

Lab 6 - Latent Growth Models

Structural Equation Modeling - 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-Lab6>

On the Github repository webpage:

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

Within R-Studio:

- c. click “NEW PROJECT”
 - 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 sources:

1. The first 3 models utilize a public use data subset the *Longitudinal Survey of American Youth (LSAY)*
[See documentation here](#)
 2. The 4th model utilizes a public use data subset the *High School Longitudinal Study (HSLS)*
[See documentation here](#)
-

Load packages

```
library(gganimate)
library(hrbrthemes)
library(tidyverse)
library(haven)
library(janitor)
library(MplusAutomation)
library(rhdf5)
library(here)
library(kableExtra)
library(gtsummary)
library(semPlot)
```

1.3 LSAY data example - Math Scores across 6 timepoints

Read in data

```
lsay_data <- read_spss(here("data", "LSAY_Lab6.sav")) %>% select(-starts_with("AB"),
  ends_with("IMP"), -contains("BIO"), -contains("PHY"), -contains("SCI"), FATHED,
  MOTHEDED) %>% clean_names() %>% rename(math_07 = amthimp, math_08 = cmthimp, math_09 = emthimp,
  math_10 = gmthimp, math_11 = imthimp, math_12 = kmthimp)

lsay_data[lsay_data == 9999] <- NA
```

View metadata

```
sjPlot::view_df(lsay_data)
```

Write a CSV file

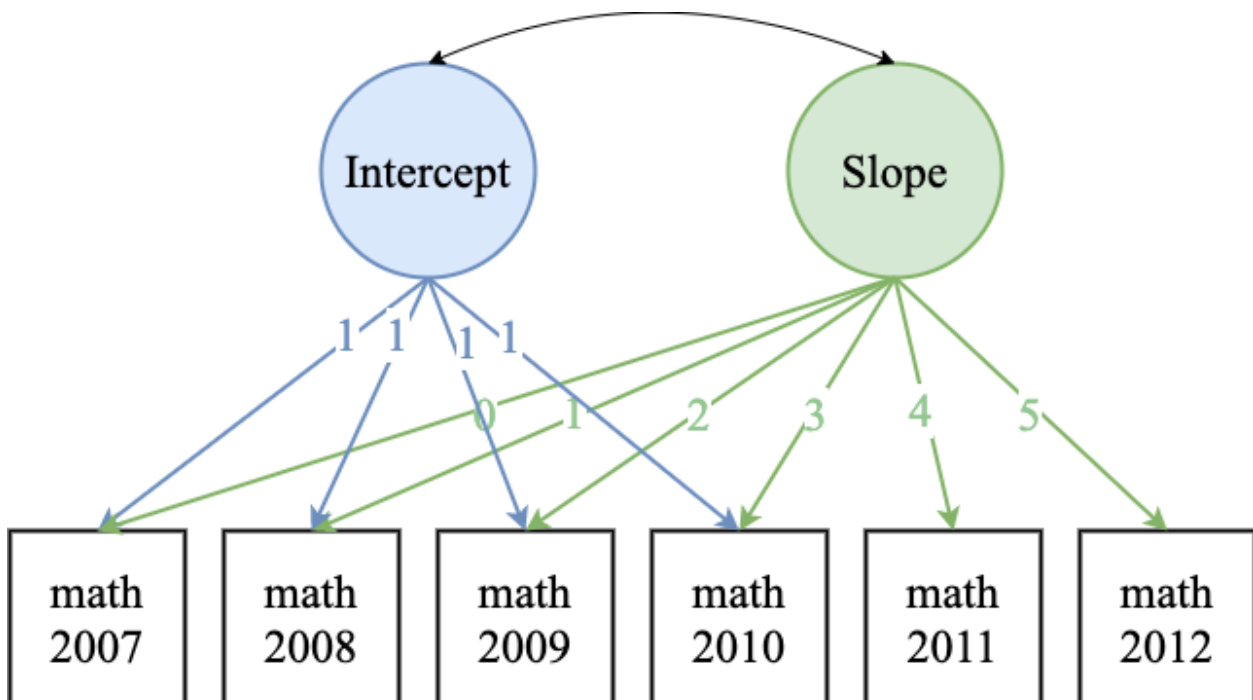
```
write_csv(lsay_data, here("data", "lsay_lab6_data.csv"))
```

