

PA2-code and figures

```
library(readr)
library(haven)
library(dplyr)
```

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
library(psych)
library(lavaan)
```

```
## This is lavaan 0.6-9
## lavaan is FREE software! Please report any bugs.
```

```
##
## Attaching package: 'lavaan'
```

```
## The following object is masked from 'package:psych':
##
##   cor2cov
```

```
library(semPlot)
library(texreg)
```

```
## Version: 1.37.5
## Date: 2020-06-17
## Author: Philip Leifeld (University of Essex)
##
## Consider submitting praise using the praise or praise_interactive functions.
## Please cite the JSS article in your publications -- see citation("texreg").
```

```
data_60 <- read_csv("~/Downloads/data_60.csv")
```

```
## Rows: 635 Columns: 18
```

```
## -- Column specification -----
## Delimiter: ","
## dbl (18): id, x1, x2, x3, x4, m1, m2, m3, m4, m5, y1, y2, y3, y4, y5, v15, v...

##
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

##Remove Unit non response

```
data_60 <- data_60[data_60$id != 145, ]
data_60 <- data_60[data_60$id != 188, ]
data_60 <- data_60[data_60$id != 190, ]
data_60 <- data_60[data_60$id != 258, ]
data_60 <- data_60[data_60$id != 268, ]
data_60 <- data_60[data_60$id != 299, ]
data_60 <- data_60[data_60$id != 314, ]
data_60 <- data_60[data_60$id != 361, ]
data_60 <- data_60[data_60$id != 453, ]
data_60 <- data_60[data_60$id != 549, ]
```

#Weird Cases

##Duplicates

```
data60_cl <- add_count(data_60, x1, x2, x3, x4, m1, m2, m3, m4, m5, y1, y2, y3, y4, y5,
                        name = "duplicate")
```

```
data60_cl <- data60_cl[order(-data60_cl$duplicate),]
```

```
data60_cl <- data60_cl[data60_cl$id != 13, ]
data60_cl <- data60_cl[data60_cl$id != 80, ]
```

```
data60_cl$weird <- 0
data60_cl$weird[data60_cl$duplicate == 2] <- 1
```

##Straight liners

```
data60_cl$sd <- apply(data60_cl[,2:15], 1, sd)
```

```
data60_cl$weird[data60_cl$sd == 0] <- 2
data60_cl <- data60_cl[order(-data60_cl$weird, data60_cl$id),]
```

##Outliers

```
summary(data60_cl)
```

```
##           id           x1           x2           x3           x4
##  Min.      : 1.0    Min.   :1.000    Min.   :1.000    Min.   :1.00    Min.   :1.000
## 1st Qu.:159.5    1st Qu.:2.000    1st Qu.:2.000    1st Qu.:2.00    1st Qu.:2.000
## Median :321.0    Median :3.000    Median :3.000    Median :3.00    Median :3.000
## Mean   :319.1    Mean   :3.064    Mean   :3.051    Mean   :3.08    Mean   :3.021
## 3rd Qu.:478.5    3rd Qu.:4.000    3rd Qu.:4.000    3rd Qu.:4.00    3rd Qu.:4.000
## Max.   :635.0    Max.   :9.000    Max.   :7.000    Max.   :9.00    Max.   :8.000
##           m1           m2           m3           m4
```

```
## Min. :1.000 Min. :1.000 Min. :1.000 Min. :1.000
## 1st Qu.:2.000 1st Qu.:2.000 1st Qu.:2.000 1st Qu.:2.000
## Median :3.000 Median :3.000 Median :3.000 Median :3.000
## Mean :2.978 Mean :3.079 Mean :2.902 Mean :2.987
## 3rd Qu.:4.000 3rd Qu.:4.000 3rd Qu.:4.000 3rd Qu.:4.000
## Max. :9.000 Max. :9.000 Max. :8.000 Max. :9.000
## m5 y1 y2 y3
## Min. :1.000 Min. :1.000 Min. :1.000 Min. :1.000
## 1st Qu.:2.000 1st Qu.:2.000 1st Qu.:2.000 1st Qu.:2.000
## Median :3.000 Median :3.000 Median :3.000 Median :3.000
## Mean :2.899 Mean :2.992 Mean :3.005 Mean :3.048
## 3rd Qu.:4.000 3rd Qu.:4.000 3rd Qu.:4.000 3rd Qu.:4.000
## Max. :9.000 Max. :8.000 Max. :9.000 Max. :8.000
## y4 y5 v15 v16
## Min. :1.000 Min. :1.000 Min. :0.000 Min. : -3.174252
## 1st Qu.:2.000 1st Qu.:2.000 1st Qu.:0.000 1st Qu.: -0.690628
## Median :3.000 Median :3.000 Median :0.000 Median : 0.019442
## Mean :3.074 Mean :3.034 Mean :0.488 Mean : -0.008646
## 3rd Qu.:4.000 3rd Qu.:4.000 3rd Qu.:1.000 3rd Qu.: 0.615984
## Max. :9.000 Max. :8.000 Max. :1.000 Max. : 2.941043
## v17 duplicate weird sd
## Min. : 0.05405 Min. :1.00 Min. :0.00000 Min. :0.3631
## 1st Qu.: 0.48987 1st Qu.:1.00 1st Qu.:0.00000 1st Qu.:0.8018
## Median : 0.98386 Median :1.00 Median :0.00000 Median :0.9169
## Mean : 2.68932 Mean :1.01 Mean :0.00642 Mean :0.9536
## 3rd Qu.: 1.96512 3rd Qu.:1.00 3rd Qu.:0.00000 3rd Qu.:1.0509
## Max. :150.77086 Max. :3.00 Max. :1.00000 Max. :3.0751
```

```
data60_cl$x1[data60_cl$x1 > 5] <- NA
data60_cl$x2[data60_cl$x2 > 5] <- NA
data60_cl$x3[data60_cl$x3 > 5] <- NA
data60_cl$x4[data60_cl$x4 > 5] <- NA
data60_cl$m1[data60_cl$m1 > 5] <- NA
data60_cl$m2[data60_cl$m2 > 5] <- NA
data60_cl$m3[data60_cl$m3 > 5] <- NA
data60_cl$m4[data60_cl$m4 > 5] <- NA
data60_cl$m5[data60_cl$m5 > 5] <- NA
data60_cl$y1[data60_cl$y1 > 5] <- NA
data60_cl$y2[data60_cl$y2 > 5] <- NA
data60_cl$y3[data60_cl$y3 > 5] <- NA
data60_cl$y4[data60_cl$y4 > 5] <- NA
data60_cl$y5[data60_cl$y5 > 5] <- NA
```

```
summary(data60_cl)
```

```
## id x1 x2 x3
## Min. : 1.0 Min. :1.000 Min. :1.000 Min. :1.000
## 1st Qu.:159.5 1st Qu.:2.000 1st Qu.:2.000 1st Qu.:2.000
## Median :321.0 Median :3.000 Median :3.000 Median :3.000
## Mean :319.1 Mean :3.032 Mean :3.034 Mean :3.048
## 3rd Qu.:478.5 3rd Qu.:4.000 3rd Qu.:4.000 3rd Qu.:4.000
## Max. :635.0 Max. :5.000 Max. :5.000 Max. :5.000
## NA's :4 NA's :3 NA's :4
## x4 m1 m2 m3 m4
```

```
## Min. :1.000 Min. :1.00 Min. :1.000 Min. :1.000 Min. :1.000
## 1st Qu.:2.000 1st Qu.:2.00 1st Qu.:2.000 1st Qu.:2.000 1st Qu.:2.000
## Median :3.000 Median :3.00 Median :3.000 Median :3.000 Median :3.000
## Mean :3.006 Mean :2.96 Mean :3.069 Mean :2.881 Mean :2.977
## 3rd Qu.:4.000 3rd Qu.:4.00 3rd Qu.:4.000 3rd Qu.:4.000 3rd Qu.:4.000
## Max. :5.000 Max. :5.00 Max. :5.000 Max. :5.000 Max. :5.000
## NA's :2 NA's :2 NA's :1 NA's :3 NA's :1
## m5 y1 y2 y3 y4
## Min. :1.000 Min. :1.000 Min. :1.00 Min. :1.000 Min. :1.000
## 1st Qu.:2.000 1st Qu.:2.000 1st Qu.:2.00 1st Qu.:2.000 1st Qu.:2.000
## Median :3.000 Median :3.000 Median :3.00 Median :3.000 Median :3.000
## Mean :2.881 Mean :2.977 Mean :2.99 Mean :3.035 Mean :3.039
## 3rd Qu.:4.000 3rd Qu.:4.000 3rd Qu.:4.00 3rd Qu.:4.000 3rd Qu.:4.000
## Max. :5.000 Max. :5.000 Max. :5.00 Max. :5.000 Max. :5.000
## NA's :2 NA's :2 NA's :2 NA's :2 NA's :5
## y5 v15 v16 v17
## Min. :1.000 Min. :0.000 Min. : -3.174252 Min. : 0.05405
## 1st Qu.:2.000 1st Qu.:0.000 1st Qu.: -0.690628 1st Qu.: 0.48987
## Median :3.000 Median :0.000 Median : 0.019442 Median : 0.98386
## Mean :3.013 Mean :0.488 Mean : -0.008646 Mean : 2.68932
## 3rd Qu.:4.000 3rd Qu.:1.000 3rd Qu.: 0.615984 3rd Qu.: 1.96512
## Max. :5.000 Max. :1.000 Max. : 2.941043 Max. :150.77086
## NA's :3
## duplicate weird sd
## Min. :1.00 Min. :0.00000 Min. :0.3631
## 1st Qu.:1.00 1st Qu.:0.00000 1st Qu.:0.8018
## Median :1.00 Median :0.00000 Median :0.9169
## Mean :1.01 Mean :0.00642 Mean :0.9536
## 3rd Qu.:1.00 3rd Qu.:0.00000 3rd Qu.:1.0509
## Max. :3.00 Max. :1.00000 Max. :3.0751
##
```

```
data60_cl$weird[data60_cl$id == 50] <- 3
data60_cl$weird[data60_cl$id == 370] <- 3
data60_cl$weird[data60_cl$id == 405] <- 3
data60_cl$weird[data60_cl$id == 431] <- 3
data60_cl$weird[data60_cl$id == 450] <- 3
data60_cl$weird[data60_cl$id == 497] <- 3

data60_cl <- data60_cl[order(-data60_cl$weird, data60_cl$id),]

skew_x1 <- round(skew(data60_cl$x1),2)
skew_x2 <- round(skew(data60_cl$x2),2)
skew_x3 <- round(skew(data60_cl$x3),2)
skew_x4 <- round(skew(data60_cl$x4),2)

skew_m1 <- round(skew(data60_cl$m1),2)
skew_m2 <- round(skew(data60_cl$m2),2)
skew_m3 <- round(skew(data60_cl$m3),2)
skew_m4 <- round(skew(data60_cl$m4),2)
skew_m5 <- round(skew(data60_cl$m5),2)

skew_y1 <- round(skew(data60_cl$y1),2)
```

```

skew_y2 <- round(skew(data60_cl$y2),2)
skew_y3 <- round(skew(data60_cl$y3),2)
skew_y4 <- round(skew(data60_cl$y4),2)
skew_y5 <- round(skew(data60_cl$y5),2)

skew_v16 <- round(skew(data60_cl$v16),2)
skew_v17 <- round(skew(data60_cl$v17),2)

data60_sk <- data60_cl

data60_sk$Z_x1 <- scale(data60_sk$x1, center=TRUE, scale=TRUE)
data60_sk$Z_x2 <- scale(data60_sk$x2, center=TRUE, scale=TRUE)
data60_sk$Z_x3 <- scale(data60_sk$x3, center=TRUE, scale=TRUE)
data60_sk$Z_x4 <- scale(data60_sk$x4, center=TRUE, scale=TRUE)

data60_sk$Z_m1 <- scale(data60_sk$m1, center=TRUE, scale=TRUE)
data60_sk$Z_m2 <- scale(data60_sk$m2, center=TRUE, scale=TRUE)
data60_sk$Z_m3 <- scale(data60_sk$m3, center=TRUE, scale=TRUE)
data60_sk$Z_m4 <- scale(data60_sk$m4, center=TRUE, scale=TRUE)
data60_sk$Z_m5 <- scale(data60_sk$m5, center=TRUE, scale=TRUE)

data60_sk$Z_y1 <- scale(data60_sk$y1, center=TRUE, scale=TRUE)
data60_sk$Z_y2 <- scale(data60_sk$y2, center=TRUE, scale=TRUE)
data60_sk$Z_y3 <- scale(data60_sk$y3, center=TRUE, scale=TRUE)
data60_sk$Z_y4 <- scale(data60_sk$y4, center=TRUE, scale=TRUE)
data60_sk$Z_y5 <- scale(data60_sk$y5, center=TRUE, scale=TRUE)

summary(data60_sk)

