Lab 3 - Moderation

Structural Equation Modeling ED 216F - Instructor: Karen Nylund-Gibson

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1 Lab preparation

1.1 Creating a version-controlled R-Project with Github

Download repository here: https://github.com/garberadamc/SEM-Lab2 On the Github repository webpage:

- a. fork your own branch of the lab repository
- b. copy the repository web URL address from the clone or download menu

Within R-Studio:

- c. click "NEW PROJECT" (upper right corner of window)
- d. choose option Version Control

- e. choose option Git
- f. paste the repository web URL path copied from the clone or download menu on Github page
- g. choose location of the R-Project (too many nested folders will result in filepath error)

1.2 Data source for example 1:

The first example utilizes the *Vocabulary and Education* dataset from the National Opinion Research Center General Social Survey. GSS Cumulative Datafile 1972-2016 (Fox, 2008) See documentation here

This dataset is available via the R-package {carData} and can be directly loaded into the R environment.

Note: All models specified in the following exercise are for demonstration only and are **not** theoretically justified or valid.

```
# equatiomatic is not yet on CRAN. Install the development version from GitHub with remotes::install_github("datalorax/equatiomatic", force = TRUE)
```

```
library(tidyverse)
library(MplusAutomation)
library(rhdf5)
library(here)
library(gt)
library(gtsummary)
library(estimatr)
library(kableExtra)
library(hrbrthemes)
library(equatiomatic)
library(effects)
library(carData)
library(Ecdat)
library(huxtable)
library(flair)
```

2 Begin lab 2 exercise

Read the dataframe into your R-environment from package {Ecdat}

```
data(Vocab)

vocab <- as.data.frame(Vocab)</pre>
```

Take a look at focal variables, make a tribble table

Name	Labels
year	Year of the survey
sex	Sex of the respondent (Female or Male)
education	Education, in years
vocabulary	Vocabulary test score: number correct on a 10-word test

check some basic descriptives with the {gtsummary} package

Characteristic	$N = 30351^1$
year	1995 (13)
sex	
Female	17148 (56%)
Male	13203~(44%)
education	13.0(3.0)
vocabulary	6.00(2.12)

¹Statistics presented: mean (SD); n (%)

Run a regression of the model with interaction of sex and education using the lm function

$$\begin{array}{ll} m1_interact <- lm(formula = \ & vocabulary \sim sex + \ \\ + \ & sex:education \;, \; \ & data = vocab) \end{array}$$

education

Print summary of regression output using package {huxtable}

```
huxreg(m1_interact, error_pos = 'right')
```

	(1)	
(Intercept)	1.788 ***	(0.064)
sexMale	-0.277 **	(0.094)
education	0.329 ***	(0.005)
sexMale:education	0.008	(0.007)
N	30351	` ` `
R2	0.231	
logLik	-61825.557	
AIC	123661.113	

^{***} p < 0.001; ** p < 0.01; * p < 0.05.

Print the Latex syntax for the regression equation using the package {equatiomatic}

```
extract_eq(m1_interact)
```

```
vocabulary = \alpha + \beta_1(\text{sex}_{\text{Male}}) + \beta_2(\text{education}) + \beta_3(\text{sex}_{\text{Male}} \times \text{education}) + \epsilon
```

```
extract_eq(m1_interact, use_coefs = TRUE)
```

```
vocabulary = 1.79 - 0.28(\text{sex}_{\text{Male}}) + 0.33(\text{education}) + 0.01(\text{sex}_{\text{Male}} \times \text{education}) + \epsilon
```

Plot the interaction effect using the package {effects}

```
int <- effect("sex:education",m1_interact)

plot(int, main="", grid=TRUE,
    x.var = "education", xlab="Education",
    ylab="Vocabulary", multiline=TRUE, confint=list(style="auto"))</pre>
```

