Results Financial Inclusion

Juan C. Correa

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```r
setwd("/home/juan/PaperBrandstrat")
```

Let's open the data

```r
library(readr)
Brandstat <- read_delim("RawDataBrandstat.csv", ",", escape_double = FALSE, trim_ws = TRUE)
```

```
##
## -- Column specification --------------------------------------------------------
## cols(
##   .default = col_double()
## )
## i Use `spec()` for the full column specifications.
```

```r
variable.names(Brandstat)
```

```
##  [1] "City"           "Sex"            "Strata"         "Age"
##  [5] "Occupation"     "Education"      "Civic_Status"   "Home_Structure"
##  [9] "Income_Level"   "FC1"            "FC2"            "FC3"
## [13] "FC4"            "Aut1"           "Aut2"           "Aut3"
## [17] "Aut4"           "Aut5"           "Aut6"           "Aut7"
```

```r
automation <- Brandstat[14:20]
FacCond <- Brandstat[10:13]
```

Sample Description

```r
library(psych)
describe.by(Brandstat$Age, group = Brandstat$Sex, mat = TRUE, digits = 2)
```

```
## Warning: describe.by is deprecated. Please use the describeBy function
```

```
##     item group1 vars   n  mean    sd median trimmed   mad min max range skew
## X11    1      1    1 248 32.69 11.76     29   31.64 11.86  18  61    43 0.65
## X12    2      2    1 230 32.97 11.87     29   31.84 11.86  18  69    51 0.73
##     kurtosis   se
## X11    -0.77 0.75
## X12    -0.59 0.78
```

```r
summary(Brandstat$Age)
```

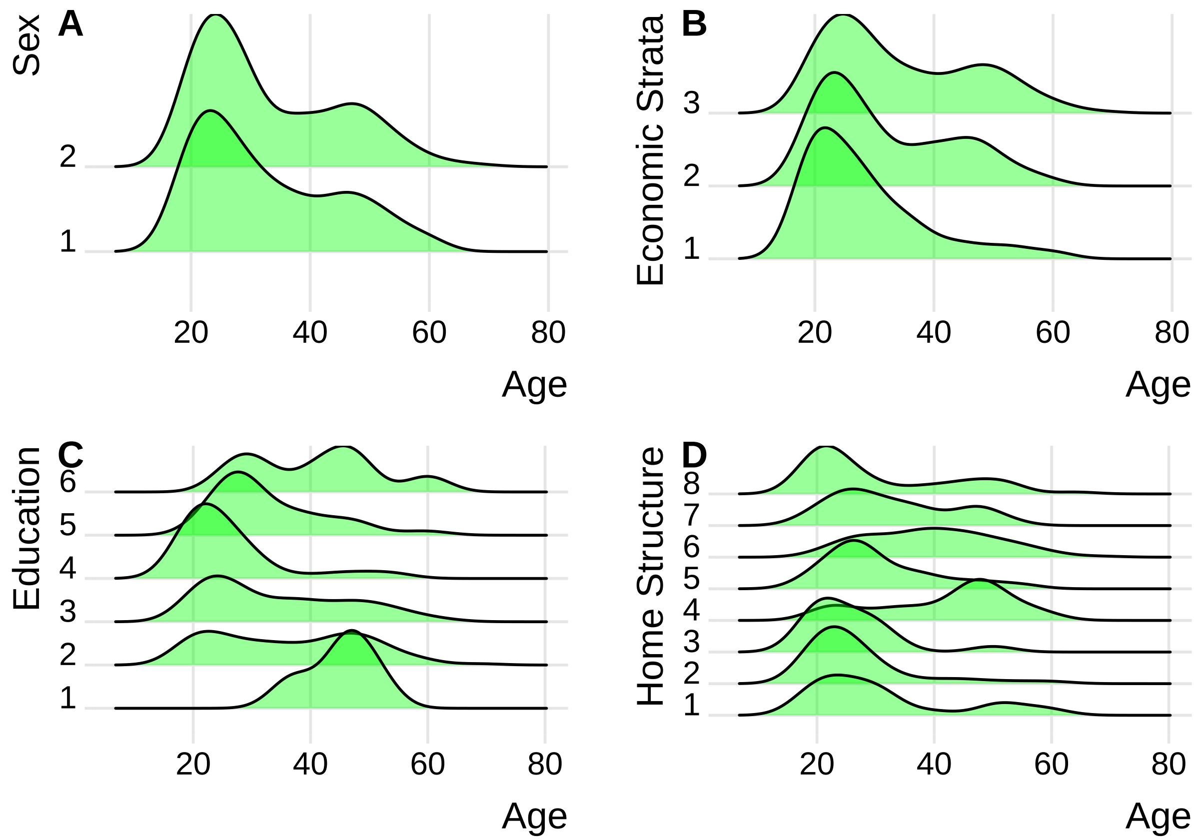
```
##    Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   18.00   23.00   29.00   32.82   42.75   69.00
```

