

Content Validation and Internal Validation for Dissertation

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This document presents the analysis of judge ratings collected during the content validation phase of a dissertation project, followed by an examination of its internal structure. This code examines the “Assessment Scale for Group Music Therapy in Substance Use Disorders (MTDQ)” (Pedrosa, 2023).

Content validity

Load Libraries an preparing data

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

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

```
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(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)

# -----
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, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37, 38, 39, 40, 41, 42, 43, 44, 45)
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
  }
}))
```

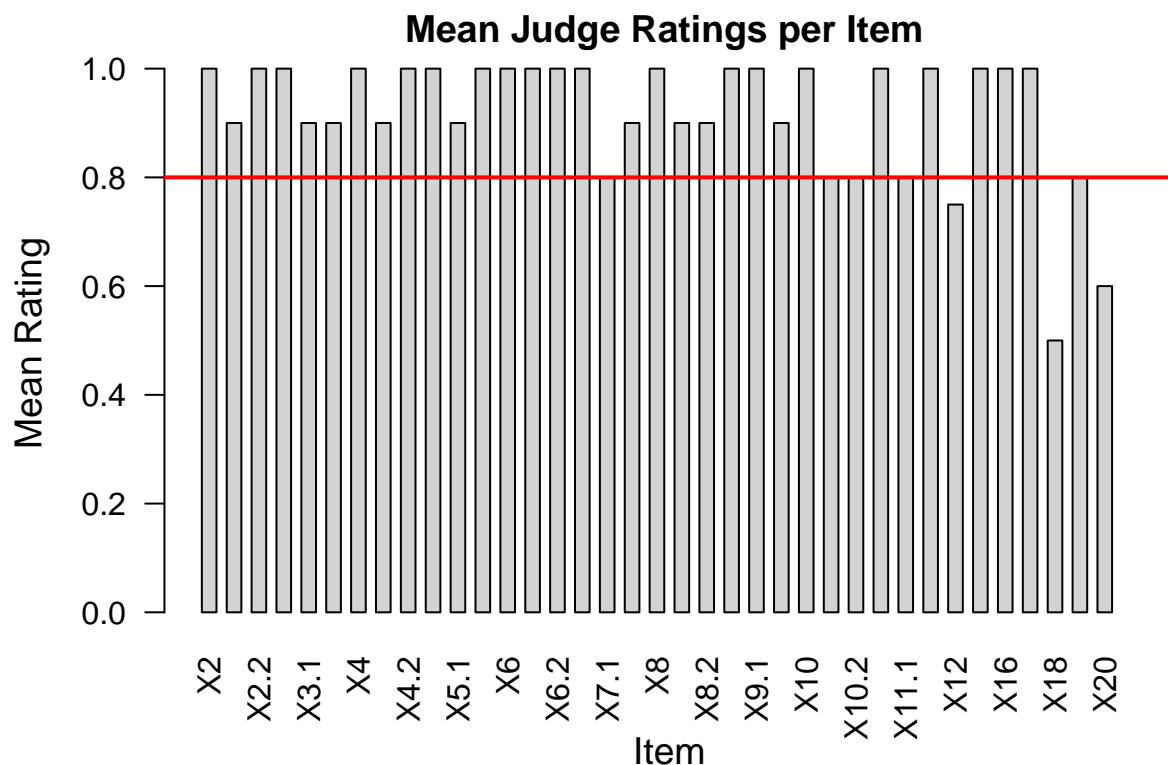
```
)))
```

```
# 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)

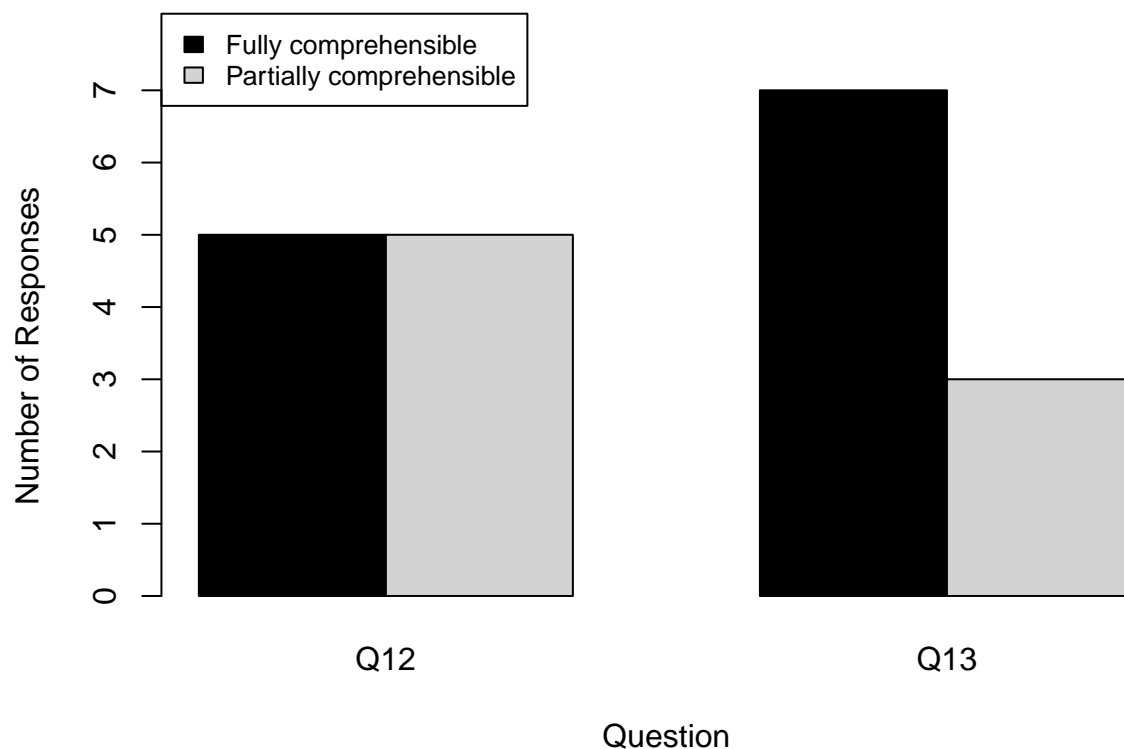
# Create a data frame for plotting
# Handle cases where a category might have zero counts by accessing table elements by name
plot_data_q12_q13 <- data.frame(
  Question = c("Q12", "Q13"),
  `Fully comprehensible` = c(freq_q12["Fully comprehensible"], freq_q13["Fully comprehensible"]),
  `Partially comprehensible`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

