

A Mild Introduction to Structural Equation Modeling Using lavaan

UseR! Workshop

28 May 2020

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Data preparation

Install and load relevant R packages

```
# Install R packages (if needed)
# install.packages(c("lavaan", "semPlot", "MPSychoR", "corrplot"))

## Load relevant libraries
library(lavaan)
library(semPlot)
library(MPSychoR)
library(corrplot)
```

Data input

```
# Select the data
data("Bergh")
View(Bergh)
attach(Bergh)

# Sample size
nrow(Bergh)

## [1] 861

## Create mean scores per construct
Bergh$Open <- (O1+O2+O3)/3
Bergh$Agree <- (A1+A2+A3)/3
Bergh$Prejudice <- (EP+SP+DP+HP)/4
```

Model 1: Regression model with manifest variables only

```
# Step 1: Model specification
model1 <- '
    # Structural model
    Prejudice ~ b1*Open + b2*Agree

    # Covariance structure of exogenous variables
    Open ~~ Open + Agree
    Agree ~~ Agree
'

# Step 2: Model estimation
model1.fit <- sem(model1,
    data = Bergh,
    meanstructure = FALSE,
    estimator = "ML")

# Step 3: Evaluate the model
```

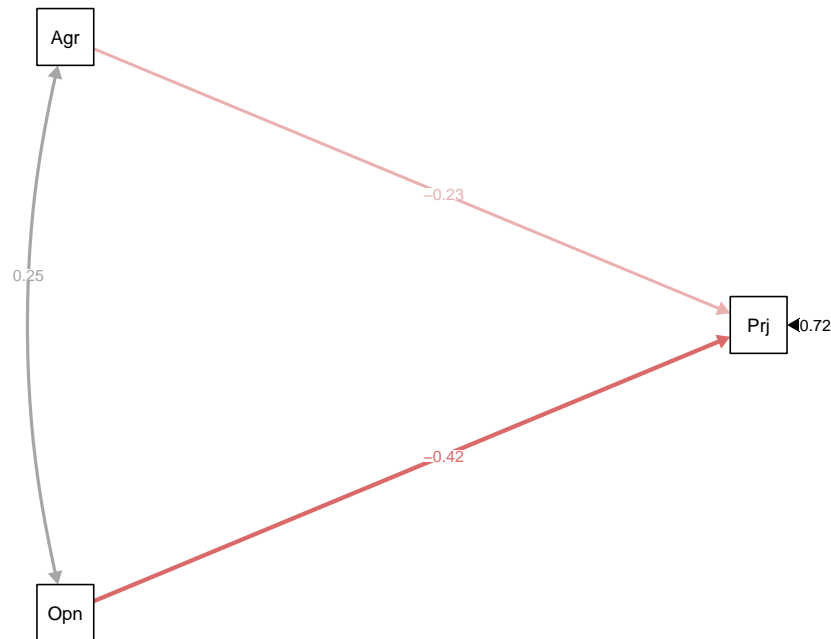
```
# Summary
summary(model1.fit, rsquare = TRUE, fit.measures = TRUE, standardized = TRUE)
```

```
## lavaan 0.6-5 ended normally after 21 iterations
##
##   Estimator                      ML
##   Optimization method          NLMINB
##   Number of free parameters      6
##
##   Number of observations          861
##
## Model Test User Model:
##
##   Test statistic                  0.000
##   Degrees of freedom              0
##
## Model Test Baseline Model:
##
##   Test statistic                  335.486
##   Degrees of freedom              3
##   P-value                        0.000
##
## User Model versus Baseline Model:
##
##   Comparative Fit Index (CFI)      1.000
##   Tucker-Lewis Index (TLI)        1.000
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)    -1689.786
##   Loglikelihood unrestricted model (H1) -1689.786
##
##   Akaike (AIC)                    3391.572
##   Bayesian (BIC)                   3420.121
##   Sample-size adjusted Bayesian (BIC) 3401.066
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                          0.000
##   90 Percent confidence interval - lower 0.000
##   90 Percent confidence interval - upper 0.000
##   P-value RMSEA <= 0.05            NA
##
## Standardized Root Mean Square Residual:
##
##   SRMR                          0.000
##
## Parameter Estimates:
##
##   Information                      Expected
##   Information saturated (h1) model Structured
##   Standard errors                  Standard
##
## Regressions:
```

```
##               Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## Prejudice ~
##   Open      (b1)  -0.612   0.043 -14.118   0.000  -0.612  -0.423
##   Agree      (b2)  -0.324   0.043  -7.522   0.000  -0.324  -0.225
##
## Covariances:
##               Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##   Open ~~
##   Agree       0.049    0.007   7.148   0.000   0.049   0.251
##
## Variances:
##               Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##   Open        0.192    0.009  20.748   0.000   0.192   1.000
##   Agree        0.194    0.009  20.748   0.000   0.194   1.000
##   .Prejudice   0.291    0.014  20.748   0.000   0.291   0.723
##
## R-Square:
##               Estimate
##   Prejudice     0.277
```

```
# Visualize the path model
```

```
semPaths(model1.fit,
  rotation = 2,
  layout = "tree2",
  what = "std",
  posCol = "black",
  edge.width = 0.5,
  style = "Lisrel",
  fade = T,
  edge.label.position = 0.55)
```



```
# Global fit indices
```

```
fitMeasures(model1.fit)
```

```
##               npar               fmin               chisq               df
```

```
##          6.000          0.000          0.000          0.000
##          pvalue      baseline.chisq      baseline.df      baseline.pvalue
##          NA          335.486          3.000          0.000
##          cfi          tli          nnfi          rfi
##          1.000          1.000          1.000          1.000
##          nfi          pnfi          ifi          rni
##          1.000          0.000          1.000          1.000
##          logl      unrestricted.logl          aic          bic
##          -1689.786      -1689.786          3391.572          3420.121
##          ntotal          bic2          rmsea      rmsea.ci.lower
##          861.000          3401.066          0.000          0.000
##          rmsea.ci.upper      rmsea.pvalue          rmr          rmr_nomean
##          0.000          NA          0.000          0.000
##          srmr          srmr_bentler      srmr_bentler_nomean          crmr
##          0.000          0.000          0.000          0.000
##          crmr_nomean          srmr_mplus      srmr_mplus_nomean          cn_05
##          0.000          0.000          0.000          1.000
##          cn_01          gfi          agfi          pgfi
##          1.000          1.000          1.000          0.000
##          mfi          ecvi
##          1.000          0.014
```

```
# Local fit measures
```

```
modificationindices(model1.fit)
```

```
## [1] lhs      op      rhs      mi      epc      sepc.lv  sepc.all sepc.noX
## <0 rows> (or 0-length row.names)
```

```
resid(model1.fit, type = "raw")
```

```
## $type
## [1] "raw"
##
## $cov
##          Prejdc Open Agree
## Prejudice 0
## Open      0      0
## Agree     0      0      0
```

```
resid(model1.fit, type = "standardized")
```

```
## $type
## [1] "standardized"
##
## $cov
##          Prejdc Open Agree
## Prejudice 0
## Open      0      0
## Agree     0      0      0
```

```
# Fitted values of the covariance matrix and the mean vector
```

```
fitted(model1.fit)
```

```
## $cov
##          Prejdc Open  Agree
## Prejudice 0.402
## Open      -0.133  0.192
```

```
## Agree      -0.093  0.049  0.194
```

```
# List all parameter values  
parameterEstimates(modell1.fit)
```

##	lhs	op	rhs	label	est	se	z	pvalue	ci.lower	ci.upper
## 1	Prejudice	~	Open	b1	-0.612	0.043	-14.118	0	-0.697	-0.527
## 2	Prejudice	~	Agree	b2	-0.324	0.043	-7.522	0	-0.408	-0.239
## 3	Open	~~	Open		0.192	0.009	20.748	0	0.174	0.210
## 4	Open	~~	Agree		0.049	0.007	7.148	0	0.035	0.062
## 5	Agree	~~	Agree		0.194	0.009	20.748	0	0.176	0.213
## 6	Prejudice	~~	Prejudice		0.291	0.014	20.748	0	0.263	0.318

```
# Check the model setup  
inspect(modell1.fit)
```

```
## $lambda  
##           Prejdc Open Agree  
## Prejudice      0      0      0  
## Open           0      0      0  
## Agree          0      0      0  
##  
## $theta  
##           Prejdc Open Agree  
## Prejudice 0  
## Open      0      0  
## Agree     0      0      0  
##  
## $psi  
##           Prejdc Open Agree  
## Prejudice 6  
## Open      0      3  
## Agree     0      4      5  
##  
## $beta  
##           Prejdc Open Agree  
## Prejudice 0      1      2  
## Open      0      0      0  
## Agree     0      0      0
```

```
# Check the starting values  
inspect(modell1.fit, what = "start")
```

```
## $lambda  
##           Prejdc Open Agree  
## Prejudice      1      0      0  
## Open           0      1      0  
## Agree          0      0      1  
##  
## $theta  
##           Prejdc Open Agree  
## Prejudice 0  
## Open      0      0  
## Agree     0      0      0  
##  
## $psi
```

```
##           Prejdc Open  Agree
## Prejudice 0.201
## Open      0.000  0.096
## Agree     0.000  0.000 0.097
```

```
##
## $beta
##           Prejdc Open Agree
## Prejudice      0    0    0
## Open           0    0    0
## Agree          0    0    0
```

```
# Check the estimates
inspect(model1.fit, what = "est")
```

```
## $lambda
##           Prejdc Open Agree
## Prejudice      1    0    0
## Open           0    1    0
## Agree          0    0    1
```

```
##
## $theta
##           Prejdc Open Agree
## Prejudice 0
## Open      0    0
## Agree     0    0    0
```

```
##
## $psi
##           Prejdc Open  Agree
## Prejudice 0.291
## Open      0.000  0.192
## Agree     0.000  0.049 0.194
```

```
##
## $beta
##           Prejdc  Open  Agree
## Prejudice      0 -0.612 -0.324
## Open           0  0.000  0.000
## Agree          0  0.000  0.000
```

```
# Step 4: Further hypothesis testing
# H0: b1=b2
lavTestWald(model1.fit, constraints = "b1==b2")
```

```
## $stat
## [1] 17.76479
##
## $df
## [1] 1
##
## $p.value
## [1] 2.499661e-05
##
## $se
## [1] "standard"
```

