  0.2893  0.4257  0.3922 -0.2217  0.54018
## nc9 -0.5027 -0.4148  0.5172  0.5945  0.4165 -0.2522  0.36426
## nc10 0.3368  0.3709 -0.4073 -0.3748 -0.3512  0.3239 -0.33175
## nc11 0.4297  0.5077 -0.3290 -0.4432 -0.3878  0.3377 -0.33241
## nc12 -0.4764 -0.4649  0.3113  0.4627  0.5715 -0.3207  0.48652
## nc13 0.5593  0.4582 -0.2936 -0.3844 -0.3790  0.2882 -0.27465
## nc14 0.4475  0.4367 -0.3134 -0.3209 -0.2902  0.3663 -0.37695
## nc15 0.3669  0.3634 -0.2799 -0.1939 -0.1684  0.2170 -0.09437
## nc16 -0.4162 -0.2270  0.2022  0.4354  0.3178 -0.2132  0.32593
## nc17 -0.3785 -0.4828  0.3381  0.5233  0.3855 -0.2966  0.37761
## nc18 0.1731  0.2349 -0.1831 -0.1204 -0.1194  0.2455 -0.23804
##      nc8      nc9      nc10      nc11      nc12      nc13      nc14
## nc1 -0.35744 -0.50275  0.33685  0.4297 -0.4764  0.55929  0.4475
## nc2 -0.44430 -0.41478  0.37092  0.5077 -0.4649  0.45815  0.4367
## nc3  0.28932  0.51718 -0.40735 -0.3290  0.3113 -0.29355 -0.3134
## nc4  0.42572  0.59450 -0.37483 -0.4432  0.4627 -0.38435 -0.3209
## nc5  0.39223  0.41649 -0.35123 -0.3878  0.5715 -0.37898 -0.2902
## nc6 -0.22168 -0.25223  0.32392  0.3377 -0.3207  0.28824  0.3663
## nc7  0.54018  0.36426 -0.33175 -0.3324  0.4865 -0.27465 -0.3769
## nc8  1.47777  0.47457 -0.33865 -0.1635  0.1333 -0.32992 -0.4156
## nc9  0.47457  1.59863 -0.44811 -0.4055  0.3340 -0.35533 -0.2058
## nc10 -0.33865 -0.44811  0.89886  0.4729 -0.2799  0.34454  0.3648
## nc11 -0.16349 -0.40550  0.47291  0.9268 -0.4029  0.45625  0.3302
## nc12 0.13331  0.33402 -0.27986 -0.4029  1.1247 -0.41348 -0.4230
## nc13 -0.32992 -0.35533  0.34454  0.4563 -0.4135  1.31916  0.5674
## nc14 -0.41557 -0.20584  0.36476  0.3302 -0.4230  0.56738  1.1994
```

```
## nc15 -0.23386 -0.23814 0.24512 0.2470 -0.2294 0.36130 0.2689
## nc16 0.20349 0.30825 -0.15575 -0.2716 0.2579 -0.06891 -0.1736
## nc17 0.49403 0.52612 -0.24190 -0.3871 0.3316 -0.13682 -0.1019
## nc18 -0.03685 -0.01581 -0.02376 0.1231 -0.1404 0.02897 0.2427
##          nc15      nc16      nc17      nc18
## nc1  0.36693 -0.41618 -0.3785 0.17306
## nc2  0.36344 -0.22696 -0.4828 0.23495
## nc3 -0.27994 0.20222 0.3381 -0.18306
## nc4 -0.19389 0.43537 0.5233 -0.12041
## nc5 -0.16836 0.31776 0.3855 -0.11943
## nc6 0.21697 -0.21324 -0.2966 0.24547
## nc7 -0.09437 0.32593 0.3776 -0.23804
## nc8 -0.23386 0.20349 0.4940 -0.03685
## nc9 -0.23814 0.30825 0.5261 -0.01581
## nc10 0.24512 -0.15575 -0.2419 -0.02376
## nc11 0.24703 -0.27161 -0.3871 0.12305
## nc12 -0.22942 0.25789 0.3316 -0.14044
## nc13 0.36130 -0.06891 -0.1368 0.02897
## nc14 0.26891 -0.17359 -0.1019 0.24269
## nc15 1.14766 -0.08549 -0.2017 0.12720
## nc16 -0.08549 1.41412 0.4410 -0.28200
## nc17 -0.20174 0.44100 1.2740 -0.12765
## nc18 0.12720 -0.28200 -0.1277 1.45874
```

```
cov(scale(NC))
```

```
##          nc1      nc2      nc3      nc4      nc5      nc6      nc7
## nc1  1.0000 0.4554 -0.2374 -0.27155 -0.30314 0.2442 -0.28711
## nc2  0.4554 1.0000 -0.4647 -0.35317 -0.49596 0.2551 -0.34272
## nc3 -0.2374 -0.4647 1.0000 0.37315 0.44680 -0.1897 0.32077
## nc4 -0.2715 -0.3532 0.3732 1.00000 0.56931 -0.1668 0.34967
## nc5 -0.3031 -0.4960 0.4468 0.56931 1.00000 -0.2241 0.38633
## nc6 0.2442 0.2551 -0.1897 -0.16684 -0.22413 1.0000 -0.24521
## nc7 -0.2871 -0.3427 0.3208 0.34967 0.38633 -0.2452 1.00000
## nc8 -0.2374 -0.3568 0.2241 0.30554 0.30908 -0.1518 0.33729
## nc9 -0.3210 -0.3203 0.3852 0.41022 0.31555 -0.1661 0.21868
## nc10 0.2868 0.3820 -0.4046 -0.34493 -0.35488 0.2844 -0.26560
## nc11 0.3603 0.5149 -0.3218 -0.40162 -0.38587 0.2920 -0.26209
## nc12 -0.3627 -0.4280 0.2764 0.38068 0.51625 -0.2518 0.34822
## nc13 0.3931 0.3895 -0.2407 -0.29196 -0.31609 0.2089 -0.18151
## nc14 0.3298 0.3894 -0.2695 -0.25564 -0.25380 0.2785 -0.26126
## nc15 0.2765 0.3312 -0.2461 -0.15791 -0.15054 0.1686 -0.06686
## nc16 -0.2825 -0.1863 0.1602 0.31941 0.25597 -0.1493 0.20804
## nc17 -0.2707 -0.4176 0.2821 0.40446 0.32716 -0.2188 0.25394
## nc18 0.1157 0.1899 -0.1427 -0.08698 -0.09472 0.1692 -0.14960
##          nc8      nc9      nc10      nc11      nc12      nc13      nc14
## nc1 -0.2374 -0.32100 0.28683 0.3603 -0.3627 0.39312 0.32985
## nc2 -0.3568 -0.32029 0.38198 0.5149 -0.4280 0.38946 0.38935
## nc3 0.2241 0.38524 -0.40465 -0.3218 0.2764 -0.24071 -0.26949
## nc4 0.3055 0.41022 -0.34493 -0.4016 0.3807 -0.29196 -0.25564
## nc5 0.3091 0.31555 -0.35488 -0.3859 0.5163 -0.31609 -0.25380
## nc6 -0.1518 -0.16608 0.28443 0.2920 -0.2518 0.20892 0.27848
## nc7 0.3373 0.21868 -0.26560 -0.2621 0.3482 -0.18151 -0.26126
## nc8 1.0000 0.30876 -0.29383 -0.1397 0.1034 -0.23630 -0.31214
```

```

## nc9    0.3088  1.00000 -0.37382 -0.3331  0.2491 -0.24468 -0.14865
## nc10   -0.2938 -0.37382  1.00000  0.5181 -0.2783  0.31641  0.35130
## nc11   -0.1397 -0.33314  0.51813  1.0000 -0.3946  0.41264  0.31315
## nc12    0.1034  0.24911 -0.27834 -0.3946  1.0000 -0.33947 -0.36420
## nc13   -0.2363 -0.24468  0.31641  0.4126 -0.3395  1.00000  0.45107
## nc14   -0.3121 -0.14865  0.35130  0.3132 -0.3642  0.45107  1.00000
## nc15   -0.1796 -0.17582  0.24134  0.2395 -0.2019  0.29364  0.22920
## nc16    0.1408  0.20501 -0.13815 -0.2373  0.2045 -0.05046 -0.13329
## nc17    0.3601  0.36866 -0.22605 -0.3563  0.2770 -0.10554 -0.08246
## nc18   -0.0251 -0.01035 -0.02075  0.1058 -0.1096  0.02089  0.18348
##          nc15      nc16      nc17      nc18
## nc1     0.27651 -0.28253 -0.27069  0.11568
## nc2     0.33123 -0.18634 -0.41760  0.18992
## nc3    -0.24610  0.16016  0.28211 -0.14274
## nc4    -0.15791  0.31941  0.40446 -0.08698
## nc5    -0.15054  0.25597  0.32716 -0.09472
## nc6     0.16861 -0.14929 -0.21878  0.16920
## nc7    -0.06686  0.20804  0.25394 -0.14960
## nc8    -0.17958  0.14077  0.36005 -0.02510
## nc9    -0.17582  0.20501  0.36866 -0.01035
## nc10     0.24134 -0.13815 -0.22605 -0.02075
## nc11     0.23952 -0.23725 -0.35628  0.10583
## nc12    -0.20194  0.20449  0.27700 -0.10965
## nc13     0.29364 -0.05046 -0.10554  0.02089
## nc14     0.22920 -0.13329 -0.08246  0.18348
## nc15     1.00000 -0.06710 -0.16685  0.09831
## nc16    -0.06710  1.00000  0.32856 -0.19634
## nc17    -0.16685  0.32856  1.00000 -0.09364
## nc18     0.09831 -0.19634 -0.09364  1.00000

(as.matrix(diag(cov(NC))))

##          [,1]
## nc1     1.5344
## nc2     1.0491
## nc3     1.1274
## nc4     1.3138
## nc5     1.0898
## nc6     1.4429
## nc7     1.7357
## nc8     1.4778
## nc9     1.5986
## nc10    0.8989
## nc11    0.9268
## nc12    1.1247
## nc13    1.3192
## nc14    1.1994
## nc15    1.1477
## nc16    1.4141
## nc17    1.2740
## nc18    1.4587

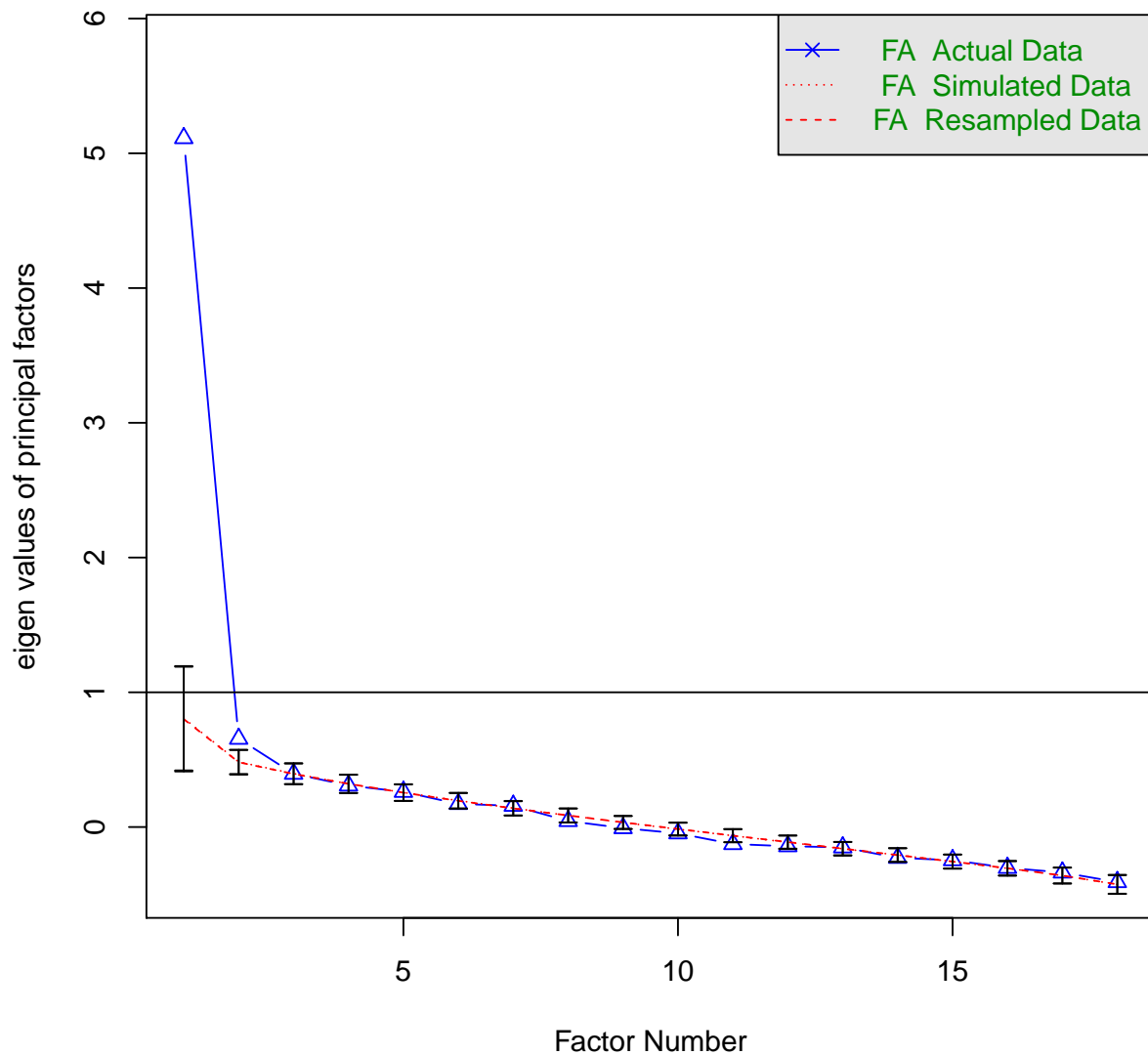
```

5 Model Modification

The Need for Cognition Scale is supposedly a measure of a single underlying latent variable. Is that true? We'll first see if one factor seems sensible using methods from exploratory factor analysis. Then we'll apply CFA and explore model modification.

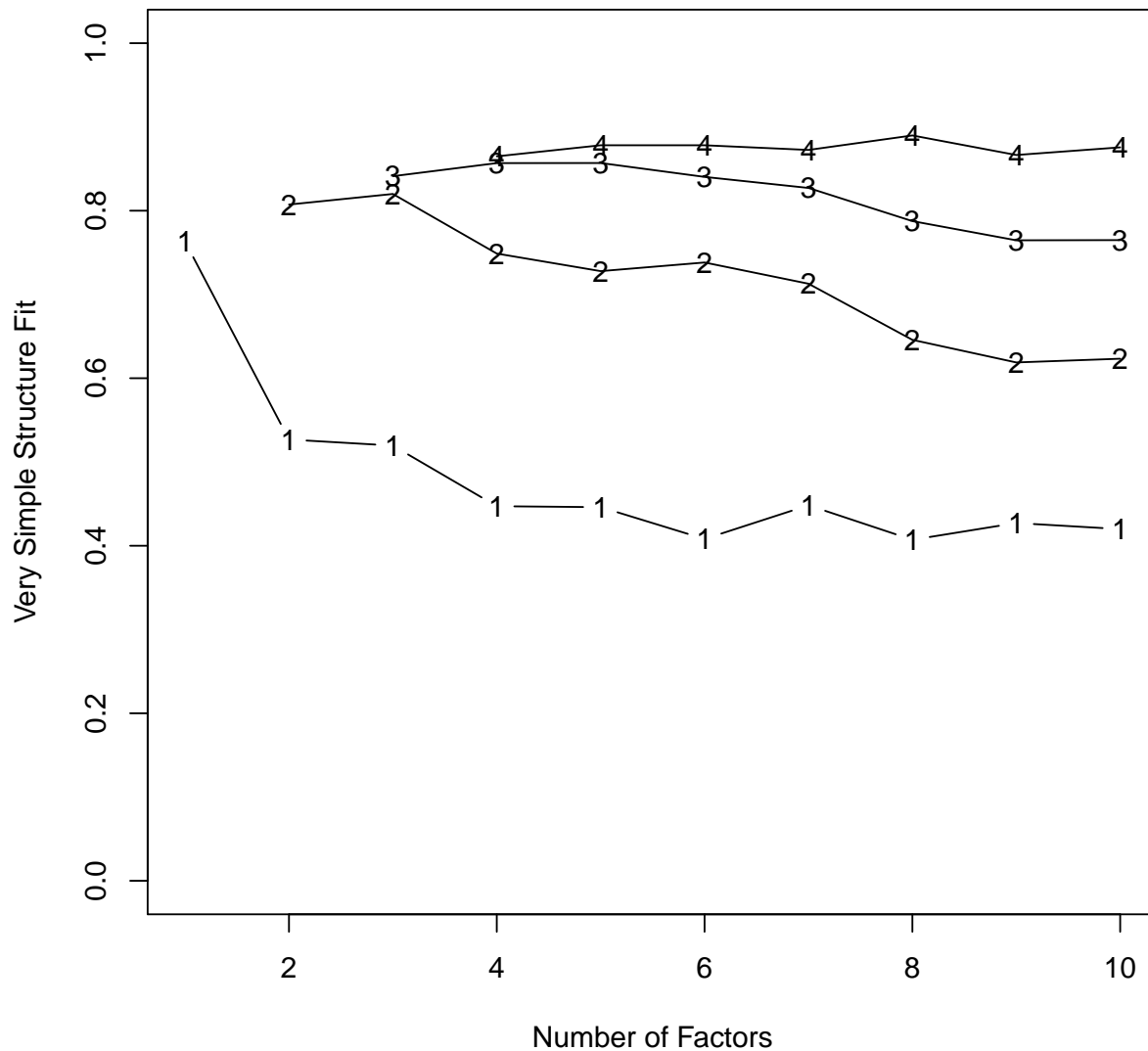
```
fa.parallel(NC, fa = "fa", fm = "ml", error.bars = TRUE, n.iter = 1000)
```

Parallel Analysis Scree Plots



```
## Parallel analysis suggests that the number of factors = 2 and the number of components = NA  
M <- vss(NC, n = 10, fm = "pc")
```

Very Simple Structure



```
plot_data <- M$map
plot_data <- as.data.frame(plot_data)
names(plot_data) <- c("map")
plot_data$component <- seq(1, 10, 1)

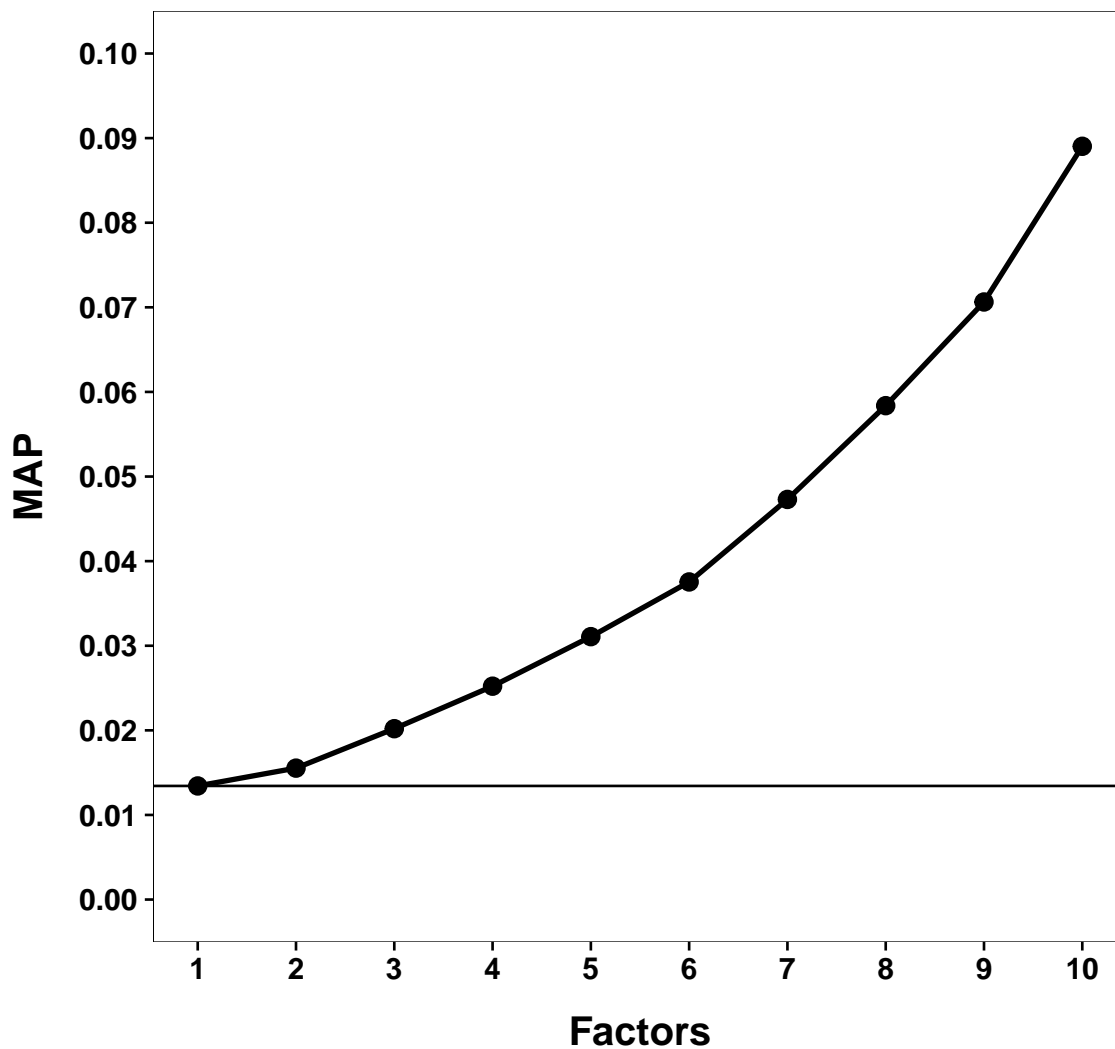
ggplot(plot_data, aes(x = component, y = map)) + geom_line(size = 1) +
  geom_point(size = 3) + coord_cartesian(xlim = c(1, 10), ylim = c(0,
0.1)) + scale_x_continuous(breaks = c(seq(1, 10, 1))) + scale_y_continuous(breaks = seq(0,
0.1, 0.01)) + xlab("Factors") + ylab("MAP") + theme(text = element_text(size = 14,
family = "sans", color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
size = 12, face = "bold", angle = 0), axis.title.x = element_text(margin = margin(15,
0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
15, 0, 0), size = 16), axis.line.x = element_blank(), axis.line.y = element_blank(),
```

```

plot.title = element_text(size = 16, face = "bold", margin = margin(0,
  0, 20, 0), hjust = 0.5), panel.background = element_rect(fill = "white",
  linetype = 1, color = "black"), panel.grid.major = element_blank(),
panel.grid.minor = element_blank(), plot.background = element_rect(fill = "white"),
plot.margin = unit(c(1, 1, 1, 1), "cm"), legend.position = "bottom",
legend.title = element_blank()) + geom_hline(yintercept = min(plot_data$map,
size = 1, linetype = 2)) + ggtitle("MAP Index as a Function of Factors")

```

MAP Index as a Function of Factors



```

fit_NC <- fa(NC, nfactors = 1, rotate = "none", fm = "ml", n.obs = 195,
  n.iter = 1000)
fit_NC

## Factor Analysis with confidence intervals using method = fa(r = NC, nfactors = 1, n.obs = 195, n.iter
##      fm = "ml")

```



```

## Factor Analysis using method = ml
## Call: fa(r = NC, nfactors = 1, n.obs = 195, n.iter = 1000, rotate = "none",
##       fm = "ml")
## Standardized loadings (pattern matrix) based upon correlation matrix
##      ML1      h2      u2 com
## nc1  -0.56 0.313 0.69   1
## nc2  -0.73 0.526 0.47   1
## nc3   0.58 0.333 0.67   1
## nc4   0.63 0.394 0.61   1
## nc5   0.68 0.457 0.54   1
## nc6  -0.39 0.155 0.84   1
## nc7   0.50 0.254 0.75   1
## nc8   0.45 0.201 0.80   1
## nc9   0.52 0.272 0.73   1
## nc10 -0.59 0.343 0.66   1
## nc11 -0.65 0.426 0.57   1
## nc12  0.59 0.353 0.65   1
## nc13 -0.52 0.272 0.73   1
## nc14 -0.51 0.256 0.74   1
## nc15 -0.37 0.134 0.87   1
## nc16  0.35 0.124 0.88   1
## nc17  0.51 0.263 0.74   1
## nc18 -0.18 0.034 0.97   1
##
##              ML1
## SS loadings    5.11
## Proportion Var 0.28
##
## Mean item complexity = 1
## Test of the hypothesis that 1 factor is sufficient.
##
## The degrees of freedom for the null model are 153 and the objective function was 5.47 with Chi Squ
## The degrees of freedom for the model are 135 and the objective function was 1.38
##
## The root mean square of the residuals (RMSR) is 0.07
## The df corrected root mean square of the residuals is 0.07
##
## The harmonic number of observations is 195 with the empirical chi square 264 with prob < 2.2e-10
## The total number of observations was 195 with Likelihood Chi Square = 257 with prob < 1.3e-09
##
## Tucker Lewis Index of factoring reliability = 0.841
## RMSEA index = 0.071 and the 90 % confidence intervals are 0.055 0.081
## BIC = -454.9
## Fit based upon off diagonal values = 0.95
## Measures of factor score adequacy
##
##              ML1
## Correlation of (regression) scores with factors 0.94
## Multiple R square of scores with factors 0.89
## Minimum correlation of possible factor scores 0.78
##
## Coefficients and bootstrapped confidence intervals
##      low ML1 upper
## nc1 -1.22 -0.56 0.58
## nc2 -1.58 -0.73 0.72

```

```
## nc3 -0.53 0.58 1.25
## nc4 -0.56 0.63 1.36
## nc5 -0.62 0.68 1.47
## nc6 -0.88 -0.39 0.42
## nc7 -0.47 0.50 1.09
## nc8 -0.43 0.45 0.97
## nc9 -0.48 0.52 1.13
## nc10 -1.28 -0.59 0.58
## nc11 -1.42 -0.65 0.65
## nc12 -0.58 0.59 1.31
## nc13 -1.14 -0.52 0.55
## nc14 -1.12 -0.51 0.55
## nc15 -0.82 -0.37 0.41
## nc16 -0.34 0.35 0.78
## nc17 -0.45 0.51 1.11
## nc18 -0.44 -0.18 0.24
```

```
NC.model.1 <- "
# Latent variable definitions.
NC =~ nc1 + nc2 + nc3 + nc4 + nc5 + nc6 + nc7 + nc8 + nc9 + nc10 +
nc11 + nc12 + nc13 + nc14 + nc15 + nc16 + nc17 + nc18
"
```

```
NC_Fit_1 <- cfa(NC.model.1, data = NC, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(NC_Fit_1, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)
```

```
## lavaan (0.6-1) converged normally after 29 iterations
##
## Number of observations              195
## Number of missing patterns          1
##
## Estimator                          ML      Robust
## Model Fit Test Statistic            267.315  228.948
## Degrees of freedom                  135      135
## P-value (Chi-square)                 0.000    0.000
## Scaling correction factor              1.168
##   for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
## Minimum Function Test Statistic      1061.451  874.878
## Degrees of freedom                   153      153
## P-value                             0.000    0.000
##
## User model versus baseline model:
##
## Comparative Fit Index (CFI)           0.854    0.870
## Tucker-Lewis Index (TLI)             0.835    0.853
##
## Robust Comparative Fit Index (CFI)      NA
## Robust Tucker-Lewis Index (TLI)        NA
##
```

```

## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)          -4984.773   -4984.773
##   Loglikelihood unrestricted model (H1)   -4850.426   -4850.426
##
##   Number of free parameters              54          54
##   Akaike (AIC)                          10077.545   10077.545
##   Bayesian (BIC)                        10254.287   10254.287
##   Sample-size adjusted Bayesian (BIC)    10083.223   10083.223
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                                0.071        0.060
##   90 Percent Confidence Interval         0.059  0.084        0.047  0.072
##   P-value RMSEA <= 0.05                0.004        0.095
##
##   Robust RMSEA                          NA
##   90 Percent Confidence Interval        NA      NA
##
## Standardized Root Mean Square Residual:
##
##   SRMR                                0.060        0.060
##
## Parameter Estimates:
##
##   Information                          Observed
##   Observed information based on        Hessian
##   Standard Errors                     Robust.huber.white
##
## Latent Variables:
##
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##   NC =~
##   nc1           1.000
##   nc2           1.072   0.155   6.922   0.000   0.741
##   nc3          -0.884   0.152  -5.818   0.000  -0.611
##   nc4          -1.038   0.189  -5.496   0.000  -0.717
##   nc5          -1.018   0.184  -5.543   0.000  -0.704
##   nc6           0.682   0.155   4.395   0.000   0.472
##   nc7          -0.957   0.164  -5.826   0.000  -0.662
##   nc8          -0.787   0.143  -5.519   0.000  -0.544
##   nc9          -0.951   0.182  -5.210   0.000  -0.657
##   nc10          0.801   0.134   5.980   0.000   0.554
##   nc11          0.906   0.149   6.095   0.000   0.626
##   nc12         -0.910   0.162  -5.602   0.000  -0.629
##   nc13          0.865   0.122   7.095   0.000   0.598
##   nc14          0.800   0.129   6.218   0.000   0.553
##   nc15          0.567   0.148   3.819   0.000   0.392
##   nc16         -0.604   0.153  -3.956   0.000  -0.418
##   nc17         -0.836   0.170  -4.912   0.000  -0.578
##   nc18          0.319   0.140   2.277   0.023   0.221
##   Std.all
##
##   0.560
##   0.725

```

```

##      -0.577
##      -0.627
##      -0.676
##       0.394
##      -0.504
##      -0.448
##      -0.521
##       0.586
##       0.652
##      -0.594
##       0.522
##       0.506
##       0.367
##      -0.352
##      -0.513
##       0.183
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      3.241   0.089  36.537   0.000   3.241
##      .nc2      3.913   0.073  53.347   0.000   3.913
##      .nc3      1.821   0.076  23.943   0.000   1.821
##      .nc4      1.974   0.082  24.054   0.000   1.974
##      .nc5      1.831   0.075  24.490   0.000   1.831
##      .nc6      3.287   0.086  38.214   0.000   3.287
##      .nc7      2.487   0.094  26.363   0.000   2.487
##      .nc8      2.518   0.087  28.924   0.000   2.518
##      .nc9      2.267   0.091  25.034   0.000   2.267
##      .nc10     4.056   0.068  59.746   0.000   4.056
##      .nc11     4.205   0.069  60.997   0.000   4.205
##      .nc12     1.754   0.076  23.094   0.000   1.754
##      .nc13     3.287   0.082  39.966   0.000   3.287
##      .nc14     3.749   0.078  47.799   0.000   3.749
##      .nc15     3.677   0.077  47.929   0.000   3.677
##      .nc16     2.292   0.085  26.918   0.000   2.292
##      .nc17     1.903   0.081  23.538   0.000   1.903
##      .nc18     3.328   0.086  38.480   0.000   3.328
##      NC        0.000
## Std.all
##      2.623
##      3.830
##      1.719
##      1.727
##      1.758
##      2.744
##      1.893
##      2.077
##      1.797
##      4.290
##      4.379
##      1.658
##      2.869
##      3.432
##      3.441

```

```

##      1.933
##      1.690
##      2.763
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      1.049   0.131   7.997   0.000   1.049
##      .nc2      0.495   0.077   6.450   0.000   0.495
##      .nc3      0.748   0.116   6.440   0.000   0.748
##      .nc4      0.793   0.129   6.153   0.000   0.793
##      .nc5      0.588   0.085   6.945   0.000   0.588
##      .nc6      1.213   0.127   9.541   0.000   1.213
##      .nc7      1.289   0.129   9.990   0.000   1.289
##      .nc8      1.175   0.111  10.615   0.000   1.175
##      .nc9      1.159   0.136   8.526   0.000   1.159
##      .nc10     0.588   0.093   6.305   0.000   0.588
##      .nc11     0.530   0.084   6.311   0.000   0.530
##      .nc12     0.724   0.125   5.766   0.000   0.724
##      .nc13     0.955   0.092  10.336   0.000   0.955
##      .nc14     0.887   0.103   8.645   0.000   0.887
##      .nc15     0.988   0.132   7.515   0.000   0.988
##      .nc16     1.232   0.145   8.477   0.000   1.232
##      .nc17     0.934   0.113   8.255   0.000   0.934
##      .nc18     1.402   0.112  12.569   0.000   1.402
##      NC        0.478   0.125   3.823   0.000   1.000
## Std.all
##      0.687
##      0.474
##      0.667
##      0.606
##      0.543
##      0.845
##      0.746
##      0.799
##      0.728
##      0.657
##      0.574
##      0.647
##      0.728
##      0.744
##      0.866
##      0.876
##      0.737
##      0.966
##      1.000
##
## R-Square:
##      Estimate
##      nc1      0.313
##      nc2      0.526
##      nc3      0.333
##      nc4      0.394
##      nc5      0.457

```

```
##      nc6      0.155
##      nc7      0.254
##      nc8      0.201
##      nc9      0.272
##      nc10     0.343
##      nc11     0.426
##      nc12     0.353
##      nc13     0.272
##      nc14     0.256
##      nc15     0.134
##      nc16     0.124
##      nc17     0.263
##      nc18     0.034

MI <- modificationIndices(NC_Fit_1)
subset(MI, mi > 10)

##      lhs op  rhs      mi      epc sepc.lv sepc.all sepc.nox
## 105 nc4 ~~ nc5 15.72  0.218   0.218   0.319   0.319
## 157 nc8 ~~ nc11 11.43  0.205   0.205   0.260   0.260
## 158 nc8 ~~ nc12 11.24 -0.234 -0.234 -0.254 -0.254
## 174 nc10 ~~ nc11 11.57  0.149   0.149   0.268   0.268
## 195 nc13 ~~ nc14 13.90  0.258   0.258   0.281   0.281
## 198 nc13 ~~ nc17 10.62  0.232   0.232   0.246   0.246
## 202 nc14 ~~ nc17 12.34  0.240   0.240   0.264   0.264
```

We will free a parameter at a time, depending on the highest modification index at each step.

