

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 Variables

##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
```

```
#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)])
```

```
#Weird Variables

dfX <- data.frame (first_column = (data60_cl$x1),
                   second_column = (data60_cl$x2),
                   third_column = (data60_cl$x3),
                   fourth_column = (data60_cl$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_cl$m1),
                   second_column = (data60_cl$m2),
                   third_column = (data60_cl$m3),
                   fourth_column = (data60_cl$m4),
                   fifth_column = (data60_cl$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
```

```
#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 16 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)    0.327   0.039   8.436   0.000    0.250    0.409
## Y ~
##   M      (b)    0.394   0.077   5.108   0.000    0.251    0.552
##   X      (cp)   0.105   0.065   1.603   0.109   -0.029    0.225
## Std.lv Std.all
##
##   0.327   0.408
##
##   0.394   0.321
##   0.105   0.106
##
## Variances:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## .M          0.301   0.019  16.147   0.000    0.263    0.338
## .Y          0.468   0.030  15.517   0.000    0.404    0.526
## Std.lv Std.all
##   0.301   0.834
##   0.468   0.858
##
## R-Square:
##           Estimate
## M          0.166
## Y          0.142
##
## Defined Parameters:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## indirect    0.129   0.037   3.522   0.000    0.068    0.215
## direct      0.105   0.065   1.603   0.109   -0.029    0.225
## total       0.234   0.042   5.556   0.000    0.149    0.310
## Std.lv Std.all
##   0.129   0.131
##   0.105   0.106
##   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 20 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) 0.333 0.040 8.366 0.000 0.258 0.413
## v15 (d1) 0.126 0.043 2.930 0.003 0.043 0.209
## v16 (f1) 0.119 0.020 6.039 0.000 0.079 0.157
## Y ~
## M (b) 0.302 0.073 4.158 0.000 0.162 0.459
## X (cp) 0.146 0.063 2.322 0.020 0.012 0.267
## v15 (d2) 0.266 0.052 5.161 0.000 0.159 0.360
## v16 (f2) 0.173 0.027 6.389 0.000 0.116 0.227
## Std.lv Std.all
##
## 0.333 0.415
## 0.126 0.105
## 0.119 0.197
##
## 0.302 0.246
## 0.146 0.148
## 0.266 0.180
## 0.173 0.232
##
## Variances:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## .M 0.284 0.018 15.853 0.000 0.248 0.318
## .Y 0.425 0.027 15.660 0.000 0.369 0.474
## Std.lv Std.all
## 0.284 0.786
## 0.425 0.780
##
## R-Square:
## Estimate
## M 0.214
## Y 0.220
##
## Defined Parameters:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper

```

```
##      indirect      0.101    0.033    3.052    0.002    0.047    0.178
##      direct       0.146    0.063    2.321    0.020    0.012    0.267
##      total        0.246    0.041    6.009    0.000    0.163    0.328
## Std.lv Std.all
##      0.101    0.102
##      0.146    0.148
##      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 32 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters    12
##
##                                     Used      Total
##      Number of observations        616        623
##
## Model Test User Model:
##
##      Test statistic                104.945
##      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)   0.333   0.037   9.062   0.000   0.259   0.405
##      v15     (d1)  0.126   0.042   2.979   0.003   0.047   0.216
##      v16     (f1)  0.119   0.021   5.744   0.000   0.079   0.160
##      Y ~
##      M      (b)   0.302   0.075   4.012   0.000   0.163   0.461
##      X      (cp)  0.146   0.064   2.265   0.024   0.021   0.261
##      v15     (d2)  0.266   0.052   5.089   0.000   0.166   0.375
##      v16     (f2)  0.173   0.028   6.154   0.000   0.116   0.227
##      v17 ~
##      M      (g1) -5.151   1.659  -3.104   0.002  -8.528  -1.923
```