Model 1 with bootstrapping of standard errors

```
# Step 2: Model estimation with bootstrapping
set.seed(616)
modell1.fit.boot <- sem(modell1,
  data = Bergh,
  meanstructure = FALSE,
  estimator = "ML",
  se = "bootstrap",
  bootstrap = 1000)

# Step 3: Evaluate the model
# Summary
summary(modell1.fit.boot,
  rsquare = TRUE,
  fit.measures = TRUE,
  standardized = TRUE,
  ci = TRUE)
```

lavaan 0.6-5 ended normally after 21 iterations

##

## Estimator	ML
## Optimization method	NLMINB
## Number of free parameters	6
##	
## Number of observations	861
##	

Model Test User Model:

##

## Test statistic	0.000
## Degrees of freedom	0
##	

Model Test Baseline Model:

##

## Test statistic	335.486
## Degrees of freedom	3
## P-value	0.000
##	

User Model versus Baseline Model:

##

## Comparative Fit Index (CFI)	1.000
## Tucker-Lewis Index (TLI)	1.000
##	

Loglikelihood and Information Criteria:

##

## Loglikelihood user model (H0)	-1689.786
## Loglikelihood unrestricted model (H1)	-1689.786
##	
## Akaike (AIC)	3391.572
## Bayesian (BIC)	3420.121
## Sample-size adjusted Bayesian (BIC)	3401.066
##	

Root Mean Square Error of Approximation:


```

##
## RMSEA 0.000
## 90 Percent confidence interval - lower 0.000
## 90 Percent confidence interval - upper 0.000
## P-value RMSEA <= 0.05 NA
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.000
##
## Parameter Estimates:
##
## Standard errors Bootstrap
## Number of requested bootstrap draws 1000
## Number of successful bootstrap draws 1000
##
## Regressions:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## Prejudice ~
## Open (b1) -0.612 0.044 -13.905 0.000 -0.704 -0.528
## Agree (b2) -0.324 0.043 -7.576 0.000 -0.410 -0.238
## Std.lv Std.all
##
## -0.612 -0.423
## -0.324 -0.225
##
## Covariances:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## Open ~~
## Agree 0.049 0.007 6.802 0.000 0.034 0.061
## Std.lv Std.all
##
## 0.049 0.251
##
## Variances:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## Open 0.192 0.009 22.345 0.000 0.174 0.209
## Agree 0.194 0.008 24.159 0.000 0.177 0.211
## .Prejudice 0.291 0.017 17.440 0.000 0.258 0.323
## Std.lv Std.all
## 0.192 1.000
## 0.194 1.000
## 0.291 0.723
##
## R-Square:
## Estimate
## Prejudice 0.277

```

List all parameter values

```

parameterEstimates(model1.fit.boot, ci = TRUE, boot.ci.type = "basic")

```

	lhs	op	rhs	label	est	se	z	pvalue	ci.lower	ci.upper
## 1	Prejudice	~	Open	b1	-0.612	0.044	-13.905	0	-0.696	-0.520
## 2	Prejudice	~	Agree	b2	-0.324	0.043	-7.576	0	-0.409	-0.238
## 3	Open	~~	Open		0.192	0.009	22.345	0	0.175	0.210

## 4	Open	~~	Agree	0.049	0.007	6.802	0	0.036	0.063
## 5	Agree	~~	Agree	0.194	0.008	24.159	0	0.178	0.212
## 6	Prejudice	~~	Prejudice	0.291	0.017	17.440	0	0.258	0.324

Model 2: Mediation model with manifest variables only (T)

```
# Step 1: Model specification
model2 <- '
    # Structural model
    Prejudice ~ b1*Open + b2*Agree
    Open ~ b3*Agree

    # Covariance structure of exogenous variables
    Agree ~~ Agree

    # New parameters (indirect and total effect)
    ind := b1*b3
    tot := ind + b2
'

# Step 2: Model estimation
model2.fit <- sem(model2,
    data = Bergh,
    meanstructure = FALSE,
    estimator = "ML")

# Step 3: Evaluate the model
# Summary
summary(model2.fit, rsquare = TRUE, fit.measures = TRUE, standardized = TRUE)
```

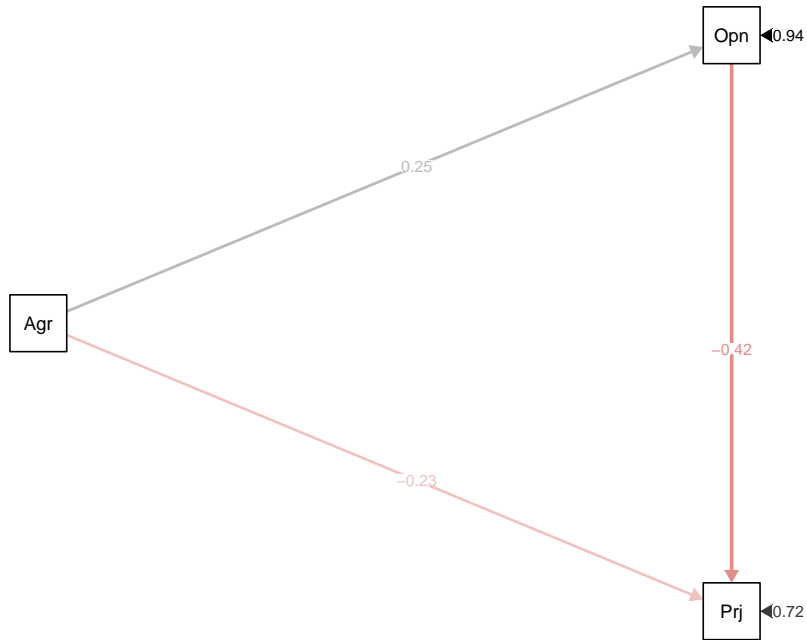
```
## lavaan 0.6-5 ended normally after 19 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of free parameters      6
##
##      Number of observations          861
##
## Model Test User Model:
##
##      Test statistic                  0.000
##      Degrees of freedom              0
##
## Model Test Baseline Model:
##
##      Test statistic                  335.486
##      Degrees of freedom              3
##      P-value                        0.000
##
## User Model versus Baseline Model:
##
```

```

## Comparative Fit Index (CFI) 1.000
## Tucker-Lewis Index (TLI) 1.000
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -1689.786
## Loglikelihood unrestricted model (H1) -1689.786
##
## Akaike (AIC) 3391.572
## Bayesian (BIC) 3420.121
## Sample-size adjusted Bayesian (BIC) 3401.066
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.000
## 90 Percent confidence interval - lower 0.000
## 90 Percent confidence interval - upper 0.000
## P-value RMSEA <= 0.05 NA
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.000
##
## Parameter Estimates:
##
## Information Expected
## Information saturated (h1) model Structured
## Standard errors Standard
##
## Regressions:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## Prejudice ~
## Open (b1) -0.612 0.043 -14.118 0.000 -0.612 -0.423
## Agree (b2) -0.324 0.043 -7.522 0.000 -0.324 -0.225
## Open ~
## Agree (b3) 0.250 0.033 7.614 0.000 0.250 0.251
##
## Variances:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## Agree 0.194 0.009 20.748 0.000 0.194 1.000
## .Prejudice 0.291 0.014 20.748 0.000 0.291 0.723
## .Open 0.180 0.009 20.748 0.000 0.180 0.937
##
## R-Square:
## Estimate
## Prejudice 0.277
## Open 0.063
##
## Defined Parameters:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## ind -0.153 0.023 -6.701 0.000 -0.153 -0.106
## tot -0.477 0.046 -10.304 0.000 -0.477 -0.331

```

```
# Visualize the path model
semPaths(model2.fit,
  rotation = 2,
  layout = "tree2",
  what = "std",
  posCol = "black",
  edge.width = 0.5,
  style = "Lisrel",
  fade = T,
  edge.label.position = 0.55)
```



