

# Regression analysis with meta-analytic data

Juan Francisco Martin Rodriguez (modified by Mike Cheung)

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Regression analysis: <https://openmx.ssri.psu.edu/node/4801>

```
library(metaSEM)
library(readr)

regression_model <- read_delim("Regression.csv",
                              ";", escape_double = FALSE,
                              col_types = cols(Type_training =
                                                col_factor(levels = c("SP", "MP", "IM"))),
                              na = "NA", trim_ws = TRUE)

# View(regression_model)

## Regression between both ES: MEP --> strength
# 1. remove the missing values (these models cannot work with MV) and exclude
# the 3 first columns
RM_new <- regression_model[-8,4:9]

## A: asymmetric paths for regression coefficients
## and factor loadings
A <- matrix(c(0, "0.1*beta1_2", 0, 0,
              0, 0, 0, 0,
              1, 0, 0, 0,
              0, 1, 0, 0),
            ncol=4, nrow=4, byrow=TRUE)

dimnames(A) <- list(c("f_strength","f_MEP",
                    "y_strength","y_MEP"),
                  c("f_strength","f_MEP",
                    "y_strength","y_MEP"))

A
```

```
##           f_strength f_MEP           y_strength y_MEP
## f_strength "0"       "0.1*beta1_2" "0"       "0"
## f_MEP      "0"       "0"         "0"       "0"
## y_strength "1"       "0"         "0"       "0"
## y_MEP      "0"       "1"         "0"       "0"
```

```
## Convert it into OpenMx matrix
A <- as.mxMatrix(A)
```

```

# S: symmetric covariances and variances
S <- mxMatrix(type="Symm", nrow=4, ncol=4, byrow=TRUE,
              free=c(TRUE,
                     FALSE,TRUE,
                     FALSE,FALSE,FALSE,
                     FALSE,FALSE,FALSE,FALSE),
              values=c(0.1,
                       0,0.1,
                       0,0,0,
                       0,0,0,0),
              labels=c("tau2_1_1",
                       NA,"tau2_2_2",
                       NA,NA,"data.v_strength",
                       NA,NA,"data.Cov_strength_MEP","data.v_MEP"),
              name = "S")

## F: select observed variables
F <- matrix(c(0, 0, 1, 0,
             0, 0, 0, 1), nrow = 2, ncol = 4, byrow = TRUE)
dimnames(F) <- list(c("y_strength","y_MEP"),
                   c("f_strength","f_MEP","y_strength",
                     "y_MEP"))
F

```

```

##           f_strength f_MEP y_strength y_MEP
## y_strength      0     0           1     0
## y_MEP           0     0           0     1

```

```

F <-as.mxMatrix(F)

## M: intercepts or means (only intercepts of the latent are estimated
## the intercepts of the observed variables are set to 0)
M <- matrix(c("0*beta1_0","0*beta2_0",0,0), nrow=1, ncol=4, byrow = TRUE)
dimnames(M)[[2]] <- c("f_strength","f_MEP",
                    "y_strength","y_MEP")
M

```

```

##           f_strength f_MEP y_strength y_MEP
## [1,] "0*beta1_0" "0*beta2_0" "0"           "0"

```

```

M <- as.mxMatrix(M)

## Formula for R2
R2 <- mxAlgebra(beta1_2^2*tau2_2_2/(beta1_2^2*tau2_2_2 + tau2_1_1),
                name="R2")

## Build the model
reg <- mxModel("Regression",
              mxData(observed=as.data.frame(RM_new), type="raw"),
              A, S, F, M, R2, mxCI("R2"),
              mxExpectationRAM(A="A", S="S",
                               F="F", M="M"),

```

```

                                dimnames = c("f_strength", "f_MEP",
                                                "y_strength", "y_MEP")),
                                mxFitFunctionML())

## Run the analysis
reg.fit <- mxRun(reg, intervals=TRUE, silent=TRUE)
summary(reg.fit)

```

```

## Summary of Regression
##
## free parameters:
##      name matrix      row      col Estimate Std.Error A
## 1 beta1_2      A f_strength  f_MEP 0.5143001 0.1166106
## 2 tau2_1_1     S      1      1 0.1432816 0.1115068
## 3 tau2_2_2     S      2      2 1.7727245 0.6569429
## 4 beta1_0      M      1 f_strength 0.6069641 0.1596404
## 5 beta2_0      M      1      f_MEP 1.0435207 0.3048190
##
## confidence intervals:
##                lbound estimate ubound note
## Regression.R2[1,1] 0.3186566 0.7659468 0.9862779
##
## Model Statistics:
##      | Parameters | Degrees of Freedom | Fit (-2lnL units)
##      Model:      5              39              123.2064
##      Saturated:  5              39              NA
##      Independence: 4              40              NA
## Number of observations/statistics: 22/44
##
## Information Criteria:
##      | df Penalty | Parameters Penalty | Sample-Size Adjusted
## AIC:  45.206414      133.2064      136.9564
## BIC:  2.655758      138.6616      123.2064
## CFI: NA
## TLI: 1 (also known as NNFI)
## RMSEA: 0 [95% CI (NA, NA)]
## Prob(RMSEA <= 0.05): NA
## To get additional fit indices, see help(mxRefModels)
## timestamp: 2022-04-01 11:38:05
## Wall clock time: 0.136869 secs
## optimizer: SLSQP
## OpenMx version number: 2.20.6
## Need help? See help(mxSummary)

