My Amazing Solutions for Problem Set 3: Inference and Multilevel Modeling

2023-10-01

This is a group assignment. Please do not include your names in your PDF submission (Canvas group membership connects your submission to your names without the teaching team seeing them!).

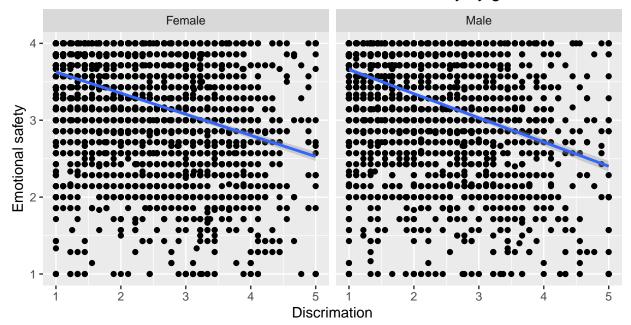
The Problems

1. Present a favorite plot or display from your work

```
mycleandata <- readRDS("./mycleandata.rds")</pre>
dat <- mycleandata
ggplot( dat, aes( x = disc, y = esafe ) ) +
  geom_point() +
  geom_smooth( method = "lm" ) +
  labs( x = "Discrimation", y = "Emotional safety" ) +
  facet_wrap( ~ gender ) + ggtitle("Association between discrimation and emotional safety by gender") +
  labs(caption =
  " This scatter plot shows the association between students sense of discrimination
  and emotional safety at school, without accounting for clustering of students
  within schools. We plotted self-identified male and female gender separately.
  The blue line represents the estimated regression line. The plots show an increased
  sense of discrimination is associated with a decreased sense of emotional security.
") +
  theme(
  plot.caption = element_text(hjust = 0)
```

'geom_smooth()' using formula = 'y ~ x'

Association between discrimation and emotional safety by gender



This scatter plot shows the association between students sense of discrimination and emotional safety at school, without accounting for clustering of students within schools. We plotted self–identified male and female gender separately. The blue line represents the estimated regression line. The plots show an increased sense of discrimination is associated with a decreased sense of emotional security.

This scatter plot shows the association between students sense of discrimination and emotional safety at school, without accounting for clustering of students within schools. We plotted self-identified male and female gender separately. The blue line represents the estimated regression line. The plots show an increased sense of discrimination is associated with a decreased sense of emotional security.

We chose to show a scatter plot over a histogram as we wanted to visualize the association between the two variables, discrimination and emotional safety, rather than the distributions of the individual variables.

2. Does sense of emotional safety vary by gender and grade?

Our model will contain a random intercept for each school, a random slope for gender while including grade as a covariate due to the following reasons: From the data exploration we saw that the data was clustered, therefore our model will include a random intercept for school ID. Since different school have surveys collected from different grades, we decided to control for grade in our model, specially using it as predictor for the random intercept (grade may explain partially why there is school clustered effect). The relationship between emotional safety and gender/grade may vary across schools, therefore we have decided to fit a random slope model. In addition, we will also include grade in our level 1 model to see whether emotional vary by gender.

The model is as follows:

$$esafe_{ij} = \beta_{0j} + \beta_{1j}(gender)_{ij} + \beta_2(grade)_{ij} + \varepsilon_{ij}$$
$$\beta_{0j} = \gamma_{00} + \gamma_{01}\overline{grade}_j + \zeta_{0j}$$
$$\beta_{1j} = \gamma_{10} + \zeta_{1j}$$

Fitting the model:

```
#standardizing esafe and disc
dat$esafe_std <- scale(dat$esafe)
dat$disc_std <- scale(dat$disc)
dat$gender <- as.factor(dat$gender)
#calculate mean grade of each school
dat <- dat %>% group_by(ID) %>% mutate(grade_mean = mean(grade), disc_mean = mean(disc))
m2 <- lmer(esafe_std ~ gender + grade + grade_mean + (1 + gender ID), dat)
summary(m2)</pre>
```

