

Supplemental Materials #1: Illustrations using a dataset from Nohe et al. (Table A1; 2015)

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Read the data and display the sample characteristics

```
## Required packages
lib2install <- c("lavaan", "semPlot", "metaSEM")

## Install them automatically if they have not been installed in your computer yet.
for (i in lib2install) {
  if (!(i %in% rownames(installed.packages()))) install.packages(i)
}

# Load the libraries for the analysis.
library(lavaan)
library(metaSEM)
library(semPlot)

## Correlation matrices
Nohe15A1$data

## $`Britt & Dawson (2005)`
##      W1   S1   W2   S2
## W1 1.00 0.29 0.58 0.22
## S1 0.29 1.00 0.24 0.57
## W2 0.58 0.24 1.00 0.27
## S2 0.22 0.57 0.27 1.00
##
## $`Demerouti et al. (2004)`
##      W1   S1   W2   S2
## W1 1.00 0.53 0.57 0.41
## S1 0.53 1.00 0.41 0.68
## W2 0.57 0.41 1.00 0.54
## S2 0.41 0.68 0.54 1.00
##
## $`Ford (2010)`
##      W1   S1   W2   S2
## W1 1.00 0.35 0.75 0.32
```

```

## S1 0.35 1.00 0.26 0.74
## W2 0.75 0.26 1.00 0.30
## S2 0.32 0.74 0.30 1.00
##
## $`Hammer et al. (2005), female subsample`
##      W1   S1   W2   S2
## W1 1.00 0.32 0.57 0.22
## S1 0.32 1.00 0.30 0.43
## W2 0.57 0.30 1.00 0.30
## S2 0.22 0.43 0.30 1.00
##
## $`Hammer et al. (2005), male subsample`
##      W1   S1   W2   S2
## W1 1.00 0.19 0.54 0.17
## S1 0.19 1.00 0.21 0.60
## W2 0.54 0.21 1.00 0.30
## S2 0.17 0.60 0.30 1.00
##
## $`Innstrand et al. (2008)`
##      W1   S1   W2   S2
## W1 1.00 0.42 0.63 0.31
## S1 0.42 1.00 0.30 0.62
## W2 0.63 0.30 1.00 0.44
## S2 0.31 0.62 0.44 1.00
##
## $`Jacobshagen et al. (2006)`
##      W1   S1   W2   S2
## W1 1.00 0.46 0.50 0.38
## S1 0.46 1.00 0.29 0.64
## W2 0.50 0.29 1.00 0.44
## S2 0.38 0.64 0.44 1.00
##
## $`Kalin et al. (2008)`
##      W1   S1   W2   S2
## W1 1.00 0.42 0.52 0.28
## S1 0.42 1.00 0.26 0.54
## W2 0.52 0.26 1.00 0.38
## S2 0.28 0.54 0.38 1.00
##
## $`Kelloway et al. (1999)`
##      W1   S1   W2   S2
## W1 1.00 0.55 0.71 0.43
## S1 0.55 1.00 0.48 0.72
## W2 0.71 0.48 1.00 0.46
## S2 0.43 0.72 0.46 1.00
##
## $`Kinnunen et al. (2010), female subsample`
##      W1   S1   W2   S2
## W1 1.00 0.11 0.57 0.18
## S1 0.11 1.00 0.18 0.71
## W2 0.57 0.18 1.00 0.22
## S2 0.18 0.71 0.22 1.00
##
## $`Kinnunen et al. (2010), male subsample`

```

```

##      W1   S1   W2   S2
## W1 1.00 0.13 0.59 0.17
## S1 0.13 1.00 0.17 0.62
## W2 0.59 0.17 1.00 0.23
## S2 0.17 0.62 0.23 1.00
##
## $`Kinnunen et al. (2004), female subsample`
##      W1   S1   W2   S2
## W1 1.00 0.28 0.71 0.31
## S1 0.28 1.00 0.27 0.61
## W2 0.71 0.27 1.00 0.34
## S2 0.31 0.61 0.34 1.00
##
## $`Kinnunen et al. (2004), male subsample`
##      W1   S1   W2   S2
## W1 1.00 0.30 0.63 0.24
## S1 0.30 1.00 0.35 0.65
## W2 0.63 0.35 1.00 0.38
## S2 0.24 0.65 0.38 1.00
##
## $`Leiter & Durup (1996)`
##      W1   S1   W2   S2
## W1 1.00 0.33 0.61 0.29
## S1 0.33 1.00 0.35 0.67
## W2 0.61 0.35 1.00 0.42
## S2 0.29 0.67 0.42 1.00
##
## $`Mauno (2010)`
##      W1   S1   W2   S2
## W1 1.00 0.54 0.66 0.45
## S1 0.54 1.00 0.34 0.56
## W2 0.66 0.34 1.00 0.65
## S2 0.45 0.56 0.65 1.00
##
## $`Meier et al. (2007)`
##      W1   S1   W2   S2
## W1 1.00 0.42 0.57 0.40
## S1 0.42 1.00 0.37 0.64
## W2 0.57 0.37 1.00 0.56
## S2 0.40 0.64 0.56 1.00
##
## $`Meier et al. (2010)`
##      W1   S1   W2   S2
## W1 1.00 0.40 0.65 0.33
## S1 0.40 1.00 0.25 0.60
## W2 0.65 0.25 1.00 0.47
## S2 0.33 0.60 0.47 1.00
##
## $`Meier et al. (2010)`
##      W1   S1   W2   S2
## W1 1.00 0.49 0.57 0.27
## S1 0.49 1.00 0.29 0.56
## W2 0.57 0.29 1.00 0.40
## S2 0.27 0.56 0.40 1.00

