

Network Meta-Analysis

MaxScott

6 January 2018

```
require(mvmeta)

## Loading required package: mvmeta
## This is mvmeta 0.4.7. For an overview type: help('mvmeta-package').
require(metaSEM)

## Loading required package: metaSEM
## Loading required package: OpenMx

## To take full advantage of multiple cores, use:
##   mxOption(NULL, 'Number of Threads', 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".

net_dat <- read.csv("Data_for_metaSEM_forum.csv", header = TRUE)

# Within each outcome, there are values for implicit_measure that are 0 and 1
table(!is.na(net_dat[, "red_act_dir"]), net_dat[, "implicit_measure"])

##
##          0    1
##  FALSE 119 208
##  TRUE   28  84

table(!is.na(net_dat[, "red_act_ind"]), net_dat[, "implicit_measure"])

##
##          0    1
##  FALSE 118 215
##  TRUE   29  77

table(!is.na(net_dat[, "red_goal"]), net_dat[, "implicit_measure"])

##
##          0    1
##  FALSE 112 261
##  TRUE   35  31

table(!is.na(net_dat[, "affirm"]), net_dat[, "implicit_measure"])

##
##          0    1
##  FALSE 145 279
##  TRUE    2   13
```

```



```

```

# Test overall model without covariates. This works
mod <- meta(y = net_dat[, 2:12],
             v = net_dat[, cov_names],
             RE.constraints = con)
summary(mod)

##
## Call:
## meta(y = net_dat[, 2:12], v = net_dat[, cov_names], RE.constraints = con)
##
## 95% confidence intervals: z statistic approximation
## Coefficients:
##              Estimate Std.Error    lbound    ubound z value Pr(>|z|)
## Intercept1 -0.2027657 0.0433266 -0.2876842 -0.1178472 -4.6799 2.870e-06
## Intercept2 -0.3415005 0.0462986 -0.4322441 -0.2507569 -7.3760 1.630e-13
## Intercept3 -0.2960616 0.0551322 -0.4041187 -0.1880045 -5.3700 7.872e-08
## Intercept4 -0.0923674 0.1138230 -0.3154563 0.1307215 -0.8115 0.417078
## Intercept5 -0.0919509 0.1514019 -0.3886931 0.2047913 -0.6073 0.543632
## Intercept6 -0.1986121 0.1309089 -0.4551887 0.0579646 -1.5172 0.129222
## Intercept7  0.0576875 0.0582595 -0.0564991 0.1718740 0.9902 0.322086
## Intercept8  0.2725811 0.0931579  0.0899949 0.4551673 2.9260 0.003433
## Intercept9  0.1710940 0.0731922  0.0276400 0.3145480 2.3376 0.019408
## Intercept10 0.2517622 0.0472683  0.1591180 0.3444064 5.3262 1.003e-07
## Intercept11 0.1829489 0.0623312  0.0607820 0.3051157 2.9351 0.003334
## a            0.1130759 0.0144787  0.0846983 0.1414536 7.8098 5.773e-15
## b            0.0436056 0.0187699  0.0068172 0.0803939 2.3232 0.020171
##
## Intercept1 ***
## Intercept2 ***
## Intercept3 ***
## Intercept4
## Intercept5
## Intercept6
## Intercept7
## Intercept8 **
## Intercept9 *
## Intercept10 ***
## Intercept11 **
## a      ***
## b      *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Q statistic on the homogeneity of effect sizes: 1503.775
## Degrees of freedom of the Q statistic: 608
## P value of the Q statistic: 0
##
## Heterogeneity indices (based on the estimated Tau2):
##              Estimate
## Intercept1: I2 (Q statistic) 0.6896
## Intercept2: I2 (Q statistic) 0.5870
## Intercept3: I2 (Q statistic) 0.6068
## Intercept4: I2 (Q statistic) 0.4465

