

SEM with Missing Data - Working Examples

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Outline

- 1 The Demonstration Data
- 2 Listwise Deletion in lavaan
- 3 Pairwise Deletion in lavaan
- 4 FIML in lavaan
- 5 No FIML in lavaan for Categorical Data
- 6 FIML in Mplus
- 7 Multiple Imputation

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The HBSC data set

- A subset of the HBSC data (Health Behaviour in School-Aged Children; Iannotti, 2005-200) will be used for demonstrating the missing data techniques implemented in lavaan (Rosseel, 2012) and Mplus (Muthén & Muthén, 2017)
- We will use responses from the 6th and 7th graders
- For demonstration purpose, we will artificially increase the missing percentage on the focal variables (this practice is considered MCAR)

```
## Loading packages
library(lavaan)
```

```
## The "data" folder contains the hbsc data
ddir <- "data"
hbsc.complete <- readRDS(file.path(ddir, "hbsc.rds"))
## hbsc contains grades 6 and 7 responses
hbsc <- hbsc.complete[hbsc.complete$Grade %in% c("6", "7"), ]
```

The HBCS data set ...

```
## The dimension of the data set "hbsc"
dim(hbsc)
```

```
[1] 4284 89
```

```
## Variables in our demonstration data "hbsc"
names(hbsc)
```

```
[1] "stud_id" "sch1_id" "Gender" "Age" "Grade" "body1_o" "body2_o" "body3_o"
[9] "body4_o" "body5_o" "phys1_o" "phys2_o" "phys3_o" "phys4_o" "phys5_o" "phys6_o"
[17] "phys7_o" "phys8_o" "depre1_o" "depre2_o" "depre3_o" "depre4_o" "depre5_o" "depre6_o"
[25] "gotBu1_o" "gotBu2_o" "gotBu3_o" "gotBu4_o" "gotBu5_o" "gotBu6_o" "gotBu7_o" "gotBu8_o"
[33] "gotBu9_o" "bu0th1_o" "bu0th2_o" "bu0th3_o" "bu0th4_o" "bu0th5_o" "bu0th6_o" "bu0th7_o"
[41] "bu0th8_o" "bu0th9_o" "alc1_o" "alc2_o" "alc3_o" "alc4_o" "alc5_o" "body1_i"
[49] "body2_i" "body3_i" "body4_i" "body5_i" "phys1_i" "phys2_i" "phys3_i" "phys4_i"
[57] "phys5_i" "phys6_i" "phys7_i" "phys8_i" "depre1_i" "depre2_i" "depre3_i" "depre4_i"
[65] "depre5_i" "depre6_i" "gotBu1_i" "gotBu2_i" "gotBu3_i" "gotBu4_i" "gotBu5_i" "gotBu6_i"
[73] "gotBu7_i" "gotBu8_i" "gotBu9_i" "bu0th1_i" "bu0th2_i" "bu0th3_i" "bu0th4_i" "bu0th5_i"
[81] "bu0th6_i" "bu0th7_i" "bu0th8_i" "bu0th9_i" "alc1_i" "alc2_i" "alc3_i" "alc4_i"
[89] "alc5_i"
```

The HBSC data set ...

```
## Creating a name vector for the depre items
varName_depre_i <- c("depre1_i", "depre2_i", "depre3_i",
  "depre4_i", "depre5_i", "depre6_i")
## Requesting summary information for the Depression items
summary(hbhc[ , varName_depre_i])
```

	depre1_i	depre2_i	depre3_i	depre4_i	depre5_i	depre6_i
Min.	:1.000	Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.000
1st Qu.:	:1.000	1st Qu. :2.000	1st Qu. :1.000	1st Qu. :1.000	1st Qu. :1.000	1st Qu. :1.000
Median	:2.000	Median :3.000	Median :1.000	Median :2.000	Median :2.000	Median :2.000
Mean	:2.332	Mean :2.684	Mean :1.817	Mean :2.209	Mean :2.485	Mean :2.473
3rd Qu.:	:3.000	3rd Qu. :3.000	3rd Qu. :3.000	3rd Qu. :3.000	3rd Qu. :4.000	3rd Qu. :3.000
Max.	:5.000	Max. :5.000	Max. :5.000	Max. :5.000	Max. :5.000	Max. :5.000
NA's	:78	NA's :87	NA's :96	NA's :113	NA's :117	NA's :95

```
## Creating a demonstration data set (depre_i_miss) with
  increased missings
depre_i <- (hbhc[ , varName_depre_i])
depre_i_miss <- as.data.frame(lapply(depre_i, function(cc)
  cc[ sample(c(TRUE, NA), prob = c(0.90, 0.10), size =
    length(cc), replace = TRUE) ]))
summary(depre_i_miss)
```

The HBSC data set ...

	depre1_i	depre2_i	depre3_i	depre4_i	depre5_i	depre6_i
Min.	:1.000	:1.000	:1.000	:1.000	:1.000	:1.000
1st Qu.	:1.000	:2.000	:1.000	:1.000	:1.000	:1.000
Median	:2.000	:3.000	:1.000	:2.000	:2.000	:2.000
Mean	:2.331	:2.678	:1.826	:2.204	:2.484	:2.475
3rd Qu.	:3.000	:3.000	:3.000	:3.000	:4.000	:3.000
Max.	:5.000	:5.000	:5.000	:5.000	:5.000	:5.000
NA's	:483	:482	:512	:524	:519	:503

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Listwise deletion in lavaan - ordinal treated as numeric

```

## Specifying the model-structure object
## One latent factor with six indicators
cfa.01.v.01 <- '
  depress =~ NA*depre1_i + depre2_i + depre3_i +
              depre4_i + depre5_i + depre6_i
  depress ~ 1*depress '

```

```

## Estimating cfa.01.v01
## with listwise deletion
cfa.01.v.01.fit.listwise <-
  cfa(model = cfa.01.v.01, data = depre_i_miss,
      mimic = "Mplus", estimator = "ML",
      missing = "listwise", meanstructure = TRUE)

```

```

## Requesting an estimation summary
## Please pay attention to
## The used and the total observations and
## The Minimum Function Test Statistics
summary(cfa.01.v.01.fit.listwise, fit.measures = TRUE,
        standardized = TRUE)

```

Listwise deletion in lavaan - ordinal treated as numeric ...

```

lavaan 0.6-3 ended normally after 17 iterations

  Optimization method           NLMINB
  Number of free parameters      18

5
  Number of observations         Used      Total
                                2221     4284

  Estimator                     ML
  Model Fit Test Statistic       121.494
  Degrees of freedom              9
  P-value (Chi-square)           0.000

10
Model test baseline model:

  Minimum Function Test Statistic 3165.412
  Degrees of freedom              15
  P-value                        0.000

15
User model versus baseline model:

  Comparative Fit Index (CFI)     0.964
  Tucker-Lewis Index (TLI)       0.940

20
Loglikelihood and Information Criteria:

  Loglikelihood user model (H0)    -20387.190
  Loglikelihood unrestricted model (H1) -20326.443

25
  Number of free parameters        18
  Akaike (AIC)                    40810.380
  Bayesian (BIC)                  40913.082
  Sample-size adjusted Bayesian (BIC) 40855.894

30

```

Listwise deletion in lavaan - ordinal treated as numeric ...

35 Root Mean Square Error of Approximation:

RMSEA		0.075
90 Percent Confidence Interval	0.063	0.087
P-value RMSEA <= 0.05		0.000

40 Standardized Root Mean Square Residual:

SRMR	0.028
------	-------

45 Parameter Estimates:

Information	Expected
Information saturated (h1) model	Structured
Standard Errors	Standard

50 Latent Variables:

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
depress =~						
depre1_i	0.741	0.024	31.427	0.000	0.741	0.664
depre2_i	0.741	0.025	29.855	0.000	0.741	0.637
depre3_i	0.788	0.026	30.755	0.000	0.788	0.652
depre4_i	0.884	0.028	31.526	0.000	0.884	0.666
depre5_i	0.740	0.031	24.233	0.000	0.740	0.534
depre6_i	0.748	0.029	25.560	0.000	0.748	0.559

55 Intercepts:

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
.depre1_i	2.334	0.024	98.567	0.000	2.334	2.091
.depre2_i	2.694	0.025	109.118	0.000	2.694	2.315
.depre3_i	1.806	0.026	70.473	0.000	1.806	1.495
.depre4_i	2.210	0.028	78.438	0.000	2.210	1.664
.depre5_i	2.485	0.029	84.546	0.000	2.485	1.794

Listwise deletion in lavaan - ordinal treated as numeric ...

.depre6_i	2.485	0.028	87.533	0.000	2.485	1.857
depress	0.000				0.000	0.000
Variances:						
	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
depress	1.000				1.000	1.000
.depre1_i	0.696	0.026	26.365	0.000	0.696	0.559
.depre2_i	0.805	0.029	27.320	0.000	0.805	0.595
.depre3_i	0.838	0.031	26.790	0.000	0.838	0.574
.depre4_i	0.982	0.037	26.300	0.000	0.982	0.557
.depre5_i	1.372	0.046	29.837	0.000	1.372	0.715
.depre6_i	1.231	0.042	29.344	0.000	1.231	0.687

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Pairwise deletion in lavaan - ordinal treated as numeric

```
## Estimating cfa.01.v01
## with pairwise deletion
cfa.01.v.01.fit.pairwise <-
  cfa(model = cfa.01.v.01, data = depre_i_miss,
      mimic = "Mplus", estimator = "ML",
      missing = "pairwise", meanstructure = TRUE)
```

```
## Requesting an estimation summary
## Please pay attention to
## The used and the total observations and
## The Minimum Function Test Statistics
summary(cfa.01.v.01.fit.pairwise, fit.measures = TRUE,
        standardized = TRUE)
```

Pairwise deletion in lavaan - ordinal treated as numeric ...

```

lavaan 0.6-3 ended normally after 17 iterations

  Optimization method          NLMINB
  Number of free parameters    18

  Number of observations      4220
  Number of missing patterns  57
  Total                        4284

  Estimator                    ML
  Model Fit Test Statistic     279.026
  Degrees of freedom           9
  P-value (Chi-square)         0.000

Model test baseline model:

  Minimum Function Test Statistic  5877.721
  Degrees of freedom               15
  P-value                           0.000

User model versus baseline model:

  Comparative Fit Index (CFI)      0.954
  Tucker-Lewis Index (TLI)        0.923

Loglikelihood and Information Criteria:

  Loglikelihood user model (H0)     -38741.991
  Loglikelihood unrestricted model (H1) -38602.479

  Number of free parameters         18
  Akaike (AIC)                      77519.983
  Bayesian (BIC)                     77634.240

```

Pairwise deletion in lavaan - ordinal treated as numeric ...

```

Sample-size adjusted Bayesian (BIC)          77577.043
35
Root Mean Square Error of Approximation:
RMSEA                                         0.084
90 Percent Confidence Interval              0.076 0.093
40 P-value RMSEA <= 0.05                    0.000

Standardized Root Mean Square Residual:
SRMR                                         0.031
45

Parameter Estimates:

Information                                  Expected
Information saturated (h1) model            Structured
50 Standard Errors                            Standard

Latent Variables:
      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
55 depress =~
   depre1_i    0.737   0.017  43.396   0.000   0.737   0.667
   depre2_i    0.713   0.018  39.600   0.000   0.713   0.619
   depre3_i    0.789   0.019  42.116   0.000   0.789   0.651
   depre4_i    0.875   0.020  42.710   0.000   0.875   0.659
   depre5_i    0.739   0.022  33.420   0.000   0.739   0.536
   depre6_i    0.732   0.021  34.166   0.000   0.732   0.546

Intercepts:
      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
55 .depre1_i    2.331   0.017  137.202   0.000   2.331   2.112
   .depre2_i    2.678   0.018  150.900   0.000   2.678   2.323
   .depre3_i    1.826   0.019   97.883   0.000   1.826   1.507
   .depre4_i    2.204   0.020  107.831   0.000   2.204   1.660

```


Pairwise deletion in lavaan - ordinal treated as numeric ...

.depre5_i	2.484	0.021	117.051	0.000	2.484	1.802
.depre6_i	2.475	0.021	119.974	0.000	2.475	1.847
depress	0.000				0.000	0.000

Variances:

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
depress	1.000				1.000	1.000
.depre1_i	0.676	0.019	35.881	0.000	0.676	0.555
.depre2_i	0.820	0.021	38.194	0.000	0.820	0.617
.depre3_i	0.846	0.023	36.728	0.000	0.846	0.576
.depre4_i	0.998	0.027	36.345	0.000	0.998	0.566
.depre5_i	1.355	0.033	40.941	0.000	1.355	0.713
.depre6_i	1.260	0.031	40.661	0.000	1.260	0.702

```
library(semTable)
labs <- c("depress" = "Depression")
cfa.ld.pw <-
  semTable(list("listwise" = cfa.01.v.01.fit.listwise,
    "pairwise" = cfa.01.v.01.fit.pairwise), paramSets =
    c("loadings", "intercepts", "residualvariances",
    "fits"), columns = c("estsestars"), file =
    file.path(tdir, "cfa.ld.pw"), type = "tex", longtable =
    TRUE, varLabels = labs)
```

Pairwise deletion in lavaan - ordinal treated as numeric ...

	listwise	pairwise
	Estimate(Std.Err.)	Estimate(Std.Err.)
<u>Factor Loadings</u>		
<u>Depression</u>		
depre1.i	0.74(0.02)***	0.74(0.02)***
depre2.i	0.74(0.02)***	0.71(0.02)***
depre3.i	0.79(0.03)***	0.79(0.02)***
depre4.i	0.88(0.03)***	0.87(0.02)***
depre5.i	0.74(0.03)***	0.74(0.02)***
depre6.i	0.75(0.03)***	0.73(0.02)***
<u>Intercepts</u>		
depre1.i	2.33(0.02)***	2.33(0.02)***
depre2.i	2.69(0.02)***	2.68(0.02)***
depre3.i	1.81(0.03)***	1.83(0.02)***
depre4.i	2.21(0.03)***	2.20(0.02)***
depre5.i	2.49(0.03)***	2.48(0.02)***
depre6.i	2.49(0.03)***	2.47(0.02)***
<u>Residual Variances</u>		
depre1.i	0.70(0.03)***	0.68(0.02)***
depre2.i	0.81(0.03)***	0.82(0.02)***
depre3.i	0.84(0.03)***	0.85(0.02)***