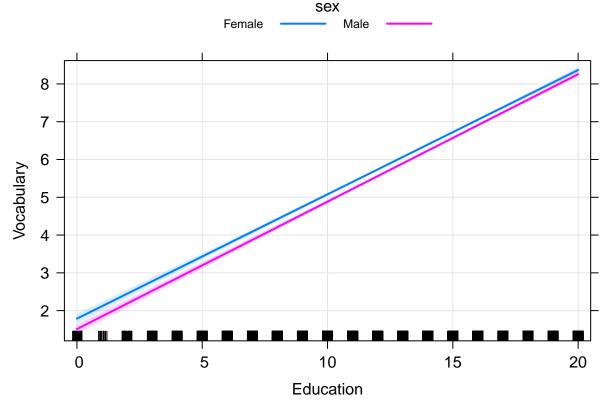


Figure. Example 1 interaction is non-significant (no moderation)

Filter year to include 1974 and 2016 (emphasizing moderation effect)

```
vocab2 <- vocab %>%
filter(year %in% c(1974, 2016)) %>%
mutate(year = droplevels(factor(year)))
```

Run regression with interaction between year and education

	(1)	
(Intercept)	1.600 ***	(0.184)
sexMale	-0.073	(0.062)
education	0.375 ****	(0.015)
year2016	0.373	(0.272)
education:year2016	-0.078 ***	(0.021)
N	3307	
R2	0.244	
logLik	-6607.022	
AIC	13226.043	

^{***} p < 0.001; ** p < 0.01; * p < 0.05.

Print the Latex syntax for the regression equation using the package {equatiomatic}

```
extract_eq(m2_interact)
```

```
\label{eq:condition} {\rm vocabulary} = \alpha + \beta_1({\rm sex_{Male}}) + \beta_2({\rm education}) + \beta_3({\rm year_{2016}}) + \beta_4({\rm education} \times {\rm year_{2016}}) + \epsilon {\rm extract\_eq(m2\_interact,\ use\_coefs\ =\ TRUE)}
```

```
vocabulary = 1.6 - 0.07(\text{sex}_{\text{Male}}) + 0.38(\text{education}) + 0.37(\text{year}_{2016}) - 0.08(\text{education} \times \text{year}_{2016}) + \epsilon
```

Plot the interaction effect using the package {effects}

```
int2 <- effect("education:year", m2_interact)

plot(int2, main="", grid=TRUE,
    x.var = "education", xlab="Education",
    ylab="Vocabulary",
    multiline=TRUE,
    confint=list(style="auto"))</pre>
```

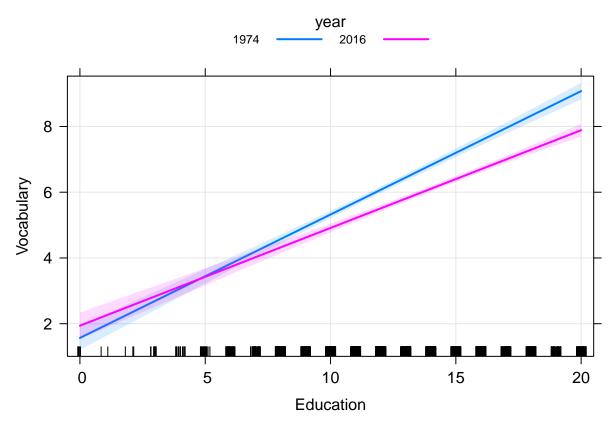
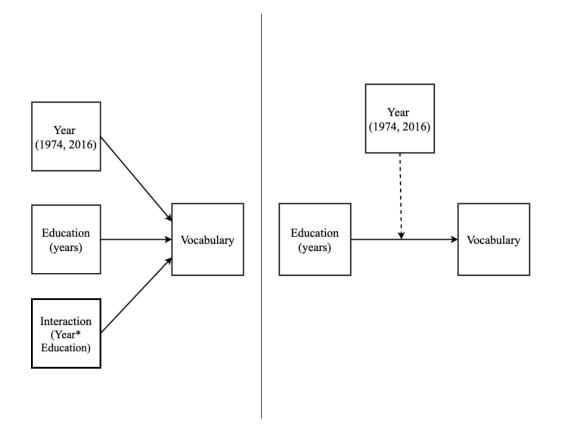


Figure. Example 2 interaction term is significant (moderation)