```r
library(ggplot2)
```

```
##
## Attaching package: 'ggplot2'

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

```r
library(ggridges)
s1 <- ggplot(Brandstat, aes(x=Age, y=as.factor(Sex))) + geom_density_ridges(fill="green", alpha = 0.4) +
theme_ridges() + ylab("Sex")

s2 <- ggplot(Brandstat, aes(x=Age, y=as.factor(Strata))) + geom_density_ridges(fill="green", alpha = 0.4
theme_ridges() + ylab("Economic Strata")

s3 <- ggplot(Brandstat, aes(x=Age, y=as.factor(Education))) + geom_density_ridges(fill="green", alpha =
theme_ridges() + ylab("Education")

s4 <- ggplot(Brandstat, aes(x=Age, y=as.factor(Home_Structure))) + geom_density_ridges(fill="green", alp
theme_ridges() + ylab("Home Structure")

library(ggpubr)

ggarrange(s1, s2, s3, s4, ncol = 2, nrow = 2, labels = c("    A","    B","    C","    D"))
```

```
## Picking joint bandwidth of 3.56

## Picking joint bandwidth of 3.56

## Picking joint bandwidth of 3.75
## Picking joint bandwidth of 3.75
```



Measurement Model of Automation

To examine the factor structure of all the scales employed, we proceed by testing the overall sampling adequacy of the items. These items are susceptible to factorization (KMO = 0.92), and the most probable psychometric structure consists of a one-factor model with freely estimated factor loadings. This emerging solution is evident through the statistical significance of the homogeneous items test (F = 1.763; p = 0.0415) which proved to be better than a psychometric structure of a one-factor model with equal factor loadings for all items captured by the tau-equivalence test (F = 1.563; p = 0.0575).

```
library(psych)
KMO(automation)
```

```
## Kaiser-Meyer-Olkin factor adequacy
## Call: KMO(r = automation)
## Overall MSA =  0.92
## MSA for each item =
## Aut1 Aut2 Aut3 Aut4 Aut5 Aut6 Aut7
## 0.89 0.94 0.93 0.93 0.90 0.93 0.94
```

```
library(coefficientalpha)
```

```
## Loading required package: rsem

## Loading required package: MASS

## Loading required package: lavaan

## This is lavaan 0.6-7
```

```
## lavaan is BETA software! Please report any bugs.
```

```
##
## Attaching package: 'lavaan'
```

```
## The following object is masked from 'package:psych':
##
##     cor2cov
```

```
##
## Attaching package: 'coefficientalpha'
```

```
## The following object is masked from 'package:ggplot2':
##
##     alpha
```

```
## The following objects are masked from 'package:psych':
##
##     alpha, omega
```

```
tau.test(automation)
```

```
## Warning: Setting row names on a tibble is deprecated.
```

```
## Test of tau equivalent
## The robust F statistic is  1.563
## with a p-value  0.0575
##
## Test of homogeneous items
## The robust F statistic is  1.763
## with a p-value  0.0415
```