```

```

##
## $`Meier et al. (2010)`
##      W1   S1   W2   S2
## W1 1.00 0.41 0.58 0.28
## S1 0.41 1.00 0.23 0.56
## W2 0.58 0.23 1.00 0.40
## S2 0.28 0.56 0.40 1.00
##
## $`Meier et al. (2010)`
##      W1   S1   W2   S2
## W1 1.00 0.42 0.56 0.33
## S1 0.42 1.00 0.28 0.53
## W2 0.56 0.28 1.00 0.51
## S2 0.33 0.53 0.51 1.00
##
## $`Meier et al. (2010)`
##      W1   S1   W2   S2
## W1 1.00 0.41 0.58 0.31
## S1 0.41 1.00 0.32 0.64
## W2 0.58 0.32 1.00 0.48
## S2 0.31 0.64 0.48 1.00
##
## $`Nohe & Sonntag (2010)`
##      W1   S1   W2   S2
## W1 1.00 0.62 0.71 0.54
## S1 0.62 1.00 0.50 0.75
## W2 0.71 0.50 1.00 0.66
## S2 0.54 0.75 0.66 1.00
##
## $`Nohe & Sonntag (2010)`
##      W1   S1   W2   S2
## W1 1.00 0.63 0.66 0.36
## S1 0.63 1.00 0.51 0.46
## W2 0.66 0.51 1.00 0.34
## S2 0.36 0.46 0.34 1.00
##
## $`Nohe & Sonntag (2014)`
##      W1   S1   W2   S2
## W1 1.00 0.68 0.75 0.59
## S1 0.68 1.00 0.60 0.82
## W2 0.75 0.60 1.00 0.69
## S2 0.59 0.82 0.69 1.00
##
## $`O'Driscoll et al. (2004)`
##      W1   S1   W2   S2
## W1 1.00 0.24 0.70 0.15
## S1 0.24 1.00 0.20 0.70
## W2 0.70 0.20 1.00 0.14
## S2 0.15 0.70 0.14 1.00
##
## $`Rantanen et al. (2008)`
##      W1   S1   W2   S2
## W1 1.00 0.14 0.54 0.07
## S1 0.14 1.00 0.24 0.51

```

```

## W2 0.54 0.24 1.00 0.16
## S2 0.07 0.51 0.16 1.00
##
## $`Schaufeli et al. (2009)`
##      W1   S1   W2   S2
## W1 1.00 0.46 0.50 0.41
## S1 0.46 1.00 0.18 0.65
## W2 0.50 0.18 1.00 0.36
## S2 0.41 0.65 0.36 1.00
##
## $`Semmer et al. (2005)`
##      W1   S1   W2   S2
## W1 1.00 0.30 0.23 0.17
## S1 0.30 1.00 0.16 0.51
## W2 0.23 0.16 1.00 0.33
## S2 0.17 0.51 0.33 1.00
##
## $`Steinmetz et al. (2008)`
##      W1   S1   W2   S2
## W1 1.00 0.25 0.82 0.25
## S1 0.25 1.00 0.34 0.62
## W2 0.82 0.34 1.00 0.39
## S2 0.25 0.62 0.39 1.00
##
## $`van der Heijden et al. (2008)`
##      W1   S1   W2   S2
## W1 1.00 0.23 0.48 0.18
## S1 0.23 1.00 0.20 0.59
## W2 0.48 0.20 1.00 0.22
## S2 0.18 0.59 0.22 1.00
##
## $`van Hooff et al. (2005)`
##      W1   S1   W2   S2
## W1 1.00 0.28 0.62 0.22
## S1 0.28 1.00 0.20 0.44
## W2 0.62 0.20 1.00 0.31
## S2 0.22 0.44 0.31 1.00
##
## $`Westman et al. (2008)`
##      W1   S1   W2   S2
## W1 1.00 0.41 0.64 0.32
## S1 0.41 1.00 0.29 0.81
## W2 0.64 0.29 1.00 0.46
## S2 0.32 0.81 0.46 1.00

## Sample sizes
Nohe15A1$n

##      [1] 489 335 328 234 234 2235 76 94 236 239 239 138 160 151
##     [15] 409 78 256 260 600 462 215 1292 470 665 403 153 201 382
##     [29] 130 946 730 66

## Number of studies
pattern.na(Nohe15A1$data, show.na = FALSE)

##      W1 S1 W2 S2