Correlation matrix

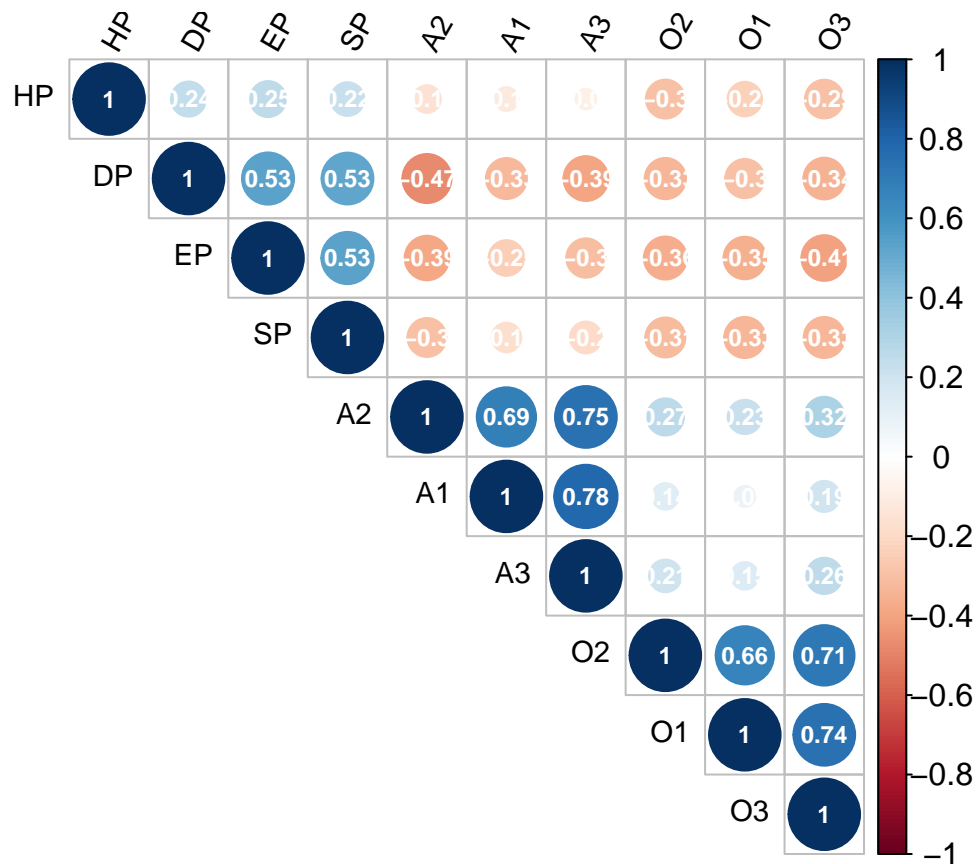
```
# Extract the correlation matrix
```

```
Bergh.cor <- cor(Bergh[,1:10], method = "pearson", use = "pairwise.complete.obs")  
Bergh.cor
```

```
##           EP           SP           HP           DP           A1           A2           A3  
## EP  1.0000000  0.5328577  0.2545270  0.5314828 -0.2486889 -0.3889079 -0.3031269  
## SP  0.5328577  1.0000000  0.2219292  0.5252140 -0.1710822 -0.2973829 -0.1987969  
## HP  0.2545270  0.2219292  1.0000000  0.2415626 -0.1120012 -0.1510590 -0.0827062  
## DP  0.5314828  0.5252140  0.2415626  1.0000000 -0.3292610 -0.4709318 -0.3936544  
## A1 -0.2486889 -0.1710822 -0.1120012 -0.3292610  1.0000000  0.6867541  0.7835360  
## A2 -0.3889079 -0.2973829 -0.1510590 -0.4709318  0.6867541  1.0000000  0.7453925  
## A3 -0.3031269 -0.1987969 -0.0827062 -0.3936544  0.7835360  0.7453925  1.0000000  
## O1 -0.3543605 -0.3317130 -0.2332906 -0.2994080  0.0861290  0.2293831  0.1488831  
## O2 -0.3622272 -0.3127873 -0.2972669 -0.3327277  0.1393367  0.2698570  0.2082816  
## O3 -0.4089230 -0.3300734 -0.2930209 -0.3407396  0.1904259  0.3178221  0.2584276  
##           O1           O2           O3  
## EP -0.3543605 -0.3622272 -0.4089230  
## SP -0.3317130 -0.3127873 -0.3300734  
## HP -0.2332906 -0.2972669 -0.2930209  
## DP -0.2994080 -0.3327277 -0.3407396  
## A1  0.0861290  0.1393367  0.1904259  
## A2  0.2293831  0.2698570  0.3178221  
## A3  0.1488831  0.2082816  0.2584276  
## O1  1.0000000  0.6624692  0.7444363  
## O2  0.6624692  1.0000000  0.7140617  
## O3  0.7444363  0.7140617  1.0000000
```

```
# Correlogram
```

```
corrplot(Bergh.cor, type = "upper", order = "hclust",  
          tl.col = "black", tl.srt = 60,  
          addCoef.col = "white",  
          number.cex = 0.75,  
          cl.cex = 1,  
          tl.cex = 0.9)
```



Model 3: Measurement model (CFA)

```
# Step 1: Model specification
model3 <- '
    # Measurement models
    OP =~ O1 + O2 + O3
    AG =~ A1 + A2 + A3
    PR =~ EP + SP + HP + DP

    # Covariance structure
    OP ~~ OP + AG + PR
    AG ~~ AG + PR
    PR ~~ PR

'

# Step 2: Model estimation
model3.fit <- sem(model3,
  data = Bergh,
  meanstructure = FALSE,
  estimator = "ML")

# Step 3: Evaluate the model
# Summary
```

```
summary(model3.fit, rsquare = TRUE, fit.measures = TRUE, standardized = TRUE)
```

```
## lavaan 0.6-5 ended normally after 54 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of free parameters      23
##
##      Number of observations          861
##
## Model Test User Model:
##
##      Test statistic                  186.620
##      Degrees of freedom              32
##      P-value (Chi-square)            0.000
##
## Model Test Baseline Model:
##
##      Test statistic                  4270.205
##      Degrees of freedom              45
##      P-value                         0.000
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)      0.963
##      Tucker-Lewis Index (TLI)        0.949
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)      -5672.807
##      Loglikelihood unrestricted model (H1) -5579.497
##
##      Akaike (AIC)                     11391.614
##      Bayesian (BIC)                    11501.050
##      Sample-size adjusted Bayesian (BIC) 11428.008
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                            0.075
##      90 Percent confidence interval - lower 0.065
##      90 Percent confidence interval - upper 0.085
##      P-value RMSEA <= 0.05              0.000
##
## Standardized Root Mean Square Residual:
##
##      SRMR                             0.054
##
## Parameter Estimates:
##
##      Information                      Expected
##      Information saturated (h1) model  Structured
##      Standard errors                   Standard
##
## Latent Variables:
```

```

##               Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## OP =~
##   O1           1.000           0.036 26.185  0.000   0.400   0.827
##   O2           0.934           0.036 26.185  0.000   0.374   0.799
##   O3           1.149           0.040 28.900  0.000   0.460   0.898
## AG =~
##   A1           1.000           0.032 28.812  0.000   0.426   0.846
##   A2           0.910           0.032 28.812  0.000   0.388   0.823
##   A3           1.030           0.032 31.899  0.000   0.439   0.914
## PR =~
##   EP           1.000           0.051 17.348  0.000   0.530   0.746
##   SP           0.886           0.051 17.348  0.000   0.469   0.686
##   HP           1.030           0.112  9.160  0.000   0.545   0.350
##   DP           0.746           0.041 18.308  0.000   0.395   0.741
##
## Covariances:
##               Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## OP ~~
##   AG           0.049           0.007  7.105  0.000   0.286   0.286
##   PR          -0.122           0.011 -11.371  0.000  -0.573  -0.573
## AG ~~
##   PR          -0.110           0.011 -10.241  0.000  -0.485  -0.485
##
## Variances:
##               Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## OP           0.160           0.011 14.156  0.000   1.000   1.000
## AG           0.182           0.012 14.822  0.000   1.000   1.000
## PR           0.281           0.025 11.385  0.000   1.000   1.000
## .O1          0.074           0.005 14.555  0.000   0.074   0.317
## .O2          0.079           0.005 15.837  0.000   0.079   0.361
## .O3          0.051           0.005  9.630  0.000   0.051   0.194
## .A1          0.072           0.005 14.461  0.000   0.072   0.284
## .A2          0.072           0.005 15.697  0.000   0.072   0.322
## .A3          0.038           0.004  9.152  0.000   0.038   0.165
## .EP          0.224           0.016 14.198  0.000   0.224   0.444
## .SP          0.248           0.015 16.146  0.000   0.248   0.530
## .HP          2.137           0.107 20.052  0.000   2.137   0.878
## .DP          0.128           0.009 14.376  0.000   0.128   0.451
##
## R-Square:
##               Estimate
##   O1           0.683
##   O2           0.639
##   O3           0.806
##   A1           0.716
##   A2           0.678
##   A3           0.835
##   EP           0.556
##   SP           0.470
##   HP           0.122
##   DP           0.549

```

```

# Global fit indices
fitMeasures(model3.fit)