The results of the confirmatory factor analysis reveal a reasonable goodness-of-fit for the scale of automation (CFI = 0.950; TLI = 0.926; RMSEA = 0.142; SRMR = 0.037)

```
library(lavaan)
aut.model <- "aut =~ Aut1 + Aut2 + Aut3 + Aut4 + Aut5 + Aut6 + Aut7"
fit1 <- lavaan::cfa(aut.model, data=Brandstat,std.lv=TRUE)
summary(fit1, fit.measures=T,standardized=T)
```

```
## lavaan 0.6-7 ended normally after 17 iterations
##
##   Estimator                                         ML
##   Optimization method                           NLMINB
##   Number of free parameters                         14
##
##   Number of observations                           478
##
## Model Test User Model:
##
##   Test statistic                               149.084
##   Degrees of freedom                                14
##   P-value (Chi-square)                           0.000
##
## Model Test Baseline Model:
##
##   Test statistic                              2741.418
##   Degrees of freedom                                21
##   P-value                                        0.000
```

```
##
## User Model versus Baseline Model:
##
##   Comparative Fit Index (CFI)                    0.950
##   Tucker-Lewis Index (TLI)                       0.926
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)              -7248.230
##   Loglikelihood unrestricted model (H1)      -7173.688
##
##   Akaike (AIC)                               14524.460
##   Bayesian (BIC)                             14582.834
##   Sample-size adjusted Bayesian (BIC)        14538.400
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                                          0.142
##   90 Percent confidence interval - lower         0.122
##   90 Percent confidence interval - upper         0.163
##   P-value RMSEA <= 0.05                          0.000
##
## Standardized Root Mean Square Residual:
##
##   SRMR                                           0.037
##
## Parameter Estimates:
##
##   Standard errors                             Standard
##   Information                                 Expected
##   Information saturated (h1) model          Structured
##
## Latent Variables:
##                    Estimate  Std.Err  z-value  P(>|z|)   Std.lv  Std.all
##   aut =~
##     Aut1              2.812    0.114   24.714    0.000    2.812    0.891
##     Aut2              2.612    0.122   21.421    0.000    2.612    0.815
##     Aut3              2.371    0.123   19.234    0.000    2.371    0.758
##     Aut4              2.387    0.115   20.760    0.000    2.387    0.799
##     Aut5              2.870    0.120   23.986    0.000    2.870    0.876
##     Aut6              2.296    0.113   20.237    0.000    2.296    0.785
##     Aut7              2.668    0.115   23.241    0.000    2.668    0.859
##
## Variances:
##                    Estimate  Std.Err  z-value  P(>|z|)   Std.lv  Std.all
##    .Aut1              2.045    0.173   11.796    0.000    2.045    0.205
##    .Aut2              3.440    0.253   13.595    0.000    3.440    0.335
##    .Aut3              4.164    0.293   14.196    0.000    4.164    0.426
##    .Aut4              3.238    0.235   13.807    0.000    3.238    0.362
##    .Aut5              2.510    0.203   12.351    0.000    2.510    0.233
##    .Aut6              3.283    0.235   13.954    0.000    3.283    0.384
##    .Aut7              2.534    0.198   12.805    0.000    2.534    0.263
##     aut               1.000                               1.000    1.000
```

```r
library(semTools)
```

```
##
## ###############################################################################
## This is semTools 0.5-3
## All users of R (or SEM) are invited to submit functions or ideas for functions.
## ###############################################################################
##
## Attaching package: 'semTools'
## The following object is masked from 'package:psych':
##
##     skew
## The following object is masked from 'package:readr':
##
##     clipboard
```

```r
reliability(fit1)
```

```
##              aut
## alpha  0.9380854
## omega  0.9386485
## omega2 0.9386485
## omega3 0.9366932
## avevar 0.6875174
```

