Modelling three multilevel models with Ime4 package in R

NHZ

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Introduction

Lets fit some models with

- 1. Outcome: Clinical decision-making
- 2. Level 1 (Individual-level factors): Age, Gender, Education, Experience, post basic (PB), continous professional development (CPD), knowledge (NKS) and critical thinking (CTS)
- 3. Level 2 (Hospital-level factors): Unit, Department size, hospital settings and hospital size
- 4. Level 3 (Regional-level factors): State, state size

Install packages

```
install.packages("lme4")
install.packages("lmerTest")
install.packages("performance")
install.packages ("reghelper")
```

Load the library

```
library(lme4) #multilevel analysis

Loading required package: Matrix

library(lmerTest) # to get p-value estimations

Attaching package: 'lmerTest'

The following object is masked from 'package:lme4':
    lmer
```

```
The following object is masked from 'package:stats':

step

library(performance) # to get ICC
library(reghelper) # to plot the interaction

Attaching package: 'reghelper'

The following object is masked from 'package:base':

beta
```

Import the data

```
library(readxl)
CTCDMR <- read_excel("CTCDMR.xlsx", sheet = "sheet1")</pre>
```

Combine hospital and department ID within each region to create a unique identifier for **hosp.dep**

```
CTCDMR$hosp.dep <- paste(CTCDMR$Hospital, CTCDMR$DepartmentID, sep = "_")
```

Select Data

```
library(dplyr)

Attaching package: 'dplyr'

The following objects are masked from 'package:stats':
    filter, lag
```

```
The following objects are masked from 'package:base':
```

intersect, setdiff, setequal, union

CTCDMR <- select(CTCDMR, NurseID, hosp.dep, DepartmentID, Hospital, Age, Gender, Education, 1 head(CTCDMR, n=50)

```
# A tibble: 50 x 20
  NurseID hosp.dep DepartmentID Hospital Age Gender Education Experience
    <dbl> <chr>
                            <dbl> <chr> <dbl> <chr> <dbl> <chr>
                                                                     <dbl>
        1 Hosp Segamat~
                                1 Hosp Se~ 31 Female Diploma
                                                                        10
        2 Hosp Segamat~
                                1 Hosp Se~ 31 Female Diploma
 2
                                                                         8
                                1 Hosp Se~ 33 Female Diploma
        3 Hosp Segamat~
                                                                         8
 4
        4 Hosp Segamat~
                               1 Hosp Se~ 32 Female Diploma
                                                                         9
5
        5 Hosp Segamat~
                               1 Hosp Se~ 35 Female Diploma
                                                                        13
                               1 Hosp Se~ 36 Female Diploma
        6 Hosp Segamat~
                                                                       13
                               1 Hosp Se~ 25 Female Diploma
7
        7 Hosp Segamat~
                                                                        4
                                1 Hosp Se~
8
        8 Hosp Segamat~
                                              30 Female Diploma
9
                               1 Hosp Se~
                                                                        10
        9 Hosp Segamat~
                                              31 Female Diploma
10
       10 Hosp Segamat~
                                1 Hosp Se~
                                              37 Female Diploma
                                                                       14
# i 40 more rows
# i 12 more variables: PB <chr>, CPD <dbl>, HospitalSet <chr>,
   HospitalSize <dbl>, DepSize <dbl>, Unit <chr>, State <chr>, Region <chr>,
```

- StateSize <dbl>, CDMS <dbl>, CTS <dbl>, NKS <dbl>

Two-level Multilevel analysis

Null model

What is the average score of the CDM for the entire group as well as for each individual?

```
m0 <- lmer(CDMS ~ 1 + (1 |hosp.dep), data = CTCDMR, REML = FALSE)
summary(m0)
```

```
Linear mixed model fit by maximum likelihood . t-tests use Satterthwaite's
 method [lmerModLmerTest]
```

Formula: CDMS ~ 1 + (1 | hosp.dep)

