

Plotting with Lavaan

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"<https://github.com/emoriebeck/R-tutorials/tree/master/sem>"

Plotting with Lavaan

```
library(lavaan)
library(psych)
library(semPlot)
library(plyr)
library(tidyverse)

data_path <- "https://github.com/longitudinal-data/1-descriptives-and-graphs-emoriebeck/raw/master/Condi
load(url(paste(data_path, "sample.RData", sep = "/")))

sample_dat_wide <- sample_dat %>%
  select(-year, -age, -CESD) %>%
  mutate(age0 = paste("A", age0, sep = ".")) %>%
  spread(key = age0, value = SensSeek) %>%
  full_join(
    sample_dat %>% select(-year, -age, -DemPweight, -SensSeek) %>%
      mutate(age0 = paste("D", age0, sep = ".")) %>%
      spread(key = age0, value = CESD)
  )
```

Time Invariant Predictors

Continuous

```
## linear growth model with a time-varying covariate
model.syntax <- '
  # intercept and slope with fixed coefficients
  i =~ 1*A.0 + 1*A.2 + 1*A.4 + 1*A.6 + 1*A.8 + 1*A.10
  s =~ 0*A.0 + 1*A.2 + 2*A.4 + 3*A.6 + 4*A.8 + 5*A.10

  # regressions
  i ~ DemPweight
```

```

      s ~ DemPweight
,

fit1 <- growth(model.syntax, data=sample_dat_wide, missing = "FIML")
summary(fit1)

```

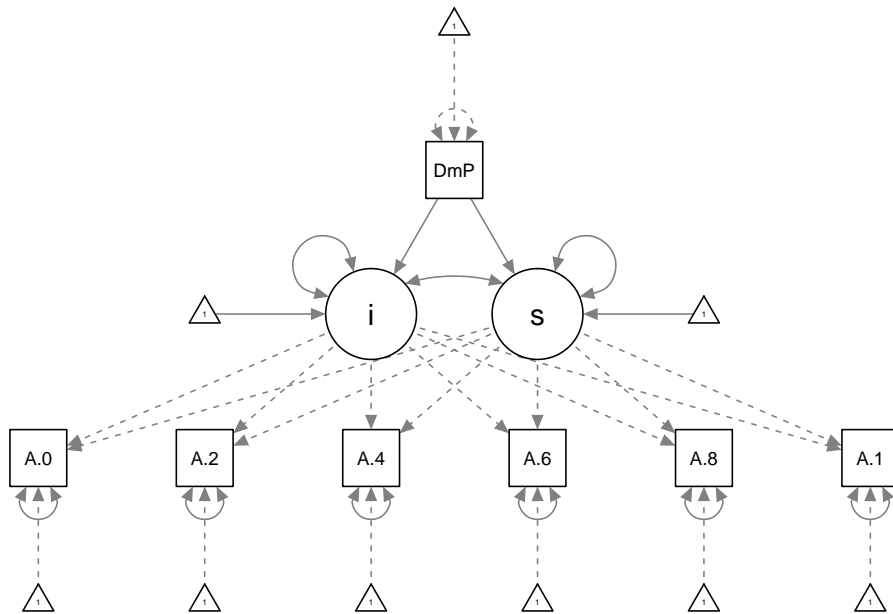
```

## lavaan (0.5-23.1097) converged normally after 38 iterations
##
##   Number of observations              924
##
##   Number of missing patterns         46
##
##   Estimator                          ML
##   Minimum Function Test Statistic    45.354
##   Degrees of freedom                 20
##   P-value (Chi-square)               0.001
##
## Parameter Estimates:
##
##   Information                        Observed
##   Standard Errors                   Standard
##
## Latent Variables:
##
##           Estimate  Std.Err  z-value  P(>|z|)
##   i =~
##     A.0             1.000
##     A.2             1.000
##     A.4             1.000
##     A.6             1.000
##     A.8             1.000
##     A.10            1.000
##   s =~
##     A.0             0.000
##     A.2             1.000
##     A.4             2.000
##     A.6             3.000
##     A.8             4.000
##     A.10            5.000
##
## Regressions:
##
##           Estimate  Std.Err  z-value  P(>|z|)
##   i ~
##     DemPweight      0.049    0.066    0.745    0.456
##   s ~
##     DemPweight     -0.017    0.023   -0.740    0.459
##
## Covariances:
##
##           Estimate  Std.Err  z-value  P(>|z|)
##   .i ~~
##     .s              -0.003    0.006   -0.553    0.581
##
## Intercepts:
##
##           Estimate  Std.Err  z-value  P(>|z|)
##   .A.0              0.000