```

```

## Intercept5: I2 (Q statistic) 0.2023
## Intercept6: I2 (Q statistic) 0.1939
## Intercept7: I2 (Q statistic) 0.6021
## Intercept8: I2 (Q statistic) 0.7652
## Intercept9: I2 (Q statistic) 0.6643
## Intercept10: I2 (Q statistic) 0.5710
## Intercept11: I2 (Q statistic) 0.4840
##
## Number of studies (or clusters): 439
## Number of observed statistics: 619
## Number of estimated parameters: 13
## Degrees of freedom: 606
## -2 log likelihood: 1606.245
## OpenMx status1: 0 ("0" or "1": The optimization is considered fine.
## Other values may indicate problems.)

# Test model with implicit_measure as a covariate. This does not work
mod1 <- meta(y = net_dat[, 2:12],
              v = net_dat[, cov_names],
              x = net_dat[, "implicit_measure"],
              RE.constraints = con)
summary(mod1)

##
## Call:
## meta(y = net_dat[, 2:12], v = net_dat[, cov_names], x = net_dat[, "implicit_measure"], RE.constraints = con)
##
## 95% confidence intervals: z statistic approximation
## Coefficients:
##             Estimate Std.Error lbound ubound z value Pr(>|z|)
## Intercept1 -0.2109650 0.0897997 -0.3869693 -0.0349608 -2.3493 0.0188095
## Intercept2 -0.2578582 0.0788417 -0.4123851 -0.1033314 -3.2706 0.0010733
## Intercept3 -0.1577604 0.0758480 -0.3064198 -0.0091009 -2.0800 0.0375298
## Intercept4 -0.0500887 0.2965706 -0.6313563 0.5311789 -0.1689 0.8658808
## Intercept5 -0.3819812 0.2444423 -0.8610792 0.0971169 -1.5627 0.1181316
## Intercept6 -0.3676466 0.1887201 -0.7375312 0.0022380 -1.9481 0.0514024
## Intercept7 0.1991309 0.1133433 -0.0230178 0.4212796 1.7569 0.0789377
## Intercept8 0.4372882 0.1316592 0.1792410 0.6953355 3.3214 0.0008958
## Intercept9 0.1176734 0.0915059 -0.0616749 0.2970217 1.2860 0.1984553
## Intercept10 0.1346010 0.0983722 -0.0582050 0.3274069 1.3683 0.1712236
## Intercept11 0.2509409 0.1205428 0.0146813 0.4872005 2.0818 0.0373646
## Slope1_1 0.0057903 0.1017686 -0.1936725 0.2052531 0.0569 0.9546273
## Slope2_1 -0.1251728 0.0966455 -0.3145946 0.0642490 -1.2952 0.1952602
## Slope3_1 -0.2785622 0.1077855 -0.4898180 -0.0673064 -2.5844 0.0097545
## Slope4_1 -0.0755864 0.3197649 -0.7023141 0.5511413 -0.2364 0.8131370
## Slope5_1 0.5505034 0.3089480 -0.0550235 1.1560303 1.7819 0.0747713
## Slope6_1 0.4094138 0.2670013 -0.1138991 0.9327266 1.5334 0.1251828
## Slope7_1 -0.2086637 0.1318182 -0.4670227 0.0496952 -1.5830 0.1134293
## Slope8_1 -0.3201655 0.1826517 -0.6781562 0.0378252 -1.7529 0.0796236
## Slope9_1 0.1556875 0.1468842 -0.1322001 0.4435752 1.0599 0.2891745
## Slope10_1 0.1499413 0.1113483 -0.0682974 0.3681800 1.3466 0.1781102
## Slope11_1 -0.1024665 0.1402228 -0.3772980 0.1723651 -0.7307 0.4649376
## a 0.1071071 0.0138687 0.0799249 0.1342893 7.7229 1.132e-14
## b 0.0446010 0.0175896 0.0101261 0.0790760 2.5356 0.0112239

