

Let R Make Your Life Easier:

An Introduction to the MplusAutomation Package

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February 1, 2017

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Talk Objectives

1. Run a single model from R and extract parameter estimates.
2. Generate unique Mplus input files for a simulation.
3. Estimate simulation files.
4. Extract simulation results.
5. Grab bag of tricks

Load MplusAutomation Library

```
library(MplusAutomation)
```

View Available Functions

```
lsp(MplusAutomation)[1:15]
```

```
## [1] "cd"                "compareModels"
## [3] "createModels"      "createSyntax"
## [5] "extract"           "extract.mplus.model"
```

```
## [7] "extract.mplusObject"      "extractModelParameters"
## [9] "extractModelSummaries"    "extractModIndices"
## [11] "getSavedata_Bparams"      "getSavedata_Data"
## [13] "getSavedata_Fileinfo"     "HTMLSummaryTable"
## [15] "LatexSummaryTable"
```

View Available Functions Cont'd

```
lsp(MplusAutomation)[16:30]
```

```
## [1] "lookupTech1Parameter"      "mplus.traceplot"
## [3] "mplusModeler"              "mplusObject"
## [5] "mplusRcov"                  "paramExtract"
## [7] "parseMplus"                 "prepareMplusData"
## [9] "readModels"                 "runModels"
## [11] "runModels_Interactive"      "showSummaryTable"
## [13] "SummaryTable"               "testBParamCompoundConstraint"
## [15] "testBParamConstraint"
```

Single Model

Step No. 1

- Read in data

```
simDat <- read.csv("templateDat.csv", header = TRUE)
```

Variable Names

- Outcome Variable and Level Identifiers

```
colnames(simDat)[1:3]
```

```
## [1] "Y"      "SSID"   "SCHID"
```

Variable Names Cont'd

- Level One Predictors

```
colnames(simDat)[4:8]
```

```
## [1] "X1" "X2" "X3" "X4" "X5"
```

Number of Units Per Level

```
# Level 2
```

```
length(unique(simDat$SCHID))
```

```
## [1] 200
```

```
# Level 1
```

```
length(unique(simDat$SSID))
```

```
## [1] 10000
```

Goal

- Estimate an organizational MLM including random effects for:
 - the intercept: γ_{00}
 - all within level slopes: $\gamma_{10} : \gamma_{50}$

Mplus Syntax – TITLE/VARIABLE

```
twoLevel <- mplusObject(  
  TITLE = "Template;",  
  VARIABLE = "  
  NAMES = Y SSID SCHID X1 X2 X3 X4 X5;  
  USEVARIABLES = Y X1 X2 X3 X4 X5;  
  CLUSTER = SCHID;  
  WITHIN = X1 X2 X3 X4 X5;",  
  ## ...
```

Mplus Syntax – ANALYSIS

```
ANALYSIS = "TYPE = TWOLEVEL RANDOM;",
```

Mplus Syntax – MODEL

```
MODEL = "  
%WITHIN%  
S1 | Y ON X1;  
S2 | Y ON X2;  
S3 | Y ON X3;  
S4 | Y ON X4;  
S5 | Y ON X5;  
%BETWEEN%  
Y WITH S1 S2 S3 S4 S5;  
S1 WITH S2 S3 S4 S5;  
S2 WITH S3 S4 S5;  
S3 WITH S4 S5;  
S4 WITH S5;  
",
```

Mplus Syntax – OUTPUT

```
OUTPUT = "TECH1;")
```

Prepare Data File

```
prepareMplusData(simDat,  
                 "template.dat",  
                 infile = TRUE, # eq "template.inp"  
                 overwrite = TRUE,  
                 interactive = FALSE)
```

Estimate Model From R via runModels()

```
runModels(recursive = FALSE,  
          filefilter = "knitr*",  
          replaceOutfile = c("always",  
                             "modifiedDate",  
                             "never"))
```

```
runModels(filefilter = "knitr*")
```

```
##
```

```
## Running model: knitr_template.inp
```

```
## System command: C:\windows\system32\cmd.exe /c cd "C:\Users\grr13002\Dropbox\Conn\Consults\McCoach" &
```

