

Content Validation and Internal Validation for Thesis

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Contents

This document presents the analysis of judge ratings collected during the content validation phase of a thesis project, followed by an examination of its internal structure. The scale examined is called “Assessment Scale for Group Music Therapy in Substance Use Disorders (MTDQ).”

##Content validity

Load Libraries an preparing data

```
library(readxl)
library(psych)
```

```
## Warning: pacote 'psych' foi compilado no R versão 4.4.3
```

```
# -----
file_path <- "~/Doutorado/content_validity.xlsx"
# -----

raw_data <- read_excel(file_path)

# Remove the first two columns and convert to data frame
df_original <- as.data.frame(raw_data[, -c(1,2)])

# Define custom column names in English
custom_colnames <- c("profession", "city", "gender", "degree",
  "1", "2", "2.1", "2.2", "3", "3.1", "3.2",
  "4", "4.1", "4.2", "5", "5.1", "5.2",
  "6", "6.1", "6.2", "7", "7.1", "7.2",
  "8", "8.1", "8.2", "9", "9.1", "9.2",
  "10", "10.1", "10.2", "11", "11.1", "11.2",
  "12", "13", "14", "15", "16", "17", "18",
  "19", "20", "21", "22")

# Assign names, making them syntactically valid
df_named <- setNames(df_original, make.names(custom_colnames))

# Display first few rows and column names to verify
print(colnames(df_named))
```

```
## [1] "profession" "city"      "gender"      "degree"      "X1"
## [6] "X2"          "X2.1"        "X2.2"        "X3"          "X3.1"
## [11] "X3.2"        "X4"          "X4.1"        "X4.2"        "X5"
## [16] "X5.1"        "X5.2"        "X6"          "X6.1"        "X6.2"
## [21] "X7"          "X7.1"        "X7.2"        "X8"          "X8.1"
## [26] "X8.2"        "X9"          "X9.1"        "X9.2"        "X10"
## [31] "X10.1"       "X10.2"       "X11"         "X11.1"       "X11.2"
## [36] "X12"         "X13"         "X14"         "X15"         "X16"
## [41] "X17"         "X18"         "X19"         "X20"         "X21"
## [46] "X22"
```

```
# Subset X: Items 2.1 to 11.2
```

```
subset_X_indices <- c(7,8,10,11,13,14,16,17,19,20,22,23,25,26,28,29,31,32,34,35)
subset_X <- df_named[, subset_X_indices]
print(colnames(subset_X))
```

```
## [1] "X2.1" "X2.2" "X3.1" "X3.2" "X4.1" "X4.2" "X5.1" "X5.2" "X6.1"
## [10] "X6.2" "X7.1" "X7.2" "X8.1" "X8.2" "X9.1" "X9.2" "X10.1" "X10.2"
## [19] "X11.1" "X11.2"
```

```
# Subset Y: Items 1 to 11.2, plus 14 to 20
```

```
subset_Y_indices <- c(6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27)
subset_Y <- df_named[, subset_Y_indices]
print(colnames(subset_Y))
```

```
## [1] "X2"      "X2.1"    "X2.2"    "X3"      "X3.1"    "X3.2"    "X4"      "X4.1"    "X4.2"
## [10] "X5"      "X5.1"    "X5.2"    "X6"      "X6.1"    "X6.2"    "X7"      "X7.1"    "X7.2"
## [19] "X8"      "X8.1"    "X8.2"    "X9"      "X9.1"    "X9.2"    "X10"     "X10.1"   "X10.2"
## [28] "X11"     "X11.1"   "X11.2"   "X12"     "X15"     "X16"     "X17"     "X18"     "X19"
## [37] "X20"     "X21"
```

```
# Subset Z: Items related to Q12/Q13
```

```
subset_Z_indices <- c(37, 38, 39, 40, 41, 42, 43, 44, 45)
subset_Z <- df_named[, subset_Z_indices]
print(colnames(subset_Z))
```

```
## [1] "X13" "X14" "X15" "X16" "X17" "X18" "X19" "X20" "X21"
```

Analysis of objective Items

This section analyzes the quantitative ratings provided by the judges for the items included in Subset Y (Items 1-11.2, 14-20). We calculate the mean rating for each item and visualize them using a bar plot. A reference line is added at 0.8, which represent a target agreement threshold.

```
# Converting columns that might be character/factor due to Excel import issues
```

```
subset_Y_numeric <- data.frame(lapply(subset_Y, function(x) {
  if(is.character(x) || is.factor(x)) {
    # Attempt conversion, return NA on failure
    suppressWarnings(as.numeric(as.character(x)))
  } else {
    as.numeric(x) # Ensure it's numeric even if already integer
  }
}))
```

```

}))

# Calculate descriptive statistics
desc_stats_Y <- describe(subset_Y_numeric)

## Warning in FUN(newX[, i], ...): nenhum argumento não faltante para min;
## retornando Inf

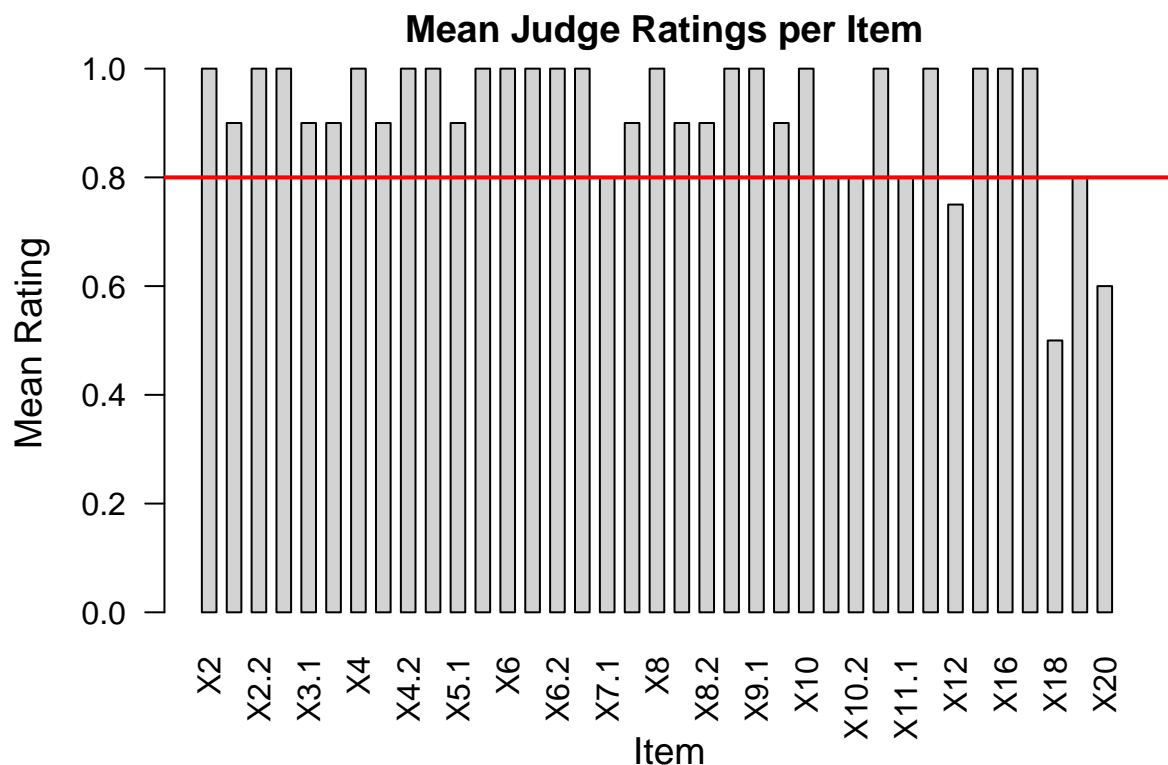
## Warning in FUN(newX[, i], ...): nenhum argumento não faltante para max;
## retornando -Inf

# Define labels for the bar plot
item_labels_y <- colnames(subset_Y) # Use actual column names from the subset

# Create the bar plot
par(cex.lab = 1.2, cex.axis = 1.0, mar = c(6, 4, 2, 2) + 0.1) # Adjust margins if labels overlap
barplot(desc_stats_Y$mean,
        width = 1, space = 0.7,
        names.arg = item_labels_y,
        main = "Mean Judge Ratings per Item ", # Added title
        xlab = "Item",
        ylab = "Mean Rating", # Changed label to reflect the metric
        col = "light grey",
        las = 2)

agreement_threshold = 0.8
abline(h = agreement_threshold, col = "red", lwd = 2)

```



Analysis of Comprehensibility Items (Q12 & Q13)

Items 12 and 13 assessed the comprehensibility of specific sections or instructions and used categorical responses. These were handled separately.

```
# Responses
responses_q12_q13 <- data.frame(
  Q12 = factor(c("Fully comprehensible", "Partially comprehensible", "Partially comprehensible",
    "Fully comprehensible", "Fully comprehensible", "Partially comprehensible",
    "Fully comprehensible", "Partially comprehensible", "Fully comprehensible",
    "Partially comprehensible"), levels = c("Fully comprehensible", "Partially comprehensible")),
  Q13 = factor(c("Fully comprehensible", "Partially comprehensible", "Fully comprehensible",
    "Partially comprehensible", "Fully comprehensible", "Partially comprehensible",
    "Fully comprehensible", "Fully comprehensible", "Fully comprehensible",
    "Fully comprehensible"), levels = c("Fully comprehensible", "Partially comprehensible"))
)

# Create frequency tables
freq_q12 <- table(responses_q12_q13$Q12)
freq_q13 <- table(responses_q12_q13$Q13)

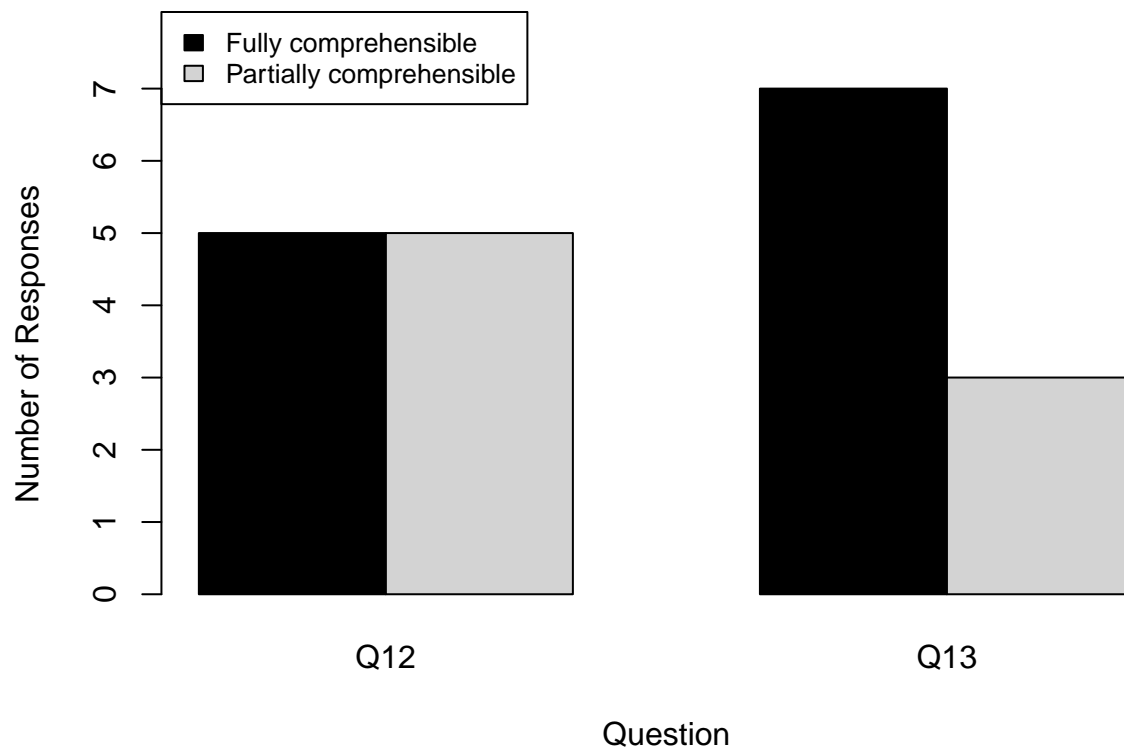
# Create a data frame for plotting
# Handle cases where a category might have zero counts by accessing table elements by name
plot_data_q12_q13 <- data.frame(
  Question = c("Q12", "Q13"),
  `Fully comprehensible` = c(freq_q12["Fully comprehensible"], freq_q13["Fully comprehensible"]),
  `Partially comprehensible` = c(freq_q12["Partially comprehensible"], freq_q13["Partially comprehensible"])
```

```

  check.names = FALSE # Prevent R from changing column names with spaces
)
# Replace NAs (if a category was missing) with 0
plot_data_q12_q13[is.na(plot_data_q12_q13)] <- 0

# Create the grouped bar plot
barplot(height = t(as.matrix(plot_data_q12_q13[, -1])),
        beside = TRUE,
        col = c("black", "light grey"),
        names.arg = plot_data_q12_q13$Question,
        main = "", # Added title
        xlab = "Question",
        ylab = "Number of Responses",
        legend.text = colnames(plot_data_q12_q13[, -1]), # Use translated colnames
        args.legend = list(x = "topleft", inset = c(0, -0.15), cex = 0.8),
        xpd = TRUE
)