```

##      Y      (g2)      4.650      1.583      2.938      0.003      1.591      7.683
##      Std.lv Std.all
##
##      0.333      0.415
##      0.126      0.105
##      0.119      0.197
##
##      0.302      0.246
##      0.146      0.148
##      0.266      0.180
##      0.173      0.232
##
##      -5.151     -0.319
##      4.650      0.354
##
## Variances:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      .M              0.284   0.018  15.691   0.000    0.245    0.316
##      .Y              0.425   0.027  16.008   0.000    0.368    0.473
##      .v17            80.743  32.466   2.487   0.013   25.038   149.576
##      Std.lv Std.all
##      0.284      0.786
##      0.425      0.780
##      80.743     0.855
##
## R-Square:
##              Estimate
##      M              0.214
##      Y              0.220
##      v17            0.145
##
## Defined Parameters:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      indirect      0.101   0.033   3.059   0.002    0.047    0.175
##      direct        0.146   0.064   2.264   0.024    0.021    0.261
##      total          0.246   0.042   5.833   0.000    0.164    0.326
##      Std.lv Std.all
##      0.101      0.102
##      0.146      0.148
##      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 17 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)   0.370   0.036  10.407   0.000   0.297   0.440
##      Y ~
##      M      (b)   0.488   0.071   6.913   0.000   0.353   0.625
##      X      (cp)  0.019   0.064   0.294   0.769  -0.120   0.138
##      Std.lv Std.all
##
##      0.370   0.456
##
##      0.488   0.401

```



```
##      0.019      0.019
##
## Variances:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      .M              0.281   0.017  16.672   0.000   0.246   0.313
##      .Y              0.438   0.029  15.269   0.000   0.380   0.491
##      Std.lv Std.all
##      0.281   0.792
##      0.438   0.832
##
## R-Square:
##              Estimate
##      M              0.208
##      Y              0.168
##
## Defined Parameters:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      indirect      0.181   0.038   4.755   0.000   0.113   0.262
##      direct        0.019   0.064   0.294   0.769  -0.120   0.138
##      total         0.200   0.042   4.794   0.000   0.116   0.275
##      Std.lv Std.all
##      0.181   0.183
##      0.019   0.019
##      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 20 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)      0.372   0.036  10.340   0.000   0.301   0.440
##      v15     (d1)     0.106   0.044   2.430   0.015   0.015   0.193
##      v16     (f1)     0.109   0.020   5.494   0.000   0.070   0.149
```

```

## Y ~
## M (b) 0.397 0.071 5.565 0.000 0.263 0.541
## X (cp) 0.058 0.062 0.927 0.354 -0.074 0.177
## v15 (d2) 0.283 0.054 5.215 0.000 0.181 0.395
## v16 (f2) 0.176 0.026 6.718 0.000 0.123 0.229
## Std.lv Std.all
##
## 0.372 0.459
## 0.106 0.089
## 0.109 0.182
##
## 0.397 0.326
## 0.058 0.058
## 0.283 0.195
## 0.176 0.241
##
## Variances:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## .M 0.267 0.017 15.541 0.000 0.231 0.300
## .Y 0.391 0.026 15.260 0.000 0.336 0.441
## Std.lv Std.all
## 0.267 0.752
## 0.391 0.744
##
## R-Square:
## Estimate
## M 0.248
## Y 0.256
##
## Defined Parameters:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## indirect 0.148 0.036 4.098 0.000 0.084 0.225
## direct 0.058 0.062 0.927 0.354 -0.074 0.177
## total 0.205 0.040 5.196 0.000 0.123 0.285
## Std.lv Std.all
## 0.148 0.150
## 0.058 0.058
## 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 33 iterations
##
## Estimator ML
## Optimization method NLMINB
## Number of model parameters 12
##
## Used Total
## Number of observations 612 613
##
## Model Test User Model:
##
## Test statistic 67.784

```