Pairwise deletion in lavaan - ordinal treated as numeric ...

depre4.i	0.98(0.04)***	1.00(0.03)***
depre5.i	1.37(0.05)***	1.35(0.03)***
depre6.i	1.23(0.04)***	1.26(0.03)***
	<u>Fit Indices</u>	
χ^2 (df)	121.49(9)***	279.03(9)***
CFI	0.96	0.95
TLI	0.94	0.92
RMSEA	0.08	0.08

⁺ Fixed parameter

* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

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FIML in lavaan - ordinal treated as numeric

```
## Estimating cfa.01.v01
## with full information maximum likelihood
cfa.01.v.01.fit.fiml <-
  cfa(model = cfa.01.v.01, data = depre_i_miss,
      mimic = "Mplus", estimator = "ML",
      missing = "ml", meanstructure = TRUE)
```

```
## Requesting an estimation summary
## Please pay attention to
## The used and the total observations and
## The Minimum Function Test Statistics
summary(cfa.01.v.01.fit.fiml, fit.measures = TRUE,
        standardized = TRUE)
```

FIML in lavaan - ordinal treated as numeric ...

```

lavaan 0.6-3 ended normally after 26 iterations

  Optimization method           NLMINB
  Number of free parameters      18

5
                                     Used      Total
  Number of observations         4220      4284
  Number of missing patterns     57

10
  Estimator                      ML
  Model Fit Test Statistic        209.383
  Degrees of freedom              9
  P-value (Chi-square)           0.000

15
Model test baseline model:

  Minimum Function Test Statistic 4905.724
  Degrees of freedom              15
  P-value                          0.000

20
User model versus baseline model:

  Comparative Fit Index (CFI)     0.959
  Tucker-Lewis Index (TLI)       0.932

25
Loglikelihood and Information Criteria:

  Loglikelihood user model (H0)    -34856.418
  Loglikelihood unrestricted model (H1) -34751.726

30
  Number of free parameters        18
  Akaike (AIC)                    69748.835
  Bayesian (BIC)                  69863.092

```

FIML in lavaan - ordinal treated as numeric ...

```

Sample-size adjusted Bayesian (BIC)           69805.895
Root Mean Square Error of Approximation:
  RMSEA                                         0.073
  90 Percent Confidence Interval             0.064  0.081
  P-value RMSEA <= 0.05                     0.000
Standardized Root Mean Square Residual:
  SRMR                                         0.030
Parameter Estimates:
  Information                                 Observed
  Observed information based on              Hessian
  Standard Errors                           Standard
Latent Variables:
  Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
depress =~
  depre1_i    0.733   0.018   40.113   0.000   0.733   0.664
  depre2_i    0.715   0.019   36.826   0.000   0.715   0.620
  depre3_i    0.791   0.020   39.307   0.000   0.791   0.653
  depre4_i    0.866   0.022   39.227   0.000   0.866   0.653
  depre5_i    0.736   0.024   30.823   0.000   0.736   0.534
  depre6_i    0.728   0.023   31.786   0.000   0.728   0.544
Intercepts:
  Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
.depre1_i    2.331   0.018   132.366  0.000   2.331   2.113
.depre2_i    2.680   0.018   145.255  0.000   2.680   2.323
.depre3_i    1.821   0.019   93.784   0.000   1.821   1.502
.depre4_i    2.208   0.021   103.874  0.000   2.208   1.666

```

FIML in lavaan - ordinal treated as numeric ...

.depre5_i	2.488	0.022	112.081	0.000	2.488	1.806
.depre6_i	2.475	0.022	114.974	0.000	2.475	1.849
depress	0.000				0.000	0.000

Variances:

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
depress	1.000				1.000	1.000
.depre1_i	0.680	0.021	32.751	0.000	0.680	0.559
.depre2_i	0.820	0.023	35.073	0.000	0.820	0.616
.depre3_i	0.843	0.025	33.508	0.000	0.843	0.574
.depre4_i	1.007	0.030	33.145	0.000	1.007	0.573
.depre5_i	1.357	0.036	37.783	0.000	1.357	0.715
.depre6_i	1.263	0.033	37.870	0.000	1.263	0.704

Comparison Table

```
library(semTable)
labs <- c("depress" = "Depression")
cfa0101tab <-
  semTable(list("listwise" = cfa.01.v.01.fit.listwise,
    "pairwise" = cfa.01.v.01.fit.pairwise, "fiml" =
    cfa.01.v.01.fit.fiml), paramSets = c("loadings",
    "intercepts", "residualvariances", "fits"), columns =
    c("estsestars"), file = file.path(tdir, "cfa0101a"),
    type = "tex", longtable = TRUE, varLabels = labs)
```

Comparison Table ...

	listwise	pairwise	fiml
	Estimate(Std.Err.)	Estimate(Std.Err.)	Estimate(Std.Err.)
<u>Factor Loadings</u>			
<u>Depression</u>			
depre1.i	0.74(0.02)***	0.74(0.02)***	0.73(0.02)***
depre2.i	0.74(0.02)***	0.71(0.02)***	0.71(0.02)***
depre3.i	0.79(0.03)***	0.79(0.02)***	0.79(0.02)***
depre4.i	0.88(0.03)***	0.87(0.02)***	0.87(0.02)***
depre5.i	0.74(0.03)***	0.74(0.02)***	0.74(0.02)***
depre6.i	0.75(0.03)***	0.73(0.02)***	0.73(0.02)***
<u>Intercepts</u>			
depre1.i	2.33(0.02)***	2.33(0.02)***	2.33(0.02)***
depre2.i	2.69(0.02)***	2.68(0.02)***	2.68(0.02)***
depre3.i	1.81(0.03)***	1.83(0.02)***	1.82(0.02)***
depre4.i	2.21(0.03)***	2.20(0.02)***	2.21(0.02)***
depre5.i	2.49(0.03)***	2.48(0.02)***	2.49(0.02)***
depre6.i	2.49(0.03)***	2.47(0.02)***	2.48(0.02)***
<u>Residual Variances</u>			
depre1.i	0.70(0.03)***	0.68(0.02)***	0.68(0.02)***
depre2.i	0.81(0.03)***	0.82(0.02)***	0.82(0.02)***
depre3.i	0.84(0.03)***	0.85(0.02)***	0.84(0.03)***

Comparison Table ...

depre4.i	0.98(0.04)***	1.00(0.03)***	1.01(0.03)***
depre5.i	1.37(0.05)***	1.35(0.03)***	1.36(0.04)***
depre6.i	1.23(0.04)***	1.26(0.03)***	1.26(0.03)***
		<u>Fit Indices</u>	
$\chi^2(df)$	121.49(9)***	279.03(9)***	209.38(9)***
CFI	0.96	0.95	0.96
TLI	0.94	0.92	0.93
RMSEA	0.08	0.08	0.07

+ Fixed parameter

* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

Outline

- 1 The Demonstration Data
- 2 Listwise Deletion in lavaan
- 3 Pairwise Deletion in lavaan
- 4 FIML in lavaan
- 5 No FIML in lavaan for Categorical Data**
- 6 FIML in Mplus
- 7 Multiple Imputation

FIML in lavaan - ordinal treated as ordinal

```

### Specifying the model-structure object
cfa.01.v.02 <- '
  depress =~ NA*depre1_i + depre2_i + depre3_i +
              depre4_i + depre5_i + depre6_i
  depress =~ 1*depress '

```

```

## Estimating cfa.01.v.02
## cfa.01.v.02.fit.fiml <-
## cfa(model = cfa.01.v.02, data = depre_i_miss,
##      mimic = "Mplus", estimator = "WLSMV",
##      missing = "ml", meanstructure = TRUE,
##      ordered = c("depre1_i", "depre2_i",
##                  "depre3_i", "depre4_i",
##                  "depre5_i", "depre6_i"))

```

```

Error in lav_options_set(opt) : lavaan ERROR:
  missing="ml" is not allowed for estimator MLM,
  MLMV, GLS, ULS, ULSM, ULSMV, DWLS, WLS, WLSM,
  WLSMV, PML

```

Pairwise deletion in lavaan - ordinal treated as ordinal

```

## Estimating cfa.01.v.02
## With pairwise deletion
cfa.01.v.02.fit.pairwise <-
  cfa(model = cfa.01.v.02, data = depre_i_miss,
      mimic = "Mplus", estimator = "WLSMV",
      missing = "pairwise", meanstructure = TRUE,
      ordered = c("depre1_i", "depre2_i",
                  "depre3_i", "depre4_i",
                  "depre5_i", "depre6_i"))
summary(cfa.01.v.02.fit.pairwise, fit.measures = TRUE,
        standardized = TRUE)

```

lavaan 0.6-3 ended normally after 11 iterations

Optimization method	NLMINB	
Number of free parameters	30	
	Used	Total
Number of observations	4220	4284
Number of missing patterns	57	
Estimator	DWLS	Robust
Model Fit Test Statistic	162.985	261.530
Degrees of freedom	9	9
P-value (Chi-square)	0.000	0.000
Scaling correction factor		0.624

Pairwise deletion in lavaan - ordinal treated as ordinal ...

```

15 Shift parameter                                0.207
   for simple second-order correction (WLSMV)

Model test baseline model:

20 Minimum Function Test Statistic             14275.959   11499.317
   Degrees of freedom                         15           15
   P-value                                     0.000        0.000

User model versus baseline model:

25 Comparative Fit Index (CFI)                 0.989        0.978
   Tucker-Lewis Index (TLI)                 0.982        0.963

   Robust Comparative Fit Index (CFI)                NA
   Robust Tucker-Lewis Index (TLI)                 NA

30 Root Mean Square Error of Approximation:

   RMSEA                                     0.064        0.082
   90 Percent Confidence Interval             0.055 0.072   0.073 0.090
   P-value RMSEA <= 0.05                     0.004        0.000

   Robust RMSEA                                NA
   90 Percent Confidence Interval             NA          NA

40 Standardized Root Mean Square Residual:

   SRMR                                     0.035        0.035

45 Weighted Root Mean Square Residual:

   WRMR                                     2.044        2.044

```

Pairwise deletion in lavaan - ordinal treated as ordinal ...

Parameter Estimates:

Information	Expected
Information saturated (h1) model	Unstructured
Standard Errors	Robust.sem

Latent Variables:

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
depress =~						
depre1_i	0.710	0.011	63.181	0.000	0.710	0.710
depre2_i	0.660	0.011	58.102	0.000	0.660	0.660
depre3_i	0.736	0.013	56.793	0.000	0.736	0.736
depre4_i	0.723	0.011	65.066	0.000	0.723	0.723
depre5_i	0.599	0.013	44.771	0.000	0.599	0.599
depre6_i	0.584	0.013	43.556	0.000	0.584	0.584

Intercepts:

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
.depre1_i	0.000				0.000	0.000
.depre2_i	0.000				0.000	0.000
.depre3_i	0.000				0.000	0.000
.depre4_i	0.000				0.000	0.000
.depre5_i	0.000				0.000	0.000
.depre6_i	0.000				0.000	0.000
depress	0.000				0.000	0.000

Thresholds:

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
depre1_i t1	-0.558	0.022	-25.930	0.000	-0.558	-0.558
depre1_i t2	0.136	0.020	6.663	0.000	0.136	0.136
depre1_i t3	1.095	0.025	43.024	0.000	1.095	1.095
depre1_i t4	1.792	0.038	47.133	0.000	1.792	1.792
depre2_i t1	-0.869	0.023	-37.153	0.000	-0.869	-0.869
depre2_i t2	-0.185	0.020	-9.043	0.000	-0.185	-0.185

Pairwise deletion in lavaan - ordinal treated as ordinal ...

depre2_i t3	0.735	0.022	32.725	0.000	0.735	0.735
depre2_i t4	1.508	0.031	47.990	0.000	1.508	1.508
depre3_i t1	0.273	0.021	13.202	0.000	0.273	0.273
depre3_i t2	0.649	0.022	29.423	0.000	0.649	0.649
depre3_i t3	1.164	0.026	44.216	0.000	1.164	1.164
depre3_i t4	1.614	0.034	47.863	0.000	1.614	1.614
depre4_i t1	-0.106	0.020	-5.184	0.000	-0.106	-0.106
depre4_i t2	0.262	0.021	12.638	0.000	0.262	0.262
depre4_i t3	0.888	0.024	37.534	0.000	0.888	0.888
depre4_i t4	1.421	0.030	47.313	0.000	1.421	1.421
depre5_i t1	-0.378	0.021	-18.025	0.000	-0.378	-0.378
depre5_i t2	0.066	0.020	3.210	0.001	0.066	0.066
depre5_i t3	0.664	0.022	29.966	0.000	0.664	0.664
depre5_i t4	1.228	0.027	45.242	0.000	1.228	1.228
depre6_i t1	-0.437	0.021	-20.691	0.000	-0.437	-0.437
depre6_i t2	0.075	0.020	3.690	0.000	0.075	0.075
depre6_i t3	0.739	0.023	32.792	0.000	0.739	0.739
depre6_i t4	1.248	0.027	45.629	0.000	1.248	1.248

Variances:

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
depress	1.000				1.000	1.000
.depre1_i	0.495				0.495	0.495
.depre2_i	0.564				0.564	0.564
.depre3_i	0.458				0.458	0.458
.depre4_i	0.478				0.478	0.478
.depre5_i	0.642				0.642	0.642
.depre6_i	0.659				0.659	0.659

Scales y*:

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
depre1_i	1.000				1.000	1.000
depre2_i	1.000				1.000	1.000
depre3_i	1.000				1.000	1.000

Pairwise deletion in lavaan - ordinal treated as ordinal ...

depre4_i	1.000	1.000	1.000
depre5_i	1.000	1.000	1.000
depre6_i	1.000	1.000	1.000

Comparison Table

```
## 20190602: bug appears in output below, don't know why!  
## 20190606: bug solved  
labs <- c("depress" = "Depression")  
cfapairwise <-  
5 semTable(list("Pairwise CFA ml" = cfa.01.v.01.fit.pairwise,  
  "Pairwise CFA wlsmv" = cfa.01.v.02.fit.pairwise),  
  paramSets = c("loadings", "thresholds", "intercepts",  
  "residualvariances", "fits"), columns = c("estsestars"),  
  file = file.path(tdir, "cfa0101b"), type = "tex",  
  longtable = TRUE, varLabels = labs)
```

Comparison Table ...

