

ASCI and Structural Equation Model

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1. Load libraries and the survey dataset.

- <http://www2.gsu.edu/~mkteer/sem2.html#loading>
- <http://lavaan.ugent.be/index.html>

```
if (!require("lavaan", quietly = TRUE)){install.packages("lavaan")}

## This is lavaan 0.5-23.1097
## lavaan is BETA software! Please report any bugs.

library(lavaan);
library(readr);
library(psych);
data_directory <- "~/Desktop/r_workspace/data-test5.csv";
data <- suppressMessages( read_csv(data_directory, progress = FALSE)[1:198,] );
data$a10 <- 10 - data$a10;    ## conformation = 10 - disconformation
```

2. Define a structural equation model and conduct fitting.

- Some of resulting standard errors are negative (Heywood cases).
- In our model, we should check the variance estimate of eta-1 (-0.12).
- <http://zencaroline.blogspot.kr/2007/05/heywood-cases-negative-error-variances.html>

```
model <- '
# measurement model
ksi =~ a1 + a2 + a3
eta1 =~ a4 + a5 + a6
eta2 =~ a7 + a8
eta3 =~ a9 + a10 + a11
eta4 =~ a15
eta5 =~ a12 + a13 + a14
# regressions
eta1 ~ ksi
eta2 ~ ksi + eta1
eta3 ~ ksi + eta1 + eta2
eta4 ~ eta3
eta5 ~ eta3 + eta4
';
fit <- sem(model, data = data);
```

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

3. Print output.

```
summary(fit, standardized=TRUE);
```

```
## lavaan (0.5-23.1097) converged normally after 105 iterations
##
##   Number of observations              198
##
##   Estimator                          ML
##   Minimum Function Test Statistic    418.023
##   Degrees of freedom                 82
##   P-value (Chi-square)               0.000
##
## Parameter Estimates:
##
##   Information                        Expected
##   Standard Errors                   Standard
##
## Latent Variables:
##
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##   ksi =~
##     a1              1.000
##     a2              3.115    1.096    2.842    0.004    0.867    0.766
##     a3              3.007    1.054    2.854    0.004    0.837    0.818
##   eta1 =~
##     a4              1.000
##     a5              0.469    0.075    6.283    0.000    0.447    0.428
##     a6              0.948    0.064   14.903    0.000    0.904    0.814
##   eta2 =~
##     a7              1.000
##     a8              1.236    0.108   11.492    0.000    0.942    0.772
##   eta3 =~
##     a9              1.000
##     a10             0.279    0.076    3.668    0.000    0.262    0.262
##     a11             1.071    0.063   17.019    0.000    1.007    0.852
##   eta4 =~
##     a15             1.000
##   eta5 =~
##     a12             1.000
##     a13             0.701    0.093    7.541    0.000    0.691    0.507
##     a14             1.033    0.097   10.666    0.000    1.018    0.663
##
## Regressions:
##
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##   eta1 ~
##     ksi              3.644    1.279    2.849    0.004    1.064    1.064
##   eta2 ~
##     ksi             -0.217    0.919   -0.236    0.813   -0.079   -0.079
##     eta1              0.848    0.268    3.165    0.002    1.060    1.060
##   eta3 ~
##     ksi              1.733    1.031    1.681    0.093    0.513    0.513
##     eta1             -0.286    1.021   -0.280    0.779   -0.290   -0.290
##     eta2              0.878    1.330    0.660    0.509    0.712    0.712
##   eta4 ~
```

```
##      eta3      -0.845    0.080  -10.535    0.000   -0.646   -0.646
##      eta5 ~
##      eta3      0.821    0.068   12.029    0.000    0.783    0.783
##      eta4     -0.204    0.045   -4.516    0.000   -0.254   -0.254
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .a1      1.869   0.188   9.953   0.000   1.869   0.960
##      .a2      0.529   0.061   8.661   0.000   0.529   0.413
##      .a3      0.347   0.045   7.754   0.000   0.347   0.331
##      .a4      0.306   0.042   7.327   0.000   0.306   0.252
##      .a5      0.895   0.091   9.848   0.000   0.895   0.817
##      .a6      0.417   0.049   8.478   0.000   0.417   0.338
##      .a7      0.435   0.054   8.078   0.000   0.435   0.428
##      .a8      0.601   0.077   7.784   0.000   0.601   0.404
##      .a9      0.202   0.030   6.819   0.000   0.202   0.186
##      .a10     0.928   0.094   9.900   0.000   0.928   0.931
##      .a11     0.384   0.047   8.104   0.000   0.384   0.275
##      .a15     0.000   0.000   0.000   0.000   0.000   0.000
##      .a12     0.281   0.052   5.406   0.000   0.281   0.224
##      .a13     1.378   0.143   9.661   0.000   1.378   0.743
##      .a14     1.320   0.143   9.245   0.000   1.320   0.560
##      ksi      0.077   0.054   1.424   0.154   1.000   1.000
##      .eta1    -0.120   0.050  -2.407   0.016  -0.132  -0.132
##      .eta2     0.028   0.044   0.640   0.522   0.048   0.048
##      .eta3     0.088   0.046   1.896   0.058   0.099   0.099
##      .eta4     0.879   0.094   9.376   0.000   0.582   0.582
##      .eta5     0.064   0.045   1.424   0.154   0.066   0.066
```