```

## 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)    0.372   0.036  10.396   0.000   0.302   0.441
##   v15     (d1)   0.106   0.043   2.495   0.013   0.024   0.193
##   v16     (f1)   0.109   0.020   5.537   0.000   0.072   0.148
## Y ~
##   M      (b)    0.397   0.072   5.501   0.000   0.259   0.534
##   X      (cp)   0.058   0.062   0.930   0.352  -0.065   0.174
##   v15     (d2)   0.283   0.053   5.340   0.000   0.175   0.382
##   v16     (f2)   0.176   0.028   6.166   0.000   0.123   0.233
## v17 ~
##   M      (g1)  -2.750   1.391  -1.977   0.048  -5.778  -0.413
##   Y      (g2)   2.440   1.367   1.785   0.074   0.061   5.432
## Std.lv Std.all
##
##   0.372   0.459
##   0.106   0.089
##   0.109   0.182
##
##   0.397   0.326
##   0.058   0.058
##   0.283   0.195
##   0.176   0.241
##
##  -2.750  -0.245
##   2.440   0.264
##
## Variances:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##   .M           0.267   0.017  15.901   0.000   0.234   0.301
##   .Y           0.391   0.026  14.768   0.000   0.338   0.443
##   .v17        41.420  19.753   2.097   0.036   7.464  85.065
## Std.lv Std.all
##   0.267   0.752
##   0.391   0.744
##  41.420   0.923
##
## R-Square:
##           Estimate
##   M           0.248
##   Y           0.256
##   v17         0.077
##
## Defined Parameters:

```

```
##               Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## indirect      0.148    0.037   4.045  0.000    0.083    0.221
## direct       0.058    0.062   0.929  0.353   -0.065    0.174
## total        0.205    0.040   5.165  0.000    0.125    0.286
## Std.lv Std.all
## 0.148 0.150
## 0.058 0.058
## 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          0.327 (0.039) ***    0.370 (0.036) ***
## Y ~ M          0.394 (0.077) ***    0.488 (0.071) ***
## Y ~ X          0.105 (0.065)          0.019 (0.064)
## M ~~ M         0.301 (0.019) ***    0.281 (0.017) ***
## Y ~~ Y         0.468 (0.030) ***    0.438 (0.029) ***
## X ~~ X         0.560 (0.000)          0.538 (0.000)
## indirect := a*b 0.129 (0.037) ***    0.181 (0.038) ***
## direct := cp    0.105 (0.065)          0.019 (0.064)
## total := a*b+cp 0.234 (0.042) ***    0.200 (0.042) ***
## -----
## agfi           1.000                  1.000
## AIC            2299.374                2200.727
## BIC            2321.491                2222.811
## 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           0.333 (0.040) ***       0.372 (0.036) ***
## M ~ v15         0.126 (0.043) **        0.106 (0.044) *
## M ~ v16         0.119 (0.020) ***       0.109 (0.020) ***
## Y ~ M           0.302 (0.073) ***       0.397 (0.071) ***
## Y ~ X           0.146 (0.063) *         0.058 (0.062)
## Y ~ v15         0.266 (0.052) ***       0.283 (0.054) ***
## Y ~ v16         0.173 (0.027) ***       0.176 (0.026) ***
## M ~~ M          0.284 (0.018) ***       0.267 (0.017) ***
## Y ~~ Y          0.425 (0.027) ***       0.391 (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  0.101 (0.033) **       0.148 (0.036) ***
## direct := cp    0.146 (0.063) *        0.058 (0.062)
## total := a*b+cp 0.246 (0.041) ***       0.205 (0.040) ***
## -----
## agfi            1.000                   1.000
## AIC             2211.480                 2108.802
## BIC             2251.290                 2148.553
## 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          0.333 (0.037) ***    0.372 (0.036) ***
## M ~ v15        0.126 (0.042) **     0.106 (0.043) *
## M ~ v16        0.119 (0.021) ***    0.109 (0.020) ***
## Y ~ M          0.302 (0.075) ***    0.397 (0.072) ***
## Y ~ X          0.146 (0.064) *      0.058 (0.062)
## Y ~ v15        0.266 (0.052) ***    0.283 (0.053) ***
## Y ~ v16        0.173 (0.028) ***    0.176 (0.028) ***
## v17 ~ M        -5.151 (1.659) **    -2.750 (1.391) *
## v17 ~ Y         4.650 (1.583) **     2.440 (1.367)
## M ~~ M         0.284 (0.018) ***    0.267 (0.017) ***
## Y ~~ Y         0.425 (0.027) ***    0.391 (0.026) ***
## v17 ~~ v17     80.743 (32.466) *     41.420 (19.753) *
## 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 0.101 (0.033) **     0.148 (0.037) ***
## direct := cp    0.146 (0.064) *      0.058 (0.062)
## total := a*b+cp 0.246 (0.042) ***    0.205 (0.040) ***
## -----
## agfi           0.364                 0.563
## AIC            6670.633              6130.530
## BIC            6723.712              6183.530
## cfi            0.792                 0.859
## chisq          104.945              67.784
## npar           12.000               12.000
## rmsea          0.235                 0.188
## rmsea.conf.high 0.274                 0.228
## srmr           0.076                 0.061
## tli            0.170                 0.436
## 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 16 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)    0.327   0.039   8.362   0.000    0.248    0.402
## Y ~
##   M      (b)    0.394   0.076   5.183   0.000    0.248    0.551
##   X      (cp)   0.105   0.067   1.575   0.115   -0.029    0.224
##   Std.lv Std.all
##
##   0.327   0.408
##
##   0.394   0.321
##   0.105   0.106
##
## Variances:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##   .M          0.301   0.019  15.880   0.000    0.264    0.337
##   .Y          0.468   0.030  15.764   0.000    0.408    0.522
##   Std.lv Std.all
##   0.301   0.834
##   0.468   0.858
##
## R-Square:
##           Estimate
##   M          0.166
##   Y          0.142
##
## Defined Parameters:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##   indirect    0.129   0.036   3.603   0.000    0.064    0.213
##   direct      0.105   0.067   1.574   0.116   -0.029    0.224
##   total       0.234   0.044   5.308   0.000    0.146    0.323
##   Std.lv Std.all
##   0.129   0.131
##   0.105   0.106
##   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 20 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) 0.333 0.040 8.358 0.000 0.252 0.409
## v15 (d1) 0.126 0.043 2.961 0.003 0.050 0.208
## v16 (f1) 0.119 0.020 5.947 0.000 0.079 0.160
## Y ~
## M (b) 0.302 0.074 4.098 0.000 0.165 0.458
## X (cp) 0.146 0.065 2.246 0.025 0.011 0.270
## v15 (d2) 0.266 0.054 4.969 0.000 0.170 0.377
## v16 (f2) 0.173 0.027 6.280 0.000 0.119 0.227
## Std.lv Std.all
##
## 0.333 0.415
## 0.126 0.105
## 0.119 0.197
##
## 0.302 0.246
## 0.146 0.148
## 0.266 0.180
## 0.173 0.232
##
## Variances:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## .M 0.284 0.018 15.796 0.000 0.246 0.318
## .Y 0.425 0.028 15.055 0.000 0.366 0.474
## Std.lv Std.all
## 0.284 0.786
## 0.425 0.780
##
## R-Square:
## Estimate
## M 0.214
## Y 0.220
##
## Defined Parameters:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper

```