```

```
## W1 32 32 32 32
## S1 32 32 32 32
## W2 32 32 32 32
## S2 32 32 32 32

## Sample sizes for each correlation coefficient
pattern.n(Nohe15A1$data, Nohe15A1$n)

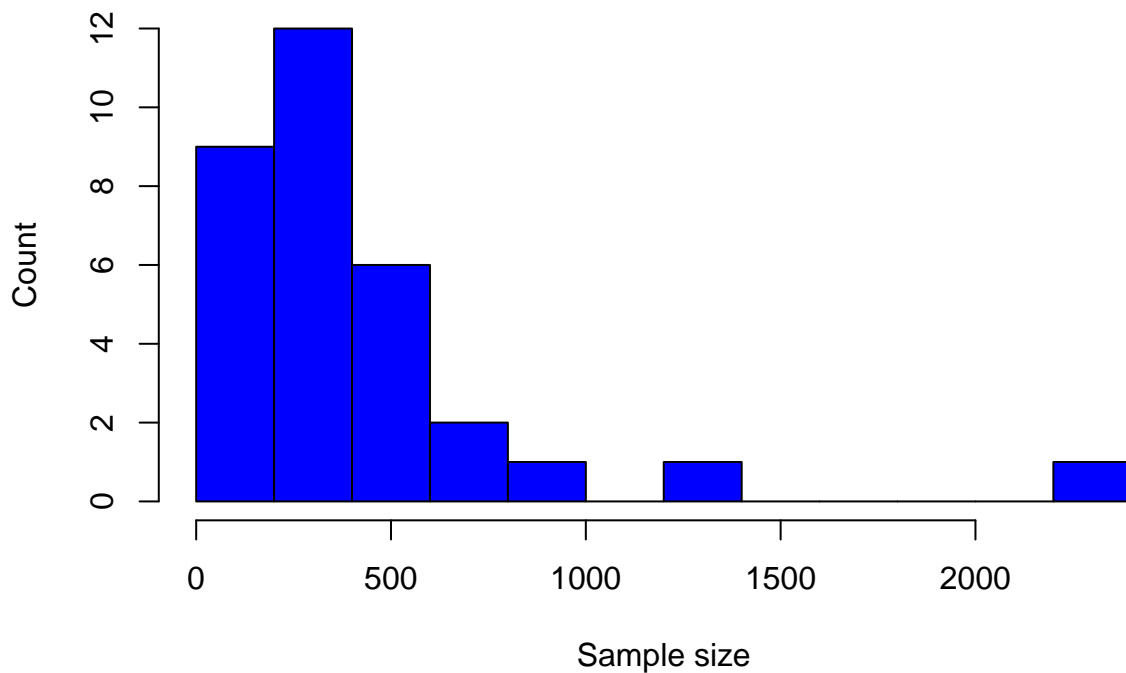
##      W1      S1      W2      S2
## W1 12906 12906 12906 12906
## S1 12906 12906 12906 12906
## W2 12906 12906 12906 12906
## S2 12906 12906 12906 12906

## Sample sizes
summary(Nohe15A1$n)

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      66.0   158.2   247.5   403.3   464.0  2235.0

## Histogram of the sample sizes
hist(Nohe15A1$n, main="Distribution of the sample sizes",
     xlab="Sample size", ylab="Count", col="blue", breaks=10)
```

