# Competing Path Models

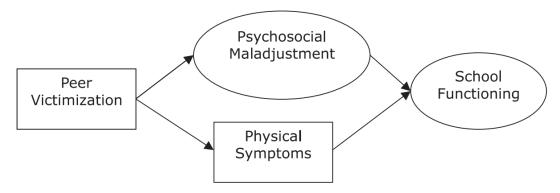
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### Preparation

Example of competing path models study from Nishina, Juvonen, Witkow (2005)



Model 1: Peer harassment as stressor

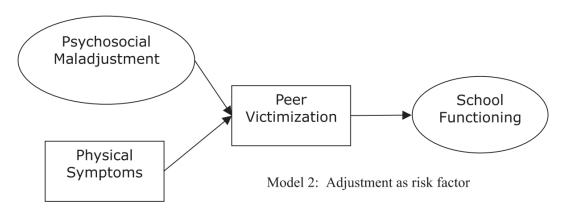


figure. Picture adapted from Nishina, Juvonen, Witkow (2005)

#### Data source:

This lab exercise utilizes the *California Test Score Data Set 1998-1999* from the California Department of Education (Stock, James, and Watson, 2003) See documentation here

This dataset is available via the R-package {Ecdat} 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.

Load packages

```
library(MplusAutomation)
library(haven)
library(tidyverse)
library(here)
library(corrplot)
library(reshape2)
library(janitor)
library(ggridges)
library(DiagrammeR)
library(semPlot)
library(setlat)
library(gt)
```

#### Change starting location to folder 13-comp-models

```
source("rep_functions.R")

change_here(glue("{project_location}/13-comp-models"))

here()

## [1] "/Users/agarber/github/NTNU-workshop/13-comp-models"

Begin lab 2 exercise

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

data(Caschool)

ca_schools <- as.data.frame(Caschool)
```

Look at the data with glimpse

```
glimpse(ca_schools)
```

Subset variables to use in path model analyses with select

#### Explore the data

K through 8th grade schools in California (N=420)

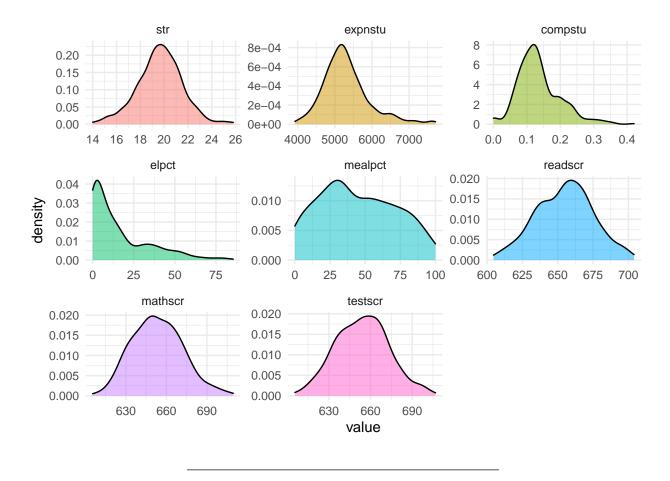
Take a look at focal variables, make a tribble table

Name	Labels
str expnstu compstu elpct mealpct readscr mathscr	student teacher ratio expenditure per student computer per student percent of English learners percent qualifying for reduced-price lunch average reading score average math score
testscr	average math score average test score (read.scr+math.scr)/2

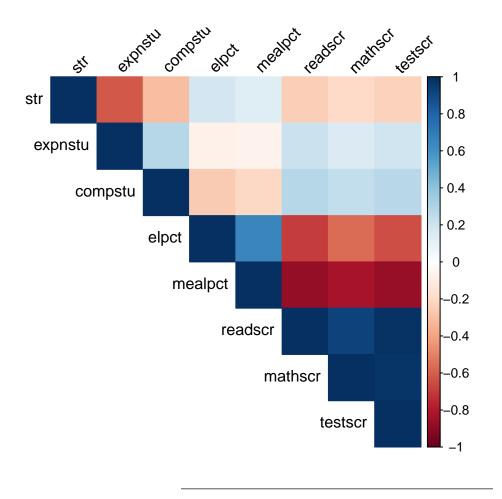
check some basic descriptives with the {gtsummary} package