Extract Output

```
singRun <- readModels()
```

```
## Reading model: C:/Users/grr13002/Dropbox/Conn/Consults/McCoach/knitr_template.out
```

```
## Reading model: C:/Users/grr13002/Dropbox/Conn/Consults/McCoach/template.out
```

```
names(singRun)
```

```
## [1] "knitr_template.out" "template.out"
```

Determine Slots Available For First Element

```
names(singRun[[1]])
```

```
## [1] "input"          "warnings"        "errors"  
## [4] "summaries"      "parameters"      "class_counts"  
## [7] "residuals"      "tech1"           "tech3"  
## [10] "tech4"          "tech7"           "tech9"  
## [13] "tech12"         "fac_score_stats" "gh5"
```

\$errors & \$warnings

- Good, no estimation warnings or errors!

```
singRun[[1]]$errors
```

```
## list()
## attr("class")
## [1] "list"          "mplus.errors"
```

```
singRun[[1]]$warnings
```

```
## list()
## attr("class")
## [1] "list"          "mplus.warnings"
```

\$summaries

```
names(singRun[[1]]$summaries)
```

```
## [1] "Mplus.version"      "Title"          "AnalysisType"
## [4] "DataType"          "Estimator"      "Observations"
## [7] "Parameters"        "LL"             "LLCorrectionFactor"
## [10] "AIC"               "BIC"            "aBIC"
## [13] "AICC"              "Filename"
```

\$parameters

```
names(singRun[[1]]$parameters)
```

```
## [1] "unstandardized"
```

Extract Parameter Estimates

```
paramEst <- singRun[[1]]$parameters$unstandardized
colnames(paramEst)
```



```
## [1] "paramHeader"  "param"          "est"            "se"
## [5] "est_se"        "pval"           "BetweenWithin"
```

Store Fixed and Random Components

```
gammaEst <- paramEst[which(paramEst$paramHeader
                           == "Means"),]
tauEst <- paramEst[which(paramEst$paramHeader
                         == "Variances"),]
sigmaEst <- paramEst[which(paramEst$paramHeader
                           == "Residual.Variances"),]
```

Combine into single data.frame

```
singleRunEST <- rbind(gammaEst,tauEst,sigmaEst)
singleRunEST[seq(1,12,2),]
```

```
##   paramHeader param  est    se est_se  pval BetweenWithin
## 17      Means     Y 0.038 0.020  1.939 0.052      Between
## 19      Means    S2 0.013 0.014  0.870 0.384      Between
## 21      Means    S4 0.180 0.014 12.965 0.000      Between
## 23  Variances     Y 0.077 0.007 10.573 0.000      Between
## 25  Variances    S2 0.041 0.004 10.105 0.000      Between
## 27  Variances    S4 0.037 0.003 10.802 0.000      Between
```

Check Summary Statistics

```
singRun[[1]]$summaries$Parameters
```

```
## [1] 28
```

```
singRun[[1]]$summaries$LL
```

```
## [1] -1090.369
```

Fixed Conditions	
Level Two Units:	50
Level One Units:	200
Total Sample Size:	10000
Fixed Components:	Γ
Level Two Random Components:	$\tau_{00,11:33} = 0.801, 0.04$
Level One Random Component:	$\sigma^2 = 0.046$
Estimator:	Full Maximum Likelihood
Varying Conditions	
X4 Slope Variance (τ_{44}):	0.04, 0.02, 0.01, 0.001, 0.000
X4 Slope Variance (τ_{55}):	0.04, 0.02, 0.01, 0.001, 0.000

Note: Random simulation facets are fully crossed, resulting in 15 conditions

```
singRun[[1]]$summaries$AIC
```

```
## [1] 2236.738
```

Simulation

Sim Conditions

Sim Recap

- Conditions
 - 15, condition specific directories
- Data file structure:
 - `dat_mplus_cond_nCond_nRep.dat`
- Model remains the same across all conditions

Steps To Run Simulation

1. Alter single run input file
2. Use R to generate all input files via `createModels()`
3. Estimate all models via `runModels()`
4. Extract all model information via `readModels()`