```

# --- 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.

```

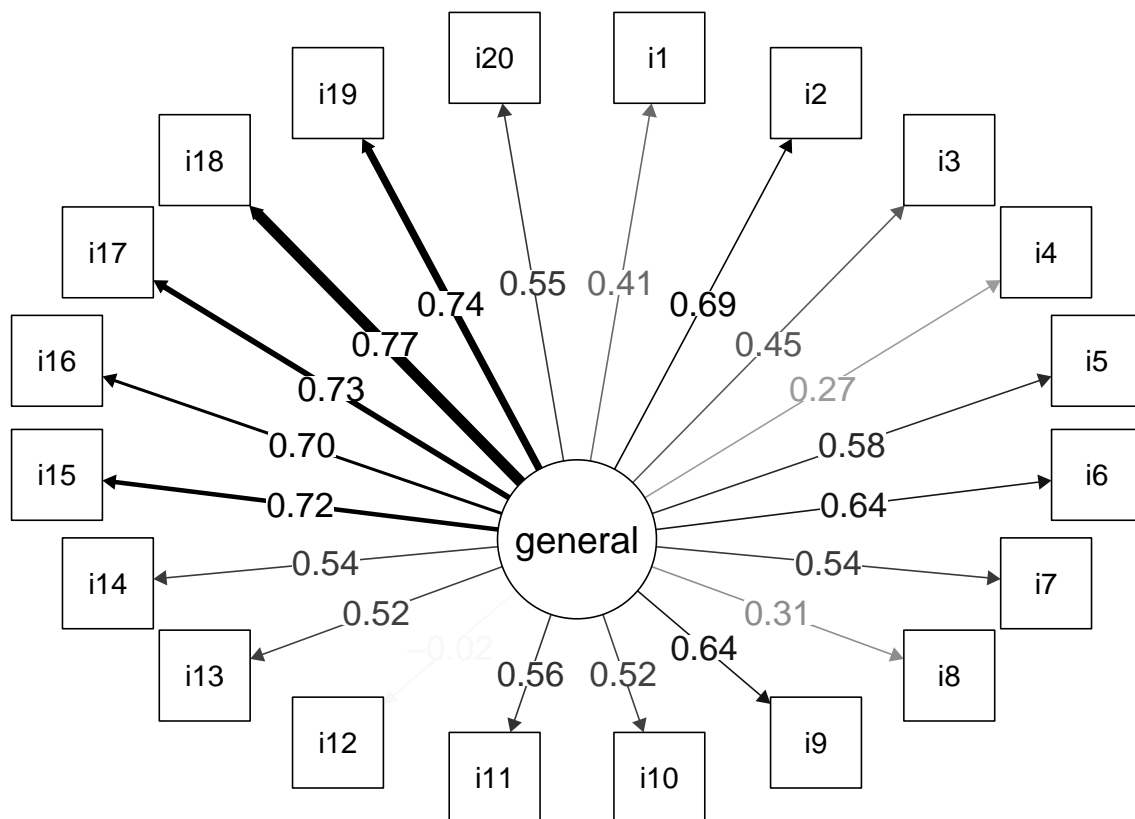
```

## # 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
```

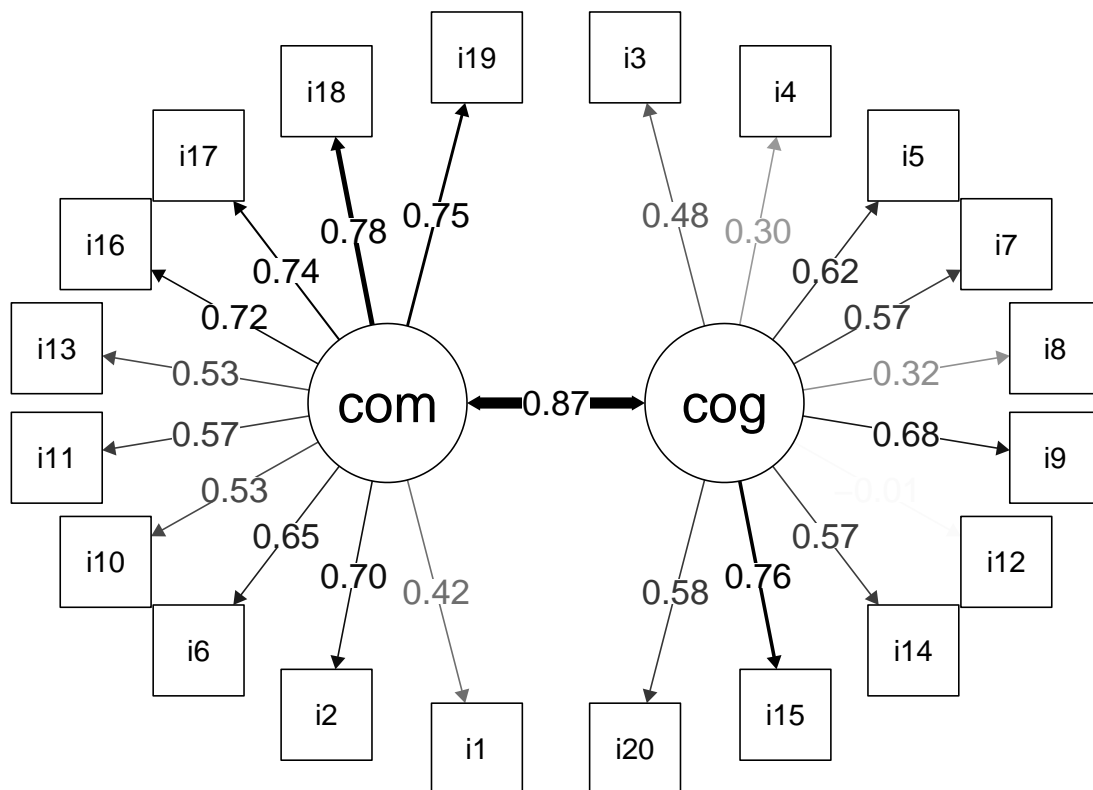
```
comp_reliability(fit_model_2f)
```