Read in the CSV file (SPSS labels removed)

```
lsay_lab6 <- read_csv(here("data", "lsay_lab6_data.csv"))
```

Table. LSAY repeated measures

Name	Labels
math_07	7th grade math score (imputed)
math_08	8th grade math score (imputed)
math_09	9th grade math score (imputed)
math_10	10th grade math score (imputed)
math_11	11th grade math score (imputed)
math_12	12th grade math score (imputed)



1.4 Model 1 - Latent growth model with fixed time effects (equal intervals)

```
m1_growth <- mplusObject(
  TITLE = "m1 growth model fixed time scores - Lab 6",
  VARIABLE =
    "usevar =
    math_07-math_12; ",
  ANALYSIS =
    "estimator = ML" ,
  MODEL =
    "i s | math_07@0 math_08@1 math_09@2 math_10@3 math_11@4 math_12@5; " ,
  OUTPUT = "sampstat standardized;",
  PLOT = "type=plot3;
    series = math_07-math_12(*)",
  usevariables = colnames(lsay_lab6),
  rdata = lsay_lab6)

m1_growth_fit <- mplusModeler(m1_growth,
  dataout=here("mplus_files", "Lab6.dat"),
  modelout=here("mplus_files", "m1_growth_Lab6.inp"),
  check=TRUE, run = TRUE, hashfilename = FALSE)
```

Load in the mplus.R functions

```
source(here("mplus.R.txt"))
```

```
## [1] "Loaded rhdf5 package"
```

1.5 Plotting using gh5 plot data generated by Mplus

1. View plots available for a given model
2. Generate plots using the `get.plot.---` function
3. Extract data and transform to tidy format
4. Plot with `ggplot`

```
mplus.view.plots(here("mplus_files", "m1_growth_Lab6.gh5"))
```

Prepare plot data

```

observed <- lsay_lab6 %>% select(starts_with("math")) %>%
  rownames_to_column() %>% drop_na()

obs100 <- observed[1:100,]

plot_obs <- obs100 %>%
  pivot_longer(`math_07`:`math_12`, # The columns I'm gathering together
              names_to = "grade", # new column name for existing names
              values_to = "value") # new column name to store values

gradelevels <- colnames(observed[,2:7])

mean_est <- as.data.frame(mplus.get.estimated_means(here("mplus_files", "m1_growth_Lab6.gh5"))) %>%
  mutate(grade = gradelevels)