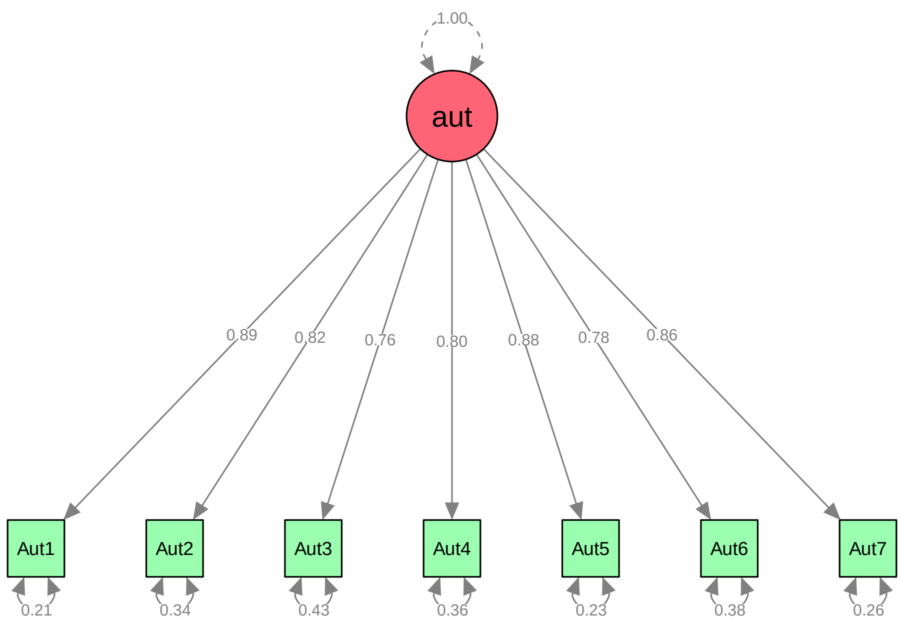
The emerging psychometric structure is depicted in the following picture.

```r
library(semPlot)
```

```
## Registered S3 methods overwritten by 'lme4':
##   method                          from
##   cooks.distance.influence.merMod car
##   influence.merMod                car
##   dfbeta.influence.merMod         car
##   dfbetas.influence.merMod        car
## Registered S3 methods overwritten by 'huge':
##   method    from
##   plot.sim  BDgraph
##   print.sim BDgraph
```

```r
semPaths(fit1, whatLabels = "stand", layout = "tree", color = list(
              lat = rgb(255, 100, 118, maxColorValue = 255),
              man = rgb(155, 253, 175, maxColorValue = 255)),
         mar = c(10, 5, 10, 5), intercepts = FALSE, residuls = FALSE, nCharNodes = 0)
```

```
## Warning in qgraph::qgraph(Edgelist, labels = nLab, bidirectional = Bidir, : The
## following arguments are not documented and likely not arguments of qgraph and
## thus ignored: residuls
```



Measurement Model Facilitating Conditions

The items of the scale of Facilitating conditions are also susceptible to factorization (KMO = 0.67), and the most probable psychometric structure consists of a one-factor model.

```
library(psych)
KMO(FacCond)
```

```
## Kaiser-Meyer-Olkin factor adequacy
## Call: KMO(r = FacCond)
## Overall MSA =  0.67
## MSA for each item =
##  FC1  FC2  FC3  FC4
## 0.65 0.74 0.73 0.64
```

```
library(coefficientalpha)
tau.test(FacCond)
```

```
## Warning: Setting row names on a tibble is deprecated.
```

```
## Test of tau equivalent
## The robust F statistic is  7.37
## with a p-value  0
##
## Test of homogeneous items
## The robust F statistic is  0.427
## with a p-value  0.6528
```