```



##Internal Structure Validity

```

library(readxl)
library(readxl)
library(dplyr)

```

##

```

## Anexando pacote: 'dplyr'

## Os seguintes objetos são mascarados por 'package:stats':
##
##     filter, lag

## Os seguintes objetos são mascarados por 'package:base':
##
##     intersect, setdiff, setequal, union

library(psych)
library(MVN)
library(semTools)

## Carregando pacotes exigidos: lavaan

## This is lavaan 0.6-19
## lavaan is FREE software! Please report any bugs.

##
## Anexando pacote: 'lavaan'

## O seguinte objeto é mascarado por 'package:psych':
##
##     cor2cov

##

## #####

## This is semTools 0.5-6

## All users of R (or SEM) are invited to submit functions or ideas for functions.

## #####

##
## Anexando pacote: 'semTools'

## Os seguintes objetos são mascarados por 'package:psych':
##
##     reliability, skew

library(lavaan)
library(semPlot)

# --- Ensure this path is correct ---
data <- read_excel("~/Doutorado/dados.xlsx")
names(data) <- c("id", paste0("i", c(1:20)) , "age", "sex")

```

```
## Warning: The 'value' argument of 'names<-()'' must have the same length as 'x' as of
## tibble 3.0.0.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```

```
## Warning: The 'value' argument of 'names<-()'' can't be empty as of tibble 3.0.0.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```

```
# -----
```

Sociodemographics

```
nrow(data)
```

```
## [1] 202
```

```
# Sample consists of 202 participants
```

```
##Age
```

```
describe(data$age)
```

```
##      vars      n mean      sd median trimmed  mad min max range skew kurtosis   se
## X1      1 190 44.68 12.65      48  45.43 13.34  18  69    51 -0.5    -0.71 0.92
```

```
# Average = 44.7, SD = 12.7, min = 18 e max = 69. Obs: 12 missings
```

```
# Sex
```

```
describe(data$sex) # 77% male participants
```

```
##      vars      n mean      sd median trimmed mad min max range skew kurtosis   se
## X1      1 199  1.77  0.42      2   1.84  0    1  2    1 -1.3    -0.31 0.03
```

Multivariate Normality Analysis

```
mvn_results <- mvn(data[, 2:21], mvnTest = "mardia")
print(mvn_results$multivariateNormality)
```

##	Test	Statistic	p value	Result
## 1	Mardia Skewness	2487.77624922984	8.76630059652458e-48	NO
## 2	Mardia Kurtosis	10.7565859820302	0	NO
## 3	MVN	<NA>	<NA>	NO

```
# Skewness: 2487.78, p < 0.001
```

```
# Kurtosis: 10.76, p < 0.001
```

```
# Data do not show multivariate normal distribution
```

```
# Use WLSMV estimator for CFA
```

Confirmatory Factor Analysis (CFA) Unidimensional Model (General Factor)

```

# Define the unidimensional model
model_uni <- '
general =~ i1 + i2 + i3 + i4 + i5 + i6 + i7 + i8 + i9 + i10 +
i11 + i12 + i13 + i14 + i15 + i16 + i17 + i18 + i19 + i20
'

# Fit the model using WLSMV estimator for ordered data
fit_model_uni <- cfa(model_uni, data = data, ordered = TRUE,
  estimator = "WLSMV", std.lv=TRUE)

# Get fit measures
fitMeasures(fit_model_uni, fit.measures = c("chisq","df","cfi", "rmsea",
  "rmsea.ci.lower", "rmsea.ci.upper"))

```

```

##          chisq          df          cfi          rmsea rmsea.ci.lower
##      283.290      170.000      0.968      0.073      0.058
## rmsea.ci.upper
##      0.087

```

```

# chisq(df=170) = 170.000, CFI = 0.968, RMSEA = 0.073 [0.058, 0.087]
# Model was not rejected

# Get summary with standardized loadings
summary(fit_model_uni, fit.measures = TRUE, standardized = TRUE)

```

```

## lavaan 0.6-19 ended normally after 13 iterations
##
##      Estimator                      DWLS
##      Optimization method          NLMINB
##      Number of model parameters      100
##
##                               Used      Total
##      Number of observations          127      202
##
## Model Test User Model:
##                               Standard      Scaled
##      Test Statistic             283.290      327.569
##      Degrees of freedom             170      170
##      P-value (Chi-square)           0.000      0.000
##      Scaling correction factor              1.164
##      Shift parameter                  84.098
##      simple second-order correction
##
## Model Test Baseline Model:
##
##      Test statistic             3742.984      1405.901
##      Degrees of freedom             190      190
##      P-value                       0.000      0.000
##      Scaling correction factor              2.922
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)           0.968      0.870

```



```

## Tucker-Lewis Index (TLI)                0.964      0.855
##
## Robust Comparative Fit Index (CFI)        0.668
## Robust Tucker-Lewis Index (TLI)          0.629
##
## Root Mean Square Error of Approximation:
##
## RMSEA                0.073      0.086
## 90 Percent confidence interval - lower    0.058      0.072
## 90 Percent confidence interval - upper    0.087      0.100
## P-value H_0: RMSEA <= 0.050              0.009      0.000
## P-value H_0: RMSEA >= 0.080              0.215      0.759
##
## Robust RMSEA                0.133
## 90 Percent confidence interval - lower    0.120
## 90 Percent confidence interval - upper    0.147
## P-value H_0: Robust RMSEA <= 0.050       0.000
## P-value H_0: Robust RMSEA >= 0.080       1.000
##
## Standardized Root Mean Square Residual:
##
## SRMR                0.097      0.097
##
## Parameter Estimates:
##
## Parameterization          Delta
## Standard errors          Robust.sem
## Information              Expected
## Information saturated (h1) model    Unstructured
##
## Latent Variables:
##
##          Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## general =~
## i1          0.407   0.070   5.803   0.000   0.407   0.407
## i2          0.687   0.055  12.515   0.000   0.687   0.687
## i3          0.449   0.070   6.422   0.000   0.449   0.449
## i4          0.272   0.085   3.205   0.001   0.272   0.272
## i5          0.585   0.059   9.907   0.000   0.585   0.585
## i6          0.638   0.056  11.499   0.000   0.638   0.638
## i7          0.537   0.064   8.416   0.000   0.537   0.537
## i8          0.308   0.082   3.749   0.000   0.308   0.308
## i9          0.636   0.052  12.273   0.000   0.636   0.636
## i10         0.519   0.067   7.699   0.000   0.519   0.519
## i11         0.559   0.061   9.105   0.000   0.559   0.559
## i12        -0.019   0.091  -0.207   0.836  -0.019  -0.019
## i13         0.522   0.067   7.794   0.000   0.522   0.522
## i14         0.538   0.080   6.737   0.000   0.538   0.538
## i15         0.716   0.045  15.849   0.000   0.716   0.716
## i16         0.703   0.049  14.335   0.000   0.703   0.703
## i17         0.726   0.046  15.812   0.000   0.726   0.726
## i18         0.771   0.044  17.430   0.000   0.771   0.771
## i19         0.740   0.046  16.025   0.000   0.740   0.740
## i20         0.547   0.067   8.172   0.000   0.547   0.547
##

```

Thresholds:

##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	i1 t1	-1.362	0.159	-8.577	0.000	-1.362	-1.362
##	i1 t2	-0.973	0.133	-7.306	0.000	-0.973	-0.973
##	i1 t3	-0.030	0.112	-0.265	0.791	-0.030	-0.030
##	i1 t4	0.527	0.117	4.483	0.000	0.527	0.527
##	i2 t1	-1.758	0.204	-8.632	0.000	-1.758	-1.758
##	i2 t2	-1.597	0.182	-8.754	0.000	-1.597	-1.597
##	i2 t3	-0.825	0.127	-6.515	0.000	-0.825	-0.825
##	i2 t4	0.109	0.112	0.972	0.331	0.109	0.109
##	i3 t1	-1.469	0.169	-8.712	0.000	-1.469	-1.469
##	i3 t2	-1.184	0.145	-8.149	0.000	-1.184	-1.184
##	i3 t3	-0.270	0.113	-2.384	0.017	-0.270	-0.270
##	i3 t4	0.527	0.117	4.483	0.000	0.527	0.527
##	i4 t1	-1.414	0.163	-8.654	0.000	-1.414	-1.414
##	i4 t2	-0.973	0.133	-7.306	0.000	-0.973	-0.973
##	i4 t3	-0.644	0.120	-5.344	0.000	-0.644	-0.644
##	i4 t4	-0.030	0.112	-0.265	0.791	-0.030	-0.030
##	i5 t1	-1.362	0.159	-8.577	0.000	-1.362	-1.362
##	i5 t2	-1.225	0.148	-8.270	0.000	-1.225	-1.225
##	i5 t3	-0.416	0.115	-3.613	0.000	-0.416	-0.416
##	i5 t4	0.460	0.116	3.962	0.000	0.460	0.460
##	i6 t1	-1.984	0.243	-8.172	0.000	-1.984	-1.984
##	i6 t2	-1.414	0.163	-8.654	0.000	-1.414	-1.414
##	i6 t3	-0.527	0.117	-4.483	0.000	-0.527	-0.527
##	i6 t4	0.290	0.113	2.560	0.010	0.290	0.290
##	i7 t1	-1.672	0.192	-8.720	0.000	-1.672	-1.672
##	i7 t2	-1.146	0.143	-8.021	0.000	-1.146	-1.146
##	i7 t3	-0.229	0.113	-2.031	0.042	-0.229	-0.229
##	i7 t4	0.596	0.119	5.001	0.000	0.596	0.596
##	i8 t1	-1.672	0.192	-8.720	0.000	-1.672	-1.672
##	i8 t2	-1.184	0.145	-8.149	0.000	-1.184	-1.184
##	i8 t3	-0.353	0.114	-3.087	0.002	-0.353	-0.353
##	i8 t4	0.353	0.114	3.087	0.002	0.353	0.353
##	i9 t1	-1.225	0.148	-8.270	0.000	-1.225	-1.225
##	i9 t2	-0.797	0.126	-6.351	0.000	-0.797	-0.797
##	i9 t3	-0.249	0.113	-2.207	0.027	-0.249	-0.249
##	i9 t4	0.573	0.119	4.829	0.000	0.573	0.573
##	i10 t1	-1.758	0.204	-8.632	0.000	-1.758	-1.758
##	i10 t2	-1.314	0.155	-8.486	0.000	-1.314	-1.314
##	i10 t3	-0.353	0.114	-3.087	0.002	-0.353	-0.353
##	i10 t4	0.527	0.117	4.483	0.000	0.527	0.527
##	i11 t1	-1.859	0.220	-8.465	0.000	-1.859	-1.859
##	i11 t2	-1.268	0.151	-8.383	0.000	-1.268	-1.268
##	i11 t3	-0.249	0.113	-2.207	0.027	-0.249	-0.249
##	i11 t4	0.504	0.117	4.310	0.000	0.504	0.504
##	i12 t1	-0.693	0.122	-5.683	0.000	-0.693	-0.693
##	i12 t2	-0.290	0.113	-2.560	0.010	-0.290	-0.290
##	i12 t3	0.049	0.112	0.442	0.659	0.049	0.049
##	i12 t4	0.504	0.117	4.310	0.000	0.504	0.504
##	i13 t1	-1.597	0.182	-8.754	0.000	-1.597	-1.597
##	i13 t2	-1.146	0.143	-8.021	0.000	-1.146	-1.146
##	i13 t3	-0.374	0.115	-3.262	0.001	-0.374	-0.374
##	i13 t4	0.395	0.115	3.438	0.001	0.395	0.395