Characteristic	N = 420
${\mathrm{str}}$	19.64 (1.89)
expnstu	5312 (634)
compstu	0.14(0.06)
elpct	16 (18)
mealpct	45(27)
readscr	655(20)
mathscr	653 (19)
testscr	654 (19)

look at shape of variable distributions



look at correlation matrix with  $\{{\tt corrplot}\}$ 



# Specifying path models using {MplusAutomation}

recall what the unrestricted variance-covariance matrix  ${\bf looks}$  like

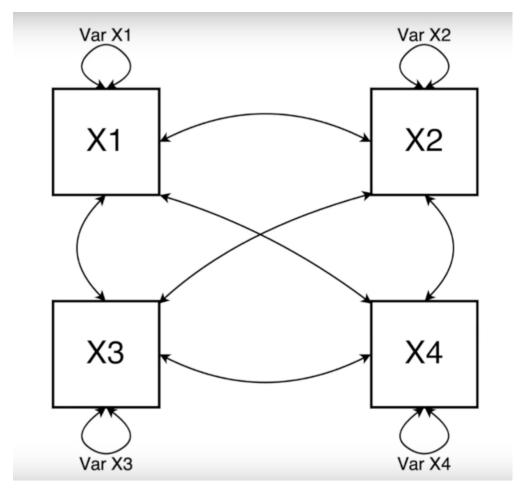


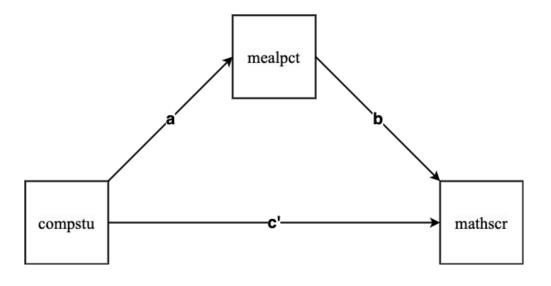
figure. Unrestricted variance covariance matrix picture from {openMX} video tutorial.

# Estimate model 1

Indirect path model:

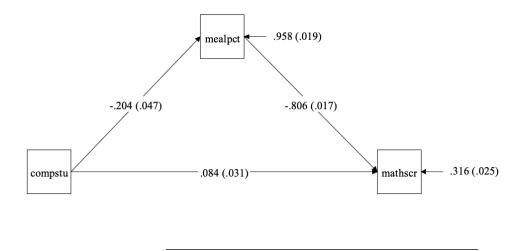
- 1. covariate: ratio of computers to students (compstu)
- 2. **mediator**: percent qualifying for reduced-price lunch (mealpct)
- 3. outcome: average math score (mathscr)

Path diagram model 1



```
m1_path <- mplusObject(</pre>
  TITLE = "m1 model indirect - Lab 1",
  VARIABLE =
   "usevar =
   compstu
                   ! covariate
                ! mediator
   mealpct
    mathscr;
                  ! outcome",
  ANALYSIS =
    "estimator = MLR" ,
  MODEL =
   "mathscr on compstu; ! direct path (c')
mathscr on mealpct; ! b path
   mathscr on mealpct;
                               ! b path
    mealpct on compstu;
                               ! a path
   Model indirect:
    mathscr ind compstu;",
  OUTPUT = "sampstat standardized modindices (ALL)",
  usevariables = colnames(path_vars),
  rdata = path_vars)
m1_path_fit <- mplusModeler(m1_path,</pre>
               dataout=here("mplus_files", "CA_schls.dat"),
               modelout=here("mplus_files", "m1_path_indirect.inp"),
               check=TRUE, run = TRUE, hashfilename = FALSE)
```

View path diagram for model 1 with standardized estimates (using Diagrammer in Mplus)

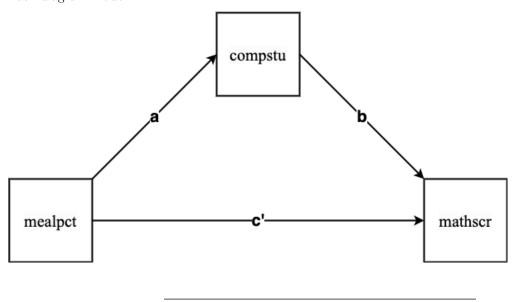


#### Estimate model 2

change variable status (**switch mediator and covariate variables**) Indirect path model:

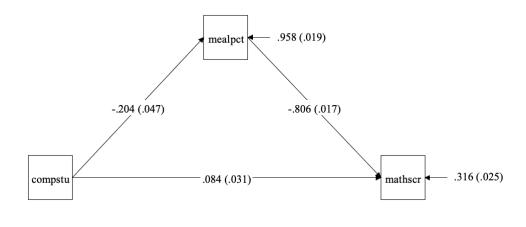
- 1. covariate: percent qualifying for reduced-price lunch (mealpct)
- 2. mediator: ratio of computers to students (compstu)
- 3. outcome: average math score (mathscr)

Path diagram model 2



```
m2_path <- mplusObject(</pre>
  TITLE = "m2 model indirect",
  VARIABLE =
   "usevar =
   mealpct
                     ! covariate
   compstu
                      ! mediator
   mathscr;
                     ! outcome",
  ANALYSIS =
    "estimator = MLR" ,
  MODEL =
                            ! direct path (c')
! b path
   "mathscr on compstu;
   mathscr on mealpct;
   mealpct on compstu;
                               ! a path
   Model indirect:
    mathscr ind compstu;",
  OUTPUT = "sampstat standardized modindices (ALL)",
  usevariables = colnames(path_vars),
  rdata = path_vars)
m2_path_fit <- mplusModeler(m2_path,</pre>
                     dataout=here("mplus_files", "CA_schls.dat"),
                    modelout=here("mplus_files", "m2_path_indirect.inp"),
                    check=TRUE, run = TRUE, hashfilename = FALSE)
```

View path diagram for model 2 with standardized estimates (using the Diagrammer in Mplus)

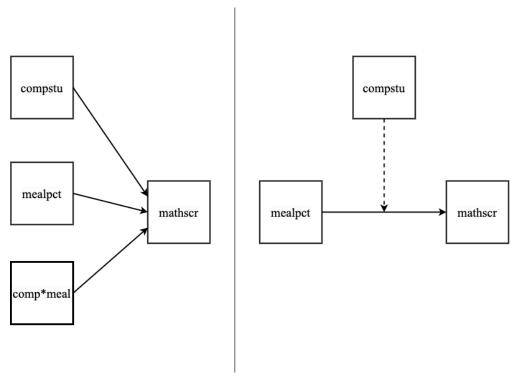


#### Estimate model 3

Path model with interaction (moderation):

- 1. covariate-moderator: percent qualifying for reduced-price lunch (mealpct)
- 2. covariate-moderator: ratio of computers to students (compstu)
- 3. outcome: average math score (mathscr)

Path diagram model 3

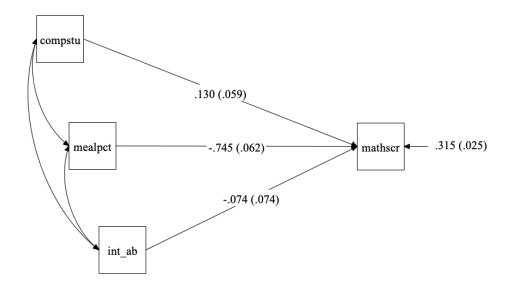


```
m3_path <- mplusObject(
  TITLE = "m3 model interact",
  VARIABLE =
  "usevar =
    compstu     ! covariate-moderator
    mealpct     ! covariate-moderator
    mathscr      ! outcome
    int_ab;     ! interaction term ",

DEFINE =
    "int_ab = compstu*mealpct; ! create interaction term" ,

ANALYSIS =
    "estimator = MLR" ,</pre>
```

View path diagram for model 3 with standardized estimates (using the Diagrammer in Mplus)