```



```
##          npar          fmin          chisq          df
##          23.000          0.108          186.620          32.000
##          pvalue      baseline.chisq      baseline.df      baseline.pvalue
##          0.000          4270.205          45.000          0.000
##          cfi          tli          nnfi          rfi
##          0.963          0.949          0.949          0.939
##          nfi          pnfi          ifi          rni
##          0.956          0.680          0.964          0.963
##          logl      unrestricted.logl          aic          bic
##          -5672.807          -5579.497          11391.614          11501.050
##          ntotal          bic2          rmsea      rmsea.ci.lower
##          861.000          11428.008          0.075          0.065
##          rmsea.ci.upper      rmsea.pvalue          rmr          rmr_nomean
##          0.085          0.000          0.025          0.025
##          srmr          srmr_bentler srmr_bentler_nomean          crmr
##          0.054          0.054          0.054          0.060
##          crmr_nomean          srmr_mplus      srmr_mplus_nomean          cn_05
##          0.060          0.054          0.054          214.124
##          cn_01          gfi          agfi          pgfi
##          247.764          0.959          0.930          0.558
##          mfi          ecvi
##          0.914          0.270
```

```
# Local fit measures
modindices(model3.fit)
```

```
##      lhs op rhs      mi      epc sepc.lv sepc.all sepc.nox
## 27 OP =~ A1 19.169 -0.131 -0.052 -0.104 -0.104
## 28 OP =~ A2 34.333 0.169 0.068 0.144 0.144
## 29 OP =~ A3 1.024 -0.028 -0.011 -0.023 -0.023
## 30 OP =~ EP 2.036 -0.109 -0.044 -0.061 -0.061
## 31 OP =~ SP 0.446 0.048 0.019 0.028 0.028
## 32 OP =~ HP 22.203 -0.842 -0.337 -0.216 -0.216
## 33 OP =~ DP 8.316 0.165 0.066 0.124 0.124
## 34 AG =~ O1 15.221 -0.111 -0.047 -0.098 -0.098
## 35 AG =~ O2 0.098 0.009 0.004 0.008 0.008
## 36 AG =~ O3 10.358 0.094 0.040 0.078 0.078
## 37 AG =~ EP 0.373 0.038 0.016 0.023 0.023
## 38 AG =~ SP 22.259 0.283 0.121 0.176 0.176
## 39 AG =~ HP 2.920 0.258 0.110 0.070 0.070
## 40 AG =~ DP 32.769 -0.268 -0.114 -0.214 -0.214
## 41 PR =~ O1 2.227 0.047 0.025 0.052 0.052
## 42 PR =~ O2 1.570 -0.039 -0.021 -0.044 -0.044
## 43 PR =~ O3 0.077 -0.009 -0.005 -0.010 -0.010
## 44 PR =~ A1 19.206 0.121 0.064 0.127 0.127
## 45 PR =~ A2 68.775 -0.220 -0.117 -0.247 -0.247
## 46 PR =~ A3 9.718 0.080 0.042 0.088 0.088
## 47 O1 ~~ O2 0.123 0.002 0.002 0.031 0.031
## 48 O1 ~~ O3 1.554 0.012 0.012 0.192 0.192
## 49 O1 ~~ A1 3.654 -0.006 -0.006 -0.085 -0.085
## 50 O1 ~~ A2 0.940 0.003 0.003 0.042 0.042
## 51 O1 ~~ A3 2.244 -0.004 -0.004 -0.080 -0.080
## 52 O1 ~~ EP 0.021 0.001 0.001 0.006 0.006
## 53 O1 ~~ SP 5.349 -0.013 -0.013 -0.099 -0.099
## 54 O1 ~~ HP 0.749 0.014 0.014 0.034 0.034
```

```

## 55 01 ~~ DP 1.246 0.005 0.005 0.050 0.050
## 56 02 ~~ 03 2.451 -0.013 -0.013 -0.208 -0.208
## 57 02 ~~ A1 1.256 -0.004 -0.004 -0.048 -0.048
## 58 02 ~~ A2 0.448 0.002 0.002 0.028 0.028
## 59 02 ~~ A3 0.109 0.001 0.001 0.017 0.017
## 60 02 ~~ EP 0.169 0.002 0.002 0.018 0.018
## 61 02 ~~ SP 0.109 0.002 0.002 0.014 0.014
## 62 02 ~~ HP 7.969 -0.044 -0.044 -0.108 -0.108
## 63 02 ~~ DP 0.908 -0.004 -0.004 -0.042 -0.042
## 64 03 ~~ A1 0.261 0.002 0.002 0.027 0.027
## 65 03 ~~ A2 0.121 0.001 0.001 0.018 0.018
## 66 03 ~~ A3 1.840 0.004 0.004 0.085 0.085
## 67 03 ~~ EP 1.762 -0.008 -0.008 -0.071 -0.071
## 68 03 ~~ SP 4.455 0.012 0.012 0.108 0.108
## 69 03 ~~ HP 3.172 -0.027 -0.027 -0.083 -0.083
## 70 03 ~~ DP 2.594 0.007 0.007 0.086 0.086
## 71 A1 ~~ A2 9.573 -0.025 -0.025 -0.344 -0.344
## 72 A1 ~~ A3 69.141 0.088 0.088 1.688 1.688
## 73 A1 ~~ EP 2.134 0.008 0.008 0.066 0.066
## 74 A1 ~~ SP 1.050 0.006 0.006 0.044 0.044
## 75 A1 ~~ HP 2.147 -0.023 -0.023 -0.058 -0.058
## 76 A1 ~~ DP 0.414 0.003 0.003 0.029 0.029
## 77 A2 ~~ A3 19.619 -0.041 -0.041 -0.778 -0.778
## 78 A2 ~~ EP 4.065 -0.011 -0.011 -0.088 -0.088
## 79 A2 ~~ SP 1.918 -0.008 -0.008 -0.058 -0.058
## 80 A2 ~~ HP 0.368 -0.009 -0.009 -0.023 -0.023
## 81 A2 ~~ DP 9.809 -0.013 -0.013 -0.137 -0.137
## 82 A3 ~~ EP 0.513 0.004 0.004 0.039 0.039
## 83 A3 ~~ SP 8.962 0.015 0.015 0.155 0.155
## 84 A3 ~~ HP 7.936 0.038 0.038 0.133 0.133
## 85 A3 ~~ DP 1.665 -0.005 -0.005 -0.070 -0.070
## 86 EP ~~ SP 5.328 0.033 0.033 0.140 0.140
## 87 EP ~~ HP 0.132 -0.011 -0.011 -0.016 -0.016
## 88 EP ~~ DP 8.308 -0.035 -0.035 -0.205 -0.205
## 89 SP ~~ HP 0.805 -0.026 -0.026 -0.036 -0.036
## 90 SP ~~ DP 3.206 0.019 0.019 0.107 0.107
## 91 HP ~~ DP 1.034 -0.023 -0.023 -0.043 -0.043

```

```
resid(model3.fit)
```

```

## $type
## [1] "raw"
##
## $cov
##      01      02      03      A1      A2      A3      EP      SP      HP      DP
## 01  0.000
## 02  0.000  0.000
## 03  0.001 -0.001  0.000
## A1 -0.028 -0.013 -0.007  0.000
## A2  0.008  0.018  0.026 -0.002  0.000
## A3 -0.016  0.000  0.006  0.002 -0.002  0.000
## EP  0.000 -0.007 -0.009  0.021 -0.030  0.009  0.000
## SP -0.002  0.000  0.008  0.038 -0.008  0.035  0.010  0.000
## HP -0.051 -0.100 -0.090  0.025 -0.008  0.054 -0.007 -0.019  0.000
## DP  0.013  0.002  0.011 -0.007 -0.044 -0.017 -0.008  0.006 -0.015  0.000