```

```

##          id          x1          x2          x3
## Min.      : 1.0    Min.    :1.000    Min.    :1.000    Min.    :1.000
## 1st Qu.:159.5    1st Qu.:2.000    1st Qu.:2.000    1st Qu.:2.000
## Median :321.0    Median :3.000    Median :3.000    Median :3.000
## Mean     :319.1    Mean     :3.032    Mean     :3.034    Mean     :3.048
## 3rd Qu.:478.5    3rd Qu.:4.000    3rd Qu.:4.000    3rd Qu.:4.000
## Max.     :635.0    Max.     :5.000    Max.     :5.000    Max.     :5.000
##          NA's      :4          NA's      :3          NA's      :4
##          x4          m1          m2          m3          m4
## Min.      :1.000    Min.      :1.00    Min.      :1.000    Min.      :1.000    Min.      :1.000
## 1st Qu.:2.000    1st Qu.:2.00    1st Qu.:2.000    1st Qu.:2.000    1st Qu.:2.000
## Median :3.000    Median :3.00    Median :3.000    Median :3.000    Median :3.000
## Mean     :3.006    Mean     :2.96    Mean     :3.069    Mean     :2.881    Mean     :2.977
## 3rd Qu.:4.000    3rd Qu.:4.00    3rd Qu.:4.000    3rd Qu.:4.000    3rd Qu.:4.000
## Max.     :5.000    Max.     :5.00    Max.     :5.000    Max.     :5.000    Max.     :5.000
## NA's      :2          NA's      :2          NA's      :1          NA's      :3          NA's      :1
##          m5          y1          y2          y3          y4
## Min.      :1.000    Min.      :1.000    Min.      :1.00    Min.      :1.000    Min.      :1.000
## 1st Qu.:2.000    1st Qu.:2.000    1st Qu.:2.00    1st Qu.:2.000    1st Qu.:2.000
## Median :3.000    Median :3.000    Median :3.00    Median :3.000    Median :3.000
## Mean     :2.881    Mean     :2.977    Mean     :2.99    Mean     :3.035    Mean     :3.039
## 3rd Qu.:4.000    3rd Qu.:4.000    3rd Qu.:4.00    3rd Qu.:4.000    3rd Qu.:4.000
## Max.     :5.000    Max.     :5.000    Max.     :5.00    Max.     :5.000    Max.     :5.000

```

```

## NA's :2      NA's :2      NA's :2      NA's :2      NA's :5
##      y5      v15      v16      v17
## Min. :1.000  Min. :0.000  Min. : -3.174252  Min. : 0.05405
## 1st Qu.:2.000  1st Qu.:0.000  1st Qu.: -0.690628  1st Qu.: 0.48987
## Median :3.000  Median :0.000  Median : 0.019442  Median : 0.98386
## Mean :3.013  Mean :0.488  Mean : -0.008646  Mean : 2.68932
## 3rd Qu.:4.000  3rd Qu.:1.000  3rd Qu.: 0.615984  3rd Qu.: 1.96512
## Max. :5.000  Max. :1.000  Max. : 2.941043  Max. :150.77086
## NA's :3
##      duplicate      weird      sd      Z_x1.V1
## Min. :1.00  Min. :0.00000  Min. :0.3631  Min. : -1.899616
## 1st Qu.:1.00  1st Qu.:0.00000  1st Qu.:0.8018  1st Qu.: -0.964908
## Median :1.00  Median :0.00000  Median :0.9169  Median : -0.030201
## Mean :1.01  Mean :0.03531  Mean :0.9536  Mean : 0.000000
## 3rd Qu.:1.00  3rd Qu.:0.00000  3rd Qu.:1.0509  3rd Qu.: 0.904507
## Max. :3.00  Max. :3.00000  Max. :3.0751  Max. : 1.839215
##      NA's :4
##      Z_x2.V1      Z_x3.V1      Z_x4.V1
## Min. : -1.9287831  Min. : -2.021582  Min. : -1.9936613
## 1st Qu.: -0.9804520  1st Qu.: -1.034705  1st Qu.: -1.0000307
## Median : -0.0321209  Median : -0.047829  Median : -0.0064002
## Mean : 0.0000000  Mean : 0.0000000  Mean : 0.0000000
## 3rd Qu.: 0.9162102  3rd Qu.: 0.939047  3rd Qu.: 0.9872303
## Max. : 1.8645413  Max. : 1.925923  Max. : 1.9808609
## NA's :3      NA's :4      NA's :2
##      Z_m1.V1      Z_m2.V1      Z_m3.V1
## Min. : -1.8523246  Min. : -1.9544331  Min. : -1.9858209
## 1st Qu.: -0.9071368  1st Qu.: -1.0098664  1st Qu.: -0.9298955
## Median : 0.0380510  Median : -0.0652996  Median : 0.1260298
## Mean : 0.0000000  Mean : 0.0000000  Mean : 0.0000000
## 3rd Qu.: 0.9832389  3rd Qu.: 0.8792671  3rd Qu.: 1.1819551
## Max. : 1.9284267  Max. : 1.8238339  Max. : 2.2378805
## NA's :2      NA's :1      NA's :3
##      Z_m4.V1      Z_m5.V1      Z_y1.V1
## Min. : -1.9128440  Min. : -2.0094371  Min. : -1.9083957
## 1st Qu.: -0.9455359  1st Qu.: -0.9410634  1st Qu.: -0.9433194
## Median : 0.0217722  Median : 0.1273102  Median : 0.0217570
## Mean : 0.0000000  Mean : 0.0000000  Mean : 0.0000000
## 3rd Qu.: 0.9890803  3rd Qu.: 1.1956839  3rd Qu.: 0.9868333
## Max. : 1.9563885  Max. : 2.2640576  Max. : 1.9519096
## NA's :1      NA's :2      NA's :2
##      Z_y2.V1      Z_y3.V1      Z_y4.V1
## Min. : -1.8626183  Min. : -1.9591454  Min. : -2.025634
## 1st Qu.: -0.9267882  1st Qu.: -0.9966222  1st Qu.: -1.032109
## Median : 0.0090418  Median : -0.0340990  Median : -0.038584
## Mean : 0.0000000  Mean : 0.0000000  Mean : 0.000000
## 3rd Qu.: 0.9448719  3rd Qu.: 0.9284241  3rd Qu.: 0.954942
## Max. : 1.8807020  Max. : 1.8909473  Max. : 1.948467
## NA's :2      NA's :2      NA's :5
##      Z_y5.V1
## Min. : -1.8511572
## 1st Qu.: -0.9315118
## Median : -0.0118664
## Mean : 0.0000000