	Pairwise CFA ml Estimate(Std.Err.)	Pairwise CFA wlsmv Estimate(Std.Err.)
<u>Factor Loadings</u>		
<u>Depression</u>		
depre1.i	0.74(0.02)***	0.71(0.01)***
depre2.i	0.71(0.02)***	0.66(0.01)***
depre3.i	0.79(0.02)***	0.74(0.01)***
depre4.i	0.87(0.02)***	0.72(0.01)***
depre5.i	0.74(0.02)***	0.60(0.01)***
depre6.i	0.73(0.02)***	0.58(0.01)***
<u>Intercepts</u>		
depre1.i	2.33(0.02)***	0.00 ⁺
depre2.i	2.68(0.02)***	0.00 ⁺
depre3.i	1.83(0.02)***	0.00 ⁺
depre4.i	2.20(0.02)***	0.00 ⁺
depre5.i	2.48(0.02)***	0.00 ⁺
depre6.i	2.47(0.02)***	0.00 ⁺
<u>Residual Variances</u>		
depre1.i	0.68(0.02)***	0.50 ⁺
depre2.i	0.82(0.02)***	0.56 ⁺
depre3.i	0.85(0.02)***	0.46 ⁺

Comparison Table ...

depre4.i	1.00(0.03)***	0.48 ⁺
depre5.i	1.35(0.03)***	0.64 ⁺
depre6.i	1.26(0.03)***	0.66 ⁺
	<u>Thresholds</u>	
depre1.i(1)		-0.56(0.02)***
depre1.i(2)		0.14(0.02)***
depre1.i(3)		1.09(0.03)***
depre1.i(4)		1.79(0.04)***
depre2.i(1)		-0.87(0.02)***
depre2.i(2)		-0.18(0.02)***
depre2.i(3)		0.73(0.02)***
depre2.i(4)		1.51(0.03)***
depre3.i(1)		0.27(0.02)***
depre3.i(2)		0.65(0.02)***
depre3.i(3)		1.16(0.03)***
depre3.i(4)		1.61(0.03)***
depre4.i(1)		-0.11(0.02)***
depre4.i(2)		0.26(0.02)***
depre4.i(3)		0.89(0.02)***
depre4.i(4)		1.42(0.03)***
depre5.i(1)		-0.38(0.02)***
depre5.i(2)		0.07(0.02)**

Comparison Table ...

depre5.i(3)		0.66(0.02)***
depre5.i(4)		1.23(0.03)***
depre6.i(1)		-0.44(0.02)***
depre6.i(2)		0.08(0.02)***
depre6.i(3)		0.74(0.02)***
depre6.i(4)		1.25(0.03)***
		<u>Fit Indices</u>
χ^2 (df)	279.03(9)***	162.99
CFI	0.95	0.99
TLI	0.92	0.98
RMSEA	0.08	0.06
Scaled χ^2 (df)		261.53(9)***

⁺Fixed parameter

* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

Outline

- 1 The Demonstration Data
- 2 Listwise Deletion in lavaan
- 3 Pairwise Deletion in lavaan
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- 6 FIML in Mplus
- 7 Multiple Imputation

FIML in Mplus - ordinal treated as numeric

When the `ESTIMATOR = ML` is specified (or MLR) in Mplus for numeric data, Mplus will apply FIML if there are missing values detected

```
Mplus VERSION 8 (Linux)
MUTHEN & MUTHEN
05/30/2018  4:51 PM
```

```
INPUT INSTRUCTIONS
```

```
TITLE:
```

```
  Example 7
Health Behavior in School Children
Confirmatory Factor Analysis with Continuous Data using MLR
1-Factor CFA Model for Depression;
```

```
DATA:
```

```
FILE IS "../.../data/hbsc-subset2/hbsc-subset2.dat";
```

```
VARIABLE:
```

```
NAMES ARE
```

```
stud_id schl_id Gender Age Grade body1r body2 body3r body4 body5r
phyhlth1 phyhlth2 phyhlth3 phyhlth4 phyhlth5 phyhlth6 phyhlth7
phyhlth8  Depress1 Depress2 Depress3 Depress4 Depress5 Depress6
Bullied1 Bullied2 Bullied3 Bullied4 Bullied5 Bullied6 Bullied7
Bullied8 Bullied9 Bullier1 Bullier2 Bullier3 Bullier4 Bullier5
Bullier6 Bullier7 Bullier8 Bullier9 Alc1 Alc2 Alc3 Alc4 Alc5;
```

```
USEVARIABLES ARE
```

```
Depress1-Depress6;
```

```
MISSING ARE all(-999);
```


FIML in Mplus - ordinal treated as numeric ...

```
ANALYSIS:
  ESTIMATOR = MLR;
```

```
MODEL:
  Depress BY Depress1* Depress2-Depress6;
  Depress@1;
```

```
OUTPUT:
  TECH1;
  STDYX;
```

```
*** WARNING
Data set contains cases with missing on all variables.
These cases were not included in the analysis.
Number of cases with missing on all variables: 140
1 WARNING(S) FOUND IN THE INPUT INSTRUCTIONS
```

```
Example 7
Health Behavior in School Children
Confirmatory Factor Analysis with Continuous Data using MLR
1-Factor CFA Model for Depression;
```

SUMMARY OF ANALYSIS

Number of groups	1
Number of observations	9087
Number of dependent variables	6

FIML in Mplus - ordinal treated as numeric ...

```

Number of independent variables           0
Number of continuous latent variables     1

Observed dependent variables

  Continuous
  DEPRESS1   DEPRESS2   DEPRESS3   DEPRESS4   DEPRESS5   DEPRESS6

Continuous latent variables
  DEPRESS

Estimator                                 MLR
Information matrix                        OBSERVED
Maximum number of iterations              1000
Convergence criterion                     0.500D-04
Maximum number of steepest descent iterations 20
Maximum number of iterations for H1       2000
Convergence criterion for H1              0.100D-03

Input data file(s)
  ../../data/hbcs-subset2/hbcs-subset2.dat

Input data format  FREE

SUMMARY OF DATA

  Number of missing data patterns         31

COVARIANCE COVERAGE OF DATA

Minimum covariance coverage value 0.100

```

FIML in Mplus - ordinal treated as numeric ...

PROPORTION OF DATA PRESENT

	Covariance Coverage				
	DEPRESS1	DEPRESS2	DEPRESS3	DEPRESS4	DEPRESS5
	-----	-----	-----	-----	-----
DEPRESS1	0.998				
DEPRESS2	0.996	0.997			
DEPRESS3	0.994	0.994	0.995		
DEPRESS4	0.992	0.991	0.990	0.993	
DEPRESS5	0.990	0.990	0.989	0.987	0.992
DEPRESS6	0.994	0.994	0.993	0.991	0.990

	Covariance Coverage
	DEPRESS6

DEPRESS6	0.996

UNIVARIATE SAMPLE STATISTICS

UNIVARIATE HIGHER-ORDER MOMENT DESCRIPTIVE STATISTICS

Variable/ Sample Size	Mean/ Variance	Skewness/ Kurtosis	Minimum/ Maximum	% with Min/Max	Percentiles		
					20%/60%	40%/80%	Median
DEPRESS1	2.404	0.389	1.000	25.83%	1.000	2.000	2.000
9070.000	1.240	-0.581	5.000	4.26%	3.000	3.000	
DEPRESS2	2.768	0.071	1.000	15.68%	2.000	3.000	3.000

FIML in Mplus - ordinal treated as numeric ...

9056.000	1.246	-0.648	5.000	6.54%	3.000	4.000	
DEPRESS3	1.895	1.171	1.000	56.14%	1.000	1.000	1.000
9043.000	1.481	0.231	5.000	5.46%	2.000	3.000	
DEPRESS4	2.249	0.639	1.000	43.77%	1.000	1.000	2.000
9020.000	1.771	-0.855	5.000	7.78%	3.000	3.000	
DEPRESS5	2.516	0.389	1.000	33.00%	1.000	2.000	2.000
9011.000	1.847	-1.074	5.000	10.75%	3.000	4.000	
DEPRESS6	2.624	0.310	1.000	27.90%	1.000	2.000	3.000
9048.000	1.799	-1.054	5.000	11.93%	3.000	4.000	

THE MODEL ESTIMATION TERMINATED NORMALLY

MODEL FIT INFORMATION

Number of Free Parameters 18

Loglikelihood

H0 Value	-82045.027
H0 Scaling Correction Factor for MLR	0.9943
H1 Value	-81755.006
H1 Scaling Correction Factor for MLR	1.0943

Information Criteria

Akaike (AIC)	164126.054
Bayesian (BIC)	164254.117
Sample-Size Adjusted BIC	164196.916
(n* = (n + 2) / 24)	

FIML in Mplus - ordinal treated as numeric ...

Chi-Square Test of Model Fit

Value	448.097*
Degrees of Freedom	9
P-Value	0.0000
Scaling Correction Factor for MLR	1.2945

- * The chi-square value for MLM, MLMV, MLR, ULSMV, WLSM and WLSMV cannot be used for chi-square difference testing in the regular way. MLM, MLR and WLSM chi-square difference testing is described on the Mplus website. MLMV, WLSMV, and ULSMV difference testing is done using the DIFFTEST option.

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.073	
90 Percent C.I.	0.068	0.079
Probability RMSEA \leq .05	0.000	

CFI/TLI

CFI	0.959
TLI	0.931

Chi-Square Test of Model Fit for the Baseline Model

Value	10695.144
Degrees of Freedom	15
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value	0.029
-------	-------

FIML in Mplus - ordinal treated as numeric ...

MODEL RESULTS

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
DEPRESS BY				
DEPRESS1	0.776	0.012	65.901	0.000
DEPRESS2	0.703	0.012	59.830	0.000
DEPRESS3	0.809	0.014	58.995	0.000
DEPRESS4	0.879	0.014	64.889	0.000
DEPRESS5	0.779	0.014	53.758	0.000
DEPRESS6	0.761	0.014	53.529	0.000
Intercepts				
DEPRESS1	2.405	0.012	205.677	0.000
DEPRESS2	2.769	0.012	236.130	0.000
DEPRESS3	1.895	0.013	148.228	0.000
DEPRESS4	2.249	0.014	160.704	0.000
DEPRESS5	2.516	0.014	175.943	0.000
DEPRESS6	2.624	0.014	186.214	0.000
Variances				
DEPRESS	1.000	0.000	999.000	999.000
Residual Variances				
DEPRESS1	0.637	0.014	45.704	0.000
DEPRESS2	0.752	0.014	52.196	0.000
DEPRESS3	0.827	0.017	47.493	0.000
DEPRESS4	0.997	0.020	49.684	0.000
DEPRESS5	1.240	0.022	56.362	0.000
DEPRESS6	1.220	0.021	57.725	0.000

FIML in Mplus - ordinal treated as numeric ...

STANDARDIZED MODEL RESULTS

STDYX Standardization

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
DEPRESS BY				
DEPRESS1	0.697	0.008	83.912	0.000
DEPRESS2	0.630	0.009	71.437	0.000
DEPRESS3	0.664	0.008	78.860	0.000
DEPRESS4	0.661	0.009	77.483	0.000
DEPRESS5	0.573	0.010	58.930	0.000
DEPRESS6	0.567	0.010	58.830	0.000
Intercepts				
DEPRESS1	2.159	0.014	152.719	0.000
DEPRESS2	2.480	0.018	138.539	0.000
DEPRESS3	1.558	0.008	203.406	0.000
DEPRESS4	1.691	0.009	186.900	0.000
DEPRESS5	1.852	0.011	169.712	0.000
DEPRESS6	1.957	0.012	163.369	0.000
Variances				
DEPRESS	1.000	0.000	999.000	999.000
Residual Variances				
DEPRESS1	0.514	0.012	44.358	0.000
DEPRESS2	0.603	0.011	54.315	0.000
DEPRESS3	0.559	0.011	49.880	0.000
DEPRESS4	0.563	0.011	49.994	0.000

FIML in Mplus - ordinal treated as numeric ...

DEPRESS5	0.672	0.011	60.243	0.000
DEPRESS6	0.678	0.011	62.017	0.000

R-SQUARE

Observed Variable	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
DEPRESS1	0.486	0.012	41.956	0.000
DEPRESS2	0.397	0.011	35.718	0.000
DEPRESS3	0.441	0.011	39.430	0.000
DEPRESS4	0.437	0.011	38.741	0.000
DEPRESS5	0.328	0.011	29.465	0.000
DEPRESS6	0.322	0.011	29.415	0.000

QUALITY OF NUMERICAL RESULTS

Condition Number for the Information Matrix (ratio of smallest to largest eigenvalue)	0.906E-01
--	-----------

TECHNICAL 1 OUTPUT

PARAMETER SPECIFICATION

NU	DEPRESS1	DEPRESS2	DEPRESS3	DEPRESS4	DEPRESS5
	-----	-----	-----	-----	-----
	1	2	3	4	5

FIML in Mplus - ordinal treated as numeric ...

```

NU
  DEPRESS6
  -----
    6

```

```

LAMBDA
  DEPRESS
  -----
DEPRESS1      7
DEPRESS2      8
DEPRESS3      9
DEPRESS4     10
DEPRESS5     11
DEPRESS6     12

```

```

THETA
  DEPRESS1    DEPRESS2    DEPRESS3    DEPRESS4    DEPRESS5
  -----
DEPRESS1     13
DEPRESS2      0          14
DEPRESS3      0          0          15
DEPRESS4      0          0          0          16
DEPRESS5      0          0          0          0          17
DEPRESS6      0          0          0          0          0

```

```

THETA
  DEPRESS6
  -----
DEPRESS6     18

```

FIML in Mplus - ordinal treated as numeric ...