The Measurement model of facilitating conditions reveals also a reasonable goodness-of-fit, as can be seen in the following results (CFI = 0.996; TLI = 0.987; RMSEA = 0.034; SRMR = 0.018)

```
library(lavaan)
fc.model <- "FC =~ FC1 + FC2 + FC3 + FC4"
fit2 <- lavaan::cfa(fc.model, data=Brandstat,std.lv=TRUE)
summary(fit2, fit.measures=T,standardized=T)
```

```
## lavaan 0.6-7 ended normally after 27 iterations
##
##   Estimator                                         ML
##   Optimization method                           NLMINB
##   Number of free parameters                          8
##
##   Number of observations                           478
##
## Model Test User Model:
##
##   Test statistic                                 3.113
##   Degrees of freedom                                 2
##   P-value (Chi-square)                           0.211
##
## Model Test Baseline Model:
##
##   Test statistic                               266.994
##   Degrees of freedom                                 6
##   P-value                                        0.000
##
## User Model versus Baseline Model:
##
##   Comparative Fit Index (CFI)                    0.996
##   Tucker-Lewis Index (TLI)                       0.987
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)              -4365.125
##   Loglikelihood unrestricted model (H1)      -4363.568
##
##   Akaike (AIC)                                8746.249
##   Bayesian (BIC)                              8779.606
##   Sample-size adjusted Bayesian (BIC)         8754.215
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                                          0.034
##   90 Percent confidence interval - lower         0.000
##   90 Percent confidence interval - upper         0.103
##   P-value RMSEA <= 0.05                          0.543
##
## Standardized Root Mean Square Residual:
##
##   SRMR                                           0.018
##
## Parameter Estimates:
##
```

```
##   Standard errors                              Standard
##   Information                                  Expected
##   Information saturated (h1) model           Structured
##
## Latent Variables:
##                    Estimate  Std.Err  z-value  P(>|z|)   Std.lv  Std.all
##   FC =~
##     FC1               1.668    0.128   13.013    0.000    1.668    0.696
##     FC2               0.620    0.170    3.651    0.000    0.620    0.196
##     FC3               1.374    0.127   10.826    0.000    1.374    0.554
##     FC4               1.607    0.120   13.408    0.000    1.607    0.725
##
## Variances:
##                    Estimate  Std.Err  z-value  P(>|z|)   Std.lv  Std.all
##    .FC1               2.957    0.344    8.604    0.000    2.957    0.515
##    .FC2               9.665    0.635   15.215    0.000    9.665    0.962
##    .FC3               4.269    0.340   12.563    0.000    4.269    0.693
##    .FC4               2.334    0.305    7.661    0.000    2.334    0.475
##     FC                1.000                               1.000    1.000
```

```
inspect(fit2, "rsquare")
```

```
##   FC1   FC2   FC3   FC4
## 0.485 0.038 0.307 0.525
```

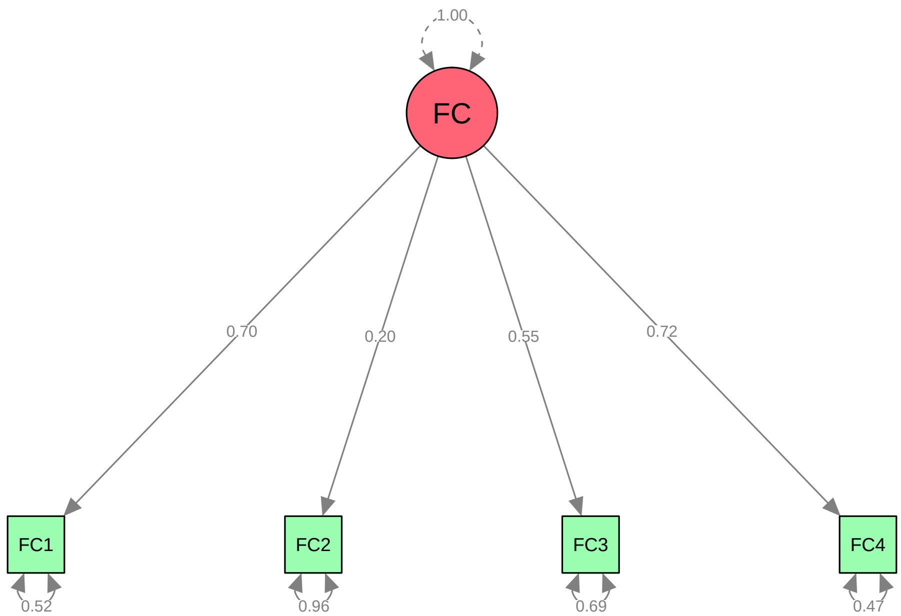
```
library(semTools)
reliability(fit2)
```

```
##               FC
## alpha  0.5714467
## omega  0.5908010
## omega2 0.5908010
## omega3 0.5905017
## avevar 0.2842474
```