```

```
## 3rd Qu.: 0.9077790
## Max.    : 1.8274244
## NA's    :3
```

#Weird Variables

```
dfX <- data.frame (first_column = (data60_c1$x1),
                    second_column = (data60_c1$x2),
                    third_column = (data60_c1$x3),
                    fourth_column = (data60_c1$x4)
                    )
```

```
dfX <- na.exclude(dfX)
names(dfX) = c("x1", "x2", "x3", "x4")
round(cor(dfX), 2)
```

```
##      x1  x2  x3  x4
## x1 1.00 0.35 0.34 0.33
## x2 0.35 1.00 0.41 0.42
## x3 0.34 0.41 1.00 0.34
## x4 0.33 0.42 0.34 1.00
```

```
alpha(dfX)
```

```
##
## Reliability analysis
## Call: alpha(x = dfX)
##
## raw_alpha std.alpha G6(smc) average_r S/N ase mean sd median_r
##      0.7      0.7      0.64      0.37 2.3 0.02   3 0.75   0.35
##
## lower alpha upper      95% confidence boundaries
## 0.66 0.7 0.74
##
## Reliability if an item is dropped:
## raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r med.r
## x1      0.66      0.66      0.57      0.39 1.9  0.024 2.3e-03 0.41
## x2      0.60      0.60      0.50      0.34 1.5  0.028 2.5e-05 0.34
## x3      0.64      0.64      0.54      0.37 1.8  0.025 2.3e-03 0.35
## x4      0.64      0.64      0.54      0.37 1.8  0.025 1.5e-03 0.35
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean sd
## x1 617 0.71 0.70 0.53 0.44 3.0 1.1
## x2 617 0.76 0.75 0.64 0.53 3.0 1.1
## x3 617 0.72 0.72 0.57 0.48 3.1 1.0
## x4 617 0.72 0.72 0.58 0.48 3.0 1.0
##
## Non missing response frequency for each item
##      1 2 3 4 5 miss
## x1 0.08 0.24 0.33 0.27 0.08 0
## x2 0.06 0.25 0.38 0.21 0.10 0
## x3 0.06 0.24 0.39 0.23 0.08 0
## x4 0.05 0.27 0.39 0.22 0.08 0
```

```
dfM <- data.frame (first_column = (data60_c1$m1),
                   second_column = (data60_c1$m2),
                   third_column = (data60_c1$m3),
                   fourth_column = (data60_c1$m4),
                   fifth_column = (data60_c1$m5)
)
dfM <- na.exclude(dfM)
names(dfM) = c("m1", "m2", "m3", "m4", "m5")
round(cor(dfM), 2)
```

```
##      m1    m2    m3    m4    m5
## m1  1.00 -0.50  0.20  0.50  0.47
## m2 -0.50  1.00 -0.17 -0.48 -0.47
## m3  0.20 -0.17  1.00  0.16  0.13
## m4  0.50 -0.48  0.16  1.00  0.47
## m5  0.47 -0.47  0.13  0.47  1.00
```

```
alpha(dfM)
```

```
## Some items ( m2 ) were negatively correlated with the total scale and
## probably should be reversed.
## To do this, run the function again with the 'check.keys=TRUE' option
```

```
##
## Reliability analysis
## Call: alpha(x = dfM)
##
##   raw_alpha std.alpha G6(smc) average_r S/N ase mean sd median_r
##   0.095      0.13      0.37      0.03 0.15 0.052  3 0.47      0.14
##
## lower alpha upper      95% confidence boundaries
## -0.01 0.1 0.2
##
## Reliability if an item is dropped:
##   raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r med.r
## m1   -0.380   -0.2983  0.069   -0.0609 -0.2297  0.088 0.144 -0.0241
## m2    0.659    0.6541  0.617    0.3210  1.8911  0.022 0.031  0.3324
## m3   -0.057   -0.0076  0.352   -0.0019 -0.0076  0.062 0.279 -0.0015
## m4   -0.360   -0.2848  0.091   -0.0587 -0.2217  0.086 0.153 -0.0241
## m5   -0.273   -0.2310  0.132   -0.0492 -0.1877  0.080 0.162 -0.0057
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean sd
## m1 619  0.71  0.70  0.73  0.35  3.0 1.06
## m2 619 -0.24 -0.26 -0.88 -0.58  3.1 1.06
## m3 619  0.54  0.55  0.25  0.15  2.9 0.95
## m4 619  0.70  0.70  0.70  0.35  3.0 1.03
## m5 619  0.66  0.67  0.64  0.32  2.9 0.94
##
## Non missing response frequency for each item
##      1 2 3 4 5 miss
## m1 0.08 0.26 0.34 0.25 0.07 0
```



```
## m2 0.06 0.25 0.34 0.26 0.09    0
## m3 0.08 0.25 0.37 0.28 0.01    0
## m4 0.08 0.25 0.35 0.26 0.06    0
## m5 0.08 0.27 0.36 0.29 0.01    0
```

```
dfY <- data.frame (first_column = (data60_cl$y1),
                   second_column = (data60_cl$y2),
                   third_column = (data60_cl$y3),
                   fourth_column = (data60_cl$y4),
                   fifth_column = (data60_cl$y5)
)
dfY <- na.exclude(dfY)
names(dfY) = c("1y", "y2", "y3", "y4", "y5")
round(cor(dfY), 2)
```

```
##      1y  y2  y3  y4  y5
## 1y 1.00 0.51 0.55 0.58 0.14
## y2 0.51 1.00 0.52 0.54 0.12
## y3 0.55 0.52 1.00 0.60 0.14
## y4 0.58 0.54 0.60 1.00 0.14
## y5 0.14 0.12 0.14 0.14 1.00
```

```
alpha(dfY)
```

```
##
## Reliability analysis
## Call: alpha(x = dfY)
##
## raw_alpha std.alpha G6(smc) average_r S/N ase mean sd median_r
##      0.75      0.76      0.74      0.38 3.1 0.016   3 0.74      0.51
##
## lower alpha upper      95% confidence boundaries
## 0.72 0.75 0.78
##
## Reliability if an item is dropped:
## raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r med.r
## 1y      0.67      0.67      0.66      0.34 2.1  0.022 0.0535 0.33
## y2      0.68      0.69      0.67      0.36 2.2  0.021 0.0581 0.34
## y3      0.67      0.67      0.65      0.34 2.0  0.022 0.0515 0.32
## y4      0.66      0.66      0.64      0.33 2.0  0.023 0.0466 0.33
## y5      0.83      0.83      0.79      0.55 4.9  0.011 0.0012 0.54
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean sd
## 1y 616 0.78 0.78 0.72 0.62 3 1.0
## y2 616 0.76 0.76 0.67 0.58 3 1.1
## y3 616 0.78 0.79 0.73 0.63 3 1.0
## y4 616 0.80 0.80 0.76 0.65 3 1.0
## y5 616 0.44 0.43 0.18 0.16 3 1.1
##
## Non missing response frequency for each item
##      1 2 3 4 5 miss
## 1y 0.07 0.26 0.38 0.21 0.08 0
```

```
## y2 0.08 0.25 0.35 0.24 0.08    0
## y3 0.07 0.21 0.43 0.20 0.09    0
## y4 0.06 0.23 0.38 0.26 0.07    0
## y5 0.06 0.28 0.32 0.23 0.10    0
```