ALPHA
DEPRESS

0

BETA
DEPRESS

DEPRESS 0

PSI
DEPRESS

DEPRESS 0

STARTING VALUES

NU	DEPRESS1	DEPRESS2	DEPRESS3	DEPRESS4	DEPRESS5
-----	-----	-----	-----	-----	-----
2.404	2.768	1.895	2.249	2.516	

NU	DEPRESS6
-----	-----
2.624	

FIML in Mplus - ordinal treated as numeric ...

LAMBDA
DEPRESS

```

DEPRESS1  1.000
DEPRESS2  1.000
DEPRESS3  1.000
DEPRESS4  1.000
DEPRESS5  1.000
DEPRESS6  1.000

```

THETA

	DEPRESS1	DEPRESS2	DEPRESS3	DEPRESS4	DEPRESS5
DEPRESS1	0.620				
DEPRESS2	0.000	0.623			
DEPRESS3	0.000	0.000	0.740		
DEPRESS4	0.000	0.000	0.000	0.886	
DEPRESS5	0.000	0.000	0.000	0.000	0.923
DEPRESS6	0.000	0.000	0.000	0.000	0.000

THETA

```

DEPRESS6
DEPRESS6  0.899

```

ALPHA

```

DEPRESS
0.000

```

FIML in Mplus - ordinal treated as numeric ...

```

      BETA
      DEPRESS
      -----
DEPRESS      0.000
  
```

```

      PSI
      DEPRESS
      -----
DEPRESS      1.000
  
```

```

Beginning Time: 16:51:22
Ending Time: 16:51:22
Elapsed Time: 00:00:00
  
```

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If users want to change the default FIML behavior in Mplus,
LISTWISE = ON in the **DATA COMMAND** can be specified

FIML in Mplus - ordinal treated as ordinal

To our knowledge, Mplus is currently the only SEM software that allows FIML estimation to be used for ordinal data with missing values

- Parameters of categorical variables are estimated (e.g., thresholds)
- Individual-level information are used, instead of covariance information, for model estimation
- Time consuming (a three-factor model can take about 20mins to converge)
- Some times large models do not converge at all

```
Mplus VERSION 8 (Linux)
MUTHEN & MUTHEN
05/30/2018  4:50 PM
```

```
INPUT INSTRUCTIONS
```

```
TITLE:
  Example 7
Health Behavior in School Children
Confirmatory Factor Analysis with Ordinal-Categorical Data using MLR
1-Factor Graded Response IRT Model for Depression;
```

```
DATA:
  FILE IS "../.../data/hbsc-subset2/hbsc-subset2.dat";
```

FIML in Mplus - ordinal treated as ordinal ...

```

VARIABLE:
  NAMES ARE
    stud_id schl_id Gender Age Grade body1r body2 body3r body4 body5r
    phyhlth1 phyhlth2 phyhlth3 phyhlth4 phyhlth5 phyhlth6 phyhlth7
    phyhlth8 Depress1 Depress2 Depress3 Depress4 Depress5 Depress6
    Bullied1 Bullied2 Bullied3 Bullied4 Bullied5 Bullied6 Bullied7
    Bullied8 Bullied9 Bullier1 Bullier2 Bullier3 Bullier4 Bullier5
    Bullier6 Bullier7 Bullier8 Bullier9 Alc1 Alc2 Alc3 Alc4 Alc5;

USEVARIABLES ARE
  Depress1-Depress6;

CATEGORICAL ARE
  Depress1-Depress6;

MISSING ARE all(-999);

ANALYSIS:
  ESTIMATOR = MLR;

MODEL:
  Depress BY Depress1* Depress2-Depress6;
  Depress@1;

OUTPUT:
  TECH1;
  STDYX;

*** WARNING
Data set contains cases with missing on all variables.
These cases were not included in the analysis.

```

FIML in Mplus - ordinal treated as ordinal ...

```
Number of cases with missing on all variables: 140
1 WARNING(S) FOUND IN THE INPUT INSTRUCTIONS
```

Example 7

```
Health Behavior in School Children
Confirmatory Factor Analysis with Ordinal-Categorical Data using MLR
1-Factor Graded Response IRT Model for Depression;
```

SUMMARY OF ANALYSIS

Number of groups	1
Number of observations	9087
Number of dependent variables	6
Number of independent variables	0
Number of continuous latent variables	1

Observed dependent variables

Binary and ordered categorical (ordinal)

DEPRESS1	DEPRESS2	DEPRESS3	DEPRESS4	DEPRESS5	DEPRESS6
----------	----------	----------	----------	----------	----------

Continuous latent variables

DEPRESS

Estimator	MLR
Information matrix	OBSERVED
Optimization Specifications for the Quasi-Newton Algorithm for Continuous Outcomes	
Maximum number of iterations	100

FIML in Mplus - ordinal treated as ordinal ...

```

Convergence criterion                0.100D-05
Optimization Specifications for the EM Algorithm
  Maximum number of iterations        500
  Convergence criteria
    Loglikelihood change              0.100D-02
    Relative loglikelihood change     0.100D-05
    Derivative                        0.100D-02
Optimization Specifications for the M step of the EM Algorithm for
Categorical Latent variables
  Number of M step iterations         1
  M step convergence criterion        0.100D-02
  Basis for M step termination        ITERATION
Optimization Specifications for the M step of the EM Algorithm for
Censored, Binary or Ordered Categorical (Ordinal), Unordered
Categorical (Nominal) and Count Outcomes
  Number of M step iterations         1
  M step convergence criterion        0.100D-02
  Basis for M step termination        ITERATION
  Maximum value for logit thresholds  15
  Minimum value for logit thresholds -15
  Minimum expected cell size for chi-square 0.100D-01
  Maximum number of iterations for H1 2000
  Convergence criterion for H1        0.100D-03
  Optimization algorithm              EMA
Integration Specifications
  Type                                STANDARD
  Number of integration points         15
  Dimensions of numerical integration  1
  Adaptive quadrature                 ON
Link                                  LOGIT
Cholesky                              ON

Input data file(s)
  ../../data/hbsc-subset2/hbsc-subset2.dat

```


FIML in Mplus - ordinal treated as ordinal ...

Input data format FREE

SUMMARY OF DATA

Number of missing data patterns	31
Number of y missing data patterns	0
Number of u missing data patterns	31

COVARIANCE COVERAGE OF DATA

Minimum covariance coverage value 0.100

PROPORTION OF DATA PRESENT FOR U

	Covariance Coverage				
	DEPRESS1	DEPRESS2	DEPRESS3	DEPRESS4	DEPRESS5
	-----	-----	-----	-----	-----
DEPRESS1	0.998				
DEPRESS2	0.996	0.997			
DEPRESS3	0.994	0.994	0.995		
DEPRESS4	0.992	0.991	0.990	0.993	
DEPRESS5	0.990	0.990	0.989	0.987	0.992
DEPRESS6	0.994	0.994	0.993	0.991	0.990

	Covariance Coverage	
	DEPRESS6	
	-----	-----
DEPRESS6	0.996	

FIML in Mplus - ordinal treated as ordinal ...

UNIVARIATE PROPORTIONS AND COUNTS FOR CATEGORICAL VARIABLES

DEPRESS1

Category 1	0.258	2343.000
Category 2	0.278	2519.000
Category 3	0.308	2792.000
Category 4	0.114	1030.000
Category 5	0.043	386.000

DEPRESS2

Category 1	0.157	1420.000
Category 2	0.230	2083.000
Category 3	0.367	3321.000
Category 4	0.181	1640.000
Category 5	0.065	592.000

DEPRESS3

Category 1	0.561	5077.000
Category 2	0.164	1482.000
Category 3	0.147	1331.000
Category 4	0.073	659.000
Category 5	0.055	494.000

DEPRESS4

Category 1	0.438	3948.000
Category 2	0.152	1375.000
Category 3	0.211	1904.000
Category 4	0.121	1091.000
Category 5	0.078	702.000

DEPRESS5

Category 1	0.330	2974.000
Category 2	0.185	1667.000
Category 3	0.231	2086.000
Category 4	0.146	1315.000
Category 5	0.108	969.000

DEPRESS6

FIML in Mplus - ordinal treated as ordinal ...

Category 1	0.279	2524.000
Category 2	0.202	1828.000
Category 3	0.254	2301.000
Category 4	0.145	1316.000
Category 5	0.119	1079.000

THE MODEL ESTIMATION TERMINATED NORMALLY

MODEL FIT INFORMATION

Number of Free Parameters 30

Loglikelihood

H0 Value	-71988.417
H0 Scaling Correction Factor for MLR	1.0190

Information Criteria

Akaike (AIC)	144036.834
Bayesian (BIC)	144250.272
Sample-Size Adjusted BIC	144154.937
(n* = (n + 2) / 24)	

Chi-Square Test of Model Fit for the Binary and Ordered Categorical
(Ordinal) Outcomes**

Pearson Chi-Square

FIML in Mplus - ordinal treated as ordinal ...

Value	21102.439
Degrees of Freedom	15496
P-Value	0.0000

Likelihood Ratio Chi-Square

Value	11118.302
Degrees of Freedom	15496
P-Value	1.0000

** Of the 40775 cells in the frequency table, 98
were deleted in the calculation of chi-square due to extreme values.

Chi-Square Test for MCAR under the Unrestricted Latent Class Indicator Model

Pearson Chi-Square

Value	2233.670
Degrees of Freedom	25120
P-Value	1.0000

Likelihood Ratio Chi-Square

Value	572.725
Degrees of Freedom	25120
P-Value	1.0000

MODEL RESULTS

Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
----------	------	-----------	-----------------------

FIML in Mplus - ordinal treated as ordinal ...

DEPRESS BY				
DEPRESS1	1.948	0.048	40.580	0.000
DEPRESS2	1.636	0.039	41.733	0.000
DEPRESS3	1.983	0.052	38.158	0.000
DEPRESS4	1.812	0.046	39.286	0.000
DEPRESS5	1.394	0.037	37.804	0.000
DEPRESS6	1.370	0.035	38.740	0.000

Thresholds

DEPRESS1\$1	-1.676	0.042	-40.228	0.000
DEPRESS1\$2	0.176	0.033	5.299	0.000
DEPRESS1\$3	2.571	0.049	52.009	0.000
DEPRESS1\$4	4.633	0.081	56.991	0.000
DEPRESS2\$1	-2.375	0.044	-53.831	0.000
DEPRESS2\$2	-0.692	0.032	-21.780	0.000
DEPRESS2\$3	1.568	0.036	43.324	0.000
DEPRESS2\$4	3.665	0.061	60.179	0.000
DEPRESS3\$1	0.368	0.034	10.665	0.000
DEPRESS3\$2	1.499	0.040	37.070	0.000
DEPRESS3\$3	2.956	0.056	52.717	0.000
DEPRESS3\$4	4.310	0.079	54.882	0.000
DEPRESS4\$1	-0.412	0.033	-12.632	0.000
DEPRESS4\$2	0.510	0.033	15.636	0.000
DEPRESS4\$3	2.048	0.043	48.020	0.000
DEPRESS4\$4	3.595	0.063	56.855	0.000
DEPRESS5\$1	-0.954	0.030	-31.303	0.000
DEPRESS5\$2	0.072	0.028	2.572	0.010
DEPRESS5\$3	1.426	0.033	42.588	0.000
DEPRESS5\$4	2.753	0.047	58.306	0.000
DEPRESS6\$1	-1.268	0.032	-39.566	0.000
DEPRESS6\$2	-0.118	0.028	-4.228	0.000
DEPRESS6\$3	1.333	0.032	41.223	0.000
DEPRESS6\$4	2.583	0.044	58.699	0.000

FIML in Mplus - ordinal treated as ordinal ...

Variances				
DEPRESS	1.000	0.000	999.000	999.000

STANDARDIZED MODEL RESULTS

STDYX Standardization

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
DEPRESS BY				
DEPRESS1	0.732	0.008	87.367	0.000
DEPRESS2	0.670	0.009	75.704	0.000
DEPRESS3	0.738	0.009	83.772	0.000
DEPRESS4	0.707	0.009	78.492	0.000
DEPRESS5	0.609	0.010	60.145	0.000
DEPRESS6	0.603	0.010	60.835	0.000
Thresholds				
DEPRESS1\$1	-0.630	0.014	-46.162	0.000
DEPRESS1\$2	0.066	0.012	5.284	0.000
DEPRESS1\$3	0.966	0.016	60.536	0.000
DEPRESS1\$4	1.741	0.025	69.189	0.000
DEPRESS2\$1	-0.972	0.016	-62.641	0.000
DEPRESS2\$2	-0.283	0.013	-22.512	0.000
DEPRESS2\$3	0.642	0.014	46.135	0.000
DEPRESS2\$4	1.500	0.022	69.312	0.000
DEPRESS3\$1	0.137	0.013	10.792	0.000
DEPRESS3\$2	0.558	0.014	40.963	0.000
DEPRESS3\$3	1.100	0.017	64.800	0.000
DEPRESS3\$4	1.604	0.023	70.095	0.000
DEPRESS4\$1	-0.161	0.013	-12.755	0.000

FIML in Mplus - ordinal treated as ordinal ...

DEPRESS4\$2	0.199	0.013	15.750	0.000
DEPRESS4\$3	0.799	0.015	54.133	0.000
DEPRESS4\$4	1.402	0.020	69.217	0.000
DEPRESS5\$1	-0.417	0.013	-32.408	0.000
DEPRESS5\$2	0.032	0.012	2.572	0.010
DEPRESS5\$3	0.623	0.014	45.602	0.000
DEPRESS5\$4	1.203	0.018	66.802	0.000
DEPRESS6\$1	-0.558	0.013	-42.056	0.000
DEPRESS6\$2	-0.052	0.012	-4.233	0.000
DEPRESS6\$3	0.586	0.014	43.417	0.000
DEPRESS6\$4	1.136	0.017	65.553	0.000

Variances

DEPRESS	1.000	0.000	999.000	999.000
---------	-------	-------	---------	---------

R-SQUARE

Observed Variable	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
DEPRESS1	0.536	0.012	43.683	0.000
DEPRESS2	0.449	0.012	37.852	0.000
DEPRESS3	0.545	0.013	41.886	0.000
DEPRESS4	0.499	0.013	39.246	0.000
DEPRESS5	0.371	0.012	30.073	0.000
DEPRESS6	0.363	0.012	30.418	0.000

QUALITY OF NUMERICAL RESULTS

Condition Number for the Information Matrix 0.162E-01
 (ratio of smallest to largest eigenvalue)

FIML in Mplus - ordinal treated as ordinal ...