##	i14 t1	-2.151	0.281	-7.657	0.000	-2.151	-2.151
##	i14 t2	-1.530	0.175	-8.748	0.000	-1.530	-1.530
##	i14 t3	-0.882	0.129	-6.837	0.000	-0.882	-0.882
##	i14 t4	-0.069	0.112	-0.619	0.536	-0.069	-0.069
##	i15 t1	-2.415	0.364	-6.629	0.000	-2.415	-2.415
##	i15 t2	-1.758	0.204	-8.632	0.000	-1.758	-1.758
##	i15 t3	-0.825	0.127	-6.515	0.000	-0.825	-0.825
##	i15 t4	0.270	0.113	2.384	0.017	0.270	0.270
##	i16 t1	-1.984	0.243	-8.172	0.000	-1.984	-1.984
##	i16 t2	-1.314	0.155	-8.486	0.000	-1.314	-1.314
##	i16 t3	-0.596	0.119	-5.001	0.000	-0.596	-0.596
##	i16 t4	0.438	0.116	3.787	0.000	0.438	0.438
##	i17 t1	-1.859	0.220	-8.465	0.000	-1.859	-1.859
##	i17 t2	-1.469	0.169	-8.712	0.000	-1.469	-1.469
##	i17 t3	-0.825	0.127	-6.515	0.000	-0.825	-0.825
##	i17 t4	-0.010	0.112	-0.088	0.930	-0.010	-0.010
##	i18 t1	-1.984	0.243	-8.172	0.000	-1.984	-1.984
##	i18 t2	-1.362	0.159	-8.577	0.000	-1.362	-1.362
##	i18 t3	-0.825	0.127	-6.515	0.000	-0.825	-0.825
##	i18 t4	-0.069	0.112	-0.619	0.536	-0.069	-0.069
##	i19 t1	-1.859	0.220	-8.465	0.000	-1.859	-1.859
##	i19 t2	-1.672	0.192	-8.720	0.000	-1.672	-1.672
##	i19 t3	-0.693	0.122	-5.683	0.000	-0.693	-0.693
##	i19 t4	0.169	0.112	1.502	0.133	0.169	0.169
##	i20 t1	-1.859	0.220	-8.465	0.000	-1.859	-1.859
##	i20 t2	-1.469	0.169	-8.712	0.000	-1.469	-1.469
##	i20 t3	-0.527	0.117	-4.483	0.000	-0.527	-0.527
##	i20 t4	0.290	0.113	2.560	0.010	0.290	0.290
##							
##	Variances:						
##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	.i1	0.835				0.835	0.835
##	.i2	0.528				0.528	0.528
##	.i3	0.799				0.799	0.799
##	.i4	0.926				0.926	0.926
##	.i5	0.658				0.658	0.658
##	.i6	0.592				0.592	0.592
##	.i7	0.712				0.712	0.712
##	.i8	0.905				0.905	0.905
##	.i9	0.596				0.596	0.596
##	.i10	0.731				0.731	0.731
##	.i11	0.688				0.688	0.688
##	.i12	1.000				1.000	1.000
##	.i13	0.728				0.728	0.728
##	.i14	0.711				0.711	0.711
##	.i15	0.488				0.488	0.488
##	.i16	0.506				0.506	0.506
##	.i17	0.473				0.473	0.473
##	.i18	0.406				0.406	0.406
##	.i19	0.452				0.452	0.452
##	.i20	0.701				0.701	0.701
##	general	1.000				1.000	1.000

```

# Descriptive statistics of factor loadings:
loadings_model_uni <- standardizedsolution(fit_model_uni, type = "std.all")
# Selecting based on row indices
print(round(loadings_model_uni[1:20, ] %>% select(est.std) %>%
  summarise(mean = mean(abs(est.std)), sd = sd(abs(est.std)),
    min = min(abs(est.std)), max = max(abs(est.std))), 2))

```

```

##   mean   sd  min  max
## 1 0.54 0.19 0.02 0.77

```

```

# Mean = 0.51, SD = 0.18, min = 0.02 and max = 0.72

```

```

# Reliability
print(round(reliability(fit_model_uni), 2))

```

```

## For constructs with categorical indicators, Zumbo et al.'s (2007) "ordinal alpha" is calculated in a

```

```

##           general
## alpha      0.85
## alpha.ord  0.89
## omega      0.86
## omega2     0.86
## omega3     0.87
## avevar     0.33

```

```

source("comp_reliability.R")
comp_reliability(fit_model_uni)

```

```

## Warning: Use of .data in tidyselect expressions was deprecated in tidyselect 1.2.0.
## i Please use "lhs" instead of '.data$lhs'
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.

```

```

## Warning: Use of .data in tidyselect expressions was deprecated in tidyselect 1.2.0.
## i Please use "est" instead of '.data$est'
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.

```

```

## Warning: Use of .data in tidyselect expressions was deprecated in tidyselect 1.2.0.
## i Please use "op" instead of '.data$op'
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.

```

```

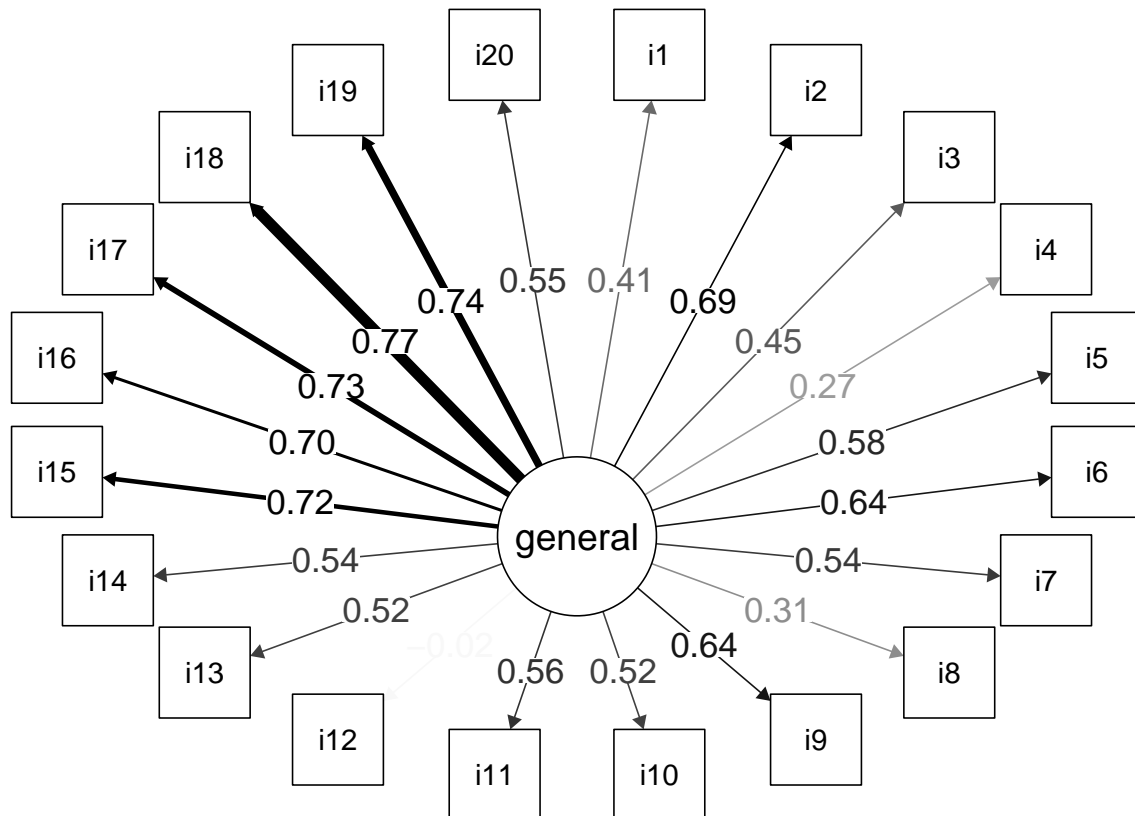
## # A tibble: 1 x 2
##   lhs      composite_reliability_ec
##   <chr>          <dbl>
## 1 general      0.745

```

```
# General: alpha = 0.85, alpha ord. = 0.89, omega McDonald = 0.86, comp. reliability = 0.74
```

```
# Plot the model
```

```
semPaths(fit_model_uni,"std",layout="circle",residuals=FALSE,sizeLat=14,sizeLat2=14,edge.color="black",
  mar=c(2.0, 2.0, 2.0, 2.0),esize=7,curvePivot = FALSE, intercepts=FALSE,thresholds = FALSE,
  nCharNodes=0,sizeMan=8, edge.label.position=0.5
)
```



Two Correlated Factors Model

```
# Define the two-factor model
```

```
model_2f <- '
cog =~ i3 + i4 + i5 + i7 + i8 + i9 + i12 + i14 + i15 + i20
com =~ i1 + i2 + i6 + i10 + i11 + i13 + i16 + i17 + i18 + i19
'
```

```
# Fit the model
```

```
fit_model_2f <- cfa(model_2f, data = data, ordered = TRUE,
  estimator = "WLSMV", std.lv=TRUE)
```

```
# Get fit measures
```

```
fitMeasures(fit_model_2f, fit.measures = c("chisq","df","cfi", "rmsea",
  "rmsea.ci.lower", "rmsea.ci.upper"))
```