##Solve M2

```
data60_cl$m2r[data60_cl$m2 == 1] <- 5
data60_cl$m2r[data60_cl$m2 == 2] <- 4
data60_cl$m2r[data60_cl$m2 == 3] <- 3
data60_cl$m2r[data60_cl$m2 == 4] <- 2
data60_cl$m2r[data60_cl$m2 == 5] <- 1

dfM2 <- data.frame (first_column = (data60_cl$m1),
                    second_column = (data60_cl$m2r),
                    third_column = (data60_cl$m3),
                    fourth_column = (data60_cl$m4),
                    fifth_column = (data60_cl$m5)
)
dfM2 <- na.exclude(dfM2)
names(dfM2) = c("m1", "m2r", "m3", "m4", "m5")
round(cor(dfM2), 2)
```

```
##      m1 m2r m3 m4 m5
## m1  1.00 0.50 0.20 0.50 0.47
## m2r 0.50 1.00 0.17 0.48 0.47
## m3  0.20 0.17 1.00 0.16 0.13
## m4  0.50 0.48 0.16 1.00 0.47
## m5  0.47 0.47 0.13 0.47 1.00
```

```
alpha(dfM2)
```

```
##
## Reliability analysis
## Call: alpha(x = dfM2)
##
## raw_alpha std.alpha G6(smc) average_r S/N ase mean sd median_r
##      0.74      0.73      0.71      0.36 2.8 0.016  2.9 0.7      0.47
##
## lower alpha upper      95% confidence boundaries
## 0.71 0.74 0.77
##
## Reliability if an item is dropped:
## raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r med.r
## m1      0.65      0.65      0.61      0.31 1.8 0.023 0.03047 0.32
## m2r      0.66      0.65      0.62      0.32 1.9 0.022 0.03092 0.33
## m3      0.79      0.79      0.74      0.48 3.7 0.014 0.00024 0.48
## m4      0.66      0.66      0.62      0.32 1.9 0.022 0.03065 0.33
## m5      0.68      0.67      0.63      0.34 2.0 0.021 0.03000 0.34
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean sd
## m1 619 0.78 0.77 0.70 0.60 3.0 1.06
## m2r 619 0.76 0.75 0.68 0.58 2.9 1.06
```

```
## m3 619 0.46 0.48 0.24 0.21 2.9 0.95
## m4 619 0.75 0.75 0.67 0.58 3.0 1.03
## m5 619 0.72 0.73 0.64 0.55 2.9 0.94
##
## Non missing response frequency for each item
##      1      2      3      4      5 miss
## m1 0.08 0.26 0.34 0.25 0.07      0
## m2r 0.09 0.26 0.34 0.25 0.06      0
## m3 0.08 0.25 0.37 0.28 0.01      0
## m4 0.08 0.25 0.35 0.26 0.06      0
## m5 0.08 0.27 0.36 0.29 0.01      0
```

```
#Scale Construction
##Construct new Scales
data60_cl$X <- rowMeans(data60_cl[,c(2,3,4,5)])

data60_cl$M <- rowMeans(data60_cl[,c(6,8,9,10,22)])

data60_cl$Y <- rowMeans(data60_cl[,c(11,12,13,14,15)])

#Analyses
##Q1 & Q2
model.1 <- "
M ~ a*X
Y ~ b*M + cp*X
indirect := a*b
direct   := cp
total    := a*b + cp
"
mediation <- sem(model.1, data = data60_cl, se = "bootstrap", bootstrap=1000)

summary(mediation, ci=T, standardized=T, rsquare=T, fit.measures=F)
```

```
## lavaan 0.6-9 ended normally after 14 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters          5
##
##                               Used      Total
##      Number of observations          616      623
##
## Model Test User Model:
##
##      Test statistic          0.000
##      Degrees of freedom          0
##
## Parameter Estimates:
##
##      Standard errors          Bootstrap
##      Number of requested bootstrap draws          1000
##      Number of successful bootstrap draws          1000
##
```

```
## Regressions:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## M ~
## X      (a)    0.185   0.050   3.698   0.000   0.089   0.280
## Y ~
## M      (b)    0.344   0.062   5.533   0.000   0.228   0.471
## X      (cp)   0.170   0.057   2.962   0.003   0.051   0.273
## Std.lv Std.all
##
##    0.185   0.197
##
##    0.344   0.327
##    0.170   0.172
##
## Variances:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## .M           0.473   0.030  15.858   0.000   0.414   0.534
## .Y           0.459   0.031  14.786   0.000   0.398   0.521
## Std.lv Std.all
##    0.473   0.961
##    0.459   0.841
##
## R-Square:
##           Estimate
## M           0.039
## Y           0.159
##
## Defined Parameters:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## indirect     0.064   0.026   2.428   0.015   0.024   0.126
## direct       0.170   0.057   2.961   0.003   0.051   0.273
## total        0.234   0.043   5.420   0.000   0.145   0.317
## Std.lv Std.all
##    0.064   0.065
##    0.170   0.172
##    0.234   0.237
```

```
##Q3 & Q4
model.2 <- "
M ~ a*X + d1*v15 +f1*v16
Y ~ b*M + cp*X + d2*v15 +f2*v16
indirect := a*b
direct   := cp
total    := a*b + cp
"
covariates <- sem(model.2, data = data60_cl, se = "bootstrap", bootstrap=1000)

summary(covariates, ci=T, standardized=T, rsquare=T, fit.measures=F)
```

```
## lavaan 0.6-9 ended normally after 18 iterations
##
## Estimator ML
## Optimization method NLMINB
## Number of model parameters 9
```

```

##
##
##           Used           Total
##   Number of observations       616       623
##
## Model Test User Model:
##
##   Test statistic           0.000
##   Degrees of freedom           0
##
## Parameter Estimates:
##
##   Standard errors           Bootstrap
##   Number of requested bootstrap draws       1000
##   Number of successful bootstrap draws       1000
##
## Regressions:
##           Estimate   Std.Err   z-value   P(>|z|)   ci.lower   ci.upper
##   M ~
##   X           (a)     0.193     0.046     4.162     0.000     0.104     0.284
##   v15         (d1)     0.167     0.053     3.132     0.002     0.059     0.274
##   v16         (f1)     0.155     0.025     6.179     0.000     0.105     0.202
##   Y ~
##   M           (b)     0.270     0.057     4.699     0.000     0.164     0.391
##   X           (cp)     0.195     0.054     3.597     0.000     0.085     0.293
##   v15         (d2)     0.259     0.053     4.853     0.000     0.155     0.365
##   v16         (f2)     0.167     0.029     5.822     0.000     0.109     0.221
##   Std.lv   Std.all
##
##   0.193     0.206
##   0.167     0.119
##   0.155     0.220
##
##   0.270     0.256
##   0.195     0.197
##   0.259     0.175
##   0.167     0.224
##
## Variances:
##           Estimate   Std.Err   z-value   P(>|z|)   ci.lower   ci.upper
##   .M           0.443     0.028     15.790     0.000     0.387     0.496
##   .Y           0.419     0.027     15.319     0.000     0.359     0.467
##   Std.lv   Std.all
##   0.443     0.900
##   0.419     0.768
##
## R-Square:
##           Estimate
##   M           0.100
##   Y           0.232
##
## Defined Parameters:
##           Estimate   Std.Err   z-value   P(>|z|)   ci.lower   ci.upper
##   indirect     0.052     0.022     2.408     0.016     0.019     0.104
##   direct       0.195     0.054     3.595     0.000     0.085     0.293

```

```
##      total      0.246    0.042    5.928    0.000    0.172    0.331
##      Std.lv Std.all
##      0.052    0.053
##      0.195    0.197
##      0.246    0.250
```

```
##Q5
model.3 <- "
M ~ a*X + d1*v15 +f1*v16
Y ~ b*M + cp*X + d2*v15 +f2*v16
v17 ~ g1*M + g2*Y
indirect := a*b
direct   := cp
total    := a*b + cp
"
covariate17 <- sem(model.3, data = data60_cl, se = "bootstrap", bootstrap=1000)
summary(covariate17, ci=T, standardized=T, rsquare=T, fit.measures=F)
```

```
## lavaan 0.6-9 ended normally after 31 iterations
##
##      Estimator      ML
##      Optimization method      NLMINB
##      Number of model parameters      12
##
##      Used      Total
##      Number of observations      616      623
##
## Model Test User Model:
##
##      Test statistic      67.638
##      Degrees of freedom      3
##      P-value (Chi-square)      0.000
##
## Parameter Estimates:
##
##      Standard errors      Bootstrap
##      Number of requested bootstrap draws      1000
##      Number of successful bootstrap draws      1000
##
## Regressions:
##      Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      M ~
##      X      (a)    0.193    0.048    3.985    0.000    0.098    0.292
##      v15     (d1)    0.167    0.052    3.206    0.001    0.073    0.275
##      v16     (f1)    0.155    0.025    6.176    0.000    0.108    0.205
##      Y ~
##      M      (b)    0.270    0.060    4.500    0.000    0.162    0.392
##      X      (cp)    0.195    0.055    3.527    0.000    0.082    0.292
##      v15     (d2)    0.259    0.053    4.883    0.000    0.161    0.364
##      v16     (f2)    0.167    0.028    5.874    0.000    0.108    0.224
##      v17 ~
##      M      (g1)   -5.687    1.781   -3.193    0.001   -8.973   -2.057
##      Y      (g2)    5.074    1.666    3.046    0.002    1.597    8.268
##      Std.lv Std.all
```

```
##
##      0.193      0.206
##      0.167      0.119
##      0.155      0.220
##
##      0.270      0.256
##      0.195      0.197
##      0.259      0.175
##      0.167      0.224
##
##     -5.687     -0.411
##      5.074      0.386
##
## Variances:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      .M              0.443   0.028  16.081   0.000   0.388   0.496
##      .Y              0.419   0.028  15.207   0.000   0.358   0.467
##      .v17            75.201  30.068   2.501   0.012  21.460  138.925
##      Std.lv Std.all
##      0.443   0.900
##      0.419   0.768
##      75.201   0.797
##
## R-Square:
##              Estimate
##      M              0.100
##      Y              0.232
##      v17            0.203
##
## Defined Parameters:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      indirect       0.052   0.023   2.295   0.022   0.019   0.108
##      direct         0.195   0.055   3.525   0.000   0.082   0.292
##      total          0.246   0.042   5.859   0.000   0.163   0.327
##      Std.lv Std.all
##      0.052   0.053
##      0.195   0.197
##      0.246   0.250
```

```
#Weird
data60_weird <- data60_cl
data60_weird1 <- data60_cl
data60_weird3 <- data60_cl
data60_weird <- data60_weird[order(-data60_weird$weird, data60_weird$id),]

data60_weird <- data60_weird[data60_weird$id != 22, ]
data60_weird <- data60_weird[data60_weird$id != 217, ]
data60_weird <- data60_weird[data60_weird$id != 345, ]
data60_weird <- data60_weird[data60_weird$id != 425, ]

data60_weird <- data60_weird[data60_weird$id != 50, ]
data60_weird <- data60_weird[data60_weird$id != 370, ]
data60_weird <- data60_weird[data60_weird$id != 405, ]
data60_weird <- data60_weird[data60_weird$id != 431, ]
```

```

data60_weird <- data60_weird[data60_weird$id != 450, ]
data60_weird <- data60_weird[data60_weird$id != 497, ]

data60_weird1 <- data60_weird1[data60_weird1$id != 50, ]
data60_weird1 <- data60_weird1[data60_weird1$id != 370, ]
data60_weird1 <- data60_weird1[data60_weird1$id != 405, ]
data60_weird1 <- data60_weird1[data60_weird1$id != 431, ]
data60_weird1 <- data60_weird1[data60_weird1$id != 450, ]
data60_weird1 <- data60_weird1[data60_weird1$id != 497, ]

data60_weird3 <- data60_weird3[data60_weird3$id != 22, ]
data60_weird3 <- data60_weird3[data60_weird3$id != 217, ]
data60_weird3 <- data60_weird3[data60_weird3$id != 345, ]
data60_weird3 <- data60_weird3[data60_weird3$id != 425, ]

mediation.W <- sem(model.1, data = data60_weird, se = "bootstrap", bootstrap=1000)
summary(mediation.W, ci=T, standardized=T, rsquare=T, fit.measures=F)