```
NC.model.2 <- "
# Latent variable definitions.
NC =~ nc1 + nc2 + nc3 + nc4 + nc5 + nc6 + nc7 + nc8 + nc9 + nc10 +
nc11 + nc12 + nc13 + nc14 + nc15 + nc16 + nc17 + nc18
nc4 ~~ nc5
"

NC_Fit_2 <- cfa(NC.model.2, data = NC, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(NC_Fit_2, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 29 iterations
##
##      Number of observations      195
##      Number of missing patterns    1
##
##      Estimator      ML      Robust
##      Model Fit Test Statistic    251.756    216.307
##      Degrees of freedom      134      134
##      P-value (Chi-square)      0.000      0.000
##      Scaling correction factor      1.164
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
```

```

## Minimum Function Test Statistic      1061.451      874.878
## Degrees of freedom                    153          153
## P-value                              0.000          0.000
##
## User model versus baseline model:
##
## Comparative Fit Index (CFI)           0.870          0.886
## Tucker-Lewis Index (TLI)             0.852          0.870
##
## Robust Comparative Fit Index (CFI)      NA
## Robust Tucker-Lewis Index (TLI)        NA
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0)          -4976.953      -4976.953
## Loglikelihood unrestricted model (H1)   -4850.426      -4850.426
##
## Number of free parameters              55           55
## Akaike (AIC)                          10063.906      10063.906
## Bayesian (BIC)                        10243.921      10243.921
## Sample-size adjusted Bayesian (BIC)    10069.689      10069.689
##
## Root Mean Square Error of Approximation:
##
## RMSEA                                0.067          0.056
## 90 Percent Confidence Interval          0.054  0.080          0.043  0.069
## P-value RMSEA <= 0.05                 0.015          0.205
##
## Robust RMSEA                          NA
## 90 Percent Confidence Interval          NA          NA
##
## Standardized Root Mean Square Residual:
##
## SRMR                                0.059          0.059
##
## Parameter Estimates:
##
## Information                          Observed
## Observed information based on          Hessian
## Standard Errors                      Robust.huber.white
##
## Latent Variables:
##
## Estimate Std.Err z-value P(>|z|) Std.lv
## NC =~
## nc1      1.000
## nc2      1.066  0.152  7.022  0.000  0.750
## nc3     -0.864  0.149 -5.800  0.000 -0.608
## nc4     -0.964  0.174 -5.548  0.000 -0.679
## nc5     -0.958  0.170 -5.632  0.000 -0.674
## nc6      0.682  0.153  4.463  0.000  0.480
## nc7     -0.932  0.158 -5.882  0.000 -0.656
## nc8     -0.771  0.139 -5.532  0.000 -0.543
## nc9     -0.928  0.174 -5.330  0.000 -0.653
## nc10     0.793  0.132  6.009  0.000  0.558

```

```

##      nc11      0.898    0.145    6.209    0.000    0.632
##      nc12     -0.886    0.158   -5.602    0.000   -0.623
##      nc13      0.861    0.119    7.223    0.000    0.606
##      nc14      0.801    0.127    6.318    0.000    0.564
##      nc15      0.573    0.146    3.920    0.000    0.403
##      nc16     -0.583    0.148   -3.924    0.000   -0.410
##      nc17     -0.815    0.165   -4.930    0.000   -0.573
##      nc18      0.320    0.138    2.313    0.021    0.225
## Std.all
##
##      0.570
##      0.735
##     -0.574
##     -0.594
##     -0.648
##      0.401
##     -0.499
##     -0.448
##     -0.518
##      0.590
##      0.658
##     -0.589
##      0.529
##      0.516
##      0.377
##     -0.346
##     -0.509
##      0.187
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc4 ~~
##      .nc5      0.220    0.070    3.134    0.002    0.220
## Std.all
##
##      0.301
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      3.241    0.089   36.537    0.000    3.241
##      .nc2      3.913    0.073   53.347    0.000    3.913
##      .nc3      1.821    0.076   23.943    0.000    1.821
##      .nc4      1.974    0.082   24.054    0.000    1.974
##      .nc5      1.831    0.075   24.490    0.000    1.831
##      .nc6      3.287    0.086   38.214    0.000    3.287
##      .nc7      2.487    0.094   26.363    0.000    2.487
##      .nc8      2.518    0.087   28.924    0.000    2.518
##      .nc9      2.267    0.091   25.034    0.000    2.267
##      .nc10     4.056    0.068   59.746    0.000    4.056
##      .nc11     4.205    0.069   60.997    0.000    4.205
##      .nc12     1.754    0.076   23.094    0.000    1.754
##      .nc13     3.287    0.082   39.966    0.000    3.287
##      .nc14     3.749    0.078   47.799    0.000    3.749
##      .nc15     3.677    0.077   47.929    0.000    3.677

```


##	.nc16	2.292	0.085	26.918	0.000	2.292
##	.nc17	1.903	0.081	23.538	0.000	1.903
##	.nc18	3.328	0.086	38.480	0.000	3.328
##	NC	0.000				0.000
##	Std.all					
##	2.623					
##	3.830					
##	1.719					
##	1.727					
##	1.758					
##	2.744					
##	1.893					
##	2.077					
##	1.797					
##	4.290					
##	4.379					
##	1.658					
##	2.869					
##	3.432					
##	3.441					
##	1.933					
##	1.690					
##	2.763					
##	0.000					
##						
##	Variances:					
##		Estimate	Std.Err	z-value	P(> z)	Std.lv
##	.nc1	1.031	0.130	7.932	0.000	1.031
##	.nc2	0.481	0.074	6.480	0.000	0.481
##	.nc3	0.752	0.116	6.460	0.000	0.752
##	.nc4	0.846	0.132	6.412	0.000	0.846
##	.nc5	0.629	0.083	7.576	0.000	0.629
##	.nc6	1.205	0.128	9.427	0.000	1.205
##	.nc7	1.296	0.129	10.044	0.000	1.296
##	.nc8	1.176	0.111	10.625	0.000	1.176
##	.nc9	1.164	0.137	8.520	0.000	1.164
##	.nc10	0.583	0.093	6.251	0.000	0.583
##	.nc11	0.523	0.084	6.246	0.000	0.523
##	.nc12	0.730	0.126	5.791	0.000	0.730
##	.nc13	0.945	0.092	10.267	0.000	0.945
##	.nc14	0.875	0.102	8.547	0.000	0.875
##	.nc15	0.979	0.131	7.474	0.000	0.979
##	.nc16	1.239	0.146	8.511	0.000	1.239
##	.nc17	0.939	0.113	8.334	0.000	0.939
##	.nc18	1.401	0.112	12.559	0.000	1.401
##	NC	0.496	0.124	3.983	0.000	1.000
##	Std.all					
##	0.675					
##	0.460					
##	0.670					
##	0.648					
##	0.580					
##	0.839					
##	0.751					

```
##      0.800
##      0.732
##      0.651
##      0.567
##      0.653
##      0.720
##      0.733
##      0.858
##      0.880
##      0.741
##      0.965
##      1.000
##
## R-Square:
##              Estimate
##      nc1          0.325
##      nc2          0.540
##      nc3          0.330
##      nc4          0.352
##      nc5          0.420
##      nc6          0.161
##      nc7          0.249
##      nc8          0.200
##      nc9          0.268
##      nc10         0.349
##      nc11         0.433
##      nc12         0.347
##      nc13         0.280
##      nc14         0.267
##      nc15         0.142
##      nc16         0.120
##      nc17         0.259
##      nc18         0.035
##
MI <- modificationIndices(NC_Fit_2)
subset(MI, mi > 10)

##      lhs op  rhs      mi      epc sepc.lv sepc.all sepc.nox
## 125 nc5 ~~ nc12 10.44  0.162   0.162   0.238   0.238
## 157 nc8 ~~ nc11 12.03  0.210   0.210   0.268   0.268
## 158 nc8 ~~ nc12 10.79 -0.231 -0.231 -0.249 -0.249
## 174 nc10 ~~ nc11 10.91  0.145   0.145   0.263   0.263
## 195 nc13 ~~ nc14 13.07  0.249   0.249   0.274   0.274
## 198 nc13 ~~ nc17 10.95  0.236   0.236   0.250   0.250
## 202 nc14 ~~ nc17 13.00  0.247   0.247   0.272   0.272
##
anova(NC_Fit_1, NC_Fit_2)

## Chi Square Difference Test
##
##              Df      AIC      BIC Chisq Chisq diff Df diff Pr(>Chisq)
## NC_Fit_2 134 10064 10244    252
## NC_Fit_1 135 10078 10254    267      15.6      1    0.00008
```

```

NC.model.3 <- "
# Latent variable definitions.
NC =~ nc1 + nc2 + nc3 + nc4 + nc5 + nc6 + nc7 + nc8 + nc9 + nc10 +
nc11 + nc12 + nc13 + nc14 + nc15 + nc16 + nc17 + nc18
nc4 ~~ nc5
nc13 ~~ nc14
"

NC_Fit_3 <- cfa(NC.model.3, data = NC, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(NC_Fit_3, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 30 iterations
##
##   Number of observations              195
##   Number of missing patterns          1
##
##   Estimator              ML          Robust
##   Model Fit Test Statistic      238.584      204.920
##   Degrees of freedom           133          133
##   P-value (Chi-square)         0.000          0.000
##   Scaling correction factor              1.164
##   for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##   Minimum Function Test Statistic      1061.451      874.878
##   Degrees of freedom           153          153
##   P-value              0.000          0.000
##
## User model versus baseline model:
##
##   Comparative Fit Index (CFI)           0.884          0.900
##   Tucker-Lewis Index (TLI)             0.866          0.885
##
##   Robust Comparative Fit Index (CFI)              NA
##   Robust Tucker-Lewis Index (TLI)                 NA
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)      -4970.333      -4970.333
##   Loglikelihood unrestricted model (H1) -4850.426      -4850.426
##
##   Number of free parameters           56          56
##   Akaike (AIC)              10052.666      10052.666
##   Bayesian (BIC)            10235.954      10235.954
##   Sample-size adjusted Bayesian (BIC) 10058.554      10058.554
##
## Root Mean Square Error of Approximation:
##
##   RMSEA              0.064          0.053
##   90 Percent Confidence Interval      0.051  0.077          0.039  0.066
##   P-value RMSEA <= 0.05              0.042          0.352

```

```

##
## Robust RMSEA NA
## 90 Percent Confidence Interval NA NA
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.057 0.057
##
## Parameter Estimates:
##
## Information Observed
## Observed information based on Hessian
## Standard Errors Robust.huber.white
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv
## NC =~
## nc1 1.000 0.699
## nc2 1.073 0.156 6.895 0.000 0.751
## nc3 -0.875 0.150 -5.825 0.000 -0.612
## nc4 -0.978 0.177 -5.511 0.000 -0.684
## nc5 -0.972 0.174 -5.576 0.000 -0.680
## nc6 0.683 0.154 4.429 0.000 0.478
## nc7 -0.944 0.160 -5.900 0.000 -0.660
## nc8 -0.775 0.141 -5.488 0.000 -0.542
## nc9 -0.944 0.178 -5.290 0.000 -0.660
## nc10 0.797 0.134 5.957 0.000 0.557
## nc11 0.903 0.149 6.072 0.000 0.631
## nc12 -0.889 0.161 -5.528 0.000 -0.621
## nc13 0.832 0.118 7.050 0.000 0.582
## nc14 0.772 0.126 6.108 0.000 0.540
## nc15 0.570 0.148 3.841 0.000 0.399
## nc16 -0.596 0.149 -4.001 0.000 -0.417
## nc17 -0.838 0.168 -4.985 0.000 -0.586
## nc18 0.321 0.140 2.301 0.021 0.225
## Std.all
##
## 0.566
## 0.735
## -0.578
## -0.598
## -0.653
## 0.399
## -0.502
## -0.447
## -0.524
## 0.589
## 0.657
## -0.588
## 0.508
## 0.494
## 0.373
## -0.352
## -0.520

```

```

##      0.187
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc4 ~~
##      .nc5      0.213   0.070   3.030   0.002   0.213
##      .nc13 ~~
##      .nc14      0.250   0.085   2.952   0.003   0.250
##      Std.all
##
##      0.295
##
##      0.267
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      3.241   0.089  36.537   0.000   3.241
##      .nc2      3.913   0.073  53.347   0.000   3.913
##      .nc3      1.821   0.076  23.943   0.000   1.821
##      .nc4      1.974   0.082  24.054   0.000   1.974
##      .nc5      1.831   0.075  24.490   0.000   1.831
##      .nc6      3.287   0.086  38.214   0.000   3.287
##      .nc7      2.487   0.094  26.363   0.000   2.487
##      .nc8      2.518   0.087  28.924   0.000   2.518
##      .nc9      2.267   0.091  25.034   0.000   2.267
##      .nc10     4.056   0.068  59.746   0.000   4.056
##      .nc11     4.205   0.069  60.997   0.000   4.205
##      .nc12     1.754   0.076  23.094   0.000   1.754
##      .nc13     3.287   0.082  39.966   0.000   3.287
##      .nc14     3.749   0.078  47.799   0.000   3.749
##      .nc15     3.677   0.077  47.929   0.000   3.677
##      .nc16     2.292   0.085  26.918   0.000   2.292
##      .nc17     1.903   0.081  23.538   0.000   1.903
##      .nc18     3.328   0.086  38.480   0.000   3.328
##      NC        0.000
##      Std.all
##      2.623
##      3.830
##      1.719
##      1.727
##      1.758
##      2.744
##      1.893
##      2.077
##      1.797
##      4.290
##      4.379
##      1.658
##      2.869
##      3.432
##      3.441
##      1.933
##      1.690
##      2.763

```

```

##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      1.038   0.130   7.964   0.000   1.038
##      .nc2      0.480   0.075   6.396   0.000   0.480
##      .nc3      0.747   0.116   6.456   0.000   0.747
##      .nc4      0.839   0.131   6.385   0.000   0.839
##      .nc5      0.622   0.082   7.552   0.000   0.622
##      .nc6      1.207   0.127   9.476   0.000   1.207
##      .nc7      1.291   0.128  10.051   0.000   1.291
##      .nc8      1.177   0.112  10.522   0.000   1.177
##      .nc9      1.154   0.136   8.459   0.000   1.154
##      .nc10     0.584   0.094   6.244   0.000   0.584
##      .nc11     0.524   0.084   6.252   0.000   0.524
##      .nc12     0.733   0.127   5.790   0.000   0.733
##      .nc13     0.974   0.092  10.613   0.000   0.974
##      .nc14     0.902   0.104   8.696   0.000   0.902
##      .nc15     0.983   0.131   7.497   0.000   0.983
##      .nc16     1.233   0.145   8.474   0.000   1.233
##      .nc17     0.924   0.112   8.241   0.000   0.924
##      .nc18     1.401   0.112  12.558   0.000   1.401
##      NC        0.489   0.125   3.919   0.000   1.000
## Std.all
##      0.680
##      0.460
##      0.666
##      0.642
##      0.574
##      0.841
##      0.748
##      0.800
##      0.726
##      0.653
##      0.568
##      0.655
##      0.742
##      0.756
##      0.861
##      0.876
##      0.729
##      0.965
##      1.000
##
## R-Square:
##      Estimate
##      nc1      0.320
##      nc2      0.540
##      nc3      0.334
##      nc4      0.358
##      nc5      0.426
##      nc6      0.159
##      nc7      0.252
##      nc8      0.200

```

```
##      nc9      0.274
##      nc10     0.347
##      nc11     0.432
##      nc12     0.345
##      nc13     0.258
##      nc14     0.244
##      nc15     0.139
##      nc16     0.124
##      nc17     0.271
##      nc18     0.035

MI <- modificationIndices(NC_Fit_3)
subset(MI, mi > 10)

##      lhs op  rhs      mi      epc sepc.lv sepc.all sepc.nox
## 126 nc5  ~~ nc12 10.37  0.161   0.161   0.239   0.239
## 158 nc8  ~~ nc11 11.89  0.210   0.210   0.267   0.267
## 159 nc8  ~~ nc12 10.58 -0.229 -0.229 -0.247 -0.247
## 175 nc10 ~~ nc11 11.14  0.147   0.147   0.266   0.266

anova(NC_Fit_2, NC_Fit_3)

## Chi Square Difference Test
##
##           Df    AIC    BIC Chisq Chisq diff Df diff Pr(>Chisq)
## NC_Fit_3 133 10053 10236   239
## NC_Fit_2 134 10064 10244   252      13.2      1   0.00028
```

```
NC.model.4 <- "
# Latent variable definitions.
NC =~ nc1 + nc2 + nc3 + nc4 + nc5 + nc6 + nc7 + nc8 + nc9 + nc10 +
nc11 + nc12 + nc13 + nc14 + nc15 + nc16 + nc17 + nc18
nc4 ~~ nc5
nc13 ~~ nc14
nc8 ~~ nc11
"
```

```
NC_Fit_4 <- cfa(NC.model.4, data = NC, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(NC_Fit_4, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 31 iterations
##
##      Number of observations      195
##      Number of missing patterns    1
##
##      Estimator      ML      Robust
##      Model Fit Test Statistic    225.806    193.454
##      Degrees of freedom      132      132
##      P-value (Chi-square)      0.000      0.000
##      Scaling correction factor      1.167
##      for the Yuan-Bentler correction (Mplus variant)
##
```

```

## Model test baseline model:
##
##   Minimum Function Test Statistic      1061.451      874.878
##   Degrees of freedom                   153          153
##   P-value                             0.000          0.000
##
## User model versus baseline model:
##
##   Comparative Fit Index (CFI)           0.897          0.915
##   Tucker-Lewis Index (TLI)             0.880          0.901
##
##   Robust Comparative Fit Index (CFI)           NA
##   Robust Tucker-Lewis Index (TLI)           NA
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)          -4963.911    -4963.911
##   Loglikelihood unrestricted model (H1)    -4850.426    -4850.426
##
##   Number of free parameters              57          57
##   Akaike (AIC)                          10041.823    10041.823
##   Bayesian (BIC)                        10228.384    10228.384
##   Sample-size adjusted Bayesian (BIC)     10047.816    10047.816
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                                0.061          0.049
##   90 Percent Confidence Interval          0.047  0.074          0.034  0.062
##   P-value RMSEA <= 0.05                 0.099          0.535
##
##   Robust RMSEA                           NA
##   90 Percent Confidence Interval          NA          NA
##
## Standardized Root Mean Square Residual:
##
##   SRMR                                0.056          0.056
##
## Parameter Estimates:
##
##   Information                          Observed
##   Observed information based on        Hessian
##   Standard Errors                     Robust.huber.white
##
## Latent Variables:
##
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##   NC =~
##   nc1           1.000
##   nc2           1.081    0.156    6.918    0.000    0.750
##   nc3          -0.870    0.151   -5.773    0.000   -0.603
##   nc4          -0.984    0.178   -5.524    0.000   -0.683
##   nc5          -0.969    0.173   -5.599    0.000   -0.672
##   nc6           0.686    0.155    4.424    0.000    0.476
##   nc7          -0.950    0.161   -5.892    0.000   -0.659
##   nc8          -0.836    0.148   -5.661    0.000   -0.580

```



```

##      nc9      -0.955    0.179   -5.340    0.000   -0.663
##      nc10     0.813    0.136    5.960    0.000    0.564
##      nc11     0.933    0.153    6.096    0.000    0.647
##      nc12    -0.875    0.160   -5.475    0.000   -0.607
##      nc13     0.842    0.119    7.053    0.000    0.584
##      nc14     0.781    0.127    6.152    0.000    0.542
##      nc15     0.573    0.150    3.817    0.000    0.398
##      nc16    -0.597    0.151   -3.957    0.000   -0.414
##      nc17    -0.854    0.170   -5.012    0.000   -0.592
##      nc18     0.314    0.141    2.230    0.026    0.218
## Std.all
##
##      0.561
##      0.734
##     -0.570
##     -0.597
##     -0.646
##      0.397
##     -0.502
##     -0.478
##     -0.525
##      0.596
##      0.674
##     -0.574
##      0.510
##      0.496
##      0.372
##     -0.349
##     -0.526
##      0.181
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc4 ~~
##      .nc5      0.218   0.070   3.135   0.002   0.218
##      .nc13 ~~
##      .nc14      0.248   0.085   2.917   0.004   0.248
##      .nc8 ~~
##      .nc11      0.213   0.059   3.620   0.000   0.213
## Std.all
##
##      0.300
##
##      0.265
##
##      0.281
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      3.241   0.089  36.537   0.000   3.241
##      .nc2      3.913   0.073  53.347   0.000   3.913
##      .nc3      1.821   0.076  23.943   0.000   1.821
##      .nc4      1.974   0.082  24.054   0.000   1.974
##      .nc5      1.831   0.075  24.490   0.000   1.831

```

##	.nc6	3.287	0.086	38.214	0.000	3.287
##	.nc7	2.487	0.094	26.363	0.000	2.487
##	.nc8	2.518	0.087	28.924	0.000	2.518
##	.nc9	2.267	0.091	25.034	0.000	2.267
##	.nc10	4.056	0.068	59.746	0.000	4.056
##	.nc11	4.205	0.069	60.997	0.000	4.205
##	.nc12	1.754	0.076	23.094	0.000	1.754
##	.nc13	3.287	0.082	39.966	0.000	3.287
##	.nc14	3.749	0.078	47.799	0.000	3.749
##	.nc15	3.677	0.077	47.929	0.000	3.677
##	.nc16	2.292	0.085	26.918	0.000	2.292
##	.nc17	1.903	0.081	23.538	0.000	1.903
##	.nc18	3.328	0.086	38.480	0.000	3.328
##	NC	0.000				0.000
##	Std.all					
##	2.623					
##	3.830					
##	1.719					
##	1.727					
##	1.758					
##	2.744					
##	1.893					
##	2.077					
##	1.797					
##	4.290					
##	4.379					
##	1.658					
##	2.869					
##	3.432					
##	3.441					
##	1.933					
##	1.690					
##	2.763					
##	0.000					
##						
##	Variances:					
##		Estimate	Std.Err	z-value	P(> z)	Std.lv
##	.nc1	1.045	0.130	8.050	0.000	1.045
##	.nc2	0.482	0.074	6.490	0.000	0.482
##	.nc3	0.758	0.117	6.466	0.000	0.758
##	.nc4	0.841	0.129	6.502	0.000	0.841
##	.nc5	0.632	0.083	7.575	0.000	0.632
##	.nc6	1.209	0.127	9.488	0.000	1.209
##	.nc7	1.292	0.127	10.184	0.000	1.292
##	.nc8	1.134	0.110	10.272	0.000	1.134
##	.nc9	1.152	0.135	8.527	0.000	1.152
##	.nc10	0.576	0.091	6.347	0.000	0.576
##	.nc11	0.503	0.080	6.253	0.000	0.503
##	.nc12	0.750	0.128	5.846	0.000	0.750
##	.nc13	0.971	0.092	10.608	0.000	0.971
##	.nc14	0.899	0.103	8.720	0.000	0.899
##	.nc15	0.984	0.130	7.543	0.000	0.984
##	.nc16	1.235	0.146	8.462	0.000	1.235
##	.nc17	0.917	0.111	8.244	0.000	0.917

```

##      .nc18      1.404    0.112    12.587    0.000    1.404
##      NC      0.481    0.124    3.895    0.000    1.000
## Std.all
##      0.685
##      0.461
##      0.676
##      0.643
##      0.583
##      0.842
##      0.748
##      0.771
##      0.724
##      0.644
##      0.546
##      0.671
##      0.740
##      0.754
##      0.862
##      0.878
##      0.723
##      0.967
##      1.000
##
## R-Square:
##      Estimate
##      nc1      0.315
##      nc2      0.539
##      nc3      0.324
##      nc4      0.357
##      nc5      0.417
##      nc6      0.158
##      nc7      0.252
##      nc8      0.229
##      nc9      0.276
##      nc10     0.356
##      nc11     0.454
##      nc12     0.329
##      nc13     0.260
##      nc14     0.246
##      nc15     0.138
##      nc16     0.122
##      nc17     0.277
##      nc18     0.033

MI <- modificationIndices(NC_Fit_4)
subset(MI, mi > 10)

##      lhs op  rhs      mi      epc sepc.lv sepc.all sepc.nox
## 127 nc5  ~~ nc12 11.45  0.171   0.171   0.248   0.248
## 159 nc8  ~~ nc12 14.05 -0.261 -0.261 -0.283 -0.283
## 175 nc10 ~~ nc11 11.31  0.144   0.144   0.268   0.268

anova(NC_Fit_3, NC_Fit_4)

## Chi Square Difference Test

```

```
##
##           Df    AIC    BIC Chisq Chisq diff Df diff Pr(>Chisq)
## NC_Fit_4 132 10042 10228   226
## NC_Fit_3 133 10053 10236   239      12.8      1    0.00035
```

```
NC.model.5 <- "
# Latent variable definitions.
NC =~ nc1 + nc2 + nc3 + nc4 + nc5 + nc6 + nc7 + nc8 + nc9 + nc10 +
nc11 + nc12 + nc13 + nc14 + nc15 + nc16 + nc17 + nc18
nc4 ~~ nc5
nc13 ~~ nc14
nc8 ~~ nc11
nc8 ~~ nc12
"
```

```
NC_Fit_5 <- cfa(NC.model.5, data = NC, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(NC_Fit_5, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 30 iterations
##
##      Number of observations              195
##      Number of missing patterns          1
##
##      Estimator                      ML      Robust
##      Model Fit Test Statistic        210.413    179.711
##      Degrees of freedom              131       131
##      P-value (Chi-square)            0.000      0.003
##      Scaling correction factor                1.171
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##      Minimum Function Test Statistic    1061.451    874.878
##      Degrees of freedom                153       153
##      P-value                          0.000      0.000
##
## User model versus baseline model:
##
##      Comparative Fit Index (CFI)        0.913      0.933
##      Tucker-Lewis Index (TLI)          0.898      0.921
##
##      Robust Comparative Fit Index (CFI)                NA
##      Robust Tucker-Lewis Index (TLI)                   NA
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)      -4956.175   -4956.175
##      Loglikelihood unrestricted model (H1) -4850.426   -4850.426
##
##      Number of free parameters          58       58
##      Akaike (AIC)                      10028.350   10028.350
```

```

## Bayesian (BIC) 10218.184 10218.184
## Sample-size adjusted Bayesian (BIC) 10034.448 10034.448
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.056 0.044
## 90 Percent Confidence Interval 0.041 0.070 0.028 0.058
## P-value RMSEA <= 0.05 0.236 0.755
##
## Robust RMSEA NA
## 90 Percent Confidence Interval NA NA
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.054 0.054
##
## Parameter Estimates:
##
## Information Observed
## Observed information based on Hessian
## Standard Errors Robust.huber.white
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv
## NC =~
## nc1 1.000 0.688
## nc2 1.081 0.154 7.011 0.000 0.744
## nc3 -0.854 0.148 -5.762 0.000 -0.587
## nc4 -0.993 0.176 -5.639 0.000 -0.683
## nc5 -0.984 0.175 -5.628 0.000 -0.677
## nc6 0.686 0.156 4.395 0.000 0.472
## nc7 -0.969 0.165 -5.865 0.000 -0.667
## nc8 -0.895 0.150 -5.955 0.000 -0.616
## nc9 -0.953 0.177 -5.392 0.000 -0.656
## nc10 0.812 0.133 6.103 0.000 0.559
## nc11 0.940 0.152 6.192 0.000 0.647
## nc12 -0.927 0.166 -5.588 0.000 -0.638
## nc13 0.851 0.121 7.051 0.000 0.586
## nc14 0.798 0.128 6.229 0.000 0.549
## nc15 0.570 0.148 3.860 0.000 0.392
## nc16 -0.595 0.150 -3.965 0.000 -0.410
## nc17 -0.863 0.167 -5.173 0.000 -0.594
## nc18 0.306 0.142 2.151 0.032 0.211
## Std.all
##
## 0.557
## 0.728
## -0.555
## -0.598
## -0.650
## 0.394
## -0.507
## -0.507

```