##	chisq	df	cfi	rmsea	rmsea.ci.lower
##	267.859	169.000	0.972	0.068	0.052

```
## rmsea.ci.upper
##          0.083
```

```
# chisq(df=169) = 267.859, CFI = 0.972, RMSEA = 0.058 [0.052, 0.083]
# Model was not rejected
```

```
# Get summary
```

```
summary(fit_model_2f, fit.measures = TRUE, standardized = TRUE)
```

```
## lavaan 0.6-19 ended normally after 20 iterations
```

```
##
##      Estimator                      DWLS
##      Optimization method            NLMINB
##      Number of model parameters      101
##
##                               Used      Total
##      Number of observations          127      202
##
## Model Test User Model:
##
##                               Standard      Scaled
##      Test Statistic                267.859    314.998
##      Degrees of freedom              169        169
##      P-value (Chi-square)            0.000        0.000
##      Scaling correction factor                1.155
##      Shift parameter                  83.036
##      simple second-order correction
##
## Model Test Baseline Model:
##
##      Test statistic                3742.984    1405.901
##      Degrees of freedom              190        190
##      P-value                        0.000        0.000
##      Scaling correction factor                2.922
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)                0.972        0.880
##      Tucker-Lewis Index (TLI)                   0.969        0.865
##
##      Robust Comparative Fit Index (CFI)                0.685
##      Robust Tucker-Lewis Index (TLI)                   0.645
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                0.068        0.083
##      90 Percent confidence interval - lower    0.052        0.069
##      90 Percent confidence interval - upper    0.083        0.097
##      P-value H_0: RMSEA <= 0.050            0.031        0.000
##      P-value H_0: RMSEA >= 0.080            0.100        0.638
##
##      Robust RMSEA                0.130
##      90 Percent confidence interval - lower    0.116
##      90 Percent confidence interval - upper    0.144
##      P-value H_0: Robust RMSEA <= 0.050            0.000
```

```

## P-value H_0: Robust RMSEA >= 0.080 1.000
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.094 0.094
##
## Parameter Estimates:
##
## Parameterization Delta
## Standard errors Robust.sem
## Information Expected
## Information saturated (h1) model Unstructured
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## cog =~
## i3 0.476 0.073 6.522 0.000 0.476 0.476
## i4 0.297 0.089 3.352 0.001 0.297 0.297
## i5 0.617 0.062 9.989 0.000 0.617 0.617
## i7 0.571 0.065 8.791 0.000 0.571 0.571
## i8 0.323 0.085 3.794 0.000 0.323 0.323
## i9 0.677 0.053 12.754 0.000 0.677 0.677
## i12 -0.015 0.094 -0.159 0.874 -0.015 -0.015
## i14 0.566 0.082 6.936 0.000 0.566 0.566
## i15 0.762 0.045 16.933 0.000 0.762 0.762
## i20 0.581 0.068 8.575 0.000 0.581 0.581
## com =~
## i1 0.417 0.071 5.857 0.000 0.417 0.417
## i2 0.697 0.056 12.510 0.000 0.697 0.697
## i6 0.651 0.056 11.632 0.000 0.651 0.651
## i10 0.530 0.068 7.794 0.000 0.530 0.530
## i11 0.570 0.061 9.285 0.000 0.570 0.570
## i13 0.533 0.067 7.918 0.000 0.533 0.533
## i16 0.718 0.049 14.774 0.000 0.718 0.718
## i17 0.742 0.045 16.329 0.000 0.742 0.742
## i18 0.782 0.043 17.982 0.000 0.782 0.782
## i19 0.754 0.046 16.542 0.000 0.754 0.754
##
## Covariances:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## cog ~~
## com 0.865 0.038 22.969 0.000 0.865 0.865
##
## Thresholds:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## i3|t1 -1.469 0.169 -8.712 0.000 -1.469 -1.469
## i3|t2 -1.184 0.145 -8.149 0.000 -1.184 -1.184
## i3|t3 -0.270 0.113 -2.384 0.017 -0.270 -0.270
## i3|t4 0.527 0.117 4.483 0.000 0.527 0.527
## i4|t1 -1.414 0.163 -8.654 0.000 -1.414 -1.414
## i4|t2 -0.973 0.133 -7.306 0.000 -0.973 -0.973
## i4|t3 -0.644 0.120 -5.344 0.000 -0.644 -0.644
## i4|t4 -0.030 0.112 -0.265 0.791 -0.030 -0.030
## i5|t1 -1.362 0.159 -8.577 0.000 -1.362 -1.362

```

##	i5 t2	-1.225	0.148	-8.270	0.000	-1.225	-1.225
##	i5 t3	-0.416	0.115	-3.613	0.000	-0.416	-0.416
##	i5 t4	0.460	0.116	3.962	0.000	0.460	0.460
##	i7 t1	-1.672	0.192	-8.720	0.000	-1.672	-1.672
##	i7 t2	-1.146	0.143	-8.021	0.000	-1.146	-1.146
##	i7 t3	-0.229	0.113	-2.031	0.042	-0.229	-0.229
##	i7 t4	0.596	0.119	5.001	0.000	0.596	0.596
##	i8 t1	-1.672	0.192	-8.720	0.000	-1.672	-1.672
##	i8 t2	-1.184	0.145	-8.149	0.000	-1.184	-1.184
##	i8 t3	-0.353	0.114	-3.087	0.002	-0.353	-0.353
##	i8 t4	0.353	0.114	3.087	0.002	0.353	0.353
##	i9 t1	-1.225	0.148	-8.270	0.000	-1.225	-1.225
##	i9 t2	-0.797	0.126	-6.351	0.000	-0.797	-0.797
##	i9 t3	-0.249	0.113	-2.207	0.027	-0.249	-0.249
##	i9 t4	0.573	0.119	4.829	0.000	0.573	0.573
##	i12 t1	-0.693	0.122	-5.683	0.000	-0.693	-0.693
##	i12 t2	-0.290	0.113	-2.560	0.010	-0.290	-0.290
##	i12 t3	0.049	0.112	0.442	0.659	0.049	0.049
##	i12 t4	0.504	0.117	4.310	0.000	0.504	0.504
##	i14 t1	-2.151	0.281	-7.657	0.000	-2.151	-2.151
##	i14 t2	-1.530	0.175	-8.748	0.000	-1.530	-1.530
##	i14 t3	-0.882	0.129	-6.837	0.000	-0.882	-0.882
##	i14 t4	-0.069	0.112	-0.619	0.536	-0.069	-0.069
##	i15 t1	-2.415	0.364	-6.629	0.000	-2.415	-2.415
##	i15 t2	-1.758	0.204	-8.632	0.000	-1.758	-1.758
##	i15 t3	-0.825	0.127	-6.515	0.000	-0.825	-0.825
##	i15 t4	0.270	0.113	2.384	0.017	0.270	0.270
##	i20 t1	-1.859	0.220	-8.465	0.000	-1.859	-1.859
##	i20 t2	-1.469	0.169	-8.712	0.000	-1.469	-1.469
##	i20 t3	-0.527	0.117	-4.483	0.000	-0.527	-0.527
##	i20 t4	0.290	0.113	2.560	0.010	0.290	0.290
##	i1 t1	-1.362	0.159	-8.577	0.000	-1.362	-1.362
##	i1 t2	-0.973	0.133	-7.306	0.000	-0.973	-0.973
##	i1 t3	-0.030	0.112	-0.265	0.791	-0.030	-0.030
##	i1 t4	0.527	0.117	4.483	0.000	0.527	0.527
##	i2 t1	-1.758	0.204	-8.632	0.000	-1.758	-1.758
##	i2 t2	-1.597	0.182	-8.754	0.000	-1.597	-1.597
##	i2 t3	-0.825	0.127	-6.515	0.000	-0.825	-0.825
##	i2 t4	0.109	0.112	0.972	0.331	0.109	0.109
##	i6 t1	-1.984	0.243	-8.172	0.000	-1.984	-1.984
##	i6 t2	-1.414	0.163	-8.654	0.000	-1.414	-1.414
##	i6 t3	-0.527	0.117	-4.483	0.000	-0.527	-0.527
##	i6 t4	0.290	0.113	2.560	0.010	0.290	0.290
##	i10 t1	-1.758	0.204	-8.632	0.000	-1.758	-1.758
##	i10 t2	-1.314	0.155	-8.486	0.000	-1.314	-1.314
##	i10 t3	-0.353	0.114	-3.087	0.002	-0.353	-0.353
##	i10 t4	0.527	0.117	4.483	0.000	0.527	0.527
##	i11 t1	-1.859	0.220	-8.465	0.000	-1.859	-1.859
##	i11 t2	-1.268	0.151	-8.383	0.000	-1.268	-1.268
##	i11 t3	-0.249	0.113	-2.207	0.027	-0.249	-0.249
##	i11 t4	0.504	0.117	4.310	0.000	0.504	0.504
##	i13 t1	-1.597	0.182	-8.754	0.000	-1.597	-1.597
##	i13 t2	-1.146	0.143	-8.021	0.000	-1.146	-1.146
##	i13 t3	-0.374	0.115	-3.262	0.001	-0.374	-0.374


```
##      i13|t4      0.395    0.115    3.438    0.001    0.395    0.395
##      i16|t1     -1.984    0.243   -8.172    0.000   -1.984   -1.984
##      i16|t2     -1.314    0.155   -8.486    0.000   -1.314   -1.314
##      i16|t3     -0.596    0.119   -5.001    0.000   -0.596   -0.596
##      i16|t4      0.438    0.116    3.787    0.000    0.438    0.438
##      i17|t1     -1.859    0.220   -8.465    0.000   -1.859   -1.859
##      i17|t2     -1.469    0.169   -8.712    0.000   -1.469   -1.469
##      i17|t3     -0.825    0.127   -6.515    0.000   -0.825   -0.825
##      i17|t4     -0.010    0.112   -0.088    0.930   -0.010   -0.010
##      i18|t1     -1.984    0.243   -8.172    0.000   -1.984   -1.984
##      i18|t2     -1.362    0.159   -8.577    0.000   -1.362   -1.362
##      i18|t3     -0.825    0.127   -6.515    0.000   -0.825   -0.825
##      i18|t4     -0.069    0.112   -0.619    0.536   -0.069   -0.069
##      i19|t1     -1.859    0.220   -8.465    0.000   -1.859   -1.859
##      i19|t2     -1.672    0.192   -8.720    0.000   -1.672   -1.672
##      i19|t3     -0.693    0.122   -5.683    0.000   -0.693   -0.693
##      i19|t4      0.169    0.112    1.502    0.133    0.169    0.169
```

```
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .i3      0.773      0.773      0.773
##      .i4      0.912      0.912      0.912
##      .i5      0.619      0.619      0.619
##      .i7      0.674      0.674      0.674
##      .i8      0.896      0.896      0.896
##      .i9      0.541      0.541      0.541
##      .i12     1.000      1.000      1.000
##      .i14     0.680      0.680      0.680
##      .i15     0.419      0.419      0.419
##      .i20     0.663      0.663      0.663
##      .i1      0.826      0.826      0.826
##      .i2      0.514      0.514      0.514
##      .i6      0.576      0.576      0.576
##      .i10     0.720      0.720      0.720
##      .i11     0.675      0.675      0.675
##      .i13     0.716      0.716      0.716
##      .i16     0.485      0.485      0.485
##      .i17     0.450      0.450      0.450
##      .i18     0.388      0.388      0.388
##      .i19     0.432      0.432      0.432
##      cog      1.000      1.000      1.000
##      com      1.000      1.000      1.000
```

```
# Descriptive statistics of factor loadings:
loadings_fit_model_2f <- standardizedsolution(fit_model_2f, type = "std.all")
# Selecting based on row indices
print("Cog Factor Loadings Summary:")
```

```
## [1] "Cog Factor Loadings Summary:"
```

```
print(round(loadings_fit_model_2f[1:10, ] %>% select(est.std) %>%
  summarise(mean = mean(abs(est.std)), sd = sd(abs(est.std)),
    min = min(abs(est.std)), max = max(abs(est.std))), 2))
```

```
##      mean    sd  min  max
## 1 0.49 0.22 0.01 0.76
```

```
# Loadings for cog: Mean = 0.49, SD = 0.16, min = 0.3 and max = 0.68
```

```
# Selecting based on row indices
print("Com Factor Loadings Summary:")
```

```
## [1] "Com Factor Loadings Summary:"
```

```
print(round(loadings_fit_model_2f[11:20, ] %>% select(est.std) %>%
  summarise(mean = mean(abs(est.std)), sd = sd(abs(est.std)),
    min = min(abs(est.std)), max = max(abs(est.std))), 2))
```

```
##      mean    sd  min  max
## 1 0.64 0.12 0.42 0.78
```

```
# Loadings for com: Mean = 0.53, SD = 0.22, min = 0.01 and max = 0.76
```

```
# Reliability
print(round(reliability(fit_model_2f), 2))
```

```
## For constructs with categorical indicators, Zumbo et al.'s (2007) "ordinal alpha" is calculated in a
```

```
##           cog  com
## alpha      0.72 0.83
## alpha.ord  0.77 0.87
## omega      0.69 0.84
## omega2     0.69 0.84
## omega3     0.65 0.85
## avevar     0.28 0.42
```