```

```

## lavaan 0.6-9 ended normally after 14 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters          5
##
##                               Used      Total
##      Number of observations          612      613
##
## Model Test User Model:
##
##      Test statistic          0.000
##      Degrees of freedom          0
##
## Parameter Estimates:
##
##      Standard errors          Bootstrap
##      Number of requested bootstrap draws          1000
##      Number of successful bootstrap draws          1000
##
## Regressions:
##
##              Estimate  Std.Err  z-value  P(>|z|)  ci.lower  ci.upper
## M ~
##   X      (a)    0.238    0.044    5.408    0.000    0.153    0.328
## Y ~
##   M      (b)    0.418    0.060    7.020    0.000    0.304    0.539
##   X      (cp)   0.100    0.055    1.816    0.069   -0.010    0.203
##   Std.lv  Std.all
##
##   0.238    0.254
##
##   0.418    0.397
##   0.100    0.101
##

```



```
## Variances:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      .M           0.443   0.027  16.283   0.000   0.392   0.497
##      .Y           0.427   0.030  14.341   0.000   0.368   0.484
## Std.lv Std.all
##    0.443   0.936
##    0.427   0.812
##
## R-Square:
##           Estimate
##      M           0.064
##      Y           0.188
##
## Defined Parameters:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## indirect      0.099   0.028   3.499   0.000   0.051   0.162
## direct        0.100   0.055   1.815   0.070  -0.010   0.203
## total         0.200   0.042   4.738   0.000   0.117   0.282
## Std.lv Std.all
##    0.099   0.101
##    0.100   0.101
##    0.200   0.202
```

```
covariates.W <- sem(model.2, data = data60_weird, se = "bootstrap", bootstrap=1000)
summary(covariates.W, ci=T, standardized=T, rsquare=T, fit.measures=F)
```

```
## lavaan 0.6-9 ended normally after 19 iterations
##
## Estimator ML
## Optimization method NLMINB
## Number of model parameters 9
##
## Used Total
## Number of observations 612 613
##
## Model Test User Model:
##
## Test statistic 0.000
## Degrees of freedom 0
##
## Parameter Estimates:
##
## Standard errors Bootstrap
## Number of requested bootstrap draws 1000
## Number of successful bootstrap draws 1000
##
## Regressions:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## M ~
## X (a) 0.240 0.046 5.278 0.000 0.155 0.333
## v15 (d1) 0.143 0.054 2.669 0.008 0.032 0.254
## v16 (f1) 0.143 0.026 5.574 0.000 0.092 0.190
## Y ~
## M (b) 0.344 0.056 6.159 0.000 0.241 0.459
```

```

##      X      (cp)    0.123    0.051    2.386    0.017    0.013    0.215
##      v15     (d2)    0.276    0.052    5.302    0.000    0.170    0.378
##      v16     (f2)    0.170    0.027    6.179    0.000    0.117    0.224
##      Std.lv  Std.all
##
##      0.240    0.256
##      0.143    0.104
##      0.143    0.207
##
##      0.344    0.327
##      0.123    0.124
##      0.276    0.190
##      0.170    0.233
##
## Variances:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      .M              0.418   0.028  15.080   0.000   0.363   0.473
##      .Y              0.384   0.026  14.812   0.000   0.330   0.428
##      Std.lv  Std.all
##      0.418    0.884
##      0.384    0.730
##
## R-Square:
##              Estimate
##      M              0.116
##      Y              0.270
##
## Defined Parameters:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      indirect      0.083   0.026   3.176   0.001   0.041   0.144
##      direct        0.123   0.051   2.385   0.017   0.013   0.215
##      total         0.205   0.039   5.305   0.000   0.124   0.277
##      Std.lv  Std.all
##      0.083    0.084
##      0.123    0.124
##      0.205    0.208

```

```

covariate17.W <- sem(model.3, data = data60_weird, se = "bootstrap", bootstrap=1000)
summary(covariate17.W, ci=T, standardized=T, rsquare=T, fit.measures=F)

```

```

## lavaan 0.6-9 ended normally after 29 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters          12
##
##              Used      Total
##      Number of observations          612          613
##
## Model Test User Model:
##
##      Test statistic          45.169
##      Degrees of freedom           3
##      P-value (Chi-square)        0.000

```

```

##
## Parameter Estimates:
##
## Standard errors                                Bootstrap
## Number of requested bootstrap draws            1000
## Number of successful bootstrap draws            1000
##
## Regressions:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## M ~
##   X      (a)      0.240   0.044   5.438   0.000   0.155   0.331
##   v15     (d1)     0.143   0.053   2.696   0.007   0.040   0.250
##   v16     (f1)     0.143   0.024   5.871   0.000   0.096   0.190
## Y ~
##   M      (b)      0.344   0.058   5.971   0.000   0.243   0.471
##   X      (cp)     0.123   0.053   2.323   0.020   0.016   0.221
##   v15     (d2)     0.276   0.052   5.288   0.000   0.173   0.377
##   v16     (f2)     0.170   0.027   6.336   0.000   0.117   0.224
## v17 ~
##   M      (g1)    -3.215   1.427  -2.253   0.024  -6.174  -0.560
##   Y      (g2)     2.804   1.401   2.002   0.045   0.137   5.735
## Std.lv Std.all
##
##   0.240   0.256
##   0.143   0.104
##   0.143   0.207
##
##   0.344   0.327
##   0.123   0.124
##   0.276   0.190
##   0.170   0.233
##
##  -3.215  -0.330
##   2.804   0.304
##
## Variances:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##   .M           0.418   0.026  16.167   0.000   0.361   0.464
##   .Y           0.384   0.025  15.319   0.000   0.330   0.428
##   .v17        39.632  17.123   2.314   0.021   7.281  75.499
## Std.lv Std.all
##   0.418   0.884
##   0.384   0.730
##  39.632   0.883
##
## R-Square:
##           Estimate
##   M           0.116
##   Y           0.270
##   v17         0.117
##
## Defined Parameters:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##   indirect     0.083   0.026   3.213   0.001   0.042   0.142

```

```
##      direct      0.123    0.053    2.322    0.020    0.016    0.221
##      total      0.205    0.040    5.098    0.000    0.128    0.285
##      Std.lv   Std.all
##      0.083    0.084
##      0.123    0.124
##      0.205    0.208
```

#Omitted Variable Bias

```
screenreg(list(mediation, mediation.W),
  custom.model.name =
    c("Model 1: Normal",
      "Model 2: Adjusted for Weird"),
  single.row = TRUE, digits = 3)
```

```
##
## =====
##              Model 1: Normal      Model 2: Adjusted for Weird
## -----
## M ~ X          0.185 (0.050) ***    0.238 (0.044) ***
## Y ~ M          0.344 (0.062) ***    0.418 (0.060) ***
## Y ~ X          0.170 (0.057) **     0.100 (0.055)
## M ~~ M         0.473 (0.030) ***    0.443 (0.027) ***
## Y ~~ Y         0.459 (0.031) ***    0.427 (0.030) ***
## X ~~ X         0.560 (0.000)         0.538 (0.000)
## indirect := a*b 0.064 (0.026) *      0.099 (0.028) ***
## direct := cp    0.170 (0.057) **     0.100 (0.055)
## total := a*b+cp 0.234 (0.043) ***    0.200 (0.042) ***
## -----
## agfi           1.000                1.000
## AIC            2565.701              2464.546
## BIC            2587.817              2486.630
## cfi            1.000                1.000
## chisq          0.000                0.000
## npar           5.000                5.000
## rmsea          0.000                0.000
## rmsea.conf.high 0.000                0.000
## srmr           0.000                0.000
## tli            1.000                1.000
## converged      1                    1
## ngroups        1                    1
## nobs           616                  612
## norig          623                  613
## nexcluded      7                    1
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05
```

```
screenreg(list(covariates, covariates.W),
  custom.model.name =
    c("Model 1: Normal",
      "Model 2: Adjusted for Weird"),
  single.row = TRUE, digits = 3)
```

```
##
```

```
## =====
##                               Model 1: Normal      Model 2: Adjusted for Weird
## -----
## M ~ X                0.193 (0.046) ***      0.240 (0.046) ***
## M ~ v15              0.167 (0.053) **       0.143 (0.054) **
## M ~ v16              0.155 (0.025) ***      0.143 (0.026) ***
## Y ~ M                0.270 (0.057) ***      0.344 (0.056) ***
## Y ~ X                0.195 (0.054) ***      0.123 (0.051) *
## Y ~ v15              0.259 (0.053) ***      0.276 (0.052) ***
## Y ~ v16              0.167 (0.029) ***      0.170 (0.027) ***
## M ~~ M               0.443 (0.028) ***      0.418 (0.028) ***
## Y ~~ Y               0.419 (0.027) ***      0.384 (0.026) ***
## X ~~ X               0.560 (0.000)          0.538 (0.000)
## X ~~ v15             -0.016 (0.000)         -0.010 (0.000)
## X ~~ v16             -0.010 (0.000)         0.001 (0.000)
## v15 ~~ v15           0.250 (0.000)          0.250 (0.000)
## v15 ~~ v16           -0.018 (0.000)         -0.020 (0.000)
## v16 ~~ v16           0.988 (0.000)          0.989 (0.000)
## indirect := a*b      0.052 (0.022) *        0.083 (0.026) **
## direct := cp         0.195 (0.054) ***      0.123 (0.051) *
## total := a*b+cp      0.246 (0.042) ***      0.205 (0.039) ***
## -----
## agfi                 1.000                  1.000
## AIC                  2477.338                2372.295
## BIC                  2517.147                2412.045
## cfi                  1.000                  1.000
## chisq                0.000                  0.000
## npar                 9.000                  9.000
## rmsea                0.000                  0.000
## rmsea.conf.high      0.000                  0.000
## srmr                 0.000                  0.000
## tli                  1.000                  1.000
## converged            1                      1
## ngroups              1                      1
## nobs                 616                    612
## norig                623                    613
## nexcluded            7                      1
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05
```

```
screenreg(list(covariate17, covariate17.W),
  custom.model.name =
    c("Model 1: Normal",
      "Model 2: Adjusted for Weird"),
  single.row = TRUE, digits = 3)
```

```
##
## =====
##                               Model 1: Normal      Model 2: Adjusted for Weird
## -----
## M ~ X                0.193 (0.048) ***      0.240 (0.044) ***
## M ~ v15              0.167 (0.052) **       0.143 (0.053) **
## M ~ v16              0.155 (0.025) ***      0.143 (0.024) ***
## Y ~ M                0.270 (0.060) ***      0.344 (0.058) ***
```

```

## Y ~ X          0.195 (0.055) ***    0.123 (0.053) *
## Y ~ v15        0.259 (0.053) ***    0.276 (0.052) ***
## Y ~ v16        0.167 (0.028) ***    0.170 (0.027) ***
## v17 ~ M        -5.687 (1.781) **    -3.215 (1.427) *
## v17 ~ Y         5.074 (1.666) **     2.804 (1.401) *
## M ~~ M         0.443 (0.028) ***    0.418 (0.026) ***
## Y ~~ Y         0.419 (0.028) ***    0.384 (0.025) ***
## v17 ~~ v17     75.201 (30.068) *     39.632 (17.123) *
## X ~~ X         0.560 (0.000)         0.538 (0.000)
## X ~~ v15       -0.016 (0.000)       -0.010 (0.000)
## X ~~ v16       -0.010 (0.000)        0.001 (0.000)
## v15 ~~ v15      0.250 (0.000)        0.250 (0.000)
## v15 ~~ v16     -0.018 (0.000)       -0.020 (0.000)
## v16 ~~ v16      0.988 (0.000)        0.989 (0.000)
## indirect := a*b  0.052 (0.023) *     0.083 (0.026) **
## direct := cp    0.195 (0.055) ***    0.123 (0.053) *
## total := a*b+cp 0.246 (0.042) ***    0.205 (0.040) ***
## -----
## agfi           0.555                 0.689
## AIC            6892.688              6367.015
## BIC            6945.767              6420.016
## cfi            0.847                 0.888
## chisq          67.638                45.169
## npar           12.000                12.000
## rmsea          0.187                 0.152
## rmsea.conf.high 0.227                 0.192
## srmr           0.063                 0.053
## tli            0.389                 0.553
## converged      1                    1
## ngroups        1                    1
## nobs           616                  612
## norig          623                  613
## nexcluded      7                    1
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05