```

```
##      .A.2          0.000
##      .A.4          0.000
##      .A.6          0.000
##      .A.8          0.000
##      .A.10         0.000
##      .i            2.735    0.047    57.593    0.000
##      .s           -0.001    0.017    -0.071    0.944
##
## Variances:
##              Estimate Std.Err  z-value  P(>|z|)
##      .A.0          0.215    0.025     8.695    0.000
##      .A.2          0.201    0.017    12.109    0.000
##      .A.4          0.222    0.021    10.569    0.000
##      .A.6          0.138    0.019     7.452    0.000
##      .A.8          0.132    0.024     5.479    0.000
##      .A.10         0.131    0.030     4.386    0.000
##      .i            0.129    0.020     6.515    0.000
##      .s            0.005    0.002     2.395    0.017
```

```
semPaths(fit1)
```



```
get_fixef <- function(fit, target.var){
  df <- parameterestimates(fit) %>%
    filter(lhs != rhs & lhs %in% target.var & op %in% c("~", "~1")) %>%
    unite(term, lhs, op, rhs, sep = "")
  v <- as.vector(df$est)
  names(v) <- df$term
  return(v)
}

# example for continuous
fixed.frame <- sample_dat %>%
  group_by(PROC_CID) %>%
  summarise(DemPweight = mean(DemPweight, na.rm = T)) %>%
  ungroup() %>%
```

```

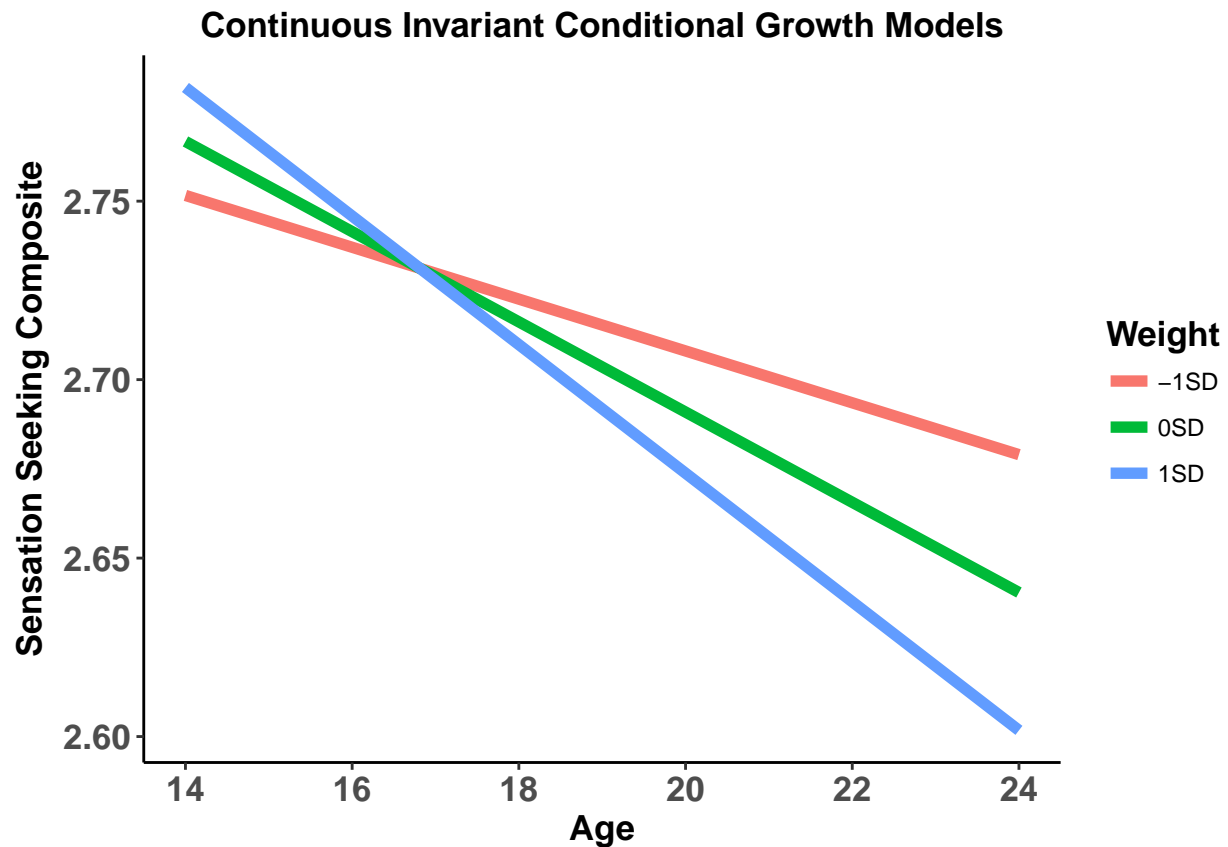
summarise(mean = mean(DemPweight, na.rm = T),