Data: CTCDMR

```
AIC BIC logLik deviance df.resid 4752.1 4765.9 -2373.1 4746.1 716
```

Scaled residuals:

Min 1Q Median 3Q Max -2.6003 -0.7153 0.0565 0.7715 2.8980

Random effects:

Groups Name Variance Std.Dev.
hosp.dep (Intercept) 14.52 3.811
Residual 38.71 6.221
Number of obs: 719, groups: hosp.dep, 36

Fixed effects:

```
Estimate Std. Error df t value Pr(>|t|)
(Intercept) 36.2638 0.6762 35.9611 53.63 <2e-16 ***
---
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

The intra-class correlation (ICC):

```
library(performance)
performance::icc(m0)
```

Intraclass Correlation Coefficient

Adjusted ICC: 0.273 Unadjusted ICC: 0.273

Random intercept model

Add level 1 and 2 variables:

Do the CDMS is predicted by the nurse's individual-level and hospital-level variables?

```
ri <- lmer(CDMS ~ NKS + Age + Gender + Education + Experience + PB + CPD + CTS + Unit + DepS: summary(ri)
```

```
Linear mixed model fit by maximum likelihood . t-tests use Satterthwaite's
  method [lmerModLmerTest]
Formula: CDMS ~ NKS + Age + Gender + Education + Experience + PB + CPD +
```

```
CTS + Unit + DepSize + HospitalSet + HospitalSize + (1 | hosp.dep)
Data: CTCDMR
```

AIC BIC logLik deviance df.resid 3905.3 3978.5 -1936.6 3873.3 703

Scaled residuals:

Min 1Q Median 3Q Max -3.523 -0.550 0.053 0.643 2.922

Random effects:

Groups Name Variance Std.Dev.
hosp.dep (Intercept) 0.1986 0.4456
Residual 12.6223 3.5528
Number of obs: 719, groups: hosp.dep, 36

Fixed effects:

	Estimate	Std. Error	df	t value	Pr(> t)	
(Intercept)	2.257e+00	2.221e+00	7.121e+02	1.016	0.3099	
NKS	1.117e-01	9.831e-03	7.161e+02	11.364	<2e-16	***
Age	-2.902e-02	7.965e-02	7.159e+02	-0.364	0.7157	
GenderMale	2.337e-02	7.909e-01	7.179e+02	0.030	0.9764	
EducationDiploma	-1.578e-01	6.274e-01	7.188e+02	-0.252	0.8015	
EducationMaster	3.576e+00	2.626e+00	7.118e+02	1.362	0.1738	
Experience	6.053e-02	8.160e-02	7.162e+02	0.742	0.4585	
PBYES	4.119e-01	3.111e-01	7.102e+02	1.324	0.1859	
CPD	1.690e-02	2.416e-02	4.105e+02	0.699	0.4848	
CTS	5.660e-01	3.215e-02	7.161e+02	17.606	<2e-16	***
UnitMedical	-4.530e-01	3.188e-01	3.781e+01	-1.421	0.1635	
DepSize	1.348e-02	1.652e-02	4.749e+01	0.816	0.4184	
${\tt HospitalSetGeneral}$	-1.589e+00	7.179e-01	3.669e+01	-2.214	0.0332	*
HospitalSize	-4.175e-04	5.965e-04	4.060e+01	-0.700	0.4880	

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Correlation matrix not shown by default, as p = 14 > 12. Use print(x, correlation=TRUE) or vcov(x) if you need it

Random Slopes model

Do the CDMS is predicted by the variables at both individual-level and hospital level?

- Do the combined effect of nurse's knowledge and experience vary across different hospital department?
 - Outcome: Clinical decision-making
 - Fixed: Knowledge, experience and hospital settings
 - Random: Knowledge and experience
- Different scale?
 - So we gonna use grand mean centering: NKS -> NKS.GM