```

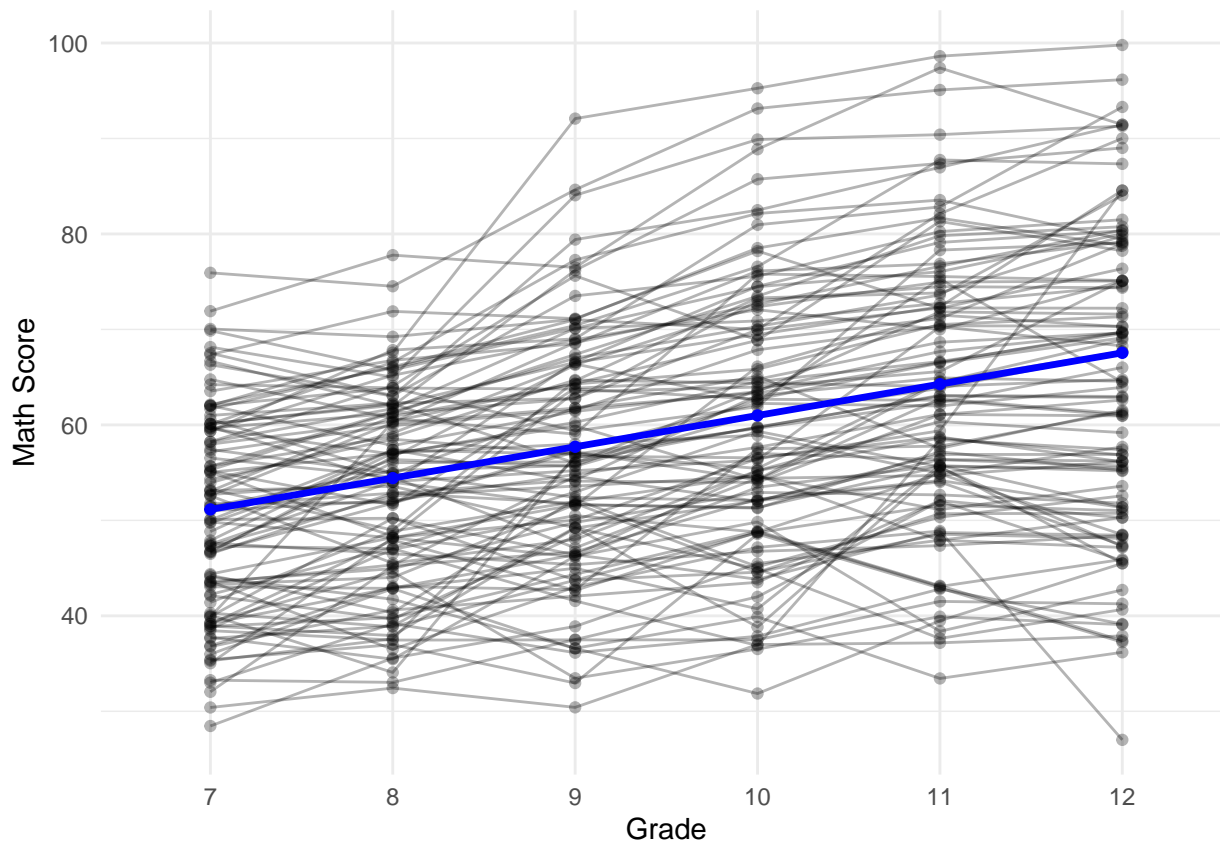
Plot the model estimated means superimposed on the observed individual values

```

growth_plot <- ggplot() +
  geom_point(data = plot_obs, aes(x = grade, y = value, group = rowname), alpha = .3) + #
  geom_line(data = plot_obs, aes(x = grade, y = value, group = rowname), alpha = .3) + #
  geom_point(data=mean_est, aes(x=grade, y = V1), color = "Blue", size = 1.5) + #
  geom_line(data=mean_est, aes(x=grade, y = V1, group = 1), color = "Blue", size = 1.2) + #
  scale_x_discrete(labels = c("7", "8", "9", "10", "11", "12")) + #
  labs(x="Grade", y="Math Score") + #
  theme_minimal()

growth_plot

```



```
ggsave(here("figures", "spaghetti_p1.png"), height = 6, width = 8, dpi = "retina")
```

Animate the plot with {gganimate}

```
growth_plot + transition_states(rowname, transition_length = 1, state_length = 1) + #
  shadow_mark(color = "Magenta", alpha = .3) #
```

```
anim_save(here("figures", "spaghetti_plot.gif"), height = 6, width = 8, dpi = "retina")
```

1.6 Model 2 - Latent growth model with freely estimated time scores (level-shape model or latent basis model)

```
m2_growth <- mplusObject(
  TITLE = "m2 growth model freely estimated time scores - Lab 6",
  VARIABLE =
    "usevar =
    math_07-math_12; ",
  ANALYSIS =
    "estimator = ML" ,
```

```

MODEL =
  "i s | math_07@0 math_08@1 math_09* math_10* math_11* math_12*; " ,

OUTPUT = "sampstat standardized;",

PLOT = "type=plot3;
        series = math_07-math_12(*)",

usevariables = colnames(lsay_lab6),
rdata = lsay_lab6)

m2_growth_fit <- mplusModeler(m2_growth,
                             dataout=here("mplus_files", "Lab6.dat"),
                             modelout=here("mplus_files", "m2_growth_Lab6.inp"),
                             check=TRUE, run = TRUE, hashfilename = FALSE)

```

Prepare plot data

```

mean_est2 <- as.data.frame(mplus.get.estimated_means(here("mplus_files", "m2_growth_Lab6.gh5"))) %>%
  mutate(grade = gradelevels)