```

```
resid(model3.fit, type = "standardized")
```

```
## $type
## [1] "standardized"
##
## $cov
##      01      02      03      A1      A2      A3      EP      SP      HP      DP
## 01  0.000
## 02  0.349  0.000
## 03  1.225 -1.598  0.000
## A1 -5.521 -2.482 -1.555  0.000
## A2  1.621  3.639  5.588 -3.240  0.000
## A3 -3.822 -0.018  1.838  7.510 -4.704  0.000
## EP -0.052 -1.051 -1.625  2.934 -4.450  1.638  0.000
## SP -0.330  0.074  1.307  5.177 -1.085  5.615  2.165  0.000
## HP -2.440 -4.947 -4.275  1.085 -0.421  2.595 -0.366 -0.904  0.000
## DP  2.764  0.352  2.630 -1.239 -8.406 -3.636 -3.017  1.694 -1.032  0.000
```

```
# Fitted values of the covariance matrix and the mean vector
fitted(model3.fit)
```

```
## $cov
##      01      02      03      A1      A2      A3      EP      SP      HP      DP
## 01  0.235
## 02  0.150  0.219
## 03  0.184  0.172  0.263
## A1  0.049  0.045  0.056  0.254
## A2  0.044  0.041  0.051  0.165  0.222
## A3  0.050  0.047  0.058  0.187  0.170  0.231
## EP -0.122 -0.114 -0.140 -0.110 -0.100 -0.113  0.505
## SP -0.108 -0.101 -0.124 -0.097 -0.088 -0.100  0.249  0.468
## HP -0.125 -0.117 -0.144 -0.113 -0.103 -0.116  0.289  0.256  2.434
## DP -0.091 -0.085 -0.104 -0.082 -0.074 -0.084  0.209  0.185  0.216  0.285
```

```
# List all parameter values
parameterEstimates(model3.fit)
```

```
##      lhs op rhs      est      se      z pvalue ci.lower ci.upper
## 1  OP == 01  1.000 0.000      NA      NA      1.000      1.000
## 2  OP == 02  0.934 0.036 26.185      0      0.864      1.004
## 3  OP == 03  1.149 0.040 28.900      0      1.071      1.227
## 4  AG == A1  1.000 0.000      NA      NA      1.000      1.000
## 5  AG == A2  0.910 0.032 28.812      0      0.848      0.972
## 6  AG == A3  1.030 0.032 31.899      0      0.967      1.094
## 7  PR == EP  1.000 0.000      NA      NA      1.000      1.000
## 8  PR == SP  0.886 0.051 17.348      0      0.786      0.986
## 9  PR == HP  1.030 0.112  9.160      0      0.809      1.250
## 10 PR == DP  0.746 0.041 18.308      0      0.666      0.826
## 11 OP == OP  0.160 0.011 14.156      0      0.138      0.182
## 12 OP == AG  0.049 0.007  7.105      0      0.035      0.062
## 13 OP == PR -0.122 0.011 -11.371      0     -0.143     -0.101
## 14 AG == AG  0.182 0.012 14.822      0      0.158      0.206
## 15 AG == PR -0.110 0.011 -10.241      0     -0.131     -0.089
## 16 PR == PR  0.281 0.025 11.385      0      0.232      0.329
## 17 01 == 01  0.074 0.005 14.555      0      0.064      0.084
```

```
## 18 O2 ~~ O2 0.079 0.005 15.837 0 0.069 0.089
## 19 O3 ~~ O3 0.051 0.005 9.630 0 0.041 0.061
## 20 A1 ~~ A1 0.072 0.005 14.461 0 0.062 0.082
## 21 A2 ~~ A2 0.072 0.005 15.697 0 0.063 0.081
## 22 A3 ~~ A3 0.038 0.004 9.152 0 0.030 0.046
## 23 EP ~~ EP 0.224 0.016 14.198 0 0.193 0.255
## 24 SP ~~ SP 0.248 0.015 16.146 0 0.218 0.278
## 25 HP ~~ HP 2.137 0.107 20.052 0 1.928 2.346
## 26 DP ~~ DP 0.128 0.009 14.376 0 0.111 0.146
```

```
# Check the model setup
```

```
inspect(model3.fit)
```

```
## $lambda
```

```
##      OP AG PR
```

```
## O1  0  0  0
```

```
## O2  1  0  0
```

```
## O3  2  0  0
```

```
## A1  0  0  0
```

```
## A2  0  3  0
```

```
## A3  0  4  0
```

```
## EP  0  0  0
```

```
## SP  0  0  5
```

```
## HP  0  0  6
```

```
## DP  0  0  7
```

```
##
```

```
## $theta
```

```
##      O1 O2 O3 A1 A2 A3 EP SP HP DP
```

```
## O1 14
```

```
## O2  0 15
```

```
## O3  0  0 16
```

```
## A1  0  0  0 17
```

```
## A2  0  0  0  0 18
```

```
## A3  0  0  0  0  0 19
```

```
## EP  0  0  0  0  0  0 20
```

```
## SP  0  0  0  0  0  0  0 21
```

```
## HP  0  0  0  0  0  0  0  0 22
```

```
## DP  0  0  0  0  0  0  0  0  0 23
```

```
##
```

```
## $psi
```

```
##      OP AG PR
```

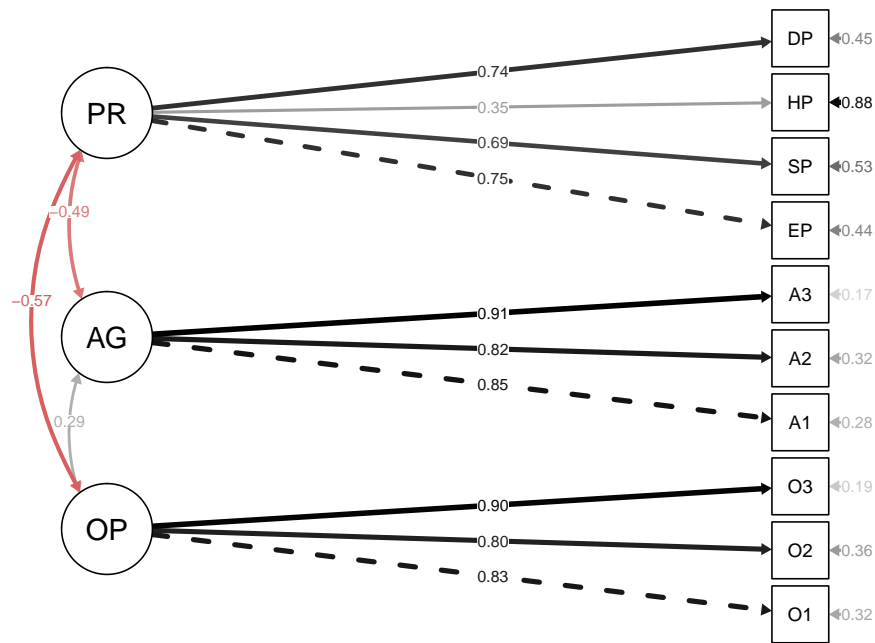
```
## OP  8
```

```
## AG  9 11
```

```
## PR 10 12 13
```

```
# Visualize the path model
```

```
semPaths(model3.fit,
  rotation = 2,
  layout = "tree2",
  what = "std",
  posCol = "black",
  edge.width = 0.5,
  style = "Lisrel",
  fade = T,
  edge.label.position = 0.55)
```



Model 3: Refined CFA of personality (T)

```
# Step 1: Model specification
model3b <- '
    # Measurement models
    OP =~ O1 + O2 + O3
    AG =~ A1 + A2 + A3
    PR =~ EP + SP + HP + DP

    # Covariance structure
    OP ~~ OP + AG + PR
    AG ~~ AG + PR
    PR ~~ PR

    # Residual covariance
    A1 ~~ A3
'

# Step 2: Model estimation
model3b.fit <- sem(model3b,
    data = Bergh,
    meanstructure = FALSE,
    estimator = "ML")

# Step 3: Evaluate the model
# Summary
summary(model3b.fit, rsquare = TRUE, fit.measures = TRUE, standardized = TRUE)

## lavaan 0.6-5 ended normally after 62 iterations
##
```

```

## Estimator ML
## Optimization method NLMINB
## Number of free parameters 24
##
## Number of observations 861
##
## Model Test User Model:
##
## Test statistic 118.256
## Degrees of freedom 31
## P-value (Chi-square) 0.000
##
## Model Test Baseline Model:
##
## Test statistic 4270.205
## Degrees of freedom 45
## P-value 0.000
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI) 0.979
## Tucker-Lewis Index (TLI) 0.970
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -5638.625
## Loglikelihood unrestricted model (H1) -5579.497
##
## Akaike (AIC) 11325.249
## Bayesian (BIC) 11439.444
## Sample-size adjusted Bayesian (BIC) 11363.226
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.057
## 90 Percent confidence interval - lower 0.046
## 90 Percent confidence interval - upper 0.068
## P-value RMSEA <= 0.05 0.131
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.043
##
## Parameter Estimates:
##
## Information Expected
## Information saturated (h1) model Structured
## Standard errors Standard
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## OP =~
## 01 1.000 0.400 0.827
## 02 0.934 0.036 26.188 0.000 0.374 0.799

```

```

##      O3              1.149    0.040   28.921    0.000    0.460    0.898
##    AG =~
##      A1              1.000
##      A2              1.361    0.086   15.756    0.000    0.471    0.999
##      A3              1.036    0.033   31.662    0.000    0.358    0.746
##    PR =~
##      EP              1.000
##      SP              0.887    0.051   17.460    0.000    0.469    0.685
##      HP              1.031    0.112    9.177    0.000    0.545    0.349
##      DP              0.750    0.040   18.535    0.000    0.397    0.744
##
## Covariances:
##              Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##    OP ~~
##      AG              0.046    0.006    7.516    0.000    0.330    0.330
##      PR             -0.121    0.011  -11.372    0.000   -0.573   -0.573
##    AG ~~
##      PR             -0.098    0.010   -9.409    0.000   -0.536   -0.536
##    .A1 ~~
##      .A3              0.066    0.008    8.266    0.000    0.066    0.560
##
## Variances:
##              Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##    OP              0.160    0.011   14.158    0.000    1.000    1.000
##    AG              0.120    0.012    9.879    0.000    1.000    1.000
##    PR              0.279    0.024   11.413    0.000    1.000    1.000
##    .O1              0.074    0.005   14.570    0.000    0.074    0.317
##    .O2              0.079    0.005   15.846    0.000    0.079    0.361
##    .O3              0.051    0.005    9.643    0.000    0.051    0.194
##    .A1              0.134    0.009   14.890    0.000    0.134    0.528
##    .A2              0.000    0.012    0.027    0.979    0.000    0.001
##    .A3              0.102    0.008   12.293    0.000    0.102    0.444
##    .EP              0.225    0.016   14.456    0.000    0.225    0.447
##    .SP              0.249    0.015   16.300    0.000    0.249    0.531
##    .HP              2.138    0.106   20.073    0.000    2.138    0.878
##    .DP              0.127    0.009   14.467    0.000    0.127    0.447
##
## R-Square:
##              Estimate
##    O1              0.683
##    O2              0.639
##    O3              0.806
##    A1              0.472
##    A2              0.999
##    A3              0.556
##    EP              0.553
##    SP              0.469
##    HP              0.122
##    DP              0.553