```
CTCDMR$NKS.GM <- scale(CTCDMR$NKS, scale = F)</pre>
CTCDMR$CTS.GM <- scale(CTCDMR$CTS, scale = F)</pre>
rs <- lmer(CDMS ~ NKS.GM + Experience + factor(HospitalSet) +
(1 + NKS.GM + Experience | hosp.dep),
data = CTCDMR, REML = FALSE,lmerControl(optimizer = 'bobyqa'))
summary(rs)
Linear mixed model fit by maximum likelihood . t-tests use Satterthwaite's
  method [lmerModLmerTest]
Formula: CDMS ~ NKS.GM + Experience + factor(HospitalSet) + (1 + NKS.GM +
    Experience | hosp.dep)
   Data: CTCDMR
Control: lmerControl(optimizer = "bobyqa")
     AIC
              BIC
                    logLik deviance df.resid
  4106.9
           4157.2 -2042.4
                             4084.9
Scaled residuals:
            1Q Median
    Min
                             30
                                    Max
-3.1879 -0.6149 0.0070 0.6467 3.0585
Random effects:
 Groups
         Name
                      Variance Std.Dev. Corr
 hosp.dep (Intercept) 1.093e+01 3.30591
          NKS.GM
                      3.031e-04 0.01741
                                          0.57
          Experience 6.959e-02 0.26380 -0.99 -0.55
 Residual
                      1.565e+01 3.95652
```

```
Number of obs: 719, groups: hosp.dep, 36
Fixed effects:
                      Estimate Std. Error
                                            df t value Pr(>|t|)
                     34.364635  0.730043  29.216304  47.072  < 2e-16 ***
(Intercept)
NKS.GM
                      Experience
factor(HospitalSet)General -1.266490 0.454201 38.192612 -2.788 0.008213 **
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Correlation of Fixed Effects:
         (Intr) NKS.GM Exprnc
NKS.GM
          0.528
Experience -0.919 -0.378
fctr(HspS)G -0.429 -0.602 0.136
```

Three-level Multilevel analysis

```
(1 \mid \text{Region/hosp.dep}) or (1 \mid \text{Region}) + (1 \mid \text{Region:DepartmentID}) Both will give the same results, but there are some analysis that need to use the longer code
```

Null model

Short version

```
L3.M0a <- lmer(CDMS ~ 1
+ (1 | Region/hosp.dep), data = CTCDMR)
summary(L3.M0a)
```

```
Linear mixed model fit by REML. t-tests use Satterthwaite's method [
lmerModLmerTest]

Formula: CDMS ~ 1 + (1 | Region/hosp.dep)

Data: CTCDMR

REML criterion at convergence: 4744.4
```

Scaled residuals:

```
Min 1Q Median
                          3Q
-2.5720 -0.7162 0.0362 0.7438 2.9257
Random effects:
                            Variance Std.Dev.
 Groups
                Name
hosp.dep:Region (Intercept) 12.436
                                   3.527
                                     1.953
Region
                (Intercept) 3.814
                                     6.222
Residual
                            38.718
Number of obs: 719, groups: hosp.dep:Region, 37; Region, 6
Fixed effects:
           Estimate Std. Error
                                  df t value Pr(>|t|)
(Intercept) 36.076
                        1.031 4.200
                                      35.01 2.43e-06 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Longer version
L3.MOb <- lmer(CDMS ~ 1
+ (1|Region)
+ (1|Region: DepartmentID), data = CTCDMR)
summary(L3.MOb)
Linear mixed model fit by REML. t-tests use Satterthwaite's method [
lmerModLmerTest]
Formula: CDMS ~ 1 + (1 | Region) + (1 | Region:DepartmentID)
  Data: CTCDMR
REML criterion at convergence: 4744.4
Scaled residuals:
         1Q Median
                            3Q
                                   Max
-2.5720 -0.7162 0.0362 0.7438 2.9257
Random effects:
Groups
                    Name
                                Variance Std.Dev.
Region:DepartmentID (Intercept) 12.436
Region
                    (Intercept) 3.814
                                         1.953
Residual
                                38.718
                                         6.222
Number of obs: 719, groups: Region:DepartmentID, 37; Region, 6
```