```

```

## 
## Intercept1  *
## Intercept2  **
## Intercept3  *
## Intercept4
## Intercept5
## Intercept6 .
## Intercept7 .
## Intercept8 ***
## Intercept9
## Intercept10
## Intercept11 *
## Slope1_1
## Slope2_1
## Slope3_1   **
## Slope4_1
## Slope5_1   .
## Slope6_1
## Slope7_1
## Slope8_1   .
## Slope9_1
## Slope10_1
## Slope11_1
## a          ***
## b          *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Q statistic on the homogeneity of effect sizes: 1503.775
## Degrees of freedom of the Q statistic: 608
## P value of the Q statistic: 0
##
## Explained variances (R2):
##           y1      y2      y3      y4      y5      y6
## Tau2 (no predictor) 0.17067 0.18236 0.19911 0.13282 0.12049 0.14493
## Tau2 (with predictors) NA     NA     NA     NA     NA     NA
## R2                  NA     NA     NA     NA     NA     NA
##           y7      y8      y9      y10     y11
## Tau2 (no predictor) 0.14255 0.14334 0.11750 0.18562 0.1359
## Tau2 (with predictors) NA     NA     NA     NA     NA
## R2                  NA     NA     NA     NA     NA
## 
## Number of studies (or clusters): 439
## Number of observed statistics: 619
## Number of estimated parameters: 24
## Degrees of freedom: 595
## -2 log likelihood: 1581.615
## OpenMx status1: 0 ("0" or "1": The optimization is considered fine.
## Other values may indicate problems.)
# Test model with implicit_measure as a covariate in mvmeta. This works
mod2 <- mvmeta(cbind(red_act_dir, red_act_ind, red_goal, affirm, pos_mood,
                     neg_mood, threat, deplete, inc_goal, inc_act_dir, inc_act_ind) ~ implicit_measure,
                S = as.matrix(net_dat[, cov_names]),
```

```

bscov = "cs",
control = list(inputna = TRUE),
data = net_dat)
summary(mod2)

## Call: mvmeta(formula = cbind(red_act_dir, red_act_ind, red_goal, affirm,
##      pos_mood, neg_mood, threat, deplete, inc_goal, inc_act_dir,
##      inc_act_ind) ~ implicit_measure, S = as.matrix(net_dat[,,
##      cov_names]), data = net_dat, bscov = "cs", control = list(inputna = TRUE))
##
## Multivariate random-effects meta-regression
## Dimension: 11
## Estimation method: REML
##
## Fixed-effects coefficients
## red_act_dir :
##              Estimate Std. Error      z Pr(>|z|) 95%ci.lb
## (Intercept) -0.2133   0.0922 -2.3128  0.0207 -0.3940
## implicit_measure 0.0077   0.1046  0.0732  0.9416 -0.1973
## 95%ci.ub
## (Intercept) -0.0325 *
## implicit_measure 0.2126
## red_act_ind :
##              Estimate Std. Error      z Pr(>|z|) 95%ci.lb
## (Intercept) -0.2577   0.0809 -3.1854  0.0014 -0.4163
## implicit_measure -0.1274   0.0991 -1.2858  0.1985 -0.3216
## 95%ci.ub
## (Intercept) -0.0991 **
## implicit_measure 0.0668
## red_goal :
##              Estimate Std. Error      z Pr(>|z|) 95%ci.lb
## (Intercept) -0.1580   0.0777 -2.0338  0.0420 -0.3103
## implicit_measure -0.2805   0.1106 -2.5366  0.0112 -0.4972
## 95%ci.ub
## (Intercept) -0.0057 *
## implicit_measure -0.0638 *
## affirm :
##              Estimate Std. Error      z Pr(>|z|) 95%ci.lb
## (Intercept) -0.0486   0.3052 -0.1592  0.8735 -0.6469
## implicit_measure -0.0770   0.3295 -0.2336  0.8153 -0.7227
## 95%ci.ub
## (Intercept) 0.5497
## implicit_measure 0.5688
## pos_mood :
##              Estimate Std. Error      z Pr(>|z|) 95%ci.lb
## (Intercept) -0.3870   0.2519 -1.5365  0.1244 -0.8807
## implicit_measure 0.5556   0.3176  1.7493  0.0802 -0.0669
## 95%ci.ub
## (Intercept) 0.1067
## implicit_measure 1.1780 .
## neg_mood :
##              Estimate Std. Error      z Pr(>|z|) 95%ci.lb
## (Intercept) -0.3676   0.1945 -1.8896  0.0588 -0.7488
## implicit_measure 0.4079   0.2742  1.4876  0.1368 -0.1295

