

Many missing correlations

Robert Balazsi (modified by Mike Cheung)

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<https://openmx.ssri.psu.edu/node/4583>

```
library(metaSEM)

## Loading required package: OpenMx

## To take full advantage of multiple cores, use:
##   mxOption(key='Number of Threads', value=parallel::detectCores()) #now
##   Sys.setenv(OMP_NUM_THREADS=parallel::detectCores()) #before library(OpenMx)

## "SLSQP" is set as the default optimizer in OpenMx.
## mxOption(NULL, "Gradient algorithm") is set at "central".
## mxOption(NULL, "Optimality tolerance") is set at "6.3e-14".
## mxOption(NULL, "Gradient iterations") is set at "2".

data <- read.table(file = "metaSEM4_0.dat", header = TRUE)
head(data)

##   Study   N v1_v2 v1_v3 v1_v4 v1_v5 v1_v6 v2_v3 v2_v4 v2_v5 v2_v6 v3_v4 v3_v5
## 1     1 174   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA 0.24
## 2     2 445   NA   NA   NA   NA   NA   NA 0.58   NA   NA   NA   NA   NA
## 3     3  32   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA
## 4     4  33   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA
## 5     5  39   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA 0.73
## 6     6  39   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA   NA 0.59
##   v3_v6 v4_v5 v4_v6 v5_v6
## 1   NA   NA   NA   NA
## 2   NA 0.40   NA   NA
## 3   NA 0.50   NA   NA
## 4   NA 0.04   NA   NA
## 5   NA   NA   NA   NA
## 6   NA   NA   NA   NA

nvar <- 6
varnames <- c("v1 ", "v2", "v3", "v4", "v5", "v6")
labels <- list(varnames, varnames)
cordat <- list()
for (i in 1:nrow(data)){
  cordat[[i]] <- vec2symMat(as.matrix(data[i,3:17]),
    diag = FALSE)
  dimnames(cordat[[i]]) <- labels
}
}
```

```

for (i in 1:length(cordat)){
  for (j in 1:nrow(cordat[[i]])){
    if (sum(is.na(cordat[[i]][j,]))==nvar-1)
      {cordat[[i]][j,j] <- NA}
  }}

```

```
cordat[[13]]
```

```

##      v1 v2 v3 v4 v5 v6
## v1  NA NA NA NA NA NA
## v2  NA NA NA NA NA NA
## v3  NA NA 1.0 NA 0.4 NA
## v4  NA NA NA NA NA NA
## v5  NA NA 0.4 NA 1.0 NA
## v6  NA NA NA NA NA NA

```

```

for (i in 1:length(cordat)){
  for (j in 1:nrow(cordat[[i]])){
    for (k in 1:nvar){
      if (is.na(cordat[[i]][j,k])==TRUE
          &is.na(cordat[[i]][j,j])!=TRUE
          &is.na(cordat[[i]][k,k])!=TRUE){
        if(sum(is.na(cordat[[i]][j,])>sum(is.na(cordat[[i]])[k,]))
            {cordat[[i]][k,k] <- NA}
        if(sum(is.na(cordat[[i]][j,])<=sum(is.na(cordat[[i]])[k,]))
            {cordat[[i]][j,j] <- NA}
      }}}

```

```

# stageifixed <- tssem1(Cov=cordat, n=data$N, method="REM", RE.type="Zero")
# summary(stageifixed)

```

```

stageirandom <- tssem1(Cov=cordat, n=data$N, method="REM")
stageirandom <- rerun(stageirandom, autofixtau2 = TRUE)