```
##      indirect      0.101    0.033    3.028    0.002    0.045    0.175
##      direct       0.146    0.065    2.245    0.025    0.011    0.270
##      total        0.246    0.042    5.809    0.000    0.165    0.329
## Std.lv Std.all
##    0.101  0.102
##    0.146  0.148
##    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 32 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters      12
##
##                                     Used      Total
##      Number of observations          616        617
##
## Model Test User Model:
##
##      Test statistic                104.945
##      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)    0.333    0.039    8.454    0.000    0.262    0.413
## v15    (d1)    0.126    0.043    2.925    0.003    0.039    0.210
## v16    (f1)    0.119    0.021    5.717    0.000    0.081    0.163
## Y ~
## M      (b)    0.302    0.078    3.879    0.000    0.171    0.474
## X      (cp)    0.146    0.065    2.238    0.025   -0.006    0.257
## v15    (d2)    0.266    0.053    5.025    0.000    0.165    0.373
## v16    (f2)    0.173    0.028    6.145    0.000    0.120    0.227
## v17 ~
## M      (g1)   -5.151    1.710   -3.012    0.003   -8.344   -1.723
## Y      (g2)    4.650    1.625    2.861    0.004    1.381    7.638
## Std.lv Std.all
##
##    0.333  0.415
##    0.126  0.105
##    0.119  0.197
##
##    0.302  0.246
##    0.146  0.148
```