```

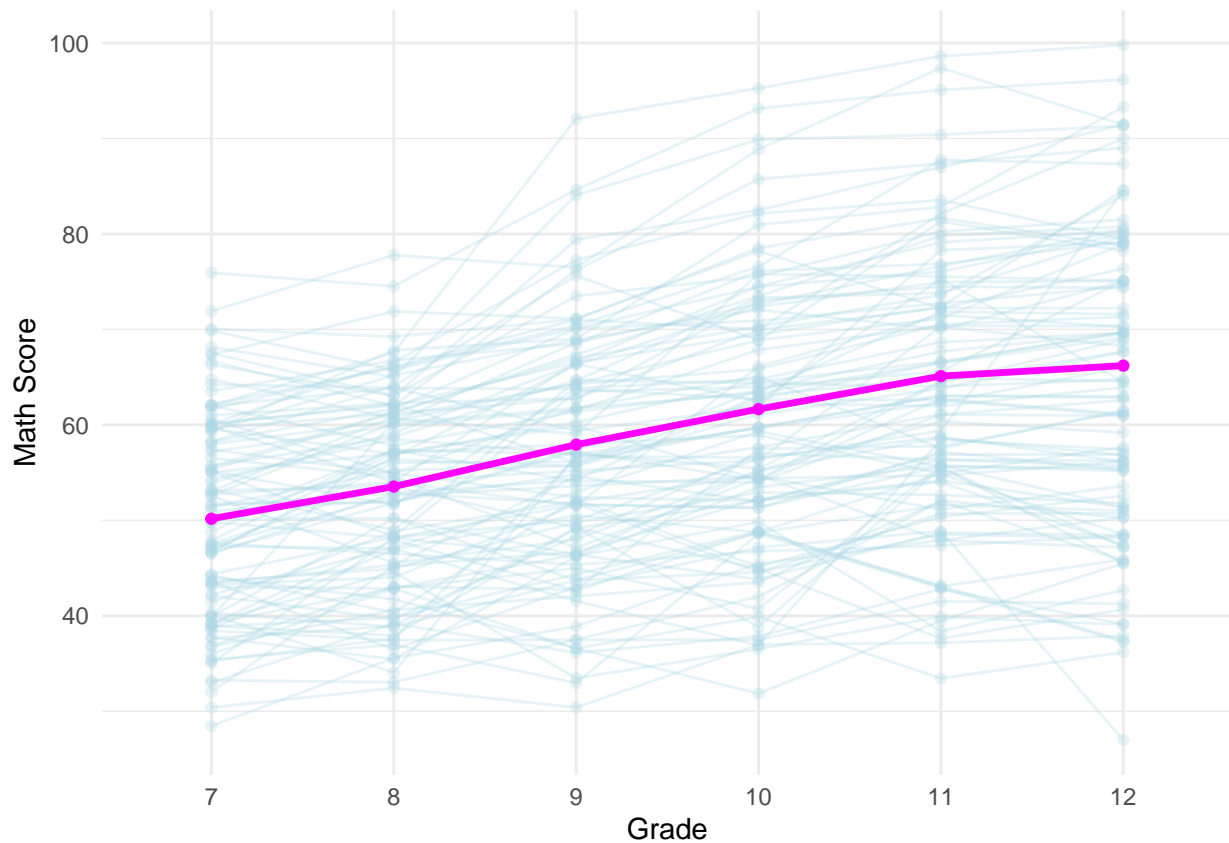
Plot the model estimated means superimposed on the observed individual values

```

growth_plot <- ggplot() +
  geom_point(data = plot_obs, aes(x = grade, y = value, group = rowname), color = "lightblue", alpha = .5) +
  geom_line(data = plot_obs, aes(x = grade, y = value, group = rowname), color = "lightblue", alpha = .5) +
  geom_point(data=mean_est2, aes(x=grade, y = V1), color = "magenta", size = 1.5) +
  geom_line(data=mean_est2, aes(x=grade, y = V1, group = 1), color = "magenta", size = 1.2) +
  scale_x_discrete(labels = c("7", "8", "9", "10", "11", "12")) +
  labs(x="Grade", y="Math Score") +
  theme_minimal()

growth_plot