```

```

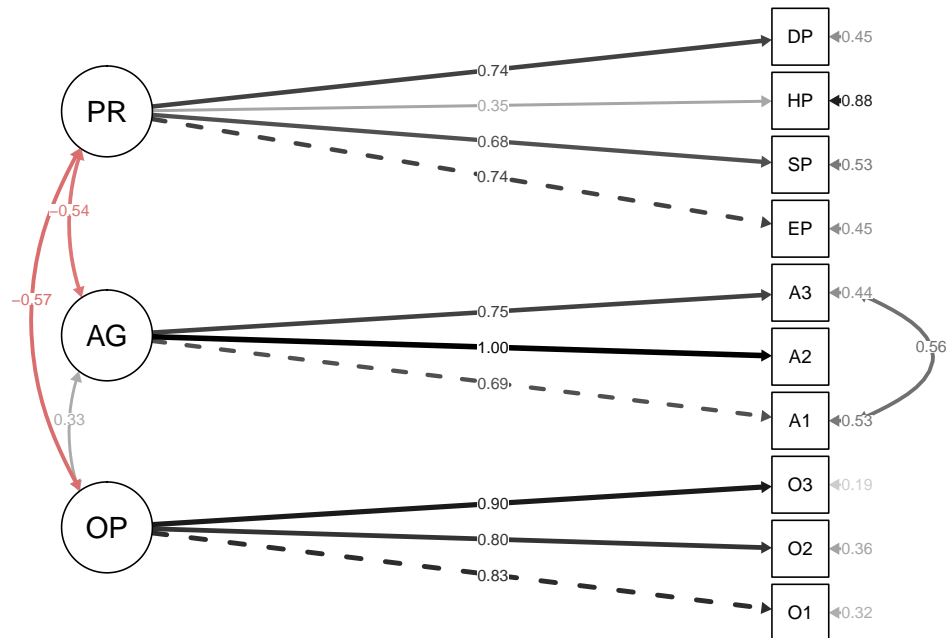
# Visualize the path model
semPaths(model3b.fit,
  rotation = 2,
  layout = "tree2",

```

```

what = "std",
posCol = "black",
edge.width = 0.5,
style = "Lisrel",
fade = T,
edge.label.position = 0.55)

```



```

## Model comparison: Model 3 vs. refined Model 3
anova(model3.fit, model3b.fit)

```

```

## Chi-Squared Difference Test
##
##           Df   AIC   BIC  Chisq Chisq diff Df diff Pr(>Chisq)
## model3b.fit 31 11325 11439 118.26
## model3.fit  32 11392 11501 186.62      68.364      1 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```


Model 4: Structural equation model

```
# Step 1: Model specification
model4 <- '
    # Measurement models
    OP =~ O1 + O2 + O3
    AG =~ A1 + A2 + A3
    PR =~ EP + SP + HP + DP

    # Residual covariance
    A1 ~~ A3

    # Structural model
    PR ~ b1*OP + b2*AG
'

# Step 2: Model estimation
model4.fit <- sem(model4,
    data = Bergh,
    meanstructure = FALSE,
    estimator = "ML")

# Step 3: Evaluate the model
# Summary
summary(model4.fit, rsquare = TRUE, fit.measures = TRUE, standardized = TRUE)

## lavaan 0.6-5 ended normally after 55 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of free parameters      24
##
##      Number of observations          861
##
## Model Test User Model:
##
##      Test statistic                  118.256
##      Degrees of freedom              31
##      P-value (Chi-square)            0.000
##
## Model Test Baseline Model:
##
##      Test statistic                  4270.205
##      Degrees of freedom              45
##      P-value                        0.000
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)      0.979
##      Tucker-Lewis Index (TLI)        0.970
##
## Loglikelihood and Information Criteria:
```

```

##
##   Loglikelihood user model (H0)                -5638.625
##   Loglikelihood unrestricted model (H1)         -5579.497
##
##   Akaike (AIC)                                11325.249
##   Bayesian (BIC)                              11439.444
##   Sample-size adjusted Bayesian (BIC)          11363.226
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                                         0.057
##   90 Percent confidence interval - lower        0.046
##   90 Percent confidence interval - upper        0.068
##   P-value RMSEA <= 0.05                        0.131
##
## Standardized Root Mean Square Residual:
##
##   SRMR                                         0.043
##
## Parameter Estimates:
##
##   Information                                Expected
##   Information saturated (h1) model            Structured
##   Standard errors                            Standard
##
## Latent Variables:
##
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##   OP =~
##     O1              1.000
##     O2              0.934    0.036   26.188    0.000    0.374    0.799
##     O3              1.149    0.040   28.921    0.000    0.460    0.898
##   AG =~
##     A1              1.000
##     A2              1.361    0.086   15.756    0.000    0.471    0.999
##     A3              1.036    0.033   31.662    0.000    0.358    0.746
##   PR =~
##     EP              1.000
##     SP              0.887    0.051   17.460    0.000    0.469    0.685
##     HP              1.031    0.112    9.177    0.000    0.545    0.349
##     DP              0.750    0.040   18.535    0.000    0.397    0.744
##
## Regressions:
##
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##   PR ~
##     OP      (b1)   -0.587    0.053  -11.106    0.000   -0.444   -0.444
##     AG      (b2)   -0.595    0.058  -10.172    0.000   -0.390   -0.390
##
## Covariances:
##
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##   .A1 ~~
##     .A3              0.066    0.008    8.266    0.000    0.066    0.560
##   OP ~~
##     AG              0.046    0.006    7.516    0.000    0.330    0.330
##

```

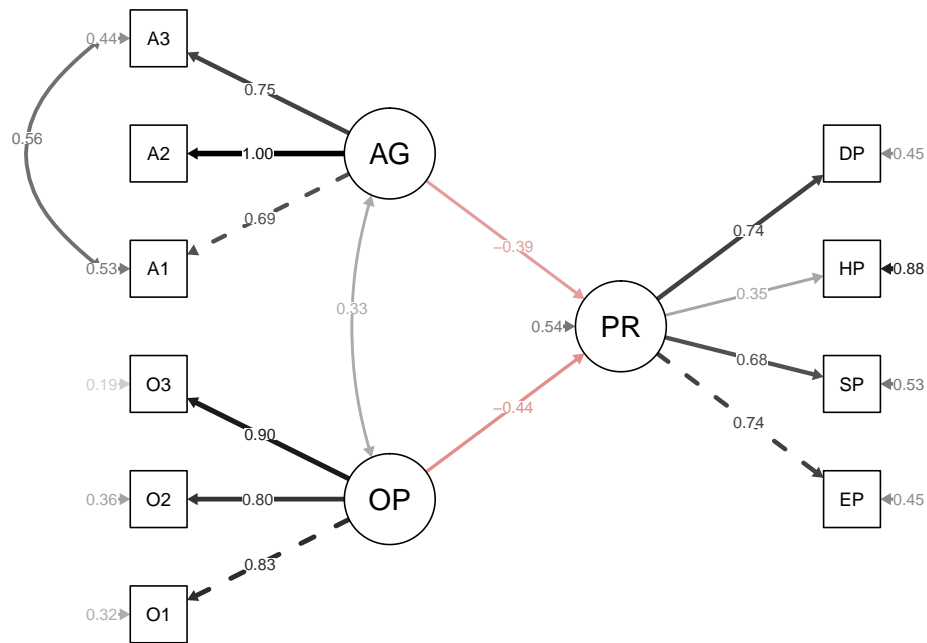
```
## Variances:
##
```

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
## .O1	0.074	0.005	14.570	0.000	0.074	0.317
## .O2	0.079	0.005	15.846	0.000	0.079	0.361
## .O3	0.051	0.005	9.643	0.000	0.051	0.194
## .A1	0.134	0.009	14.890	0.000	0.134	0.528
## .A2	0.000	0.012	0.027	0.979	0.000	0.001
## .A3	0.102	0.008	12.293	0.000	0.102	0.444
## .EP	0.225	0.016	14.456	0.000	0.225	0.447
## .SP	0.249	0.015	16.300	0.000	0.249	0.531
## .HP	2.138	0.106	20.073	0.000	2.138	0.878
## .DP	0.127	0.009	14.467	0.000	0.127	0.447
## OP	0.160	0.011	14.158	0.000	1.000	1.000
## AG	0.120	0.012	9.879	0.000	1.000	1.000
## .PR	0.150	0.015	9.937	0.000	0.536	0.536

```
##
## R-Square:
##
```

	Estimate
## O1	0.683
## O2	0.639
## O3	0.806
## A1	0.472
## A2	0.999
## A3	0.556
## EP	0.553
## SP	0.469
## HP	0.122
## DP	0.553
## PR	0.464

```
# Visualize the path model
semPaths(model4.fit,
  rotation = 2,
  layout = "tree2",
  what = "std",
  posCol = "black",
  edge.width = 0.5,
  style = "Lisrel",
  fade = T,
  edge.label.position = 0.55)
```



```
## Hypothesis testing
lavTestWald(model4.fit, constraints = "b1 == b2")
```

```
## $stat
## [1] 0.009016331
##
## $df
## [1] 1
##
## $p.value
## [1] 0.9243511
##
## $se
## [1] "standard"
```

Model 5: Multi-group SEM (Gender differences in the structural parameters)

```
# Step 1: Model specification
model5 <- '
    # Measurement models
    OP =~ O1 + O2 + O3
    AG =~ A1 + A2 + A3
    PR =~ EP + SP + HP + DP