```

##      0.266      0.180
##      0.173      0.232
##
##     -5.151     -0.319
##      4.650      0.354
##
## Variances:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      .M              0.284   0.018  15.679   0.000   0.248   0.317
##      .Y              0.425   0.028  15.442   0.000   0.364   0.476
##      .v17            80.743  32.368   2.495   0.013  21.330  148.926
##      Std.lv Std.all
##      0.284   0.786
##      0.425   0.780
##      80.743   0.855
##
## R-Square:
##              Estimate
##      M              0.214
##      Y              0.220
##      v17            0.145
##
## Defined Parameters:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      indirect       0.101   0.035   2.846   0.004   0.050   0.188
##      direct         0.146   0.065   2.237   0.025  -0.006   0.257
##      total          0.246   0.041   5.946   0.000   0.166   0.324
##      Std.lv Std.all
##      0.101   0.102
##      0.146   0.148
##      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          0.370 (0.036) ***          0.327 (0.039) ***
## Y ~ M          0.488 (0.071) ***          0.394 (0.076) ***
## Y ~ X          0.019 (0.064)              0.105 (0.067)
## M ~~ M         0.281 (0.017) ***          0.301 (0.019) ***
## Y ~~ Y         0.438 (0.029) ***          0.468 (0.030) ***
## X ~~ X         0.538 (0.000)              0.560 (0.000)
## indirect := a*b 0.181 (0.038) ***          0.129 (0.036) ***
## direct := cp    0.019 (0.064)              0.105 (0.067)
## total := a*b+cp 0.200 (0.042) ***          0.234 (0.044) ***
## -----
## agfi           1.000                      1.000
## AIC            2200.727                  2299.374

```

```
## BIC                2222.811                2321.491
## 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          0.372 (0.036) ***          0.333 (0.040) ***
## M ~ v15        0.106 (0.044) *            0.126 (0.043) **
## M ~ v16        0.109 (0.020) ***          0.119 (0.020) ***
## Y ~ M          0.397 (0.071) ***          0.302 (0.074) ***
## Y ~ X          0.058 (0.062)              0.146 (0.065) *
## Y ~ v15        0.283 (0.054) ***          0.266 (0.054) ***
## Y ~ v16        0.176 (0.026) ***          0.173 (0.027) ***
## M ~~ M         0.267 (0.017) ***          0.284 (0.018) ***
## Y ~~ Y         0.391 (0.026) ***          0.425 (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 0.148 (0.036) ***          0.101 (0.033) **
## direct := cp    0.058 (0.062)              0.146 (0.065) *
## total := a*b+cp 0.205 (0.040) ***          0.246 (0.042) ***
## -----
## agfi           1.000                      1.000
## AIC            2108.802                    2211.480
## BIC            2148.553                    2251.290
## 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 0.372 (0.036) *** 0.333 (0.039) ***
## M ~ v15 0.106 (0.043) * 0.126 (0.043) **
## M ~ v16 0.109 (0.020) *** 0.119 (0.021) ***
## Y ~ M 0.397 (0.072) *** 0.302 (0.078) ***
## Y ~ X 0.058 (0.062) 0.146 (0.065) *
## Y ~ v15 0.283 (0.053) *** 0.266 (0.053) ***
## Y ~ v16 0.176 (0.028) *** 0.173 (0.028) ***
## v17 ~ M -2.750 (1.391) * -5.151 (1.710) **
## v17 ~ Y 2.440 (1.367) 4.650 (1.625) **
## M ~~ M 0.267 (0.017) *** 0.284 (0.018) ***
## Y ~~ Y 0.391 (0.026) *** 0.425 (0.028) ***
## v17 ~~ v17 41.420 (19.753) * 80.743 (32.368) *
## 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 0.148 (0.037) *** 0.101 (0.035) **
## direct := cp 0.058 (0.062) 0.146 (0.065) *
## total := a*b+cp 0.205 (0.040) *** 0.246 (0.041) ***
## -----
## agfi 0.563 0.364
## AIC 6130.530 6670.633
## BIC 6183.530 6723.712
## cfi 0.859 0.792
## chisq 67.784 104.945
## npar 12.000 12.000
## rmsea 0.188 0.235
## rmsea.conf.high 0.228 0.274
## srmr 0.061 0.076
## tli 0.436 0.170
## 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 17 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)    0.370    0.036   10.157   0.000    0.296    0.442
##      Y ~
##      M      (b)    0.488    0.074    6.624   0.000    0.352    0.637
##      X      (cp)   0.019    0.066    0.288   0.773   -0.125    0.140
##      Std.lv Std.all
##
##      0.370    0.456
##
##      0.488    0.401
##      0.019    0.019
##
## Variances:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      .M      0.281    0.017   16.401   0.000    0.245    0.316
##      .Y      0.438    0.031   14.264   0.000    0.378    0.498
##      Std.lv Std.all
##      0.281    0.792
##      0.438    0.832
##
## R-Square:
```