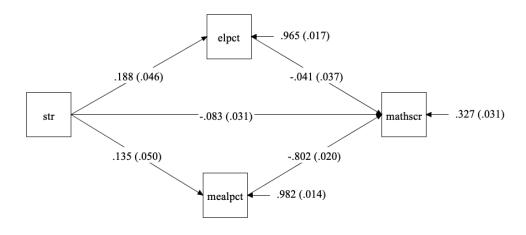
#### Estimate model 4

Two indirect paths or "mediators"

```
m4_path <- mplusObject(
  TITLE = "m4 model indirect (two mediators)",
  VARIABLE =
  "usevar =
   str          ! covariate
   elpct         ! mediator
   mealpct     ! mediator</pre>
```

```
mathscr
                      ! outcome",
  DEFINE =
    "int_ab = compstu*mealpct; ! create interaction term" ,
  ANALYSIS =
    "estimator = MLR" ,
  MODEL =
   "mathscr on str;
                               ! direct path (c')
    mathscr on elpct mealpct; ! b paths
    elpct mealpct on str;
                               ! a paths
    Model indirect:
    mathscr ind str;" ,
  OUTPUT = "sampstat standardized modindices (ALL)",
  usevariables = colnames(path_vars),
  rdata = path_vars)
m4_path_fit <- mplusModeler(m4_path,</pre>
                    dataout=here("mplus_files", "CA_schls.dat"),
                    modelout=here("mplus_files", "m4_path_indirect.inp"),
                    check=TRUE, run = TRUE, hashfilename = FALSE)
```

View path diagram for model 4 with standardized estimates (using the Diagrammer in Mplus)

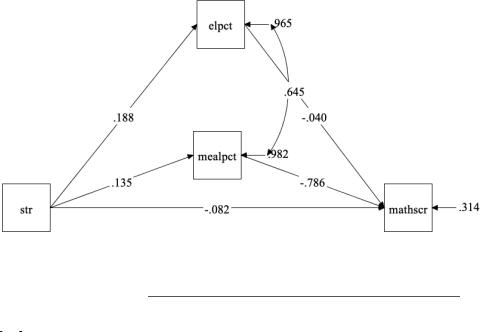


#### Estimate model 5

Add modification statement - correlate mediators mealpct with elpct

```
m5_path <- mplusObject(</pre>
 TITLE = "m5 model indirect (correlate mediators)",
 VARIABLE =
  "usevar =
   str
                   ! covariate
   elpct
                   ! mediator
   mealpct
                    ! mediator
   mathscr
                   ! outcome",
 DEFINE =
   "int_ab = compstu*mealpct; ! create interaction term" ,
 ANALYSIS =
   "estimator = MLR" ,
 MODEL =
  "mathscr on str; ! direct path (c')
   mathscr on elpct mealpct; ! b paths
   elpct mealpct on str;
                             ! a paths
   mealpct with elpct ! modification statement
   Model indirect:
   mathscr ind str; ",
 OUTPUT = "sampstat standardized modindices (ALL)",
 usevariables = colnames(path_vars),
 rdata = path_vars)
m5_path_fit <- mplusModeler(m5_path,</pre>
                   dataout=here("mplus_files", "CA_schls.dat"),
                   modelout=here("mplus_files", "m5_path_indirect.inp"),
                   check=TRUE, run = TRUE, hashfilename = FALSE)
```

View path diagram for model 5 with standardized estimates (using the Diagrammer in Mplus)



## End

#### References

Hallquist, M. N., & Wiley, J. F. (2018). MplusAutomation: An R Package for Facilitating Large-Scale Latent Variable Analyses in Mplus. Structural equation modeling: a multidisciplinary journal, 25(4), 621-638.

 $Horst,\,A.\,\,(2020).\,\,Course\,\,\&\,\,Workshop\,\,Materials.\,\,Git Hub\,\,Repositories,\,https://https://allisonhorst.github.io/Leadings.com/leadings$ 

Ingels, S. J., Pratt, D. J., Herget, D. R., Burns, L. J., Dever, J. A., Ottem, R., . . . & Leinwand, S. (2011). High School Longitudinal Study of 2009 (HSLS: 09): Base-Year Data File Documentation. NCES 2011-328. National Center for Education Statistics.

Muthén, L.K. and Muthén, B.O. (1998-2017). Mplus User's Guide. Eighth Edition. Los Angeles, CA: Muthén & Muthén

R Core Team (2017). R: A language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, Austria. URL http://www.R-project.org/

Wickham et al., (2019). Welcome to the tidy verse. Journal of Open Source Software, 4(43), 1686, https://doi.org/10.21105/joss.01686