```

##      -0.520
##      0.591
##      0.674
##     -0.603
##      0.511
##      0.503
##      0.367
##     -0.345
##     -0.528
##      0.175
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc4 ~~
##      .nc5      0.215   0.069   3.135   0.002   0.215
##      .nc13 ~~
##      .nc14      0.243   0.083   2.906   0.004   0.243
##      .nc8 ~~
##      .nc11      0.240   0.058   4.126   0.000   0.240
##      .nc12     -0.266   0.064  -4.165   0.000  -0.266
##      Std.all
##
##      0.297
##
##      0.261
##
##      0.324
##     -0.301
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      3.241   0.089  36.537   0.000   3.241
##      .nc2      3.913   0.073  53.347   0.000   3.913
##      .nc3      1.821   0.076  23.943   0.000   1.821
##      .nc4      1.974   0.082  24.054   0.000   1.974
##      .nc5      1.831   0.075  24.490   0.000   1.831
##      .nc6      3.287   0.086  38.214   0.000   3.287
##      .nc7      2.487   0.094  26.363   0.000   2.487
##      .nc8      2.518   0.087  28.924   0.000   2.518
##      .nc9      2.267   0.091  25.034   0.000   2.267
##      .nc10     4.056   0.068  59.746   0.000   4.056
##      .nc11     4.205   0.069  60.997   0.000   4.205
##      .nc12     1.754   0.076  23.094   0.000   1.754
##      .nc13     3.287   0.082  39.966   0.000   3.287
##      .nc14     3.749   0.078  47.799   0.000   3.749
##      .nc15     3.677   0.077  47.929   0.000   3.677
##      .nc16     2.292   0.085  26.918   0.000   2.292
##      .nc17     1.903   0.081  23.538   0.000   1.903
##      .nc18     3.328   0.086  38.480   0.000   3.328
##      NC        0.000
##      Std.all
##      2.623
##      3.830
##      1.719

```

```

##      1.727
##      1.758
##      2.744
##      1.893
##      2.074
##      1.797
##      4.290
##      4.379
##      1.658
##      2.869
##      3.432
##      3.441
##      1.933
##      1.690
##      2.763
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      1.053   0.129   8.139   0.000   1.053
##      .nc2      0.490   0.074   6.636   0.000   0.490
##      .nc3      0.777   0.120   6.494   0.000   0.777
##      .nc4      0.840   0.128   6.590   0.000   0.840
##      .nc5      0.626   0.083   7.580   0.000   0.626
##      .nc6      1.212   0.128   9.503   0.000   1.212
##      .nc7      1.282   0.126  10.176   0.000   1.282
##      .nc8      1.095   0.108  10.175   0.000   1.095
##      .nc9      1.160   0.135   8.583   0.000   1.160
##      .nc10     0.582   0.089   6.504   0.000   0.582
##      .nc11     0.504   0.080   6.265   0.000   0.504
##      .nc12     0.712   0.126   5.642   0.000   0.712
##      .nc13     0.969   0.090  10.755   0.000   0.969
##      .nc14     0.891   0.101   8.787   0.000   0.891
##      .nc15     0.988   0.129   7.666   0.000   0.988
##      .nc16     1.239   0.147   8.451   0.000   1.239
##      .nc17     0.914   0.110   8.325   0.000   0.914
##      .nc18     1.407   0.112  12.608   0.000   1.407
##      NC        0.474   0.122   3.889   0.000   1.000
## Std.all
##      0.690
##      0.469
##      0.692
##      0.643
##      0.577
##      0.845
##      0.743
##      0.742
##      0.729
##      0.651
##      0.546
##      0.636
##      0.739
##      0.747
##      0.865

```

```

##      0.881
##      0.721
##      0.969
##      1.000
##
## R-Square:
##              Estimate
##      nc1          0.310
##      nc2          0.531
##      nc3          0.308
##      nc4          0.357
##      nc5          0.423
##      nc6          0.155
##      nc7          0.257
##      nc8          0.258
##      nc9          0.271
##      nc10         0.349
##      nc11         0.454
##      nc12         0.364
##      nc13         0.261
##      nc14         0.253
##      nc15         0.135
##      nc16         0.119
##      nc17         0.279
##      nc18         0.031

MI <- modificationIndices(NC_Fit_5)
subset(MI, mi > 10)

##      lhs op  rhs      mi  epc sepc.lv sepc.all sepc.nox
## 175 nc10 ~~ nc11 10.54 0.14    0.14    0.258    0.258

anova(NC_Fit_4, NC_Fit_5)

## Chi Square Difference Test
##
##              Df    AIC    BIC Chisq Chisq diff Df diff Pr(>Chisq)
## NC_Fit_5 131 10028 10218   210
## NC_Fit_4 132 10042 10228   226      15.4      1 0.000087

NC.model.6 <- "
# Latent variable definitions.
NC =~ nc1 + nc2 + nc3 + nc4 + nc5 + nc6 + nc7 + nc8 + nc9 + nc10 +
nc11 + nc12 + nc13 + nc14 + nc15 + nc16 + nc17 + nc18
nc4 ~~ nc5
nc13 ~~ nc14
nc8 ~~ nc11
nc8 ~~ nc12
nc10 ~~ nc11
"

```



```

NC_Fit_6 <- cfa(NC.model.6, data = NC, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(NC_Fit_6, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 30 iterations
##
##   Number of observations              195
##   Number of missing patterns          1
##
##   Estimator              ML          Robust
##   Model Fit Test Statistic      200.120      171.533
##   Degrees of freedom            130          130
##   P-value (Chi-square)          0.000          0.009
##   Scaling correction factor              1.167
##   for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##   Minimum Function Test Statistic      1061.451      874.878
##   Degrees of freedom                    153          153
##   P-value                              0.000          0.000
##
## User model versus baseline model:
##
##   Comparative Fit Index (CFI)          0.923          0.942
##   Tucker-Lewis Index (TLI)            0.909          0.932
##
##   Robust Comparative Fit Index (CFI)              NA
##   Robust Tucker-Lewis Index (TLI)                NA
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)          -4951.002      -4951.002
##   Loglikelihood unrestricted model (H1)    -4850.426      -4850.426
##
##   Number of free parameters              59          59
##   Akaike (AIC)                          10020.004      10020.004
##   Bayesian (BIC)                        10213.111      10213.111
##   Sample-size adjusted Bayesian (BIC)      10026.208      10026.208
##
## Root Mean Square Error of Approximation:
##
##   RMSEA              0.053          0.041
##   90 Percent Confidence Interval      0.038  0.067          0.023  0.055
##   P-value RMSEA <= 0.05              0.365          0.849
##
##   Robust RMSEA              NA
##   90 Percent Confidence Interval      NA          NA
##
## Standardized Root Mean Square Residual:
##
##   SRMR              0.053          0.053
##
## Parameter Estimates:

```

```

##
##      Information                      Observed
##      Observed information based on      Hessian
##      Standard Errors                   Robust.huber.white
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##      NC =~
##      nc1          1.000
##      nc2          1.078    0.153    7.031    0.000    0.749
##      nc3         -0.848    0.147   -5.777    0.000   -0.589
##      nc4         -0.987    0.177   -5.584    0.000   -0.686
##      nc5         -0.985    0.177   -5.574    0.000   -0.684
##      nc6          0.676    0.157    4.317    0.000    0.470
##      nc7         -0.970    0.165   -5.881    0.000   -0.674
##      nc8         -0.882    0.148   -5.943    0.000   -0.613
##      nc9         -0.940    0.178   -5.278    0.000   -0.653
##      nc10          0.756    0.128    5.922    0.000    0.526
##      nc11          0.899    0.146    6.143    0.000    0.625
##      nc12         -0.928    0.166   -5.595    0.000   -0.645
##      nc13          0.838    0.121    6.944    0.000    0.582
##      nc14          0.787    0.127    6.182    0.000    0.546
##      nc15          0.564    0.146    3.876    0.000    0.392
##      nc16         -0.599    0.148   -4.057    0.000   -0.416
##      nc17         -0.863    0.166   -5.199    0.000   -0.600
##      nc18          0.320    0.142    2.247    0.025    0.222
##      Std.all
##
##      0.562
##      0.733
##      -0.556
##      -0.600
##      -0.657
##      0.392
##      -0.513
##      -0.506
##      -0.518
##      0.556
##      0.651
##      -0.610
##      0.508
##      0.500
##      0.367
##      -0.351
##      -0.533
##      0.184
##
## Covariances:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##      .nc4 ~~
##      .nc5          0.208    0.069    3.002    0.003    0.208
##      .nc13 ~~
##      .nc14          0.246    0.084    2.924    0.003    0.246
##      .nc8 ~~

```

```

##      .nc11      0.219    0.061    3.570    0.000    0.219
##      .nc12     -0.257    0.063   -4.076    0.000   -0.257
##      .nc10 ~~
##      .nc11      0.140    0.074    1.896    0.058    0.140
##      Std.all
##
##      0.290
##
##      0.264
##
##      0.288
##     -0.294
##
##      0.245
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      3.241    0.089   36.537    0.000    3.241
##      .nc2      3.913    0.073   53.347    0.000    3.913
##      .nc3      1.821    0.076   23.943    0.000    1.821
##      .nc4      1.974    0.082   24.054    0.000    1.974
##      .nc5      1.831    0.075   24.490    0.000    1.831
##      .nc6      3.287    0.086   38.214    0.000    3.287
##      .nc7      2.487    0.094   26.363    0.000    2.487
##      .nc8      2.518    0.087   28.924    0.000    2.518
##      .nc9      2.267    0.091   25.034    0.000    2.267
##      .nc10     4.056    0.068   59.746    0.000    4.056
##      .nc11     4.205    0.069   60.997    0.000    4.205
##      .nc12     1.754    0.076   23.094    0.000    1.754
##      .nc13     3.287    0.082   39.966    0.000    3.287
##      .nc14     3.749    0.078   47.799    0.000    3.749
##      .nc15     3.677    0.077   47.929    0.000    3.677
##      .nc16     2.292    0.085   26.918    0.000    2.292
##      .nc17     1.903    0.081   23.538    0.000    1.903
##      .nc18     3.328    0.086   38.480    0.000    3.328
##      NC        0.000
##      Std.all
##      2.623
##      3.830
##      1.719
##      1.727
##      1.758
##      2.744
##      1.893
##      2.079
##      1.797
##      4.290
##      4.381
##      1.658
##      2.869
##      3.432
##      3.441
##      1.933
##      1.690

```

```

##      2.763
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      1.044   0.130   8.012   0.000   1.044
##      .nc2      0.482   0.073   6.646   0.000   0.482
##      .nc3      0.775   0.119   6.493   0.000   0.775
##      .nc4      0.837   0.128   6.511   0.000   0.837
##      .nc5      0.616   0.083   7.444   0.000   0.616
##      .nc6      1.215   0.128   9.503   0.000   1.215
##      .nc7      1.272   0.126  10.118   0.000   1.272
##      .nc8      1.091   0.107  10.166   0.000   1.091
##      .nc9      1.164   0.136   8.538   0.000   1.164
##      .nc10     0.618   0.100   6.203   0.000   0.618
##      .nc11     0.531   0.087   6.088   0.000   0.531
##      .nc12     0.703   0.125   5.631   0.000   0.703
##      .nc13     0.974   0.090  10.788   0.000   0.974
##      .nc14     0.895   0.102   8.770   0.000   0.895
##      .nc15     0.988   0.129   7.637   0.000   0.988
##      .nc16     1.234   0.146   8.437   0.000   1.234
##      .nc17     0.908   0.110   8.269   0.000   0.908
##      .nc18     1.402   0.112  12.519   0.000   1.402
##      NC        0.483   0.124   3.878   0.000   1.000
## Std.all
##      0.684
##      0.462
##      0.691
##      0.640
##      0.568
##      0.846
##      0.737
##      0.744
##      0.732
##      0.691
##      0.576
##      0.628
##      0.742
##      0.750
##      0.865
##      0.877
##      0.716
##      0.966
##      1.000
##
## R-Square:
##      Estimate
##      nc1      0.316
##      nc2      0.538
##      nc3      0.309
##      nc4      0.360
##      nc5      0.432
##      nc6      0.154
##      nc7      0.263

```

```
##      nc8      0.256
##      nc9      0.268
##      nc10     0.309
##      nc11     0.424
##      nc12     0.372
##      nc13     0.258
##      nc14     0.250
##      nc15     0.135
##      nc16     0.123
##      nc17     0.284
##      nc18     0.034

MI <- modificationIndices(NC_Fit_6)
subset(MI, mi > 5)

##      lhs op  rhs      mi      epc sepc.lv sepc.all sepc.nox
## 80  nc2  ~~  nc4  7.524  0.134   0.134   0.211   0.211
## 99  nc3  ~~  nc8  5.883 -0.167  -0.167  -0.181  -0.181
## 101 nc3  ~~  nc10 6.198 -0.125  -0.125  -0.180  -0.180
## 102 nc3  ~~  nc11 5.041  0.107   0.107   0.167   0.167
## 113 nc4  ~~  nc9  5.728  0.169   0.169   0.172   0.172
## 129 nc5  ~~  nc12 8.664  0.145   0.145   0.220   0.220
## 198 nc13 ~~  nc17 5.849  0.164   0.164   0.174   0.174
## 202 nc14 ~~  nc17 8.331  0.187   0.187   0.208   0.208
## 207 nc16 ~~  nc17 6.678  0.203   0.203   0.192   0.192

anova(NC_Fit_5, NC_Fit_6)

## Chi Square Difference Test
##
##      Df      AIC      BIC Chisq Chisq diff Df diff Pr(>Chisq)
## NC_Fit_6 130 10020 10213   200
## NC_Fit_5 131 10028 10218   210      10.3      1    0.0013
```

```
NC.model.7 <- "
# Latent variable definitions.
NC =~ nc1 + nc2 + nc3 + nc4 + nc5 + nc6 + nc7 + nc8 + nc9 + nc10 +
nc11 + nc12 + nc13 + nc14 + nc15 + nc16 + nc17 + nc18
nc4 ~~ nc5
nc13 ~~ nc14
nc8 ~~ nc11
nc8 ~~ nc12
nc10 ~~ nc11
nc5 ~~ nc12
"
```

```
NC_Fit_7 <- cfa(NC.model.7, data = NC, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(NC_Fit_7, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 31 iterations
##
##      Number of observations      195
```

```

##      Number of missing patterns              1
##
##      Estimator              ML      Robust
##      Model Fit Test Statistic      191.642      164.452
##      Degrees of freedom              129              129
##      P-value (Chi-square)              0.000              0.019
##      Scaling correction factor              1.165
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##      Minimum Function Test Statistic      1061.451      874.878
##      Degrees of freedom              153              153
##      P-value              0.000              0.000
##
## User model versus baseline model:
##
##      Comparative Fit Index (CFI)              0.931              0.951
##      Tucker-Lewis Index (TLI)              0.918              0.942
##
##      Robust Comparative Fit Index (CFI)              NA
##      Robust Tucker-Lewis Index (TLI)              NA
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)      -4946.741      -4946.741
##      Loglikelihood unrestricted model (H1)      -4850.426      -4850.426
##
##      Number of free parameters              60              60
##      Akaike (AIC)              10013.482      10013.482
##      Bayesian (BIC)              10209.862      10209.862
##      Sample-size adjusted Bayesian (BIC)      10019.791      10019.791
##
## Root Mean Square Error of Approximation:
##
##      RMSEA              0.050              0.038
##      90 Percent Confidence Interval      0.034      0.064              0.018      0.053
##      P-value RMSEA <= 0.05              0.484              0.908
##
##      Robust RMSEA              NA
##      90 Percent Confidence Interval              NA      NA
##
## Standardized Root Mean Square Residual:
##
##      SRMR              0.052              0.052
##
## Parameter Estimates:
##
##      Information              Observed
##      Observed information based on              Hessian
##      Standard Errors              Robust.huber.white
##
## Latent Variables:
##      Estimate      Std.Err      z-value      P(>|z|)      Std.lv

```

```

##      NC =~
##      nc1          1.000          0.151      7.090      0.000      0.700
##      nc2          1.071      0.151      7.090      0.000      0.750
##      nc3         -0.843      0.148     -5.716      0.000     -0.590
##      nc4         -0.980      0.175     -5.586      0.000     -0.686
##      nc5         -0.940      0.165     -5.693      0.000     -0.658
##      nc6          0.674      0.155      4.359      0.000      0.472
##      nc7         -0.956      0.161     -5.953      0.000     -0.669
##      nc8         -0.874      0.147     -5.953      0.000     -0.612
##      nc9         -0.944      0.177     -5.347      0.000     -0.661
##      nc10          0.758      0.129      5.872      0.000      0.531
##      nc11          0.899      0.145      6.181      0.000      0.629
##      nc12         -0.889      0.157     -5.662      0.000     -0.622
##      nc13          0.833      0.119      6.980      0.000      0.583
##      nc14          0.783      0.127      6.166      0.000      0.548
##      nc15          0.569      0.146      3.887      0.000      0.399
##      nc16         -0.598      0.148     -4.041      0.000     -0.418
##      nc17         -0.863      0.166     -5.197      0.000     -0.604
##      nc18          0.321      0.142      2.261      0.024      0.225
##      Std.all
##
##      0.567
##      0.734
##      -0.557
##      -0.600
##      -0.633
##      0.394
##      -0.509
##      -0.506
##      -0.524
##      0.561
##      0.655
##      -0.589
##      0.509
##      0.502
##      0.373
##      -0.353
##      -0.537
##      0.186
##
## Covariances:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##      .nc4  ~~
##      .nc5          0.219      0.065      3.365      0.001      0.219
##      .nc13  ~~
##      .nc14          0.245      0.085      2.867      0.004      0.245
##      .nc8   ~~
##      .nc11          0.216      0.061      3.532      0.000      0.216
##      .nc12         -0.233      0.062     -3.731      0.000     -0.233
##      .nc10  ~~
##      .nc11          0.136      0.074      1.846      0.065      0.136
##      .nc5   ~~
##      .nc12          0.144      0.065      2.233      0.026      0.144
##      Std.all

```

```

##
##      0.298
##
##      0.263
##
##      0.285
##     -0.261
##
##      0.239
##
##      0.210
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      3.241   0.089  36.537   0.000   3.241
##      .nc2      3.913   0.073  53.347   0.000   3.913
##      .nc3      1.821   0.076  23.943   0.000   1.821
##      .nc4      1.974   0.082  24.054   0.000   1.974
##      .nc5      1.831   0.075  24.490   0.000   1.831
##      .nc6      3.287   0.086  38.214   0.000   3.287
##      .nc7      2.487   0.094  26.363   0.000   2.487
##      .nc8      2.518   0.087  28.924   0.000   2.518
##      .nc9      2.267   0.091  25.034   0.000   2.267
##      .nc10     4.056   0.068  59.746   0.000   4.056
##      .nc11     4.205   0.069  60.997   0.000   4.205
##      .nc12     1.754   0.076  23.094   0.000   1.754
##      .nc13     3.287   0.082  39.966   0.000   3.287
##      .nc14     3.749   0.078  47.799   0.000   3.749
##      .nc15     3.677   0.077  47.929   0.000   3.677
##      .nc16     2.292   0.085  26.918   0.000   2.292
##      .nc17     1.903   0.081  23.538   0.000   1.903
##      .nc18     3.328   0.086  38.480   0.000   3.328
##      NC        0.000
##      Std.all
##      2.623
##      3.830
##      1.719
##      1.727
##      1.761
##      2.744
##      1.893
##      2.082
##      1.797
##      4.290
##      4.380
##      1.660
##      2.869
##      3.432
##      3.441
##      1.933
##      1.690
##      2.763
##      0.000
##

```



```

## Variances:
##           Estimate Std.Err  z-value  P(>|z|)  Std.lv
##      .nc1          1.037    0.129    8.006    0.000    1.037
##      .nc2          0.481    0.072    6.644    0.000    0.481
##      .nc3          0.773    0.119    6.474    0.000    0.773
##      .nc4          0.837    0.129    6.501    0.000    0.837
##      .nc5          0.647    0.089    7.300    0.000    0.647
##      .nc6          1.213    0.128    9.465    0.000    1.213
##      .nc7          1.279    0.125   10.192    0.000    1.279
##      .nc8          1.088    0.106   10.216    0.000    1.088
##      .nc9          1.154    0.135    8.567    0.000    1.154
##      .nc10         0.613    0.100    6.127    0.000    0.613
##      .nc11         0.526    0.088    6.007    0.000    0.526
##      .nc12         0.728    0.125    5.842    0.000    0.728
##      .nc13         0.972    0.091   10.644    0.000    0.972
##      .nc14         0.893    0.103    8.685    0.000    0.893
##      .nc15         0.983    0.130    7.558    0.000    0.983
##      .nc16         1.232    0.146    8.437    0.000    1.232
##      .nc17         0.903    0.109    8.278    0.000    0.903
##      .nc18         1.401    0.112   12.519    0.000    1.401
##      NC            0.490    0.124    3.940    0.000    1.000
## Std.all
##      0.679
##      0.461
##      0.689
##      0.640
##      0.599
##      0.845
##      0.741
##      0.744
##      0.726
##      0.685
##      0.570
##      0.653
##      0.741
##      0.748
##      0.861
##      0.876
##      0.712
##      0.965
##      1.000
##
## R-Square:
##           Estimate
##      nc1          0.321
##      nc2          0.539
##      nc3          0.311
##      nc4          0.360
##      nc5          0.401
##      nc6          0.155
##      nc7          0.259
##      nc8          0.256
##      nc9          0.274
##      nc10         0.315

```

```
##      nc11      0.430
##      nc12      0.347
##      nc13      0.259
##      nc14      0.252
##      nc15      0.139
##      nc16      0.124
##      nc17      0.288
##      nc18      0.035

MI <- modificationIndices(NC_Fit_7)
subset(MI, mi > 5)

##      lhs op  rhs      mi      epc sepc.lv sepc.all sepc.nox
## 81  nc2  ~~  nc4 9.007  0.148  0.148  0.233  0.233
## 97  nc3  ~~  nc5 6.003  0.122  0.122  0.172  0.172
## 100 nc3  ~~  nc8 5.889 -0.167 -0.167 -0.182 -0.182
## 102 nc3  ~~  nc10 5.941 -0.122 -0.122 -0.178 -0.178
## 103 nc3  ~~  nc11 5.313  0.110  0.110  0.173  0.173
## 104 nc3  ~~  nc12 5.233 -0.127 -0.127 -0.170 -0.170
## 198 nc13 ~~  nc17 6.127  0.168  0.168  0.179  0.179
## 202 nc14 ~~  nc17 8.709  0.191  0.191  0.213  0.213
## 207 nc16 ~~  nc17 6.552  0.201  0.201  0.190  0.190

anova(NC_Fit_6, NC_Fit_7)

## Chi Square Difference Test
##
##      Df      AIC      BIC Chisq Chisq diff Df diff Pr(>Chisq)
## NC_Fit_7 129 10013 10210    192
## NC_Fit_6 130 10020 10213    200      8.48      1      0.0036
```

```
NC.model.8 <- "
# Latent variable definitions.
NC =~ nc1 + nc2 + nc3 + nc4 + nc5 + nc6 + nc7 + nc8 + nc9 + nc10 +
nc11 + nc12 + nc13 + nc14 + nc15 + nc16 + nc17 + nc18
nc4 ~~ nc5
nc13 ~~ nc14
nc8 ~~ nc11
nc8 ~~ nc12
nc10 ~~ nc11
nc5 ~~ nc12
nc2 ~~ nc4
"
```

```
NC_Fit_8 <- cfa(NC.model.8, data = NC, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(NC_Fit_8, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 31 iterations
##
##      Number of observations      195
##      Number of missing patterns      1
##
```

```

##      Estimator                      ML      Robust
##      Model Fit Test Statistic      182.118    156.424
##      Degrees of freedom              128       128
##      P-value (Chi-square)           0.001     0.044
##      Scaling correction factor              1.164
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##      Minimum Function Test Statistic      1061.451    874.878
##      Degrees of freedom              153       153
##      P-value              0.000     0.000
##
## User model versus baseline model:
##
##      Comparative Fit Index (CFI)           0.940     0.961
##      Tucker-Lewis Index (TLI)            0.929     0.953
##
##      Robust Comparative Fit Index (CFI)              NA
##      Robust Tucker-Lewis Index (TLI)              NA
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)      -4941.954    -4941.954
##      Loglikelihood unrestricted model (H1) -4850.426    -4850.426
##
##      Number of free parameters              61       61
##      Akaike (AIC)              10005.909    10005.909
##      Bayesian (BIC)              10205.562    10205.562
##      Sample-size adjusted Bayesian (BIC)      10012.323    10012.323
##
## Root Mean Square Error of Approximation:
##
##      RMSEA              0.047     0.034
##      90 Percent Confidence Interval      0.030  0.062     0.010  0.050
##      P-value RMSEA <= 0.05              0.626     0.954
##
##      Robust RMSEA              NA
##      90 Percent Confidence Interval      NA      NA
##
## Standardized Root Mean Square Residual:
##
##      SRMR              0.052     0.052
##
## Parameter Estimates:
##
##      Information              Observed
##      Observed information based on      Hessian
##      Standard Errors              Robust.huber.white
##
## Latent Variables:
##
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##      NC =~
##      nc1              1.000              0.694

```

```

##      nc2      1.104    0.158    6.998    0.000    0.766
##      nc3     -0.853    0.149   -5.738    0.000   -0.592
##      nc4     -1.036    0.182   -5.690    0.000   -0.719
##      nc5     -0.945    0.168   -5.634    0.000   -0.656
##      nc6      0.664    0.155    4.271    0.000    0.461
##      nc7     -0.958    0.160   -5.977    0.000   -0.665
##      nc8     -0.873    0.145   -6.034    0.000   -0.606
##      nc9     -0.952    0.178   -5.351    0.000   -0.661
##      nc10     0.756    0.130    5.826    0.000    0.525
##      nc11     0.904    0.147    6.159    0.000    0.627
##      nc12    -0.891    0.155   -5.730    0.000   -0.618
##      nc13     0.834    0.120    6.956    0.000    0.579
##      nc14     0.782    0.128    6.122    0.000    0.543
##      nc15     0.572    0.148    3.867    0.000    0.397
##      nc16    -0.604    0.148   -4.085    0.000   -0.419
##      nc17    -0.879    0.167   -5.256    0.000   -0.610
##      nc18     0.326    0.142    2.295    0.022    0.226
## Std.all
##
##      0.562
##      0.750
##     -0.559
##     -0.627
##     -0.630
##      0.385
##     -0.506
##     -0.501
##     -0.524
##      0.555
##      0.653
##     -0.585
##      0.505
##      0.497
##      0.371
##     -0.354
##     -0.542
##      0.188
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc4 ~~
##      .nc5      0.212   0.064   3.339   0.001   0.212
##      .nc13 ~~
##      .nc14      0.250   0.086   2.903   0.004   0.250
##      .nc8 ~~
##      .nc11      0.212   0.061   3.497   0.000   0.212
##      .nc12     -0.228   0.062  -3.682   0.000  -0.228
##      .nc10 ~~
##      .nc11      0.142   0.074   1.929   0.054   0.142
##      .nc5 ~~
##      .nc12      0.156   0.064   2.430   0.015   0.156
##      .nc2 ~~
##      .nc4      0.148   0.058   2.534   0.011   0.148
## Std.all

```

```

##
##      0.294
##
##      0.267
##
##      0.278
##     -0.255
##
##      0.248
##
##      0.226
##
##      0.245
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      3.241   0.089  36.537   0.000   3.241
##      .nc2      3.913   0.073  53.347   0.000   3.913
##      .nc3      1.821   0.076  23.943   0.000   1.821
##      .nc4      1.974   0.082  24.054   0.000   1.974
##      .nc5      1.831   0.075  24.490   0.000   1.831
##      .nc6      3.287   0.086  38.214   0.000   3.287
##      .nc7      2.487   0.094  26.363   0.000   2.487
##      .nc8      2.518   0.087  28.924   0.000   2.518
##      .nc9      2.267   0.091  25.034   0.000   2.267
##      .nc10     4.056   0.068  59.746   0.000   4.056
##      .nc11     4.205   0.069  60.997   0.000   4.205
##      .nc12     1.754   0.076  23.094   0.000   1.754
##      .nc13     3.287   0.082  39.966   0.000   3.287
##      .nc14     3.749   0.078  47.799   0.000   3.749
##      .nc15     3.677   0.077  47.929   0.000   3.677
##      .nc16     2.292   0.085  26.918   0.000   2.292
##      .nc17     1.903   0.081  23.538   0.000   1.903
##      .nc18     3.328   0.086  38.480   0.000   3.328
##      NC        0.000
## Std.all
##      2.623
##      3.830
##      1.719
##      1.723
##      1.759
##      2.744
##      1.893
##      2.083
##      1.797
##      4.290
##      4.379
##      1.659
##      2.869
##      3.432
##      3.441
##      1.933
##      1.690
##      2.763

```

```

##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      1.045   0.129   8.120   0.000   1.045
##      .nc2      0.457   0.070   6.489   0.000   0.457
##      .nc3      0.771   0.120   6.427   0.000   0.771
##      .nc4      0.796   0.127   6.274   0.000   0.796
##      .nc5      0.653   0.090   7.282   0.000   0.653
##      .nc6      1.223   0.128   9.570   0.000   1.223
##      .nc7      1.285   0.124  10.332   0.000   1.285
##      .nc8      1.095   0.106  10.296   0.000   1.095
##      .nc9      1.153   0.135   8.560   0.000   1.153
##      .nc10     0.619   0.100   6.179   0.000   0.619
##      .nc11     0.529   0.088   6.028   0.000   0.529
##      .nc12     0.735   0.124   5.917   0.000   0.735
##      .nc13     0.978   0.091  10.710   0.000   0.978
##      .nc14     0.899   0.103   8.686   0.000   0.899
##      .nc15     0.984   0.130   7.552   0.000   0.984
##      .nc16     1.231   0.145   8.509   0.000   1.231
##      .nc17     0.895   0.110   8.139   0.000   0.895
##      .nc18     1.400   0.112  12.525   0.000   1.400
##      NC        0.482   0.123   3.914   0.000   1.000
## Std.all
##      0.684
##      0.438
##      0.688
##      0.606
##      0.603
##      0.852
##      0.744
##      0.749
##      0.725
##      0.692
##      0.573
##      0.658
##      0.745
##      0.753
##      0.862
##      0.875
##      0.706
##      0.965
##      1.000
##
## R-Square:
##      Estimate
##      nc1      0.316
##      nc2      0.562
##      nc3      0.312
##      nc4      0.394
##      nc5      0.397
##      nc6      0.148
##      nc7      0.256
##      nc8      0.251

```