          sd = sd(DemPweight, na.rm = T))

fixed.frame <-
  data.frame(
    expand.grid(
      # here, you add values for your time variable and predictors
      age0 = seq(0,10,2),
      DemPweight = c(fixed.frame$mean-fixed.frame$sd,
                     fixed.frame$mean,
                     fixed.frame$mean+fixed.frame$sd))) %>%
# now take care of interactions and add an intercept
mutate(`age0:DemPweight` = age0*DemPweight,
      Intercept = 1) %>%
# reordering everything
select(DemPweight, `age0:DemPweight`, Intercept, age0)

fixed.frame$y <- as.vector(as.matrix(fixed.frame) %*% get_fixef(fit1, c("i", "s")))

tbl_df(fixed.frame) %>%
  mutate(Weight = factor(DemPweight, levels = unique(DemPweight), labels = c("-1SD", "0SD", "1SD")),
        age = age0 + 14) %>%
  ggplot(aes(x = age, y = y, color = Weight)) +
    geom_line(size = 2) +
    labs(x = "Age", y = "Sensation Seeking Composite",
         title = "Continuous Invariant Conditional Growth Models") +
    theme_classic() +
    theme(axis.text = element_text(face = "bold", size = rel(1.2)),
          axis.title = element_text(face = "bold", size = rel(1.2)),
          legend.title = element_text(face = "bold", size = rel(1.2)),
          plot.title = element_text(face = "bold", size = rel(1.2), hjust = .5))