2.1 Estimate moderation example 1

- 1. covariate: Years of education (education)
- 2. moderator: Year of the survey with 2-levels 1974 and 2016 (year)
- 3. outcome: Vocabulary test score number correct on a 10-word test (vocabulary)



```
education_2sd <- 2*sqrt(9.85) # 6.28
m1_model <- mplusObject(</pre>
 TITLE = "m5 model indirect - Lab 3",
 VARIABLE =
  "usevar =
                    ! covariate/moderator
   year
   education
                    ! covariate
   vocabulary
                    ! outcome
   int_yred;
                    ! interaction of year and education",
 DEFINE =
   "center education (grandmean);
    int_yred = year*education; ! create interaction term ",
  ANALYSIS =
    "estimator = MLR" ,
 MODEL =
  "[vocabulary](b0);
   vocabulary on
   year(b1)
   education(b2)
   int_yred(b3); " ,
```

```
MODELCONSTRAINT =
  "LOOP(x,-1,1,0.01);
  PLOT(y1974 y2016);
  new(hi_y1974 lo_y1974 hi_y2016 lo_y2016 diff_hi);
  y1974 = b0 + b2*x;
  y2016 = b0 + b1 + (b2+b3)*x;
  hi_y1974 = b0 + b2*(6.28);
  lo y1974 = b0 + b2*(-6.28);
  hi_y2016 = b0 + b1 + (b2 + b3)*(6.28);
  lo_y2016 = b0 + b1 + (b2 + b3)*(-6.28);
   diff_hi = hi_y2016 - hi_y1974; ",
  OUTPUT = "sampstat standardized modindices (3.84)",
  PLOT = "type=plot3;",
  usevariables = colnames(vocab2),
 rdata = vocab2)
m1_model_fit <- mplusModeler(m1_model,</pre>
                     dataout=here("mplus_files", "Lab3.dat"),
                    modelout=here("mplus_files", "model1_Lab3.inp"),
                    check=TRUE, run = TRUE, hashfilename = FALSE)
```

2.2 Create the simple slope plot from Mplus model output

Extract the output parameters generated using the model constraint

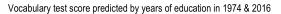
```
simp_slope <- data.frame(m1_model_fit[["results"]][["parameters"]][["unstandardized"]]) %>%
   filter(paramHeader == "New.Additional.Parameters") %>%
   filter(param!= "DIFF_HI") %>%
   select(param, est, se) %>%
   mutate(year = case_when(
      param %in% c("HI_Y1974", "LO_Y1974") ~ "1974",
      param %in% c("HI_Y2016", "LO_Y2016") ~ "2016")) %>%
   mutate(education = case_when(
      param %in% c("HI_Y1974", "HI_Y2016") ~ 9,
      param %in% c("IO_Y1974", "LO_Y2016") ~ -9))
```

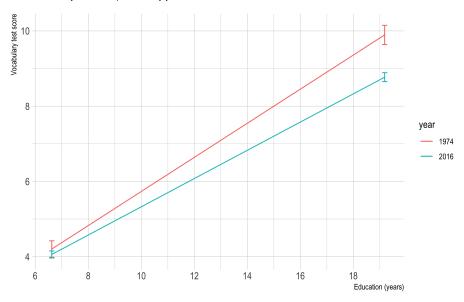
Plot the interaction effect with ggplot using theme from {hrbrthemes} package

```
# un-center 'education' so values on x-axis are on the original scale
plot_data <- simp_slope %>% mutate(education = education + 12.9)

ggplot(plot_data, aes(x=education, y=est, color=year, group=year)) +
    geom_point(size=0) +
```

Simple Slopes Graph





2.3 Data source for example 2:

The next example utilizes the Effects on Learning of Small Class Sizes (Star) dataset from the Introduction to Econometrics textbook. (Stock et al., 2003) See documentation here

This dataset is available via the R-package {Ecdat} and can be directly loaded into the R environment.