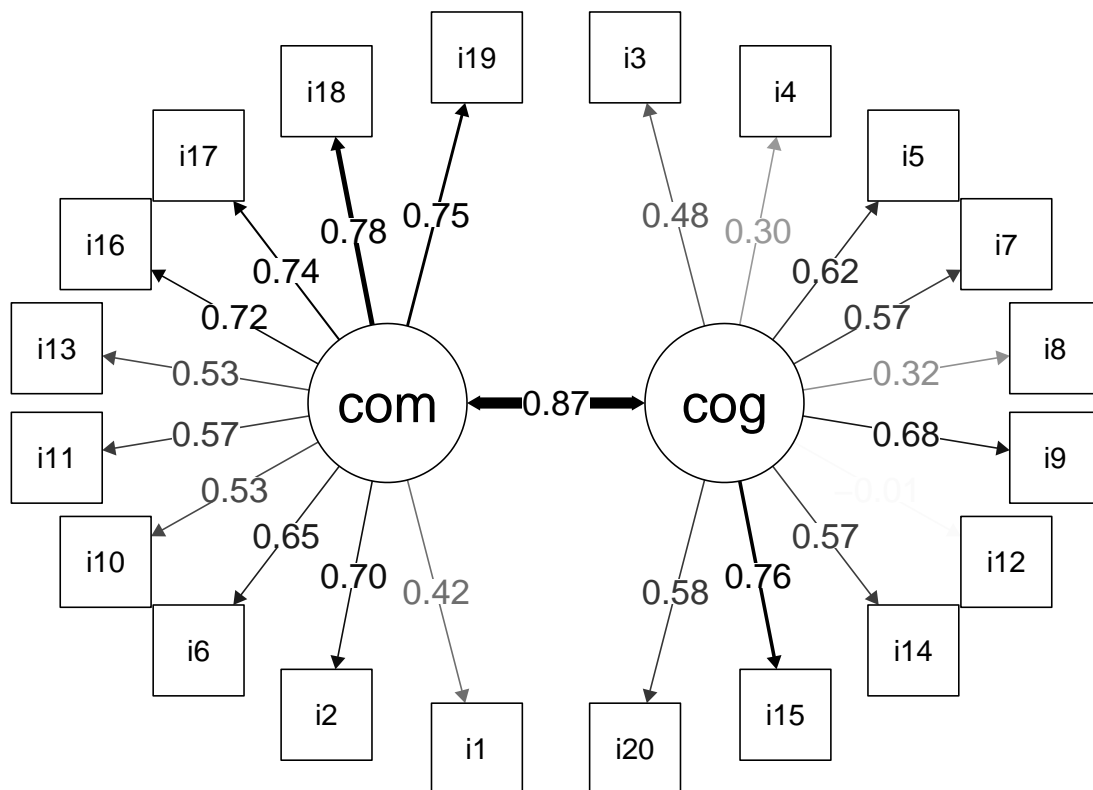
```
comp_reliability(fit_model_2f)
```

```
## # A tibble: 2 x 2
##   lhs      composite_reliability_ec
##   <chr>                <dbl>
## 1 cog                  0.523
## 2 com                  0.702
```

```
# Cog: alpha = 0.72, alpha ord. = 0.77, omega McDonald = 0.69, comp. reliability = 0.52
# Com: alpha = 0.83, alpha ord. = 0.87, omega McDonald = 0.84, comp. reliability = 0.70
```

```
# Plot the model
```

```
semPaths(fit_model_2f,"std",layout="circle",residuals=FALSE,sizeLat=14,sizeLat2=14,edge.color="black",e
  mar=c(2.5, 2.5, 2.5, 2.5),esize=7,curvePivot = TRUE, intercepts=FALSE,thresholds = FALSE,
  nCharNodes=0,sizeMan=8, edge.label.position=0.5
)
```



Bifactor Model (Orthogonal Factors)

```
# Define the bifactor model
model_bifactor <- '
cog =~ i3 + i4 + i5 + i7 + i8 + i9 + i12 + i14 + i15 + i20
com =~ i1 + i2 + i6 + i10 + i11 + i13 + i16 + i17 + i18 + i19

general =~ i1 + i2 + i3 + i4 + i5 + i6 + i7 + i8 + i9 + i10 +
          i11 + i12 + i13 + i14 + i15 + i16 + i17 + i18 + i19 + i20
'

# Fit the bifactor model with orthogonal factors
fit_model_bifactor <- cfa(model_bifactor, data = data, ordered = TRUE, orthogonal = TRUE,
                           estimator = "WLSMV", std.lv=TRUE)

fitMeasures(fit_model_bifactor, fit.measures = c("chisq","df","cfi", "rmsea",
                                                  "rmsea.ci.lower", "rmsea.ci.upper"))
```

```
##          chisq          df          cfi          rmsea rmsea.ci.lower
##          188.811        150.000        0.989        0.045          0.020
## rmsea.ci.upper
##          0.064
```

```
# chisq(df=150) = 188.811, CFI = 0.989, RMSEA = 0.045 [0.020, 0.064]
# Model was not rejected
```

```
# Get summary
```

```
summary(fit_model_bifactor, fit.measures = TRUE, standardized = TRUE)
```

```
## lavaan 0.6-19 ended normally after 42 iterations
```

```
##
```

##	Estimator	DWLS	
##	Optimization method	NLMINB	
##	Number of model parameters	120	
##			
##		Used	Total
##	Number of observations	127	202

```
##
```

```
## Model Test User Model:
```

##		Standard	Scaled
##	Test Statistic	188.811	264.567
##	Degrees of freedom	150	150
##	P-value (Chi-square)	0.017	0.000
##	Scaling correction factor		0.956
##	Shift parameter		66.963
##	simple second-order correction		

```
##
```

```
## Model Test Baseline Model:
```

##			
##	Test statistic	3742.984	1405.901
##	Degrees of freedom	190	190
##	P-value	0.000	0.000
##	Scaling correction factor		2.922

```
##
```

```
## User Model versus Baseline Model:
```

##			
##	Comparative Fit Index (CFI)	0.989	0.906
##	Tucker-Lewis Index (TLI)	0.986	0.881
##			
##	Robust Comparative Fit Index (CFI)		0.751
##	Robust Tucker-Lewis Index (TLI)		0.684

```
##
```

```
## Root Mean Square Error of Approximation:
```

```
##
```

##	RMSEA	0.045	0.078
##	90 Percent confidence interval - lower	0.020	0.062
##	90 Percent confidence interval - upper	0.064	0.093
##	P-value H_0: RMSEA <= 0.050	0.637	0.003
##	P-value H_0: RMSEA >= 0.080	0.001	0.421

```
##
```

##	Robust RMSEA		0.123
##	90 Percent confidence interval - lower		0.108
##	90 Percent confidence interval - upper		0.138
##	P-value H_0: Robust RMSEA <= 0.050		0.000
##	P-value H_0: Robust RMSEA >= 0.080		1.000

```
##
```

```

## Standardized Root Mean Square Residual:
##
##   SRMR                                0.078      0.078
##
## Parameter Estimates:
##
##   Parameterization                    Delta
##   Standard errors                    Robust.sem
##   Information                        Expected
##   Information saturated (h1) model    Unstructured
##
## Latent Variables:
##
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
## cog =~
##   i3              0.133    0.103    1.298    0.194    0.133    0.133
##   i4              0.442    0.122    3.634    0.000    0.442    0.442
##   i5             -0.280    0.099   -2.825    0.005   -0.280   -0.280
##   i7              0.183    0.092    1.978    0.048    0.183    0.183
##   i8              0.584    0.101    5.801    0.000    0.584    0.584
##   i9              0.014    0.106    0.135    0.893    0.014    0.014
##   i12             0.467    0.125    3.738    0.000    0.467    0.467
##   i14             0.424    0.106    3.988    0.000    0.424    0.424
##   i15             0.012    0.100    0.116    0.907    0.012    0.012
##   i20             0.143    0.113    1.263    0.207    0.143    0.143
## com =~
##   i1              0.236    0.121    1.944    0.052    0.236    0.236
##   i2             -0.010    0.113   -0.092    0.927   -0.010   -0.010
##   i6              0.198    0.110    1.800    0.072    0.198    0.198
##   i10             0.284    0.093    3.049    0.002    0.284    0.284
##   i11             0.274    0.112    2.442    0.015    0.274    0.274
##   i13             0.475    0.088    5.392    0.000    0.475    0.475
##   i16             0.363    0.077    4.715    0.000    0.363    0.363
##   i17             0.675    0.067   10.144    0.000    0.675    0.675
##   i18             0.137    0.089    1.548    0.122    0.137    0.137
##   i19             0.403    0.082    4.883    0.000    0.403    0.403
## general =~
##   i1              0.357    0.078    4.545    0.000    0.357    0.357
##   i2              0.724    0.056   12.868    0.000    0.724    0.724
##   i3              0.464    0.071    6.527    0.000    0.464    0.464
##   i4              0.262    0.090    2.914    0.004    0.262    0.262
##   i5              0.644    0.058   11.024    0.000    0.644    0.644
##   i6              0.608    0.064    9.567    0.000    0.608    0.608
##   i7              0.553    0.066    8.375    0.000    0.553    0.553
##   i8              0.280    0.090    3.101    0.002    0.280    0.280
##   i9              0.670    0.053   12.528    0.000    0.670    0.670
##   i10             0.454    0.073    6.220    0.000    0.454    0.454
##   i11             0.500    0.072    6.895    0.000    0.500    0.500
##   i12            -0.061    0.098   -0.621    0.534   -0.061   -0.061
##   i13             0.407    0.077    5.256    0.000    0.407    0.407
##   i14             0.522    0.089    5.868    0.000    0.522    0.522
##   i15             0.760    0.046   16.568    0.000    0.760    0.760
##   i16             0.627    0.059   10.543    0.000    0.627    0.627
##   i17             0.559    0.066    8.470    0.000    0.559    0.559
##   i18             0.767    0.053   14.402    0.000    0.767    0.767

```

##	i19	0.640	0.063	10.102	0.000	0.640	0.640
##	i20	0.566	0.069	8.228	0.000	0.566	0.566
##							
##	Covariances:						
##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	cog ~~						
##	com	0.000				0.000	0.000
##	general	0.000				0.000	0.000
##	com ~~						
##	general	0.000				0.000	0.000
##							
##	Thresholds:						
##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	i3 t1	-1.469	0.169	-8.712	0.000	-1.469	-1.469
##	i3 t2	-1.184	0.145	-8.149	0.000	-1.184	-1.184
##	i3 t3	-0.270	0.113	-2.384	0.017	-0.270	-0.270
##	i3 t4	0.527	0.117	4.483	0.000	0.527	0.527
##	i4 t1	-1.414	0.163	-8.654	0.000	-1.414	-1.414
##	i4 t2	-0.973	0.133	-7.306	0.000	-0.973	-0.973
##	i4 t3	-0.644	0.120	-5.344	0.000	-0.644	-0.644
##	i4 t4	-0.030	0.112	-0.265	0.791	-0.030	-0.030
##	i5 t1	-1.362	0.159	-8.577	0.000	-1.362	-1.362
##	i5 t2	-1.225	0.148	-8.270	0.000	-1.225	-1.225
##	i5 t3	-0.416	0.115	-3.613	0.000	-0.416	-0.416
##	i5 t4	0.460	0.116	3.962	0.000	0.460	0.460
##	i7 t1	-1.672	0.192	-8.720	0.000	-1.672	-1.672
##	i7 t2	-1.146	0.143	-8.021	0.000	-1.146	-1.146
##	i7 t3	-0.229	0.113	-2.031	0.042	-0.229	-0.229
##	i7 t4	0.596	0.119	5.001	0.000	0.596	0.596
##	i8 t1	-1.672	0.192	-8.720	0.000	-1.672	-1.672
##	i8 t2	-1.184	0.145	-8.149	0.000	-1.184	-1.184
##	i8 t3	-0.353	0.114	-3.087	0.002	-0.353	-0.353
##	i8 t4	0.353	0.114	3.087	0.002	0.353	0.353
##	i9 t1	-1.225	0.148	-8.270	0.000	-1.225	-1.225
##	i9 t2	-0.797	0.126	-6.351	0.000	-0.797	-0.797
##	i9 t3	-0.249	0.113	-2.207	0.027	-0.249	-0.249
##	i9 t4	0.573	0.119	4.829	0.000	0.573	0.573
##	i12 t1	-0.693	0.122	-5.683	0.000	-0.693	-0.693
##	i12 t2	-0.290	0.113	-2.560	0.010	-0.290	-0.290
##	i12 t3	0.049	0.112	0.442	0.659	0.049	0.049
##	i12 t4	0.504	0.117	4.310	0.000	0.504	0.504
##	i14 t1	-2.151	0.281	-7.657	0.000	-2.151	-2.151
##	i14 t2	-1.530	0.175	-8.748	0.000	-1.530	-1.530
##	i14 t3	-0.882	0.129	-6.837	0.000	-0.882	-0.882
##	i14 t4	-0.069	0.112	-0.619	0.536	-0.069	-0.069
##	i15 t1	-2.415	0.364	-6.629	0.000	-2.415	-2.415
##	i15 t2	-1.758	0.204	-8.632	0.000	-1.758	-1.758
##	i15 t3	-0.825	0.127	-6.515	0.000	-0.825	-0.825
##	i15 t4	0.270	0.113	2.384	0.017	0.270	0.270
##	i20 t1	-1.859	0.220	-8.465	0.000	-1.859	-1.859
##	i20 t2	-1.469	0.169	-8.712	0.000	-1.469	-1.469
##	i20 t3	-0.527	0.117	-4.483	0.000	-0.527	-0.527
##	i20 t4	0.290	0.113	2.560	0.010	0.290	0.290
##	i1 t1	-1.362	0.159	-8.577	0.000	-1.362	-1.362