4. Detailed outputs.

```
# Unstandardized solution matrix (factor loading, or lambda).
inspect(fit,what="est")$lambda
```

```
##      ksi eta1 eta2 eta3 eta4 eta5
## a1  1.000 0.000 0.000 0.000 0 0.000
## a2  3.115 0.000 0.000 0.000 0 0.000
## a3  3.007 0.000 0.000 0.000 0 0.000
## a4  0.000 1.000 0.000 0.000 0 0.000
## a5  0.000 0.469 0.000 0.000 0 0.000
## a6  0.000 0.948 0.000 0.000 0 0.000
## a7  0.000 0.000 1.000 0.000 0 0.000
## a8  0.000 0.000 1.236 0.000 0 0.000
## a9  0.000 0.000 0.000 1.000 0 0.000
## a10 0.000 0.000 0.000 0.279 0 0.000
## a11 0.000 0.000 0.000 1.071 0 0.000
## a15 0.000 0.000 0.000 0.000 1 0.000
## a12 0.000 0.000 0.000 0.000 0 1.000
## a13 0.000 0.000 0.000 0.000 0 0.701
## a14 0.000 0.000 0.000 0.000 0 1.033
```

```
# Standardized solution matrix (factor loading, or lambda).
inspect(fit,what="std")$lambda
```

```
##      ksi  eta1  eta2  eta3 eta4  eta5
## a1  0.200 0.000 0.000 0.000    0 0.000
## a2  0.766 0.000 0.000 0.000    0 0.000
## a3  0.818 0.000 0.000 0.000    0 0.000
## a4  0.000 0.865 0.000 0.000    0 0.000
## a5  0.000 0.428 0.000 0.000    0 0.000
## a6  0.000 0.814 0.000 0.000    0 0.000
## a7  0.000 0.000 0.756 0.000    0 0.000
## a8  0.000 0.000 0.772 0.000    0 0.000
## a9  0.000 0.000 0.000 0.902    0 0.000
## a10 0.000 0.000 0.000 0.262    0 0.000
## a11 0.000 0.000 0.000 0.852    0 0.000
## a15 0.000 0.000 0.000 0.000    1 0.000
## a12 0.000 0.000 0.000 0.000    0 0.881
## a13 0.000 0.000 0.000 0.000    0 0.507
## a14 0.000 0.000 0.000 0.000    0 0.663
```

5. Goodness of fit of the model.

```
fitmeasures(fit)
```

```
##      npar      fmin      chisq
##      38.000      1.056      418.023
##      df      pvalue baseline.chisq
##      82.000      0.000      2180.067
##      baseline.df baseline.pvalue      cfi
##      105.000      0.000      0.838
##      tli      nnfi      rfi
##      0.793      0.793      0.754
##      nfi      pnfi      ifi
##      0.808      0.631      0.840
##      rni      logl unrestricted.logl
##      0.838      -3768.288      -3559.276
##      aic      bic      ntotal
##      7612.575      7737.529      198.000
##      bic2      rmsea      rmsea.ci.lower
##      7617.145      0.144      0.130
##      rmsea.ci.upper      rmsea.pvalue      rmr
##      0.158      0.000      0.137
##      rmr_nomean      srmr      srmr_bentler
##      0.137      0.102      0.102
##      srmr_bentler_nomean      srmr_bollen      srmr_bollen_nomean
##      0.102      0.102      0.102
##      srmr_mplus      srmr_mplus_nomean      cn_05
##      0.102      0.102      50.326
##      cn_01      gfi      agfi
##      55.326      0.785      0.685
##      pgfi      mfi      ecvi
##      0.536      0.428      2.495
```

6. ASCI (simplified fomula).

```
# Load weighted of eta-3.
numerator <- sum(inspect(fit,what="est")$lambda[9:11,4] * colMeans(data[,11:13])) - sum(inspect(fit,what="est")$lambda[9:11,4]) * 9
denominator <- sum(inspect(fit,what="est")$lambda[9:11,4]) * 9

ASCI <- (numerator / denominator) *100
ASCI

## [1] 62.91129
```