    # Covariance structure
    OP ~~ OP + AG
    AG ~~ AG

    # Residual covariance
    A1 ~~ A3

    # Structural model
    PR ~ c(a1,b1)*OP + c(a2,b2)*AG
'

# Step 2: Model estimation
# Only allow for differences in the structural parameters
# Keep all other parameters equal (measurement invariance)
model5.fit <- sem(model5,
    data = Bergh,
    meanstructure = FALSE,
    estimator = "ML",
    group = "gender",
    group.equal = c("loadings", "residuals"))

# Step 3: Evaluate the model
# Summary
summary(model5.fit, rsquare = TRUE, fit.measures = TRUE, standardized = TRUE)

## lavaan 0.6-5 ended normally after 60 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of free parameters      48
##      Number of equality constraints   17
##      Row rank of the constraints matrix 17
##
##      Number of observations per group:
##      male                          249
##      female                        612
##
## Model Test User Model:
##
##      Test statistic          208.998
##      Degrees of freedom       79
##      P-value (Chi-square)     0.000
```

```

## Test statistic for each group:
##   male      83.323
##   female    125.675
##
## Model Test Baseline Model:
##
## Test statistic      4207.254
## Degrees of freedom      90
## P-value      0.000
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI)      0.968
## Tucker-Lewis Index (TLI)      0.964
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0)      -5575.804
## Loglikelihood unrestricted model (H1)      -5471.305
##
## Akaike (AIC)      11213.608
## Bayesian (BIC)      11361.109
## Sample-size adjusted Bayesian (BIC)      11262.661
##
## Root Mean Square Error of Approximation:
##
## RMSEA      0.062
## 90 Percent confidence interval - lower      0.052
## 90 Percent confidence interval - upper      0.072
## P-value RMSEA <= 0.05      0.028
##
## Standardized Root Mean Square Residual:
##
## SRMR      0.067
##
## Parameter Estimates:
##
## Information      Expected
## Information saturated (h1) model      Structured
## Standard errors      Standard
##
##
## Group 1 [male]:
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## OP =~
## O1      1.000
## O2      0.932 0.036 26.006 0.000 0.382 0.806
## O3      1.148 0.040 28.727 0.000 0.471 0.902
## AG =~
## A1      1.000
## A2      1.327 0.089 14.919 0.000 0.459 0.993
## A3      1.021 0.033 31.242 0.000 0.353 0.744

```

```

## PR =~
## EP 1.000 0.552 0.761
## SP (.p8.) 0.822 0.049 16.706 0.000 0.454 0.685
## HP (.p9.) 1.029 0.114 9.020 0.000 0.568 0.363
## DP (.10.) 0.733 0.041 17.857 0.000 0.405 0.748
##
## Regressions:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## PR ~
## OP (a1) -0.620 0.097 -6.403 0.000 -0.461 -0.461
## AG (a2) -0.510 0.108 -4.719 0.000 -0.320 -0.320
##
## Covariances:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## OP ~~
## AG 0.059 0.011 5.315 0.000 0.414 0.414
## .A1 ~~
## .A3 0.058 0.009 6.338 0.000 0.058 0.502
##
## Variances:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## OP 0.168 0.019 9.083 0.000 1.000 1.000
## AG 0.120 0.015 7.753 0.000 1.000 1.000
## .01 (.17.) 0.074 0.005 14.561 0.000 0.074 0.306
## .02 (.18.) 0.079 0.005 15.870 0.000 0.079 0.351
## .03 (.19.) 0.051 0.005 9.612 0.000 0.051 0.186
## .A1 (.20.) 0.132 0.009 14.304 0.000 0.132 0.525
## .A2 (.21.) 0.003 0.012 0.244 0.807 0.003 0.013
## .A3 (.22.) 0.101 0.008 11.859 0.000 0.101 0.447
## .EP (.23.) 0.221 0.016 14.053 0.000 0.221 0.421
## .SP (.24.) 0.233 0.014 16.504 0.000 0.233 0.531
## .HP (.25.) 2.133 0.106 20.044 0.000 2.133 0.869
## .DP (.26.) 0.129 0.009 14.552 0.000 0.129 0.440
## .PR 0.172 0.026 6.590 0.000 0.564 0.564
##
## R-Square:
## Estimate
## O1 0.694
## O2 0.649
## O3 0.814
## A1 0.475
## A2 0.987
## A3 0.553
## EP 0.579
## SP 0.469
## HP 0.131
## DP 0.560
## PR 0.436
##
##
## Group 2 [female]:
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all

```

```

## OP =~
## O1      1.000      0.394  0.822
## O2      (.p2.) 0.932  0.036 26.006  0.000 0.367  0.794
## O3      (.p3.) 1.148  0.040 28.727  0.000 0.452  0.895
## AG =~
## A1      1.000      0.343  0.686
## A2      (.p5.) 1.327  0.089 14.919  0.000 0.455  0.993
## A3      (.p6.) 1.021  0.033 31.242  0.000 0.351  0.741
## PR =~
## EP      1.000      0.511  0.736
## SP      (.p8.) 0.822  0.049 16.706  0.000 0.420  0.656
## HP      (.p9.) 1.029  0.114  9.020  0.000 0.526  0.339
## DP      (.10.) 0.733  0.041 17.857  0.000 0.375  0.722
##
## Regressions:
##           Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## PR ~
## OP      (b1) -0.586  0.060 -9.722  0.000 -0.451 -0.451
## AG      (b2) -0.568  0.066 -8.614  0.000 -0.381 -0.381
##
## Covariances:
##           Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## OP ~~
## AG      0.037  0.007  5.683  0.000  0.276  0.276
## .A1 ~~
## .A3      0.067  0.008  8.011  0.000  0.067  0.576
##
## Variances:
##           Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## OP      0.155  0.012 12.660  0.000  1.000  1.000
## AG      0.118  0.013  9.258  0.000  1.000  1.000
## .O1      (.17.) 0.074  0.005 14.561  0.000  0.074  0.324
## .O2      (.18.) 0.079  0.005 15.870  0.000  0.079  0.370
## .O3      (.19.) 0.051  0.005  9.612  0.000  0.051  0.199
## .A1      (.20.) 0.132  0.009 14.304  0.000  0.132  0.529
## .A2      (.21.) 0.003  0.012  0.244  0.807  0.003  0.014
## .A3      (.22.) 0.101  0.008 11.859  0.000  0.101  0.450
## .EP      (.23.) 0.221  0.016 14.053  0.000  0.221  0.459
## .SP      (.24.) 0.233  0.014 16.504  0.000  0.233  0.569
## .HP      (.25.) 2.133  0.106 20.044  0.000  2.133  0.885
## .DP      (.26.) 0.129  0.009 14.552  0.000  0.129  0.478
## .PR      0.145  0.017  8.743  0.000  0.556  0.556
##
## R-Square:
##           Estimate
## O1      0.676
## O2      0.630
## O3      0.801
## A1      0.471
## A2      0.986
## A3      0.550
## EP      0.541
## SP      0.431
## HP      0.115

```



```
##      DP      0.522
##      PR      0.444

# Hypothesis testing
lavTestWald(model5.fit, constraints = "a1==b1")

## $stat
## [1] 0.0955577
##
## $df
## [1] 1
##
## $p.value
## [1] 0.7572271
##
## $se
## [1] "standard"

lavTestWald(model5.fit, constraints = "a2==b2")

## $stat
## [1] 0.2275054
##
## $df
## [1] 1
##
## $p.value
## [1] 0.6333798
##
## $se
## [1] "standard"
```

Model 6: Multi-group SEM with equal structural parameters

```
# Step 1: Model specification
model6 <- '
    # Measurement models
    OP =~ O1 + O2 + O3
    AG =~ A1 + A2 + A3
    PR =~ EP + SP + HP + DP

    # Covariance structure
    OP ~~ OP + AG
    AG ~~ AG

    # Residual covariance
    A1 ~~ A3

    # Structural model
    PR ~ OP + AG
'

# Step 2: Model estimation
```

```

model6.fit <- sem(model6,
  data = Bergh,
  meanstructure = FALSE,
  estimator = "ML",
  group = "gender",
  group.equal = c("loadings",
                  "residuals",
                  "regressions"))

# Summary
summary(model6.fit, rsquare = TRUE, fit.measures = TRUE, standardized = TRUE)

```

```

## lavaan 0.6-5 ended normally after 59 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of free parameters      48
##      Number of equality constraints   19
##      Row rank of the constraints matrix 19
##
##      Number of observations per group:
##      male                          249
##      female                        612
##
## Model Test User Model:
##
##      Test statistic                209.237
##      Degrees of freedom              81
##      P-value (Chi-square)           0.000
##      Test statistic for each group:
##      male                          83.388
##      female                        125.849
##
## Model Test Baseline Model:
##
##      Test statistic                4207.254
##      Degrees of freedom              90
##      P-value                        0.000
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)    0.969
##      Tucker-Lewis Index (TLI)      0.965
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)   -5575.924
##      Loglikelihood unrestricted model (H1) -5471.305
##
##      Akaike (AIC)                  11209.847
##      Bayesian (BIC)                 11347.832
##      Sample-size adjusted Bayesian (BIC) 11255.736
##
## Root Mean Square Error of Approximation:

```