```

```
reg.fit@output$status[[1]]
```

```
## [1] 0
```

An alternative approach with the developmental version at <https://github.com/mikewlcheung/metasem>

```
model <- "## Latent effect sizes
  f_strength =~ 1*y_strength
  f_MEP =~ 1*y_MEP

  ## Regression coefficient
  f_strength ~ beta1_2*f_MEP

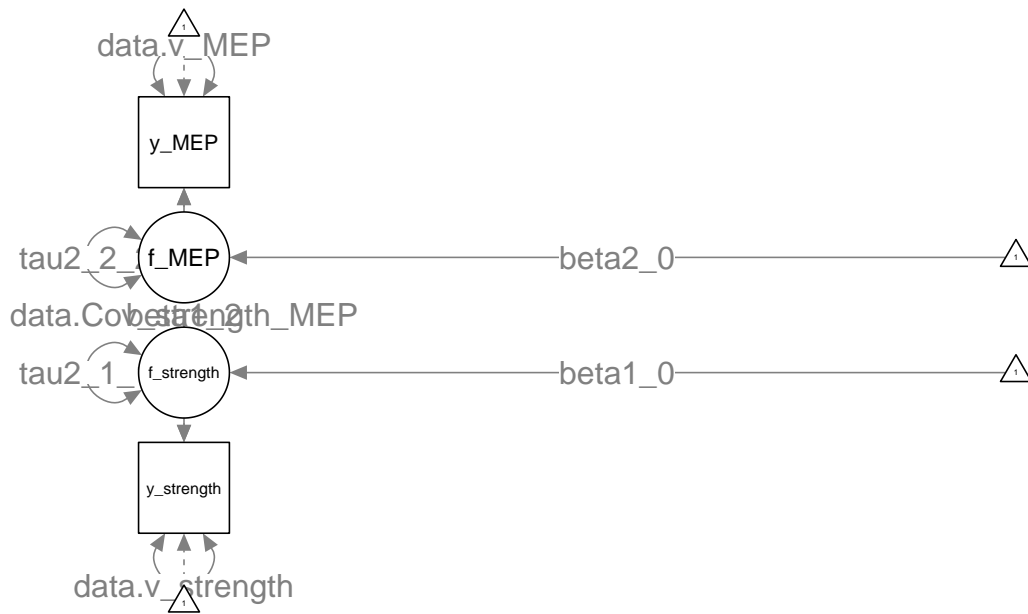
  ## Mean and intercept
  f_strength ~ beta1_0*1
  f_MEP ~ beta2_0*1

  ## Latent variances
  f_strength ~~ tau2_1_1*f_strength
  f_MEP ~~ tau2_2_2*f_MEP

  ## Known sampling variances and covariance
  y_strength ~~ data.v_strength*y_strength
  y_MEP ~~ data.v_MEP*y_MEP
  y_strength ~~ data.Cov_strength_MEP*y_MEP

  ## Define an R2
  R2 := beta1_2^2*tau2_2_2/(beta1_2^2*tau2_2_2 + tau2_1_1)
"

plot(model, layout="tree")
```



```
RAM <- lavaan2RAM(model, obs.variables = c("y_strength", "y_MEP"), std.lv=FALSE)
mx.fit <- create.mxModel("Regressoin", RAM=RAM, data=as.data.frame(RM_new),
                          intervals.type="LB")
summary(mx.fit)
```

```
## Summary of Regressoin
##
## free parameters:
##      name matrix      row      col Estimate Std.Error A
## 1 beta1_2 Amatrix f_strength      f_MEP 0.5143001 0.1166106
## 2 tau2_1_1 Smatrix f_strength f_strength 0.1432816 0.1115068
## 3 tau2_2_2 Smatrix      f_MEP      f_MEP 1.7727243 0.6569429
## 4 beta1_0 Mmatrix          1 f_strength 0.6069641 0.1596404
## 5 beta2_0 Mmatrix          1      f_MEP 1.0435208 0.3048189
##
## confidence intervals:
##              lbound estimate  ubound note
## beta1_2          0.289792684 0.5143001 0.7709697
## tau2_1_1          0.007250631 0.1432816 0.5182668
## tau2_2_2          0.872768675 1.7727243 3.8024010
## beta1_0          0.284552909 0.6069641 0.9506220
## beta2_0          0.431701676 1.0435208 1.6889816
## Regressoin.R2[1,1] 0.318656620 0.7659468 0.9862779
##
```

```

## Model Statistics:
##           | Parameters | Degrees of Freedom | Fit (-2lnL units)
##   Model:   |         5         |          39         |          123.2064
##   Saturated: |         5         |          39         |           NA
## Independence: |         4         |          40         |           NA
## Number of observations/statistics: 22/44
##
## Information Criteria:
##           | df Penalty | Parameters Penalty | Sample-Size Adjusted
##   AIC:    | 45.206414  |          133.2064  |          136.9564
##   BIC:    |  2.655758  |          138.6616  |          123.2064
##   CFI: NA
##   TLI: 1   (also known as NNFI)
##   RMSEA: 0 [95% CI (NA, NA)]
##   Prob(RMSEA <= 0.05): NA
## To get additional fit indices, see help(mxRefModels)
## timestamp: 2022-04-01 11:38:07
## Wall clock time: 0.1395154 secs
## optimizer: SLSQP
## OpenMx version number: 2.20.6
## Need help? See help(mxSummary)