```
##      nc9      0.275
##      nc10     0.308
##      nc11     0.427
##      nc12     0.342
##      nc13     0.255
##      nc14     0.247
##      nc15     0.138
##      nc16     0.125
##      nc17     0.294
##      nc18     0.035

MI <- modificationIndices(NC_Fit_8)
subset(MI, mi > 5)

##      lhs op  rhs      mi      epc sepc.lv sepc.all sepc.nox
## 97  nc3  ~~  nc5 6.074  0.121   0.121   0.171   0.171
## 100 nc3  ~~  nc8 5.406 -0.158 -0.158 -0.172 -0.172
## 102 nc3  ~~  nc10 6.193 -0.124 -0.124 -0.180 -0.180
## 103 nc3  ~~  nc11 5.173  0.108  0.108  0.168  0.168
## 104 nc3  ~~  nc12 5.042 -0.124 -0.124 -0.165 -0.165
## 198 nc13 ~~  nc17 6.102  0.166  0.166  0.178  0.178
## 202 nc14 ~~  nc17 8.607  0.189  0.189  0.211  0.211
## 207 nc16 ~~  nc17 6.355  0.196  0.196  0.187  0.187

anova(NC_Fit_7, NC_Fit_8)

## Chi Square Difference Test
##
##      Df      AIC      BIC Chisq Chisq diff Df diff Pr(>Chisq)
## NC_Fit_8 128 10006 10206   182
## NC_Fit_7 129 10013 10210   192      9.52      1      0.002
```

```
NC.model.9 <- "
# Latent variable definitions.
NC =~ nc1 + nc2 + nc3 + nc4 + nc5 + nc6 + nc7 + nc8 + nc9 + nc10 +
nc11 + nc12 + nc13 + nc14 + nc15 + nc16 + nc17 + nc18
nc4 ~~ nc5
nc13 ~~ nc14
nc8 ~~ nc11
nc8 ~~ nc12
nc10 ~~ nc11
nc5 ~~ nc12
nc2 ~~ nc4
nc14 ~~ nc17
"
```

```
NC_Fit_9 <- cfa(NC.model.9, data = NC, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(NC_Fit_9, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 33 iterations
##
##      Number of observations      195
```

```

##      Number of missing patterns              1
##
##      Estimator              ML      Robust
##      Model Fit Test Statistic      172.279      147.771
##      Degrees of freedom              127      127
##      P-value (Chi-square)              0.005      0.100
##      Scaling correction factor              1.166
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##      Minimum Function Test Statistic      1061.451      874.878
##      Degrees of freedom              153      153
##      P-value              0.000      0.000
##
## User model versus baseline model:
##
##      Comparative Fit Index (CFI)              0.950      0.971
##      Tucker-Lewis Index (TLI)              0.940      0.965
##
##      Robust Comparative Fit Index (CFI)              NA
##      Robust Tucker-Lewis Index (TLI)              NA
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)      -4937.009      -4937.009
##      Loglikelihood unrestricted model (H1)      -4850.426      -4850.426
##
##      Number of free parameters              62      62
##      Akaike (AIC)              9998.019      9998.019
##      Bayesian (BIC)              10200.945      10200.945
##      Sample-size adjusted Bayesian (BIC)      10004.538      10004.538
##
## Root Mean Square Error of Approximation:
##
##      RMSEA              0.043      0.029
##      90 Percent Confidence Interval      0.025      0.058      0.000      0.046
##      P-value RMSEA <= 0.05              0.762      0.982
##
##      Robust RMSEA              NA
##      90 Percent Confidence Interval      0.000      NA
##
## Standardized Root Mean Square Residual:
##
##      SRMR              0.050      0.050
##
## Parameter Estimates:
##
##      Information              Observed
##      Observed information based on              Hessian
##      Standard Errors              Robust.huber.white
##
## Latent Variables:
##      Estimate      Std.Err      z-value      P(>|z|)      Std.lv

```



```

##      NC =~
##      nc1          1.000          0.154      7.131      0.000      0.694
##      nc2          1.101      0.154      7.131      0.000      0.764
##      nc3         -0.847      0.148     -5.740      0.000     -0.588
##      nc4         -1.030      0.179     -5.762      0.000     -0.715
##      nc5         -0.936      0.165     -5.690      0.000     -0.650
##      nc6          0.670      0.155      4.318      0.000      0.465
##      nc7         -0.956      0.158     -6.036      0.000     -0.664
##      nc8         -0.883      0.144     -6.123      0.000     -0.613
##      nc9         -0.946      0.175     -5.409      0.000     -0.656
##      nc10          0.755      0.129      5.862      0.000      0.524
##      nc11          0.901      0.144      6.267      0.000      0.625
##      nc12         -0.893      0.155     -5.759      0.000     -0.619
##      nc13          0.834      0.120      6.973      0.000      0.578
##      nc14          0.817      0.129      6.334      0.000      0.567
##      nc15          0.572      0.147      3.888      0.000      0.397
##      nc16         -0.607      0.147     -4.142      0.000     -0.421
##      nc17         -0.897      0.164     -5.458      0.000     -0.622
##      nc18          0.332      0.142      2.339      0.019      0.230
##      Std.all
##
##      0.562
##      0.748
##      -0.555
##      -0.623
##      -0.624
##      0.388
##      -0.505
##      -0.507
##      -0.520
##      0.554
##      0.651
##      -0.586
##      0.505
##      0.523
##      0.371
##      -0.355
##      -0.552
##      0.191
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc4 ~~
##      .nc5          0.221      0.065      3.422      0.001      0.221
##      .nc13 ~~
##      .nc14          0.184      0.084      2.177      0.029      0.184
##      .nc8 ~~
##      .nc11          0.214      0.060      3.556      0.000      0.214
##      .nc12         -0.234      0.061     -3.831      0.000     -0.234
##      .nc10 ~~
##      .nc11          0.143      0.073      1.952      0.051      0.143
##      .nc5 ~~
##      .nc12          0.159      0.065      2.447      0.014      0.159
##      .nc2 ~~

```

```

##      .nc4              0.146    0.058    2.502    0.012    0.146
##      .nc14 ~~
##      .nc17              0.209    0.066    3.173    0.002    0.209
##      Std.all
##
##      0.303
##
##      0.201
##
##      0.282
##      -0.262
##
##      0.249
##
##      0.229
##
##      0.240
##
##      0.241
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      3.241    0.089   36.537   0.000    3.241
##      .nc2      3.913    0.073   53.347   0.000    3.913
##      .nc3      1.821    0.076   23.943   0.000    1.821
##      .nc4      1.974    0.082   24.054   0.000    1.974
##      .nc5      1.831    0.075   24.490   0.000    1.831
##      .nc6      3.287    0.086   38.214   0.000    3.287
##      .nc7      2.487    0.094   26.363   0.000    2.487
##      .nc8      2.518    0.087   28.924   0.000    2.518
##      .nc9      2.267    0.091   25.034   0.000    2.267
##      .nc10     4.056    0.068   59.746   0.000    4.056
##      .nc11     4.205    0.069   60.997   0.000    4.205
##      .nc12     1.754    0.076   23.094   0.000    1.754
##      .nc13     3.287    0.082   39.966   0.000    3.287
##      .nc14     3.749    0.078   47.799   0.000    3.749
##      .nc15     3.677    0.077   47.929   0.000    3.677
##      .nc16     2.292    0.085   26.918   0.000    2.292
##      .nc17     1.903    0.081   23.538   0.000    1.903
##      .nc18     3.328    0.086   38.480   0.000    3.328
##      NC        0.000
##      Std.all
##      2.623
##      3.830
##      1.719
##      1.722
##      1.759
##      2.744
##      1.893
##      2.083
##      1.797
##      4.290
##      4.380
##      1.659

```

```

##      2.869
##      3.461
##      3.441
##      1.933
##      1.690
##      2.763
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      1.045   0.128   8.165   0.000   1.045
##      .nc2      0.460   0.071   6.492   0.000   0.460
##      .nc3      0.776   0.121   6.432   0.000   0.776
##      .nc4      0.804   0.127   6.315   0.000   0.804
##      .nc5      0.661   0.091   7.270   0.000   0.661
##      .nc6      1.219   0.128   9.525   0.000   1.219
##      .nc7      1.286   0.124  10.358   0.000   1.286
##      .nc8      1.086   0.105  10.390   0.000   1.086
##      .nc9      1.160   0.135   8.579   0.000   1.160
##      .nc10     0.620   0.100   6.216   0.000   0.620
##      .nc11     0.531   0.088   6.022   0.000   0.531
##      .nc12     0.734   0.125   5.879   0.000   0.734
##      .nc13     0.978   0.091  10.724   0.000   0.978
##      .nc14     0.852   0.095   8.931   0.000   0.852
##      .nc15     0.984   0.130   7.566   0.000   0.984
##      .nc16     1.229   0.144   8.519   0.000   1.229
##      .nc17     0.881   0.108   8.154   0.000   0.881
##      .nc18     1.398   0.112  12.502   0.000   1.398
##      NC        0.481   0.122   3.955   0.000   1.000
## Std.all
##      0.685
##      0.441
##      0.692
##      0.611
##      0.610
##      0.849
##      0.745
##      0.743
##      0.729
##      0.693
##      0.576
##      0.657
##      0.745
##      0.726
##      0.862
##      0.874
##      0.695
##      0.963
##      1.000
##
## R-Square:
##      Estimate
##      nc1      0.315
##      nc2      0.559

```

```
##      nc3      0.308
##      nc4      0.389
##      nc5      0.390
##      nc6      0.151
##      nc7      0.255
##      nc8      0.257
##      nc9      0.271
##      nc10     0.307
##      nc11     0.424
##      nc12     0.343
##      nc13     0.255
##      nc14     0.274
##      nc15     0.138
##      nc16     0.126
##      nc17     0.305
##      nc18     0.037
```

```
MI <- modificationIndices(NC_Fit_9)
subset(MI, mi > 5)
```

```
##      lhs op  rhs      mi      epc sepc.lv sepc.all sepc.nox
## 98   nc3 ~~  nc5  6.463  0.125   0.125   0.175   0.175
## 101  nc3 ~~  nc8  5.543 -0.159 -0.159  -0.174  -0.174
## 103  nc3 ~~  nc10 6.312 -0.126 -0.126  -0.181  -0.181
## 105  nc3 ~~  nc12 5.097 -0.124 -0.124  -0.165  -0.165
## 198 nc13 ~~  nc16 5.082  0.177  0.177   0.161   0.161
## 199 nc13 ~~  nc17 12.611 0.249  0.249   0.268   0.268
## 207 nc16 ~~  nc17 6.012  0.186  0.186   0.179   0.179
```

```
anova(NC_Fit_8, NC_Fit_9)
```

```
## Chi Square Difference Test
```

```
##
```

```
##      Df    AIC    BIC Chisq Chisq diff Df diff Pr(>Chisq)
## NC_Fit_9 127  9998 10201   172
## NC_Fit_8 128 10006 10206   182      9.84      1    0.0017
```

An alternative would be to specify a model alternative that might be conceptually viable.

```
NC.model.10 <- "
# Latent variable definitions.
NC_N =~ NA*nc1 + nc2 + nc6 + nc10 + nc11 + nc13 + nc14 + nc15 + nc18
NC_P =~ NA*nc3 + nc4 + nc5 + nc7 + nc8 + nc9 + nc12 + nc16 + nc17
NC_N ~~ 1*NC_N
NC_P ~~ 1*NC_P

NC_N ~~ NC_P
"
```

```
NC_Fit_10 <- cfa(NC.model.10, data = NC, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(NC_Fit_10, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 22 iterations
```

```

##
##   Number of observations                195
##   Number of missing patterns           1
##
##   Estimator                           ML      Robust
##   Model Fit Test Statistic             230.668  198.045
##   Degrees of freedom                   134      134
##   P-value (Chi-square)                 0.000    0.000
##   Scaling correction factor             1.165
##   for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##   Minimum Function Test Statistic      1061.451  874.878
##   Degrees of freedom                   153      153
##   P-value                             0.000    0.000
##
## User model versus baseline model:
##
##   Comparative Fit Index (CFI)           0.894    0.911
##   Tucker-Lewis Index (TLI)             0.879    0.899
##
##   Robust Comparative Fit Index (CFI)      NA
##   Robust Tucker-Lewis Index (TLI)        NA
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)          -4966.355  -4966.355
##   Loglikelihood unrestricted model (H1)   -4850.426  -4850.426
##
##   Number of free parameters              55      55
##   Akaike (AIC)                          10042.710  10042.710
##   Bayesian (BIC)                        10222.725  10222.725
##   Sample-size adjusted Bayesian (BIC)    10048.493  10048.493
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                                0.061    0.050
##   90 Percent Confidence Interval         0.047  0.074    0.035  0.063
##   P-value RMSEA <= 0.05                 0.088    0.504
##
##   Robust RMSEA                          NA
##   90 Percent Confidence Interval        NA      NA
##
## Standardized Root Mean Square Residual:
##
##   SRMR                                0.055    0.055
##
## Parameter Estimates:
##
##   Information                          Observed
##   Observed information based on         Hessian
##   Standard Errors                      Robust.huber.white
##

```

```

## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      NC_N =~
##      nc1      0.722   0.088   8.193   0.000   0.722
##      nc2      0.757   0.076  10.002   0.000   0.757
##      nc6      0.493   0.096   5.148   0.000   0.493
##      nc10     0.570   0.087   6.567   0.000   0.570
##      nc11     0.663   0.087   7.643   0.000   0.663
##      nc13     0.658   0.083   7.882   0.000   0.658
##      nc14     0.599   0.100   6.022   0.000   0.599
##      nc15     0.433   0.089   4.880   0.000   0.433
##      nc18     0.224   0.099   2.258   0.024   0.224
##      NC_P =~
##      nc3      0.617   0.075   8.199   0.000   0.617
##      nc4      0.793   0.089   8.948   0.000   0.793
##      nc5      0.760   0.068  11.201   0.000   0.760
##      nc7      0.694   0.082   8.483   0.000   0.694
##      nc8      0.554   0.091   6.051   0.000   0.554
##      nc9      0.679   0.101   6.748   0.000   0.679
##      nc12     0.626   0.094   6.649   0.000   0.626
##      nc16     0.452   0.103   4.400   0.000   0.452
##      nc17     0.609   0.104   5.845   0.000   0.609
##      Std.all
##
##      0.585
##      0.741
##      0.411
##      0.603
##      0.691
##      0.574
##      0.549
##      0.405
##      0.186
##
##      0.583
##      0.694
##      0.730
##      0.528
##      0.457
##      0.539
##      0.592
##      0.381
##      0.541
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      NC_N ~~
##      NC_P      -0.820   0.051 -15.934   0.000  -0.820
##      Std.all
##
##      -0.820
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv

```

##	.nc1	3.241	0.089	36.537	0.000	3.241
##	.nc2	3.913	0.073	53.347	0.000	3.913
##	.nc6	3.287	0.086	38.214	0.000	3.287
##	.nc10	4.056	0.068	59.746	0.000	4.056
##	.nc11	4.205	0.069	60.997	0.000	4.205
##	.nc13	3.287	0.082	39.966	0.000	3.287
##	.nc14	3.749	0.078	47.799	0.000	3.749
##	.nc15	3.677	0.077	47.929	0.000	3.677
##	.nc18	3.328	0.086	38.480	0.000	3.328
##	.nc3	1.821	0.076	23.943	0.000	1.821
##	.nc4	1.974	0.082	24.054	0.000	1.974
##	.nc5	1.831	0.075	24.490	0.000	1.831
##	.nc7	2.487	0.094	26.363	0.000	2.487
##	.nc8	2.518	0.087	28.924	0.000	2.518
##	.nc9	2.267	0.091	25.034	0.000	2.267
##	.nc12	1.754	0.076	23.094	0.000	1.754
##	.nc16	2.292	0.085	26.918	0.000	2.292
##	.nc17	1.903	0.081	23.538	0.000	1.903
##	NC_N	0.000				0.000
##	NC_P	0.000				0.000
##	Std.all					
##	2.623					
##	3.830					
##	2.744					
##	4.290					
##	4.379					
##	2.869					
##	3.432					
##	3.441					
##	2.763					
##	1.719					
##	1.727					
##	1.758					
##	1.893					
##	2.077					
##	1.797					
##	1.658					
##	1.933					
##	1.690					
##	0.000					
##	0.000					
##						
##	Variances:					
##		Estimate	Std.Err	z-value	P(> z)	Std.lv
##	NC_N	1.000				1.000
##	NC_P	1.000				1.000
##	.nc1	1.005	0.133	7.569	0.000	1.005
##	.nc2	0.471	0.076	6.211	0.000	0.471
##	.nc6	1.193	0.130	9.141	0.000	1.193
##	.nc10	0.569	0.093	6.104	0.000	0.569
##	.nc11	0.482	0.081	5.969	0.000	0.482
##	.nc13	0.879	0.090	9.768	0.000	0.879
##	.nc14	0.834	0.103	8.114	0.000	0.834
##	.nc15	0.954	0.132	7.209	0.000	0.954

```
##      .nc18      1.401    0.112   12.546    0.000    1.401
##      .nc3      0.741    0.118    6.256    0.000    0.741
##      .nc4      0.678    0.123    5.506    0.000    0.678
##      .nc5      0.507    0.086    5.873    0.000    0.507
##      .nc7      1.245    0.130    9.566    0.000    1.245
##      .nc8      1.164    0.113   10.296    0.000    1.164
##      .nc9      1.129    0.139    8.150    0.000    1.129
##      .nc12     0.727    0.125    5.839    0.000    0.727
##      .nc16     1.203    0.146    8.233    0.000    1.203
##      .nc17     0.896    0.114    7.843    0.000    0.896
## Std.all
##      1.000
##      1.000
##      0.658
##      0.451
##      0.831
##      0.637
##      0.523
##      0.670
##      0.699
##      0.836
##      0.965
##      0.661
##      0.518
##      0.467
##      0.721
##      0.791
##      0.710
##      0.650
##      0.855
##      0.707
##
## R-Square:
##      Estimate
##      nc1      0.342
##      nc2      0.549
##      nc6      0.169
##      nc10     0.363
##      nc11     0.477
##      nc13     0.330
##      nc14     0.301
##      nc15     0.164
##      nc18     0.035
##      nc3      0.339
##      nc4      0.482
##      nc5      0.533
##      nc7      0.279
##      nc8      0.209
##      nc9      0.290
##      nc12     0.350
##      nc16     0.145
##      nc17     0.293
```

```
MI <- modificationIndices(NC_Fit_10)
```



```
subset(MI, mi > 5)

##      lhs op  rhs      mi      epc sepc.lv sepc.all sepc.nox
## 61  NC_N =~ nc4  5.233  0.403   0.403   0.353   0.353
## 66  NC_N =~ nc12 6.590 -0.435 -0.435  -0.411  -0.411
## 126 nc10 ~~ nc11 8.545  0.130   0.130   0.249   0.249
## 130 nc10 ~~ nc18 6.271 -0.169 -0.169  -0.190  -0.190
## 148 nc11 ~~ nc8  10.047 0.187   0.187   0.250   0.250
## 153 nc13 ~~ nc14 9.301  0.206   0.206   0.240   0.240
## 164 nc13 ~~ nc17 7.484  0.187   0.187   0.210   0.210
## 175 nc14 ~~ nc17 8.947  0.198   0.198   0.229   0.229
## 203 nc4  ~~ nc5  5.381  0.127   0.127   0.216   0.216
## 213 nc5  ~~ nc12 6.709  0.136   0.136   0.225   0.225
## 222 nc8  ~~ nc12 12.317 -0.250 -0.250  -0.272  -0.272
## 224 nc8  ~~ nc17 5.053  0.175   0.175   0.172   0.172
## 230 nc16 ~~ nc17 5.384  0.182   0.182   0.175   0.175

anova(NC_Fit_9, NC_Fit_10)

## Chi Square Difference Test
##
##           Df    AIC    BIC Chisq Chisq diff Df diff Pr(>Chisq)
## NC_Fit_9  127  9998 10201   172
## NC_Fit_10 134 10043 10223   231      58.4      7  3.2e-10

anova(NC_Fit_1, NC_Fit_9)

## Chi Square Difference Test
##
##           Df    AIC    BIC Chisq Chisq diff Df diff Pr(>Chisq)
## NC_Fit_9  127  9998 10201   172
## NC_Fit_1 135 10078 10254   267      95      8  <2e-16

anova(NC_Fit_1, NC_Fit_10)

## Chi Square Difference Test
##
##           Df    AIC    BIC Chisq Chisq diff Df diff Pr(>Chisq)
## NC_Fit_10 134 10043 10223   231
## NC_Fit_1  135 10078 10254   267      36.6      1  1.4e-09
```

6 Reliability

The reliability of a scale can be obtained from CFA output. It is first important to score the items in a consistent direction.

```
NC_rescaled <- as.data.frame(NC)
NC_rescaled$nc1 <- 6 - NC_rescaled$nc1
NC_rescaled$nc2 <- 6 - NC_rescaled$nc2
NC_rescaled$nc6 <- 6 - NC_rescaled$nc6
NC_rescaled$nc10 <- 6 - NC_rescaled$nc10
NC_rescaled$nc11 <- 6 - NC_rescaled$nc11
NC_rescaled$nc13 <- 6 - NC_rescaled$nc13
```

```

NC_rescaled$nc14 <- 6 - NC_rescaled$nc14
NC_rescaled$nc15 <- 6 - NC_rescaled$nc15
NC_rescaled$nc18 <- 6 - NC_rescaled$nc18
cor(NC_rescaled)

```

```

##          nc1      nc2      nc3      nc4      nc5      nc6      nc7      nc8
## nc1  1.0000  0.4554  0.2374  0.27155  0.30314  0.2442  0.28711  0.2374
## nc2  0.4554  1.0000  0.4647  0.35317  0.49596  0.2551  0.34272  0.3568
## nc3  0.2374  0.4647  1.0000  0.37315  0.44680  0.1897  0.32077  0.2241
## nc4  0.2715  0.3532  0.3732  1.00000  0.56931  0.1668  0.34967  0.3055
## nc5  0.3031  0.4960  0.4468  0.56931  1.00000  0.2241  0.38633  0.3091
## nc6  0.2442  0.2551  0.1897  0.16684  0.22413  1.0000  0.24521  0.1518
## nc7  0.2871  0.3427  0.3208  0.34967  0.38633  0.2452  1.00000  0.3373
## nc8  0.2374  0.3568  0.2241  0.30554  0.30908  0.1518  0.33729  1.0000
## nc9  0.3210  0.3203  0.3852  0.41022  0.31555  0.1661  0.21868  0.3088
## nc10 0.2868  0.3820  0.4046  0.34493  0.35488  0.2844  0.26560  0.2938
## nc11 0.3603  0.5149  0.3218  0.40162  0.38587  0.2920  0.26209  0.1397
## nc12 0.3627  0.4280  0.2764  0.38068  0.51625  0.2518  0.34822  0.1034
## nc13 0.3931  0.3895  0.2407  0.29196  0.31609  0.2089  0.18151  0.2363
## nc14 0.3298  0.3894  0.2695  0.25564  0.25380  0.2785  0.26126  0.3121
## nc15 0.2765  0.3312  0.2461  0.15791  0.15054  0.1686  0.06686  0.1796
## nc16 0.2825  0.1863  0.1602  0.31941  0.25597  0.1493  0.20804  0.1408
## nc17 0.2707  0.4176  0.2821  0.40446  0.32716  0.2188  0.25394  0.3601
## nc18 0.1157  0.1899  0.1427  0.08698  0.09472  0.1692  0.14960  0.0251
##          nc9      nc10      nc11      nc12      nc13      nc14      nc15
## nc1  0.32100  0.28683  0.3603  0.3627  0.39312  0.32985  0.27651
## nc2  0.32029  0.38198  0.5149  0.4280  0.38946  0.38935  0.33123
## nc3  0.38524  0.40465  0.3218  0.2764  0.24071  0.26949  0.24610
## nc4  0.41022  0.34493  0.4016  0.3807  0.29196  0.25564  0.15791
## nc5  0.31555  0.35488  0.3859  0.5163  0.31609  0.25380  0.15054
## nc6  0.16608  0.28443  0.2920  0.2518  0.20892  0.27848  0.16861
## nc7  0.21868  0.26560  0.2621  0.3482  0.18151  0.26126  0.06686
## nc8  0.30876  0.29383  0.1397  0.1034  0.23630  0.31214  0.17958
## nc9  1.00000  0.37382  0.3331  0.2491  0.24468  0.14865  0.17582
## nc10 0.37382  1.00000  0.5181  0.2783  0.31641  0.35130  0.24134
## nc11 0.33314  0.51813  1.0000  0.3946  0.41264  0.31315  0.23952
## nc12 0.24911  0.27834  0.3946  1.0000  0.33947  0.36420  0.20194
## nc13 0.24468  0.31641  0.4126  0.3395  1.00000  0.45107  0.29364
## nc14 0.14865  0.35130  0.3132  0.3642  0.45107  1.00000  0.22920
## nc15 0.17582  0.24134  0.2395  0.2019  0.29364  0.22920  1.00000
## nc16 0.20501  0.13815  0.2373  0.2045  0.05046  0.13329  0.06710
## nc17 0.36866  0.22605  0.3563  0.2770  0.10554  0.08246  0.16685
## nc18 0.01035 -0.02075  0.1058  0.1096  0.02089  0.18348  0.09831
##          nc16      nc17      nc18
## nc1  0.28253  0.27069  0.11568
## nc2  0.18634  0.41760  0.18992
## nc3  0.16016  0.28211  0.14274
## nc4  0.31941  0.40446  0.08698
## nc5  0.25597  0.32716  0.09472
## nc6  0.14929  0.21878  0.16920
## nc7  0.20804  0.25394  0.14960
## nc8  0.14077  0.36005  0.02510
## nc9  0.20501  0.36866  0.01035
## nc10 0.13815  0.22605 -0.02075

```

```
## nc11 0.23725 0.35628 0.10583
## nc12 0.20449 0.27700 0.10965
## nc13 0.05046 0.10554 0.02089
## nc14 0.13329 0.08246 0.18348
## nc15 0.06710 0.16685 0.09831
## nc16 1.00000 0.32856 0.19634
## nc17 0.32856 1.00000 0.09364
## nc18 0.19634 0.09364 1.00000
```

6.1 Single Factor Model

A single factor model is estimated in order to get the raw information for calculating reliability.