The emerging psychometric structure is depicted in the following picture.

```
library(semPlot)
semPaths(fit2, whatLabels = "stand", layout = "tree", color = list(
              lat = rgb(255, 100, 118, maxColorValue = 255),
              man = rgb(155, 253, 175, maxColorValue = 255)),
          mar = c(10, 5, 10, 5), intercepts = FALSE, residuls = FALSE, nCharNodes = 0)
```

```
## Warning in qgraph::qgraph(Edgelist, labels = nLab, bidirectional = Bidir, : The
## following arguments are not documented and likely not arguments of qgraph and
## thus ignored: residuls
```



Structural Model: Facilitating Conditions as Predictor of Automation

Our structural model posits that facilitating conditions predicts the acceptance of technology through the automation of financial processes. This set of relationships are summarized and illustrated as follows

```
library(lavaan);
modelData <- Brandstat
 model<-"
! regressions
   Automation=~Aut__Aut1*Aut1
   Automation=~Aut__Aut2*Aut2
   Automation=~Aut__Aut3*Aut3
   Automation=~Aut__Aut4*Aut4
   Automation=~Aut__Aut5*Aut5
   Automation=~Aut__Aut6*Aut6
   Automation=~Aut__Aut7*Aut7
   Facilitating_Conditions=~Facilitating_Conditions__FC1*FC1
   Facilitating_Conditions=~Facilitating_Conditions__FC4*FC4
   Facilitating_Conditions=~Facilitating_Conditions__FC2*FC2
   Facilitating_Conditions=~Facilitating_Conditions__FC3*FC3
   Facilitating_Conditions=~Facilitating_Conditions__Automation*Automation
! residuals, variances and covariances
   Automation ~~ 1.0*Automation
   Aut1 ~~ VAR_Aut1*Aut1
```

```
   Aut2 ~~ VAR_Aut2*Aut2
   Aut3 ~~ VAR_Aut3*Aut3
   Aut4 ~~ VAR_Aut4*Aut4
   Aut5 ~~ VAR_Aut5*Aut5
   Aut6 ~~ VAR_Aut6*Aut6
   Aut7 ~~ VAR_Aut7*Aut7
   Facilitating_Conditions ~~ 1.0*Facilitating_Conditions
   FC1 ~~ VAR_FC1*FC1
   FC4 ~~ VAR_FC4*FC4
   FC2 ~~ VAR_FC2*FC2
   FC3 ~~ VAR_FC3*FC3
! observed means
   Aut1~1;
   Aut2~1;
   Aut3~1;
   Aut4~1;
   Aut5~1;
   Aut6~1;
   Aut7~1;
   FC1~1;
   FC4~1;
   FC2~1;
   FC3~1;
";
result<- sem(model, data=modelData);
summary(result, fit.measures=TRUE);
```

```
## lavaan 0.6-7 ended normally after 33 iterations
##
##   Estimator                                         ML
##   Optimization method                           NLMINB
##   Number of free parameters                         32
##
##   Number of observations                           478
##
## Model Test User Model:
##
##   Test statistic                               574.732
##   Degrees of freedom                                45
##   P-value (Chi-square)                           0.000
##
## Model Test Baseline Model:
##
##   Test statistic                              3117.042
##   Degrees of freedom                                55
##   P-value                                        0.000
##
## User Model versus Baseline Model:
##
##   Comparative Fit Index (CFI)                    0.827
##   Tucker-Lewis Index (TLI)                       0.789
##
## Loglikelihood and Information Criteria:
##
```

```
##   Loglikelihood user model (H0)            -11770.307
##   Loglikelihood unrestricted model (H1)    -11482.941
##
##   Akaike (AIC)                              23604.615
##   Bayesian (BIC)                            23738.042
##   Sample-size adjusted Bayesian (BIC)       23636.478
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                                         0.157