```

```

##          95%ci.ub
## (Intercept) 0.0137 .
## implicit_measure 0.9453
## threat :
##             Estimate Std. Error      z Pr(>|z|) 95%ci.lb
## (Intercept) 0.1989  0.1164  1.7087   0.0875 -0.0292
## implicit_measure -0.2089  0.1351 -1.5464   0.1220 -0.4738
##          95%ci.ub
## (Intercept) 0.4270 .
## implicit_measure 0.0559
## deplete :
##             Estimate Std. Error      z Pr(>|z|) 95%ci.lb
## (Intercept) 0.4405  0.1346  3.2720   0.0011 0.1766
## implicit_measure -0.3221  0.1872 -1.7206   0.0853 -0.6890
##          95%ci.ub
## (Intercept) 0.7043 **
## implicit_measure 0.0448 .
## inc_goal :
##             Estimate Std. Error      z Pr(>|z|) 95%ci.lb
## (Intercept) 0.1194  0.0940  1.2704   0.2039 -0.0648
## implicit_measure 0.1563  0.1511  1.0346   0.3009 -0.1398
##          95%ci.ub
## (Intercept) 0.3037
## implicit_measure 0.4525
## inc_act_dir :
##             Estimate Std. Error      z Pr(>|z|) 95%ci.lb
## (Intercept) 0.1340  0.1010  1.3262   0.1848 -0.0640
## implicit_measure 0.1519  0.1144  1.3281   0.1842 -0.0723
##          95%ci.ub
## (Intercept) 0.3320
## implicit_measure 0.3762
## inc_act_ind :
##             Estimate Std. Error      z Pr(>|z|) 95%ci.lb
## (Intercept) 0.2511  0.1239  2.0270   0.0427 0.0083
## implicit_measure -0.1036  0.1441 -0.7193   0.4720 -0.3860
##          95%ci.ub
## (Intercept) 0.4940 *
## implicit_measure 0.1787
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Between-study random-effects (co)variance components
## Structure: Compound symmetry
##             Std. Dev       Corr
## red_act_dir 0.3401 red_act_dir red_act_ind red_goal affirm
## red_act_ind 0.3401        0.4016
## red_goal    0.3401 0.4016        0.4016
## affirm      0.3401 0.4016 0.4016        0.4016
## pos_mood    0.3401 0.4016 0.4016 0.4016        0.4016
## neg_mood    0.3401 0.4016 0.4016 0.4016 0.4016
## threat      0.3401 0.4016 0.4016 0.4016 0.4016
## deplete     0.3401 0.4016 0.4016 0.4016 0.4016
## inc_goal    0.3401 0.4016 0.4016 0.4016 0.4016
## inc_act_dir 0.3401 0.4016 0.4016 0.4016 0.4016

```

```

## inc_act_ind    0.3401      0.4016      0.4016    0.4016  0.4016
##
## red_act_dir  pos_mood  neg_mood  threat  deplete  inc_goal  inc_act_dir
## red_act_ind
## red_goal
## affirm
## pos_mood
## neg_mood    0.4016
## threat      0.4016    0.4016
## deplete     0.4016    0.4016  0.4016
## inc_goal    0.4016    0.4016  0.4016    0.4016
## inc_act_dir 0.4016    0.4016  0.4016    0.4016    0.4016
## inc_act_ind  0.4016    0.4016  0.4016    0.4016    0.4016    0.4016
##
## Multivariate Cochran Q-test for residual heterogeneity:
## Q = 1440.5680 (df = 597), p-value = 0.0000
## I-square statistic = 58.6%
##
## 439 studies, 619 observations, 22 fixed and 2 random-effects parameters
##   logLik          AIC          BIC
## -34749.9483    69547.8966   69653.3026

sessionInfo()

## R version 3.4.3 (2017-11-30)
## Platform: x86_64-pc-linux-gnu (64-bit)
## Running under: Linux Mint 18.2
##
## Matrix products: default
## BLAS: /usr/lib/openblas-base/libblas.so.3
## LAPACK: /usr/lib/libopenblas-r0.2.18.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_0.9.16 OpenMx_2.8.3   mvmeta_0.4.7
##
## loaded via a namespace (and not attached):
## [1] Rcpp_0.12.14    mvtnorm_1.0-6   lattice_0.20-35 digest_0.6.12
## [5] rprojroot_1.3-2 MASS_7.3-48    grid_3.4.3    backports_1.1.2
## [9] magrittr_1.5    ellipse_0.4.1   evaluate_0.10.1 stringi_1.1.6
## [13] Matrix_1.2-12   rmarkdown_1.8   tools_3.4.3   stringr_1.2.0
## [17] yaml_2.1.16    parallel_3.4.3 compiler_3.4.3 htmltools_0.3.6
## [21] knitr_1.18

```