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
  Formula: esafe_std ~ gender + grade + grade_mean + (1 + gender | ID)
##
      Data: dat
##
## REML criterion at convergence: 21727.4
##
## Scaled residuals:
##
       Min
                1Q Median
                                3Q
                                       Max
##
   -4.1038 -0.6039 0.2552 0.7356
##
## Random effects:
##
   Groups
                         Variance Std.Dev. Corr
##
             (Intercept) 0.118763 0.34462
   TD
##
             genderMale 0.004856 0.06969
##
   Residual
                         0.868768 0.93208
  Number of obs: 8002, groups: ID, 39
##
##
## Fixed effects:
##
                 Estimate Std. Error
                                             df t value Pr(>|t|)
## (Intercept) -1.506e-01 2.904e-01 4.038e+01
                                                 -0.519
                                                           0.6069
## genderMale
                9.156e-02 2.676e-02 2.001e+01
                                                  3.422
                                                           0.0027 **
                4.102e-02 7.336e-03 7.930e+03
## grade
                                                  5.592 2.32e-08 ***
## grade mean -1.917e-02 3.363e-02 4.458e+01
                                                -0.570
                                                           0.5714
```

```
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##
              (Intr) gndrMl grade
## genderMale -0.016
              -0.001 0.025
## grade
## grade_mean -0.956 -0.004 -0.218
m2 %>% tidy()
## # A tibble: 8 x 8
##
     effect
              group
                        term
                                        estimate std.error statistic
                                                                           df
                                                                               p.value
##
     <chr>>
              <chr>>
                        <chr>
                                            <dbl>
                                                      <dbl>
                                                                 <dbl>
                                                                        <dbl>
                                                                                  <dbl>
## 1 fixed
              <NA>
                        (Intercept)
                                          -0.151
                                                    0.290
                                                                -0.519
                                                                         40.4
                                                                               6.07e-1
## 2 fixed
              <NA>
                        genderMale
                                           0.0916
                                                    0.0268
                                                                 3.42
                                                                         20.0
                                                                               2.70e-3
## 3 fixed
              <NA>
                                                    0.00734
                                                                 5.59 7930.
                                                                               2.32e-8
                        grade
                                          0.0410
## 4 fixed
              <NA>
                        grade mean
                                          -0.0192
                                                    0.0336
                                                                -0.570
                                                                         44.6 5.71e-1
## 5 ran_pars ID
                        sd__(Intercept)
                                           0.345
                                                               NA
                                                                         NA
                                                                              NA
                                                   NA
## 6 ran_pars ID
                        cor__(Intercep~
                                           0.223
                                                   NA
                                                               NA
                                                                         NA
                                                                              NA
```

7 ran_pars ID

The model indicates there are gender differences. Controlling for grade, on average females have 0.0916 units increase in the SD of emotional safety score as compared to males. The P-value of 0.003 < 0.05 significance level, and a t-value of 3.422 > 1.96 also indicating statistical significance. We can reject the null hypothesis that there is no difference in emotional safety score between males and females, controlling for grade. Controlling for gender, on average each 1 grade higher predicts a 0.0410 units increase in the SD of emotional safety score.

0.0697

0.932

NA

NA

NA

NA

NA

NA

NA

NA

sd__genderMale

8 ran_pars Residual sd__Observation

3. Do those who feel discriminated against feel less safe?

Answer: The statistical model should allow the relationship between emotional safety and feeling discriminated to vary across school. Since gender and grade are predictors of emotional safety, we will include them in our model to control for them. We will use a multilevel model to account for the nested structure of the data. The model is as follows:

$$esafe_{ij} = \beta_{0j} + \beta_{1j}(disc)_{ij} + \beta_{2}(gender)_{ij} + \beta_{2}(grade)_{ij} + \varepsilon_{ij}$$
$$\beta_{0j} = \gamma_{00} + \gamma_{01}\overline{disc}_{j} + \gamma_{02}\overline{grade}_{j} + \zeta_{0j}$$
$$\beta_{1j} = \gamma_{10} + \zeta_{1j}$$

where i indexes students and j indexes schools.

```
#create a group centered discrimination variable
dat<-dat%>%
  mutate(disc_dev=disc_std-disc_mean)
```

```
m3_b <- lmer(esafe_std ~ disc_dev + disc_mean + gender + grade + grade_mean + (1+disc_dev|ID), dat)
m3_b %>% tidy() %>% knitr::kable(digits = 4)
```

effect	group	term	estimate	std.error	statistic	df	p.value
fixed	NA	(Intercept)	0.0822	0.2248	0.3659	35.0367	0.7166
fixed	NA	$\operatorname{disc_dev}$	-0.3663	0.0205	-17.8627	27.4154	0.0000
fixed	NA	disc_mean	-0.6661	0.0869	-7.6627	30.4478	0.0000
fixed	NA	genderMale	0.0257	0.0212	1.2117	7886.0774	0.2257
fixed	NA	grade	0.0549	0.0069	7.9713	7946.0246	0.0000
fixed	NA	grade_mean	0.0023	0.0220	0.1051	42.7374	0.9168
ran_pars	ID	sd (Intercept)	0.2894	NA	NA	NA	NA
ran_pars	ID	cor(Intercept).disc_dev	0.7204	NA	NA	NA	NA
ran_pars	ID	$sd__disc_dev$	0.0917	NA	NA	NA	NA
ran_pars	Residual	sd Observation	0.8716	NA	NA	NA	NA