```

##Omitted Var Bias 1

```

mediation.W1 <- sem(model.1, data = data60_weird1, se = "bootstrap", bootstrap=1000)
summary(mediation.W1, ci=T, standardized=T, rsquare=T, fit.measures=F)

```

```

## lavaan 0.6-9 ended normally after 14 iterations
##
##      Estimator              ML
##      Optimization method    NLMINB
##      Number of model parameters    5
##
##                               Used      Total
##      Number of observations      616      617
##
## Model Test User Model:
##
##      Test statistic      0.000
##      Degrees of freedom      0

```

```
##
## Parameter Estimates:
##
## Standard errors                                Bootstrap
## Number of requested bootstrap draws            1000
## Number of successful bootstrap draws            1000
##
## Regressions:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## M ~
## X      (a)    0.185   0.049   3.774   0.000   0.089   0.283
## Y ~
## M      (b)    0.344   0.061   5.655   0.000   0.239   0.477
## X      (cp)   0.170   0.057   3.003   0.003   0.055   0.270
## Std.lv Std.all
##
## 0.185   0.197
##
## 0.344   0.327
## 0.170   0.172
##
## Variances:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## .M          0.473   0.030  15.851   0.000   0.412   0.532
## .Y          0.459   0.030  15.383   0.000   0.397   0.512
## Std.lv Std.all
## 0.473   0.961
## 0.459   0.841
##
## R-Square:
##           Estimate
## M          0.039
## Y          0.159
##
## Defined Parameters:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## indirect    0.064   0.026   2.486   0.013   0.025   0.126
## direct      0.170   0.057   3.002   0.003   0.055   0.270
## total       0.234   0.043   5.405   0.000   0.150   0.319
## Std.lv Std.all
## 0.064   0.065
## 0.170   0.172
## 0.234   0.237
```

```
covariates.W1 <- sem(model.2, data = data60_weird1, se = "bootstrap", bootstrap=1000)
summary(covariates.W1, ci=T, standardized=T, rsquare=T, fit.measures=F)
```

```
## lavaan 0.6-9 ended normally after 18 iterations
```

```
##
## Estimator                                ML
## Optimization method                      NLMINB
## Number of model parameters                9
##
##                                         Used      Total
```

```

##      Number of observations                616          617
##
## Model Test User Model:
##
##      Test statistic                0.000
##      Degrees of freedom                0
##
## Parameter Estimates:
##
##      Standard errors                Bootstrap
##      Number of requested bootstrap draws        1000
##      Number of successful bootstrap draws        1000
##
## Regressions:
##
##              Estimate  Std.Err  z-value  P(>|z|)  ci.lower  ci.upper
## M ~
##   X      (a)    0.193    0.048    3.991    0.000    0.095    0.283
##   v15    (d1)    0.167    0.055    3.034    0.002    0.056    0.276
##   v16    (f1)    0.155    0.026    6.063    0.000    0.105    0.204
## Y ~
##   M      (b)    0.270    0.060    4.463    0.000    0.156    0.404
##   X      (cp)   0.195    0.056    3.481    0.001    0.076    0.295
##   v15    (d2)   0.259    0.052    5.004    0.000    0.153    0.357
##   v16    (f2)   0.167    0.028    6.040    0.000    0.112    0.218
##   Std.lv  Std.all
##
##   0.193    0.206
##   0.167    0.119
##   0.155    0.220
##
##   0.270    0.256
##   0.195    0.197
##   0.259    0.175
##   0.167    0.224
##
## Variances:
##
##              Estimate  Std.Err  z-value  P(>|z|)  ci.lower  ci.upper
##   .M              0.443    0.028    15.751    0.000    0.385    0.496
##   .Y              0.419    0.028    15.070    0.000    0.360    0.466
##   Std.lv  Std.all
##   0.443    0.900
##   0.419    0.768
##
## R-Square:
##
##              Estimate
##   M              0.100
##   Y              0.232
##
## Defined Parameters:
##
##              Estimate  Std.Err  z-value  P(>|z|)  ci.lower  ci.upper
##   indirect          0.052    0.023    2.296    0.022    0.017    0.105
##   direct            0.195    0.056    3.479    0.001    0.076    0.295
##   total            0.246    0.041    5.953    0.000    0.161    0.326
##   Std.lv  Std.all