7. Cronbach's alpha.

```
alpha(data[,3:5], check.keys = TRUE);

##
## Reliability analysis
## Call: alpha(x = data[, 3:5], check.keys = TRUE)
##
##   raw_alpha std.alpha G6(smc) average_r S/N   ase mean  sd
##      0.49      0.54    0.54      0.28 1.2 0.066  6.5 0.84
##
##   lower alpha upper      95% confidence boundaries
## 0.37 0.49 0.62
##
## Reliability if an item is dropped:
##   raw_alpha std.alpha G6(smc) average_r S/N alpha se
## a1      0.78      0.78    0.636      0.636 3.50  0.032
## a2      0.21      0.22    0.126      0.126 0.29  0.107
## a3      0.15      0.16    0.085      0.085 0.18  0.118
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean  sd
## a1 198  0.64  0.56  0.14  0.11  6.5 1.4
## a2 198  0.75  0.79  0.71  0.42  6.4 1.1
## a3 198  0.76  0.81  0.74  0.48  6.5 1.0
##
## Non missing response frequency for each item
##      3  4  5  6  7  8  9 10 miss
## a1 0.02 0.07 0.17 0.24 0.24 0.23 0.05 0.00  0
## a2 0.02 0.03 0.18 0.26 0.37 0.14 0.01 0.00  0
## a3 0.01 0.02 0.14 0.30 0.40 0.13 0.01 0.01  0

alpha(data[,6:8], check.keys = TRUE);

##
## Reliability analysis
## Call: alpha(x = data[, 6:8], check.keys = TRUE)
##
##   raw_alpha std.alpha G6(smc) average_r S/N   ase mean  sd
##      0.71      0.71    0.68      0.45 2.5 0.036  6.5 0.87
##
##   lower alpha upper      95% confidence boundaries
```

```

## 0.64 0.71 0.78
##
## Reliability if an item is dropped:
##   raw_alpha std.alpha G6(smc) average_r S/N alpha se
## a4      0.49      0.49   0.32      0.32 0.95   0.073
## a5      0.84      0.84   0.72      0.72 5.18   0.023
## a6      0.47      0.47   0.31      0.31 0.89   0.075
##
## Item statistics
##   n raw.r std.r r.cor r.drop mean sd
## a4 198 0.86 0.85 0.79 0.64 6.3 1.1
## a5 198 0.67 0.68 0.38 0.34 6.8 1.0
## a6 198 0.86 0.86 0.80 0.65 6.3 1.1
##
## Non missing response frequency for each item
##   3 4 5 6 7 8 9 miss
## a4 0.01 0.03 0.22 0.26 0.35 0.13 0.01 0
## a5 0.00 0.02 0.11 0.23 0.41 0.20 0.04 0
## a6 0.01 0.04 0.20 0.29 0.32 0.13 0.01 0

#alpha(data[,9:10], check.keys = TRUE);

alpha(data[,11:13], check.keys = TRUE);

##
## Reliability analysis
## Call: alpha(x = data[, 11:13], check.keys = TRUE)
##
##   raw_alpha std.alpha G6(smc) average_r S/N ase mean sd
##   0.66      0.65      0.67      0.38 1.8 0.042 6.7 0.83
##
## lower alpha upper      95% confidence boundaries
## 0.57 0.66 0.74
##
## Reliability if an item is dropped:
##   raw_alpha std.alpha G6(smc) average_r S/N alpha se
## a9      0.31      0.31   0.19      0.19 0.45   0.097
## a10     0.87      0.87   0.77      0.77 6.79   0.018
## a11     0.30      0.30   0.18      0.18 0.44   0.099
##
## Item statistics
##   n raw.r std.r r.cor r.drop mean sd
## a9 198 0.86 0.85 0.83 0.65 6.8 1.0
## a10 198 0.56 0.59 0.21 0.19 6.9 1.0
## a11 198 0.87 0.85 0.83 0.63 6.4 1.2
##
## Non missing response frequency for each item
##   3 4 5 6 7 8 9 miss
## a9 0.01 0.01 0.11 0.20 0.40 0.26 0.02 0
## a10 0.00 0.01 0.08 0.20 0.43 0.24 0.04 0
## a11 0.01 0.06 0.14 0.24 0.36 0.18 0.01 0

alpha(data[,14:16], check.keys = TRUE);

##
## Reliability analysis

```

```

## Call: alpha(x = data[, 14:16], check.keys = TRUE)
##
##   raw_alpha std.alpha G6(smc) average_r S/N   ase mean  sd
##      0.73      0.74    0.66      0.49 2.9 0.032  5.2 1.1
##
##   lower alpha upper      95% confidence boundaries
## 0.67 0.73 0.8
##
## Reliability if an item is dropped:
##   raw_alpha std.alpha G6(smc) average_r S/N alpha se
## a12      0.62      0.62    0.45      0.45 1.7   0.053
## a13      0.68      0.70    0.54      0.54 2.4   0.042
## a14      0.64      0.65    0.48      0.48 1.9   0.050
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean  sd
## a12 198  0.80  0.83  0.70   0.60  6.4 1.1
## a13 198  0.79  0.79  0.61   0.53  3.5 1.4
## a14 198  0.84  0.82  0.67   0.57  5.8 1.5
##
## Non missing response frequency for each item
##      1    2    3    4    5    6    7    8    9   10 miss
## a12 0.00 0.00 0.01 0.05 0.16 0.29 0.35 0.14 0.02 0.00    0
## a13 0.07 0.16 0.35 0.16 0.20 0.04 0.02 0.00 0.00 0.00    0
## a14 0.01 0.02 0.04 0.12 0.23 0.21 0.26 0.09 0.02 0.01    0

```