TECHNICAL 1 OUTPUT

PARAMETER SPECIFICATION

```

TAU
DEPRESS1    DEPRESS1    DEPRESS1    DEPRESS1    DEPRESS2
-----
          7          8          9         10         11

TAU
DEPRESS2    DEPRESS2    DEPRESS2    DEPRESS3    DEPRESS3
-----
          12         13         14         15         16

TAU
DEPRESS3    DEPRESS3    DEPRESS4    DEPRESS4    DEPRESS4
-----
          17         18         19         20         21

TAU
DEPRESS4    DEPRESS5    DEPRESS5    DEPRESS5    DEPRESS5
-----
          22         23         24         25         26

TAU
DEPRESS6    DEPRESS6    DEPRESS6    DEPRESS6
-----

```


FIML in Mplus - ordinal treated as ordinal ...

27 28 29 30

```

NU
DEPRESS1            DEPRESS2            DEPRESS3            DEPRESS4            DEPRESS5
-----
0                    0                    0                    0                    0

```

```

NU
DEPRESS6
-----
0

```

```

LAMBDA
DEPRESS
-----

```

```

DEPRESS1            1
DEPRESS2            2
DEPRESS3            3
DEPRESS4            4
DEPRESS5            5
DEPRESS6            6

```

```

THETA
DEPRESS1            DEPRESS2            DEPRESS3            DEPRESS4            DEPRESS5
-----
DEPRESS1            0
DEPRESS2            0                    0
DEPRESS3            0                    0                    0
DEPRESS4            0                    0                    0                    0
DEPRESS5            0                    0                    0                    0                    0

```

FIML in Mplus - ordinal treated as ordinal ...

```
DEPRESS6      0      0      0      0      0
```

```
THETA
```

```
  DEPRESS6
```

```
DEPRESS6  -----
           0
```

```
ALPHA
```

```
  DEPRESS
```

```
  -----
           0
```

```
BETA
```

```
  DEPRESS
```

```
DEPRESS  -----
           0
```

```
PSI
```

```
  DEPRESS
```

```
DEPRESS  -----
           0
```

```
STARTING VALUES
```

```
TAU
```

```
  DEPRESS1
```

```
  DEPRESS1
```

```
  DEPRESS1
```

```
  DEPRESS1
```

```
  DEPRESS2
```

```
  -----
 -1.055
```

```
  -----
  0.144
```

```
  -----
  1.687
```

```
  -----
  3.113
```

```
  -----
 -1.682
```

FIML in Mplus - ordinal treated as ordinal ...

TAU					
DEPRESS2	DEPRESS2	DEPRESS2	DEPRESS3	DEPRESS3	
-----	-----	-----	-----	-----	
-0.461	1.118	2.660	0.247	0.971	

TAU					
DEPRESS3	DEPRESS3	DEPRESS4	DEPRESS4	DEPRESS4	
-----	-----	-----	-----	-----	
1.923	2.851	-0.251	0.365	1.394	

TAU					
DEPRESS4	DEPRESS5	DEPRESS5	DEPRESS5	DEPRESS5	
-----	-----	-----	-----	-----	
2.472	-0.708	0.060	1.080	2.116	

TAU					
DEPRESS6	DEPRESS6	DEPRESS6	DEPRESS6		
-----	-----	-----	-----		
-0.950	-0.076	1.022	2.000		

NU					
DEPRESS1	DEPRESS2	DEPRESS3	DEPRESS4	DEPRESS5	
-----	-----	-----	-----	-----	
0.000	0.000	0.000	0.000	0.000	

NU	
DEPRESS6	

FIML in Mplus - ordinal treated as ordinal ...

```
-----
0.000
```

```
LAMBDA
DEPRESS
```

```
-----
DEPRESS1    1.000
DEPRESS2    1.000
DEPRESS3    1.000
DEPRESS4    1.000
DEPRESS5    1.000
DEPRESS6    1.000
```

```
THETA
```

	DEPRESS1	DEPRESS2	DEPRESS3	DEPRESS4	DEPRESS5
DEPRESS1	1.000				
DEPRESS2	0.000	1.000			
DEPRESS3	0.000	0.000	1.000		
DEPRESS4	0.000	0.000	0.000	1.000	
DEPRESS5	0.000	0.000	0.000	0.000	1.000
DEPRESS6	0.000	0.000	0.000	0.000	0.000

```
THETA
```

```
DEPRESS6
-----
DEPRESS6    1.000
```

```
ALPHA
```

```
DEPRESS
```

FIML in Mplus - ordinal treated as ordinal ...

```
-----
0.000
```

```
BETA
DEPRESS
```

```
-----
DEPRESS 0.000
```

```
PSI
DEPRESS
```

```
-----
DEPRESS 1.000
```

```
Beginning Time: 16:50:29
Ending Time: 16:50:30
Elapsed Time: 00:00:01
```

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FIML in Mplus - ordinal treated as ordinal ...

If users want to change the default FIML behavior in Mplus, `LISTWISE = ON` in the `DATA COMMAND` can be specified

Outline

- 1 The Demonstration Data
- 2 Listwise Deletion in lavaan
- 3 Pairwise Deletion in lavaan
- 4 FIML in lavaan
- 5 No FIML in lavaan for Categorical Data
- 6 FIML in Mplus
- 7 Multiple Imputation**

Two Types of Imputation

- mice: “fcs” fully conditional specification
 - very slow with lots of ordinal variables
- Amelia: multivariate normal approximation
 - faster!

Packages you'll need

I installed “mice”, “amelia”, “mitools” to get this going. Probably several others will come along with those if you install them.

Mice ran slowly in this case

- The imputation with mice is launched as follows. This takes several hours, so don't launch it unless you really mean it!
- I only asked for default number of completed sets, $m=5$, because I hoped it would be fast. Usually I'd want more. But even this was slow.

```
library(mice)
## I was careless, did not look at hbsc carefully first
set5 <- mice(hbsc, m = 5)
## To see what you get, do this
par(ask = TRUE)
plot(set5)
```

After inspecting the initial set with 5 iterations, I concluded the results were inadequate, so I ran several additional commands to add more samples. To add more iterations, use “mice.mids”.

Mice ran slowly in this case ...

```
set15 <- mice.mids(set5, maxit = 10)
## That was not adequate keep going
set25 <- mice.mids(set15, maxit = 10)
## Inadequate, keep going
set30 <- mice.mids(set25, maxit = 5)
## Expectations crashed that was not good
set45 <- mice.mids(set45, maxit = 25)
## Stop there, I had enough
saveRDS(set45, file.path(wdir, "mice-45.rds"))
```

Load the saved mice file from workingdata

```
wdir <- "workingdata"  
list.files(wdir)
```

```
[1] "amelia-imps.rds" "mice-05.rds"      "mice-15.rds"      "mice-25.rds"      "mice-35.rds"  
[6] "mice-45.rds"
```

```
library(mice)  
miceimputed <- readRDS(file.path(wdir, "mice-45.rds"))  
if(interactive()){  
  par(ask=TRUE)  
  plot(miceimputed)  
}
```

Load the saved mice file from workingdata ...

- I want to fit this structural equation model

```
sem.01.int <- '  
## the measurement model  
gotBully =~ NA*gotBu1_i + gotBu2_i + gotBu3_i  
           + gotBu4_i + gotBu5_i + gotBu6_i  
           + gotBu7_i + gotBu8_i + gotBu9_i  
gotBully ~ 1*gotBully  
  
alcohol =~ NA*alc1_i + alc2_i + alc3_i  
          + alc4_i + alc5_i  
alcohol ~ 1*alcohol  
  
# regress  
alcohol ~ gotBully  
,
```

- mice offers a way to run a set of models, one for each imputed data set. However in this case it does not work.

Load the saved mice file from workingdata ...

```
fits.mice <- with(data = miceimputed, exp =  
  sem(model = sem.01.int, mimic = "Mplus",  
    estimator = "ML"))
```

- That returns without error, but further inspection indicates the model was not estimated and that the usual function “summary(fits.mice)” does not understand lavaan objects.
- So we fall back to manual method.

Step 1: pull a completed data set for each imputation

Create a new list called `imps` (“imputations”) and use the `complete` function to get the data frames

```
imps <- list()
for(i in 1:5){
  imps[[i]] <- complete(miceimputed, action = i)
}
```

- When the imputation ran, it only filled in values for the ordinal “_o” variables, it left NAs on the integer ones.

```
summary(imps[[1]])
```

Step 1: pull a completed data set for each imputation ...

stud_id	schl_id		Gender	Age	Grade	body1_o	body2_o		
Min. : 1	Min. : 2.0	Min. : 0.000	Min. :11.0	Min. :6.000	SA: 807	SA: 847			
1st Qu.:2278	1st Qu.: 59.0	1st Qu.:0.000	1st Qu.:12.0	1st Qu.:6.000	A : 804	A : 696			
Median :4588	Median :122.0	Median :1.000	Median :12.0	Median :6.000	U : 755	U : 725			
Mean :4606	Mean :117.2	Mean :0.507	Mean :12.1	Mean :6.439	D : 751	D :1135			
3rd Qu.:6925	3rd Qu.:172.0	3rd Qu.:1.000	3rd Qu.:13.0	3rd Qu.:7.000	SD:1167	SD: 881			
Max. :9224	Max. :227.0	Max. :1.000	Max. :17.0	Max. :7.000					
body3_o	body4_o	body5_o	phys1_o	phys2_o	phys3_o				
SA: 763	SA: 857	SA: 728	EveryDay : 328	EveryDay : 214	EveryDay : 310				
A : 449	A : 596	A : 354	EveryWeek : 443	EveryWeek : 384	EveryWeek : 315				
U : 620	U : 736	U : 723	MoreOnceWeek: 425	MoreOnceWeek: 423	MoreOnceWeek: 342				
D : 592	D : 983	D : 491	EveryMonth :1117	EveryMonth :1263	EveryMonth : 685				
SD:1860	SD:1112	SD:1988	Never :1971	Never :2000	Never :2632				
phys4_o		phys5_o		phys6_o		phys7_o			
EveryDay : 341		EveryDay : 558		EveryDay : 488		EveryDay : 731			
EveryWeek : 323		EveryWeek : 541		EveryWeek : 505		EveryWeek : 422			
MoreOnceWeek: 351		MoreOnceWeek: 541		MoreOnceWeek: 654		MoreOnceWeek: 416			
EveryMonth : 742		EveryMonth : 981		EveryMonth :1000		EveryMonth : 612			
Never :2527		Never :1663		Never :1637		Never :2103			
phys8_o		depre1_o		depre2_o		depre3_o		depre4_o	
EveryDay : 256		Never :1229		Never : 816		Never :2600		Never :1939	
EveryWeek : 278		Seldom :1132		Seldom : 987		Seldom : 584		Seldom : 611	
MoreOnceWeek: 263		Sometimes:1323		Sometimes:1470		Sometimes: 569		Sometimes: 911	
EveryMonth : 577		Often : 428		Often : 711		Often : 294		Often : 489	
Never :2910		Always : 172		Always : 300		Always : 237		Always : 334	
depre5_o		depre6_o		gotBu1_o		gotBu2_o			

Step 1: pull a completed data set for each imputation ...

```

Never      :1505  Never      :1416  never      :1831  never      :2049
Seldom     : 738  Seldom     : 843  1 or 2     : 668  1 or 2     : 539
Sometimes  : 934  Sometimes  :1019  2-3 p/month : 402  2-3 p/month : 603
Often      : 628  Often      : 540  1 p/week   : 625  1 p/week   : 413
Always     : 479  Always     : 466  Several p/week: 758  Several p/week: 680

```

```

      gotBu3_o          gotBu4_o          gotBu5_o          gotBu6_o
never      :2452  never      :1864  never      :2519  never      :2669
1 or 2     : 305  1 or 2     : 629  1 or 2     : 262  1 or 2     : 173
2-3 p/month : 475  2-3 p/month : 469  2-3 p/month : 429  2-3 p/month : 150
1 p/week   : 520  1 p/week   : 574  1 p/week   : 484  1 p/week   : 995
Several p/week: 532  Several p/week: 748  Several p/week: 590  Several p/week: 297

```

```

      gotBu7_o          gotBu8_o          gotBu9_o          bu0th1_o
never      :2217  never      :2723  never      :2785  never      :1966
1 or 2     : 388  1 or 2     : 146  1 or 2     : 97  1 or 2     : 705
2-3 p/month : 409  2-3 p/month : 460  2-3 p/month : 507  2-3 p/month : 387
1 p/week   : 660  1 p/week   : 484  1 p/week   : 581  1 p/week   : 646
Several p/week: 610  Several p/week: 471  Several p/week: 314  Several p/week: 580

```

```

      bu0th2_o          bu0th3_o          bu0th4_o          bu0th5_o
never      :2273  never      :2525  never      :2555  never      :2704
1 or 2     : 477  1 or 2     : 262  1 or 2     : 297  1 or 2     : 159
2-3 p/month : 449  2-3 p/month : 189  2-3 p/month : 405  2-3 p/month : 386
1 p/week   : 509  1 p/week   : 536  1 p/week   : 511  1 p/week   : 495
Several p/week: 576  Several p/week: 772  Several p/week: 516  Several p/week: 540

```

```

      bu0th6_o          bu0th7_o          bu0th8_o          bu0th9_o
never      :2767  never      :2618  never      :2758  never      :2819
1 or 2     : 190  1 or 2     : 216  1 or 2     : 144  1 or 2     : 94

```

Step 1: pull a completed data set for each imputation ...

```

2-3 p/month : 613  2-3 p/month : 204  2-3 p/month : 345  2-3 p/month : 579
1 p/week    : 409  1 p/week    : 678  1 p/week    : 633  1 p/week    : 217
Several p/week: 305  Several p/week: 568  Several p/week: 404  Several p/week: 575