```
##           Estimate
##      M           0.208
##      Y           0.168
##
## Defined Parameters:
##           Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      indirect      0.181   0.039   4.574   0.000   0.111   0.269
##      direct        0.019   0.066   0.288   0.773  -0.125   0.140
##      total         0.200   0.043   4.688   0.000   0.115   0.281
## Std.lv Std.all
##    0.181  0.183
##    0.019  0.019
##    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 20 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)    0.372   0.037  10.127   0.000   0.291   0.439
##      v15     (d1)   0.106   0.041   2.592   0.010   0.027   0.184
##      v16     (f1)   0.109   0.019   5.637   0.000   0.069   0.148
##      Y ~
##      M      (b)    0.397   0.072   5.532   0.000   0.254   0.548
##      X      (cp)   0.058   0.064   0.900   0.368  -0.079   0.177
##      v15     (d2)   0.283   0.052   5.472   0.000   0.183   0.389
##      v16     (f2)   0.176   0.026   6.655   0.000   0.122   0.224
## Std.lv Std.all
##
##    0.372  0.459
##    0.106  0.089
##    0.109  0.182
##
```

```

##      0.397      0.326
##      0.058      0.058
##      0.283      0.195
##      0.176      0.241
##
## Variances:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      .M              0.267   0.017  15.265   0.000   0.231   0.299
##      .Y              0.391   0.026  14.778   0.000   0.335   0.443
##      Std.lv Std.all
##      0.267   0.752
##      0.391   0.744
##
## R-Square:
##              Estimate
##      M              0.248
##      Y              0.256
##
## Defined Parameters:
##              Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
##      indirect      0.148   0.036   4.049   0.000   0.083   0.230
##      direct        0.058   0.064   0.899   0.369  -0.079   0.177
##      total         0.205   0.042   4.866   0.000   0.114   0.283
##      Std.lv Std.all
##      0.148   0.150
##      0.058   0.058
##      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 33 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters          12
##
##                                Used      Total
##      Number of observations          612      619
##
## Model Test User Model:
##
##      Test statistic          67.784
##      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) 0.372 0.037 10.183 0.000 0.299 0.440
## v15 (d1) 0.106 0.042 2.531 0.011 0.024 0.192
## v16 (f1) 0.109 0.019 5.652 0.000 0.070 0.147
## Y ~
## M (b) 0.397 0.071 5.580 0.000 0.263 0.547
## X (cp) 0.058 0.063 0.918 0.359 -0.071 0.179
## v15 (d2) 0.283 0.051 5.579 0.000 0.183 0.378
## v16 (f2) 0.176 0.027 6.567 0.000 0.122 0.225
## v17 ~
## M (g1) -2.750 1.380 -1.993 0.046 -5.499 -0.153
## Y (g2) 2.440 1.362 1.792 0.073 -0.133 5.114
## Std.lv Std.all
##
## 0.372 0.459
## 0.106 0.089
## 0.109 0.182
##
## 0.397 0.326
## 0.058 0.058
## 0.283 0.195
## 0.176 0.241
##
## -2.750 -0.245
## 2.440 0.264
##
## Variances:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## .M 0.267 0.017 15.688 0.000 0.233 0.300
## .Y 0.391 0.024 16.079 0.000 0.339 0.435
## .v17 41.420 19.207 2.157 0.031 6.770 79.628
## Std.lv Std.all
## 0.267 0.752
## 0.391 0.744
## 41.420 0.923
##
## R-Square:
## Estimate
## M 0.248
## Y 0.256
## v17 0.077
##
## Defined Parameters:
## Estimate Std.Err z-value P(>|z|) ci.lower ci.upper
## indirect 0.148 0.036 4.071 0.000 0.086 0.230
## direct 0.058 0.063 0.917 0.359 -0.071 0.179
## total 0.205 0.041 5.016 0.000 0.124 0.277
## Std.lv Std.all
## 0.148 0.150
## 0.058 0.058
## 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          0.370 (0.036) ***          0.370 (0.036) ***
## Y ~ M          0.488 (0.071) ***          0.488 (0.074) ***
## Y ~ X          0.019 (0.064)              0.019 (0.066)
## M ~~ M         0.281 (0.017) ***          0.281 (0.017) ***
## Y ~~ Y         0.438 (0.029) ***          0.438 (0.031) ***
## X ~~ X         0.538 (0.000)              0.538 (0.000)
## indirect := a*b 0.181 (0.038) ***          0.181 (0.039) ***
## direct := cp    0.019 (0.064)              0.019 (0.066)
## total := a*b+cp 0.200 (0.042) ***          0.200 (0.043) ***
## -----
## agfi           1.000                      1.000
## AIC            2200.727                  2200.727
## BIC            2222.811                  2222.811
## 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          0.372 (0.036) ***          0.372 (0.037) ***
## M ~ v15        0.106 (0.044) *            0.106 (0.041) **
## M ~ v16        0.109 (0.020) ***          0.109 (0.019) ***
## Y ~ M          0.397 (0.071) ***          0.397 (0.072) ***
## Y ~ X          0.058 (0.062)              0.058 (0.064)
## Y ~ v15        0.283 (0.054) ***          0.283 (0.052) ***

```

