

EFA in lavaan

lavaan (LAatent VARIable ANalysis)

Full documentation available at (<https://cran.r-project.org/web/packages/lavaan/lavaan.pdf>)

Some Useful lavaan Notations

The “ $=\sim$ ” operator can be used to define (continuous) latent variables. This is to define a reflexive factor.

The “ $\sim\sim$ ” (‘double tilde’) operator specifies (residual) variances of an observed or latent variable, or a set of covariances.

The “ $<\sim$ ” operator can be used to define a formative factor.

The “ $|$ ” operator can be used to define the thresholds of categorical endogenous variables.

As in Mplus, you can have multiple variables on the left side of the operator. If you do, you will need to add “ $+$ ” in between them (if you list them without the “ $+$ ” (as you would in Mplus), you will get an error message).

lavaan Defaults to Know About

As in Mplus, this varies by the model being estimated, we’ll go over these as needed but all is noted in the lavaan official documentation. Many are identical to Mplus’ defaults, but it’s worth checking them first.

- Default estimator is maximum likelihood;
- The factor loading of the first indicator of a latent variable is fixed to 1;
- Residual variances are freely estimated;
- All exogenous variables are allowed to covary.

Fixing covariances in lavaan

Building on the above, you can specify an orthogonal (zero) covariance between two latent or observed variables: $f1 \sim\sim 0*f2$

If you have categorical indicators

Muthen & Muthen recommend weighted least squares (WLS) when you have many factors and not so many factor indicators. They recommend maximum likelihood (ML, MLR) when you have few factors and many factor indicators. Both MLR and WLS can deal with categorical and continuous outcomes.

Mplus resources for EFA

Not just code, the stats too

<http://www.statmodel.com/discussion/messages/8/8.html>

Load packages

```
library(lavaan)  # for the loadings

## This is lavaan 0.6-9
## lavaan is FREE software! Please report any bugs.
```

Load file

```
efa_data <- read.table("IH validation 1-21.dat")
```

Specify the model

```
# these are exploratory blocks (you can give them any
# name), you'll need to specify them
efa_model <- "
  efa(\"block1\")*F1 =~ V1 + V2 + V3 + V4 + V5 + V6
  efa(\"block1\")*F2 =~ V1 + V2 + V3 + V4 + V5 + V6
"
```

Estimate the Model

```
efa_f2 <- sem(model = efa_model, data = efa_data, rotation = "oblimin",
  estimator = "MLR")
```

```
## Warning in lav_object_post_check(object): lavaan WARNING: some estimated ov
## variances are negative
```

Request the Output

```
summary(efa_f2, fit.measures = TRUE, standardized = TRUE)
```

```

## lavaan 0.6-9 ended normally after 27 iterations
##
## Estimator ML
## Optimization method NLMINB
## Number of model parameters 17
##
## Rotation method OBLIMIN OBLIQUE
## Oblimin gamma 0
## Rotation algorithm (rstarts) GPA (100)
## Standardized metric TRUE
## Row weights None
##
## Number of observations 200
##
## Model Test User Model:
## Standard Robust
## Test Statistic 13.105 14.567
## Degrees of freedom 4 4
## P-value (Chi-square) 0.011 0.006
## Scaling correction factor 0.900
## Yuan-Bentler correction (Mplus variant)
##
## Model Test Baseline Model:
##
## Test statistic 260.859 203.404
## Degrees of freedom 15 15
## P-value 0.000 0.000
## Scaling correction factor 1.282
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI) 0.963 0.944
## Tucker-Lewis Index (TLI) 0.861 0.790
##
## Robust Comparative Fit Index (CFI) 0.961
## Robust Tucker-Lewis Index (TLI) 0.852
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -1437.700 -1437.700
## Scaling correction factor 1.473
## for the MLR correction
## Loglikelihood unrestricted model (H1) -1431.147 -1431.147
## Scaling correction factor 1.364
## for the MLR correction
##
## Akaike (AIC) 2909.400 2909.400
## Bayesian (BIC) 2965.471 2965.471
## Sample-size adjusted Bayesian (BIC) 2911.613 2911.613
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.107 0.115
## 90 Percent confidence interval - lower 0.046 0.053

```

```

## 90 Percent confidence interval - upper      0.173      0.185
## P-value RMSEA <= 0.05                     0.060      0.045
##
## Robust RMSEA                                0.109
## 90 Percent confidence interval - lower      0.053
## 90 Percent confidence interval - upper      0.172
##
## Standardized Root Mean Square Residual:
##
## SRMR                                0.038      0.038
##
## Parameter Estimates:
##
## Standard errors                        Sandwich
## Information bread                      Observed
## Observed information based on          Hessian
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## F1 =~ block1
## V1          1.402   1.997   0.702   0.483   1.402   1.356
## V2          0.065   0.139   0.467   0.641   0.065   0.084
## V3         -0.080   0.118  -0.677   0.498  -0.080  -0.099
## V4          0.168   0.326   0.514   0.607   0.168   0.181
## V5         -0.063   0.097  -0.648   0.517  -0.063  -0.073
## V6          0.033   0.147   0.222   0.825   0.033   0.034
## F2 =~ block1
## V1          0.001   0.006   0.146   0.884   0.001   0.001
## V2          0.535   0.101   5.302   0.000   0.535   0.693
## V3          0.394   0.111   3.551   0.000   0.394   0.491
## V4          0.434   0.175   2.484   0.013   0.434   0.470
## V5          0.675   0.112   6.022   0.000   0.675   0.781
## V6          0.486   0.113   4.311   0.000   0.486   0.503
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## F1 ~~
## F2          0.273   0.337   0.809   0.418   0.273   0.273
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .V1        -0.897   5.601  -0.160   0.873  -0.897  -0.839
## .V2         0.286   0.047   6.053   0.000   0.286   0.481
## .V3         0.500   0.086   5.810   0.000   0.500   0.776
## .V4         0.598   0.090   6.627   0.000   0.598   0.700
## .V5         0.309   0.101   3.075   0.002   0.309   0.415
## .V6         0.687   0.093   7.370   0.000   0.687   0.737
## F1          1.000
## F2          1.000

```