```



Categorical

```
## linear growth model with a time-varying covariate
model.syntax <- '
  # intercept and slope with fixed coefficients
  i =~ 1*A.0 + 1*A.2 + 1*A.4 + 1*A.6 + 1*A.8 + 1*A.10
  s =~ 0*A.0 + 1*A.2 + 2*A.4 + 3*A.6 + 4*A.8 + 5*A.10

  # regressions
  i ~ groups
  s ~ groups
'

fit2 <- growth(model.syntax, data=sample_dat_wide %>% filter(groups != "CommServ"), missing = "FIML")
summary(fit2)
```

```
## lavaan (0.5-23.1097) converged normally after 55 iterations
##
##   Number of observations              689
##
##   Number of missing patterns          46
##
##   Estimator                           ML
##   Minimum Function Test Statistic     38.403
##   Degrees of freedom                   20
##   P-value (Chi-square)                 0.008
```

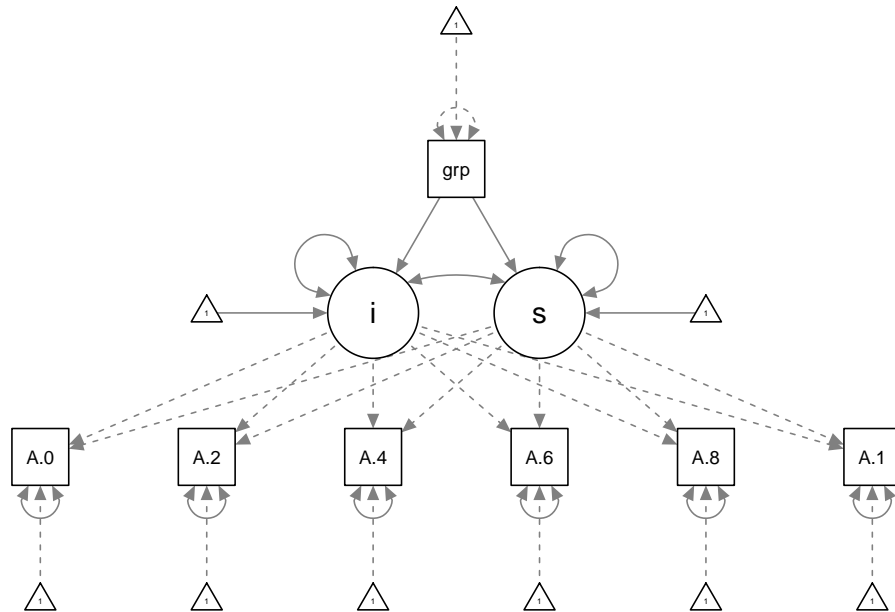
```

##
## Parameter Estimates:
##
## Information Observed
## Standard Errors Standard
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|)
## i =~
## A.0 1.000
## A.2 1.000
## A.4 1.000
## A.6 1.000
## A.8 1.000
## A.10 1.000
## s =~
## A.0 0.000
## A.2 1.000
## A.4 2.000
## A.6 3.000
## A.8 4.000
## A.10 5.000
##
## Regressions:
## Estimate Std.Err z-value P(>|z|)
## i ~
## groups 0.092 0.048 1.897 0.058
## s ~
## groups -0.011 0.016 -0.689 0.491
##
## Covariances:
## Estimate Std.Err z-value P(>|z|)
## .i ~~
## .s -0.004 0.006 -0.679 0.497
##
## Intercepts:
## Estimate Std.Err z-value P(>|z|)
## .A.0 0.000
## .A.2 0.000
## .A.4 0.000
## .A.6 0.000
## .A.8 0.000
## .A.10 0.000
## .i 2.627 0.079 33.071 0.000
## .s 0.001 0.027 0.029 0.977
##
## Variances:
## Estimate Std.Err z-value P(>|z|)
## .A.0 0.228 0.030 7.645 0.000
## .A.2 0.196 0.019 10.128 0.000
## .A.4 0.228 0.025 9.211 0.000
## .A.6 0.143 0.022 6.593 0.000
## .A.8 0.111 0.025 4.418 0.000
## .A.10 0.145 0.034 4.322 0.000