```



1.7 Model 3 - Latent growth model with covariate and freely estimated time scores

```

m3_growth <- mplusObject(
  TITLE = "m3 growth model with covariate and freely estimated time scores - Lab 6",
  VARIABLE =
    "usevar =
    math_07-math_12 fathed; ",

  ANALYSIS =
    "estimator = ML" ,

  DEFINE = "center fathed (grandmean);",

  MODEL =
    "i s | math_07@0 math_08@1 math_09* math_10* math_11* math_12*;
    i s on fathed; " ,

  OUTPUT = "sampstat standardized;",

  PLOT = "type=plot3;

```



```

series = math_07-math_12(*)",

usevariables = colnames(lsay_lab6),
rdata = lsay_lab6)

m3_growth_fit <- mplusModeler(m3_growth,
  dataout=here("mplus_files", "Lab6.dat"),
  modelout=here("mplus_files", "m3_growth_Lab6.inp"),
  check=TRUE, run = TRUE, hashfilename = FALSE)

```

Check the path diagram of the model with {semPlot}

```

m3_output <- readModels(here("mplus_files", "m3_growth_Lab6.out"))

```

```

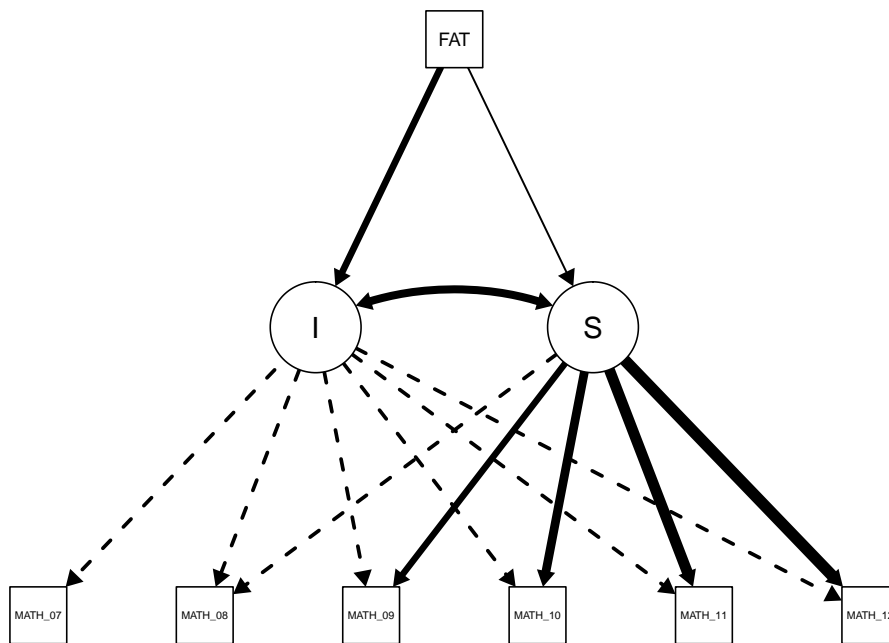
## Reading model: /Users/agarber/Desktop/SEM_S20/Lab6_SEM/mplus_files/m3_growth_Lab6.out

```

```

semPaths(m3_output, "est", intercepts = FALSE, residuals = FALSE, fade = FALSE, edge.color = "black",
  edgeLabels = "")

```



1.8 HSLS data example - Academic expectations

```
hsls_rep <- read_csv(here("data", "hsls_rep_lab6.csv"))
```

Table. HSLs repeated measures

Question stem - Highest level of education expected...