```
arm::display(m3_b)
```

```
## lmer(formula = esafe_std ~ disc_dev + disc_mean + gender + grade +
       grade_mean + (1 + disc_dev | ID), data = dat)
##
##
               coef.est coef.se
               0.08
                         0.22
## (Intercept)
## disc_dev
               -0.37
                         0.02
## disc_mean
               -0.67
                         0.09
## genderMale
                0.03
                         0.02
## grade
                0.05
                         0.01
## grade_mean
                0.00
                         0.02
##
## Error terms:
  Groups
                         Std.Dev. Corr
             (Intercept) 0.29
##
  ID
##
             disc_dev
                         0.09
                                  0.72
## Residual
                         0.87
## ---
```

```
## number of obs: 8002, groups: ID, 39
## AIC = 20675.7, DIC = 20587.9
## deviance = 20621.8

#confidence intervals
confint(m3_b, method = c("Wald")) %>% knitr::kable(digits = 3)
```

	2.5 %	97.5 %
.sig01	NA	NA
.sig02	NA	NA
.sig03	NA	NA
.sigma	NA	NA
(Intercept)	-0.358	0.523
$\operatorname{disc}_{\operatorname{dev}}$	-0.407	-0.326
$\operatorname{disc_mean}$	-0.836	-0.496
genderMale	-0.016	0.067
grade	0.041	0.068
${\rm grade_mean}$	-0.041	0.045

The model indicates that there is a negative relationship between emotional safety and feeling discriminated against. Also note the are not assuming that the relationship between discrimination and emotional safety is the same across schools, allowing each to vary (have its own slope and its own intercept.)

Within schools, on average, a one-unit increase in SD of discrimination score predicts a 0.37 (95%CI:-0.407, -0.324) units **decrease** in the SD of emotional safety, adjusting for gender, grade, and school average grade. This is statistically significant (95% CI does not include 0).

Schools whose average discrimination score is 1SD higher are predicted to have 0.67 (95%CI:-0.490, -0.147) units **decrease** in the SD of emotional safety, adjusting for grade, school average grade, and gender. This is also statistically significant (95% CI does not include 0).

. Contextual effects for discrimination?

```
m4 <- lmer(esafe_std ~ disc_std + disc_mean + gender + grade + grade_mean + (1 | ID), dat)
summary(m4)
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: esafe_std ~ disc_std + disc_mean + gender + grade + grade_mean +
##
       (1 | ID)
##
     Data: dat
##
## REML criterion at convergence: 20692.3
##
## Scaled residuals:
               1Q Median
      Min
                               3Q
                                      Max
## -4.5170 -0.5779 0.2148 0.6696 3.0844
## Random effects:
## Groups
            Name
                        Variance Std.Dev.
## ID
             (Intercept) 0.04109 0.2027
## Residual
                        0.76625 0.8754
## Number of obs: 8002, groups: ID, 39
##
## Fixed effects:
                Estimate Std. Error
                                            df t value Pr(>|t|)
## (Intercept) 8.303e-02 2.237e-01 3.466e+01
                                                0.371 0.71271
              -3.540e-01 1.077e-02 7.957e+03 -32.875 < 2e-16 ***
## disc_std
## disc_mean
              -2.995e-01 8.830e-02 3.247e+01
                                               -3.392 0.00184 **
              2.900e-02 2.129e-02 7.905e+03
                                                 1.362 0.17312
## genderMale
## grade
               5.658e-02 6.897e-03 7.955e+03
                                                 8.202 2.72e-16 ***
## grade_mean
              5.239e-04 2.211e-02 4.376e+01
                                                 0.024 0.98121
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Correlation of Fixed Effects:
##
             (Intr) dsc_st dsc_mn gndrMl grade
## disc_std
              0.101
## disc_mean -0.581 -0.129
## genderMale -0.012 0.082 -0.042
## grade
             -0.007 -0.068 0.008 0.024
## grade_mean -0.622 0.021 -0.203 -0.005 -0.312
confint(m4, method = c("Wald")) %>% knitr::kable(digits = 3)
```

	2.5~%	97.5 %
.sigma	NA	NA
(Intercept)	-0.355	0.521
$\operatorname{disc_std}$	-0.375	-0.333
$\operatorname{disc_mean}$	-0.473	-0.126
genderMale	-0.013	0.071
grade	0.043	0.070
${\rm grade_mean}$	-0.043	0.044

a) Is there a contextual effect for discrimination with regard to emotional safety?

Yes, we see a contextual effect for discrimination with regard to emotional safety.

Comparing two students of the same individual discrimination score, a student that goes to a school whose average discrimination score is 1 SD higher is predicted to have an emotional safety score 0.30 points lower (p = 0.0018).

b) Once you take out the contextual effect, if any, does the relationship at the student level of sense of discrimination and emotional safety change?