```

```
##      .i                0.142    0.024    6.030    0.000
##      .s                0.005    0.002    2.010    0.044
```

```
semPaths(fit2)
```



```
get_fixef <- function(fit, target.var){
  df <- parameterestimates(fit) %>%
    filter(lhs != rhs & lhs %in% target.var & op %in% c("~", "~1")) %>%
    unite(term, lhs, op, rhs, sep = "")
  v <- as.vector(df$est)
  names(v) <- df$term
  return(v)
}

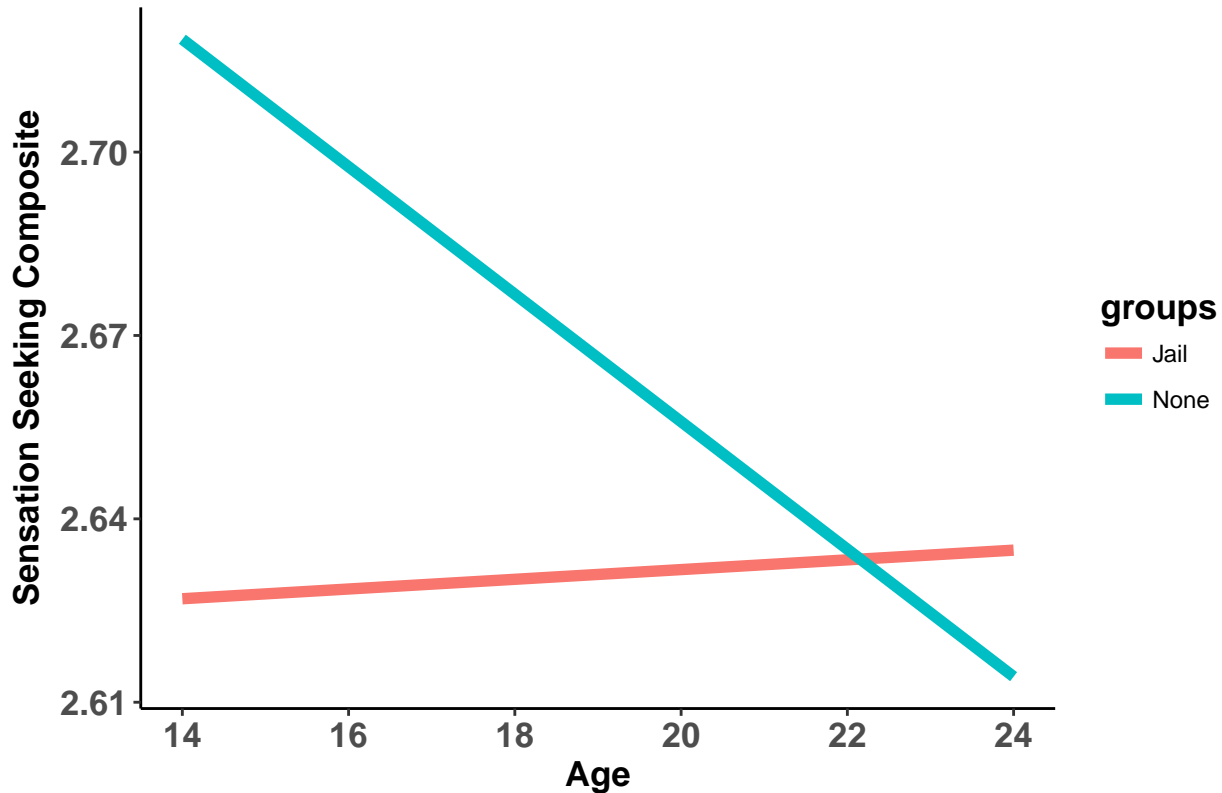
# example for categorical
fixed.frame <-
  data.frame(
    expand.grid(
      # here, you add values for your time variable and predictors
      Intercept = 1,
      age0 = seq(0,10,2),
      groupsNone = c(0,1))) %>%
  # now take care of interactions and add an intercept
  mutate(`age0:groupsNone` = age0*groupsNone) %>%
  select(groupsNone, `age0:groupsNone`, Intercept, age0)

fixed.frame$y <- as.vector(as.matrix(fixed.frame) %*% get_fixef(fit2, c("i", "s")))

fixed.frame %>%
  mutate(groups = factor(groupsNone, levels = c(0,1), labels = c("Jail", "None")),
         age = age0 + 14) %>%
  ggplot(aes(x = age, y = y, color = groups)) +
  geom_line(size = 2) +
  labs(x = "Age", y = "Sensation Seeking Composite",
       title = "2 Group Time Invariant Conditional Growth Models") +
```

```
theme_classic() +
  theme(axis.text = element_text(face = "bold", size = rel(1.2)),
        axis.title = element_text(face = "bold", size = rel(1.2)),
        legend.title = element_text(face = "bold", size = rel(1.2)),
        plot.title = element_text(face = "bold", size = rel(1.2), hjust = .5))
```