```

```

      alc1_o      alc2_o      alc3_o      alc4_o      alc5_o      body1_i
Never :3597  Never :3455  Never :3766  Never :3469  Never :3589  Min. :1.000
Rarely : 448  Rarely : 578  Rarely : 254  Rarely : 463  Rarely : 408  1st Qu.:3.000
Monthly: 85  Monthly: 87  Monthly: 83  Monthly: 146  Monthly: 88  Median :4.000
Weekly : 37  Weekly : 63  Weekly : 73  Weekly : 78  Weekly : 76  Mean  :3.623
Daily  :117  Daily  :101  Daily  :108  Daily  :128  Daily  :123  3rd Qu.:5.000
                                           Max.  :5.000
                                           NA's  :1300

      body2_i      body3_i      body4_i      body5_i      phys1_i      phys2_i
Min. :1.000  Min. :1.000  Min. :1.00  Min. :1.000  Min. :1.000  Min. :1.000
1st Qu.:3.000  1st Qu.:4.000  1st Qu.:3.00  1st Qu.:4.000  1st Qu.:3.000  1st Qu.:4.000
Median :4.000  Median :5.000  Median :4.00  Median :5.000  Median :4.000  Median :4.000
Mean  :3.558  Mean  :4.167  Mean  :3.68  Mean  :4.256  Mean  :3.938  Mean  :4.053
3rd Qu.:5.000  3rd Qu.:5.000  3rd Qu.:5.00  3rd Qu.:5.000  3rd Qu.:5.000  3rd Qu.:5.000
Max.  :5.000  Max.  :5.000  Max.  :5.00  Max.  :5.000  Max.  :5.000  Max.  :5.000
NA's  :1315  NA's  :1308  NA's  :1311  NA's  :1307  NA's  :78  NA's  :90

      phys3_i      phys4_i      phys5_i      phys6_i      phys7_i      phys8_i
Min. :1.000  Min. :1.000  Min. :1.000  Min. :1.000  Min. :1.000  Min. :1.000
1st Qu.:4.000  1st Qu.:4.000  1st Qu.:2.000  1st Qu.:3.000  1st Qu.:2.000  1st Qu.:4.000
Median :5.000  Median :5.000  Median :4.000  Median :4.000  Median :4.000  Median :5.000
Mean  :4.186  Mean  :4.135  Mean  :3.638  Mean  :3.666  Mean  :3.693  Mean  :4.325
3rd Qu.:5.000  3rd Qu.:5.000  3rd Qu.:5.000  3rd Qu.:5.000  3rd Qu.:5.000  3rd Qu.:5.000
Max.  :5.000  Max.  :5.000  Max.  :5.000  Max.  :5.000  Max.  :5.000  Max.  :5.000
NA's  :107  NA's  :140  NA's  :109  NA's  :121  NA's  :100  NA's  :111

      depre1_i      depre2_i      depre3_i      depre4_i      depre5_i      depre6_i
Min. :1.000  Min. :1.000  Min. :1.000  Min. :1.000  Min. :1.000  Min. :1.000
1st Qu.:1.000  1st Qu.:2.000  1st Qu.:1.000  1st Qu.:1.000  1st Qu.:1.000  1st Qu.:1.000
Median :2.000  Median :3.000  Median :1.000  Median :2.000  Median :2.000  Median :2.000
Mean  :2.332  Mean  :2.684  Mean  :1.817  Mean  :2.209  Mean  :2.485  Mean  :2.473

```

Step 1: pull a completed data set for each imputation ...

```

3rd Qu.:3.000 3rd Qu.:3.000 3rd Qu.:3.000 3rd Qu.:3.000 3rd Qu.:4.000 3rd Qu.:3.000
Max. :5.000 Max. :5.000 Max. :5.000 Max. :5.000 Max. :5.000 Max. :5.000
NA's :78 NA's :87 NA's :96 NA's :113 NA's :117 NA's :95
gotBu1_i gotBu2_i gotBu3_i gotBu4_i gotBu5_i gotBu6_i
Min. :1.000 Min. :1.000 Min. :1.000 Min. :1.000 Min. :1.000 Min. :1.000
1st Qu.:1.000 1st Qu.:1.000 1st Qu.:1.000 1st Qu.:1.000 1st Qu.:1.000 1st Qu.:1.000
Median :1.000 Median :1.000 Median :1.000 Median :1.000 Median :1.000 Median :1.000
Mean :1.848 Mean :1.652 Mean :1.398 Mean :1.759 Mean :1.353 Mean :1.235
3rd Qu.:2.000 3rd Qu.:2.000 3rd Qu.:1.000 3rd Qu.:2.000 3rd Qu.:1.000 3rd Qu.:1.000
Max. :5.000 Max. :5.000 Max. :5.000 Max. :5.000 Max. :5.000 Max. :5.000
NA's :1322 NA's :1329 NA's :1340 NA's :1331 NA's :1351 NA's :1359
gotBu7_i gotBu8_i gotBu9_i bu0th1_i bu0th2_i bu0th3_i
Min. :1.000 Min. :1.000 Min. :1.000 Min. :1.000 Min. :1.00 Min. :1.000
1st Qu.:1.000 1st Qu.:1.000 1st Qu.:1.000 1st Qu.:1.000 1st Qu.:1.00 1st Qu.:1.000
Median :1.000 Median :1.000 Median :1.000 Median :1.000 Median :1.00 Median :1.000
Mean :1.568 Mean :1.195 Mean :1.166 Mean :1.579 Mean :1.42 Mean :1.292
3rd Qu.:2.000 3rd Qu.:1.000 3rd Qu.:1.000 3rd Qu.:2.000 3rd Qu.:1.00 3rd Qu.:1.000
Max. :5.000 Max. :5.000 Max. :5.000 Max. :5.000 Max. :5.00 Max. :5.000
NA's :1347 NA's :1357 NA's :1355 NA's :1339 NA's :1349 NA's :1364
bu0th4_i bu0th5_i bu0th6_i bu0th7_i bu0th8_i bu0th9_i
Min. :1.000 Min. :1.0 Min. :1.000 Min. :1.00 Min. :1.000 Min. :1.000
1st Qu.:1.000 1st Qu.:1.0 1st Qu.:1.000 1st Qu.:1.00 1st Qu.:1.000 1st Qu.:1.000
Median :1.000 Median :1.0 Median :1.000 Median :1.00 Median :1.000 Median :1.000
Mean :1.264 Mean :1.2 Mean :1.158 Mean :1.25 Mean :1.155 Mean :1.147
3rd Qu.:1.000 3rd Qu.:1.0 3rd Qu.:1.000 3rd Qu.:1.00 3rd Qu.:1.000 3rd Qu.:1.000
Max. :5.000 Max. :5.0 Max. :5.000 Max. :5.00 Max. :5.000 Max. :5.000
NA's :1358 NA's :1355 NA's :1356 NA's :1359 NA's :1365 NA's :1357
alc1_i alc2_i alc3_i alc4_i alc5_i
Min. :1.000 Min. :1.000 Min. :1.000 Min. :1.000 Min. :1.000
1st Qu.:1.000 1st Qu.:1.000 1st Qu.:1.000 1st Qu.:1.000 1st Qu.:1.000
Median :1.000 Median :1.000 Median :1.000 Median :1.000 Median :1.000
Mean :1.191 Mean :1.224 Mean :1.152 Mean :1.251 Mean :1.206
3rd Qu.:1.000 3rd Qu.:1.000 3rd Qu.:1.000 3rd Qu.:1.000 3rd Qu.:1.000
Max. :5.000 Max. :5.000 Max. :5.000 Max. :5.000 Max. :5.000

```

Step 1: pull a completed data set for each imputation ...

```
NA's :192    NA's :207    NA's :257    NA's :246    NA's :206
```

- Replace the “_i” variables with fillins from “_o” data (as.integer coerces to integer).

```
ordvars <- grep("^.*_o$", colnames(imps[[1]]), value = TRUE)
## Re-create the integer coded variables
for(i in seq_along(imps)){
  oneds <- imps[[i]]
  for(j in ordvars){
    newvar <- as.integer(oneds[ , j])
    newname <- gsub("_o$", "_i", j)
    oneds[ , newname] <- as.integer(oneds[ , j])
  }
  imps[[i]] <- oneds
}
```

Step 2: Run each data set

```
sem.01.int.fits <- list()
for(i in seq_along(imps)){
  sem.01.int.fits[[i]] <- sem(model = sem.01.int, mimic =
    "Mplus",
                             estimator = "ML", data = imps[[i]])
}
```

Step 3: Pool the fits

```
library(mitools)
int.betas <- MIextract(sem.01.int.fits, fun = coef)
int.vars <- MIextract(sem.01.int.fits, fun = vcov)
int.fits.pooled <- summary(MIcombine(int.betas, int.vars))
```

And the result is

```
int.fits.pooled
```

	results	se	(lower	upper)	missInfo
gotBully~gotBui_i	1.1982834	0.025817251	1.14548816	1.2510786	41 %
gotBully~gotBu2_i	1.2450029	0.027455181	1.18693907	1.3030667	54 %
gotBully~gotBu3_i	1.3261447	0.027159753	1.26782438	1.3844651	59 %
gotBully~gotBu4_i	1.2742133	0.023171474	1.22774321	1.3206833	30 %
gotBully~gotBu5_i	1.3822795	0.024417522	1.33151645	1.4330425	48 %
gotBully~gotBu6_i	1.3933757	0.032275120	1.31864261	1.4681088	77 %
gotBully~gotBu7_i	1.3330777	0.029120884	1.27003544	1.3961200	62 %
gotBully~gotBu8_i	1.3018800	0.034611116	1.22027251	1.3834875	80 %
gotBully~gotBu9_i	1.2643364	0.022848186	1.21635226	1.3123205	52 %
alcohol~alc1_i	0.6476732	0.015491549	0.61347024	0.6818762	67 %
alcohol~alc2_i	0.6086073	0.020824494	0.55889546	0.6583191	82 %
alcohol~alc3_i	0.7075967	0.019749221	0.66025848	0.7549349	83 %
alcohol~alc4_i	0.7256755	0.021750158	0.67421283	0.7771382	81 %
alcohol~alc5_i	0.7334636	0.019139446	0.68869426	0.7782330	79 %
alcohol~gotBully	0.1355003	0.022911952	0.08691018	0.1840905	55 %
gotBu1_i~gotBu1_i	1.0353404	0.033400361	0.96454949	1.1061312	55 %
gotBu2_i~gotBu2_i	0.8295723	0.023851361	0.78103522	0.8781094	39 %
gotBu3_i~gotBu3_i	0.5582119	0.021624500	0.51050299	0.6059207	67 %
gotBu4_i~gotBu4_i	0.8244160	0.027230818	0.76663301	0.8821990	56 %
gotBu5_i~gotBu5_i	0.4669792	0.014498018	0.43749403	0.4964644	38 %
gotBu6_i~gotBu6_i	0.2842870	0.012326163	0.25745551	0.3111184	63 %
gotBu7_i~gotBu7_i	0.6816414	0.026360862	0.62313504	0.7401477	68 %
gotBu8_i~gotBu8_i	0.4134766	0.035781117	0.32026300	0.5066903	94 %
gotBu9_i~gotBu9_i	0.3682245	0.016117252	0.33193495	0.4045141	71 %
alc1_i~alc1_i	0.1776069	0.006686170	0.16328234	0.1919314	59 %
alc2_i~alc2_i	0.2215927	0.008605126	0.20263426	0.2405512	66 %
alc3_i~alc3_i	0.1044154	0.005932963	0.09089545	0.1179353	74 %
alc4_i~alc4_i	0.2137228	0.013176746	0.18133817	0.2461073	86 %
alc5_i~alc5_i	0.1330404	0.008305480	0.11322733	0.1528534	82 %

And the result is ...

gotBu1_i~1	2.4945845	0.025778243	2.44379283	2.5453762	14 %
gotBu2_i~1	2.3512138	0.026920128	2.29756704	2.4048606	25 %
gotBu3_i~1	2.1739496	0.027201024	2.11944390	2.2284553	29 %
gotBu4_i~1	2.4711951	0.024249163	2.42365625	2.5187340	3 %
gotBu5_i~1	2.1598973	0.028645275	2.10191830	2.2178763	36 %
gotBu6_i~1	2.0893091	0.029804433	2.02768726	2.1509308	46 %
gotBu7_i~1	2.3285714	0.027479934	2.27375776	2.3833851	26 %
gotBu8_i~1	2.0049953	0.028322632	1.94686221	2.0631285	43 %
gotBu9_i~1	1.9584967	0.024992128	1.90845074	2.0085427	29 %
alc1_i~1	1.2745098	0.013620703	1.24734524	1.3016744	26 %
alc2_i~1	1.3049953	0.013959416	1.27695972	1.3330309	31 %
alc3_i~1	1.2450980	0.015150098	1.21408709	1.2761090	41 %
alc4_i~1	1.3452848	0.014556527	1.31649282	1.3740767	19 %
alc5_i~1	1.3007470	0.014168749	1.27260883	1.3288851	22 %

Ordinal version of the SEM

```
sem.01.ord <- '  
  ## the measurement model  
  gotBully =~ NA*gotBu1_o + gotBu2_o + gotBu3_o  
              + gotBu4_o + gotBu5_o + gotBu6_o  
              + gotBu7_o + gotBu8_o + gotBu9_o  
  gotBully =~ 1*gotBully  
  alcohol =~ NA*alc1_o + alc2_o + alc3_o  
            + alc4_o + alc5_o  
  alcohol =~ 1*alcohol  
  # regress  
  alcohol ~ gotBully  
,  
  
## Run on each imputed data set  
sem.01.ord.fits <- list()  
for(i in seq_along(imps)){  
  sem.01.ord.fits[[i]] <- sem(model = sem.01.ord, mimic =  
    "Mplus",  
                               estimator = "DWLS", data = imps[[i]])  
}
```

Ordinal version of the SEM ...

```
ord.betas <- MIextract(sem.01.ord.fits, fun = coef)
ord.vars <- MIextract(sem.01.ord.fits, fun = vcov)
ord.fits.pooled <- summary(MIcombine(ord.betas, ord.vars))
```