```



```
##      0.052    0.053
##      0.195    0.197
##      0.246    0.250
```

```
covariate17.W1 <- sem(model.3, data = data60_weird1, se = "bootstrap", bootstrap=1000)
summary(covariate17.W1, ci=T, standardized=T, rsquare=T, fit.measures=F)
```

```
## lavaan 0.6-9 ended normally after 31 iterations
```

```
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters          12
##
##
##      Used      Total
##      Number of observations          616      617
##
```

```
## Model Test User Model:
```

```
##
##      Test statistic          67.638
##      Degrees of freedom           3
##      P-value (Chi-square)       0.000
##
```

```
## Parameter Estimates:
```

```
##
##      Standard errors          Bootstrap
##      Number of requested bootstrap draws          1000
##      Number of successful bootstrap draws          1000
##
```

```
## Regressions:
```

```
##      Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      M ~
##      X      (a)    0.193    0.047    4.071    0.000    0.099    0.284
##      v15     (d1)    0.167    0.051    3.293    0.001    0.058    0.264
##      v16     (f1)    0.155    0.025    6.287    0.000    0.107    0.200
##      Y ~
##      M      (b)    0.270    0.060    4.479    0.000    0.158    0.391
##      X      (cp)    0.195    0.053    3.689    0.000    0.089    0.299
##      v15     (d2)    0.259    0.053    4.914    0.000    0.159    0.366
##      v16     (f2)    0.167    0.027    6.080    0.000    0.111    0.221
##      v17 ~
##      M      (g1)   -5.687    1.695   -3.356    0.001   -9.149   -2.532
##      Y      (g2)    5.074    1.587    3.197    0.001    1.946    8.267
##      Std.lv  Std.all
##
##      0.193    0.206
##      0.167    0.119
##      0.155    0.220
##
##      0.270    0.256
##      0.195    0.197
##      0.259    0.175
##      0.167    0.224
##
##      -5.687   -0.411
```

```
##      5.074      0.386
##
## Variances:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      .M              0.443   0.028  15.979   0.000   0.388   0.497
##      .Y              0.419   0.028  15.077   0.000   0.360   0.470
##      .v17            75.201  28.936   2.599   0.009  25.236  135.692
##      Std.lv Std.all
##      0.443   0.900
##      0.419   0.768
##      75.201   0.797
##
## R-Square:
##              Estimate
##      M              0.100
##      Y              0.232
##      v17            0.203
##
## Defined Parameters:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      indirect      0.052   0.022   2.367   0.018   0.018   0.106
##      direct        0.195   0.053   3.687   0.000   0.089   0.299
##      total         0.246   0.040   6.158   0.000   0.170   0.326
##      Std.lv Std.all
##      0.052   0.053
##      0.195   0.197
##      0.246   0.250
```

```
screenreg(list(mediation.W, mediation.W1),
  custom.model.name =
    c("Model 1: All Weird removed",
      "Model 2: Flatliners removed"),
  single.row = TRUE, digits = 3)
```

```
##
## =====
##              Model 1: All Weird removed  Model 2: Flatliners removed
## -----
## M ~ X          0.238 (0.044) ***          0.185 (0.049) ***
## Y ~ M          0.418 (0.060) ***          0.344 (0.061) ***
## Y ~ X          0.100 (0.055)              0.170 (0.057) **
## M ~~ M         0.443 (0.027) ***          0.473 (0.030) ***
## Y ~~ Y         0.427 (0.030) ***          0.459 (0.030) ***
## X ~~ X         0.538 (0.000)              0.560 (0.000)
## indirect := a*b 0.099 (0.028) ***          0.064 (0.026) *
## direct := cp    0.100 (0.055)              0.170 (0.057) **
## total := a*b+cp 0.200 (0.042) ***          0.234 (0.043) ***
## -----
## agfi           1.000                      1.000
## AIC            2464.546                    2565.701
## BIC            2486.630                    2587.817
## cfi            1.000                      1.000
## chisq          0.000                      0.000
## npar           5.000                      5.000
```

```
## rmsea          0.000          0.000
## rmsea.conf.high 0.000          0.000
## srmr           0.000          0.000
## tli            1.000          1.000
## converged      1            1
## ngroups        1            1
## nobs           612          616
## norig          613          617
## nexcluded      1            1
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05
```

```
screenreg(list(covariates.W, covariates.W1),
  custom.model.name =
    c("Model 1: All Weird removed",
      "Model 2: Flatliners removed"),
  single.row = TRUE, digits = 3)
```

```
##
## =====
##              Model 1: All Weird removed  Model 2: Flatliners removed
## -----
## M ~ X          0.240 (0.046) ***          0.193 (0.048) ***
## M ~ v15        0.143 (0.054) **           0.167 (0.055) **
## M ~ v16        0.143 (0.026) ***          0.155 (0.026) ***
## Y ~ M          0.344 (0.056) ***          0.270 (0.060) ***
## Y ~ X          0.123 (0.051) *            0.195 (0.056) ***
## Y ~ v15        0.276 (0.052) ***          0.259 (0.052) ***
## Y ~ v16        0.170 (0.027) ***          0.167 (0.028) ***
## M ~~ M         0.418 (0.028) ***          0.443 (0.028) ***
## Y ~~ Y         0.384 (0.026) ***          0.419 (0.028) ***
## X ~~ X         0.538 (0.000)              0.560 (0.000)
## X ~~ v15       -0.010 (0.000)             -0.016 (0.000)
## X ~~ v16       0.001 (0.000)             -0.010 (0.000)
## v15 ~~ v15     0.250 (0.000)              0.250 (0.000)
## v15 ~~ v16    -0.020 (0.000)             -0.018 (0.000)
## v16 ~~ v16     0.989 (0.000)              0.988 (0.000)
## indirect := a*b 0.083 (0.026) **          0.052 (0.023) *
## direct := cp    0.123 (0.051) *            0.195 (0.056) ***
## total := a*b+cp 0.205 (0.039) ***          0.246 (0.041) ***
## -----
## agfi           1.000          1.000
## AIC            2372.295        2477.338
## BIC            2412.045        2517.147
## cfi            1.000          1.000
## chisq          0.000          0.000
## npar           9.000          9.000
## rmsea          0.000          0.000
## rmsea.conf.high 0.000          0.000
## srmr           0.000          0.000
## tli            1.000          1.000
## converged      1            1
## ngroups        1            1
## nobs           612          616
```

```
## norig          613          617
## nexcluded      1          1
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05
```

```
screenreg(list(covariate17.W, covariate17.W1),
  custom.model.name =
    c("Model 1: All Weird removed",
      "Model 2: Flatliners removed"),
  single.row = TRUE, digits = 3)
```

```
##
## =====
##               Model 1: All Weird removed  Model 2: Flatliners removed
## -----
## M ~ X          0.240 (0.044) ***          0.193 (0.047) ***
## M ~ v15        0.143 (0.053) **           0.167 (0.051) ***
## M ~ v16        0.143 (0.024) ***           0.155 (0.025) ***
## Y ~ M          0.344 (0.058) ***           0.270 (0.060) ***
## Y ~ X          0.123 (0.053) *             0.195 (0.053) ***
## Y ~ v15        0.276 (0.052) ***           0.259 (0.053) ***
## Y ~ v16        0.170 (0.027) ***           0.167 (0.027) ***
## v17 ~ M        -3.215 (1.427) *             -5.687 (1.695) ***
## v17 ~ Y         2.804 (1.401) *              5.074 (1.587) **
## M ~~ M          0.418 (0.026) ***           0.443 (0.028) ***
## Y ~~ Y          0.384 (0.025) ***           0.419 (0.028) ***
## v17 ~~ v17     39.632 (17.123) *             75.201 (28.936) **
## X ~~ X          0.538 (0.000)              0.560 (0.000)
## X ~~ v15       -0.010 (0.000)             -0.016 (0.000)
## X ~~ v16        0.001 (0.000)             -0.010 (0.000)
## v15 ~~ v15      0.250 (0.000)              0.250 (0.000)
## v15 ~~ v16     -0.020 (0.000)             -0.018 (0.000)
## v16 ~~ v16      0.989 (0.000)              0.988 (0.000)
## indirect := a*b  0.083 (0.026) **           0.052 (0.022) *
## direct := cp     0.123 (0.053) *            0.195 (0.053) ***
## total := a*b+cp  0.205 (0.040) ***          0.246 (0.040) ***
## -----
## agfi           0.689          0.555
## AIC            6367.015       6892.688
## BIC            6420.016       6945.767
## cfi            0.888          0.847
## chisq          45.169         67.638
## npar           12.000         12.000
## rmsea          0.152          0.187
## rmsea.conf.high 0.192          0.227
## srmr           0.053          0.063
## tli            0.553          0.389
## converged      1              1
## ngroups        1              1
## nobs           612           616
## norig          613           617
## nexcluded      1              1
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05
```