2 Group Time Invariant Conditional Growth Models



Tables

```
table_fun <- function(mod, int.var, slp.var, covar){
  var <- c(int.var, slp.var, covar)
  df <- parameterestimates(mod) %>%
    filter((lhs != rhs & lhs %in% var & op %in% c("~", "~1")) |
           (lhs == rhs & lhs %in% var & op == "~~")) %>%
    mutate(term = ifelse(lhs %in% int.var & rhs %in% covar, rhs,
                        ifelse(lhs %in% slp.var & rhs %in% covar, paste("Slope", rhs, sep = ":"),
                        ifelse(op %in% c("~1", "~~") , plyr::mapvalues(lhs, var, c("Intercept", "Slope", covar),
                        lhs))),
            type = plyr::mapvalues(op, c("~", "~1", "~~"), c("Fixed", "Fixed", "Random"))) %>%
    select(term, type, est, ci.lower, ci.upper)

  fitmeas <- data.frame(fitmeasures(mod)) %>%
    mutate(term = rownames(.), type = "fitmeas") %>%
    dplyr::rename(est = fitmeasures.mod.) %>%
    filter(term %in% c("cfi", "rmsea", "chisq", "df")) %>%
```



```
mutate(term = mapvalues(term, unique(term), c("$\\chi^2$", "df", "CFI", "RMSEA")))

df <- df %>% full_join(fitmeas)
}
```

Continuous Predictor

```
tab <- table_fun(fit1, "i", "s", "DemPweight")

(form.tab <- tab %>%
  filter(type != "fitmeas") %>%
  mutate(sig = ifelse(sign(ci.lower) == sign(ci.upper), "sig", "nonsig")) %>%
  mutate_at(vars(est:ci.upper), funs(sprintf("%.2f", .))) %>%
  mutate(CI = sprintf("[%s, %s]", ci.lower, ci.upper)) %>%
  mutate_at(vars(est, CI), funs(ifelse(sig == "sig", sprintf("\\textbf{%s}", .), .))) %>%
  select(term:est, CI) %>%
  full_join(tab %>% filter(type == "fitmeas") %>%
    mutate(est = sprintf("%.2f", est)) %>%
    select(term:est)) %>%
  mutate(type = factor(type, levels = c("Fixed", "Random", "fitmeas"))) %>%
  arrange(type))
```

##	term	type	est	CI
## 1	DemPweight	Fixed	0.05	[-0.08, 0.18]
## 2	Slope:DemPweight	Fixed	-0.02	[-0.06, 0.03]
## 3	DemPweight	Fixed	\\textbf{0.66}	\\textbf{[0.66, 0.66]}
## 4	Intercept	Fixed	\\textbf{2.73}	\\textbf{[2.64, 2.83]}
## 5	Slope	Fixed	-0.00	[-0.03, 0.03]
## 6	Intercept	Random	\\textbf{0.13}	\\textbf{[0.09, 0.17]}
## 7	Slope	Random	\\textbf{0.01}	\\textbf{[0.00, 0.01]}
## 8	DemPweight	Random	\\textbf{0.10}	\\textbf{[0.10, 0.10]}
## 9	\$\\chi^2\$	fitmeas	45.35	<NA>
## 10	df	fitmeas	20.00	<NA>
## 11	CFI	fitmeas	0.92	<NA>
## 12	RMSEA	fitmeas	0.04	<NA>

```
options(knitr.kable.NA = '')
form.tab %>% select(-type) %>%
  papaja::apa_table(caption = "Sample Lavaan Table",
    stub_indents = list("Fixed" = c(1:5),
      "Random" = c(6:8),
      "Fit Measures" = c(9:12)))
```