And the result is

```
ord.fits.pooled
```

	results	se	(lower	upper)	missInfo
gotBully~gotBui_o	0.84958774	0.0063444951	0.834605389	0.86457008	80 %
gotBully~gotBu2_o	0.88391019	0.006161330	0.869160713	0.89865966	82 %
gotBully~gotBu3_o	0.90993262	0.004768570	0.899023760	0.92084148	74 %
gotBully~gotBu4_o	0.88371316	0.006461694	0.868053337	0.89937298	84 %
gotBully~gotBu5_o	0.93298054	0.003428420	0.925764495	0.94019658	53 %
gotBully~gotBu6_o	0.94237464	0.003834359	0.933823365	0.95092592	69 %
gotBully~gotBu7_o	0.89415037	0.006826189	0.877197891	0.91110285	88 %
gotBully~gotBu8_o	0.93080094	0.006285709	0.915263005	0.94633887	87 %
gotBully~gotBu9_o	0.92745771	0.005222476	0.915054060	0.93986135	81 %
alcohol~alc1_o	0.89462489	0.005797322	0.883249783	0.90600000	6 %
alcohol~alc2_o	0.84983885	0.007716939	0.834169168	0.86550853	37 %
alcohol~alc3_o	0.96226982	0.005996541	0.950088279	0.97445136	38 %
alcohol~alc4_o	0.91009138	0.006728283	0.896163608	0.92401915	46 %
alcohol~alc5_o	0.93570462	0.005306686	0.925230649	0.94617859	16 %
alcohol~gotBully	0.13621288	0.019672222	0.083554569	0.18887120	96 %
gotBui_o t1	-0.18025162	0.019848098	-0.219192761	-0.14131047	6 %
gotBui_o t2	0.20648468	0.020346800	0.166484034	0.24648532	10 %
gotBui_o t3	0.44699079	0.024149766	0.398111417	0.49587016	36 %
gotBui_o t4	0.92536957	0.026863698	0.871229173	0.97950998	33 %
gotBu2_o t1	-0.05842792	0.019585078	-0.096835130	-0.02002071	4 %
gotBu2_o t2	0.25661456	0.020393365	0.216530774	0.29669834	10 %
gotBu2_o t3	0.64642332	0.023177563	0.600416892	0.69242976	22 %
gotBu2_o t4	0.95377132	0.036924031	0.871842355	1.03570029	68 %
gotBu3_o t1	0.17620880	0.019708607	0.137556871	0.21486073	5 %
gotBu3_o t2	0.36111032	0.020595593	0.320637161	0.40158348	10 %
gotBu3_o t3	0.66554355	0.032212191	0.595179233	0.73590786	64 %
gotBu3_o t4	1.11739421	0.037715373	1.034872364	1.19991605	64 %
gotBu4_o t1	-0.16682771	0.020046256	-0.206190214	-0.12746520	8 %
gotBu4_o t2	0.20457132	0.019985784	0.165345464	0.24379718	7 %

And the result is ...

gotBu4_o t3	0.50652007	0.023907583	0.458394041	0.55464609	32 %
gotBu4_o t4	0.91648471	0.030057757	0.853816238	0.97915318	49 %
gotBu5_o t1	0.22132930	0.019573921	0.182957534	0.25970106	3 %
gotBu5_o t2	0.37840721	0.021549667	0.335807367	0.42100705	18 %
gotBu5_o t3	0.65724665	0.032582587	0.585762639	0.72873066	65 %
gotBu5_o t4	1.08342164	0.033644126	1.012162629	1.15468064	55 %
gotBu6_o t1	0.31206147	0.019982841	0.272868070	0.35125486	5 %
gotBu6_o t2	0.42573197	0.020966849	0.384490407	0.46697354	11 %
gotBu6_o t3	0.51517916	0.022955533	0.469435865	0.56092245	25 %
gotBu6_o t4	1.45569057	0.104533640	1.181433079	1.72994806	94 %
gotBu7_o t1	0.04073535	0.019458141	0.002587177	0.07888353	3 %
gotBu7_o t2	0.27504051	0.019969890	0.235864824	0.31421620	6 %
gotBu7_o t3	0.51497543	0.029971895	0.450334428	0.57961643	61 %
gotBu7_o t4	1.03725488	0.034444277	0.963288705	1.11122105	59 %
gotBu8_o t1	0.34516588	0.020247499	0.305429630	0.38490212	7 %
gotBu8_o t2	0.43844745	0.020415602	0.398395506	0.47849940	6 %
gotBu8_o t3	0.83578886	0.058537977	0.688547829	0.98302990	89 %
gotBu8_o t4	1.23989330	0.044668817	1.138618682	1.34116792	73 %
gotBu9_o t1	0.38305674	0.020457485	0.342891571	0.42322192	8 %
gotBu9_o t2	0.44761238	0.020359360	0.407680912	0.48754385	5 %
gotBu9_o t3	0.82460462	0.042447256	0.725171596	0.92403763	79 %
gotBu9_o t4	1.43632886	0.044858570	1.337739615	1.53491811	66 %
alc1_o t1	1.00068096	0.024685676	0.952064348	1.04929756	13 %
alc1_o t2	1.59633019	0.036545128	1.523121350	1.66953904	29 %
alc1_o t3	1.81698110	0.044636984	1.726485983	1.90747622	37 %
alc1_o t4	1.94090120	0.042844866	1.856559580	2.02524282	13 %
alc2_o t1	0.87375797	0.023377110	0.827772782	0.91974316	11 %
alc2_o t2	1.58979976	0.039621433	1.508565753	1.67103377	42 %
alc2_o t3	1.79894166	0.043221272	1.711759492	1.88612382	34 %
alc2_o t4	2.01736685	0.049045720	1.919576104	2.11515759	26 %
alc3_o t1	1.18149348	0.029036807	1.123349481	1.23963747	29 %
alc3_o t2	1.54035669	0.036845312	1.465707848	1.61500552	36 %
alc3_o t3	1.74254979	0.046671227	1.645072080	1.84002751	50 %
alc3_o t4	1.98039030	0.049366118	1.881069379	2.07971122	32 %

And the result is ...

```
55 | alc4_olt1      0.88650811 0.023037441 0.841278008 0.93173822    8 %  
    | alc4_olt2      1.39286736 0.030031505 1.333606443 1.45212827   16 %  
    | alc4_olt3      1.67533889 0.040814007 1.592315132 1.75836265   38 %  
    | alc4_olt4      1.89882735 0.042206396 1.815520394 1.98213430   16 %  
70 | alc5_olt1      0.98269055 0.024306087 0.934874273 1.03050682   12 %  
    | alc5_olt2      1.49938979 0.032866291 1.434206701 1.56457288   21 %  
    | alc5_olt3      1.70311709 0.041466716 1.618843963 1.78739021   38 %  
    | alc5_olt4      1.93267970 0.045629373 1.841740117 2.02361929   25 %
```

For comparison

```
sem.01.ord.ld <- sem(model = sem.01.ord, mimic = "Mplus",  
                    estimator = "DWLS", data = hbsc,  
                    missing = "listwise")  
sem.01.ord.pair <- sem(model = sem.01.ord, mimic = "Mplus",  
                      estimator = "DWLS", data = hbsc,  
                      missing = "pairwise")
```

```
table509 <- semTable(list("listwise" = sem.01.ord.ld,  
                         "pairwise" = sem.01.ord.pair),  
                    columns = c("estsestars", "rsquare"), type =  
                              "tex", longtable=TRUE,  
                    file = file.path(odir, "semtable509"),  
                    print.results=FALSE)
```

Double-check a couple of the fitted models

Results on coefficients are quite different.

- 1 Listwise deletion much like pairwise complete analysis
- 2 Note Regression slope with listwise deletion twice as large, is this meaningful?
- 3 Wondered if there is something wrong in the pooling, check a couple of models from the list

```
summary(sem.01.ord.fits[[1]])
```

```
lavaan 0.6-3 ended normally after 25 iterations
```

Optimization method	NLMINB
Number of free parameters	71
Number of observations	4284
Estimator	DWLS
Model Fit Test Statistic	655.170
Degrees of freedom	76
P-value (Chi-square)	0.000

Double-check a couple of the fitted models ...

Parameter Estimates:

Information	Expected
Information saturated (h1) model	Unstructured
Standard Errors	Standard

Latent Variables:

	Estimate	Std.Err	z-value	P(> z)
gotBully =~				
gotBu1_o	0.844	0.003	261.198	0.000
gotBu2_o	0.875	0.003	293.967	0.000
gotBu3_o	0.905	0.003	335.418	0.000
gotBu4_o	0.875	0.003	294.915	0.000
gotBu5_o	0.932	0.002	373.082	0.000
gotBu6_o	0.943	0.002	406.795	0.000
gotBu7_o	0.891	0.003	323.355	0.000
gotBu8_o	0.938	0.003	375.064	0.000
gotBu9_o	0.927	0.003	366.829	0.000
alcohol =~				
alc1_o	0.896	0.005	164.537	0.000
alc2_o	0.856	0.006	141.664	0.000
alc3_o	0.958	0.005	201.311	0.000
alc4_o	0.912	0.005	182.528	0.000
alc5_o	0.934	0.005	194.569	0.000

Regressions:

Double-check a couple of the fitted models ...

```

40      alcohol ~
      gotBully          Estimate   Std.Err   z-value   P(>|z|)
                                0.156     0.004    36.084    0.000

Intercepts:
                                Estimate   Std.Err   z-value   P(>|z|)
45      .gotBu1_o          0.000
      .gotBu2_o          0.000
      .gotBu3_o          0.000
      .gotBu4_o          0.000
      .gotBu5_o          0.000
50      .gotBu6_o          0.000
      .gotBu7_o          0.000
      .gotBu8_o          0.000
      .gotBu9_o          0.000
      .alc1_o            0.000
      .alc2_o            0.000
55      .alc3_o            0.000
      .alc4_o            0.000
      .alc5_o            0.000
      gotBully           0.000
      alcohol            0.000

60      Thresholds:
                                Estimate   Std.Err   z-value   P(>|z|)
      gotBu1_o|t1        -0.183     0.019    -9.497    0.000

```

Double-check a couple of the fitted models ...

55	gotBu1_o t2	0.210	0.019	10.900	0.000
	gotBu1_o t3	0.460	0.020	23.100	0.000
	gotBu1_o t4	0.927	0.022	41.271	0.000
	gotBu2_o t1	-0.054	0.019	-2.841	0.004
	gotBu2_o t2	0.264	0.019	13.611	0.000
70	gotBu2_o t3	0.658	0.021	31.749	0.000
	gotBu2_o t4	1.000	0.023	43.335	0.000
	gotBu3_o t1	0.182	0.019	9.467	0.000
	gotBu3_o t2	0.368	0.020	18.747	0.000
	gotBu3_o t3	0.689	0.021	32.951	0.000
75	gotBu3_o t4	1.154	0.025	46.939	0.000
	gotBu4_o t1	-0.163	0.019	-8.490	0.000
	gotBu4_o t2	0.207	0.019	10.717	0.000
	gotBu4_o t3	0.500	0.020	24.935	0.000
	gotBu4_o t4	0.936	0.023	41.540	0.000
80	gotBu5_o t1	0.222	0.019	11.509	0.000
	gotBu5_o t2	0.383	0.020	19.474	0.000
	gotBu5_o t3	0.672	0.021	32.306	0.000
	gotBu5_o t4	1.091	0.024	45.588	0.000
	gotBu6_o t1	0.313	0.019	16.075	0.000
85	gotBu6_o t2	0.422	0.020	21.319	0.000
	gotBu6_o t3	0.520	0.020	25.835	0.000
	gotBu6_o t4	1.481	0.029	50.847	0.000
	gotBu7_o t1	0.044	0.019	2.291	0.022
	gotBu7_o t2	0.274	0.019	14.129	0.000
90	gotBu7_o t3	0.535	0.020	26.494	0.000

Double-check a couple of the fitted models ...

gotBu7_o t4	1.070	0.024	45.102	0.000
gotBu8_o t1	0.347	0.020	17.715	0.000
gotBu8_o t2	0.439	0.020	22.135	0.000
gotBu8_o t3	0.762	0.021	35.763	0.000
gotBu8_o t4	1.227	0.025	48.244	0.000
gotBu9_o t1	0.386	0.020	19.595	0.000
gotBu9_o t2	0.447	0.020	22.527	0.000
gotBu9_o t3	0.810	0.022	37.474	0.000
gotBu9_o t4	1.452	0.029	50.703	0.000
alc1_o t1	0.993	0.023	43.154	0.000
alc1_o t2	1.591	0.031	51.042	0.000
alc1_o t3	1.800	0.036	49.973	0.000
alc1_o t4	1.922	0.040	48.562	0.000
alc2_o t1	0.865	0.022	39.326	0.000
alc2_o t2	1.567	0.031	51.046	0.000
alc2_o t3	1.771	0.035	50.225	0.000
alc2_o t4	1.985	0.042	47.635	0.000
alc3_o t1	1.170	0.025	47.250	0.000
alc3_o t2	1.541	0.030	51.022	0.000
alc3_o t3	1.725	0.034	50.560	0.000
alc3_o t4	1.956	0.041	48.071	0.000
alc4_o t1	0.877	0.022	39.714	0.000
alc4_o t2	1.391	0.028	50.273	0.000
alc4_o t3	1.664	0.033	50.875	0.000
alc4_o t4	1.883	0.038	49.074	0.000
alc5_o t1	0.985	0.023	42.945	0.000

Double-check a couple of the fitted models ...

```

alc5_o|t2      1.499    0.029    50.916    0.000
alc5_o|t3      1.680    0.033    50.806    0.000
alc5_o|t4      1.900    0.039    48.853    0.000

```

Variances:

```

                Estimate  Std.Err  z-value  P(>|z|)
gotBully          1.000
.alcohol          1.000
.gotBu1_o         0.287
.gotBu2_o         0.234
.gotBu3_o         0.182
.gotBu4_o         0.234
.gotBu5_o         0.131
.gotBu6_o         0.111
.gotBu7_o         0.207
.gotBu8_o         0.120
.gotBu9_o         0.141
.alc1_o           0.178
.alc2_o           0.250
.alc3_o           0.060
.alc4_o           0.149
.alc5_o           0.107