Distribution of the sample sizes



FIMASEM

```
## Set seed for reproducibility
set.seed(39128482)
```

```
## Average the correlation coefficients with the univariate-r approach
uni1 <- uniR1(Nohe15A1$data, Nohe15A1$n)
uni1
```

```
##
## Total sample sizes: 12906
## Harmonic mean of the sample sizes: 12906
##
## Average correlation matrix:
##      W1      S1      W2      S2
## W1 1.0000000 0.4074043 0.6126003 0.3191213
## S1 0.4074043 1.0000000 0.3146358 0.6182109
## W2 0.6126003 0.3146358 1.0000000 0.4161150
## S2 0.3191213 0.6182109 0.4161150 1.0000000
##
## Sampling error (SE) of the average correlation matrix:
##      W1      S1      W2      S2
## W1      NA 0.04158106 0.03114614 0.04477883
## S1 0.04158106      NA 0.04492055 0.03080186
## W2 0.03114614 0.04492055      NA 0.04122343
## S2 0.04477883 0.03080186 0.04122343      NA
##
## Population heterogeneity (SD) of the average correlation matrix:
##      W1      S1      W2      S2
## W1      NA 0.14342795 0.0954675 0.11902793
## S1 0.1434279      NA 0.1129741 0.09397904
## W2 0.0954675 0.1129741      NA 0.14548415
## S2 0.1190279 0.09397904 0.1454842      NA
```

```
## Generate random correlation matrices with parametric bootstrap
boot.cor <- bootuniR1(uni1, Rep=500)
```

```
## Display the summary of the generated correlation matrices
summary(boot.cor)
```

```
## Population Sigma:
##      W1      S1      W2      S2
## W1 1.0000000 0.4074043 0.6126003 0.3191213
## S1 0.4074043 1.0000000 0.3146358 0.6182109
## W2 0.6126003 0.3146358 1.0000000 0.4161150
## S2 0.3191213 0.6182109 0.4161150 1.0000000
##
## Population V:
##      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]
## [1,] 0.02057158 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000
## [2,] 0.00000000 0.009114044 0.00000000 0.00000000 0.00000000 0.00000000
## [3,] 0.00000000 0.00000000 0.01416765 0.00000000 0.00000000 0.00000000
## [4,] 0.00000000 0.00000000 0.00000000 0.01276316 0.00000000 0.00000000
## [5,] 0.00000000 0.00000000 0.00000000 0.00000000 0.00883206 0.00000000
## [6,] 0.00000000 0.00000000 0.00000000 0.00000000 0.00000000 0.02116564
##
## Sample R (or S):
##      W1      S1      W2      S2
## W1 1.0000000 0.3950748 0.6074085 0.3275886
## S1 0.3950748 1.0000000 0.3115342 0.6166034
```

```

## W2 0.6074085 0.3115342 1.0000000 0.4002967
## S2 0.3275886 0.6166034 0.4002967 1.0000000
##
## Sample V:
##           [,1]           [,2]           [,3]           [,4]           [,5]
## [1,] 1.983929e-02 -0.0006395606 4.321777e-05 -3.712378e-04 -0.0006069681
## [2,] -6.395606e-04 0.0092941981 1.337491e-03 6.517486e-04 -0.0005487748
## [3,] 4.321777e-05 0.0013374908 1.524212e-02 6.315026e-05 0.0002264202
## [4,] -3.712378e-04 0.0006517486 6.315026e-05 1.279211e-02 0.0005280506
## [5,] -6.069681e-04 -0.0005487748 2.264202e-04 5.280506e-04 0.0086174415
## [6,] -2.065454e-03 -0.0011528208 9.548620e-04 -1.196961e-04 0.0001615825
##           [,6]
## [1,] -0.0020654543
## [2,] -0.0011528208
## [3,] 0.0009548620
## [4,] -0.0001196961
## [5,] 0.0001615825
## [6,] 0.0190725054
##
## Method to handle non-positive definite matrices: replace
## Number of samples: 500
## Count of non-positive definite matrices: 44
## Proposed saturated model
model1 <- 'W2 + S2 ~ W1 + S1'

boot1.fit <- bootuniR2(model=model1, data=boot.cor, n=uni1$n.harmonic)
summary(boot1.fit)

## Summary of the coefficients:
##           Mean           SD Quantile: 0% Quantile: 10% Quantile: 50%
## W2~W1 0.591406 0.135107 -0.030510 0.423918 0.587254
## W2~S1 0.074748 0.176960 -0.687941 -0.165838 0.094188
## S2~W1 0.089386 0.191929 -0.712109 -0.159295 0.102208
## S2~S1 0.594912 0.141675 0.181928 0.423943 0.590140
## W2~~W2 0.591081 0.121878 0.156670 0.433570 0.596585
## S2~~S2 0.574565 0.124050 0.191970 0.399188 0.590155
## W2~~S2 0.157501 0.173790 -0.283522 -0.075474 0.167106
##           Quantile: 90% Quantile: 100%
## W2~W1 0.760348 1.1736
## W2~S1 0.292507 0.6286
## S2~W1 0.318158 0.5799
## S2~S1 0.765699 1.2256
## W2~~W2 0.746833 0.9053
## S2~~S2 0.729497 0.8285
## W2~~S2 0.377395 0.5924
##
## Summary of the goodness-of-fit indices:
##           Mean           SD Quantile: 0% Quantile: 10% Quantile: 50%
## chisq 1.0317e-11 1.4721e-11 0.0000e+00 0.0000e+00 0.0000e+00
## cfi 1.0000e+00 8.9998e-16 1.0000e+00 1.0000e+00 1.0000e+00
## srmr 7.0806e-09 5.9148e-09 4.5142e-13 1.2181e-09 5.2271e-09
## rmsea 0.0000e+00 0.0000e+00 0.0000e+00 0.0000e+00 0.0000e+00
##           Quantile: 90% Quantile: 100%
## chisq 3.4388e-11 0