##	i1 t2	-0.973	0.133	-7.306	0.000	-0.973	-0.973
##	i1 t3	-0.030	0.112	-0.265	0.791	-0.030	-0.030
##	i1 t4	0.527	0.117	4.483	0.000	0.527	0.527
##	i2 t1	-1.758	0.204	-8.632	0.000	-1.758	-1.758
##	i2 t2	-1.597	0.182	-8.754	0.000	-1.597	-1.597
##	i2 t3	-0.825	0.127	-6.515	0.000	-0.825	-0.825
##	i2 t4	0.109	0.112	0.972	0.331	0.109	0.109
##	i6 t1	-1.984	0.243	-8.172	0.000	-1.984	-1.984
##	i6 t2	-1.414	0.163	-8.654	0.000	-1.414	-1.414
##	i6 t3	-0.527	0.117	-4.483	0.000	-0.527	-0.527
##	i6 t4	0.290	0.113	2.560	0.010	0.290	0.290
##	i10 t1	-1.758	0.204	-8.632	0.000	-1.758	-1.758
##	i10 t2	-1.314	0.155	-8.486	0.000	-1.314	-1.314
##	i10 t3	-0.353	0.114	-3.087	0.002	-0.353	-0.353
##	i10 t4	0.527	0.117	4.483	0.000	0.527	0.527
##	i11 t1	-1.859	0.220	-8.465	0.000	-1.859	-1.859
##	i11 t2	-1.268	0.151	-8.383	0.000	-1.268	-1.268
##	i11 t3	-0.249	0.113	-2.207	0.027	-0.249	-0.249
##	i11 t4	0.504	0.117	4.310	0.000	0.504	0.504
##	i13 t1	-1.597	0.182	-8.754	0.000	-1.597	-1.597
##	i13 t2	-1.146	0.143	-8.021	0.000	-1.146	-1.146
##	i13 t3	-0.374	0.115	-3.262	0.001	-0.374	-0.374
##	i13 t4	0.395	0.115	3.438	0.001	0.395	0.395
##	i16 t1	-1.984	0.243	-8.172	0.000	-1.984	-1.984
##	i16 t2	-1.314	0.155	-8.486	0.000	-1.314	-1.314
##	i16 t3	-0.596	0.119	-5.001	0.000	-0.596	-0.596
##	i16 t4	0.438	0.116	3.787	0.000	0.438	0.438
##	i17 t1	-1.859	0.220	-8.465	0.000	-1.859	-1.859
##	i17 t2	-1.469	0.169	-8.712	0.000	-1.469	-1.469
##	i17 t3	-0.825	0.127	-6.515	0.000	-0.825	-0.825
##	i17 t4	-0.010	0.112	-0.088	0.930	-0.010	-0.010
##	i18 t1	-1.984	0.243	-8.172	0.000	-1.984	-1.984
##	i18 t2	-1.362	0.159	-8.577	0.000	-1.362	-1.362
##	i18 t3	-0.825	0.127	-6.515	0.000	-0.825	-0.825
##	i18 t4	-0.069	0.112	-0.619	0.536	-0.069	-0.069
##	i19 t1	-1.859	0.220	-8.465	0.000	-1.859	-1.859
##	i19 t2	-1.672	0.192	-8.720	0.000	-1.672	-1.672
##	i19 t3	-0.693	0.122	-5.683	0.000	-0.693	-0.693
##	i19 t4	0.169	0.112	1.502	0.133	0.169	0.169

##

Variances:

##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	.i3	0.767				0.767	0.767
##	.i4	0.736				0.736	0.736
##	.i5	0.506				0.506	0.506
##	.i7	0.661				0.661	0.661
##	.i8	0.581				0.581	0.581
##	.i9	0.551				0.551	0.551
##	.i12	0.778				0.778	0.778
##	.i14	0.547				0.547	0.547
##	.i15	0.422				0.422	0.422
##	.i20	0.659				0.659	0.659
##	.i1	0.817				0.817	0.817
##	.i2	0.476				0.476	0.476

```
##      .i6      0.591      0.591      0.591
##      .i10     0.713      0.713      0.713
##      .i11     0.675      0.675      0.675
##      .i13     0.609      0.609      0.609
##      .i16     0.475      0.475      0.475
##      .i17     0.232      0.232      0.232
##      .i18     0.392      0.392      0.392
##      .i19     0.429      0.429      0.429
##      cog      1.000      1.000      1.000
##      com      1.000      1.000      1.000
##      general  1.000      1.000      1.000
```

```
# Descriptive statistics of factor loadings:
loadings_fit_model_bifactor <- standardizedsolution(fit_model_bifactor, type = "std.all")
```

```
# Selecting based on row indices
print(round(loadings_fit_model_bifactor[1:10, ] %>% select(est.std) %>%
  summarise(mean = mean(abs(est.std)), sd = sd(abs(est.std)),
    min = min(abs(est.std)), max = max(abs(est.std))), 2))
```

```
##      mean  sd  min  max
## 1 0.27 0.2 0.01 0.58
```

```
# Loadings for cog: Mean = 0.27, SD = 0.2, min = 0.01 and max = 0.58
```

```
# Selecting based on row indices
print(round(loadings_fit_model_bifactor[11:20, ] %>% select(est.std) %>%
  summarise(mean = mean(abs(est.std)), sd = sd(abs(est.std)),
    min = min(abs(est.std)), max = max(abs(est.std))), 2))
```

```
##      mean  sd  min  max
## 1 0.31 0.19 0.01 0.67
```

```
# Loadings for com: Mean = 0.31, SD = 0.19, min = 0.01 and max = 0.67
```

```
# Selecting based on row indices
print(round(loadings_fit_model_bifactor[21:40, ] %>% select(est.std) %>%
  summarise(mean = mean(abs(est.std)), sd = sd(abs(est.std)),
    min = min(abs(est.std)), max = max(abs(est.std))), 2))
```

```
##      mean  sd  min  max
## 1 0.52 0.18 0.06 0.77
```

```
# Loadings for general: Mean = 0.52, SD = 0.18, min = 0.06 and max = 0.77
```

```
# Reliability of the latent variables in the model
print(round(reliability(fit_model_bifactor), 2))
```

```
## For constructs with categorical indicators, Zumbo et al.'s (2007) "ordinal alpha" is calculated in a
```

```
##      cog  com  general
```



```
## alpha      0.72 0.83      0.85
## alpha.ord  0.77 0.87      0.89
## omega      0.30 0.39      0.76
## omega2     0.15 0.19      0.77
## omega3     0.15 0.19      0.78
## avevar      NA  NA       NA
```

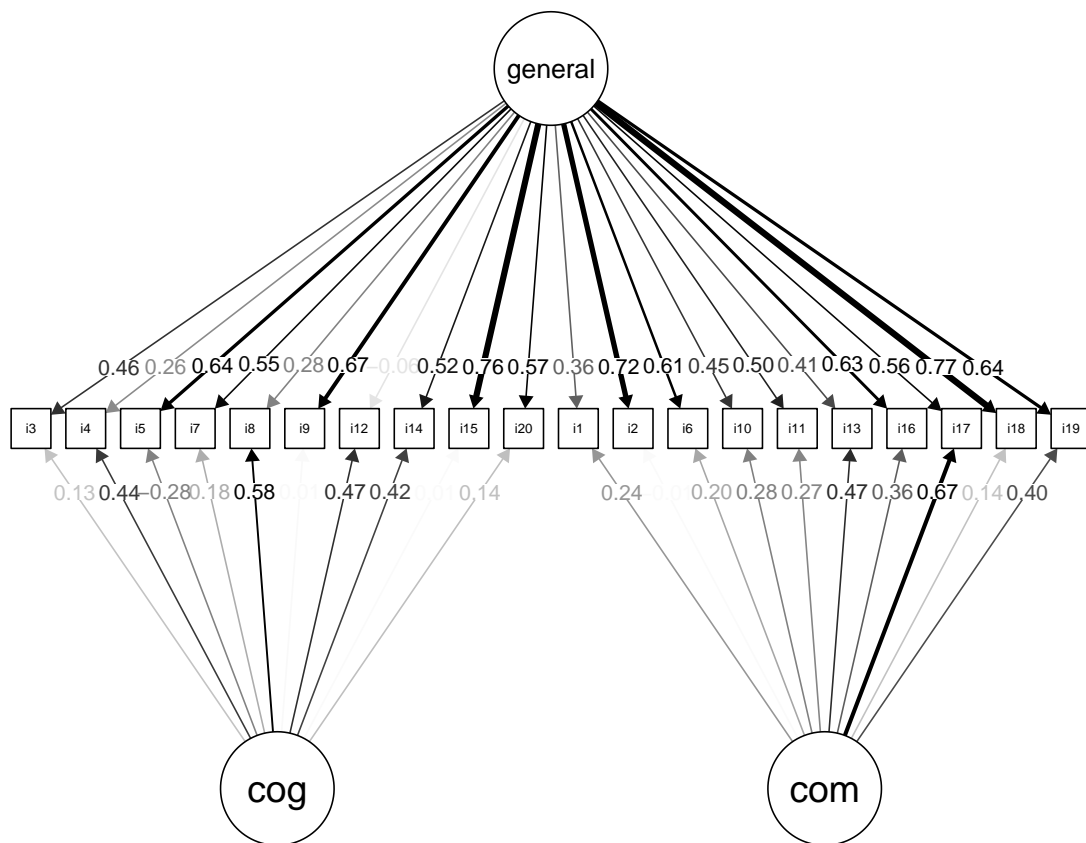
```
comp_reliability(fit_model_bifactor)
```

```
## # A tibble: 3 x 2
##   lhs      composite_reliability
##   <chr>          <dbl>
## 1 cog          0.420
## 2 com          0.630
## 3 general      0.901
```

```
# cog: alpha = 0.72, alpha ord. = 0.77, omega McDonald = 0.15, comp. reliability = 0.42
# com: alpha = 0.83, alpha ord. = 0.87, omega McDonald = 0.19, comp. reliability = 0.63
# general: alpha = 0.87, alpha ord. = 0.90, omega McDonald = 0.78, comp. reliability = 0.90
```

```
# Plot the model
```

```
semPaths(fit_model_bifactor,"std",layout="tree2",residuals=FALSE,sizeLat=10,sizeLat2=10,edge.color="black",
  mar=c(2, 2, 2, 2),esize=4,curvePivot = FALSE, intercepts=FALSE,thresholds = FALSE,
  nCharNodes=0,sizeMan=3.5, edge.label.position=0.85, bifactor = "general") # Specify general fa
```