Step 1: init

- The **init** section goes on top of your standard mplus input file.

```
[[init]]
iterators = conds rep;
conds = 1:15;
rep = 1:10;
filename = "cond_[conds]_[rep].inp";
outputDirectory =
"C:/Users/grr13002/Dropbox/Conn/Consults...
/McCoach/Simulation/c[[conds]]";
[/init]]
```

Create Input Files

```
createModels("knitr_template.txt")
```

Run Models

- From the top level directory, we can set the recursive logical to TRUE
- The argument, “*modifiedDate*”, tells MplusAutomation to estimate models for which the modified date of the input file is more recent than its respective output file.

```
runModels(recursive = TRUE,
          replaceOutfile = "modifiedDate")
```

Read in Models

```
mplus.Extract <- readModels(recursive = TRUE)
```

- Determine all output files have been read in:
 - $150 = (10 \text{ rep} * 15 \text{ conds})$

```
length(mplus.Extract)
```

```
## [1] 150
```

Investigate list object

- Notice the same slot names are available as before

```
names(mplus.Extract[[1]])
```

```
## [1] "input"          "warnings"       "errors"
## [4] "summaries"      "parameters"     "class_counts"
## [7] "residuals"      "tech1"          "tech3"
## [10] "tech4"          "tech7"          "tech9"
## [13] "tech12"         "fac_score_stats" "gh5"
```

What information to store

1. All fixed components:
2. All random components:
3. Model Summaries:
 - Number of Free Parameters
 - Loglikelihood
 - Akaike's information criterion
 - Bayesian information criterion
4. Data set identifier
 - Rep & Condition Number

Tip

- Extract the needed information for a single replication
 - Once you do, its easy to do so for all 150 models

Model 2: Fixed and Random Components

- Where is the Standard Error Estimate?!

```
mpplus.Extract[[2]]$parameters$unstandardized[1:6,]
```

```
##           paramHeader param      est BetweenWithin
## 1 Residual.Variances      Y  0.045           Within
## 2                Y.WITH   S1  0.012           Between
## 3                Y.WITH   S2  0.010           Between
## 4                Y.WITH   S3  0.010           Between
## 5                Y.WITH   S4  0.000           Between
## 6                Y.WITH   S5 -0.001           Between
```

- We should check the errors...

Consult Errors

- How many errors?

```
length(mpplus.Extract[[2]]$errors)
```

```
## [1] 3
```

Error No. 1

```
mpplus.Extract[[2]]$errors[[1]]
```

```
## [1] "THE MODEL ESTIMATION DID NOT TERMINATE NORMALLY DUE TO AN ILL-CONDITIONED"
```

```
## [2] "FISHER INFORMATION MATRIX.  CHANGE YOUR MODEL AND/OR STARTING VALUES."
```

Error No. 2

```
mpplus.Extract[[2]]$errors[[2]]
```

```
## [1] "THE MODEL ESTIMATION DID NOT TERMINATE NORMALLY DUE TO A NON-POSITIVE"
```

```
## [2] "DEFINITE FISHER INFORMATION MATRIX.  THIS MAY BE DUE TO THE STARTING VALUES"
```

```
## [3] "BUT MAY ALSO BE AN INDICATION OF MODEL NONIDENTIFICATION. THE CONDITION"
## [4] "NUMBER IS          0.541D-10."
```

Error No. 3

```
mplus.Extract[[2]]$errors[[3]]
```

```
## [1] "THE STANDARD ERRORS OF THE MODEL PARAMETER ESTIMATES COULD NOT BE"
## [2] "COMPUTED. THIS IS OFTEN DUE TO THE STARTING VALUES BUT MAY ALSO BE"
## [3] "AN INDICATION OF MODEL NONIDENTIFICATION. CHANGE YOUR MODEL AND/OR"
## [4] "STARTING VALUES. PROBLEM INVOLVING THE FOLLOWING PARAMETER:"
## [5] "Parameter 3, %BETWEEN%: [ S2 ]"
```

Where is the Standard Error Estimate?!