```



```

## cfi      1.0000e+00      1
## srmr     1.6702e-08      0
## rmsea    0.0000e+00      0
##
## Chisq test (df): 0
## Percentage of rejection ( 0.05 ): NaN
## Percentage of CFI > 0.9 : 100
## Percentage of SRMR < 0.1 : 100
## Percentage of RMSEA < 0.05 : 100
## Number of total replications: 500
## Number of successful replications: 500

## Proposed model with equal regression coefficients
model2 <- 'W2 ~ Same*W1 + Cross*S1
          S2 ~ Cross*W1 + Same*S1
          W1 ~~ Cor1*S1
          W2 ~~ Cor2*S2'

boot2.fit <- bootuniR2(model=model2, data=boot.cor, n=uni1$n.harmonic)
summary(boot2.fit)

## Summary of the coefficients:
##           Mean           SD Quantile: 0% Quantile: 10% Quantile: 50%
## Same      6.0793e-01  1.0324e-01  3.2339e-01  4.8337e-01  6.0005e-01
## Cross      8.3327e-02  1.5355e-01 -4.8197e-01 -1.1581e-01  9.4636e-02
## Cross      8.3327e-02  1.5355e-01 -4.8197e-01 -1.1581e-01  9.4636e-02
## Same      6.0793e-01  1.0324e-01  3.2339e-01  4.8337e-01  6.0005e-01
## Cor1       3.9504e-01  1.4084e-01 -3.7642e-02  2.1759e-01  3.9667e-01
## Cor2       1.6516e-01  1.7516e-01 -2.7973e-01 -6.9321e-02  1.7292e-01
## W2~~W2      6.0852e-01  1.2991e-01  1.7256e-01  4.4094e-01  6.0912e-01
## S2~~S2      5.9060e-01  1.2988e-01  1.9718e-01  4.1072e-01  5.9971e-01
## W1~~W1      9.9992e-01  1.6969e-07  9.9992e-01  9.9992e-01  9.9992e-01
## S1~~S1      9.9992e-01  1.6969e-07  9.9992e-01  9.9992e-01  9.9992e-01
##           Quantile: 90% Quantile: 100%
## Same       7.2303e-01      1.0301
## Cross       2.5639e-01      0.4667
## Cross       2.5639e-01      0.4667
## Same       7.2303e-01      1.0301
## Cor1        5.7121e-01      0.7622
## Cor2        3.8778e-01      0.5988
## W2~~W2      7.6367e-01      1.0853
## S2~~S2      7.5113e-01      1.0150
## W1~~W1      9.9992e-01      0.9999
## S1~~S1      9.9992e-01      0.9999
##
## Summary of the goodness-of-fit indices:
##           Mean           SD Quantile: 0% Quantile: 10% Quantile: 50%
## chisq      7.3840e+02  8.5148e+02  9.8525e-01  6.3337e+01  4.4021e+02
## cfi        9.6477e-01  3.6509e-02  8.0167e-01  9.1332e-01  9.7703e-01
## srmr       6.1901e-02  4.6919e-02  3.7242e-03  2.0187e-02  5.1474e-02
## rmsea      1.4585e-01  8.5276e-02  0.0000e+00  4.8747e-02  1.3030e-01
##           Quantile: 90% Quantile: 100%
## chisq      1.9679e+03      5914.1954
## cfi         9.9655e-01      1.0000
## srmr        1.0990e-01      0.4745

```

```
## rmsea      2.7597e-01      0.4786
##
## Chisq test (df): 2
## Percentage of rejection ( 0.05 ): 99.4
## Percentage of CFI > 0.9 : 92.4
## Percentage of SRMR < 0.1 : 86.6
## Percentage of RMSEA < 0.05 : 10.8
## Number of total replications: 500
## Number of successful replications: 500
```