Constrained Bifactor Model (Orthogonal Factors)

```

# Define the constrained bifactor model (fixing negative loadings to 0)
model_bifactor <- '
cog =~ i3 + i4 + 0*i5 + i7 + i8 + i9 + i12 + i14 + i15 + i20
com =~ i1 + 0*i2 + i6 + i10 + i11 + i13 + i16 + i17 + i18 + i19

general =~ i1 + i2 + i3 + i4 + i5 + i6 + i7 + i8 + i9 + i10 +
          i11 + 0*i12 + i13 + i14 + i15 + i16 + i17 + i18 + i19 + i20
'

# Fit the constrained bifactor model
fit_model_bifactor <- cfa(model_bifactor, data = data, ordered = TRUE, orthogonal = TRUE,
                          estimator = "WLSMV", std.lv=TRUE)

# Get fit measures
fitMeasures(fit_model_bifactor, fit.measures = c("chisq","df","cfi", "rmsea",

##          chisq          df          cfi          rmsea rmsea.ci.lower
##          199.014        153.000        0.987        0.049        0.027
## rmsea.ci.upper
##          0.067

# chisq(df=153) = 199.014, CFI = 0.987, RMSEA = 0.045 [0.020, 0.064]

# Get summary
summary(fit_model_bifactor, fit.measures = TRUE, standardized = TRUE)

## lavaan 0.6-19 ended normally after 40 iterations
##
##      Estimator                      DWLS
##      Optimization method          NLMINB
##      Number of model parameters      117
##
##                                     Used      Total
##      Number of observations          127        202
##
## Model Test User Model:
##                                     Standard      Scaled
##      Test Statistic                 199.014    256.157
##      Degrees of freedom              153        153
##      P-value (Chi-square)            0.007      0.000
##      Scaling correction factor                1.090
##      Shift parameter                   73.497
##      simple second-order correction
##
## Model Test Baseline Model:
##
##      Test statistic                 3742.984    1405.901
##      Degrees of freedom              190        190
##      P-value                        0.000      0.000
##      Scaling correction factor                2.922
##

```

```

## User Model versus Baseline Model:
##
##   Comparative Fit Index (CFI)                0.987      0.915
##   Tucker-Lewis Index (TLI)                  0.984      0.895
##
##   Robust Comparative Fit Index (CFI)                0.747
##   Robust Tucker-Lewis Index (TLI)                0.685
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                0.049      0.073
##   90 Percent confidence interval - lower      0.027      0.057
##   90 Percent confidence interval - upper      0.067      0.089
##   P-value H_0: RMSEA <= 0.050                0.525      0.010
##   P-value H_0: RMSEA >= 0.080                0.001      0.242
##
##   Robust RMSEA                0.123
##   90 Percent confidence interval - lower      0.108
##   90 Percent confidence interval - upper      0.138
##   P-value H_0: Robust RMSEA <= 0.050          0.000
##   P-value H_0: Robust RMSEA >= 0.080          1.000
##
## Standardized Root Mean Square Residual:
##
##   SRMR                0.080      0.080
##
## Parameter Estimates:
##
##   Parameterization                Delta
##   Standard errors                Robust.sem
##   Information                Expected
##   Information saturated (h1) model      Unstructured
##
## Latent Variables:
##
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
## cog =~
##   i3           0.141   0.107   1.321   0.187   0.141   0.141
##   i4           0.465   0.131   3.544   0.000   0.465   0.465
##   i5           0.000           1.609   0.108   0.155   0.155
##   i7           0.155   0.096   1.609   0.108   0.155   0.155
##   i8           0.611   0.119   5.114   0.000   0.611   0.611
##   i9          -0.001   0.105  -0.005   0.996  -0.001  -0.001
##   i12          0.396   0.144   2.747   0.006   0.396   0.396
##   i14          0.413   0.109   3.769   0.000   0.413   0.413
##   i15          -0.027   0.102  -0.270   0.787  -0.027  -0.027
##   i20          0.198   0.111   1.779   0.075   0.198   0.198
## com =~
##   i1           0.237   0.121   1.961   0.050   0.237   0.237
##   i2           0.000           2.974   0.003   0.279   0.279
##   i6           0.193   0.109   1.771   0.077   0.193   0.193
##   i10          0.279   0.094   2.974   0.003   0.279   0.279
##   i11          0.273   0.110   2.484   0.013   0.273   0.273
##   i13          0.475   0.088   5.427   0.000   0.475   0.475
##   i16          0.349   0.079   4.411   0.000   0.349   0.349

```

```

##      i17      0.674    0.068    9.877    0.000    0.674    0.674
##      i18      0.119    0.096    1.239    0.215    0.119    0.119
##      i19      0.395    0.081    4.873    0.000    0.395    0.395
## general =~
##      i1      0.356    0.079    4.534    0.000    0.356    0.356
##      i2      0.721    0.055   13.223    0.000    0.721    0.721
##      i3      0.460    0.071    6.458    0.000    0.460    0.460
##      i4      0.251    0.092    2.741    0.006    0.251    0.251
##      i5      0.615    0.061   10.075    0.000    0.615    0.615
##      i6      0.611    0.063    9.748    0.000    0.611    0.611
##      i7      0.552    0.066    8.320    0.000    0.552    0.552
##      i8      0.265    0.086    3.094    0.002    0.265    0.265
##      i9      0.672    0.054   12.477    0.000    0.672    0.672
##     i10      0.457    0.073    6.229    0.000    0.457    0.457
##     i11      0.502    0.072    6.967    0.000    0.502    0.502
##     i12      0.000                    0.000    0.000
##     i13      0.409    0.076    5.359    0.000    0.409    0.409
##     i14      0.515    0.087    5.934    0.000    0.515    0.515
##     i15      0.767    0.046   16.773    0.000    0.767    0.767
##     i16      0.634    0.059   10.763    0.000    0.634    0.634
##     i17      0.563    0.064    8.774    0.000    0.563    0.563
##     i18      0.776    0.054   14.462    0.000    0.776    0.776
##     i19      0.645    0.060   10.669    0.000    0.645    0.645
##     i20      0.557    0.070    7.923    0.000    0.557    0.557
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## cog ~~
##   com      0.000                    0.000    0.000
##   general  0.000                    0.000    0.000
## com ~~
##   general  0.000                    0.000    0.000
##
## Thresholds:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## i3|t1     -1.469    0.169   -8.712    0.000   -1.469   -1.469
## i3|t2     -1.184    0.145   -8.149    0.000   -1.184   -1.184
## i3|t3     -0.270    0.113   -2.384    0.017   -0.270   -0.270
## i3|t4      0.527    0.117    4.483    0.000    0.527    0.527
## i4|t1     -1.414    0.163   -8.654    0.000   -1.414   -1.414
## i4|t2     -0.973    0.133   -7.306    0.000   -0.973   -0.973
## i4|t3     -0.644    0.120   -5.344    0.000   -0.644   -0.644
## i4|t4     -0.030    0.112   -0.265    0.791   -0.030   -0.030
## i5|t1     -1.362    0.159   -8.577    0.000   -1.362   -1.362
## i5|t2     -1.225    0.148   -8.270    0.000   -1.225   -1.225
## i5|t3     -0.416    0.115   -3.613    0.000   -0.416   -0.416
## i5|t4      0.460    0.116    3.962    0.000    0.460    0.460
## i7|t1     -1.672    0.192   -8.720    0.000   -1.672   -1.672
## i7|t2     -1.146    0.143   -8.021    0.000   -1.146   -1.146
## i7|t3     -0.229    0.113   -2.031    0.042   -0.229   -0.229
## i7|t4      0.596    0.119    5.001    0.000    0.596    0.596
## i8|t1     -1.672    0.192   -8.720    0.000   -1.672   -1.672
## i8|t2     -1.184    0.145   -8.149    0.000   -1.184   -1.184
## i8|t3     -0.353    0.114   -3.087    0.002   -0.353   -0.353

```

##	i8 t4	0.353	0.114	3.087	0.002	0.353	0.353
##	i9 t1	-1.225	0.148	-8.270	0.000	-1.225	-1.225
##	i9 t2	-0.797	0.126	-6.351	0.000	-0.797	-0.797
##	i9 t3	-0.249	0.113	-2.207	0.027	-0.249	-0.249
##	i9 t4	0.573	0.119	4.829	0.000	0.573	0.573
##	i12 t1	-0.693	0.122	-5.683	0.000	-0.693	-0.693
##	i12 t2	-0.290	0.113	-2.560	0.010	-0.290	-0.290
##	i12 t3	0.049	0.112	0.442	0.659	0.049	0.049
##	i12 t4	0.504	0.117	4.310	0.000	0.504	0.504
##	i14 t1	-2.151	0.281	-7.657	0.000	-2.151	-2.151
##	i14 t2	-1.530	0.175	-8.748	0.000	-1.530	-1.530
##	i14 t3	-0.882	0.129	-6.837	0.000	-0.882	-0.882
##	i14 t4	-0.069	0.112	-0.619	0.536	-0.069	-0.069
##	i15 t1	-2.415	0.364	-6.629	0.000	-2.415	-2.415
##	i15 t2	-1.758	0.204	-8.632	0.000	-1.758	-1.758
##	i15 t3	-0.825	0.127	-6.515	0.000	-0.825	-0.825
##	i15 t4	0.270	0.113	2.384	0.017	0.270	0.270
##	i20 t1	-1.859	0.220	-8.465	0.000	-1.859	-1.859
##	i20 t2	-1.469	0.169	-8.712	0.000	-1.469	-1.469
##	i20 t3	-0.527	0.117	-4.483	0.000	-0.527	-0.527
##	i20 t4	0.290	0.113	2.560	0.010	0.290	0.290
##	i1 t1	-1.362	0.159	-8.577	0.000	-1.362	-1.362
##	i1 t2	-0.973	0.133	-7.306	0.000	-0.973	-0.973
##	i1 t3	-0.030	0.112	-0.265	0.791	-0.030	-0.030
##	i1 t4	0.527	0.117	4.483	0.000	0.527	0.527
##	i2 t1	-1.758	0.204	-8.632	0.000	-1.758	-1.758
##	i2 t2	-1.597	0.182	-8.754	0.000	-1.597	-1.597
##	i2 t3	-0.825	0.127	-6.515	0.000	-0.825	-0.825
##	i2 t4	0.109	0.112	0.972	0.331	0.109	0.109
##	i6 t1	-1.984	0.243	-8.172	0.000	-1.984	-1.984
##	i6 t2	-1.414	0.163	-8.654	0.000	-1.414	-1.414
##	i6 t3	-0.527	0.117	-4.483	0.000	-0.527	-0.527
##	i6 t4	0.290	0.113	2.560	0.010	0.290	0.290
##	i10 t1	-1.758	0.204	-8.632	0.000	-1.758	-1.758
##	i10 t2	-1.314	0.155	-8.486	0.000	-1.314	-1.314
##	i10 t3	-0.353	0.114	-3.087	0.002	-0.353	-0.353
##	i10 t4	0.527	0.117	4.483	0.000	0.527	0.527
##	i11 t1	-1.859	0.220	-8.465	0.000	-1.859	-1.859
##	i11 t2	-1.268	0.151	-8.383	0.000	-1.268	-1.268
##	i11 t3	-0.249	0.113	-2.207	0.027	-0.249	-0.249
##	i11 t4	0.504	0.117	4.310	0.000	0.504	0.504
##	i13 t1	-1.597	0.182	-8.754	0.000	-1.597	-1.597
##	i13 t2	-1.146	0.143	-8.021	0.000	-1.146	-1.146
##	i13 t3	-0.374	0.115	-3.262	0.001	-0.374	-0.374
##	i13 t4	0.395	0.115	3.438	0.001	0.395	0.395
##	i16 t1	-1.984	0.243	-8.172	0.000	-1.984	-1.984
##	i16 t2	-1.314	0.155	-8.486	0.000	-1.314	-1.314
##	i16 t3	-0.596	0.119	-5.001	0.000	-0.596	-0.596
##	i16 t4	0.438	0.116	3.787	0.000	0.438	0.438
##	i17 t1	-1.859	0.220	-8.465	0.000	-1.859	-1.859
##	i17 t2	-1.469	0.169	-8.712	0.000	-1.469	-1.469
##	i17 t3	-0.825	0.127	-6.515	0.000	-0.825	-0.825
##	i17 t4	-0.010	0.112	-0.088	0.930	-0.010	-0.010
##	i18 t1	-1.984	0.243	-8.172	0.000	-1.984	-1.984

```
##      i18|t2      -1.362    0.159   -8.577    0.000   -1.362   -1.362
##      i18|t3      -0.825    0.127   -6.515    0.000   -0.825   -0.825
##      i18|t4      -0.069    0.112   -0.619    0.536   -0.069   -0.069
##      i19|t1      -1.859    0.220   -8.465    0.000   -1.859   -1.859
##      i19|t2      -1.672    0.192   -8.720    0.000   -1.672   -1.672
##      i19|t3      -0.693    0.122   -5.683    0.000   -0.693   -0.693
##      i19|t4       0.169    0.112    1.502    0.133    0.169    0.169
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .i3      0.768
##      .i4      0.721
##      .i5      0.622
##      .i7      0.672
##      .i8      0.557
##      .i9      0.549
##      .i12     0.843
##      .i14     0.564
##      .i15     0.411
##      .i20     0.650
##      .i1      0.817
##      .i2      0.480
##      .i6      0.590
##      .i10     0.713
##      .i11     0.674
##      .i13     0.607
##      .i16     0.477
##      .i17     0.229
##      .i18     0.384
##      .i19     0.428
##      cog      1.000
##      com      1.000
##      general  1.000
```

```
# Descriptive statistics of factor loadings:
loadings_fit_model_bifactor <- standardizedsolution(fit_model_bifactor, type = "std.all")

print(round(loadings_fit_model_bifactor[1:9, ] %>% select(est.std) %>%
  summarise(mean = mean(abs(est.std)), sd = sd(abs(est.std)),
    min = min(abs(est.std)), max = max(abs(est.std))), 2))
```

```
##      mean    sd min  max
## 1 0.25 0.23  0 0.61
```

```
# Mean = 0.27, SD = 0.2, min = 0.01 and max = 0.58
```

```
# Selecting based on row indices
print(round(loadings_fit_model_bifactor[10:18, ] %>% select(est.std) %>%
  summarise(mean = mean(abs(est.std)), sd = sd(abs(est.std)),
    min = min(abs(est.std)), max = max(abs(est.std))), 2))
```

```
##      mean    sd min  max
## 1  0.3 0.19  0 0.67
```

```
# Mean = 0.31, SD = 0.19, min = 0.01 and max = 0.67
```

```
# Selecting based on row indices
```

```
print(round(loadings_fit_model_bifactor[19:36, ] %>% select(est.std) %>%  
  summarise(mean = mean(abs(est.std)), sd = sd(abs(est.std)),  
    min = min(abs(est.std)), max = max(abs(est.std))), 2))
```

```
##   mean   sd min  max  
## 1 0.46 0.21  0 0.77
```

```
# Mean = 0.52, SD = 0.18, min = 0.06 and max = 0.77
```

```
# Reliability
```

```
print(round(reliability(fit_model_bifactor), 2))
```

```
## For constructs with categorical indicators, Zumbo et al.'s (2007) "ordinal alpha" is calculated in a
```

```
##           cog  com general  
## alpha      0.71 0.81    0.87  
## alpha.ord  0.76 0.86    0.90  
## omega      0.36 0.40    0.78  
## omega2     0.21 0.22    0.79  
## omega3     0.21 0.22    0.80  
## avevar      NA  NA     NA
```