```

## Y ~ v16          0.176 (0.026) ***          0.176 (0.026) ***
## M ~~ M           0.267 (0.017) ***          0.267 (0.017) ***
## Y ~~ Y           0.391 (0.026) ***          0.391 (0.026) ***
## 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   0.148 (0.036) ***          0.148 (0.036) ***
## direct := cp      0.058 (0.062)            0.058 (0.064)
## total := a*b+cp   0.205 (0.040) ***          0.205 (0.042) ***
## -----
## agfi              1.000                    1.000
## AIC                2108.802                2108.802
## BIC                2148.553                2148.553
## 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          0.372 (0.036) ***          0.372 (0.037) ***
## M ~ v15        0.106 (0.043) *           0.106 (0.042) *
## M ~ v16        0.109 (0.020) ***          0.109 (0.019) ***
## Y ~ M          0.397 (0.072) ***          0.397 (0.071) ***
## Y ~ X          0.058 (0.062)              0.058 (0.063)
## Y ~ v15        0.283 (0.053) ***          0.283 (0.051) ***
## Y ~ v16        0.176 (0.028) ***          0.176 (0.027) ***
## v17 ~ M        -2.750 (1.391) *           -2.750 (1.380) *
## v17 ~ Y        2.440 (1.367)              2.440 (1.362)
## M ~~ M         0.267 (0.017) ***          0.267 (0.017) ***
## Y ~~ Y         0.391 (0.026) ***          0.391 (0.024) ***
## v17 ~~ v17     41.420 (19.753) *           41.420 (19.207) *
## 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     0.148 (0.037) ***       0.148 (0.036) ***
## direct := cp        0.058 (0.062)          0.058 (0.063)
## total := a*b+cp     0.205 (0.040) ***       0.205 (0.041) ***
## -----
## agfi                0.563                  0.563
## AIC                 6130.530                6130.530
## BIC                 6183.530                6183.530
## cfi                 0.859                  0.859
## chisq               67.784                 67.784
## npar                12.000                 12.000
## rmsea               0.188                  0.188
## rmsea.conf.high     0.228                  0.228
## srmr                0.061                  0.061
## tli                 0.436                  0.436
## converged           1                     1
## ngroups             1                     1
## nobs                612                   612
## norig               613                   619
## nexcluded           1                     7
## =====
## *** p < 0.001; ** p < 0.01; * p < 0.05

```