```

```
## Running TSSEM1 Correlation with 30 parameters
```

```
##
```

```
## Beginning initial fit attempt
```

```
## Running TSSEM1 Correlation with 30 parameters
```

```
##
```

```
## Lowest minimum so far: -341.750622295448
```

```
## Not all eigenvalues of the Hessian are positive: 401135.607944602, 147491.847684654, 76762.11504876
```

```
##
```

```
## Beginning fit attempt 1 of at maximum 10 extra tries
```

```
## Running TSSEM1 Correlation with 30 parameters
```

```
## Not all eigenvalues of the Hessian are positive: 401135.608918055, 147491.849048463, 76762.11685192
```

```
##
```

```
## Beginning fit attempt 2 of at maximum 10 extra tries
```

```
## Running TSSEM1 Correlation with 30 parameters
```

```
##
```

```
## Lowest minimum so far: -341.750622296416
## Not all eigenvalues of the Hessian are positive: 401135.836153742, 147491.754562791, 76762.29615357
##
## Beginning fit attempt 3 of at maximum 10 extra tries
## Running TSSEM1 Correlation with 30 parameters
##
## Lowest minimum so far: -341.750622296958
## Not all eigenvalues of the Hessian are positive: 401135.80845603, 147492.325551991, 76762.229089849
##
## Beginning fit attempt 4 of at maximum 10 extra tries
## Running TSSEM1 Correlation with 30 parameters
## Not all eigenvalues of the Hessian are positive: 401135.806585092, 147492.322151334, 76762.22625201
##
## Beginning fit attempt 5 of at maximum 10 extra tries
## Running TSSEM1 Correlation with 30 parameters
## Not all eigenvalues of the Hessian are positive: 401135.490994888, 147475.919946766, 76762.63351326
##
## Beginning fit attempt 6 of at maximum 10 extra tries
## Running TSSEM1 Correlation with 30 parameters
## Not all eigenvalues of the Hessian are positive: 401135.790675054, 147491.763703245, 76762.02238162
##
## Beginning fit attempt 7 of at maximum 10 extra tries
## Running TSSEM1 Correlation with 30 parameters
##
## Lowest minimum so far: -341.750622297403
## Not all eigenvalues of the Hessian are positive: 401135.828641763, 147491.952194547, 76762.18997703
##
## Beginning fit attempt 8 of at maximum 10 extra tries
## Running TSSEM1 Correlation with 30 parameters
## Not all eigenvalues of the Hessian are positive: 401135.830588388, 147491.954158329, 76762.19174933
##
## Beginning fit attempt 9 of at maximum 10 extra tries
## Running TSSEM1 Correlation with 30 parameters
## Not all eigenvalues of the Hessian are positive: 401135.887447516, 147492.203907797, 76762.33697663
##
## Beginning fit attempt 10 of at maximum 10 extra tries
## Running TSSEM1 Correlation with 30 parameters
## Not all eigenvalues of the Hessian are positive: 401135.783411033, 147491.621011274, 76762.31152382
##
## Retry limit reached
```

```
##
## Retry limit reached; Best fit=-341.75062 (started at -341.75062) (11 attempt(s): 11 valid, 0 errors)
## Running TSSEM1 Correlation with 29 parameters
##
## Beginning initial fit attempt
## Running TSSEM1 Correlation with 29 parameters
##
## Solution found
##
## Solution found! Final fit=-341.75062 (started at -341.75062) (1 attempt(s): 1 valid, 0 errors)
```

`summary(stage1random)`

```
##
## 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 (robust=FALSE)
## Coefficients:
```