```
NC.model.11 <- "
# Latent variable definitions.
NC =~ nc1 + nc2 + nc3 + nc4 + nc5 + nc6 + nc7 + nc8 + nc9 + nc10 +
nc11 + nc12 + nc13 + nc14 + nc15 + nc16 + nc17 + nc18
"
```

```
NC_Fit_11 <- cfa(NC.model.11, data = NC_rescaled, missing = "ML",
  estimator = "MLR", likelihood = "wishart", representation = "LISREL")
summary(NC_Fit_11, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)
```

```
## lavaan (0.6-1) converged normally after 29 iterations
##
##   Number of observations              195
##   Number of missing patterns          1
##
##   Estimator                        ML      Robust
##   Model Fit Test Statistic          267.315    228.948
##   Degrees of freedom                 135       135
##   P-value (Chi-square)               0.000     0.000
##   Scaling correction factor          1.168
##   for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##   Minimum Function Test Statistic    1061.451    874.878
##   Degrees of freedom                 153       153
##   P-value                           0.000     0.000
##
## User model versus baseline model:
##
##   Comparative Fit Index (CFI)         0.854     0.870
##   Tucker-Lewis Index (TLI)           0.835     0.853
##
##   Robust Comparative Fit Index (CFI)          NA
##   Robust Tucker-Lewis Index (TLI)            NA
##
## Loglikelihood and Information Criteria:
##
```

```

##      Loglikelihood user model (H0)                -4984.773   -4984.773
##      Loglikelihood unrestricted model (H1)         -4850.426   -4850.426
##
##      Number of free parameters                    54          54
##      Akaike (AIC)                                10077.545   10077.545
##      Bayesian (BIC)                               10254.287   10254.287
##      Sample-size adjusted Bayesian (BIC)          10083.223   10083.223
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                                         0.071          0.060
##      90 Percent Confidence Interval               0.059  0.084          0.047  0.072
##      P-value RMSEA <= 0.05                       0.004          0.095
##
##      Robust RMSEA                                  NA
##      90 Percent Confidence Interval               NA      NA
##
## Standardized Root Mean Square Residual:
##
##      SRMR                                         0.060          0.060
##
## Parameter Estimates:
##
##      Information                                Observed
##      Observed information based on                Hessian
##      Standard Errors                            Robust.huber.white
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##      NC =~
##      nc1       1.000
##      nc2       1.072    0.155    6.922    0.000    0.741
##      nc3       0.884    0.152    5.818    0.000    0.611
##      nc4       1.038    0.189    5.496    0.000    0.717
##      nc5       1.018    0.184    5.543    0.000    0.704
##      nc6       0.682    0.155    4.395    0.000    0.472
##      nc7       0.957    0.164    5.826    0.000    0.662
##      nc8       0.787    0.143    5.519    0.000    0.544
##      nc9       0.951    0.182    5.210    0.000    0.657
##      nc10      0.801    0.134    5.980    0.000    0.554
##      nc11      0.906    0.149    6.095    0.000    0.626
##      nc12      0.910    0.162    5.602    0.000    0.629
##      nc13      0.865    0.122    7.095    0.000    0.598
##      nc14      0.800    0.129    6.218    0.000    0.553
##      nc15      0.567    0.148    3.819    0.000    0.392
##      nc16      0.604    0.153    3.956    0.000    0.418
##      nc17      0.836    0.170    4.912    0.000    0.578
##      nc18      0.319    0.140    2.277    0.023    0.221
##      Std.all
##
##      0.560
##      0.725
##      0.577

```

```

##      0.627
##      0.676
##      0.394
##      0.504
##      0.448
##      0.521
##      0.586
##      0.652
##      0.594
##      0.522
##      0.506
##      0.367
##      0.352
##      0.513
##      0.183
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      2.759   0.089  31.103   0.000   2.759
##      .nc2      2.087   0.073  28.456   0.000   2.087
##      .nc3      1.821   0.076  23.943   0.000   1.821
##      .nc4      1.974   0.082  24.054   0.000   1.974
##      .nc5      1.831   0.075  24.490   0.000   1.831
##      .nc6      2.713   0.086  31.537   0.000   2.713
##      .nc7      2.487   0.094  26.363   0.000   2.487
##      .nc8      2.518   0.087  28.924   0.000   2.518
##      .nc9      2.267   0.091  25.034   0.000   2.267
##      .nc10     1.944   0.068  28.627   0.000   1.944
##      .nc11     1.795   0.069  26.035   0.000   1.795
##      .nc12     1.754   0.076  23.094   0.000   1.754
##      .nc13     2.713   0.082  32.983   0.000   2.713
##      .nc14     2.251   0.078  28.705   0.000   2.251
##      .nc15     2.323   0.077  30.281   0.000   2.323
##      .nc16     2.292   0.085  26.918   0.000   2.292
##      .nc17     1.903   0.081  23.538   0.000   1.903
##      .nc18     2.672   0.086  30.891   0.000   2.672
##      NC        0.000
##      Std.all
##      2.233
##      2.043
##      1.719
##      1.727
##      1.758
##      2.264
##      1.893
##      2.077
##      1.797
##      2.055
##      1.869
##      1.658
##      2.368
##      2.061
##      2.174
##      1.933

```

```

##      1.690
##      2.218
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .nc1      1.049   0.131   7.997   0.000   1.049
##      .nc2      0.495   0.077   6.450   0.000   0.495
##      .nc3      0.748   0.116   6.440   0.000   0.748
##      .nc4      0.793   0.129   6.153   0.000   0.793
##      .nc5      0.588   0.085   6.945   0.000   0.588
##      .nc6      1.213   0.127   9.541   0.000   1.213
##      .nc7      1.289   0.129   9.990   0.000   1.289
##      .nc8      1.175   0.111  10.615   0.000   1.175
##      .nc9      1.159   0.136   8.526   0.000   1.159
##      .nc10     0.588   0.093   6.305   0.000   0.588
##      .nc11     0.530   0.084   6.311   0.000   0.530
##      .nc12     0.724   0.125   5.766   0.000   0.724
##      .nc13     0.955   0.092  10.336   0.000   0.955
##      .nc14     0.887   0.103   8.645   0.000   0.887
##      .nc15     0.988   0.132   7.515   0.000   0.988
##      .nc16     1.232   0.145   8.477   0.000   1.232
##      .nc17     0.934   0.113   8.255   0.000   0.934
##      .nc18     1.402   0.112  12.569   0.000   1.402
##      NC        0.478   0.125   3.823   0.000   1.000
## Std.all
##      0.687
##      0.474
##      0.667
##      0.606
##      0.543
##      0.845
##      0.746
##      0.799
##      0.728
##      0.657
##      0.574
##      0.647
##      0.728
##      0.744
##      0.866
##      0.876
##      0.737
##      0.966
##      1.000
##
## R-Square:
##      Estimate
##      nc1      0.313
##      nc2      0.526
##      nc3      0.333
##      nc4      0.394
##      nc5      0.457
##      nc6      0.155

```

```
##      nc7      0.254
##      nc8      0.201
##      nc9      0.272
##      nc10     0.343
##      nc11     0.426
##      nc12     0.353
##      nc13     0.272
##      nc14     0.256
##      nc15     0.134
##      nc16     0.124
##      nc17     0.263
##      nc18     0.034
```

6.2 Coefficient Alpha: Traditional

```
psych::alpha(NC_rescaled)

##
## Reliability analysis
## Call: psych::alpha(x = NC_rescaled)
##
##      raw_alpha std.alpha G6(smc) average_r S/N   ase mean   sd
##      0.86      0.87      0.88      0.27 6.6 0.014  2.2 0.62
## median_r
##      0.27
##
## lower alpha upper      95% confidence boundaries
## 0.84 0.86 0.89
##
## Reliability if an item is dropped:
##      raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r
## nc1      0.85      0.86      0.88      0.26 6.1   0.015 0.014
## nc2      0.85      0.85      0.87      0.25 5.8   0.016 0.012
## nc3      0.85      0.86      0.88      0.26 6.1   0.015 0.013
## nc4      0.85      0.86      0.87      0.26 6.0   0.015 0.012
## nc5      0.85      0.86      0.87      0.26 5.9   0.015 0.012
## nc6      0.86      0.87      0.88      0.27 6.4   0.014 0.014
## nc7      0.86      0.86      0.88      0.27 6.2   0.015 0.014
## nc8      0.86      0.86      0.88      0.27 6.3   0.015 0.013
## nc9      0.86      0.86      0.88      0.27 6.2   0.015 0.013
## nc10     0.86      0.86      0.87      0.26 6.1   0.015 0.013
## nc11     0.85      0.86      0.87      0.26 6.0   0.015 0.012
## nc12     0.85      0.86      0.87      0.26 6.1   0.015 0.013
## nc13     0.86      0.86      0.88      0.27 6.2   0.015 0.013
## nc14     0.86      0.86      0.88      0.27 6.2   0.015 0.013
## nc15     0.86      0.87      0.88      0.28 6.5   0.014 0.013
## nc16     0.86      0.87      0.88      0.28 6.5   0.014 0.013
## nc17     0.86      0.86      0.88      0.27 6.2   0.015 0.013
## nc18     0.87      0.87      0.89      0.29 6.9   0.014 0.010
##      med.r
## nc1      0.26
## nc2      0.26
```

```

## nc3 0.27
## nc4 0.26
## nc5 0.26
## nc6 0.28
## nc7 0.28
## nc8 0.27
## nc9 0.27
## nc10 0.26
## nc11 0.26
## nc12 0.27
## nc13 0.27
## nc14 0.27
## nc15 0.28
## nc16 0.28
## nc17 0.27
## nc18 0.29
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean  sd
## nc1 195 0.61 0.60 0.57 0.53 2.8 1.24
## nc2 195 0.72 0.73 0.72 0.67 2.1 1.02
## nc3 195 0.59 0.60 0.57 0.53 1.8 1.06
## nc4 195 0.65 0.64 0.63 0.58 2.0 1.15
## nc5 195 0.67 0.67 0.66 0.61 1.8 1.04
## nc6 195 0.47 0.47 0.41 0.38 2.7 1.20
## nc7 195 0.56 0.55 0.51 0.47 2.5 1.32
## nc8 195 0.51 0.50 0.47 0.43 2.5 1.22
## nc9 195 0.56 0.56 0.52 0.48 2.3 1.26
## nc10 195 0.59 0.60 0.58 0.53 1.9 0.95
## nc11 195 0.64 0.66 0.65 0.59 1.8 0.96
## nc12 195 0.60 0.61 0.59 0.54 1.8 1.06
## nc13 195 0.55 0.55 0.52 0.47 2.7 1.15
## nc14 195 0.56 0.56 0.53 0.48 2.3 1.10
## nc15 195 0.42 0.43 0.37 0.34 2.3 1.07
## nc16 195 0.44 0.43 0.37 0.34 2.3 1.19
## nc17 195 0.56 0.55 0.52 0.48 1.9 1.13
## nc18 195 0.29 0.28 0.20 0.18 2.7 1.21
##
## Non missing response frequency for each item
##      1 2 3 4 5 miss
## nc1 0.12 0.42 0.18 0.14 0.14 0
## nc2 0.31 0.44 0.14 0.08 0.03 0
## nc3 0.50 0.31 0.09 0.07 0.03 0
## nc4 0.44 0.32 0.10 0.09 0.05 0
## nc5 0.49 0.33 0.07 0.09 0.02 0
## nc6 0.14 0.36 0.24 0.15 0.11 0
## nc7 0.30 0.26 0.16 0.19 0.08 0
## nc8 0.21 0.39 0.14 0.18 0.07 0
## nc9 0.32 0.40 0.06 0.15 0.08 0
## nc10 0.35 0.46 0.11 0.05 0.03 0
## nc11 0.45 0.41 0.07 0.04 0.03 0
## nc12 0.56 0.25 0.10 0.06 0.03 0
## nc13 0.14 0.34 0.28 0.15 0.09 0
## nc14 0.26 0.44 0.14 0.12 0.04 0

```



```
## nc15 0.22 0.42 0.23 0.07 0.06 0
## nc16 0.28 0.40 0.12 0.13 0.06 0
## nc17 0.49 0.27 0.11 0.09 0.04 0
## nc18 0.16 0.37 0.18 0.21 0.08 0
```

6.3 Coefficient Alpha: CFA

```
# Extract standardized loadings and error variances.
NC_Loadings <- inspect(NC_Fit_11, what = "std")$lambda
NC_error_variances <- diag(inspect(NC_Fit_11, what = "std")$theta)

# Standardized coefficient alpha, version 1.
Mean_Item_Reliability <- mean(NC_Loadings^2)
(18 * Mean_Item_Reliability)/(1 + 17 * Mean_Item_Reliability)

## [1] 0.8771

# Standardized coefficient alpha, version 2.
sum(NC_Loadings^2)/(sum(NC_Loadings^2) + sum(NC_error_variances))

## [1] 0.8705
```

7 Cross-Validation

7.1 Across Samples

The most convincing cross-validation occurs across independent samples. For this demonstration, the state self-esteem data will be used. Cross-validation in this case is usually described in the context of measurement invariance. Several increasingly stringent forms of invariance are possible, estimated from a set of models that place increasing equality constraints across groups.

- 1. The first model simply requires estimation of the same basic factor model in each group. Configural invariance is said to exist if this first model has a good fit and the same loadings are significant in all groups.*
- 2. The next constraint is to require the factor loadings to be equal across groups. This constraint is called metric or weak invariance and tests whether respondents across groups attribute the same meaning to the latent constructs.*
- 3. Next the intercepts are constrained to be equal. This is called scalar or strong invariance and implies that the meaning of the construct (the factor loadings), and the levels of the underlying items (intercepts) are equal in both groups. If this constraint holds, the groups can be compared on their latent variable scores.*
- 4. Finally, the residual variances are fixed to be equal across groups. This is called strict invariance and means that the explained variance for every item is the same across groups. In other words, the latent construct is measured identically across groups. If error variances are not equal, groups can still be compared on the latent variable, but this is measured with different amounts of error between groups.*

It is also possible to extend the sequence of tests by requiring the latent means, variances, and covariances to be equal.

The sequence of tests compares each newly constrained model to the previous one. The level of invariance is determined by the furthest step at which model equivalence holds.

It is possible that partial invariance holds, occurring when a level of invariance is found for some, but not all items.

Comparison of models is usually conducted using both chi-square difference tests as well as comparison of CFI. A nonsignificant chi-square difference and a CFI difference less than .01 are conventionally taken as evidence for invariance at a particular step.

```
HS.model <- "
Visual =~ x1+x2+x3
Textual =~ x4+x5+x6
Speed =~ x7+x8+x9
"
no_groups <- cfa(HS.model, data = HS, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(no_groups, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 36 iterations
##
##      Number of observations              301
##      Number of missing patterns          1
##
##      Estimator              ML          Robust
##      Model Fit Test Statistic      85.022      86.842
##      Degrees of freedom              24          24
##      P-value (Chi-square)           0.000          0.000
##      Scaling correction factor              0.979
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##      Minimum Function Test Statistic      915.799      877.158
##      Degrees of freedom              36          36
##      P-value              0.000          0.000
##
## User model versus baseline model:
##
##      Comparative Fit Index (CFI)           0.931          0.925
##      Tucker-Lewis Index (TLI)            0.896          0.888
##
##      Robust Comparative Fit Index (CFI)              NA
##      Robust Tucker-Lewis Index (TLI)              NA
##
## Loglikelihood and Information Criteria:
##
```

```

##      Loglikelihood user model (H0)                -3737.745   -3737.745
##      Loglikelihood unrestricted model (H1)         -3695.092   -3695.092
##
##      Number of free parameters                    30          30
##      Akaike (AIC)                                7535.490      7535.490
##      Bayesian (BIC)                               7646.703      7646.703
##      Sample-size adjusted Bayesian (BIC)          7551.560      7551.560
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                                         0.092          0.093
##      90 Percent Confidence Interval               0.071  0.114      0.073  0.115
##      P-value RMSEA <= 0.05                       0.001          0.001
##
##      Robust RMSEA                                  NA
##      90 Percent Confidence Interval               NA      NA
##
## Standardized Root Mean Square Residual:
##
##      SRMR                                         0.060          0.060
##
## Parameter Estimates:
##
##      Information                                Observed
##      Observed information based on              Hessian
##      Standard Errors                          Robust.huber.white
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##      Visual =~
##      x1         1.000
##      x2         0.554    0.132    4.184    0.000    0.498
##      x3         0.729    0.141    5.161    0.000    0.656
##      Textual =~
##      x4         1.000
##      x5         1.113    0.066   16.917    0.000    1.102
##      x6         0.926    0.061   15.064    0.000    0.917
##      Speed =~
##      x7         1.000
##      x8         1.180    0.131    9.031    0.000    0.731
##      x9         1.082    0.267    4.053    0.000    0.670
##      Std.all
##
##      0.772
##      0.424
##      0.581
##
##      0.852
##      0.855
##      0.838
##
##      0.570
##      0.723

```

```

##      0.665
##
## Covariances:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv
##      Visual ~~
##      Textual      0.408     0.099     4.104     0.000     0.459
##      Speed        0.262     0.060     4.359     0.000     0.471
##      Textual ~~
##      Speed        0.173     0.056     3.076     0.002     0.283
##      Std.all
##
##      0.459
##      0.471
##
##      0.283
##
## Intercepts:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv
##      .x1         4.936     0.067    73.351     0.000     4.936
##      .x2         6.088     0.068    89.705     0.000     6.088
##      .x3         2.250     0.065    34.522     0.000     2.250
##      .x4         3.061     0.067    45.618     0.000     3.061
##      .x5         4.341     0.074    58.355     0.000     4.341
##      .x6         2.186     0.063    34.609     0.000     2.186
##      .x7         4.186     0.063    66.655     0.000     4.186
##      .x8         5.527     0.058    94.697     0.000     5.527
##      .x9         5.374     0.058    92.392     0.000     5.374
##      Visual      0.000
##      Textual     0.000
##      Speed       0.000
##      Std.all
##      4.235
##      5.179
##      1.993
##      2.634
##      3.369
##      1.998
##      3.848
##      5.467
##      5.334
##      0.000
##      0.000
##      0.000
##
## Variances:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv
##      .x1         0.549     0.157     3.503     0.000     0.549
##      .x2         1.134     0.112    10.118     0.000     1.134
##      .x3         0.844     0.100     8.405     0.000     0.844
##      .x4         0.371     0.050     7.370     0.000     0.371
##      .x5         0.446     0.057     7.857     0.000     0.446
##      .x6         0.356     0.047     7.645     0.000     0.356
##      .x7         0.799     0.097     8.209     0.000     0.799
##      .x8         0.488     0.120     4.073     0.000     0.488

```

```

##      .x9      0.566    0.119    4.760    0.000    0.566
##      Visual    0.809    0.181    4.479    0.000    1.000
##      Textual    0.979    0.121    8.062    0.000    1.000
##      Speed     0.384    0.107    3.590    0.000    1.000
## Std.all
##      0.404
##      0.821
##      0.662
##      0.275
##      0.269
##      0.298
##      0.676
##      0.477
##      0.558
##      1.000
##      1.000
##      1.000
##
## R-Square:
##      Estimate
##      x1      0.596
##      x2      0.179
##      x3      0.338
##      x4      0.725
##      x5      0.731
##      x6      0.702
##      x7      0.324
##      x8      0.523
##      x9      0.442

config <- cfa(HS.model, data = HS, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL", group = "school")
summary(config, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 66 iterations
##
##      Number of observations per group
##      Pasteur      156
##      Grant-White  145
##      Number of missing patterns per group
##      Pasteur      1
##      Grant-White  1
##
##      Estimator      ML      Robust
##      Model Fit Test Statistic      115.084      120.934
##      Degrees of freedom      48      48
##      P-value (Chi-square)      0.000      0.000
##      Scaling correction factor      0.952
##      for the Yuan-Bentler correction (Mplus variant)
##
## Chi-square for each group:
##
##      Pasteur      63.897      67.145
##      Grant-White  51.187      53.789

```

```

##
## Model test baseline model:
##
##   Minimum Function Test Statistic          951.384      928.080
##   Degrees of freedom                      72          72
##   P-value                                0.000        0.000
##
## User model versus baseline model:
##
##   Comparative Fit Index (CFI)              0.924        0.915
##   Tucker-Lewis Index (TLI)                0.886        0.872
##
##   Robust Comparative Fit Index (CFI)              NA
##   Robust Tucker-Lewis Index (TLI)              NA
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)             -3682.198    -3682.198
##   Loglikelihood unrestricted model (H1)       -3624.272    -3624.272
##
##   Number of free parameters                  60          60
##   Akaike (AIC)                             7484.395    7484.395
##   Bayesian (BIC)                           7706.822    7706.822
##   Sample-size adjusted Bayesian (BIC)        7516.536    7516.536
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                                0.097        0.101
##   90 Percent Confidence Interval          0.074  0.119      0.078  0.124
##   P-value RMSEA <= 0.05                0.001        0.000
##
##   Robust RMSEA                                NA
##   90 Percent Confidence Interval          NA      NA
##
## Standardized Root Mean Square Residual:
##
##   SRMR                                0.068        0.068
##
## Parameter Estimates:
##
##   Information                                Observed
##   Observed information based on              Hessian
##   Standard Errors                          Robust.huber.white
##
##
## Group 1 [Pasteur]:
##
## Latent Variables:
##
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##   Visual =~
##     x1           1.000
##     x2           0.394    0.198    1.990    0.047    0.412
##     x3           0.570    0.202    2.828    0.005    0.597
##   Textual =~

```

```

##      x4      1.000
##      x5      1.183      0.106      11.176      0.000      1.119
##      x6      0.875      0.094      9.357      0.000      0.827
## Speed =~
##      x7      1.000
##      x8      1.125      0.263      4.283      0.000      0.665
##      x9      0.922      0.298      3.089      0.002      0.545
## Std.all
##
##      0.887
##      0.336
##      0.515
##
##      0.823
##      0.856
##      0.838
##
##      0.547
##      0.682
##      0.551
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
## Visual ~~
##      Textual      0.479      0.144      3.323      0.001      0.484
##      Speed      0.185      0.076      2.441      0.015      0.299
## Textual ~~
##      Speed      0.182      0.076      2.394      0.017      0.325
## Std.all
##
##      0.484
##      0.299
##
##      0.325
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      4.941      0.095      52.075      0.000      4.941
##      .x2      5.984      0.099      60.746      0.000      5.984
##      .x3      2.487      0.093      26.689      0.000      2.487
##      .x4      2.823      0.092      30.587      0.000      2.823
##      .x5      3.995      0.105      38.056      0.000      3.995
##      .x6      1.922      0.079      24.240      0.000      1.922
##      .x7      4.432      0.087      51.011      0.000      4.432
##      .x8      5.563      0.078      70.977      0.000      5.563
##      .x9      5.418      0.079      68.212      0.000      5.418
##      Visual      0.000
##      Textual      0.000
##      Speed      0.000
## Std.all
##      4.183
##      4.880
##      2.144
##      2.457

```

```

##      3.057
##      1.947
##      4.098
##      5.702
##      5.480
##      0.000
##      0.000
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      0.298   0.339   0.879   0.379   0.298
##      .x2      1.334   0.180   7.416   0.000   1.334
##      .x3      0.989   0.154   6.422   0.000   0.989
##      .x4      0.425   0.073   5.848   0.000   0.425
##      .x5      0.456   0.086   5.286   0.000   0.456
##      .x6      0.290   0.055   5.250   0.000   0.290
##      .x7      0.820   0.133   6.190   0.000   0.820
##      .x8      0.510   0.101   5.031   0.000   0.510
##      .x9      0.680   0.126   5.388   0.000   0.680
##      Visual    1.097   0.380   2.889   0.004   1.000
##      Textual   0.894   0.161   5.558   0.000   1.000
##      Speed     0.350   0.140   2.494   0.013   1.000
## Std.all
##      0.214
##      0.887
##      0.735
##      0.322
##      0.267
##      0.297
##      0.701
##      0.535
##      0.696
##      1.000
##      1.000
##      1.000
##
## R-Square:
##      Estimate
##      x1      0.786
##      x2      0.113
##      x3      0.265
##      x4      0.678
##      x5      0.733
##      x6      0.703
##      x7      0.299
##      x8      0.465
##      x9      0.304
##
##
## Group 2 [Grant-White]:
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv

```



```

## Visual =~
##      x1          1.000          0.777
##      x2          0.736      0.191      3.857      0.000      0.572
##      x3          0.925      0.217      4.267      0.000      0.719
## Textual =~
##      x4          1.000          0.971
##      x5          0.990      0.089     11.142      0.000      0.961
##      x6          0.963      0.091     10.552      0.000      0.935
## Speed =~
##      x7          1.000          0.679
##      x8          1.226      0.163      7.535      0.000      0.833
##      x9          1.058      0.251      4.213      0.000      0.719
## Std.all
##
##      0.677
##      0.517
##      0.694
##
##      0.866
##      0.829
##      0.826
##
##      0.659
##      0.796
##      0.701
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
## Visual ~~
##      Textual      0.408      0.116      3.523      0.000      0.541
##      Speed        0.276      0.101      2.722      0.006      0.523
## Textual ~~
##      Speed        0.222      0.105      2.108      0.035      0.336
## Std.all
##
##      0.541
##      0.523
##
##      0.336
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1        4.930      0.096     51.524      0.000      4.930
##      .x2        6.200      0.092     67.192      0.000      6.200
##      .x3        1.996      0.086     23.118      0.000      1.996
##      .x4        3.317      0.093     35.507      0.000      3.317
##      .x5        4.712      0.097     48.823      0.000      4.712
##      .x6        2.469      0.094     26.189      0.000      2.469
##      .x7        3.921      0.086     45.666      0.000      3.921
##      .x8        5.488      0.087     62.963      0.000      5.488
##      .x9        5.327      0.085     62.363      0.000      5.327
##      Visual      0.000          0.000
##      Textual      0.000          0.000
##      Speed        0.000          0.000

```

```

## Std.all
## 4.293
## 5.599
## 1.926
## 2.959
## 4.068
## 2.182
## 3.805
## 5.246
## 5.196
## 0.000
## 0.000
## 0.000
##
## Variances:
## Estimate Std.Err z-value P(>|z|) Std.lv
## .x1 0.715 0.182 3.925 0.000 0.715
## .x2 0.899 0.142 6.317 0.000 0.899
## .x3 0.557 0.122 4.552 0.000 0.557
## .x4 0.315 0.066 4.748 0.000 0.315
## .x5 0.419 0.072 5.857 0.000 0.419
## .x6 0.406 0.076 5.345 0.000 0.406
## .x7 0.601 0.100 5.976 0.000 0.601
## .x8 0.401 0.160 2.508 0.012 0.401
## .x9 0.535 0.141 3.801 0.000 0.535
## Visual 0.604 0.195 3.094 0.002 1.000
## Textual 0.942 0.163 5.787 0.000 1.000
## Speed 0.461 0.121 3.826 0.000 1.000
## Std.all
## 0.542
## 0.733
## 0.519
## 0.251
## 0.312
## 0.317
## 0.566
## 0.367
## 0.509
## 1.000
## 1.000
## 1.000
##
## R-Square:
## Estimate
## x1 0.458
## x2 0.267
## x3 0.481
## x4 0.749
## x5 0.688
## x6 0.683
## x7 0.434
## x8 0.633
## x9 0.491

```

```

weak <- cfa(HS.model, data = HS, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL", group = "school",
  group.equal = "loadings")
summary(weak, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 52 iterations
##
##   Number of observations per group
##   Pasteur                        156
##   Grant-White                   145
##   Number of missing patterns per group
##   Pasteur                        1
##   Grant-White                   1
##
##   Estimator                      ML      Robust
##   Model Fit Test Statistic       123.222  125.163
##   Degrees of freedom              54      54
##   P-value (Chi-square)            0.000    0.000
##   Scaling correction factor
##   for the Yuan-Bentler correction (Mplus variant)
##   0.984
##
## Chi-square for each group:
##
##   Pasteur                        68.384    69.461
##   Grant-White                   54.838    55.702
##
## Model test baseline model:
##
##   Minimum Function Test Statistic  951.384    928.080
##   Degrees of freedom               72      72
##   P-value                           0.000    0.000
##
## User model versus baseline model:
##
##   Comparative Fit Index (CFI)       0.921    0.917
##   Tucker-Lewis Index (TLI)         0.895    0.889
##
##   Robust Comparative Fit Index (CFI)      NA
##   Robust Tucker-Lewis Index (TLI)         NA
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)       -3686.294  -3686.294
##   Loglikelihood unrestricted model (H1) -3624.272  -3624.272
##
##   Number of free parameters           54      54
##   Akaike (AIC)                        7480.587  7480.587
##   Bayesian (BIC)                      7680.771  7680.771
##   Sample-size adjusted Bayesian (BIC)    7509.514  7509.514
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                             0.093    0.094
##   90 Percent Confidence Interval      0.071  0.114    0.072  0.116

```

```

##      P-value RMSEA <= 0.05                0.001      0.001
##
##      Robust RMSEA                        NA
##      90 Percent Confidence Interval      NA      NA
##
## Standardized Root Mean Square Residual:
##
##      SRMR                0.072      0.072
##
## Parameter Estimates:
##
##      Information                        Observed
##      Observed information based on      Hessian
##      Standard Errors                  Robust.huber.white
##
##
## Group 1 [Pasteur]:
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##      Visual =~
##      x1          1.000
##      x2      (.p2.)  0.599    0.140    4.271    0.000    0.537
##      x3      (.p3.)  0.784    0.153    5.129    0.000    0.704
##      Textual =~
##      x4          1.000
##      x5      (.p5.)  1.083    0.070   15.557    0.000    1.035
##      x6      (.p6.)  0.912    0.069   13.239    0.000    0.871
##      Speed =~
##      x7          1.000
##      x8      (.p8.)  1.201    0.136    8.855    0.000    0.663
##      x9      (.p9.)  1.038    0.207    5.006    0.000    0.573
##      Std.all
##
##      0.771
##      0.432
##      0.600
##
##      0.823
##      0.824
##      0.860
##
##      0.514
##      0.679
##      0.577
##
## Covariances:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##      Visual ~~
##      Textual      0.416    0.135    3.070    0.002    0.485
##      Speed        0.169    0.066    2.556    0.011    0.340
##      Textual ~~
##      Speed        0.176    0.061    2.895    0.004    0.333
##      Std.all

```

```

##
##      0.485
##      0.340
##
##      0.333
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      4.941   0.095  52.075   0.000   4.941
##      .x2      5.984   0.099  60.746   0.000   5.984
##      .x3      2.487   0.093  26.689   0.000   2.487
##      .x4      2.823   0.092  30.587   0.000   2.823
##      .x5      3.995   0.105  38.056   0.000   3.995
##      .x6      1.922   0.079  24.240   0.000   1.922
##      .x7      4.432   0.087  51.011   0.000   4.432
##      .x8      5.563   0.078  70.977   0.000   5.563
##      .x9      5.418   0.079  68.212   0.000   5.418
##      Visual    0.000
##      Textual    0.000
##      Speed      0.000
## Std.all
##      4.243
##      4.812
##      2.119
##      2.432
##      3.180
##      1.898
##      4.126
##      5.692
##      5.457
##      0.000
##      0.000
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      0.551   0.176   3.133   0.002   0.551
##      .x2      1.258   0.166   7.587   0.000   1.258
##      .x3      0.882   0.129   6.824   0.000   0.882
##      .x4      0.434   0.072   6.013   0.000   0.434
##      .x5      0.508   0.080   6.366   0.000   0.508
##      .x6      0.266   0.057   4.660   0.000   0.266
##      .x7      0.849   0.113   7.495   0.000   0.849
##      .x8      0.515   0.095   5.430   0.000   0.515
##      .x9      0.658   0.118   5.555   0.000   0.658
##      Visual    0.805   0.227   3.547   0.000   1.000
##      Textual    0.913   0.139   6.559   0.000   1.000
##      Speed      0.305   0.083   3.689   0.000   1.000
## Std.all
##      0.406
##      0.813
##      0.640
##      0.322
##      0.322

```

```

##      0.260
##      0.736
##      0.539
##      0.667
##      1.000
##      1.000
##      1.000
##
## R-Square:
##              Estimate
##      x1              0.594
##      x2              0.187
##      x3              0.360
##      x4              0.678
##      x5              0.678
##      x6              0.740
##      x7              0.264
##      x8              0.461
##      x9              0.333
##
##
## Group 2 [Grant-White]:
##
## Latent Variables:
##              Estimate  Std.Err  z-value  P(>|z|)  Std.lv
## Visual =~
##      x1              1.000              0.850
##      x2      (.p2.)    0.599      0.140      4.271      0.000      0.509
##      x3      (.p3.)    0.784      0.153      5.129      0.000      0.667
## Textual =~
##      x4              1.000              0.952
##      x5      (.p5.)    1.083      0.070     15.557      0.000      1.031
##      x6      (.p6.)    0.912      0.069     13.239      0.000      0.868
## Speed =~
##      x7              1.000              0.689
##      x8      (.p8.)    1.201      0.136      8.855      0.000      0.828
##      x9      (.p9.)    1.038      0.207      5.006      0.000      0.715
## Std.all
##
##      0.727
##      0.466
##      0.651
##
##      0.857
##      0.857
##      0.795
##
##      0.665
##      0.793
##      0.700
##
## Covariances:
##              Estimate  Std.Err  z-value  P(>|z|)  Std.lv
## Visual ~~

```

```

##      Textual      0.437    0.108    4.054    0.000    0.540
##      Speed      0.314    0.097    3.242    0.001    0.536
##      Textual ~~
##      Speed      0.226    0.099    2.280    0.023    0.345
##      Std.all
##
##      0.540
##      0.536
##
##      0.345
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      4.930   0.096  51.524   0.000   4.930
##      .x2      6.200   0.092  67.192   0.000   6.200
##      .x3      1.996   0.086  23.118   0.000   1.996
##      .x4      3.317   0.093  35.507   0.000   3.317
##      .x5      4.712   0.097  48.823   0.000   4.712
##      .x6      2.469   0.094  26.189   0.000   2.469
##      .x7      3.921   0.086  45.666   0.000   3.921
##      .x8      5.488   0.087  62.963   0.000   5.488
##      .x9      5.327   0.085  62.363   0.000   5.327
##      Visual    0.000
##      Textual    0.000
##      Speed      0.000
##      Std.all
##      4.216
##      5.679
##      1.948
##      2.985
##      3.918
##      2.263
##      3.783
##      5.253
##      5.214
##      0.000
##      0.000
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      0.645   0.171   3.780   0.000   0.645
##      .x2      0.933   0.146   6.392   0.000   0.933
##      .x3      0.605   0.113   5.352   0.000   0.605
##      .x4      0.329   0.066   4.999   0.000   0.329
##      .x5      0.384   0.072   5.324   0.000   0.384
##      .x6      0.437   0.075   5.838   0.000   0.437
##      .x7      0.599   0.095   6.297   0.000   0.599
##      .x8      0.406   0.145   2.805   0.005   0.406
##      .x9      0.532   0.126   4.216   0.000   0.532
##      Visual    0.722   0.176   4.094   0.000   1.000
##      Textual    0.906   0.147   6.172   0.000   1.000
##      Speed      0.475   0.113   4.193   0.000   1.000
##      Std.all

```