TSSEM

```
rand1 <- tssem1(Nohe15A1$data, Nohe15A1$n, method="REM", RE.type="Diag")
summary(rand1)
```

```
##
## Call:
## meta(y = ES, v = acovR, RE.constraints = Diag(paste0(RE.startvalues,
##      "*Tau2_", 1:no.es, "_", 1:no.es)), RE.lbound = RE.lbound,
##      I2 = I2, model.name = model.name, suppressWarnings = TRUE,
##      silent = silent, run = run)
##
## 95% confidence intervals: z statistic approximation
## Coefficients:
##      Estimate Std.Error    lbound    ubound z value Pr(>|z|)
## Intercept1 0.3804522 0.0225615 0.3362323 0.4246720 16.8629 < 2.2e-16 ***
## Intercept2 0.6051298 0.0180362 0.5697794 0.6404802 33.5508 < 2.2e-16 ***
## Intercept3 0.3032290 0.0178803 0.2681842 0.3382738 16.9588 < 2.2e-16 ***
## Intercept4 0.3036392 0.0178408 0.2686718 0.3386065 17.0194 < 2.2e-16 ***
## Intercept5 0.6166503 0.0166427 0.5840312 0.6492694 37.0523 < 2.2e-16 ***
## Intercept6 0.3954085 0.0216645 0.3529469 0.4378701 18.2515 < 2.2e-16 ***
## Tau2_1_1    0.0134777 0.0038704 0.0058919 0.0210635  3.4823 0.0004972 ***
## Tau2_2_2    0.0087592 0.0025260 0.0038083 0.0137102  3.4676 0.0005252 ***
## Tau2_3_3    0.0071123 0.0022470 0.0027082 0.0115163  3.1652 0.0015496 **
## Tau2_4_4    0.0070585 0.0022121 0.0027229 0.0113941  3.1909 0.0014183 **
## Tau2_5_5    0.0072633 0.0021092 0.0031293 0.0113974  3.4436 0.0005740 ***
## Tau2_6_6    0.0122813 0.0034848 0.0054513 0.0191113  3.5243 0.0004246 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Q statistic on the homogeneity of effect sizes: 1466.161
## Degrees of freedom of the Q statistic: 186
## P value of the Q statistic: 0
##
## Heterogeneity indices (based on the estimated Tau2):
##      Estimate
## Intercept1: I2 (Q statistic) 0.8829
## Intercept2: I2 (Q statistic) 0.8973
## Intercept3: I2 (Q statistic) 0.7743
## Intercept4: I2 (Q statistic) 0.7718
## Intercept5: I2 (Q statistic) 0.8810
## Intercept6: I2 (Q statistic) 0.8748
```

```

##
## Number of studies (or clusters): 32
## Number of observed statistics: 192
## Number of estimated parameters: 12
## Degrees of freedom: 180
## -2 log likelihood: -300.1702
## OpenMx status1: 0 ("0" or "1": The optimization is considered fine.
## Other values may indicate problems.)

## SDs of the correlations (heterogeneity)
sqrt(coef(rand1, select="rand"))

##   Tau2_1_1   Tau2_2_2   Tau2_3_3   Tau2_4_4   Tau2_5_5   Tau2_6_6
## 0.11609350 0.09359083 0.08433432 0.08401477 0.08522529 0.11082105

## Cross-lagged panel model
model3 <- 'W2 ~ W2W*W1 + S2W*S1
          S2 ~ W2S*W1 + S2S*S1
          W1 ~~ W1cS1*S1
          W2 ~~ W2cS2*S2
          W1 ~~ 1*W1
          S1 ~~ 1*S1'

## Convert the lavaan model to the RAM model
RAM1 <- lavaan2RAM(model3, obs.variables=c("W1", "S1", "W2", "S2"))

rand2a <- tssem2(rand1, Amatrix=RAM1$A, Smatrix=RAM1$S)
summary(rand2a)

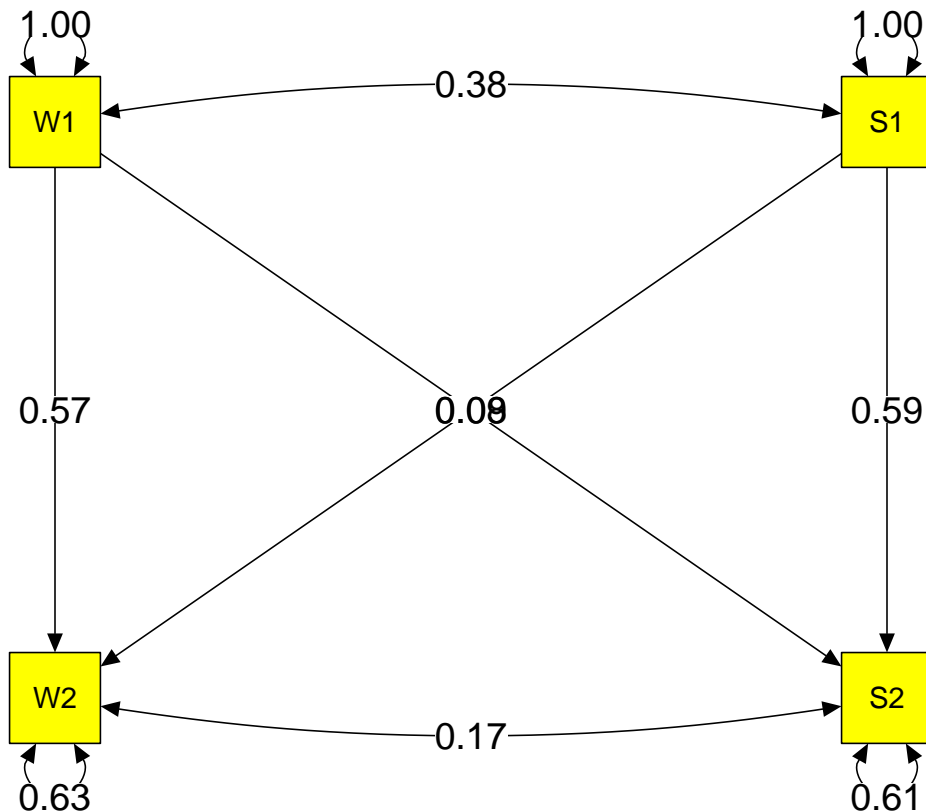
##
## Call:
## wls(Cov = pooledS, aCov = aCov, n = tssem1.obj$total.n, Amatrix = Amatrix,
##      Smatrix = Smatrix, Fmatrix = Fmatrix, diag.constraints = diag.constraints,
##      cor.analysis = cor.analysis, intervals.type = intervals.type,
##      mx.algebras = mx.algebras, model.name = model.name, suppressWarnings = suppressWarnings,
##      silent = silent, run = run)
##
## 95% confidence intervals: z statistic approximation
## Coefficients:
##      Estimate Std.Error   lbound   ubound z value  Pr(>|z|)
## S2S    0.586124  0.020790  0.545376  0.626872 28.1926 < 2.2e-16 ***
## W2S    0.080237  0.024842  0.031547  0.128926  3.2299 0.0012385 **
## S2W    0.085841  0.024796  0.037242  0.134440  3.4619 0.0005363 ***
## W2W    0.572471  0.022265  0.528834  0.616109 25.7122 < 2.2e-16 ***
## W1cS1  0.380452  0.022562  0.336232  0.424672 16.8629 < 2.2e-16 ***
## W2cS2  0.168885  0.025232  0.119431  0.218338  6.6933 2.182e-11 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Goodness-of-fit indices:
##
##      Sample size      Value
## Chi-square of target model      0
## DF of target model      0
## p value of target model      0