##Omitted Variable Bias Weird3

```
mediation.W3 <- sem(model.1, data = data60_weird3, se = "bootstrap", bootstrap=1000)
summary(mediation.W3, ci=T, standardized=T, rsquare=T, fit.measures=F)
```

```
## lavaan 0.6-9 ended normally after 14 iterations
```

```
##
```

```
## Estimator ML
```

```
## Optimization method NLMINB
```

```
## Number of model parameters 5
```

```
##
```

```
## Used Total
```

```
## Number of observations 612 619
```

```
##
```

```
## Model Test User Model:
```

```
##
```

```
## Test statistic 0.000
```

```
## Degrees of freedom 0
```

```
##
```

```
## Parameter Estimates:
```

```
##
```

```
## Standard errors Bootstrap
```

```
## Number of requested bootstrap draws 1000
```

```
## Number of successful bootstrap draws 1000
```

```
##
```

```
## Regressions:
```

```
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
```

```
## M ~
```

```
## X (a) 0.238 0.046 5.204 0.000 0.144 0.326
```

```
## Y ~
```

```
## M (b) 0.418 0.060 6.978 0.000 0.306 0.541
```

```
## X (cp) 0.100 0.056 1.797 0.072 -0.019 0.201
```

```
## Std.lv Std.all
```

```
##
```

```
## 0.238 0.254
```

```
##
```

```
## 0.418 0.397
```

```
## 0.100 0.101
```

```
##
```

```
## Variances:
```

```
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
```

```
## .M 0.443 0.028 15.895 0.000 0.388 0.498
```

```
## .Y 0.427 0.032 13.514 0.000 0.362 0.492
```

```
## Std.lv Std.all
```

```
## 0.443 0.936
```

```
## 0.427 0.812
```

```
##
```

```
## R-Square:
```

```
## Estimate
```

```
## M 0.064
```

```
## Y 0.188
```

```
##
```

```
## Defined Parameters:
```

```
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## indirect      0.099   0.030   3.358   0.001    0.050    0.163
## direct        0.100   0.056   1.796   0.072   -0.019    0.201
## total         0.200   0.042   4.780   0.000    0.113    0.280
## Std.lv Std.all
## 0.099 0.101
## 0.100 0.101
## 0.200 0.202
```

```
covariates.W3 <- sem(model.2, data = data60_weird3, se = "bootstrap", bootstrap=1000)
summary(covariates.W3, ci=T, standardized=T, rsquare=T, fit.measures=F)
```

```
## lavaan 0.6-9 ended normally after 19 iterations
```

```
##
## Estimator ML
## Optimization method NLMINB
## Number of model parameters 9
##
## Used Total
## Number of observations 612 619
##
```

```
## Model Test User Model:
```

```
##
## Test statistic 0.000
## Degrees of freedom 0
##
```

```
## Parameter Estimates:
```

```
##
## Standard errors Bootstrap
## Number of requested bootstrap draws 1000
## Number of successful bootstrap draws 1000
##
```

```
## Regressions:
```

```
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## M ~
## X (a) 0.240 0.046 5.212 0.000 0.149 0.327
## v15 (d1) 0.143 0.052 2.731 0.006 0.043 0.250
## v16 (f1) 0.143 0.025 5.837 0.000 0.098 0.194
## Y ~
## M (b) 0.344 0.059 5.814 0.000 0.228 0.464
## X (cp) 0.123 0.053 2.319 0.020 0.017 0.222
## v15 (d2) 0.276 0.053 5.196 0.000 0.167 0.378
## v16 (f2) 0.170 0.028 6.073 0.000 0.116 0.225
## Std.lv Std.all
##
## 0.240 0.256
## 0.143 0.104
## 0.143 0.207
##
## 0.344 0.327
## 0.123 0.124
## 0.276 0.190
## 0.170 0.233
```

```
##
## Variances:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      .M           0.418   0.028  14.881   0.000   0.362   0.475
##      .Y           0.384   0.025  15.456   0.000   0.332   0.430
##      Std.lv Std.all
##      0.418   0.884
##      0.384   0.730
##
## R-Square:
##           Estimate
##      M           0.116
##      Y           0.270
##
## Defined Parameters:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      indirect      0.083   0.026   3.127   0.002   0.037   0.140
##      direct        0.123   0.053   2.318   0.020   0.017   0.222
##      total         0.205   0.039   5.307   0.000   0.128   0.281
##      Std.lv Std.all
##      0.083   0.084
##      0.123   0.124
##      0.205   0.208
```

```
covariate17.W3 <- sem(model.3, data = data60_weird3, se = "bootstrap", bootstrap=1000)
summary(covariate17.W3, ci=T, standardized=T, rsquare=T, fit.measures=F)
```

```
## lavaan 0.6-9 ended normally after 29 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters          12
##
##                               Used      Total
##      Number of observations          612      619
##
## Model Test User Model:
##
##      Test statistic          45.169
##      Degrees of freedom           3
##      P-value (Chi-square)        0.000
##
## Parameter Estimates:
##
##      Standard errors          Bootstrap
##      Number of requested bootstrap draws          1000
##      Number of successful bootstrap draws          1000
##
## Regressions:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      M ~
##      X      (a)    0.240   0.044   5.448   0.000   0.154   0.326
##      v15     (d1)   0.143   0.052   2.765   0.006   0.048   0.252
##      v16     (f1)   0.143   0.025   5.740   0.000   0.092   0.193
```

```

## Y ~
## M (b) 0.344 0.059 5.868 0.000 0.237 0.470
## X (cp) 0.123 0.053 2.294 0.022 0.006 0.220
## v15 (d2) 0.276 0.051 5.400 0.000 0.168 0.373
## v16 (f2) 0.170 0.028 6.024 0.000 0.110 0.224
## v17 ~
## M (g1) -3.215 1.398 -2.300 0.021 -5.846 -0.530
## Y (g2) 2.804 1.374 2.040 0.041 0.216 5.584
## Std.lv Std.all
##
## 0.240 0.256
## 0.143 0.104
## 0.143 0.207
##
## 0.344 0.327
## 0.123 0.124
## 0.276 0.190
## 0.170 0.233
##
## -3.215 -0.330
## 2.804 0.304
##
## Variances:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## .M 0.418 0.026 15.820 0.000 0.365 0.470
## .Y 0.384 0.026 14.951 0.000 0.328 0.430
## .v17 39.632 16.888 2.347 0.019 6.933 73.262
## Std.lv Std.all
## 0.418 0.884
## 0.384 0.730
## 39.632 0.883
##
## R-Square:
## Estimate
## M 0.116
## Y 0.270
## v17 0.117
##
## Defined Parameters:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## indirect 0.083 0.026 3.193 0.001 0.040 0.142
## direct 0.123 0.053 2.293 0.022 0.006 0.220
## total 0.205 0.041 5.024 0.000 0.125 0.281
## Std.lv Std.all
## 0.083 0.084
## 0.123 0.124
## 0.205 0.208

```

```

screenreg(list(mediation.W, mediation.W3),
  custom.model.name =
    c("Model 1: All Weird removed",
      "Model 2: Partly Item Non Response removed"),
  single.row = TRUE, digits = 3)

```



```
##
## =====
##               Model 1: All Weird removed  Model 2: Partly Item Non Response removed
## -----
## M ~ X          0.238 (0.044) ***          0.238 (0.046) ***
## Y ~ M          0.418 (0.060) ***          0.418 (0.060) ***
## Y ~ X          0.100 (0.055)              0.100 (0.056)
## M ~~ M         0.443 (0.027) ***          0.443 (0.028) ***
## Y ~~ Y         0.427 (0.030) ***          0.427 (0.032) ***
## X ~~ X         0.538 (0.000)              0.538 (0.000)
## indirect := a*b 0.099 (0.028) ***          0.099 (0.030) ***
## direct := cp    0.100 (0.055)              0.100 (0.056)
## total := a*b+cp 0.200 (0.042) ***          0.200 (0.042) ***
## -----
## agfi           1.000                      1.000
## AIC            2464.546                  2464.546
## BIC            2486.630                  2486.630
## cfi            1.000                      1.000
## chisq          0.000                      0.000
## npar           5.000                      5.000
## rmsea          0.000                      0.000
## rmsea.conf.high 0.000                      0.000
## srmr           0.000                      0.000
## tli            1.000                      1.000
## converged      1                        1
## ngroups        1                        1
## nobs           612                      612
## norig          613                      619
## nexcluded      1                        7
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05
```

```
screenreg(list(covariates.W, covariates.W3),
  custom.model.name =
    c("Model 1: All Weird removed",
      "Model 2: Partly Item Non Response removed"),
  single.row = TRUE, digits = 3)
```

```
##
## =====
##               Model 1: All Weird removed  Model 2: Partly Item Non Response removed
## -----
## M ~ X          0.240 (0.046) ***          0.240 (0.046) ***
## M ~ v15        0.143 (0.054) **           0.143 (0.052) **
## M ~ v16        0.143 (0.026) ***          0.143 (0.025) ***
## Y ~ M          0.344 (0.056) ***          0.344 (0.059) ***
## Y ~ X          0.123 (0.051) *            0.123 (0.053) *
## Y ~ v15        0.276 (0.052) ***          0.276 (0.053) ***
## Y ~ v16        0.170 (0.027) ***          0.170 (0.028) ***
## M ~~ M         0.418 (0.028) ***          0.418 (0.028) ***
## Y ~~ Y         0.384 (0.026) ***          0.384 (0.025) ***
## X ~~ X         0.538 (0.000)              0.538 (0.000)
## X ~~ v15       -0.010 (0.000)             -0.010 (0.000)
## X ~~ v16       0.001 (0.000)              0.001 (0.000)
```

```
## v15 ~~ v15          0.250 (0.000)          0.250 (0.000)
## v15 ~~ v16         -0.020 (0.000)         -0.020 (0.000)
## v16 ~~ v16          0.989 (0.000)          0.989 (0.000)
## indirect := a*b      0.083 (0.026) **       0.083 (0.026) **
## direct := cp         0.123 (0.051) *        0.123 (0.053) *
## total := a*b+cp      0.205 (0.039) ***      0.205 (0.039) ***
## -----
## agfi                 1.000                 1.000
## AIC                 2372.295                2372.295
## BIC                 2412.045                2412.045
## cfi                 1.000                 1.000
## chisq               0.000                 0.000
## npar                9.000                 9.000
## rmsea               0.000                 0.000
## rmsea.conf.high     0.000                 0.000
## srmr                0.000                 0.000
## tli                 1.000                 1.000
## converged           1                     1
## ngroups             1                     1
## nobs                612                   612
## norig               613                   619
## nexcluded           1                     7
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05
```

```
screenreg(list(covariate17.W, covariate17.W3),
  custom.model.name =
    c("Model 1: All Weird removed",
      "Model 2: Partly Item Non Response removed"),
  single.row = TRUE, digits = 3)
```

```
##
## =====
##               Model 1: All Weird removed  Model 2: Partly Item Non Response removed
## -----
## M ~ X          0.240 (0.044) ***          0.240 (0.044) ***
## M ~ v15        0.143 (0.053) **           0.143 (0.052) **
## M ~ v16        0.143 (0.024) ***           0.143 (0.025) ***
## Y ~ M          0.344 (0.058) ***           0.344 (0.059) ***
## Y ~ X          0.123 (0.053) *             0.123 (0.053) *
## Y ~ v15        0.276 (0.052) ***           0.276 (0.051) ***
## Y ~ v16        0.170 (0.027) ***           0.170 (0.028) ***
## v17 ~ M        -3.215 (1.427) *             -3.215 (1.398) *
## v17 ~ Y        2.804 (1.401) *             2.804 (1.374) *
## M ~~ M         0.418 (0.026) ***           0.418 (0.026) ***
## Y ~~ Y         0.384 (0.025) ***           0.384 (0.026) ***
## v17 ~~ v17     39.632 (17.123) *            39.632 (16.888) *
## X ~~ X         0.538 (0.000)              0.538 (0.000)
## X ~~ v15       -0.010 (0.000)             -0.010 (0.000)
## X ~~ v16       0.001 (0.000)              0.001 (0.000)
## v15 ~~ v15     0.250 (0.000)              0.250 (0.000)
## v15 ~~ v16     -0.020 (0.000)             -0.020 (0.000)
## v16 ~~ v16     0.989 (0.000)              0.989 (0.000)
## indirect := a*b 0.083 (0.026) **           0.083 (0.026) **
```

```

## direct := cp      0.123 (0.053) *      0.123 (0.053) *
## total := a*b+cp   0.205 (0.040) ***    0.205 (0.041) ***
## -----
## agfi              0.689                0.689
## AIC                6367.015            6367.015
## BIC                6420.016            6420.016
## cfi                0.888                0.888
## chisq              45.169              45.169
## npar               12.000              12.000
## rmsea              0.152                0.152
## rmsea.conf.high    0.192                0.192
## srmr               0.053                0.053
## tli                0.553                0.553
## converged          1                    1
## ngroups            1                    1
## nobs               612                  612
## norig              613                  619
## nexcluded          1                    7
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05

```