```

##      0.472
##      0.783
##      0.577
##      0.266
##      0.265
##      0.367
##      0.558
##      0.372
##      0.510
##      1.000
##      1.000
##      1.000
##
## R-Square:
##              Estimate
##      x1              0.528
##      x2              0.217
##      x3              0.423
##      x4              0.734
##      x5              0.735
##      x6              0.633
##      x7              0.442
##      x8              0.628
##      x9              0.490

strong <- cfa(HS.model, data = HS, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL", group = "school",
  group.equal = c("loadings", "intercepts"))
summary(strong, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 61 iterations
##
##      Number of observations per group
##      Pasteur                      156
##      Grant-White                  145
##      Number of missing patterns per group
##      Pasteur                      1
##      Grant-White                  1
##
##      Estimator              ML      Robust
##      Model Fit Test Statistic      163.015      165.643
##      Degrees of freedom              60      60
##      P-value (Chi-square)            0.000      0.000
##      Scaling correction factor              0.984
##      for the Yuan-Bentler correction (Mplus variant)
##
## Chi-square for each group:
##
##      Pasteur              89.632      91.077
##      Grant-White          73.383      74.566
##
## Model test baseline model:
##
##      Minimum Function Test Statistic      951.384      928.080

```



```

## Degrees of freedom              72          72
## P-value                        0.000        0.000
##
## User model versus baseline model:
##
## Comparative Fit Index (CFI)      0.883        0.877
## Tucker-Lewis Index (TLI)        0.859        0.852
##
## Robust Comparative Fit Index (CFI)          NA
## Robust Tucker-Lewis Index (TLI)          NA
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0)      -3706.323   -3706.323
## Loglikelihood unrestricted model (H1) -3624.272   -3624.272
##
## Number of free parameters          48          48
## Akaike (AIC)                      7508.647   7508.647
## Bayesian (BIC)                    7686.588   7686.588
## Sample-size adjusted Bayesian (BIC) 7534.359   7534.359
##
## Root Mean Square Error of Approximation:
##
## RMSEA                            0.107        0.109
## 90 Percent Confidence Interval      0.088 0.127        0.089 0.129
## P-value RMSEA <= 0.05              0.000        0.000
##
## Robust RMSEA                      NA
## 90 Percent Confidence Interval      NA        NA
##
## Standardized Root Mean Square Residual:
##
## SRMR                            0.082        0.082
##
## Parameter Estimates:
##
## Information                      Observed
## Observed information based on      Hessian
## Standard Errors                   Robust.huber.white
##
## Group 1 [Pasteur]:
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv
## Visual =~
##   x1          1.000
##   x2      (.p2.) 0.576 0.131 4.380 0.000 0.514
##   x3      (.p3.) 0.798 0.173 4.612 0.000 0.712
## Textual =~
##   x4          1.000
##   x5      (.p5.) 1.120 0.068 16.568 0.000 1.050
##   x6      (.p6.) 0.932 0.064 14.603 0.000 0.874
## Speed =~

```

```

##      x7      1.000      0.568
##      x8      (.p8.) 1.130    0.134    8.459    0.000    0.641
##      x9      (.p9.) 1.009    0.208    4.848    0.000    0.573
## Std.all
##
##      0.768
##      0.411
##      0.591
##
##      0.815
##      0.829
##      0.862
##
##      0.516
##      0.657
##      0.578
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
## Visual ~~
## Textual      0.410    0.135    3.033    0.002    0.490
## Speed        0.178    0.067    2.652    0.008    0.351
## Textual ~~
## Speed        0.180    0.063    2.880    0.004    0.338
## Std.all
##
##      0.490
##      0.351
##
##      0.338
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
## .x1      (.25.) 5.001    0.095   52.874    0.000    5.001
## .x2      (.26.) 6.151    0.087   70.631    0.000    6.151
## .x3      (.27.) 2.271    0.095   23.937    0.000    2.271
## .x4      (.28.) 2.778    0.087   32.065    0.000    2.778
## .x5      (.29.) 4.035    0.104   38.965    0.000    4.035
## .x6      (.30.) 1.926    0.075   25.665    0.000    1.926
## .x7      (.31.) 4.242    0.079   53.523    0.000    4.242
## .x8      (.32.) 5.630    0.075   74.753    0.000    5.630
## .x9      (.33.) 5.465    0.072   75.861    0.000    5.465
## Visual      0.000      0.000
## Textual      0.000      0.000
## Speed        0.000      0.000
## Std.all
##      4.302
##      4.925
##      1.885
##      2.413
##      3.184
##      1.900
##      3.855
##      5.771

```

```

##      5.516
##      0.000
##      0.000
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      0.555   0.185   3.007   0.003   0.555
##      .x2      1.296   0.162   8.021   0.000   1.296
##      .x3      0.944   0.147   6.424   0.000   0.944
##      .x4      0.445   0.073   6.093   0.000   0.445
##      .x5      0.502   0.081   6.177   0.000   0.502
##      .x6      0.263   0.058   4.510   0.000   0.263
##      .x7      0.888   0.128   6.929   0.000   0.888
##      .x8      0.541   0.089   6.052   0.000   0.541
##      .x9      0.654   0.115   5.666   0.000   0.654
##      Visual    0.796   0.239   3.325   0.001   1.000
##      Textual   0.879   0.135   6.491   0.000   1.000
##      Speed     0.322   0.092   3.488   0.000   1.000
## Std.all
##      0.411
##      0.831
##      0.650
##      0.336
##      0.313
##      0.256
##      0.734
##      0.568
##      0.666
##      1.000
##      1.000
##      1.000
##
## R-Square:
##      Estimate
##      x1      0.589
##      x2      0.169
##      x3      0.350
##      x4      0.664
##      x5      0.687
##      x6      0.744
##      x7      0.266
##      x8      0.432
##      x9      0.334
##
##
## Group 2 [Grant-White]:
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      Visual =~
##      x1      1.000
##      x2      (.p2.) 0.576   0.131   4.380   0.000   0.484
##      x3      (.p3.) 0.798   0.173   4.612   0.000   0.672

```

```

##      Textual =~
##      x4          1.000          0.933
##      x5      (.p5.)  1.120    0.068   16.568    0.000   1.045
##      x6      (.p6.)  0.932    0.064   14.603    0.000   0.869
##      Speed =~
##      x7          1.000          0.711
##      x8      (.p8.)  1.130    0.134    8.459    0.000   0.803
##      x9      (.p9.)  1.009    0.208    4.848    0.000   0.717
##      Std.all
##
##      0.721
##      0.442
##      0.643
##
##      0.847
##      0.862
##      0.796
##
##      0.668
##      0.773
##      0.704
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      Visual ~~
##      Textual      0.427    0.107    3.990    0.000    0.544
##      Speed        0.329    0.101    3.257    0.001    0.550
##      Textual ~~
##      Speed        0.236    0.098    2.415    0.016    0.356
##      Std.all
##
##      0.544
##      0.550
##
##      0.356
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      (.25.)  5.001    0.095   52.874    0.000    5.001
##      .x2      (.26.)  6.151    0.087   70.631    0.000    6.151
##      .x3      (.27.)  2.271    0.095   23.937    0.000    2.271
##      .x4      (.28.)  2.778    0.087   32.065    0.000    2.778
##      .x5      (.29.)  4.035    0.104   38.965    0.000    4.035
##      .x6      (.30.)  1.926    0.075   25.665    0.000    1.926
##      .x7      (.31.)  4.242    0.079   53.523    0.000    4.242
##      .x8      (.32.)  5.630    0.075   74.753    0.000    5.630
##      .x9      (.33.)  5.465    0.072   75.861    0.000    5.465
##      Visual    -0.148    0.141   -1.050    0.294   -0.176
##      Textual    0.576    0.119    4.824    0.000    0.618
##      Speed     -0.177    0.104   -1.706    0.088   -0.250
##      Std.all
##      4.286
##      5.618
##      2.174

```

```

##      2.522
##      3.330
##      1.763
##      3.991
##      5.422
##      5.369
##     -0.176
##      0.618
##     -0.250
##
## Variances:
##           Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1           0.654   0.178   3.679   0.000   0.654
##      .x2           0.964   0.152   6.330   0.000   0.964
##      .x3           0.641   0.130   4.933   0.000   0.641
##      .x4           0.343   0.065   5.258   0.000   0.343
##      .x5           0.376   0.073   5.146   0.000   0.376
##      .x6           0.437   0.074   5.902   0.000   0.437
##      .x7           0.625   0.106   5.882   0.000   0.625
##      .x8           0.434   0.148   2.938   0.003   0.434
##      .x9           0.522   0.126   4.149   0.000   0.522
##      Visual       0.708   0.187   3.791   0.000   1.000
##      Textual      0.870   0.144   6.048   0.000   1.000
##      Speed        0.505   0.125   4.040   0.000   1.000
## Std.all
##      0.480
##      0.804
##      0.587
##      0.283
##      0.256
##      0.366
##      0.553
##      0.403
##      0.504
##      1.000
##      1.000
##      1.000
##
## R-Square:
##           Estimate
##      x1           0.520
##      x2           0.196
##      x3           0.413
##      x4           0.717
##      x5           0.744
##      x6           0.634
##      x7           0.447
##      x8           0.597
##      x9           0.496

strict <- cfa(HS.model, data = HS, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL", group = "school",
  group.equal = c("loadings", "intercepts", "residuals"))
summary(strict, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

```

```

## lavaan (0.6-1) converged normally after 60 iterations
##
##   Number of observations per group
##   Pasteur                        156
##   Grant-White                    145
##   Number of missing patterns per group
##   Pasteur                        1
##   Grant-White                    1
##
##   Estimator                      ML      Robust
##   Model Fit Test Statistic        180.305  180.223
##   Degrees of freedom              69      69
##   P-value (Chi-square)            0.000    0.000
##   Scaling correction factor        1.000
##   for the Yuan-Bentler correction (Mplus variant)
##
## Chi-square for each group:
##
##   Pasteur                        92.496    92.454
##   Grant-White                    87.809    87.769
##
## Model test baseline model:
##
##   Minimum Function Test Statistic  951.384    928.080
##   Degrees of freedom              72      72
##   P-value                        0.000    0.000
##
## User model versus baseline model:
##
##   Comparative Fit Index (CFI)      0.873    0.870
##   Tucker-Lewis Index (TLI)        0.868    0.864
##
##   Robust Comparative Fit Index (CFI)      NA
##   Robust Tucker-Lewis Index (TLI)         NA
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)      -3715.028  -3715.028
##   Loglikelihood unrestricted model (H1) -3624.272  -3624.272
##
##   Number of free parameters          39      39
##   Akaike (AIC)                      7508.055  7508.055
##   Bayesian (BIC)                     7652.632  7652.632
##   Sample-size adjusted Bayesian (BIC)  7528.947  7528.947
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                          0.104    0.104
##   90 Percent Confidence Interval    0.085  0.123    0.085  0.123
##   P-value RMSEA <= 0.05           0.000    0.000
##
##   Robust RMSEA                      NA
##   90 Percent Confidence Interval    NA      NA

```

```

##
## Standardized Root Mean Square Residual:
##
##   SRMR                      0.088      0.088
##
## Parameter Estimates:
##
##   Information                      Observed
##   Observed information based on      Hessian
##   Standard Errors                  Robust.huber.white
##
##
## Group 1 [Pasteur]:
##
## Latent Variables:
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##   Visual =~
##     x1           1.000
##     x2      (.p2.)  0.591    0.142    4.159    0.000    0.518
##     x3      (.p3.)  0.837    0.213    3.936    0.000    0.733
##   Textual =~
##     x4           1.000
##     x5      (.p5.)  1.125    0.066   16.932    0.000    1.064
##     x6      (.p6.)  0.933    0.061   15.245    0.000    0.882
##   Speed =~
##     x7           1.000
##     x8      (.p8.)  1.121    0.137    8.180    0.000    0.654
##     x9      (.p9.)  1.028    0.220    4.664    0.000    0.600
##   Std.all
##
##     0.739
##     0.438
##     0.641
##
##     0.837
##     0.850
##     0.829
##
##     0.554
##     0.678
##     0.620
##
## Covariances:
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##   Visual ~~
##     Textual      0.367    0.150    2.455    0.014    0.444
##     Speed      0.174    0.070    2.480    0.013    0.341
##   Textual ~~
##     Speed      0.176    0.063    2.809    0.005    0.319
##   Std.all
##
##     0.444
##     0.341
##

```

```

##      0.319
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      (.25.)   5.012   0.095  52.568   0.000   5.012
##      .x2      (.26.)   6.133   0.085  71.826   0.000   6.133
##      .x3      (.27.)   2.314   0.097  23.915   0.000   2.314
##      .x4      (.28.)   2.784   0.087  32.046   0.000   2.784
##      .x5      (.29.)   4.029   0.102  39.323   0.000   4.029
##      .x6      (.30.)   1.927   0.075  25.696   0.000   1.927
##      .x7      (.31.)   4.271   0.078  54.872   0.000   4.271
##      .x8      (.32.)   5.622   0.074  75.588   0.000   5.622
##      .x9      (.33.)   5.461   0.072  76.041   0.000   5.461
##      Visual           0.000           0.000
##      Textual          0.000           0.000
##      Speed            0.000           0.000
## Std.all
##      4.227
##      5.186
##      2.023
##      2.464
##      3.219
##      1.811
##      4.056
##      5.834
##      5.644
##      0.000
##      0.000
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      (.10.)   0.638   0.176   3.616   0.000   0.638
##      .x2      (.11.)   1.130   0.111  10.192   0.000   1.130
##      .x3      (.12.)   0.771   0.134   5.751   0.000   0.771
##      .x4      (.13.)   0.383   0.050   7.642   0.000   0.383
##      .x5      (.14.)   0.435   0.057   7.568   0.000   0.435
##      .x6      (.15.)   0.354   0.047   7.484   0.000   0.354
##      .x7      (.16.)   0.769   0.094   8.185   0.000   0.769
##      .x8      (.17.)   0.501   0.092   5.445   0.000   0.501
##      .x9      (.18.)   0.576   0.099   5.841   0.000   0.576
##      Visual           0.767   0.225   3.416   0.001   1.000
##      Textual          0.894   0.135   6.601   0.000   1.000
##      Speed            0.340   0.103   3.312   0.001   1.000
## Std.all
##      0.454
##      0.808
##      0.589
##      0.300
##      0.278
##      0.312
##      0.693
##      0.540
##      0.616

```



```

##      1.000
##      1.000
##      1.000
##
## R-Square:
##              Estimate
##      x1              0.546
##      x2              0.192
##      x3              0.411
##      x4              0.700
##      x5              0.722
##      x6              0.688
##      x7              0.307
##      x8              0.460
##      x9              0.384
##
##
## Group 2 [Grant-White]:
##
## Latent Variables:
##              Estimate  Std.Err  z-value  P(>|z|)  Std.lv
## Visual =~
##      x1              1.000              0.810
##      x2      (.p2.)    0.591    0.142    4.159    0.000    0.479
##      x3      (.p3.)    0.837    0.213    3.936    0.000    0.678
## Textual =~
##      x4              1.000              0.936
##      x5      (.p5.)    1.125    0.066   16.932    0.000    1.053
##      x6      (.p6.)    0.933    0.061   15.245    0.000    0.874
## Speed =~
##      x7              1.000              0.692
##      x8      (.p8.)    1.121    0.137    8.180    0.000    0.775
##      x9      (.p9.)    1.028    0.220    4.664    0.000    0.711
## Std.all
##
##      0.712
##      0.411
##      0.611
##
##      0.834
##      0.847
##      0.827
##
##      0.619
##      0.738
##      0.684
##
## Covariances:
##              Estimate  Std.Err  z-value  P(>|z|)  Std.lv
## Visual ~~
##      Textual          0.422    0.109    3.888    0.000    0.556
##      Speed            0.331    0.100    3.309    0.001    0.590
## Textual ~~
##      Speed            0.236    0.098    2.404    0.016    0.364

```

```

## Std.all
##
## 0.556
## 0.590
##
## 0.364
##
## Intercepts:
##          Estimate Std.Err z-value P(>|z|) Std.lv
## .x1      (.25.)   5.012   0.095  52.568   0.000   5.012
## .x2      (.26.)   6.133   0.085  71.826   0.000   6.133
## .x3      (.27.)   2.314   0.097  23.915   0.000   2.314
## .x4      (.28.)   2.784   0.087  32.046   0.000   2.784
## .x5      (.29.)   4.029   0.102  39.323   0.000   4.029
## .x6      (.30.)   1.927   0.075  25.696   0.000   1.927
## .x7      (.31.)   4.271   0.078  54.872   0.000   4.271
## .x8      (.32.)   5.622   0.074  75.588   0.000   5.622
## .x9      (.33.)   5.461   0.072  76.041   0.000   5.461
## Visual                -0.157   0.144  -1.096   0.273  -0.194
## Textual                0.575   0.118   4.856   0.000   0.614
## Speed                 -0.176   0.105  -1.675   0.094  -0.255
## Std.all
## 4.404
## 5.260
## 2.086
## 2.481
## 3.243
## 1.824
## 3.825
## 5.356
## 5.249
## -0.194
## 0.614
## -0.255
##
## Variances:
##          Estimate Std.Err z-value P(>|z|) Std.lv
## .x1      (.10.)   0.638   0.176   3.616   0.000   0.638
## .x2      (.11.)   1.130   0.111  10.192   0.000   1.130
## .x3      (.12.)   0.771   0.134   5.751   0.000   0.771
## .x4      (.13.)   0.383   0.050   7.642   0.000   0.383
## .x5      (.14.)   0.435   0.057   7.568   0.000   0.435
## .x6      (.15.)   0.354   0.047   7.484   0.000   0.354
## .x7      (.16.)   0.769   0.094   8.185   0.000   0.769
## .x8      (.17.)   0.501   0.092   5.445   0.000   0.501
## .x9      (.18.)   0.576   0.099   5.841   0.000   0.576
## Visual                0.657   0.227   2.892   0.004   1.000
## Textual                0.876   0.143   6.140   0.000   1.000
## Speed                0.478   0.126   3.801   0.000   1.000
## Std.all
## 0.493
## 0.831
## 0.626
## 0.304

```

```
##      0.282
##      0.317
##      0.616
##      0.455
##      0.533
##      1.000
##      1.000
##      1.000
```

```
##
## R-Square:
##              Estimate
##      x1          0.507
##      x2          0.169
##      x3          0.374
##      x4          0.696
##      x5          0.718
##      x6          0.683
##      x7          0.384
##      x8          0.545
##      x9          0.467
```

```
strict_2 <- cfa(HS.model, data = HS, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL", group = "school",
  group.equal = c("loadings", "intercepts", "residuals", "means"))
summary(strict_2, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)
```

```
## lavaan (0.6-1) converged normally after 57 iterations
##
##      Number of observations per group
##      Pasteur                      156
##      Grant-White                  145
##      Number of missing patterns per group
##      Pasteur                      1
##      Grant-White                  1
##
##      Estimator                    ML      Robust
##      Model Fit Test Statistic      219.863  219.162
##      Degrees of freedom              72      72
##      P-value (Chi-square)           0.000    0.000
##      Scaling correction factor      1.003
##      for the Yuan-Bentler correction (Mplus variant)
##
## Chi-square for each group:
##
##      Pasteur                      110.637  110.284
##      Grant-White                  109.226  108.878
##
## Model test baseline model:
##
##      Minimum Function Test Statistic  951.384  928.080
##      Degrees of freedom              72      72
##      P-value                        0.000    0.000
##
```

```

## User model versus baseline model:
##
##   Comparative Fit Index (CFI)                0.832      0.828
##   Tucker-Lewis Index (TLI)                  0.832      0.828
##
##   Robust Comparative Fit Index (CFI)          NA
##   Robust Tucker-Lewis Index (TLI)            NA
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)              -3734.939   -3734.939
##   Loglikelihood unrestricted model (H1)       -3624.272   -3624.272
##
##   Number of free parameters                  36         36
##   Akaike (AIC)                             7541.879   7541.879
##   Bayesian (BIC)                           7675.335   7675.335
##   Sample-size adjusted Bayesian (BIC)        7561.163   7561.163
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                                0.117      0.117
##   90 Percent Confidence Interval          0.100  0.135      0.099  0.135
##   P-value RMSEA <= 0.05                  0.000      0.000
##
##   Robust RMSEA                              NA
##   90 Percent Confidence Interval          NA      NA
##
## Standardized Root Mean Square Residual:
##
##   SRMR                                0.114      0.114
##
## Parameter Estimates:
##
##   Information                                Observed
##   Observed information based on              Hessian
##   Standard Errors                          Robust.huber.white
##
## Group 1 [Pasteur]:
##
## Latent Variables:
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv
## Visual =~
##   x1           1.000
##   x2      (.p2.)  0.571    0.144    3.967    0.000    0.521
##   x3      (.p3.)  0.753    0.166    4.535    0.000    0.688
## Textual =~
##   x4           1.000
##   x5      (.p5.)  1.113    0.066   16.761    0.000    1.106
##   x6      (.p6.)  0.926    0.062   14.952    0.000    0.920
## Speed =~
##   x7           1.000
##   x8      (.p8.)  1.187    0.130    9.111    0.000    0.668
##   x9      (.p9.)  1.084    0.265    4.092    0.000    0.610

```

```

## Std.all
##
## 0.770
## 0.441
## 0.602
##
## 0.852
## 0.856
## 0.839
##
## 0.533
## 0.693
## 0.630
##
## Covariances:
## Estimate Std.Err z-value P(>|z|) Std.lv
## Visual ~~
## Textual 0.385 0.154 2.494 0.013 0.424
## Speed 0.181 0.069 2.621 0.009 0.351
## Textual ~~
## Speed 0.150 0.062 2.444 0.015 0.269
## Std.all
##
## 0.424
## 0.351
##
## 0.269
##
## Intercepts:
## Estimate Std.Err z-value P(>|z|) Std.lv
## .x1 (.25.) 4.939 0.077 64.457 0.000 4.939
## .x2 (.26.) 6.090 0.069 88.522 0.000 6.090
## .x3 (.27.) 2.253 0.071 31.566 0.000 2.253
## .x4 (.28.) 3.070 0.073 42.211 0.000 3.070
## .x5 (.29.) 4.350 0.083 52.704 0.000 4.350
## .x6 (.30.) 2.194 0.070 31.346 0.000 2.194
## .x7 (.31.) 4.201 0.067 62.871 0.000 4.201
## .x8 (.32.) 5.545 0.066 84.185 0.000 5.545
## .x9 (.33.) 5.391 0.061 88.256 0.000 5.391
## Visual 0.000 0.000
## Textual 0.000 0.000
## Speed 0.000 0.000
## Std.all
## 4.162
## 5.150
## 1.973
## 2.635
## 3.368
## 2.001
## 3.972
## 5.746
## 5.562
## 0.000
## 0.000

```

```

##      0.000
##
## Variances:
##              Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      (.10.)    0.574   0.172   3.333   0.001   0.574
##      .x2      (.11.)    1.126   0.114   9.842   0.000   1.126
##      .x3      (.12.)    0.830   0.110   7.539   0.000   0.830
##      .x4      (.13.)    0.371   0.051   7.348   0.000   0.371
##      .x5      (.14.)    0.446   0.058   7.628   0.000   0.446
##      .x6      (.15.)    0.357   0.048   7.418   0.000   0.357
##      .x7      (.16.)    0.802   0.096   8.374   0.000   0.802
##      .x8      (.17.)    0.485   0.115   4.202   0.000   0.485
##      .x9      (.18.)    0.567   0.119   4.779   0.000   0.567
##      Visual              0.835   0.219   3.816   0.000   1.000
##      Textual             0.986   0.151   6.517   0.000   1.000
##      Speed               0.317   0.100   3.180   0.001   1.000
## Std.all
##      0.407
##      0.805
##      0.637
##      0.273
##      0.267
##      0.297
##      0.716
##      0.520
##      0.603
##      1.000
##      1.000
##      1.000
##
## R-Square:
##              Estimate
##      x1              0.593
##      x2              0.195
##      x3              0.363
##      x4              0.727
##      x5              0.733
##      x6              0.703
##      x7              0.284
##      x8              0.480
##      x9              0.397
##
##
## Group 2 [Grant-White]:
##
## Latent Variables:
##              Estimate Std.Err z-value P(>|z|) Std.lv
##      Visual =~
##      x1              1.000              0.855
##      x2      (.p2.)    0.571   0.144   3.967   0.000   0.488
##      x3      (.p3.)    0.753   0.166   4.535   0.000   0.643
##      Textual =~
##      x4              1.000              0.986
##      x5      (.p5.)    1.113   0.066  16.761   0.000   1.098

```

```

##      x6      (.p6.)      0.926      0.062      14.952      0.000      0.913
## Speed =~
##      x7              1.000              0.672
##      x8      (.p8.)      1.187      0.130      9.111      0.000      0.797
##      x9      (.p9.)      1.084      0.265      4.092      0.000      0.728
## Std.all
##
##      0.748
##      0.418
##      0.577
##
##      0.851
##      0.854
##      0.837
##
##      0.600
##      0.753
##      0.695
##
## Covariances:
##              Estimate Std.Err z-value P(>|z|) Std.lv
## Visual ~~
##      Textual      0.417      0.114      3.670      0.000      0.494
##      Speed      0.342      0.093      3.664      0.000      0.596
## Textual ~~
##      Speed      0.197      0.099      1.996      0.046      0.297
## Std.all
##
##      0.494
##      0.596
##
##      0.297
##
## Intercepts:
##              Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      (.25.)      4.939      0.077      64.457      0.000      4.939
##      .x2      (.26.)      6.090      0.069      88.522      0.000      6.090
##      .x3      (.27.)      2.253      0.071      31.566      0.000      2.253
##      .x4      (.28.)      3.070      0.073      42.211      0.000      3.070
##      .x5      (.29.)      4.350      0.083      52.704      0.000      4.350
##      .x6      (.30.)      2.194      0.070      31.346      0.000      2.194
##      .x7      (.31.)      4.201      0.067      62.871      0.000      4.201
##      .x8      (.32.)      5.545      0.066      84.185      0.000      5.545
##      .x9      (.33.)      5.391      0.061      88.256      0.000      5.391
##      Visual      0.000              0.000
##      Textual      0.000              0.000
##      Speed      0.000              0.000
## Std.all
##      4.324
##      5.214
##      2.020
##      2.649
##      3.386

```

```
##      2.011
##      3.754
##      5.240
##      5.148
##      0.000
##      0.000
##      0.000
##
## Variances:
##              Estimate Std.Err  z-value  P(>|z|)  Std.lv
##      .x1      (.10.)    0.574    0.172    3.333    0.001    0.574
##      .x2      (.11.)    1.126    0.114    9.842    0.000    1.126
##      .x3      (.12.)    0.830    0.110    7.539    0.000    0.830
##      .x4      (.13.)    0.371    0.051    7.348    0.000    0.371
##      .x5      (.14.)    0.446    0.058    7.628    0.000    0.446
##      .x6      (.15.)    0.357    0.048    7.418    0.000    0.357
##      .x7      (.16.)    0.802    0.096    8.374    0.000    0.802
##      .x8      (.17.)    0.485    0.115    4.202    0.000    0.485
##      .x9      (.18.)    0.567    0.119    4.779    0.000    0.567
##      Visual              0.731    0.222    3.296    0.001    1.000
##      Textual             0.972    0.162    5.990    0.000    1.000
##      Speed               0.451    0.130    3.479    0.001    1.000
## Std.all
##      0.440
##      0.826
##      0.667
##      0.276
##      0.270
##      0.300
##      0.640
##      0.433
##      0.517
##      1.000
##      1.000
##      1.000
##
## R-Square:
##              Estimate
##      x1              0.560
##      x2              0.174
##      x3              0.333
##      x4              0.724
##      x5              0.730
##      x6              0.700
##      x7              0.360
##      x8              0.567
##      x9              0.483
```

```
strict_3 <- cfa(HS.model, data = HS, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL", group = "school",
  group.equal = c("loadings", "intercepts", "residuals", "means",
    "lv.variances", "lv.covariances"))
summary(strict_3, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)
```



```

## lavaan (0.6-1) converged normally after 55 iterations
##
##   Number of observations per group
##   Pasteur                        156
##   Grant-White                    145
##   Number of missing patterns per group
##   Pasteur                        1
##   Grant-White                    1
##
##   Estimator                      ML      Robust
##   Model Fit Test Statistic      225.436  223.115
##   Degrees of freedom             78      78
##   P-value (Chi-square)           0.000    0.000
##   Scaling correction factor      1.010
##   for the Yuan-Bentler correction (Mplus variant)
##
## Chi-square for each group:
##
##   Pasteur                        113.022  111.858
##   Grant-White                    112.414  111.257
##
## Model test baseline model:
##
##   Minimum Function Test Statistic  951.384  928.080
##   Degrees of freedom               72      72
##   P-value                          0.000    0.000
##
## User model versus baseline model:
##
##   Comparative Fit Index (CFI)      0.832    0.830
##   Tucker-Lewis Index (TLI)        0.845    0.844
##
##   Robust Comparative Fit Index (CFI)      NA
##   Robust Tucker-Lewis Index (TLI)        NA
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)      -3737.745  -3737.745
##   Loglikelihood unrestricted model (H1) -3624.272  -3624.272
##
##   Number of free parameters          30      30
##   Akaike (AIC)                       7535.490  7535.490
##   Bayesian (BIC)                      7646.703  7646.703
##   Sample-size adjusted Bayesian (BIC)  7551.560  7551.560
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                            0.112    0.112
##   90 Percent Confidence Interval      0.095  0.130    0.095  0.129
##   P-value RMSEA <= 0.05              0.000    0.000
##
##   Robust RMSEA                      NA
##   90 Percent Confidence Interval      NA      NA

```