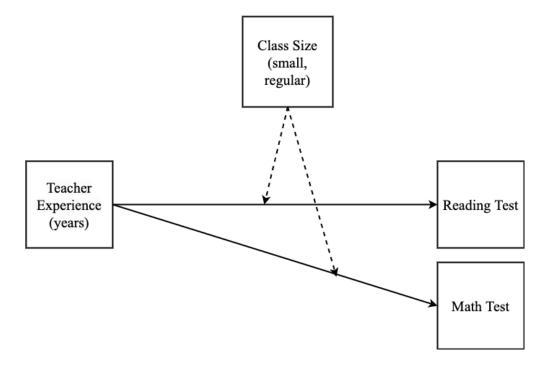
Read the dataframe into your R-environment from package {Ecdat}

```
data(Star)
star_data <- as.data.frame(Star)</pre>
```

Take a look at the variables in the Star dataset

Name	Labels
tmathssk	total math scaled score
treadssk	total reading scaled score
classk	type of class (small, regular, regular with aide)
totexpk	years of total teaching experience

Subset and recode variables to use in moderation model with select, mutate, and case_when



2.4 Estimate moderation example 2

- 1. covariate: Years of education (totexpk)
- 2. moderator: type of class (small, regular) (classk)
- 3. outcome 1: total math scaled score (tmathssk)
- 4. outcome 2: total reading scaled score (treadssk)

```
teach_exp_2sd <- sqrt(33.261) # 5.77
m2_model <- mplusObject(</pre>
 TITLE = "m2 model indirect - Lab 3",
  VARIABLE =
  "usevar =
   totexpk classk
   tmathssk, treadssk
   tchXclas; ",
 DEFINE =
    "center totexpk (grandmean);
    tchXclas = totexpk*classk; ! create interaction term" ,
  ANALYSIS =
    "estimator = mlr; ",
  MODEL =
   "treadssk on classk totexpk tchXclas;
    [tmathssk](b0);
   tmathssk on
   classk (b1)
   totexpk (b2)
   tchXclas (b3); ",
  MODELCONSTRAINT =
  "LOOP(x,-1,1,0.01);
  PLOT(small regular);
  new(hi_small lo_small hi_regular lo_regular diff_hi);
     small = b0 + b2*x;
  regular = b0 + b1 + (b2+b3)*x;
    hi_small = b0 + b2*(9.3);
    lo_small = b0 + b2*(-9.3);
  hi_regular = b0 + b1 + (b2 + b3)*(9.3);
  lo_regular = b0 + b1 + (b2 + b3)*(-9.3);
      diff_hi = hi_small - hi_regular; ",
  OUTPUT = "sampstat standardized modindices (3.84)",
 PLOT = "type=plot3;",
 usevariables = colnames(mod_data),
 rdata = mod_data)
m2_model_fit <- mplusModeler(m2_model,</pre>
                     dataout=here("mplus_files", "Lab3_caschools.dat"),
                    modelout=here("mplus_files", "model2_Lab2.inp"),
                    check=TRUE, run = TRUE, hashfilename = FALSE)
```

2.5 Create the simple slope plot from Mplus model output

```
simp slope2 <- data.frame(m2 model fit[["results"]][["parameters"]][["unstandardized"]]) %>%
  filter(paramHeader == "New.Additional.Parameters") %>%
  filter(param!= "DIFF_HI") %>%
  select(param, est, se) %>%
  mutate(size = case when(
     param %in% c("HI_SMALL", "LO_SMALL") ~ "Small",
     param %in% c("HI_REGUL", "LO_REGUL") ~ "Regular")) %>%
  mutate(experience = case_when(
     param %in% c("HI_SMALL", "HI_REGUL") ~ 9.3,
     param %in% c("LO_SMALL", "LO_REGUL") ~ -9.3))
\# un-center 'experience' so values on x-axis are on the original scale
mean_exp <- mean(mod_data$totexpk)</pre>
plot_data2 <- simp_slope2 %>% mutate(experience = experience + mean_exp)
ggplot(plot_data2, aes(x=experience, y=est, color=size, group=size)) +
 geom_point(size=0) +
  geom line() +
  geom_errorbar(aes(ymin=est-se, ymax=est+se), width=.25) +
  scale_x_continuous(breaks = c(seq(0,18,2))) +
  labs(title = "Simple Slopes Graph",
       subtitle = "Math test score predicted by years of teaching experience in small & regular classro
           x = "Teaching Experience (years)",
           y = "Math test score") +
  theme_ipsum()
ggsave(here("figures", "m2_simple_slope.png"), height = 6, width = 8)
```

Simple Slopes Graph

Math test score predicted by years of teaching experience in small & regular classrooms