```
Fixed effects:
```

```
Estimate Std. Error df t value Pr(>|t|)
                        1.031 4.200 35.01 2.43e-06 ***
(Intercept) 36.076
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Random intercept model

Do the CDMS is predicted by the nurse's individual and hospital department variables?

```
L3.ri <- lmer(CDMS ~ CTS + NKS + HospitalSize + (1|Region)
+ (1|Region: DepartmentID), data = CTCDMR, REML = FALSE)
summary(L3.ri)
```

```
Linear mixed model fit by maximum likelihood . t-tests use Satterthwaite's
 method [lmerModLmerTest]
Formula:
CDMS ~ CTS + NKS + HospitalSize + (1 | Region) + (1 | Region:DepartmentID)
  Data: CTCDMR
     AIC
             BIC
                  logLik deviance df.resid
          3938.6 -1946.3
                            3892.6
  3906.6
Scaled residuals:
           1Q Median
                            3Q
-3.5960 -0.5589 0.0381 0.6226 2.6628
Random effects:
 Groups
                    Name
                                Variance Std.Dev.
Region:DepartmentID (Intercept) 0.27588 0.5252
Region
                     (Intercept) 0.06633 0.2575
```

```
Residual
                                 12.87336 3.5879
```

Number of obs: 719, groups: Region:DepartmentID, 37; Region, 6

Fixed effects:

```
Estimate Std. Error
                                         df t value Pr(>|t|)
(Intercept)
             1.847e+00 8.332e-01 3.356e+02 2.217
                                                    0.0273 *
             5.957e-01 3.131e-02 7.092e+02 19.024
CTS
                                                     <2e-16 ***
NKS
             1.075e-01 8.622e-03 4.433e+02 12.469
                                                     <2e-16 ***
HospitalSize -9.103e-04 3.323e-04 4.838e+01 -2.739
                                                     0.0086 **
```

Random Slope model

Do the CDMS is predicted by the variables at both individual-level, hospital level, and regional level? - Do the CTS effect vary across different region?

```
L3.RS <-lmer(CDMS ~ CTS.GM + Experience + Unit + HospitalSet
+ (1+ CTS.GM|Region)
+ (1|Region: DepartmentID),
data=CTCDMR, REML=FALSE,lmerControl(optimizer = 'bobyqa'))
summary(L3.RS)
Linear mixed model fit by maximum likelihood . t-tests use Satterthwaite's
  method [lmerModLmerTest]
Formula: CDMS ~ CTS.GM + Experience + Unit + HospitalSet + (1 + CTS.GM |
    Region) + (1 | Region:DepartmentID)
   Data: CTCDMR
Control: lmerControl(optimizer = "bobyqa")
                    logLik deviance df.resid
     AIC
             BIC
  4018.9
           4064.6 -1999.4
                            3998.9
Scaled residuals:
            1Q Median
                             30
                                    Max
-3.3170 -0.6134 -0.0153 0.6388 2.8700
Random effects:
 Groups
                     Name
                                 Variance Std.Dev. Corr
 Region:DepartmentID (Intercept) 0.161755 0.40219
 Region
                     (Intercept) 0.716231 0.84630
                                  0.006558 0.08098 -0.01
                     CTS.GM
 Residual
                                 14.734358 3.83854
Number of obs: 719, groups: Region:DepartmentID, 37; Region, 6
```

Fixed effects:

```
Estimate Std. Error
                                             df t value Pr(>|t|)
                   34.07860 0.56786 25.76862 60.012 < 2e-16 ***
(Intercept)
CTS.GM
                              0.04292
                                       7.81381 19.188 7.46e-08 ***
                   0.82348
                   0.10075
                              0.02371 487.04291
                                                 4.250 2.56e-05 ***
Experience
                              0.32014 27.42463 -0.132 0.895744
UnitMedical
                   -0.04234
                   1.35417
                              0.34664 36.16166
                                                 3.907 0.000394 ***
HospitalSetGeneral
```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Correlation of Fixed Effects:

(Intr) CTS.GM Exprnc UntMdc

CTS.GM 0.212

Experience -0.611 -0.268

UnitMedical -0.361 -0.014 0.143

HsptlStGnrl -0.356 -0.226 0.100 -0.004

Interactions and Cross-Level Interactions

 ${\bf Cross\text{-}level}$ interaction: Experience (level 1) and hospital Set (level 2)

Interactions: NKS and experience, CTS and NKS, CTS and Experience

```
L3.int <- lmer (CDMS ~ NKS.GM * Experience + CTS.GM * NKS.GM + CTS.GM * Experience + Experience + (1|Region: DepartmentID), data = CTCDMR, REML = FALSE)
summary(L3.int)
```

```
Linear mixed model fit by maximum likelihood . t-tests use Satterthwaite's method [lmerModLmerTest]
```

```
Formula: CDMS ~ NKS.GM * Experience + CTS.GM * NKS.GM + CTS.GM * Experience + Experience * HospitalSet + (1 | Region) + (1 | Region:DepartmentID)

Data: CTCDMR
```

```
AIC BIC logLik deviance df.resid 3858.7 3913.7 -1917.4 3834.7 707
```

Scaled residuals:

```
Min 1Q Median 3Q Max -3.7588 -0.5765 0.0370 0.6624 2.7531
```

Random effects:

```
Groups Name Variance Std.Dev.
Region:DepartmentID (Intercept) 0.34686 0.5889
Region (Intercept) 0.05199 0.2280
Residual 11.82346 3.4385
```

Number of obs: 719, groups: Region:DepartmentID, 37; Region, 6

Fixed effects:

	Estimate	Std. Error	df	t value Pr(> t)
(Intercept)	3.371e+01	5.668e-01	1.407e+02	59.464 < 2e-16
NKS.GM	8.787e-02	1.792e-02	6.863e+02	4.903 1.18e-06
Experience	2.408e-01	3.557e-02	4.891e+02	6.771 3.69e-11
CTS.GM	7.858e-01	6.637e-02	7.146e+02	11.839 < 2e-16
HospitalSetGeneral	1.939e+00	7.337e-01	2.421e+02	2.643 0.00876
NKS.GM:Experience	1.701e-03	1.246e-03	7.141e+02	1.365 0.17257
NKS.GM:CTS.GM	2.589e-03	9.005e-04	6.880e+02	2.875 0.00416
Experience: CTS.GM	-1.942e-02	4.393e-03	7.142e+02	-4.420 1.14e-05
Experience: HospitalSetGeneral	-2.473e-01	4.937e-02	6.301e+02	-5.010 7.08e-07

 (Intercept)

 NKS.GM

 Experience

 CTS.GM

 HospitalSetGeneral
 **

 NKS.GM:Experience
 NKS.GM:CTS.GM

Experience:CTS.GM ***
Experience:HospitalSetGeneral ***

Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1

Correlation of Fixed Effects:

(Intr) NKS.GM Exprnc CTS.GM HsptSG NKS.GM:E NKS.GM:C E:CTS.

NKS.GM 0.227

Experience -0.865 -0.244

CTS.GM -0.137 -0.618 0.181

HsptlStGnrl -0.717 -0.446 0.657 0.104

NKS.GM:Expr -0.144 -0.851 0.251 0.464 0.404

NKS.GM:CTS. -0.310 0.319 0.251 0.278 -0.081 -0.340

Expr:CTS.GM 0.242 0.515 -0.351 -0.870 -0.169 -0.500 -0.269

Exprnc:HsSG 0.576 0.418 -0.714 -0.135 -0.832 -0.555 0.047 0.236

Plot interaction

```
graph_model(
  model = L3.int,
  y = CDMS,
  x = Experience,
  lines = HospitalSet,
  errorbars = "none")
```