	Estimate	Std.Error	lbound	ubound	z value	Pr(> z)
## Intercept1	3.5674e-02	3.8227e-02	-3.9248e-02	1.1060e-01	0.9332	0.350700
## Intercept2	1.0534e-01	1.6971e-01	-2.2728e-01	4.3796e-01	0.6207	0.534775
## Intercept3	4.9244e-01	5.1627e-02	3.9126e-01	5.9363e-01	9.5385	< 2.2e-16
## Intercept4	7.9443e-02	5.8848e-02	-3.5897e-02	1.9478e-01	1.3500	0.177028
## Intercept5	1.4851e-01	1.0990e-01	-6.6900e-02	3.6391e-01	1.3512	0.176618
## Intercept6	1.3393e-01	2.4096e-02	8.6707e-02	1.8116e-01	5.5583	2.724e-08
## Intercept7	2.2790e-01	3.4890e-02	1.5952e-01	2.9629e-01	6.5320	6.491e-11
## Intercept8	2.1995e-01	1.0410e-02	1.9954e-01	2.4035e-01	21.1278	< 2.2e-16
## Intercept9	2.8060e-01	3.7708e-02	2.0669e-01	3.5450e-01	7.4412	9.970e-14
## Intercept10	1.6877e-01	9.4175e-02	-1.5814e-02	3.5334e-01	1.7920	0.073127
## Intercept11	3.1030e-01	2.5045e-02	2.6122e-01	3.5939e-01	12.3899	< 2.2e-16
## Intercept12	1.2881e-01	1.2664e-01	-1.1941e-01	3.7702e-01	1.0171	0.309117
## Intercept13	2.6362e-01	2.4469e-02	2.1567e-01	3.1158e-01	10.7739	< 2.2e-16
## Intercept14	-5.7899e-02	5.7696e-02	-1.7098e-01	5.5184e-02	-1.0035	0.315614
## Intercept15	3.5509e-01	3.3831e-02	2.8878e-01	4.2139e-01	10.4959	< 2.2e-16
## Tau2_2_2	5.1881e-02	5.7494e-02	-6.0805e-02	1.6457e-01	0.9024	0.366860
## Tau2_3_3	1.0072e-10	5.7266e-03	-1.1224e-02	1.1224e-02	0.0000	1.000000
## Tau2_4_4	3.5346e-02	1.8504e-02	-9.2209e-04	7.1614e-02	1.9101	0.056116
## Tau2_5_5	7.2718e-02	4.6164e-02	-1.7762e-02	1.6320e-01	1.5752	0.115208
## Tau2_6_6	8.1903e-03	3.7012e-03	9.3599e-04	1.5445e-02	2.2128	0.026908
## Tau2_7_7	1.9862e-02	9.1421e-03	1.9440e-03	3.7780e-02	2.1726	0.029810
## Tau2_8_8	1.7525e-02	2.2351e-03	1.3144e-02	2.1906e-02	7.8408	4.441e-15
## Tau2_9_9	1.2313e-02	7.2248e-03	-1.8471e-03	2.6474e-02	1.7043	0.088325
## Tau2_10_10	3.0139e-02	2.6186e-02	-2.1185e-02	8.1463e-02	1.1509	0.249754
## Tau2_11_11	2.6842e-02	6.8011e-03	1.3512e-02	4.0172e-02	3.9466	7.926e-05
## Tau2_12_12	8.4210e-02	5.6044e-02	-2.5635e-02	1.9405e-01	1.5026	0.132954
## Tau2_13_13	1.5596e-02	5.1774e-03	5.4482e-03	2.5743e-02	3.0123	0.002593
## Tau2_14_14	4.3752e-02	2.0684e-02	3.2112e-03	8.4293e-02	2.1152	0.034412
## Tau2_15_15	9.6711e-03	5.6447e-03	-1.3923e-03	2.0735e-02	1.7133	0.086657

```

##
## Intercept1
## Intercept2
## Intercept3 ***
## Intercept4
## Intercept5
## Intercept6 ***
## Intercept7 ***
## Intercept8 ***
## Intercept9 ***
## Intercept10 .
## Intercept11 ***
## Intercept12
## Intercept13 ***
## Intercept14
## Intercept15 ***
## Tau2_2_2
## Tau2_3_3
## Tau2_4_4 .
## Tau2_5_5
## Tau2_6_6 *
## Tau2_7_7 *
## Tau2_8_8 ***
## Tau2_9_9 .
## Tau2_10_10
## Tau2_11_11 ***
## Tau2_12_12
## Tau2_13_13 **
## Tau2_14_14 *
## Tau2_15_15 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Q statistic on the homogeneity of effect sizes: 2673.115
## Degrees of freedom of the Q statistic: 423
## P value of the Q statistic: 0
##
## Heterogeneity indices (based on the estimated Tau2):
##
##           Estimate
## Intercept1: I2 (Q statistic)  0.0000
## Intercept2: I2 (Q statistic)  0.9474
## Intercept3: I2 (Q statistic)  0.0000
## Intercept4: I2 (Q statistic)  0.9246
## Intercept5: I2 (Q statistic)  0.9619
## Intercept6: I2 (Q statistic)  0.7398
## Intercept7: I2 (Q statistic)  0.8733
## Intercept8: I2 (Q statistic)  0.8643
## Intercept9: I2 (Q statistic)  0.8104
## Intercept10: I2 (Q statistic) 0.9127
## Intercept11: I2 (Q statistic) 0.9032
## Intercept12: I2 (Q statistic) 0.9669
## Intercept13: I2 (Q statistic) 0.8441
## Intercept14: I2 (Q statistic) 0.9382
## Intercept15: I2 (Q statistic) 0.7705

```

```
##
## Number of studies (or clusters): 379
## Number of observed statistics: 438
## Number of estimated parameters: 29
## Degrees of freedom: 409
## -2 log likelihood: -341.7506
## OpenMx status1: 0 ("0" or "1": The optimization is considered fine.
## Other values may indicate problems.)
```

```
A<-create.mxMatrix (c(0,0,0,0,0,0
,0,0,0,0,0,0,
0,0,0,0,0,0,
"0.1*b41", "0.1*b42", 0,0,0,0,
0,0, "0.1*b53", "0.1*b54", 0,0,
0,0, "0.1*b63", "0.1*b64", 0,0),
type = "Full",
  nrow = 6,
  ncol = 6,
  byrow = TRUE,
  name = "A",
  dimnames = list(varnames, varnames))
```

```
S <- create.mxMatrix(
c(1,
".1*p21", 1,
".1*p31", ".1*p32", 1,
0,0,0, "1*p44",
0,0,0,0, "1*p55",
0,0,0,0,0, "1*p66"),
type="Symm", byrow = TRUE, name="S",
  dimnames = list(varnames, varnames))
```

```
stage2 <- tssem2(stage1random, Amatrix=A, Smatrix=S,
  diag.constraints=FALSE, intervals="LB")
summary(stage2)
```

```
##
## Call:
## wls(Cov = pooledS, aCov = aCov, n = tssem1.obj$total.n, RAM = RAM,
## 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: Likelihood-based statistic
```

```
## Coefficients:
```