```

##
## RMSEA 0.061
## 90 Percent confidence interval - lower 0.051
## 90 Percent confidence interval - upper 0.071
## P-value RMSEA <= 0.05 0.041
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.067
##
## Parameter Estimates:
##
## Information Expected
## Information saturated (h1) model Structured
## Standard errors Standard
##
##
## Group 1 [male]:
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## OP =~
## O1 1.000 0.411 0.833
## O2 (.p2.) 0.932 0.036 26.007 0.000 0.383 0.806
## O3 (.p3.) 1.148 0.040 28.723 0.000 0.471 0.902
## AG =~
## A1 1.000 0.346 0.690
## A2 (.p5.) 1.322 0.088 14.987 0.000 0.458 0.991
## A3 (.p6.) 1.022 0.033 31.242 0.000 0.354 0.745
## PR =~
## EP 1.000 0.554 0.762
## SP (.p8.) 0.822 0.049 16.716 0.000 0.456 0.686
## HP (.p9.) 1.029 0.114 9.019 0.000 0.570 0.364
## DP (.10.) 0.733 0.041 17.855 0.000 0.406 0.749
##
## Regressions:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## PR ~
## OP (.15.) -0.593 0.053 -11.116 0.000 -0.439 -0.439
## AG (.16.) -0.554 0.058 -9.488 0.000 -0.346 -0.346
##
## Covariances:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## OP ~~
## AG 0.059 0.011 5.315 0.000 0.414 0.414
## .A1 ~~
## .A3 0.058 0.009 6.303 0.000 0.058 0.500
##
## Variances:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## OP 0.169 0.019 9.099 0.000 1.000 1.000
## AG 0.120 0.015 7.760 0.000 1.000 1.000
## .O1 (.17.) 0.074 0.005 14.560 0.000 0.074 0.306
## .O2 (.18.) 0.079 0.005 15.867 0.000 0.079 0.351

```

```

##      .O3      (.19.)    0.051    0.005    9.610    0.000    0.051    0.186
##      .A1      (.20.)    0.132    0.009   14.287    0.000    0.132    0.524
##      .A2      (.21.)    0.004    0.012    0.312    0.755    0.004    0.017
##      .A3      (.22.)    0.100    0.008   11.834    0.000    0.100    0.445
##      .EP      (.23.)    0.221    0.016   14.050    0.000    0.221    0.419
##      .SP      (.24.)    0.233    0.014   16.496    0.000    0.233    0.529
##      .HP      (.25.)    2.133    0.106   20.044    0.000    2.133    0.868
##      .DP      (.26.)    0.129    0.009   14.562    0.000    0.129    0.439
##      .PR              0.172    0.026    6.611    0.000    0.561    0.561
##
## R-Square:
##              Estimate
##      O1              0.694
##      O2              0.649
##      O3              0.814
##      A1              0.476
##      A2              0.983
##      A3              0.555
##      EP              0.581
##      SP              0.471
##      HP              0.132
##      DP              0.561
##      PR              0.439
##
##
## Group 2 [female]:
##
## Latent Variables:
##              Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      OP =~
##      O1              1.000
##      O2      (.p2.)    0.932    0.036   26.007    0.000    0.367    0.794
##      O3      (.p3.)    1.148    0.040   28.723    0.000    0.452    0.895
##      AG =~
##      A1              1.000
##      A2      (.p5.)    1.322    0.088   14.987    0.000    0.455    0.991
##      A3      (.p6.)    1.022    0.033   31.242    0.000    0.351    0.743
##      PR =~
##      EP              1.000
##      SP      (.p8.)    0.822    0.049   16.716    0.000    0.420    0.656
##      HP      (.p9.)    1.029    0.114    9.019    0.000    0.525    0.338
##      DP      (.10.)    0.733    0.041   17.855    0.000    0.374    0.722
##
## Regressions:
##              Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      PR ~
##      OP      (.15.)   -0.593    0.053  -11.116    0.000   -0.457   -0.457
##      AG      (.16.)   -0.554    0.058   -9.488    0.000   -0.374   -0.374
##
## Covariances:
##              Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      OP ~~
##      AG              0.038    0.007    5.698    0.000    0.277    0.277
##      .A1 ~~

```

```
##      .A3              0.066    0.008    7.978    0.000    0.066    0.575
##
## Variances:
##              Estimate Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      OP              0.155    0.012   12.667    0.000    1.000    1.000
##      AG              0.118    0.013    9.279    0.000    1.000    1.000
##      .O1      (.17.)    0.074    0.005   14.560    0.000    0.074    0.324
##      .O2      (.18.)    0.079    0.005   15.867    0.000    0.079    0.370
##      .O3      (.19.)    0.051    0.005    9.610    0.000    0.051    0.199
##      .A1      (.20.)    0.132    0.009   14.287    0.000    0.132    0.527
##      .A2      (.21.)    0.004    0.012    0.312    0.755    0.004    0.017
##      .A3      (.22.)    0.100    0.008   11.834    0.000    0.100    0.448
##      .EP      (.23.)    0.221    0.016   14.050    0.000    0.221    0.459
##      .SP      (.24.)    0.233    0.014   16.496    0.000    0.233    0.569
##      .HP      (.25.)    2.133    0.106   20.044    0.000    2.133    0.885
##      .DP      (.26.)    0.129    0.009   14.562    0.000    0.129    0.479
##      .PR              0.145    0.017    8.747    0.000    0.557    0.557
##
## R-Square:
##              Estimate
##      O1              0.676
##      O2              0.630
##      O3              0.801
##      A1              0.473
##      A2              0.983
##      A3              0.552
##      EP              0.541
##      SP              0.431
##      HP              0.115
##      DP              0.521
##      PR              0.443
```

```
# Model comparison
```

```
anova(model5.fit, model6.fit)
```

```
## Chi-Squared Difference Test
```

```
##
##              Df    AIC    BIC  Chisq Chisq diff Df diff Pr(>Chisq)
## model5.fit  79 11214 11361 209.00
## model6.fit  81 11210 11348 209.24      0.2392      2      0.8873
```

R session info

```
sessionInfo()
```

```
## R version 3.6.3 (2020-02-29)
## Platform: x86_64-apple-darwin15.6.0 (64-bit)
## Running under: macOS Sierra 10.12.6
##
## Matrix products: default
## BLAS: /Library/Frameworks/R.framework/Versions/3.6/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/3.6/Resources/lib/libRlapack.dylib
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
## attached base packages:
## [1] stats      graphics  grDevices  utils      datasets  methods   base
##
## other attached packages:
## [1] corrplot_0.84    MPsychR_0.10-7  semPlot_1.1.2   lavaan_0.6-5
##
## loaded via a namespace (and not attached):
## [1] nlme_3.1-144      RColorBrewer_1.1-2  mi_1.0
## [4] tools_3.6.3       backports_1.1.6     R6_2.4.1
## [7] d3Network_0.5.2.1 rpart_4.1-15        Hmisc_4.3-1
## [10] colorspace_1.4-1  nnet_7.3-12         tidyselect_1.0.0
## [13] gridExtra_2.3     mnormt_1.5-7        compiler_3.6.3
## [16] qgraph_1.6.5      fdrtool_1.2.15      htmlTable_1.13.3
## [19] regsem_1.5.2      scales_1.1.0        checkmate_2.0.0
## [22] psych_1.9.12.31   pbapply_1.4-2       sem_3.1-9
## [25] stringr_1.4.0     digest_0.6.25       pbivnorm_0.6.0
## [28] foreign_0.8-75    minqa_1.2.4         rmarkdown_2.1
## [31] base64enc_0.1-3   jpeg_0.1-8.1        pkgconfig_2.0.3
## [34] htmltools_0.4.0   lme4_1.1-23         lisrelToR_0.1.4
## [37] htmlwidgets_1.5.1 rlang_0.4.6         huge_1.3.4
## [40] rstudioapi_0.11   gtools_3.8.1        acepack_1.4.1
## [43] dplyr_0.8.5       zip_2.0.4           magrittr_1.5
## [46] OpenMx_2.17.3     Formula_1.2-3       Matrix_1.2-18
## [49] Rcpp_1.0.4.6      munsell_0.5.0       abind_1.4-5
## [52] rockchalk_1.8.144 lifecycle_0.2.0     whisker_0.4
## [55] stringi_1.4.6     yaml_2.2.1          carData_3.0-3
## [58] MASS_7.3-51.5     plyr_1.8.6          matrixcalc_1.0-3
## [61] grid_3.6.3        parallel_3.6.3      crayon_1.3.4
## [64] lattice_0.20-40   kutils_1.69         splines_3.6.3
## [67] knitr_1.28        pillar_1.4.3        igraph_1.2.5
## [70] rjson_0.2.20      boot_1.3-24         corpcor_1.6.9
## [73] BDgraph_2.62      codetools_0.2-16    reshape2_1.4.4
## [76] stats4_3.6.3      XML_3.99-0.3        glue_1.4.0
## [79] evaluate_0.14     latticeExtra_0.6-29 data.table_1.12.8
## [82] png_0.1-7         vctr_0.2.4          nloptr_1.2.2.1
## [85] gtable_0.3.0      purrr_0.3.4         assertthat_0.2.1
## [88] ggplot2_3.3.0     xfun_0.13           openxlsx_4.1.4
## [91] xtable_1.8-4      coda_0.19-3         Rsolnp_1.16
```

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## [94] survival_3.1-8      glasso_1.11          truncnorm_1.0-8
## [97] tibble_3.0.1         arm_1.10-1          cluster_2.1.0
## [100] statmod_1.4.34       ellipsis_0.3.0
```