- This model did not converge, therefore, no standard errors are printed.
- This will cause problems when extracting estimates across all 150 models.
- Figure out the number of columns to make things easier later:

```
ncol(mplus.Extract[[2]]$parameters$unstandardized)
```

```
## [1] 4
```

Extract Parameter Name, Estimate, & Standard error

```
r2c1.pe <-unlist(mplus.Extract[[2]]
$parameters
$unstandardized[c(1,17:28),3])

r2c1.se <- ifelse(
ncol(mplus.Extract[[2]]$parameters$unstandardized) == 4,
NA,
unlist(mplus.Extract[[2]]
$parameters
```

```
$unstandardized[c(1,17:28),4])
)
```

Check Parameter Est. & SE Values

```
comb.PeSe <- rbind(r2c1.pe,r2c1.se)
comb.PeSe[,1:4]
```

```
##           [,1] [,2] [,3] [,4]
## r2c1.pe 0.045 0.037 -0.041 0.001
## r2c1.se    NA    NA     NA    NA
```

Extract Parameter Names: The Problem

```
tail(mplus.Extract[[2]]$parameters$unstandardized)
```

```
##      paramHeader param      est BetweenWithin
## 23  Variances      Y 0.064      Between
## 24  Variances      S1 0.035      Between
## 25  Variances      S2 0.044      Between
## 26  Variances      S3 0.039      Between
## 27  Variances      S4 0.000      Between
## 28  Variances      S5 0.000      Between
```

The Fix

```
paramID.1 <- unlist(mplus.Extract[[2]]
$parameters
$unstandardized[c(1,17:28),1])
  paramID.2 <- unlist(mplus.Extract[[2]]
$parameters
$unstandardized[c(1,17:28),2])
paramName <- paste0(paramID.1,".",paramID.2)
```

Check Parameter Names from *paste0()*

```
sample(paramName,3)

## [1] "Variances.Y" "Variances.S2" "Means.S3"
```

Incorporate Parameter Name and Estimates

```
colnames(comb.PeSe) <- paramName
comb.PeSe[,1:4]

##           Residual.Variances.Y Means.Y Means.S1 Means.S2
## r2c1.pe           0.045    0.037    -0.041    0.001
## r2c1.se           NA      NA      NA      NA
```

Extract Model Summary

- Recall that this model did not converge, therefore, there will be fewer values available from `$summaries`
 - Determine number of values

```
length(mplus.Extract[[2]]$summaries)

## [1] 7

length(mplus.Extract[[3]]$summaries) # Converged

## [1] 14
```

No. Params & LL

```
r2c1.nParm <- ifelse(
length(mplus.Extract[[2]]$summaries) < 14,
NA,
unlist(mplus.Extract[[2]]
$summaries
$Parameters)
```



```
)

r2c1.ll <- ifelse(
  length(mplus.Extract[[i]]$summaries) < 14,
  NA,
  unlist(mplus.Extract[[i]]
    $summaries
    $LL)
)
```

AIC & BIC

```
r2c1.aic <- ifelse(
  length(mplus.Extract[[2]]$summaries) < 14,
  NA,
  unlist(mplus.Extract[[2]]
    $summaries
    $AIC)
)

r2c1.bic <- ifelse(
  length(mplus.Extract[[2]]$summaries) < 14,
  NA,
  unlist(mplus.Extract[[2]]
    $summaries
    $BIC)
)
```

Extract Rep and Condition IDs

```
repID <- unlist(mplus.Extract[[2]]$summaries$Filename)
```

- Notice that this is a character value, but we need the numeric value...

```
repID
```

```
## [1] "cond_1_10.out"
```

- Notice how the condition number is between __ & __?

strsplit()

```
splt <- strsplit(repID,c("_",".out"))
```

```
length(splt[[1]])
```

```
## [1] 3
```

```
splt
```

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

```
## [1] "cond"    "1"        "10.out"
```

```
print(idCond <- as.numeric(splt[[1]][2]))
```

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

Pull Numeric Values

```
splt[[1]][3]
```

```
## [1] "10.out"
```

```
print(idRep <- as.numeric(
  strsplit(splt[[1]][3],".out")[1]))
```

```
## [1] 10
```