Name	Labels	Levels
s1eduexp	9th grade (2009)	1 = less HS, 2 = HS, 3 = Bach, 5 = Master, 6 = Ph.D
s2eduexp	11th grade (2012)	1 = less HS, 2 = HS, 3 = Bach, 5 = Master, 6 = Ph.D
s4eduexp	3 years post high school (2016)	1 = less HS, 2 = HS, 3 = Bach, 5 = Master, 6 = Ph.D

1.9 Model 4 - Latent growth model with categorical outcomes

```
m4_growth <- mplusObject(  
  TITLE = "m4 growth model - HSLs - Lab 6",  
  
  VARIABLE =  
    "usevar = s1eduexp-s4eduexp;  
    categorical = s1eduexp-s4eduexp;",  
  
  ANALYSIS = "" ,  
  
  MODEL =  
    "! 0=09 1=10 2=11 3=12 | 4=13 5=14 6=15 7=16  
  
    i s | s1eduexp@0 s2eduexp@3 s4eduexp@7; ",  
  
  OUTPUT = "sampstat standardized;",  
  
  PLOT = "type=plot3;  
    series = s1eduexp-s4eduexp(*)",  
  
  usevariables = colnames(hsls_rep),  
  rdata = hsls_rep)  
  
m4_growth_fit <- mplusModeler(m4_growth,  
  dataout=here("mplus_files", "Lab6.dat"),  
  modelout=here("mplus_files", "m4_growth_Lab6.inp"),  
  check=TRUE, run = TRUE, hashfilename = FALSE)
```

Prepare plot data

```

loop_data <- lapply(1:6, function(k) {
  probs <- mplus.get.estimated_probabilities(here("mplus_files", "m4_growth_Lab6.gh5"), 'process1', k, k)

  loop_data <- as.data.frame(probs) %>%
    mutate(cat = factor(k))
})

plot_data <- bind_rows(loop_data)

observed <- hsls_rep %>% select(contains("eduexp")) %>%
  rownames_to_column() %>% drop_na()

obs100 <- observed[1:100,]

plot_obs <- obs100 %>%
  pivot_longer(`s1eduexp`:`s4eduexp`, # The columns I'm gathering together
              names_to = "year",      # new column name for existing names
              values_to = "value") %>% # new column name to store values
  mutate(year = case_when(
    year == "s1eduexp" ~ 1,
    year == "s2eduexp" ~ 2,
    year == "s4eduexp" ~ 3,
  ))

yearlevels <- colnames(observed[,2:4])

prob_est <- plot_data %>%
  mutate(year = rep(1:3, 6))

```

Plot the model estimated probabilities (categorical outcomes)

```

ggplot(data=prob_est, aes(x=year, y=V1, fill=cat)) + #
  geom_area(alpha=0.3 , size=.4, colour="black") + #
  scale_x_continuous(breaks = 1:3, #
                    labels = c("9th grade (2009)", "11th grade (2012)", "3 years post-HS (2016)")) + #
  scale_y_continuous("Probability") + #
  scale_fill_discrete("", #
                    labels = c("< High School", "High School", "Associates", "Bachelor", "Masters", "Ph.D")) + #
  labs(title="Highest level of education expected", #
       subtitle = "High School Longitudinal Study (N=21,758)", y="Probability", x="") + #
  theme_minimal() #

```

Highest level of education expected

High School Longitudinal Study (N=21,758)



```
ggsave(here("figures", "cat_growth_plot.png"), height = 6, width = 8, dpi = "retina")
```

Create an animated plot with {gganimate}

```
cat_plot <- ggplot(data = plot_obs, aes(x = year, y = value, group = rowname)) + #
  geom_jitter(color = "black", alpha = 0, width = 0.1, height = .3) + #
  geom_line(color = "black") + #
  scale_x_continuous(breaks = 1:3, #
    labels = c("9th grade (2009)", "11th grade (2012)", "3 years post-HS (2016)")) + #
  scale_y_reverse(breaks = 1:6, labels = c("< HS", "HS", "AA", "BA/BS", "MA", "Ph.D")) + #
  theme_minimal() + theme(panel.grid.minor = element_blank()) + #
  labs(y="", title="Highest level of education expected", #
    subtitle = "High School Longitudinal Study (N=100, sub-sample)") #

cat_plot + transition_states(rowname, transition_length = 3, state_length = 3) + #
  shadow_mark(color = "blue", alpha = .15) #
```

```
anim_save(here("figures", "cat_growth_anim.gif"), height = 6, width = 8, dpi = "retina")
```

2 References

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