```

##
## Standardized Root Mean Square Residual:
##
##   SRMR                                0.119      0.119
##
## Parameter Estimates:
##
##   Information                                Observed
##   Observed information based on              Hessian
##   Standard Errors                          Robust.huber.white
##
##
## Group 1 [Pasteur]:
##
## Latent Variables:
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##   Visual =~
##     x1           1.000
##     x2      (.p2.)  0.553    0.133    4.177    0.000    0.498
##     x3      (.p3.)  0.729    0.142    5.152    0.000    0.656
##   Textual =~
##     x4           1.000
##     x5      (.p5.)  1.113    0.066   16.889    0.000    1.102
##     x6      (.p6.)  0.926    0.062   15.039    0.000    0.917
##   Speed =~
##     x7           1.000
##     x8      (.p8.)  1.180    0.131    9.015    0.000    0.731
##     x9      (.p9.)  1.082    0.267    4.047    0.000    0.670
##   Std.all
##
##     0.772
##     0.424
##     0.581
##
##     0.852
##     0.855
##     0.838
##
##     0.570
##     0.723
##     0.665
##
## Covariances:
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##   Visual ~~
##     Textual (.22.)  0.408    0.100    4.097    0.000    0.459
##     Speed   (.23.)  0.262    0.060    4.352    0.000    0.471
##   Textual ~~
##     Speed   (.24.)  0.173    0.056    3.071    0.002    0.283
##   Std.all
##
##     0.459
##     0.471
##

```

```

##      0.283
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      (.25.)   4.936   0.067  73.229   0.000   4.936
##      .x2      (.26.)   6.088   0.068  89.556   0.000   6.088
##      .x3      (.27.)   2.250   0.065  34.464   0.000   2.250
##      .x4      (.28.)   3.061   0.067  45.542   0.000   3.061
##      .x5      (.29.)   4.341   0.075  58.258   0.000   4.341
##      .x6      (.30.)   2.186   0.063  34.552   0.000   2.186
##      .x7      (.31.)   4.186   0.063  66.544   0.000   4.186
##      .x8      (.32.)   5.527   0.058  94.539   0.000   5.527
##      .x9      (.33.)   5.374   0.058  92.238   0.000   5.374
##      Visual           0.000           0.000
##      Textual          0.000           0.000
##      Speed            0.000           0.000
## Std.all
##      4.235
##      5.179
##      1.993
##      2.634
##      3.369
##      1.998
##      3.848
##      5.467
##      5.334
##      0.000
##      0.000
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      (.10.)   0.549   0.157   3.497   0.000   0.549
##      .x2      (.11.)   1.134   0.112  10.101   0.000   1.134
##      .x3      (.12.)   0.844   0.101   8.391   0.000   0.844
##      .x4      (.13.)   0.371   0.050   7.357   0.000   0.371
##      .x5      (.14.)   0.446   0.057   7.844   0.000   0.446
##      .x6      (.15.)   0.356   0.047   7.632   0.000   0.356
##      .x7      (.16.)   0.799   0.098   8.195   0.000   0.799
##      .x8      (.17.)   0.488   0.120   4.066   0.000   0.488
##      .x9      (.18.)   0.566   0.119   4.752   0.000   0.566
##      Visual   (.19.)   0.809   0.181   4.471   0.000   1.000
##      Textual  (.20.)   0.979   0.122   8.048   0.000   1.000
##      Speed    (.21.)   0.384   0.107   3.584   0.000   1.000
## Std.all
##      0.404
##      0.821
##      0.662
##      0.275
##      0.269
##      0.298
##      0.676
##      0.477
##      0.558

```

```

##      1.000
##      1.000
##      1.000
##
## R-Square:
##              Estimate
##      x1              0.596
##      x2              0.179
##      x3              0.338
##      x4              0.725
##      x5              0.731
##      x6              0.702
##      x7              0.324
##      x8              0.523
##      x9              0.442
##
##
## Group 2 [Grant-White]:
##
## Latent Variables:
##              Estimate  Std.Err  z-value  P(>|z|)  Std.lv
## Visual =~
##      x1              1.000              0.900
##      x2      (.p2.)    0.553    0.133    4.177    0.000    0.498
##      x3      (.p3.)    0.729    0.142    5.152    0.000    0.656
## Textual =~
##      x4              1.000              0.990
##      x5      (.p5.)    1.113    0.066   16.889    0.000    1.102
##      x6      (.p6.)    0.926    0.062   15.039    0.000    0.917
## Speed =~
##      x7              1.000              0.619
##      x8      (.p8.)    1.180    0.131    9.015    0.000    0.731
##      x9      (.p9.)    1.082    0.267    4.047    0.000    0.670
## Std.all
##
##      0.772
##      0.424
##      0.581
##
##      0.852
##      0.855
##      0.838
##
##      0.570
##      0.723
##      0.665
##
## Covariances:
##              Estimate  Std.Err  z-value  P(>|z|)  Std.lv
## Visual ~~
##      Textual (.22.)    0.408    0.100    4.097    0.000    0.459
##      Speed   (.23.)    0.262    0.060    4.352    0.000    0.471
## Textual ~~
##      Speed   (.24.)    0.173    0.056    3.071    0.002    0.283

```

```

## Std.all
##
## 0.459
## 0.471
##
## 0.283
##
## Intercepts:
##          Estimate Std.Err z-value P(>|z|) Std.lv
## .x1      (.25.)   4.936   0.067  73.229   0.000   4.936
## .x2      (.26.)   6.088   0.068  89.556   0.000   6.088
## .x3      (.27.)   2.250   0.065  34.464   0.000   2.250
## .x4      (.28.)   3.061   0.067  45.542   0.000   3.061
## .x5      (.29.)   4.341   0.075  58.258   0.000   4.341
## .x6      (.30.)   2.186   0.063  34.552   0.000   2.186
## .x7      (.31.)   4.186   0.063  66.544   0.000   4.186
## .x8      (.32.)   5.527   0.058  94.539   0.000   5.527
## .x9      (.33.)   5.374   0.058  92.238   0.000   5.374
## Visual              0.000              0.000
## Textual             0.000              0.000
## Speed               0.000              0.000
## Std.all
## 4.235
## 5.179
## 1.993
## 2.634
## 3.369
## 1.998
## 3.848
## 5.467
## 5.334
## 0.000
## 0.000
## 0.000
##
## Variances:
##          Estimate Std.Err z-value P(>|z|) Std.lv
## .x1      (.10.)   0.549   0.157   3.497   0.000   0.549
## .x2      (.11.)   1.134   0.112  10.101   0.000   1.134
## .x3      (.12.)   0.844   0.101   8.391   0.000   0.844
## .x4      (.13.)   0.371   0.050   7.357   0.000   0.371
## .x5      (.14.)   0.446   0.057   7.844   0.000   0.446
## .x6      (.15.)   0.356   0.047   7.632   0.000   0.356
## .x7      (.16.)   0.799   0.098   8.195   0.000   0.799
## .x8      (.17.)   0.488   0.120   4.066   0.000   0.488
## .x9      (.18.)   0.566   0.119   4.752   0.000   0.566
## Visual   (.19.)   0.809   0.181   4.471   0.000   1.000
## Textual  (.20.)   0.979   0.122   8.048   0.000   1.000
## Speed    (.21.)   0.384   0.107   3.584   0.000   1.000
## Std.all
## 0.404
## 0.821
## 0.662
## 0.275

```

```
##      0.269
##      0.298
##      0.676
##      0.477
##      0.558
##      1.000
##      1.000
##      1.000
##
## R-Square:
##              Estimate
##      x1          0.596
##      x2          0.179
##      x3          0.338
##      x4          0.725
##      x5          0.731
##      x6          0.702
##      x7          0.324
##      x8          0.523
##      x9          0.442
```

```
anova(config, weak, strong, strict, strict_2, strict_3)
```

```
## Chi Square Difference Test
##
##      Df  AIC  BIC Chisq Chisq diff Df diff Pr(>Chisq)
## config  48 7484 7707   115
## weak    54 7481 7681   123      8.1      6      0.228
## strong  60 7509 7687   163     39.8      6 0.000000500
## strict  69 7508 7653   180     17.3      9      0.044
## strict_2 72 7542 7675   220     39.6      3 0.000000013
## strict_3 78 7535 7647   225      5.6      6      0.473
```

```
measurementInvariance(HS.model, data = HS, group = "school")
```

```
##
## Measurement invariance models:
##
## Model 1 : fit.configural
## Model 2 : fit.loadings
## Model 3 : fit.intercepts
## Model 4 : fit.means
##
## Chi Square Difference Test
##
##      Df  AIC  BIC Chisq Chisq diff Df diff Pr(>Chisq)
## fit.configural 48 7484 7707   116
## fit.loadings   54 7481 7681   124      8.2      6      0.22
## fit.intercepts 60 7509 7687   164     40.1      6 4.4e-07
## fit.means      63 7543 7710   205     40.5      3 8.3e-09
##
##
## Fit measures:
```

```
##
##               cfi rmsea cfi.delta rmsea.delta
## fit.configural 0.923 0.097      NA      NA
## fit.loadings   0.921 0.093    0.002    0.004
## fit.intercepts 0.882 0.107    0.038    0.015
## fit.means      0.840 0.122    0.042    0.015
```

7.2 Over Time, Within Samples

Invariance can also be tested within a sample, over time. The self-esteem data provide a good example. First, to establish configural invariance, a single-factor model is estimated at each time.

```
SE_Model_1 <- "
SE_1 =~ r_1_1 + r_1_2 + r_1_3 + r_1_4 + r_1_5 + r_1_6 + r_1_7 + r_1_8 + r_1_9 + r_1_10
SE_2 =~ r_2_1 + r_2_2 + r_2_3 + r_2_4 + r_2_5 + r_2_6 + r_2_7 + r_2_8 + r_2_9 + r_2_10
SE_1 ~ 0*1
SE_2 ~ 0*1
"

config <- cfa(SE_Model_1, data = SE, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(config, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 51 iterations
##
##   Number of observations              94
##   Number of missing patterns          1
##
##   Estimator                        ML      Robust
##   Model Fit Test Statistic          545.371  512.917
##   Degrees of freedom                 169      169
##   P-value (Chi-square)               0.000    0.000
##   Scaling correction factor          1.063
##   for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##   Minimum Function Test Statistic    1469.287  1325.006
##   Degrees of freedom                 190      190
##   P-value                           0.000    0.000
##
## User model versus baseline model:
##
##   Comparative Fit Index (CFI)        0.706    0.697
##   Tucker-Lewis Index (TLI)          0.669    0.659
##
##   Robust Comparative Fit Index (CFI)      NA
##   Robust Tucker-Lewis Index (TLI)        NA
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)        -1540.180  -1540.180
##   Loglikelihood unrestricted model (H1)  -1264.562  -1264.562
```

```

##
##   Number of free parameters                61          61
##   Akaike (AIC)                        3202.360      3202.360
##   Bayesian (BIC)                      3357.501      3357.501
##   Sample-size adjusted Bayesian (BIC)    3164.924      3164.924
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                                0.155          0.148
##   90 Percent Confidence Interval        0.140  0.169      0.134  0.162
##   P-value RMSEA <= 0.05                0.000          0.000
##
##   Robust RMSEA                                NA
##   90 Percent Confidence Interval        NA          NA
##
## Standardized Root Mean Square Residual:
##
##   SRMR                                0.084          0.084
##
## Parameter Estimates:
##
##   Information                                Observed
##   Observed information based on              Hessian
##   Standard Errors                          Robust.huber.white
##
## Latent Variables:
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv
## SE_1 =~
##   r_1_1            1.000
##   r_1_2            0.715    0.117    6.134    0.000    0.330
##   r_1_3            1.032    0.144    7.190    0.000    0.477
##   r_1_4            0.913    0.111    8.215    0.000    0.422
##   r_1_5            1.001    0.196    5.105    0.000    0.462
##   r_1_6            1.189    0.163    7.306    0.000    0.549
##   r_1_7            1.069    0.149    7.193    0.000    0.494
##   r_1_8            1.294    0.282    4.593    0.000    0.598
##   r_1_9            1.226    0.249    4.924    0.000    0.566
##   r_1_10           1.374    0.275    4.993    0.000    0.635
## SE_2 =~
##   r_2_1            1.000
##   r_2_2            0.786    0.091    8.607    0.000    0.303
##   r_2_3            1.093    0.169    6.465    0.000    0.421
##   r_2_4            0.881    0.132    6.668    0.000    0.340
##   r_2_5            1.258    0.265    4.747    0.000    0.485
##   r_2_6            1.329    0.221    6.022    0.000    0.512
##   r_2_7            1.229    0.248    4.962    0.000    0.474
##   r_2_8            1.712    0.300    5.710    0.000    0.660
##   r_2_9            1.752    0.331    5.297    0.000    0.676
##   r_2_10           1.641    0.363    4.516    0.000    0.633
## Std.all
##
##   0.782
##   0.702
##   0.786

```



```

##      0.682
##      0.563
##      0.793
##      0.746
##      0.606
##      0.639
##      0.673
##
##      0.769
##      0.686
##      0.689
##      0.489
##      0.632
##      0.750
##      0.740
##      0.690
##      0.715
##      0.660
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
## SE_1 ~~
## SE_2      0.165   0.046   3.614   0.000   0.926
## Std.all
##      0.926
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
## SE_1      0.000
## SE_2      0.000
## .r_1_1    3.670   0.061  59.940   0.000   3.670
## .r_1_2    3.723   0.049  76.318   0.000   3.723
## .r_1_3    3.596   0.063  57.124   0.000   3.596
## .r_1_4    3.372   0.064  52.575   0.000   3.372
## .r_1_5    3.457   0.085  40.622   0.000   3.457
## .r_1_6    3.202   0.072  44.552   0.000   3.202
## .r_1_7    3.202   0.069  46.667   0.000   3.202
## .r_1_8    2.755   0.102  26.951   0.000   2.755
## .r_1_9    2.638   0.092  28.734   0.000   2.638
## .r_1_10   3.064   0.098  31.327   0.000   3.064
## .r_2_1    3.702   0.052  71.163   0.000   3.702
## .r_2_2    3.734   0.046  81.499   0.000   3.734
## .r_2_3    3.564   0.063  56.229   0.000   3.564
## .r_2_4    3.351   0.072  46.493   0.000   3.351
## .r_2_5    3.479   0.080  43.676   0.000   3.479
## .r_2_6    3.181   0.071  44.874   0.000   3.181
## .r_2_7    3.191   0.066  48.058   0.000   3.191
## .r_2_8    2.904   0.099  29.258   0.000   2.904
## .r_2_9    2.979   0.098  30.395   0.000   2.979
## .r_2_10   3.266   0.099  32.861   0.000   3.266
## Std.all
##      0.000
##      0.000

```

```

##      6.215
##      7.914
##      5.923
##      5.452
##      4.212
##      4.620
##      4.839
##      2.795
##      2.980
##      3.248
##      7.379
##      8.451
##      5.831
##      4.821
##      4.529
##      4.653
##      4.983
##      3.034
##      3.152
##      3.408
##
## Variances:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv
##      .r_1_1      0.135     0.031     4.429     0.000     0.135
##      .r_1_2      0.112     0.022     5.147     0.000     0.112
##      .r_1_3      0.141     0.024     5.997     0.000     0.141
##      .r_1_4      0.205     0.030     6.935     0.000     0.205
##      .r_1_5      0.460     0.145     3.173     0.002     0.460
##      .r_1_6      0.179     0.027     6.664     0.000     0.179
##      .r_1_7      0.194     0.027     7.218     0.000     0.194
##      .r_1_8      0.615     0.089     6.888     0.000     0.615
##      .r_1_9      0.463     0.079     5.843     0.000     0.463
##      .r_1_10     0.486     0.104     4.680     0.000     0.486
##      .r_2_1      0.103     0.026     3.896     0.000     0.103
##      .r_2_2      0.103     0.027     3.840     0.000     0.103
##      .r_2_3      0.196     0.034     5.826     0.000     0.196
##      .r_2_4      0.368     0.103     3.559     0.000     0.368
##      .r_2_5      0.355     0.127     2.801     0.005     0.355
##      .r_2_6      0.205     0.031     6.600     0.000     0.205
##      .r_2_7      0.186     0.028     6.691     0.000     0.186
##      .r_2_8      0.480     0.068     7.036     0.000     0.480
##      .r_2_9      0.437     0.069     6.347     0.000     0.437
##      .r_2_10     0.518     0.121     4.293     0.000     0.518
##      SE_1        0.213     0.081     2.632     0.008     1.000
##      SE_2        0.149     0.043     3.418     0.001     1.000
##      Std.all
##      0.388
##      0.507
##      0.383
##      0.535
##      0.683
##      0.372
##      0.443
##      0.633

```

```
##      0.591
##      0.547
##      0.409
##      0.530
##      0.525
##      0.761
##      0.601
##      0.438
##      0.452
##      0.524
##      0.489
##      0.564
##      1.000
##      1.000
##
## R-Square:
##              Estimate
##      r_1_1          0.612
##      r_1_2          0.493
##      r_1_3          0.617
##      r_1_4          0.465
##      r_1_5          0.317
##      r_1_6          0.628
##      r_1_7          0.557
##      r_1_8          0.367
##      r_1_9          0.409
##      r_1_10         0.453
##      r_2_1          0.591
##      r_2_2          0.470
##      r_2_3          0.475
##      r_2_4          0.239
##      r_2_5          0.399
##      r_2_6          0.562
##      r_2_7          0.548
##      r_2_8          0.476
##      r_2_9          0.511
##      r_2_10         0.436
```

Across time, the corresponding items are constrained to have the same loadings.

```
SE_Model_2 <- "
SE_1 =~ a*r_1_1 + b*r_1_2 + c*r_1_3 + d*r_1_4 + e*r_1_5 + f*r_1_6 + g*r_1_7 + h*r_1_8 + i*r_1_9 + j*r_1_10
SE_2 =~ a*r_2_1 + b*r_2_2 + c*r_2_3 + d*r_2_4 + e*r_2_5 + f*r_2_6 + g*r_2_7 + h*r_2_8 + i*r_2_9 + j*r_2_10
SE_1 ~ 0*1
SE_2 ~ 0*1
"
weak <- cfa(SE_Model_2, data = SE, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(weak, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 40 iterations
##
##      Number of observations          94
##      Number of missing patterns      1
```

```

##
##      Estimator                      ML      Robust
##      Model Fit Test Statistic      550.029    525.677
##      Degrees of freedom              178      178
##      P-value (Chi-square)            0.000      0.000
##      Scaling correction factor              1.046
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##      Minimum Function Test Statistic      1469.287    1325.006
##      Degrees of freedom              190      190
##      P-value              0.000      0.000
##
## User model versus baseline model:
##
##      Comparative Fit Index (CFI)            0.709      0.694
##      Tucker-Lewis Index (TLI)              0.690      0.673
##
##      Robust Comparative Fit Index (CFI)              NA
##      Robust Tucker-Lewis Index (TLI)              NA
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)            -1542.534    -1542.534
##      Loglikelihood unrestricted model (H1)      -1264.562    -1264.562
##
##      Number of free parameters              52      52
##      Akaike (AIC)              3189.068    3189.068
##      Bayesian (BIC)              3321.319    3321.319
##      Sample-size adjusted Bayesian (BIC)      3157.155    3157.155
##
## Root Mean Square Error of Approximation:
##
##      RMSEA              0.150      0.145
##      90 Percent Confidence Interval      0.136    0.164      0.131    0.159
##      P-value RMSEA <= 0.05              0.000      0.000
##
##      Robust RMSEA              NA
##      90 Percent Confidence Interval              NA      NA
##
## Standardized Root Mean Square Residual:
##
##      SRMR              0.089      0.089
##
## Parameter Estimates:
##
##      Information              Observed
##      Observed information based on      Hessian
##      Standard Errors              Robust.huber.white
##
## Latent Variables:
##
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##
##      SE_1 =~

```

```

##      r_1_1      (a)      1.000                                0.430
##      r_1_2      (b)      0.753      0.067      11.190      0.000      0.324
##      r_1_3      (c)      1.076      0.124      8.708      0.000      0.463
##      r_1_4      (d)      0.926      0.101      9.171      0.000      0.399
##      r_1_5      (e)      1.134      0.162      7.010      0.000      0.488
##      r_1_6      (f)      1.264      0.143      8.855      0.000      0.544
##      r_1_7      (g)      1.150      0.158      7.284      0.000      0.495
##      r_1_8      (h)      1.513      0.252      5.999      0.000      0.651
##      r_1_9      (i)      1.480      0.233      6.358      0.000      0.637
##      r_1_10     (j)      1.514      0.271      5.578      0.000      0.651
## SE_2 =~
##      r_2_1      (a)      1.000                                0.409
##      r_2_2      (b)      0.753      0.067      11.190      0.000      0.308
##      r_2_3      (c)      1.076      0.124      8.708      0.000      0.440
##      r_2_4      (d)      0.926      0.101      9.171      0.000      0.379
##      r_2_5      (e)      1.134      0.162      7.010      0.000      0.464
##      r_2_6      (f)      1.264      0.143      8.855      0.000      0.517
##      r_2_7      (g)      1.150      0.158      7.284      0.000      0.470
##      r_2_8      (h)      1.513      0.252      5.999      0.000      0.618
##      r_2_9      (i)      1.480      0.233      6.358      0.000      0.605
##      r_2_10     (j)      1.514      0.271      5.578      0.000      0.619
## Std.all
##
##      0.754
##      0.692
##      0.775
##      0.657
##      0.585
##      0.788
##      0.745
##      0.641
##      0.685
##      0.686
##
##      0.789
##      0.694
##      0.704
##      0.531
##      0.613
##      0.752
##      0.737
##      0.664
##      0.670
##      0.650
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
## SE_1 ~~
## SE_2      0.164      0.046      3.585      0.000      0.931
## Std.all
##
##      0.931
##

```

```

## Intercepts:
##           Estimate Std.Err z-value P(>|z|) Std.lv
## SE_1          0.000
## SE_2          0.000
## .r_1_1        3.670    0.061   59.940   0.000   3.670
## .r_1_2        3.723    0.049   76.318   0.000   3.723
## .r_1_3        3.596    0.063   57.124   0.000   3.596
## .r_1_4        3.372    0.064   52.575   0.000   3.372
## .r_1_5        3.457    0.085   40.622   0.000   3.457
## .r_1_6        3.202    0.072   44.552   0.000   3.202
## .r_1_7        3.202    0.069   46.667   0.000   3.202
## .r_1_8        2.755    0.102   26.951   0.000   2.755
## .r_1_9        2.638    0.092   28.734   0.000   2.638
## .r_1_10       3.064    0.098   31.327   0.000   3.064
## .r_2_1        3.702    0.052   71.163   0.000   3.702
## .r_2_2        3.734    0.046   81.499   0.000   3.734
## .r_2_3        3.564    0.063   56.229   0.000   3.564
## .r_2_4        3.351    0.072   46.493   0.000   3.351
## .r_2_5        3.479    0.080   43.676   0.000   3.479
## .r_2_6        3.181    0.071   44.874   0.000   3.181
## .r_2_7        3.191    0.066   48.058   0.000   3.191
## .r_2_8        2.904    0.099   29.258   0.000   2.904
## .r_2_9        2.979    0.098   30.395   0.000   2.979
## .r_2_10       3.266    0.099   32.861   0.000   3.266
## Std.all
## 0.000
## 0.000
## 6.432
## 7.947
## 6.021
## 5.561
## 4.145
## 4.641
## 4.823
## 2.713
## 2.840
## 3.227
## 7.147
## 8.416
## 5.707
## 4.697
## 4.597
## 4.629
## 5.001
## 3.116
## 3.298
## 3.432
##
## Variances:
##           Estimate Std.Err z-value P(>|z|) Std.lv
## .r_1_1        0.141    0.032    4.341   0.000   0.141
## .r_1_2        0.115    0.022    5.289   0.000   0.115
## .r_1_3        0.142    0.024    6.030   0.000   0.142
## .r_1_4        0.209    0.030    6.890   0.000   0.209

```

##	.r_1_5	0.458	0.143	3.194	0.001	0.458
##	.r_1_6	0.180	0.026	6.875	0.000	0.180
##	.r_1_7	0.196	0.027	7.268	0.000	0.196
##	.r_1_8	0.608	0.089	6.852	0.000	0.608
##	.r_1_9	0.458	0.079	5.801	0.000	0.458
##	.r_1_10	0.477	0.101	4.721	0.000	0.477
##	.r_2_1	0.101	0.026	3.937	0.000	0.101
##	.r_2_2	0.102	0.026	3.917	0.000	0.102
##	.r_2_3	0.196	0.034	5.805	0.000	0.196
##	.r_2_4	0.366	0.104	3.526	0.000	0.366
##	.r_2_5	0.358	0.124	2.876	0.004	0.358
##	.r_2_6	0.205	0.031	6.625	0.000	0.205
##	.r_2_7	0.186	0.027	6.788	0.000	0.186
##	.r_2_8	0.486	0.069	7.021	0.000	0.486
##	.r_2_9	0.449	0.069	6.529	0.000	0.449
##	.r_2_10	0.523	0.119	4.391	0.000	0.523
##	SE_1	0.185	0.058	3.202	0.001	1.000
##	SE_2	0.167	0.043	3.925	0.000	1.000
##	Std.all					
##	0.432					
##	0.522					
##	0.399					
##	0.568					
##	0.658					
##	0.379					
##	0.445					
##	0.589					
##	0.530					
##	0.529					
##	0.377					
##	0.518					
##	0.504					
##	0.718					
##	0.624					
##	0.434					
##	0.457					
##	0.560					
##	0.551					
##	0.577					
##	1.000					
##	1.000					
##						
##	R-Square:					
##		Estimate				
##	r_1_1	0.568				
##	r_1_2	0.478				
##	r_1_3	0.601				
##	r_1_4	0.432				
##	r_1_5	0.342				
##	r_1_6	0.621				
##	r_1_7	0.555				
##	r_1_8	0.411				
##	r_1_9	0.470				
##	r_1_10	0.471				

```
##      r_2_1          0.623
##      r_2_2          0.482
##      r_2_3          0.496
##      r_2_4          0.282
##      r_2_5          0.376
##      r_2_6          0.566
##      r_2_7          0.543
##      r_2_8          0.440
##      r_2_9          0.449
##      r_2_10         0.423
```

Next the corresponding items are constrained to have the same intercepts.

```
SE_Model_3 <- "
SE_1 =~ a*r_1_1 + b*r_1_2 + c*r_1_3 + d*r_1_4 + e*r_1_5 + f*r_1_6 + g*r_1_7 + h*r_1_8 + i*r_1_9 + j*r_1_10
SE_2 =~ a*r_2_1 + b*r_2_2 + c*r_2_3 + d*r_2_4 + e*r_2_5 + f*r_2_6 + g*r_2_7 + h*r_2_8 + i*r_2_9 + j*r_2_10
r_1_1 ~ aa*1
r_2_1 ~ aa*1
r_1_2 ~ bb*1
r_2_2 ~ bb*1
r_1_3 ~ cc*1
r_2_3 ~ cc*1
r_1_4 ~ dd*1
r_2_4 ~ dd*1
r_1_5 ~ ee*1
r_2_5 ~ ee*1
r_1_6 ~ ff*1
r_2_6 ~ ff*1
r_1_7 ~ gg*1
r_2_7 ~ gg*1
r_1_8 ~ hh*1
r_2_8 ~ hh*1
r_1_9 ~ ii*1
r_2_9 ~ ii*1
r_1_10 ~ jj*1
r_2_10 ~ jj*1
SE_1 ~ 0*1
SE_2 ~ 0*1
"

strong <- cfa(SE_Model_3, data = SE, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(strong, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 65 iterations
##
##      Number of observations          94
##      Number of missing patterns      1
##
##      Estimator              ML      Robust
##      Model Fit Test Statistic    566.714    550.388
##      Degrees of freedom          188      188
##      P-value (Chi-square)         0.000      0.000
##      Scaling correction factor              1.030
##      for the Yuan-Bentler correction (Mplus variant)
```



```

##
## Model test baseline model:
##
##   Minimum Function Test Statistic           1469.287      1325.006
##   Degrees of freedom                        190          190
##   P-value                                   0.000          0.000
##
## User model versus baseline model:
##
##   Comparative Fit Index (CFI)                0.704          0.681
##   Tucker-Lewis Index (TLI)                  0.701          0.677
##
##   Robust Comparative Fit Index (CFI)                NA
##   Robust Tucker-Lewis Index (TLI)                NA
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)                -1550.966      -1550.966
##   Loglikelihood unrestricted model (H1)          -1264.562      -1264.562
##
##   Number of free parameters                     42          42
##   Akaike (AIC)                                 3185.933      3185.933
##   Bayesian (BIC)                               3292.751      3292.751
##   Sample-size adjusted Bayesian (BIC)           3160.157      3160.157
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                                0.147          0.144
##   90 Percent Confidence Interval          0.133  0.161          0.130  0.158
##   P-value RMSEA <= 0.05                  0.000          0.000
##
##   Robust RMSEA                                NA
##   90 Percent Confidence Interval          NA          NA
##
## Standardized Root Mean Square Residual:
##
##   SRMR                                0.092          0.092
##
## Parameter Estimates:
##
##   Information                                Observed
##   Observed information based on              Hessian
##   Standard Errors                          Robust.huber.white
##
## Latent Variables:
##
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv
## SE_1 =~
##   r_1_1      (a)    1.000
##   r_1_2      (b)    0.753    0.068   11.161    0.000    0.433
##   r_1_3      (c)    1.070    0.123    8.690    0.000    0.463
##   r_1_4      (d)    0.926    0.101    9.212    0.000    0.401
##   r_1_5      (e)    1.128    0.162    6.975    0.000    0.488
##   r_1_6      (f)    1.259    0.143    8.808    0.000    0.545
##   r_1_7      (g)    1.146    0.158    7.259    0.000    0.496

```

```

##      r_1_8      (h)      1.505      0.251      6.006      0.000      0.652
##      r_1_9      (i)      1.469      0.229      6.422      0.000      0.636
##      r_1_10     (j)      1.502      0.268      5.605      0.000      0.650
## SE_2 =~
##      r_2_1      (a)      1.000
##      r_2_2      (b)      0.753      0.068      11.161      0.000      0.309
##      r_2_3      (c)      1.070      0.123      8.690      0.000      0.439
##      r_2_4      (d)      0.926      0.101      9.212      0.000      0.380
##      r_2_5      (e)      1.128      0.162      6.975      0.000      0.463
##      r_2_6      (f)      1.259      0.143      8.808      0.000      0.516
##      r_2_7      (g)      1.146      0.158      7.259      0.000      0.470
##      r_2_8      (h)      1.505      0.251      6.006      0.000      0.617
##      r_2_9      (i)      1.469      0.229      6.422      0.000      0.602
##      r_2_10     (j)      1.502      0.268      5.605      0.000      0.616
## Std.all
##
##      0.758
##      0.696
##      0.775
##      0.661
##      0.585
##      0.790
##      0.747
##      0.639
##      0.672
##      0.681
##
##      0.792
##      0.697
##      0.702
##      0.532
##      0.612
##      0.751
##      0.737
##      0.661
##      0.656
##      0.644
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
## SE_1 ~~
## SE_2      0.164   0.046   3.597   0.000   0.926
## Std.all
##
##      0.926
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv
## .r_1_1      (aa)   3.693   0.050  74.423   0.000   3.693
## .r_2_1      (aa)   3.693   0.050  74.423   0.000   3.693
## .r_1_2      (bb)   3.733   0.041  91.257   0.000   3.733
## .r_2_2      (bb)   3.733   0.041  91.257   0.000   3.733
## .r_1_3      (cc)   3.589   0.056  63.898   0.000   3.589

```