```

```
## Number of constraints imposed on "Smatrix"      0
## DF manually adjusted                          0
## Chi-square of independence model              3079
## DF of independence model                      6
## RMSEA                                         0
## RMSEA lower 95% CI                          0
## RMSEA upper 95% CI                          0
## SRMR                                         0
## TLI                                         -Inf
## CFI                                           1
## AIC                                           0
## BIC                                           0
## OpenMx status1: 0 ("0" or "1": The optimization is considered fine.
## Other values indicate problems.)
```

```
## Plot the model
```

```
my.plot1 <- meta2semPlot(rand2a)
semPaths(my.plot1, whatLabels="est", sizeMan=8, edge.label.cex=1.5,
         color="yellow", edge.color = "black", weighted=FALSE)
```



```
## Model with equal regression coefficients
```

```
model4 <- 'W2 ~ Same*W1 + Cross*S1
          S2 ~ Cross*W1 + Same*S1
          W1 ~~ W1cS1*S1
          W2 ~~ W2cS2*S2
          W1 ~~ 1*W1
          S1 ~~ 1*S1'
```

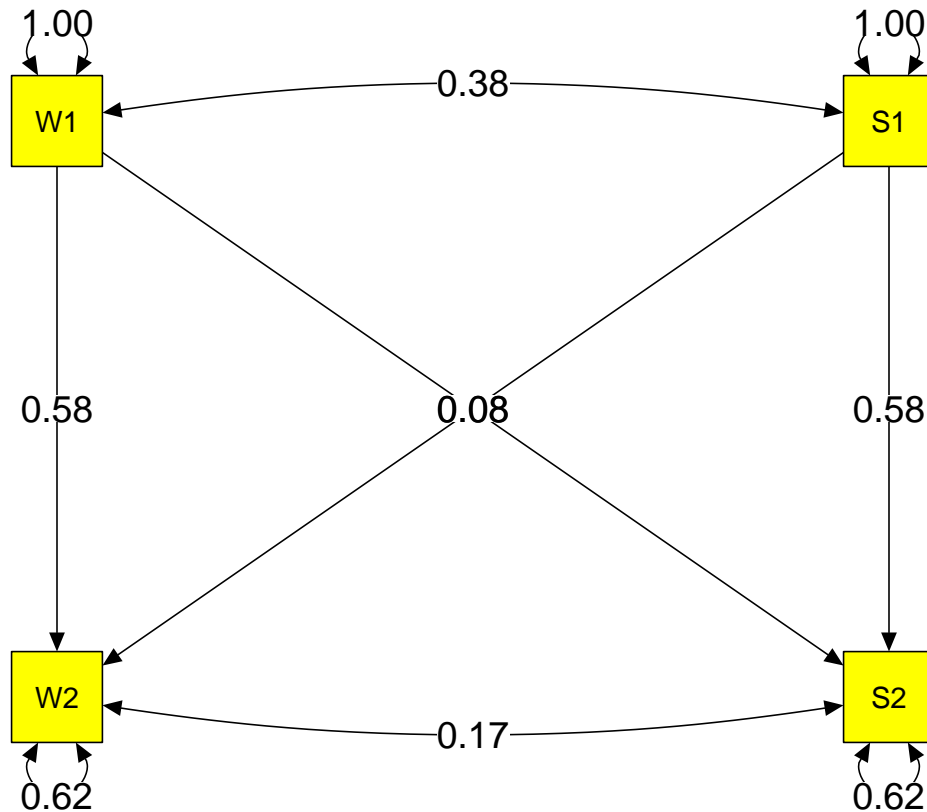
```
RAM2 <- lavaan2RAM(model4, obs.variables=c("W1", "S1", "W2", "S2"))
```

```

rand2b <- tssem2(rand1, Amatrix=RAM2$A, Smatrix=RAM2$S)
summary(rand2b)

##
## Call:
## wls(Cov = pooledS, aCov = aCov, n = tssem1.obj$total.n, Amatrix = Amatrix,
##      Smatrix = Smatrix, Fmatrix = Fmatrix, diag.constraints = diag.constraints,
##      cor.analysis = cor.analysis, intervals.type = intervals.type,
##      mx.algebras = mx.algebras, model.name = model.name, suppressWarnings = suppressWarnings,
##      silent = silent, run = run)
##
## 95% confidence intervals: z statistic approximation
## Coefficients:
##      Estimate Std.Error   lbound   ubound z value Pr(>|z|)
## Cross 0.082840  0.019712 0.044206 0.121474  4.2026 2.638e-05 ***
## Same  0.579844  0.015217 0.550020 0.609669 38.1054 < 2.2e-16 ***
## W1cS1 0.380451  0.022562 0.336231 0.424671 16.8628 < 2.2e-16 ***
## W2cS2 0.168824  0.025241 0.119353 0.218295  6.6886 2.254e-11 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Goodness-of-fit indices:
##
##                                     Value
## Sample size                        12906.0000
## Chi-square of target model          0.2247
## DF of target model                  2.0000
## p value of target model             0.8937
## Number of constraints imposed on "Smatrix" 0.0000
## DF manually adjusted                0.0000
## Chi-square of independence model    3078.9538
## DF of independence model            6.0000
## RMSEA                              0.0000
## RMSEA lower 95% CI                 0.0000
## RMSEA upper 95% CI                 0.0079
## SRMR                              0.0033
## TLI                               1.0017
## CFI                               1.0000
## AIC                               -3.7753
## BIC                               -18.7062
## OpenMx status1: 0 ("0" or "1": The optimization is considered fine.
## Other values indicate problems.)
my.plot2 <- meta2semPlot(rand2b)
semPaths(my.plot2, whatLabels="est", sizeMan=8, edge.label.cex=1.5,
         color="yellow", edge.color = "black", weighted=FALSE)