```

```
sessionInfo()
```

```

## R version 4.1.3 (2022-03-10)
## Platform: x86_64-pc-linux-gnu (64-bit)
## Running under: Ubuntu 20.04.4 LTS
##
## Matrix products: default
## BLAS: /usr/lib/x86_64-linux-gnu/blas/libblas.so.3.9.0
## LAPACK: /usr/lib/x86_64-linux-gnu/lapack/liblapack.so.3.9.0
##
## locale:
##  [1] LC_CTYPE=C.UTF-8          LC_NUMERIC=C             LC_TIME=C.UTF-8
##  [4] LC_COLLATE=C.UTF-8       LC_MONETARY=C.UTF-8     LC_MESSAGES=C.UTF-8
##  [7] LC_PAPER=C.UTF-8        LC_NAME=C               LC_ADDRESS=C
## [10] LC_TELEPHONE=C          LC_MEASUREMENT=C.UTF-8 LC_IDENTIFICATION=C
##
## attached base packages:
## [1] stats      graphics  grDevices  utils      datasets  methods    base
##
## other attached packages:
## [1] readr_2.1.2      metaSEM_1.2.5.9 OpenMx_2.20.6
##
## loaded via a namespace (and not attached):
##  [1] minqa_1.2.4      colorspace_2.0-2      ellipsis_0.3.2
##  [4] htmlTable_2.4.0  corpcor_1.6.10        base64enc_0.1-3
##  [7] rstudioapi_0.13  lavaan_0.6-11         bit64_4.0.5
## [10] fansi_1.0.2      mvtnorm_1.1-3         splines_4.1.3
## [13] mnormt_2.0.2     knitr_1.37            glasso_1.11
## [16] Formula_1.2-4    nloptr_1.2.2.3        cluster_2.1.2
## [19] png_0.1-7        compiler_4.1.3        backports_1.4.1
## [22] assertthat_0.2.1 Matrix_1.4-0          fastmap_1.1.0
## [25] cli_3.1.1        htmltools_0.5.2       tools_4.1.3

```

```

## [28] igraph_1.2.11      coda_0.19-4      gtable_0.3.0
## [31] glue_1.6.2         reshape2_1.4.4   dplyr_1.0.8
## [34] Rcpp_1.0.8.3       carData_3.0-5    vctrs_0.3.8
## [37] nlme_3.1-155       lisrelToR_0.1.4  psych_2.2.3
## [40] xfun_0.30          stringr_1.4.0    openxlsx_4.2.5
## [43] lme4_1.1-28        lifecycle_1.0.1  gtools_3.9.2
## [46] XML_3.99-0.8       MASS_7.3-56      scales_1.1.1
## [49] vroom_1.5.7        hms_1.1.1        kutils_1.70
## [52] parallel_4.1.3     RColorBrewer_1.1-2 yaml_2.3.5
## [55] pbapply_1.5-0      gridExtra_2.3    ggplot2_3.3.5
## [58] rpart_4.1.16       latticeExtra_0.6-29 stringi_1.7.6
## [61] sem_3.1-14         checkmate_2.0.0  boot_1.3-28
## [64] zip_2.2.0          rlang_1.0.2      pkgconfig_2.0.3
## [67] arm_1.12-2         evaluate_0.15    lattice_0.20-45
## [70] purrr_0.3.4        htmlwidgets_1.5.4 bit_4.0.4
## [73] tidysselect_1.1.2  plyr_1.8.6       magrittr_2.0.2
## [76] R6_2.5.1           generics_0.1.2   Hmisc_4.6-0
## [79] DBI_1.1.2          pillar_1.7.0     foreign_0.8-82
## [82] rockchalk_1.8.151  survival_3.2-13  semPlot_1.1.5
## [85] abind_1.4-5        nnet_7.3-17      tibble_3.1.6
## [88] crayon_1.4.2       fdrtool_1.2.17  utf8_1.2.2
## [91] ellipse_0.4.2      tmvnsim_1.0-2    tzdb_0.2.0
## [94] rmarkdown_2.13     jpeg_0.1-9       grid_4.1.3
## [97] qgraph_1.9.2       data.table_1.14.2 pbivnorm_0.6.0
## [100] digest_0.6.29      xtable_1.8-4     mi_1.0
## [103] RcppParallel_5.1.5 stats4_4.1.3     munsell_0.5.0

```