```

##      .r_2_3      (cc)      3.589      0.056      63.898      0.000      3.589
##      .r_1_4      (dd)      3.372      0.056      59.732      0.000      3.372
##      .r_2_4      (dd)      3.372      0.056      59.732      0.000      3.372
##      .r_1_5      (ee)      3.475      0.066      52.456      0.000      3.475
##      .r_2_5      (ee)      3.475      0.066      52.456      0.000      3.475
##      .r_1_6      (ff)      3.200      0.063      50.756      0.000      3.200
##      .r_2_6      (ff)      3.200      0.063      50.756      0.000      3.200
##      .r_1_7      (gg)      3.203      0.058      55.628      0.000      3.203
##      .r_2_7      (gg)      3.203      0.058      55.628      0.000      3.203
##      .r_1_8      (hh)      2.845      0.096      29.731      0.000      2.845
##      .r_2_8      (hh)      2.845      0.096      29.731      0.000      2.845
##      .r_1_9      (ii)      2.818      0.093      30.353      0.000      2.818
##      .r_2_9      (ii)      2.818      0.093      30.353      0.000      2.818
##      .r_1_10     (jj)      3.169      0.093      34.226      0.000      3.169
##      .r_2_10     (jj)      3.169      0.093      34.226      0.000      3.169
##      SE_1                0.000
##      SE_2                0.000
## Std.all
##      6.464
##      7.131
##      7.960
##      8.418
##      6.006
##      5.749
##      5.557
##      4.725
##      4.162
##      4.594
##      4.636
##      4.655
##      4.822
##      5.019
##      2.789
##      3.048
##      2.980
##      3.071
##      3.317
##      3.314
##      0.000
##      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv
##      .r_1_1      0.139   0.032   4.304   0.000   0.139
##      .r_1_2      0.114   0.022   5.274   0.000   0.114
##      .r_1_3      0.143   0.023   6.116   0.000   0.143
##      .r_1_4      0.207   0.030   6.932   0.000   0.207
##      .r_1_5      0.458   0.144   3.182   0.001   0.458
##      .r_1_6      0.179   0.026   6.768   0.000   0.179
##      .r_1_7      0.195   0.027   7.263   0.000   0.195
##      .r_1_8      0.616   0.091   6.784   0.000   0.616
##      .r_1_9      0.490   0.086   5.728   0.000   0.490
##      .r_1_10     0.490   0.105   4.675   0.000   0.490

```

##	.r_2_1	0.100	0.026	3.865	0.000	0.100
##	.r_2_2	0.101	0.026	3.862	0.000	0.101
##	.r_2_3	0.198	0.035	5.639	0.000	0.198
##	.r_2_4	0.365	0.105	3.482	0.000	0.365
##	.r_2_5	0.358	0.125	2.874	0.004	0.358
##	.r_2_6	0.206	0.032	6.490	0.000	0.206
##	.r_2_7	0.186	0.028	6.688	0.000	0.186
##	.r_2_8	0.491	0.070	7.020	0.000	0.491
##	.r_2_9	0.480	0.072	6.703	0.000	0.480
##	.r_2_10	0.535	0.118	4.522	0.000	0.535
##	SE_1	0.187	0.059	3.201	0.001	1.000
##	SE_2	0.168	0.043	3.950	0.000	1.000
##	Std.all					
##	0.426					
##	0.516					
##	0.400					
##	0.563					
##	0.658					
##	0.376					
##	0.442					
##	0.592					
##	0.548					
##	0.537					
##	0.373					
##	0.515					
##	0.507					
##	0.717					
##	0.626					
##	0.436					
##	0.457					
##	0.563					
##	0.569					
##	0.585					
##	1.000					
##	1.000					
##						
##	R-Square:					
##		Estimate				
##	r_1_1	0.574				
##	r_1_2	0.484				
##	r_1_3	0.600				
##	r_1_4	0.437				
##	r_1_5	0.342				
##	r_1_6	0.624				
##	r_1_7	0.558				
##	r_1_8	0.408				
##	r_1_9	0.452				
##	r_1_10	0.463				
##	r_2_1	0.627				
##	r_2_2	0.485				
##	r_2_3	0.493				
##	r_2_4	0.283				
##	r_2_5	0.374				
##	r_2_6	0.564				

##	r_2_7	0.543
##	r_2_8	0.437
##	r_2_9	0.431
##	r_2_10	0.415

Finally, the corresponding items are constrained to have the same loadings, the same intercepts, and their error variances are constrained to be equal as well.

```
SE_Model_4 <- "
SE_1 =~ a*r_1_1 + b*r_1_2 + c*r_1_3 + d*r_1_4 + e*r_1_5 + f*r_1_6 + g*r_1_7 + h*r_1_8 + i*r_1_9 + j*r_1_10
SE_2 =~ a*r_2_1 + b*r_2_2 + c*r_2_3 + d*r_2_4 + e*r_2_5 + f*r_2_6 + g*r_2_7 + h*r_2_8 + i*r_2_9 + j*r_2_10
r_1_1 ~ aa*1
r_2_1 ~ aa*1
r_1_2 ~ bb*1
r_2_2 ~ bb*1
r_1_3 ~ cc*1
r_2_3 ~ cc*1
r_1_4 ~ dd*1
r_2_4 ~ dd*1
r_1_5 ~ ee*1
r_2_5 ~ ee*1
r_1_6 ~ ff*1
r_2_6 ~ ff*1
r_1_7 ~ gg*1
r_2_7 ~ gg*1
r_1_8 ~ hh*1
r_2_8 ~ hh*1
r_1_9 ~ ii*1
r_2_9 ~ ii*1
r_1_10 ~ jj*1
r_2_10 ~ jj*1
r_1_1 ~~ aaa*r_1_1
r_2_1 ~~ aaa*r_2_1
r_1_2 ~~ bbb*r_1_2
r_2_2 ~~ bbb*r_2_2
r_1_3 ~~ ccc*r_1_3
r_2_3 ~~ ccc*r_2_3
r_1_4 ~~ ddd*r_1_4
r_2_4 ~~ ddd*r_2_4
r_1_5 ~~ eee*r_1_5
r_2_5 ~~ eee*r_2_5
r_1_6 ~~ fff*r_1_6
r_2_6 ~~ fff*r_2_6
r_1_7 ~~ ggg*r_1_7
r_2_7 ~~ ggg*r_2_7
r_1_8 ~~ hhh*r_1_8
r_2_8 ~~ hhh*r_2_8
r_1_9 ~~ iii*r_1_9
r_2_9 ~~ iii*r_2_9
r_1_10 ~~ jjj*r_1_10
r_2_10 ~~ jjj*r_2_10
SE_1 ~ 0*1
SE_2 ~ 1
"
```

```

strict_1 <- cfa(SE_Model_4, data = SE, missing = "ML", estimator = "MLR",
  likelihood = "wishart", representation = "LISREL")
summary(strict_1, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)

## lavaan (0.6-1) converged normally after 75 iterations
##
##   Number of observations                94
##   Number of missing patterns             1
##
##   Estimator                            ML      Robust
##   Model Fit Test Statistic              578.691  556.710
##   Degrees of freedom                    197      197
##   P-value (Chi-square)                   0.000      0.000
##   Scaling correction factor              1.039
##   for the Yuan-Bentler correction (Mplus variant)
##
## Model test baseline model:
##
##   Minimum Function Test Statistic        1469.287  1325.006
##   Degrees of freedom                      190      190
##   P-value                                0.000      0.000
##
## User model versus baseline model:
##
##   Comparative Fit Index (CFI)             0.702      0.683
##   Tucker-Lewis Index (TLI)               0.712      0.694
##
##   Robust Comparative Fit Index (CFI)      NA
##   Robust Tucker-Lewis Index (TLI)        NA
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)          -1557.019  -1557.019
##   Loglikelihood unrestricted model (H1)   -1264.562  -1264.562
##
##   Number of free parameters               33      33
##   Akaike (AIC)                           3180.039  3180.039
##   Bayesian (BIC)                         3263.967  3263.967
##   Sample-size adjusted Bayesian (BIC)     3159.786  3159.786
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                                0.144      0.140
##   90 Percent Confidence Interval          0.131  0.158      0.127  0.154
##   P-value RMSEA <= 0.05                  0.000      0.000
##
##   Robust RMSEA                          NA
##   90 Percent Confidence Interval          NA      NA
##
## Standardized Root Mean Square Residual:
##
##   SRMR                                0.092      0.092
##
## Parameter Estimates:

```

```

##
##      Information                      Observed
##      Observed information based on      Hessian
##      Standard Errors                   Robust.huber.white
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##      SE_1 =~
##      r_1_1      (a)    1.000
##      r_1_2      (b)    0.743    0.074   10.105    0.000    0.322
##      r_1_3      (c)    1.061    0.125    8.469    0.000    0.459
##      r_1_4      (d)    0.895    0.103    8.683    0.000    0.388
##      r_1_5      (e)    1.124    0.157    7.167    0.000    0.487
##      r_1_6      (f)    1.251    0.143    8.759    0.000    0.542
##      r_1_7      (g)    1.140    0.150    7.582    0.000    0.494
##      r_1_8      (h)    1.493    0.253    5.901    0.000    0.646
##      r_1_9      (i)    1.485    0.231    6.420    0.000    0.643
##      r_1_10     (j)    1.515    0.268    5.662    0.000    0.656
##      SE_2 =~
##      r_2_1      (a)    1.000
##      r_2_2      (b)    0.743    0.074   10.105    0.000    0.306
##      r_2_3      (c)    1.061    0.125    8.469    0.000    0.436
##      r_2_4      (d)    0.895    0.103    8.683    0.000    0.368
##      r_2_5      (e)    1.124    0.157    7.167    0.000    0.462
##      r_2_6      (f)    1.251    0.143    8.759    0.000    0.515
##      r_2_7      (g)    1.140    0.150    7.582    0.000    0.469
##      r_2_8      (h)    1.493    0.253    5.901    0.000    0.614
##      r_2_9      (i)    1.485    0.231    6.420    0.000    0.611
##      r_2_10     (j)    1.515    0.268    5.662    0.000    0.623
##      Std.all
##
##      0.779
##      0.697
##      0.744
##      0.584
##      0.607
##      0.776
##      0.748
##      0.657
##      0.683
##      0.679
##
##      0.763
##      0.678
##      0.727
##      0.565
##      0.587
##      0.760
##      0.731
##      0.637
##      0.664
##      0.660
##
## Covariances:

```

```

##              Estimate Std.Err z-value P(>|z|) Std.lv
## SE_1 ~~
## SE_2          0.166   0.048   3.472   0.001   0.934
## Std.all
##
## 0.934
##
## Intercepts:
##              Estimate Std.Err z-value P(>|z|) Std.lv
## .r_1_1 (aa)   3.668   0.055  66.963   0.000   3.668
## .r_2_1 (aa)   3.668   0.055  66.963   0.000   3.668
## .r_1_2 (bb)   3.715   0.043  86.143   0.000   3.715
## .r_2_2 (bb)   3.715   0.043  86.143   0.000   3.715
## .r_1_3 (cc)   3.560   0.060  59.096   0.000   3.560
## .r_2_3 (cc)   3.560   0.060  59.096   0.000   3.560
## .r_1_4 (dd)   3.345   0.060  55.309   0.000   3.345
## .r_2_4 (dd)   3.345   0.060  55.309   0.000   3.345
## .r_1_5 (ee)   3.447   0.070  49.204   0.000   3.447
## .r_2_5 (ee)   3.447   0.070  49.204   0.000   3.447
## .r_1_6 (ff)   3.168   0.068  46.456   0.000   3.168
## .r_2_6 (ff)   3.168   0.068  46.456   0.000   3.168
## .r_1_7 (gg)   3.176   0.064  49.556   0.000   3.176
## .r_2_7 (gg)   3.176   0.064  49.556   0.000   3.176
## .r_1_8 (hh)   2.802   0.098  28.696   0.000   2.802
## .r_2_8 (hh)   2.802   0.098  28.696   0.000   2.802
## .r_1_9 (ii)   2.781   0.088  31.438   0.000   2.781
## .r_2_9 (ii)   2.781   0.088  31.438   0.000   2.781
## .r_1_10 (jj)  3.137   0.092  34.025   0.000   3.137
## .r_2_10 (jj)  3.137   0.092  34.025   0.000   3.137
## SE_1      0.000
## SE_2      0.037   0.026   1.412   0.158   0.091
## Std.all
## 6.596
## 6.799
## 8.046
## 8.243
## 5.767
## 5.928
## 5.043
## 5.128
## 4.299
## 4.378
## 4.537
## 4.675
## 4.813
## 4.949
## 2.846
## 2.907
## 2.953
## 3.022
## 3.245
## 3.320
## 0.000

```



```

##      0.091
##
## Variances:
##           Estimate Std.Err  z-value  P(>|z|)  Std.lv
##   .r_1_1  (aaa)    0.122    0.025    4.894    0.000    0.122
##   .r_2_1  (aaa)    0.122    0.025    4.894    0.000    0.122
##   .r_1_2  (bbb)    0.110    0.020    5.540    0.000    0.110
##   .r_2_2  (bbb)    0.110    0.020    5.540    0.000    0.110
##   .r_1_3  (ccc)    0.170    0.023    7.479    0.000    0.170
##   .r_2_3  (ccc)    0.170    0.023    7.479    0.000    0.170
##   .r_1_4  (ddd)    0.290    0.059    4.922    0.000    0.290
##   .r_2_4  (ddd)    0.290    0.059    4.922    0.000    0.290
##   .r_1_5  (eee)    0.406    0.094    4.338    0.000    0.406
##   .r_2_5  (eee)    0.406    0.094    4.338    0.000    0.406
##   .r_1_6  (fff)    0.194    0.021    9.143    0.000    0.194
##   .r_2_6  (fff)    0.194    0.021    9.143    0.000    0.194
##   .r_1_7  (ggg)    0.192    0.021    9.153    0.000    0.192
##   .r_2_7  (ggg)    0.192    0.021    9.153    0.000    0.192
##   .r_1_8  (hhh)    0.551    0.069    8.048    0.000    0.551
##   .r_2_8  (hhh)    0.551    0.069    8.048    0.000    0.551
##   .r_1_9  (iii)    0.474    0.069    6.835    0.000    0.474
##   .r_2_9  (iii)    0.474    0.069    6.835    0.000    0.474
##   .r_1_10 (jjj)    0.504    0.093    5.422    0.000    0.504
##   .r_2_10 (jjj)    0.504    0.093    5.422    0.000    0.504
##   SE_1      0.187    0.061    3.082    0.002    1.000
##   SE_2      0.169    0.044    3.852    0.000    1.000
## Std.all
##   0.394
##   0.418
##   0.514
##   0.540
##   0.447
##   0.472
##   0.659
##   0.681
##   0.632
##   0.655
##   0.398
##   0.423
##   0.440
##   0.465
##   0.569
##   0.594
##   0.534
##   0.559
##   0.539
##   0.565
##   1.000
##   1.000
##
## R-Square:
##           Estimate
##   r_1_1      0.606
##   r_2_1      0.582

```

```

##      r_1_2      0.486
##      r_2_2      0.460
##      r_1_3      0.553
##      r_2_3      0.528
##      r_1_4      0.341
##      r_2_4      0.319
##      r_1_5      0.368
##      r_2_5      0.345
##      r_1_6      0.602
##      r_2_6      0.577
##      r_1_7      0.560
##      r_2_7      0.535
##      r_1_8      0.431
##      r_2_8      0.406
##      r_1_9      0.466
##      r_2_9      0.441
##      r_1_10     0.461
##      r_2_10     0.435

anova(config, weak, strong, strict_1)

## Chi Square Difference Test
##
##           Df  AIC  BIC Chisq Chisq diff Df diff Pr(>Chisq)
## config    169 3202 3358   545
## weak      178 3189 3321   550      4.66      9      0.863
## strong    188 3186 3293   567     16.69     10      0.082
## strict_1  197 3180 3264   579     11.98      9      0.215

anova(config, weak)

## Chi Square Difference Test
##
##           Df  AIC  BIC Chisq Chisq diff Df diff Pr(>Chisq)
## config    169 3202 3358   545
## weak      178 3189 3321   550      4.66      9      0.86

anova(config, strong)

## Chi Square Difference Test
##
##           Df  AIC  BIC Chisq Chisq diff Df diff Pr(>Chisq)
## config    169 3202 3358   545
## strong    188 3186 3293   567     21.3     19      0.32

anova(config, strict_1)

## Chi Square Difference Test
##
##           Df  AIC  BIC Chisq Chisq diff Df diff Pr(>Chisq)
## config    169 3202 3358   545
## strict_1  197 3180 3264   579     33.3     28      0.22

```

Across time, the corresponding items are constrained to have the same loadings, the same intercepts, and their error variances are constrained to be equal as well. In addi-

tion, the latent means are constrained to be equal (and 0).

```
SE_Model_5 <- "  
SE_1 =~ a*r_1_1 + b*r_1_2 + c*r_1_3 + d*r_1_4 + e*r_1_5 + f*r_1_6 + g*r_1_7 + h*r_1_8 + i*r_1_9 + j*r_1_10  
SE_2 =~ a*r_2_1 + b*r_2_2 + c*r_2_3 + d*r_2_4 + e*r_2_5 + f*r_2_6 + g*r_2_7 + h*r_2_8 + i*r_2_9 + j*r_2_10  
r_1_1 ~ aa*1  
r_2_1 ~ aa*1  
r_1_2 ~ bb*1  
r_2_2 ~ bb*1  
r_1_3 ~ cc*1  
r_2_3 ~ cc*1  
r_1_4 ~ dd*1  
r_2_4 ~ dd*1  
r_1_5 ~ ee*1  
r_2_5 ~ ee*1  
r_1_6 ~ ff*1  
r_2_6 ~ ff*1  
r_1_7 ~ gg*1  
r_2_7 ~ gg*1  
r_1_8 ~ hh*1  
r_2_8 ~ hh*1  
r_1_9 ~ ii*1  
r_2_9 ~ ii*1  
r_1_10 ~ jj*1  
r_2_10 ~ jj*1  
r_1_1 ~~ aaa*r_1_1  
r_2_1 ~~ aaa*r_2_1  
r_1_2 ~~ bbb*r_1_2  
r_2_2 ~~ bbb*r_2_2  
r_1_3 ~~ ccc*r_1_3  
r_2_3 ~~ ccc*r_2_3  
r_1_4 ~~ ddd*r_1_4  
r_2_4 ~~ ddd*r_2_4  
r_1_5 ~~ eee*r_1_5  
r_2_5 ~~ eee*r_2_5  
r_1_6 ~~ fff*r_1_6  
r_2_6 ~~ fff*r_2_6  
r_1_7 ~~ ggg*r_1_7  
r_2_7 ~~ ggg*r_2_7  
r_1_8 ~~ hhh*r_1_8  
r_2_8 ~~ hhh*r_2_8  
r_1_9 ~~ iii*r_1_9  
r_2_9 ~~ iii*r_2_9  
r_1_10 ~~ jjj*r_1_10  
r_2_10 ~~ jjj*r_2_10  
SE_1 ~ 0*1  
SE_2 ~ 0*1  
"  
  
strict_2 <- cfa(SE_Model_5, data = SE[, 2:21], missing = "ML", estimator = "MLR",  
  likelihood = "wishart", representation = "LISREL")  
  
## Error in '[.data.frame'(SE, , 2:21): undefined columns selected  
  
summary(strict_2, standardized = TRUE, rsq = TRUE, fit.measures = TRUE)  
  
## lavaan (0.6-1) converged normally after 57 iterations
```

```

##
## Number of observations per group
## Pasteur 156
## Grant-White 145
## Number of missing patterns per group
## Pasteur 1
## Grant-White 1
##
## Estimator ML Robust
## Model Fit Test Statistic 219.863 219.162
## Degrees of freedom 72 72
## P-value (Chi-square) 0.000 0.000
## Scaling correction factor 1.003
## for the Yuan-Bentler correction (Mplus variant)
##
## Chi-square for each group:
##
## Pasteur 110.637 110.284
## Grant-White 109.226 108.878
##
## Model test baseline model:
##
## Minimum Function Test Statistic 951.384 928.080
## Degrees of freedom 72 72
## P-value 0.000 0.000
##
## User model versus baseline model:
##
## Comparative Fit Index (CFI) 0.832 0.828
## Tucker-Lewis Index (TLI) 0.832 0.828
##
## Robust Comparative Fit Index (CFI) NA
## Robust Tucker-Lewis Index (TLI) NA
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -3734.939 -3734.939
## Loglikelihood unrestricted model (H1) -3624.272 -3624.272
##
## Number of free parameters 36 36
## Akaike (AIC) 7541.879 7541.879
## Bayesian (BIC) 7675.335 7675.335
## Sample-size adjusted Bayesian (BIC) 7561.163 7561.163
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.117 0.117
## 90 Percent Confidence Interval 0.100 0.135 0.099 0.135
## P-value RMSEA <= 0.05 0.000 0.000
##
## Robust RMSEA NA
## 90 Percent Confidence Interval NA NA
##
## Standardized Root Mean Square Residual:

```

```

##
##      SRMR                                0.114      0.114
##
## Parameter Estimates:
##
##      Information                                Observed
##      Observed information based on              Hessian
##      Standard Errors                        Robust.huber.white
##
##
## Group 1 [Pasteur]:
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##      Visual =~
##      x1          1.000
##      x2      (.p2.)  0.571   0.144   3.967   0.000   0.521
##      x3      (.p3.)  0.753   0.166   4.535   0.000   0.688
##      Textual =~
##      x4          1.000
##      x5      (.p5.)  1.113   0.066  16.761   0.000   1.106
##      x6      (.p6.)  0.926   0.062  14.952   0.000   0.920
##      Speed =~
##      x7          1.000
##      x8      (.p8.)  1.187   0.130   9.111   0.000   0.668
##      x9      (.p9.)  1.084   0.265   4.092   0.000   0.610
##      Std.all
##
##      0.770
##      0.441
##      0.602
##
##      0.852
##      0.856
##      0.839
##
##      0.533
##      0.693
##      0.630
##
## Covariances:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv
##      Visual ~~
##      Textual      0.385   0.154   2.494   0.013   0.424
##      Speed        0.181   0.069   2.621   0.009   0.351
##      Textual ~~
##      Speed        0.150   0.062   2.444   0.015   0.269
##      Std.all
##
##      0.424
##      0.351
##
##      0.269
##

```

```

## Intercepts:
##           Estimate Std.Err z-value P(>|z|) Std.lv
##   .x1      (.25.)   4.939   0.077  64.457   0.000   4.939
##   .x2      (.26.)   6.090   0.069  88.522   0.000   6.090
##   .x3      (.27.)   2.253   0.071  31.566   0.000   2.253
##   .x4      (.28.)   3.070   0.073  42.211   0.000   3.070
##   .x5      (.29.)   4.350   0.083  52.704   0.000   4.350
##   .x6      (.30.)   2.194   0.070  31.346   0.000   2.194
##   .x7      (.31.)   4.201   0.067  62.871   0.000   4.201
##   .x8      (.32.)   5.545   0.066  84.185   0.000   5.545
##   .x9      (.33.)   5.391   0.061  88.256   0.000   5.391
##   Visual                0.000                0.000
##   Textual                0.000                0.000
##   Speed                  0.000                0.000
## Std.all
##   4.162
##   5.150
##   1.973
##   2.635
##   3.368
##   2.001
##   3.972
##   5.746
##   5.562
##   0.000
##   0.000
##   0.000
##
## Variances:
##           Estimate Std.Err z-value P(>|z|) Std.lv
##   .x1      (.10.)   0.574   0.172   3.333   0.001   0.574
##   .x2      (.11.)   1.126   0.114   9.842   0.000   1.126
##   .x3      (.12.)   0.830   0.110   7.539   0.000   0.830
##   .x4      (.13.)   0.371   0.051   7.348   0.000   0.371
##   .x5      (.14.)   0.446   0.058   7.628   0.000   0.446
##   .x6      (.15.)   0.357   0.048   7.418   0.000   0.357
##   .x7      (.16.)   0.802   0.096   8.374   0.000   0.802
##   .x8      (.17.)   0.485   0.115   4.202   0.000   0.485
##   .x9      (.18.)   0.567   0.119   4.779   0.000   0.567
##   Visual                0.835   0.219   3.816   0.000   1.000
##   Textual                0.986   0.151   6.517   0.000   1.000
##   Speed                  0.317   0.100   3.180   0.001   1.000
## Std.all
##   0.407
##   0.805
##   0.637
##   0.273
##   0.267
##   0.297
##   0.716
##   0.520
##   0.603
##   1.000
##   1.000

```

```

##      1.000
##
## R-Square:
##           Estimate
##      x1          0.593
##      x2          0.195
##      x3          0.363
##      x4          0.727
##      x5          0.733
##      x6          0.703
##      x7          0.284
##      x8          0.480
##      x9          0.397
##
##
## Group 2 [Grant-White]:
##
## Latent Variables:
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv
## Visual =~
##      x1          1.000                0.855
##      x2      (.p2.)  0.571    0.144    3.967    0.000    0.488
##      x3      (.p3.)  0.753    0.166    4.535    0.000    0.643
## Textual =~
##      x4          1.000                0.986
##      x5      (.p5.)  1.113    0.066   16.761    0.000    1.098
##      x6      (.p6.)  0.926    0.062   14.952    0.000    0.913
## Speed =~
##      x7          1.000                0.672
##      x8      (.p8.)  1.187    0.130    9.111    0.000    0.797
##      x9      (.p9.)  1.084    0.265    4.092    0.000    0.728
## Std.all
##
##      0.748
##      0.418
##      0.577
##
##      0.851
##      0.854
##      0.837
##
##      0.600
##      0.753
##      0.695
##
## Covariances:
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv
## Visual ~~
##      Textual      0.417    0.114    3.670    0.000    0.494
##      Speed       0.342    0.093    3.664    0.000    0.596
## Textual ~~
##      Speed       0.197    0.099    1.996    0.046    0.297
## Std.all
##

```

```

##      0.494
##      0.596
##
##      0.297
##
## Intercepts:
##              Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      (.25.)   4.939   0.077  64.457   0.000   4.939
##      .x2      (.26.)   6.090   0.069  88.522   0.000   6.090
##      .x3      (.27.)   2.253   0.071  31.566   0.000   2.253
##      .x4      (.28.)   3.070   0.073  42.211   0.000   3.070
##      .x5      (.29.)   4.350   0.083  52.704   0.000   4.350
##      .x6      (.30.)   2.194   0.070  31.346   0.000   2.194
##      .x7      (.31.)   4.201   0.067  62.871   0.000   4.201
##      .x8      (.32.)   5.545   0.066  84.185   0.000   5.545
##      .x9      (.33.)   5.391   0.061  88.256   0.000   5.391
##      Visual              0.000              0.000
##      Textual              0.000              0.000
##      Speed                0.000              0.000
## Std.all
##      4.324
##      5.214
##      2.020
##      2.649
##      3.386
##      2.011
##      3.754
##      5.240
##      5.148
##      0.000
##      0.000
##      0.000
##
## Variances:
##              Estimate Std.Err z-value P(>|z|) Std.lv
##      .x1      (.10.)   0.574   0.172   3.333   0.001   0.574
##      .x2      (.11.)   1.126   0.114   9.842   0.000   1.126
##      .x3      (.12.)   0.830   0.110   7.539   0.000   0.830
##      .x4      (.13.)   0.371   0.051   7.348   0.000   0.371
##      .x5      (.14.)   0.446   0.058   7.628   0.000   0.446
##      .x6      (.15.)   0.357   0.048   7.418   0.000   0.357
##      .x7      (.16.)   0.802   0.096   8.374   0.000   0.802
##      .x8      (.17.)   0.485   0.115   4.202   0.000   0.485
##      .x9      (.18.)   0.567   0.119   4.779   0.000   0.567
##      Visual              0.731   0.222   3.296   0.001   1.000
##      Textual              0.972   0.162   5.990   0.000   1.000
##      Speed                0.451   0.130   3.479   0.001   1.000
## Std.all
##      0.440
##      0.826
##      0.667
##      0.276
##      0.270
##      0.300

```



```

##      0.640
##      0.433
##      0.517
##      1.000
##      1.000
##      1.000
##
## R-Square:
##              Estimate
##      x1              0.560
##      x2              0.174
##      x3              0.333
##      x4              0.724
##      x5              0.730
##      x6              0.700
##      x7              0.360
##      x8              0.567
##      x9              0.483

anova(config, weak, strong, strict_1, strict_2)

## Warning in lavTestLRT(object = new("lavaan", version = "0.6.1", call = lavaan::lavaan(model
= SE_Model_1, : lavaan WARNING: some models are based on a different set of observed variables

## Chi Square Difference Test
##
##           Df  AIC  BIC Chisq Chisq diff Df diff Pr(>Chisq)
## strict_2   72 7542 7675   220
## config    169 3202 3358   545          326      97      <2e-16
## weak      178 3189 3321   550           5       9      0.863
## strong    188 3186 3293   567          17      10      0.082
## strict_1  197 3180 3264   579          12       9      0.215

anova(config, weak)

## Chi Square Difference Test
##
##           Df  AIC  BIC Chisq Chisq diff Df diff Pr(>Chisq)
## config    169 3202 3358   545
## weak      178 3189 3321   550          4.66      9      0.86

anova(config, strong)

## Chi Square Difference Test
##
##           Df  AIC  BIC Chisq Chisq diff Df diff Pr(>Chisq)
## config    169 3202 3358   545
## strong    188 3186 3293   567          21.3     19      0.32

anova(config, strict_1)

## Chi Square Difference Test
##
##           Df  AIC  BIC Chisq Chisq diff Df diff Pr(>Chisq)
## config    169 3202 3358   545
## strict_1  197 3180 3264   579          33.3     28      0.22

```

```
anova(config, strict_2)

## Warning in lavTestLRT(object = new("lavaan", version = "0.6.1", call = lavaan::lavaan(model
= SE_Model_1, : lavaan WARNING: some models are based on a different set of observed variables

## Chi Square Difference Test
##
##           Df  AIC   BIC Chisq Chisq diff Df diff Pr(>Chisq)
## strict_2   72 7542 7675    220
## config    169 3202 3358    545          326      97    <2e-16
```