##   90 Percent confidence interval - lower        0.146
##   90 Percent confidence interval - upper        0.169
##   P-value RMSEA <= 0.05                         0.000
##
## Standardized Root Mean Square Residual:
##
##   SRMR                                          0.169
##
## Parameter Estimates:
##
##   Standard errors                            Standard
##   Information                                Expected
##   Information saturated (h1) model         Structured
##
## Latent Variables:
##                               Estimate  Std.Err  z-value  P(>|z|)
##   Automation =~
##     At1    (A__A1)               1.000
##     At2    (A__A2)               1.948    0.109   17.934    0.000
##     At3    (A__A3)               1.828    0.108   16.939    0.000
##     At4    (A__A4)               1.825    0.101   18.047    0.000
##     At5    (A__A5)               2.101    0.108   19.395    0.000
##     At6    (A__A6)               1.769    0.099   17.792    0.000
##     At7    (A__A7)               2.031    0.102   19.997    0.000
##   Facilitating_Conditions =~
##     FC1 (F_C__FC1)               1.000
##     FC4 (F_C__FC4)               1.435    0.126   11.353    0.000
##     FC2 (F_C__FC2)               0.682    0.179    3.815    0.000
##     FC3 (F_C__FC3)               1.323    0.137    9.642    0.000
##     Atm   (F_C__A)               0.597    0.078    7.618    0.000
##
## Intercepts:
##                     Estimate  Std.Err  z-value  P(>|z|)
##    .Aut1               4.977    0.113   44.215    0.000
##    .Aut2               5.073    0.135   37.566    0.000
##    .Aut3               4.339    0.133   32.680    0.000
##    .Aut4               4.228    0.126   33.570    0.000
##    .Aut5               5.103    0.137   37.282    0.000
##    .Aut6               4.105    0.123   33.256    0.000
##    .Aut7               4.757    0.129   36.832    0.000
##    .FC1                7.709    0.101   76.419    0.000
##    .FC4                8.130    0.099   81.897    0.000
##    .FC2                6.816    0.145   47.116    0.000
##    .FC3                7.255    0.112   64.858    0.000
```

```
##    .Automation        0.000
##     Fclttng_Cndtns    0.000
##
## Variances:
##                    Estimate  Std.Err  z-value  P(>|z|)
##    .Atmt              1.000
##    .Aut1  (VAR_A1)    4.700    0.312   15.058    0.000
##    .Aut2  (VAR_A2)    3.573    0.269   13.270    0.000
##    .Aut3  (VAR_A3)    3.895    0.284   13.695    0.000
##    .Aut4  (VAR_A4)    3.062    0.232   13.215    0.000
##    .Aut5  (VAR_A5)    2.966    0.240   12.377    0.000
##    .Aut6  (VAR_A6)    3.039    0.228   13.338    0.000
##    .Aut7  (VAR_A7)    2.379    0.201   11.864    0.000
##     Fc_C              1.000
##    .FC1  (VAR_FC1)    3.865    0.277   13.931    0.000
##    .FC4  (VAR_FC4)    2.651    0.310    8.561    0.000
##    .FC2  (VAR_FC2)    9.538    0.634   15.040    0.000
##    .FC3  (VAR_FC3)    4.230    0.364   11.629    0.000
```

```
standardizedSolution(result)
```

```
##                         lhs op                     rhs est.std    se      z
## 1                Automation =~                    Aut1   0.473 0.018 26.664
## 2                Automation =~                    Aut2   0.768 0.021 36.151
## 3                Automation =~                    Aut3   0.733 0.024 31.167
## 4                Automation =~                    Aut4   0.772 0.021 36.779
## 5                Automation =~                    Aut5   0.818 0.018 45.619
## 6                Automation =~                    Aut6   0.763 0.022 35.377
## 7                Automation =~                    Aut7   0.838 0.017 50.512
## 8  Facilitating_Conditions =~                     FC1   0.453 0.013 35.072
## 9  Facilitating_Conditions =~                     FC4   0.661 0.049 13.474
## 10 Facilitating_Conditions =~                     FC2   0.216 0.055  3.902
## 11 Facilitating_Conditions =~                     FC3   0.541 0.049 11.008
## 12 Facilitating_Conditions =~              Automation   0.513 0.050 10.333
## 13               Automation ~~              Automation   0.737 0.051 14.495
## 14                     Aut1 ~~                    Aut1   0.776 0.017 46.196
## 15                     Aut2 ~~                    Aut2   0.410 0.033 12.553
## 16                     