Compile All Results

- Must create empty matrices to store all information:
1. Parameter estimates - 13 cols
 2. Standard Errors - 13 cols

3. Rep & Cond IDs - 2 cols
 4. Model Information - 4 cols
- Each matrix needs 150 rows, corresponding to each simulation run

Matrix Storage

```
paramE <- matrix(NA,ncol = 13, nrow = 150)
seE <- matrix(NA,ncol = 13, nrow = 150)
simR.C <- matrix(NA,ncol = 2, nrow = 150)
sumStat <- matrix(NA,ncol = 4,nrow = 150)
```

Utilize an incremental loop

- Because we want to automate this process, we must alter our code minimally:

```
for (i in 1:length(mplus.Extract)){
  paramE[i, ] <- unlist(mplus.Extract[[i]]
  $parameters
  $unstandardized[c(1,17:28),3])
  ...
}
```

Let's Run Through It

- You should have all the necessary files.

Grab Bag

Error Variances

- Given a set of manifest variables, R can generate the appropriate Mplus syntax to model:
 - Homogeneous

- Heterogeneous
- Compound Symmetry
- Toeplitz
- Autoregressive
- Unstructured

```
timeV
```

```
## [1] "t1" "t2" "t3" "t4" "t5"
```

MplusAutomation Implementation

- Homogeneous

```
mplusRcov(timeV,"homogenous",collapse = TRUE)
```

```
## t1 t2 t3 t4 t5 (e);
## t1 WITH t2@0 t3@0 t4@0 t5@0;
## t2 WITH t3@0 t4@0 t5@0;
## t3 WITH t4@0 t5@0;
## t4 WITH t5@0;
```

MplusAutomation Implementation

- Heterogeneous

```
mplusRcov(timeV,"heterogenous",collapse = TRUE)
```

```
## t1 t2 t3 t4 t5;
## t1 WITH t2@0 t3@0 t4@0 t5@0;
## t2 WITH t3@0 t4@0 t5@0;
## t3 WITH t4@0 t5@0;
## t4 WITH t5@0;
```

MplusAutomation Implementation

- Compound Symmetry

```
mplusRcov(timeV,"cs",collapse = TRUE)
```

```
## t1 t2 t3 t4 t5 (e);  
## t1 t2 t3 t4 PWITH t2 t3 t4 t5 (rho);  
## t1 t2 t3 PWITH t3 t4 t5 (rho);  
## t1 t2 PWITH t4 t5 (rho);  
## t1 PWITH t5 (rho);
```

MplusAutomation Implementation

- Toeplitz

```
mplusRcov(timeV,"toeplitz",collapse = TRUE)
```

```
## t1 t2 t3 t4 t5 (e);  
## t1 t2 t3 t4 PWITH t2 t3 t4 t5 (rho);  
## t1 t2 t3 PWITH t3 t4 t5 (rho2);  
## t1 t2 PWITH t4 t5 (rho3);  
## t1 PWITH t5 (rho4);
```

MplusAutomation Implementation

- Autoregressive

```
mplusRcov(timeV,"ar",collapse = TRUE)
```

```
## t1 t2 t3 t4 t5 (e);  
## t1 t2 t3 t4 PWITH t2 t3 t4 t5 (rho);  
## t1 t2 t3 PWITH t3 t4 t5 (rho2);  
## t1 t2 PWITH t4 t5 (rho3);  
## t1 PWITH t5 (rho4);  
## MODEL CONSTRAINT:  
##   rho2 = ((rho/e)^2) * e;  
##   rho3 = ((rho/e)^3) * e;  
##   rho4 = ((rho/e)^4) * e;
```

MplusAutomation Implementation

- Unstructured: $\frac{p*(p+1)}{2}$
 - p = No. of parameters

```
mplusRcov(timeV,"un",collapse = TRUE)
```

```
## t1 t2 t3 t4 t5;  
## t1 WITH t2 t3 t4 t5;  
## t2 WITH t3 t4 t5;  
## t3 WITH t4 t5;  
## t4 WITH t5;
```

Interactive Run Models

```
runModels_Interactive()
```