```
m5 <- lmer(esafe_std ~ disc_std + gender + grade + grade_mean + (1|ID), dat)
summary(m5)</pre>
```

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
  Formula: esafe_std ~ disc_std + gender + grade + grade_mean + (1 | ID)
##
##
##
## REML criterion at convergence: 20698.9
##
## Scaled residuals:
##
      Min
                1Q Median
                                ЗQ
                                       Max
##
  -4.5272 -0.5767 0.2173 0.6692
##
## Random effects:
##
   Groups
                         Variance Std.Dev.
##
   ID
             (Intercept) 0.05899 0.2429
   Residual
                         0.76598 0.8752
## Number of obs: 8002, groups: ID, 39
##
## Fixed effects:
##
                 Estimate Std. Error
                                             df t value Pr(>|t|)
## (Intercept) -3.396e-01 2.106e-01
                                      3.916e+01
                                                 -1.613
                                                           0.115
## disc_std
               -3.574e-01
                           1.070e-02
                                      7.892e+03
                                                -33.398
                                                         < 2e-16 ***
## genderMale
                2.695e-02
                           2.130e-02
                                      7.956e+03
                                                  1.265
                                                           0.206
                5.671e-02
                           6.896e-03
                                      7.959e+03
                                                  8.223
                                                         2.3e-16 ***
## grade
## grade_mean -1.682e-02 2.475e-02 4.558e+01
                                                 -0.680
                                                           0.500
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Correlation of Fixed Effects:
```

```
## (Intr) dsc_st gndrMl grade
## disc_std 0.028
## genderMale -0.038 0.079
## grade -0.003 -0.067 0.024
## grade mean -0.940 -0.004 -0.013 -0.277
```

```
screenreg(list(m4,m5))
```

```
##
  _____
##
                                   Model 2
                     Model 1
##
                                       -0.34
##
                          0.08
  (Intercept)
##
                         (0.22)
                                       (0.21)
                         -0.35 ***
                                       -0.36 ***
## disc_std
##
                         (0.01)
                                       (0.01)
## disc_mean
                         -0.30 ***
##
                         (0.09)
                          0.03
                                       0.03
##
  genderMale
                         (0.02)
                                       (0.02)
##
                                        0.06 ***
##
  grade
                          0.06 ***
##
                         (0.01)
                                       (0.01)
                          0.00
                                       -0.02
##
  grade_mean
##
                         (0.02)
                                       (0.02)
##
## AIC
                      20708.26
                                    20712.89
## BIC
                      20764.16
                                    20761.80
## Log Likelihood
                     -10346.13
                                   -10349.44
## Num. obs.
                       8002
                                     8002
## Num. groups: ID
                         39
                                       39
## Var: ID (Intercept)
                          0.04
                                        0.06
## Var: Residual
                          0 77
                                        0 77
## *** p < 0.001; ** p < 0.01; * p < 0.05
```

When accounting for the contextual effect, the student level effect of sense of discrimination on emotional safety is -0.35 with SD 0.011 (p <0.0001) and when we remove the contextual effect is -0.36 and SD 0.011 (p <0.0001). Removing the contextual effect does not result in a large change in the effect of discrimination on emotional safety at the student level.

Once you have answered this, reflect on the following:

c) What, if anything, does it add substantively to do this within-school and contextual effect analysis?

The within school analysis allows us to look at how a change in individual discrimination affects emotional safety within schools. This allows for us to address clustering and account for systematic differences of schools that we may have been unable to measure.

The contextual effect analysis allows us to look at the effect of a group level variable such as mean discrimination on an individual level outcome, controlling for individual level discrimination as well. In other words looking at differences between schools, holding individual discrimination scores constant, we are able to look at the effect of school level discrimination score on predicted individual reading comprehension scores.

Overall, these analyses allow us to answer specific research questions in the presence of clustering.