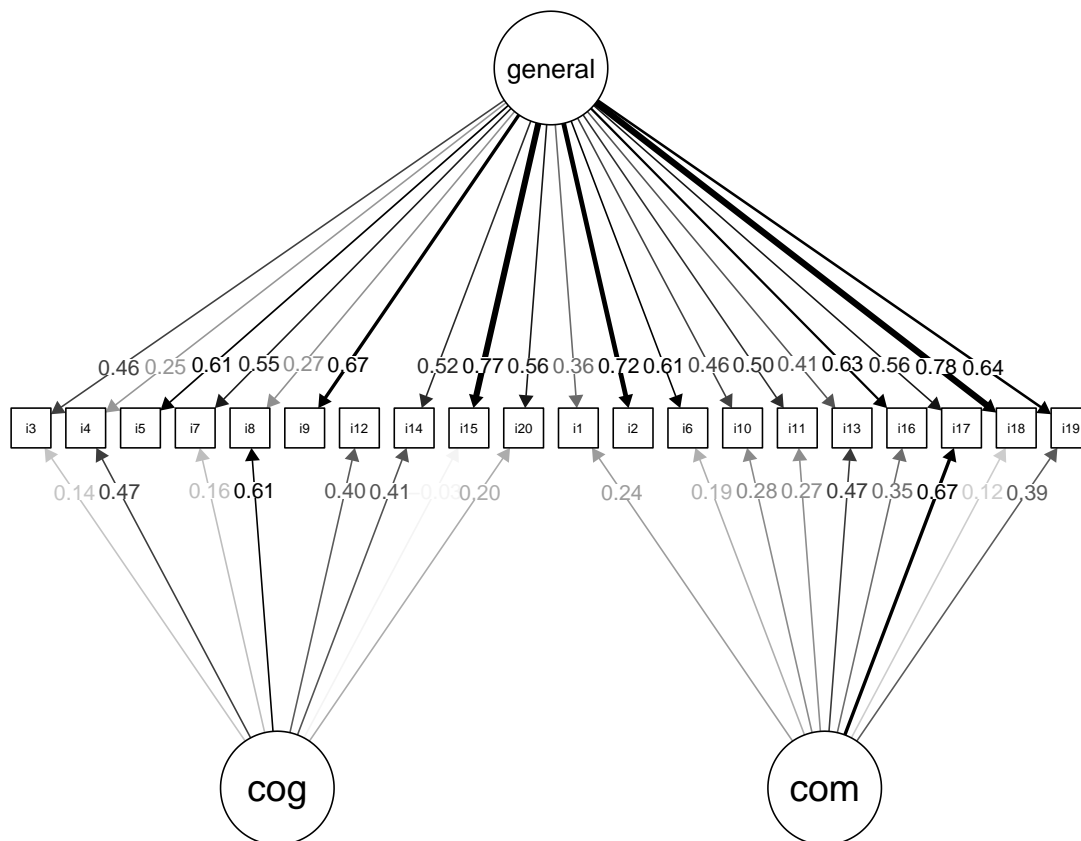
```
comp_reliability(fit_model_bifactor)
```

```
## # A tibble: 3 x 2  
##   lhs      composite_reliability  
##   <chr>                <dbl>  
## 1 cog                    0.465  
## 2 com                    0.624  
## 3 general                0.901
```

```
# cog: alpha = 0.71, alpha ord. = 0.76, omega McDonald = 0.36, comp. reliability = 0.46  
# com: alpha = 0.81, alpha ord. = 0.86, omega McDonald = 0.40, comp. reliability = 0.62  
# geral: alpha = 0.87, alpha ord. = 0.90, omega McDonald = 0.78, comp. reliability = 0.90
```

```
# Plot the model
```

```
semPaths(fit_model_bifactor,"std",layout="tree2",residuals=FALSE,sizeLat=10,sizeLat2=10,edge.color="bla  
  mar=c(2, 2, 2, 2),esize=4,curvePivot = FALSE, intercepts=FALSE,thresholds = FALSE,  
  nCharNodes=0,sizeMan=3.5, edge.label.position=0.85, bifactor = "general")
```



Model Comparison

O melhor modelo foi o bifatorial

```
lavTestLRT(fit_model_uni, fit_model_2f, fit_model_bifactor)
```

```
##
## Scaled Chi-Squared Difference Test (method = "satorra.2000")
##
## lavaan->lavTestLRT():
##   lavaan NOTE: The "Chisq" column contains standard test statistics, not the
##   robust test that should be reported per model. A robust difference test is
##   a function of two standard (not robust) statistics.
##
##           Df AIC BIC  Chisq Chisq diff Df diff Pr(>Chisq)
## fit_model_bifactor 153      199.01
## fit_model_2f      169      267.86      58.819      16 8.265e-07 ***
## fit_model_uni      170      283.29      13.562       1 0.0002308 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
sessionInfo()
```

```
## R version 4.4.2 (2024-10-31 ucrt)
## Platform: x86_64-w64-mingw32/x64
## Running under: Windows 11 x64 (build 26100)
```



```

##
## Matrix products: default
##
##
## locale:
## [1] LC_COLLATE=Portuguese_Brazil.utf8 LC_CTYPE=Portuguese_Brazil.utf8
## [3] LC_MONETARY=Portuguese_Brazil.utf8 LC_NUMERIC=C
## [5] LC_TIME=Portuguese_Brazil.utf8
##
## time zone: America/Sao_Paulo
## tzcode source: internal
##
## attached base packages:
## [1] stats      graphics  grDevices  utils      datasets  methods    base
##
## other attached packages:
## [1] semPlot_1.1.6  semTools_0.5-6 lavaan_0.6-19 MVN_5.9      dplyr_1.1.4
## [6] psych_2.5.3    readxl_1.4.3
##
## loaded via a namespace (and not attached):
## [1] Rdpack_2.6.4      mnormt_2.1.1      pbapply_1.7-2
## [4] gridExtra_2.3     fdrtool_1.2.18    sandwich_3.1-1
## [7] rlang_1.1.5       magrittr_2.0.3    multcomp_1.4-26
## [10] rockchalk_1.8.157 compiler_4.4.2    png_0.1-8
## [13] vctrs_0.6.5       reshape2_1.4.4    OpenMx_2.21.13
## [16] gsl_2.1-8         quadprog_1.5-8    stringr_1.5.1
## [19] crayon_1.5.3      pkgconfig_2.0.3   fastmap_1.2.0
## [22] arm_1.14-4        backports_1.5.0   energy_1.7-12
## [25] utf8_1.2.4        pbivnorm_0.6.0    rmarkdown_2.29
## [28] nloptr_2.2.1      xfun_0.52         kutils_1.73
## [31] jpeg_0.1-10       parallel_4.4.2    cluster_2.1.6
## [34] R6_2.6.1          stringi_1.8.7     car_3.1-3
## [37] boot_1.3-31       rpart_4.1.23      cellranger_1.1.0
## [40] estimability_1.5.1 Rcpp_1.0.14       knitr_1.50
## [43] zoo_1.8-14        base64enc_0.1-3    Matrix_1.7-1
## [46] splines_4.4.2     nnet_7.3-19       igraph_2.1.4
## [49] tidyselect_1.2.1  rstudioapi_0.17.1 abind_1.4-8
## [52] yaml_2.3.10       codetools_0.2-20  qgraph_1.9.8
## [55] lattice_0.22-6    tibble_3.2.1      plyr_1.8.9
## [58] withr_3.0.2       coda_0.19-4.1     evaluate_1.0.3
## [61] moments_0.14.1    foreign_0.8-87    survival_3.7-0
## [64] RcppParallel_5.1.9 zip_2.3.1         pillar_1.10.2
## [67] carData_3.0-5     checkmate_2.3.2   nortest_1.0-4
## [70] stats4_4.4.2      reformulas_0.4.0  generics_0.1.3
## [73] ggplot2_3.5.2     munsell_0.5.1     scales_1.3.0
## [76] minqa_1.2.8       gtools_3.9.5      xtable_1.8-4
## [79] glue_1.8.0        mi_1.1            emmeans_1.10.6
## [82] Hmisc_5.2-3       tools_4.4.2       data.table_1.17.0
## [85] lme4_1.1-37       openxlsx_4.2.7.1  mvtnorm_1.3-3
## [88] XML_3.99-0.18     grid_4.4.2        sem_3.1-16
## [91] rbibutils_2.3     colorspace_2.1-1  nlme_3.1-166
## [94] htmlTable_2.4.3   Formula_1.2-5     cli_3.6.4
## [97] glasso_1.11       corpcor_1.6.10    gtable_0.3.6
## [100] digest_0.6.37     TH.data_1.1-3     htmlwidgets_1.6.4

```

```
## [103] htmltools_0.5.8.1 lifecycle_1.0.4    lisrelToR_0.3  
## [106] MASS_7.3-61
```