```



sessionInfo()

```

## R version 3.5.1 (2018-07-02)
## Platform: x86_64-pc-linux-gnu (64-bit)
## Running under: Ubuntu 18.04 LTS
##
## Matrix products: default
## BLAS: /usr/lib/x86_64-linux-gnu/blas/libblas.so.3.7.1
## LAPACK: /usr/lib/x86_64-linux-gnu/lapack/liblapack.so.3.7.1
##
## locale:
##  [1] LC_CTYPE=en_US.utf8      LC_NUMERIC=C
##  [3] LC_TIME=en_US.utf8      LC_COLLATE=en_US.utf8
##  [5] LC_MONETARY=en_US.utf8  LC_MESSAGES=en_US.utf8
##  [7] LC_PAPER=en_US.utf8     LC_NAME=C
##  [9] LC_ADDRESS=C            LC_TELEPHONE=C
## [11] LC_MEASUREMENT=en_US.utf8 LC_IDENTIFICATION=C
##
## attached base packages:
## [1] stats      graphics  grDevices  utils      datasets  methods    base
##
## other attached packages:
## [1] semPlot_1.1      metaSEM_1.1.1    OpenMx_2.9.9     lavaan_0.6-1
## [5] rmarkdown_1.10
##
## loaded via a namespace (and not attached):
##  [1] nlme_3.1-137      RColorBrewer_1.1-2  rprojroot_1.3-2
##  [4] mi_1.0            tools_3.5.1         backports_1.1.2

```

## [7]	R6_2.2.2	d3Network_0.5.2.1	rpart_4.1-13
## [10]	Hmisc_4.1-1	lazyeval_0.2.1	colorspace_1.3-2
## [13]	nnet_7.3-12	tidyselect_0.2.4	gridExtra_2.3
## [16]	mnormt_1.5-5	curl_3.2	compiler_3.5.1
## [19]	fdrtool_1.2.15	qgraph_1.5	htmlTable_1.12
## [22]	network_1.13.0.1	scales_0.5.0	checkmate_1.8.5
## [25]	mvtnorm_1.0-8	psych_1.8.4	pbapply_1.3-4
## [28]	sem_3.1-9	stringr_1.3.1	digest_0.6.15
## [31]	pbivnorm_0.6.0	foreign_0.8-70	minqa_1.2.4
## [34]	rio_0.5.10	base64enc_0.1-3	jpeg_0.1-8
## [37]	pkgconfig_2.0.1	htmltools_0.3.6	lme4_1.1-17
## [40]	lisrelToR_0.1.4	htmlwidgets_1.2	rlang_0.2.1
## [43]	readxl_1.1.0	huge_1.2.7	rstudioapi_0.7
## [46]	bindr_0.1.1	gtools_3.8.1	statnet.common_4.1.4
## [49]	acepack_1.4.1	dplyr_0.7.6	zip_1.0.0
## [52]	car_3.0-0	magrittr_1.5	Formula_1.2-3
## [55]	Matrix_1.2-14	Rcpp_0.12.17	munsell_0.5.0
## [58]	abind_1.4-5	rockchalk_1.8.111	whisker_0.3-2
## [61]	stringi_1.2.3	yaml_2.1.19	carData_3.0-1
## [64]	MASS_7.3-50	plyr_1.8.4	matrixcalc_1.0-3
## [67]	grid_3.5.1	parallel_3.5.1	forcats_0.3.0
## [70]	lattice_0.20-35	haven_1.1.2	splines_3.5.1
## [73]	hms_0.4.2	sna_2.4	knitr_1.20
## [76]	pillar_1.2.3	igraph_1.2.1	rjson_0.2.20
## [79]	boot_1.3-20	corpcor_1.6.9	BDgraph_2.51
## [82]	reshape2_1.4.3	stats4_3.5.1	XML_3.98-1.11
## [85]	glue_1.2.0	evaluate_0.10.1	latticeExtra_0.6-28
## [88]	data.table_1.11.4	png_0.1-7	nloptr_1.0.4
## [91]	cellranger_1.1.0	gtable_0.2.0	purrr_0.2.5
## [94]	assertthat_0.2.0	ggplot2_3.0.0	openxlsx_4.1.0
## [97]	semTools_0.5-0	coda_0.19-1	glasso_1.8
## [100]	survival_2.42-3	tibble_1.4.2	arm_1.10-1
## [103]	ggm_2.3	ellipse_0.4.1	bindrcpp_0.2.2
## [106]	cluster_2.0.7-1		