Aut3 ~~                    Aut3   0.462 0.035 13.393
## 17                     Aut4 ~~                    Aut4   0.404 0.032 12.459
## 18                     Aut5 ~~                    Aut5   0.331 0.029 11.301
## 19                     Aut6 ~~                    Aut6   0.417 0.033 12.672
## 20                     Aut7 ~~                    Aut7   0.298 0.028 10.738
## 21 Facilitating_Conditions ~~ Facilitating_Conditions   1.000 0.000     NA
## 22                      FC1 ~~                     FC1   0.794 0.012 67.769
## 23                      FC4 ~~                     FC4   0.563 0.065  8.670
## 24                      FC2 ~~                     FC2   0.953 0.024 39.979
## 25                      FC3 ~~                     FC3   0.707 0.053 13.294
## 26                     Aut1 ~1                           2.022 0.071 28.540
## 27                     Aut2 ~1                           1.718 0.072 24.030
## 28                     Aut3 ~1                           1.495 0.066 22.568
## 29                     Aut4 ~1                           1.535 0.067 22.881
## 30                     Aut5 ~1                           1.705 0.071 23.999
## 31                     Aut6 ~1                           1.521 0.067 22.774
## 32                     Aut7 ~1                           1.685 0.070 23.898
## 33                      FC1 ~1                           3.495 0.110 31.876
```

```
## 34                    FC4 ~1                          3.746 0.129 29.095
## 35                    FC2 ~1                          2.155 0.083 25.851
## 36                    FC3 ~1                          2.967 0.106 27.978
## 37             Automation ~1                          0.000 0.000     NA
## 38 Facilitating_Conditions ~1                         0.000 0.000     NA
##    pvalue ci.lower ci.upper
## 1       0    0.438    0.508
## 2       0    0.727    0.810
## 3       0    0.687    0.779
## 4       0    0.731    0.813
## 5       0    0.783    0.853
## 6       0    0.721    0.806
## 7       0    0.805    0.870
## 8       0    0.428    0.479
## 9       0    0.565    0.757
## 10      0    0.107    0.324
## 11      0    0.445    0.637
## 12      0    0.415    0.610
## 13      0    0.638    0.837
## 14      0    0.743    0.809
## 15      0    0.346    0.474
## 16      0    0.395    0.530
## 17      0    0.340    0.467
## 18      0    0.274    0.389
## 19      0    0.353    0.482
## 20      0    0.244    0.353
## 21     NA    1.000    1.000
## 22      0    0.771    0.817
## 23      0    0.436    0.690
## 24      0    0.907    1.000
## 25      0    0.603    0.811
## 26      0    1.883    2.161
## 27      0    1.578    1.858
## 28      0    1.365    1.625
## 29      0    1.404    1.667
## 30      0    1.566    1.845
## 31      0    1.390    1.652
## 32      0    1.546    1.823
## 33      0    3.280    3.710
## 34      0    3.494    3.998
## 35      0    1.992    2.318
## 36      0    2.759    3.174
## 37     NA    0.000    0.000
## 38     NA    0.000    0.000
```

```
inspect(result, "rsquare")
```

```
## Automation       Aut1       Aut2       Aut3       Aut4       Aut5       Aut6
##      0.263      0.224      0.590      0.538      0.596      0.669      0.583
##       Aut7        FC1        FC4        FC2        FC3
##      0.702      0.206      0.437      0.047      0.293
```

```
library(semPlot)
semPaths(result, whatLabels = "std", layout = "spring", color = list(
              lat = rgb(255, 100, 118, maxColorValue = 255),
```

```r
            man = rgb(155, 253, 175, maxColorValue = 255)),
        mar = c(10, 5, 10, 5), intercepts = FALSE, residuls = FALSE, nCharNodes = 0)
```

```
## Warning in qgraph::qgraph(Edgelist, labels = nLab, bidirectional = Bidir, : The
## following arguments are not documented and likely not arguments of qgraph and
## thus ignored: residuls
```