##	Estimate	Std.Error	lbound	ubound	z value	Pr(> z)
## b41	0.545374	NA	0.443233	0.643550	NA	NA
## b42	0.301758	NA	0.229143	0.373562	NA	NA
## b53	0.406600	NA	0.352499	0.465238	NA	NA
## b54	0.364211	NA	0.306955	0.426759	NA	NA
## b63	0.645278	NA	0.505137	0.790270	NA	NA
## b64	0.193998	NA	0.068458	0.311900	NA	NA
## p21	0.036556	NA	-0.039310	0.112407	NA	NA

```

## p31 -0.331146      NA -0.491288 -0.153820      NA      NA
## p32  0.197413      NA  0.152893  0.241686      NA      NA
##
## Goodness-of-fit indices:
##                                     Value
## Sample size                        1.3240e+05
## Chi-square of target model          8.3600e+01
## DF of target model                  6.0000e+00
## p value of target model             0.0000e+00
## Number of constraints imposed on "Smatrix" 0.0000e+00
## DF manually adjusted                0.0000e+00
## Chi-square of independence model    1.0520e+03
## DF of independence model            1.5000e+01
## RMSEA                               9.9000e-03
## RMSEA lower 95% CI                  8.1000e-03
## RMSEA upper 95% CI                  1.1800e-02
## SRMR                                2.0770e-01
## TLI                                 8.1290e-01
## CFI                                 9.2520e-01
## AIC                                 7.1600e+01
## BIC                                 1.2839e+01
## OpenMx status1: 0 ("0" or "1": The optimization is considered fine.
## Other values indicate problems.)

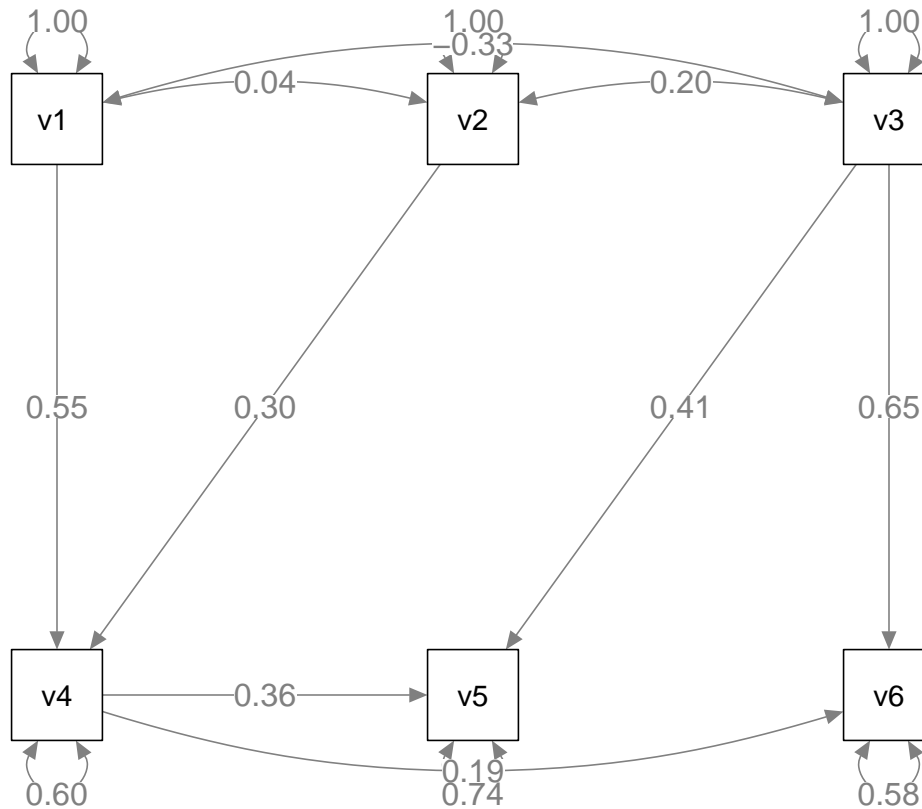
```

```
plot(stage2)
```

```

## Registered S3 methods overwritten by 'huge':
##   method      from
##   plot.sim    BDgraph
##   print.sim   BDgraph