```
## # 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
```

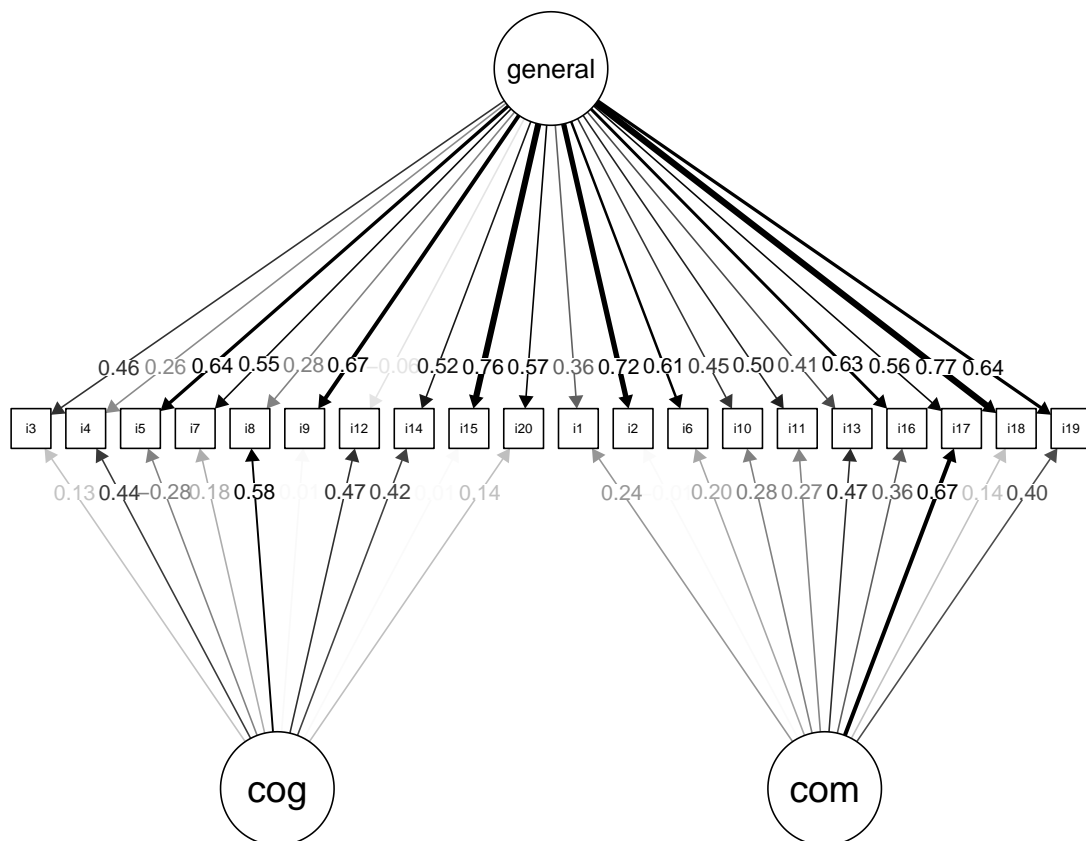
```
comp_reliability(fit_model_bifactor)
```

```
## # 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
##   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
##   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
```

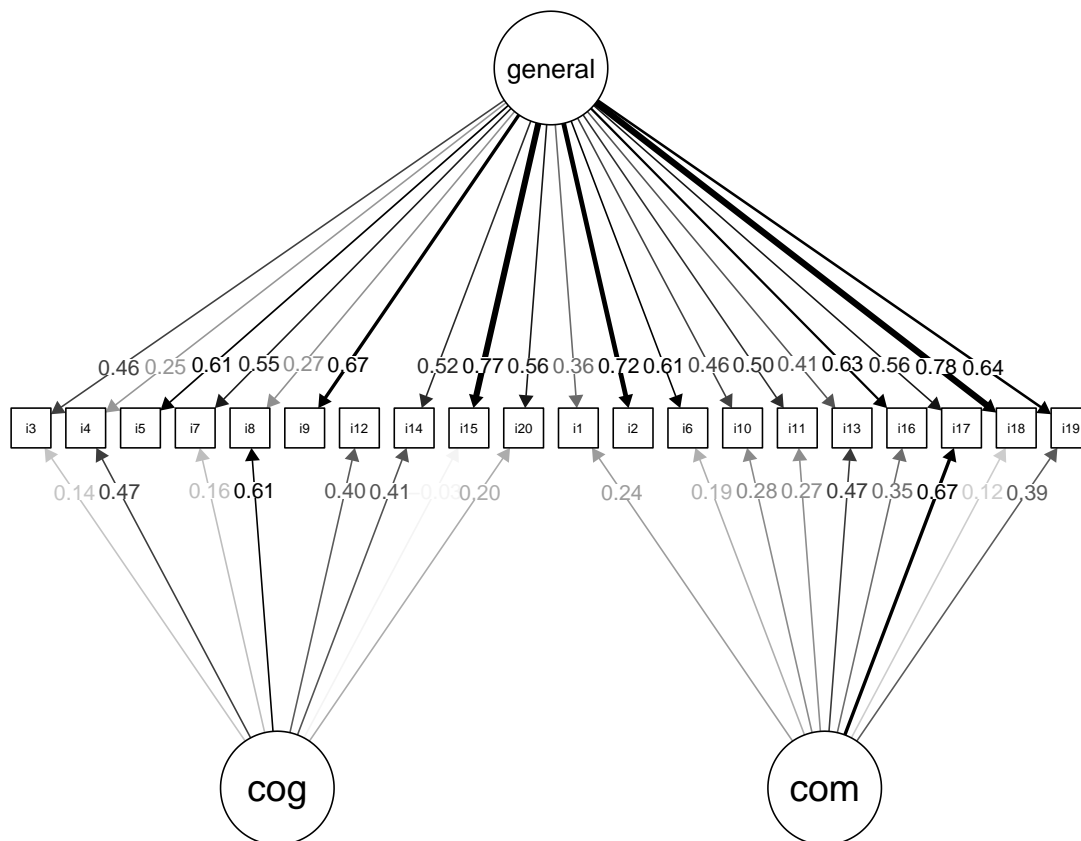
```
comp_reliability(fit_model_bifactor)
```

```
## # 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="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")
```



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
```