Categorical Predictor

```
tab <- table_fun(fit2, "i", "s", "groups")

(form.tab <- tab %>%
  filter(type != "fitmeas") %>%
  mutate(sig = ifelse(sign(ci.lower) == sign(ci.upper), "sig", "nonsig")) %>%
  mutate_at(vars(est:ci.upper), funs(sprintf("%.2f", .))) %>%
```

Table 1: Sample Lavaan Table

term	est	CI
Fixed		
DemPweight	0.05	[-0.08, 0.18]
Slope:DemPweight	-0.02	[-0.06, 0.03]
DemPweight	0.66	[0.66, 0.66]
Intercept	2.73	[2.64, 2.83]
Slope	-0.00	[-0.03, 0.03]
Random		
Intercept	0.13	[0.09, 0.17]
Slope	0.01	[0.00, 0.01]
DemPweight	0.10	[0.10, 0.10]
Fit Measures		
χ^2	45.35	
df	20.00	
CFI	0.92	
RMSEA	0.04	

```

mutate(CI = sprintf("[%s, %s]", ci.lower, ci.upper)) %>%
mutate_at(vars(est, CI), funs(ifelse(sig == "sig", sprintf("\\textbf{%s}", .), .))) %>%
select(term:est, CI) %>%
full_join(tab %>% filter(type == "fitmeas") %>%
           mutate(est = sprintf("%.2f", est)) %>%
           select(term:est)) %>%
mutate(type = factor(type, levels = c("Fixed", "Random", "fitmeas"))) %>%
arrange(type))

```

```

##      term      type      est      CI
## 1  groups  Fixed      0.09  [-0.00, 0.19]
## 2 Slope:groups Fixed     -0.01  [-0.04, 0.02]
## 3  groups  Fixed  \\textbf{1.56}  \\textbf{[1.56, 1.56]}
## 4  Intercept Fixed  \\textbf{2.63}  \\textbf{[2.47, 2.78]}
## 5  Slope    Fixed      0.00  [-0.05, 0.05]
## 6  Intercept Random  \\textbf{0.14}  \\textbf{[0.10, 0.19]}
## 7  Slope    Random  \\textbf{0.00}  \\textbf{[0.00, 0.01]}
## 8  groups   Random  \\textbf{0.25}  \\textbf{[0.25, 0.25]}
## 9  $\\chi^2$ fitmeas      38.40      <NA>
## 10 df      fitmeas      20.00      <NA>
## 11 CFI      fitmeas      0.93      <NA>
## 12 RMSEA    fitmeas      0.04      <NA>

```

```

options(knitr.kable.NA = '')
form.tab %>% select(-type) %>%
  papaja::apa_table(caption = "Sample Lavaan Table",
    stub_indents = list("Fixed" = c(1:5),
                        "Random" = c(6:8),
                        "Fit Measures" = c(9:12)))

```

Table 2: Sample Lavaan Table

term	est	CI
Fixed		
groups	0.09	[-0.00, 0.19]
Slope:groups	-0.01	[-0.04, 0.02]
groups	1.56	[1.56, 1.56]
Intercept	2.63	[2.47, 2.78]
Slope	0.00	[-0.05, 0.05]
Random		
Intercept	0.14	[0.10, 0.19]
Slope	0.00	[0.00, 0.01]
groups	0.25	[0.25, 0.25]
Fit Measures		
χ^2	38.40	
df	20.00	
CFI	0.93	
RMSEA	0.04	