```



```
sessionInfo()
```

```
## R version 3.6.2 (2019-12-12)
## Platform: x86_64-pc-linux-gnu (64-bit)
## Running under: Linux Mint 19.1
##
## Matrix products: default
## BLAS: /usr/lib/x86_64-linux-gnu/openblas/libblas.so.3
## LAPACK: /usr/lib/x86_64-linux-gnu/libopenblas-r0.2.20.so
##
## locale:
## [1] LC_CTYPE=en_SG.UTF-8      LC_NUMERIC=C
## [3] LC_TIME=en_SG.UTF-8      LC_COLLATE=en_SG.UTF-8
## [5] LC_MONETARY=en_SG.UTF-8  LC_MESSAGES=en_SG.UTF-8
## [7] LC_PAPER=en_SG.UTF-8     LC_NAME=C
## [9] LC_ADDRESS=C             LC_TELEPHONE=C
## [11] LC_MEASUREMENT=en_SG.UTF-8 LC_IDENTIFICATION=C
##
## attached base packages:
## [1] stats    graphics grDevices utils      datasets methods  base
##
## other attached packages:
## [1] metaSEM_1.2.3.1 OpenMx_2.15.5
##
## loaded via a namespace (and not attached):
## [1] nlme_3.1-143      RColorBrewer_1.1-2 mi_1.0
```



```

## [4] tools_3.6.2          backports_1.1.5      R6_2.4.1
## [7] d3Network_0.5.2.1    rpart_4.1-15        Hmisc_4.3-0
## [10] lazyeval_0.2.2       colorspace_1.4-1    nnet_7.3-12
## [13] tidyselect_0.2.5     gridExtra_2.3        mnormt_1.5-5
## [16] compiler_3.6.2       qgraph_1.6.4        fdrtool_1.2.15
## [19] htmlTable_1.13.3     regsem_1.3.9         scales_1.1.0
## [22] checkmate_1.9.4      mvtnorm_1.0-12      psych_1.9.12.31
## [25] pbapply_1.4-2        sem_3.1-9            stringr_1.4.0
## [28] digest_0.6.23        pbivnorm_0.6.0      foreign_0.8-75
## [31] minqa_1.2.4          rmarkdown_2.1        base64enc_0.1-3
## [34] jpeg_0.1-8           pkgconfig_2.0.3     htmltools_0.4.0
## [37] lme4_1.1-21          lisrelToR_0.1.4     htmlwidgets_1.5.1
## [40] rlang_0.4.2          huge_1.3.4           rstudioapi_0.10
## [43] gtools_3.8.1         acepack_1.4.1        dplyr_0.8.3
## [46] zip_2.0.4            magrittr_1.5         Formula_1.2-3
## [49] Matrix_1.2-18        Rcpp_1.0.3           munsell_0.5.0
## [52] abind_1.4-5          rockchalk_1.8.144   lifecycle_0.1.0
## [55] whisker_0.4          stringi_1.4.5        yaml_2.2.0
## [58] carData_3.0-3        MASS_7.3-51.5        plyr_1.8.5
## [61] matrixcalc_1.0-3     lavaan_0.6-5         grid_3.6.2
## [64] parallel_3.6.2       crayon_1.3.4         lattice_0.20-38
## [67] semPlot_1.1.3        kutils_1.69          splines_3.6.2
## [70] knitr_1.27           pillar_1.4.3         igraph_1.2.4.2
## [73] rjson_0.2.20         boot_1.3-24          corpcor_1.6.9
## [76] BDgraph_2.62         reshape2_1.4.3       stats4_3.6.2
## [79] XML_3.98-1.20        glue_1.3.1           evaluate_0.14
## [82] latticeExtra_0.6-29 data.table_1.12.8    png_0.1-7
## [85] nloptr_1.2.1         gtable_0.3.0         purrr_0.3.3
## [88] assertthat_0.2.1    ggplot2_3.2.1        xfun_0.12
## [91] openxlsx_4.1.4       xtable_1.8-4         coda_0.19-3
## [94] Rsolnp_1.16          glasso_1.11          survival_3.1-8
## [97] truncnorm_1.0-8     tibble_2.1.3         arm_1.10-1
## [100] ggm_2.3              ellipse_0.4.1        cluster_2.1.0

```