```

Scales y*:

```

                Estimate  Std.Err  z-value  P(>|z|)
gotBu1_o          1.000

```

Double-check a couple of the fitted models ...

```
45 gotBu2_o      1.000
   gotBu3_o      1.000
   gotBu4_o      1.000
   gotBu5_o      1.000
   gotBu6_o      1.000
   gotBu7_o      1.000
   gotBu8_o      1.000
50 gotBu9_o      1.000
   alc1_o        1.000
   alc2_o        1.000
   alc3_o        1.000
   alc4_o        1.000
55 alc5_o        1.000
```

```
summary(sem.01.ord.fits[[4]])
```

Double-check a couple of the fitted models ...

```
lavaan 0.6-3 ended normally after 28 iterations
```

```

Optimization method           NLMINB
Number of free parameters      71
Number of observations         4284
Estimator                      DWLS
Model Fit Test Statistic      642.392
Degrees of freedom             76
P-value (Chi-square)          0.000

```

```
Parameter Estimates:
```

```

Information                    Expected
Information saturated (h1) model Unstructured
Standard Errors                 Standard

```

```
Latent Variables:
```

	Estimate	Std.Err	z-value	P(> z)
gotBully =~				
gotBu1_o	0.851	0.003	272.363	0.000
gotBu2_o	0.887	0.003	307.660	0.000
gotBu3_o	0.910	0.003	340.966	0.000
gotBu4_o	0.884	0.003	305.526	0.000

Double-check a couple of the fitted models ...

```

gotBu5_o      0.936    0.002   379.021    0.000
gotBu6_o      0.941    0.002   404.475    0.000
gotBu7_o      0.889    0.003   320.817    0.000
gotBu8_o      0.924    0.003   345.769    0.000
gotBu9_o      0.927    0.003   361.097    0.000
alcohol =~
alc1_o        0.895    0.006   159.786    0.000
alc2_o        0.848    0.006   134.403    0.000
alc3_o        0.960    0.005   197.400    0.000
alc4_o        0.911    0.005   178.514    0.000
alc5_o        0.939    0.005   193.104    0.000

Regressions:
      Estimate  Std.Err  z-value  P(>|z|)
alcohol ~
gotBully      0.135    0.004    31.283    0.000

Intercepts:
      Estimate  Std.Err  z-value  P(>|z|)
.gotBu1_o    0.000
.gotBu2_o    0.000
.gotBu3_o    0.000
.gotBu4_o    0.000
.gotBu5_o    0.000
.gotBu6_o    0.000
.gotBu7_o    0.000

```

Double-check a couple of the fitted models ...

```

.gotBu8_o      0.000
.gotBu9_o      0.000
.alc1_o        0.000
55 .alc2_o        0.000
.alc3_o        0.000
.alc4_o        0.000
.alc5_o        0.000
.gotBully      0.000
50 .alcohol      0.000

```

Thresholds:

	Estimate	Std.Err	z-value	P(> z)
gotBu1_o t1	-0.186	0.019	-9.650	0.000
gotBu1_o t2	0.199	0.019	10.321	0.000
gotBu1_o t3	0.436	0.020	21.984	0.000
gotBu1_o t4	0.928	0.022	41.298	0.000
gotBu2_o t1	-0.064	0.019	-3.330	0.001
gotBu2_o t2	0.249	0.019	12.850	0.000
gotBu2_o t3	0.643	0.021	31.101	0.000
gotBu2_o t4	0.934	0.023	41.486	0.000
gotBu3_o t1	0.174	0.019	9.039	0.000
gotBu3_o t2	0.355	0.020	18.110	0.000
gotBu3_o t3	0.691	0.021	33.038	0.000
gotBu3_o t4	1.120	0.024	46.228	0.000
gotBu4_o t1	-0.175	0.019	-9.100	0.000
gotBu4_o t2	0.197	0.019	10.199	0.000

Double-check a couple of the fitted models ...

gotBu4_o t3	0.512	0.020	25.505	0.000
gotBu4_o t4	0.916	0.022	40.946	0.000
gotBu5_o t1	0.217	0.019	11.235	0.000
gotBu5_o t2	0.365	0.020	18.625	0.000
gotBu5_o t3	0.650	0.021	31.425	0.000
gotBu5_o t4	1.066	0.024	45.003	0.000
gotBu6_o t1	0.305	0.019	15.680	0.000
gotBu6_o t2	0.422	0.020	21.319	0.000
gotBu6_o t3	0.510	0.020	25.385	0.000
gotBu6_o t4	1.509	0.030	50.951	0.000
gotBu7_o t1	0.037	0.019	1.955	0.051
gotBu7_o t2	0.269	0.019	13.885	0.000
gotBu7_o t3	0.535	0.020	26.494	0.000
gotBu7_o t4	1.036	0.023	44.281	0.000
gotBu8_o t1	0.337	0.020	17.230	0.000
gotBu8_o t2	0.431	0.020	21.773	0.000
gotBu8_o t3	0.880	0.022	39.824	0.000
gotBu8_o t4	1.199	0.025	47.768	0.000
gotBu9_o t1	0.375	0.020	19.080	0.000
gotBu9_o t2	0.441	0.020	22.225	0.000
gotBu9_o t3	0.852	0.022	38.880	0.000
gotBu9_o t4	1.419	0.028	50.494	0.000
alc1_o t1	0.999	0.023	43.309	0.000
alc1_o t2	1.591	0.031	51.042	0.000
alc1_o t3	1.809	0.036	49.888	0.000
alc1_o t4	1.949	0.040	48.186	0.000

Double-check a couple of the fitted models ...

alc2_o t1	0.873	0.022	39.575	0.000
alc2_o t2	1.587	0.031	51.044	0.000
alc2_o t3	1.794	0.036	50.027	0.000
alc2_o t4	2.033	0.043	46.839	0.000
alc3_o t1	1.186	0.025	47.533	0.000
alc3_o t2	1.539	0.030	51.019	0.000
alc3_o t3	1.749	0.035	50.397	0.000
alc3_o t4	1.981	0.042	47.700	0.000
alc4_o t1	0.888	0.022	40.072	0.000
alc4_o t2	1.391	0.028	50.273	0.000
alc4_o t3	1.666	0.033	50.866	0.000
alc4_o t4	1.893	0.039	48.943	0.000
alc5_o t1	0.980	0.023	42.788	0.000
alc5_o t2	1.497	0.029	50.910	0.000
alc5_o t3	1.702	0.034	50.696	0.000
alc5_o t4	1.937	0.040	48.353	0.000

Variances:

	Estimate	Std.Err	z-value	P(> z)
gotBully	1.000			
.alcohol	1.000			
.gotBu1_o	0.275			
.gotBu2_o	0.214			
.gotBu3_o	0.172			
.gotBu4_o	0.218			
.gotBu5_o	0.123			

Double-check a couple of the fitted models ...

```
30 .gotBu6_o      0.114
    .gotBu7_o      0.209
    .gotBu8_o      0.146
    .gotBu9_o      0.140
    .alc1_o        0.184
35 .alc2_o        0.267
    .alc3_o        0.061
    .alc4_o        0.154
    .alc5_o        0.103
```

40 Scales y*:

	Estimate	Std.Err	z-value	P(> z)
gotBu1_o	1.000			
gotBu2_o	1.000			
gotBu3_o	1.000			
45 gotBu4_o	1.000			
gotBu5_o	1.000			
gotBu6_o	1.000			
gotBu7_o	1.000			
gotBu8_o	1.000			
50 gotBu9_o	1.000			
alc1_o	1.000			
alc2_o	1.000			
alc3_o	1.000			
alc4_o	1.000			
55 alc5_o	1.000			

Double-check a couple of the fitted models ...

Multivariate Normal Imputation

```
library(Amelia)
## Drop the redundant integer versions of the variables
dato <- hbsc[ , grep("^.*_i$", colnames(hbsc), invert = TRUE)]
## Don't have good reason for emburn setting, but
## run was quitting so quickly I was not comfortable
datoamelia <- amelia(dato, m = 5L, p2s=0, idvars = "stud_id",
                    ords = grep("^.*_o$", colnames(hbsc),
                                value = TRUE),
                    noms = c("Gender", "Grade"),
                    emburn = c(100L, 10L))

## Extract the imputed complete data sets
imps <- datoamelia$imputations
## Now re-create the integer coded variables
ordvars <- grep("^.*_o$", colnames(imps[[1]]), value = TRUE)
for(i in seq_along(imps)){
  oneds <- imps[[i]]
  for(j in ordvars){
    newvar <- as.integer(oneds[ , j])
    newname <- gsub("_o$", "_i", j)
    oneds[ , newname] <- as.integer(oneds[ , j])
  }
}
```

Multivariate Normal Imputation ...

```
    imps[[i]] <- oneds
  }
saveRDS(imps, file.path(wdir, "amelia-imps.rds"))
```

Run Models with Saved Amelia Imputations

```
imps <- readRDS(file.path(wdir, "amelia-imps.rds"))
```

```
## Run same model for each imputed data set
ord.amelia.fits <- list()
for (i in seq_along(imps)){
  ord.amelia.fits[[i]] <- sem(model = sem.01.ord, data =
    imps[[i]],
    mimic = "Mplus", estimator = "DWLS")
}
```

```
ord.amelia.betas <- MIextract(ord.amelia.fits, fun = coef)
ord.amelia.vars <- MIextract(ord.amelia.fits, fun = vcov)
ord.amelia.fits.pooled <- summary(MIcombine(ord.amelia.betas,
  ord.amelia.vars))
```

Run Models with Saved Amelia Imputations

```
Multiple imputation results:
      MIcombine.default(ord.amelia.betas, ord.amelia.vars)
              results          se      (lower      upper)
              missInfo
gotBully=~gotBu1_o 0.6991060 0.014690604 0.6652397635 0.7329723      76
%
gotBully=~gotBu2_o 0.7243062 0.014202636 0.6920209397 0.7565914      73
%
gotBully=~gotBu3_o 0.7389645 0.011076978 0.7160359961 0.7618930      46
%
gotBully=~gotBu4_o 0.7362147 0.014396653 0.7032196288 0.7692097      75
%
gotBully=~gotBu5_o 0.7717544 0.012933985 0.7436246697 0.7998842      63
%
gotBully=~gotBu6_o 0.7975737 0.009336515 0.7792017503 0.8159457      12
%
gotBully=~gotBu7_o 0.7123489 0.011458063 0.6882973027 0.7364005      52
%
gotBully=~gotBu8_o 0.7681203 0.017041575 0.7290167509 0.8072238      75
%
gotBully=~gotBu9_o 0.7625237 0.017669718 0.7218136093 0.8032338      76
%
alcohol=~alc1_o    0.8184650 0.010053358 0.7984271789 0.8385027      25
%
```


Run Models with Saved Amelia Imputations ...

alcohol=~alc2_o	0.7531594	0.010397894	0.7325270365	0.7737918	22
%					
alcohol=~alc3_o	0.8963321	0.010472284	0.8749751740	0.9176891	40
%					
alcohol=~alc4_o	0.8342438	0.012003417	0.8086249009	0.8598628	57
%					
alcohol=~alc5_o	0.8709425	0.009381282	0.8522183869	0.8896667	27
%					
alcohol~gotBully	0.2783286	0.019147646	0.2299428918	0.3267144	90
%					
gotBu1_o t1	0.1186529	0.048091023	-0.0007469421	0.2380527	88
%					
gotBu1_o t2	0.6728061	0.064265091	0.5075447513	0.8380675	92
%					
gotBu1_o t3	0.9953136	0.068846094	0.8191122512	1.1715150	92
%					
gotBu1_o t4	1.3462670	0.058942318	1.2042684901	1.4882655	83
%					
gotBu2_o t1	0.2721201	0.056432193	0.1282898763	0.4159502	91
%					
gotBu2_o t2	0.8106289	0.067512230	0.6367509433	0.9845068	92
%					
gotBu2_o t3	1.1539149	0.073731234	0.9651324268	1.3426973	92
%					
gotBu2_o t4	1.5588790	0.058184826	1.4235723569	1.6941856	78
%					

Run Models with Saved Amelia Imputations ...

gotBu3_o t1	0.6540871	0.034721587	0.5763373553	0.7318369	70
%					
gotBu3_o t2	1.1352009	0.035439489	1.0594056868	1.2109961	58
%					
gotBu3_o t3	1.4975565	0.041633297	1.4092835583	1.5858295	55
%					
gotBu3_o t4	1.8481843	0.042570916	1.7633933573	1.9329752	25
%					
gotBu4_o t1	0.1920224	0.043319456	0.0869304765	0.2971144	84
%					
gotBu4_o t2	0.7822849	0.049881329	0.6603305509	0.9042393	86
%					
gotBu4_o t3	1.1556472	0.040862486	1.0644153123	1.2468791	69
%					
gotBu4_o t4	1.5234609	0.039862383	1.4405296675	1.6063921	48
%					
gotBu5_o t1	0.7505417	0.042038828	0.6517395218	0.8493438	79
%					
gotBu5_o t2	1.2074801	0.044195285	1.1071336105	1.3078267	73
%					
gotBu5_o t3	1.5503738	0.036405629	1.4769764757	1.6237710	33
%					
gotBu5_o t4	1.9065720	0.040104990	1.8279089877	1.9852350	5
%					
gotBu6_o t1	0.9320516	0.065301696	0.7656765430	1.0984267	91
%					

Run Models with Saved Amelia Imputations ...

40	gotBu6_o t2	1.4160054	0.052945798	1.2931994510	1.5388114	77
	%					
	gotBu6_o t3	1.7967755	0.046768174	1.7002307643	1.8933202	45
	%					
	gotBu6_o t4	2.1329846	0.053211634	2.0273388751	2.2386304	22
	%					
	gotBu7_o t1	0.5002629	0.045196983	0.3905328519	0.6099929	85
	%					
	gotBu7_o t2	0.9791734	0.050717131	0.8565431823	1.1018037	84
	%					
45	gotBu7_o t3	1.2960440	0.044000713	1.1975933131	1.3944947	70
	%					
	gotBu7_o t4	1.6678122	0.041616117	1.5825504498	1.7530740	42
	%					
	gotBu8_o t1	1.0504286	0.065349344	0.8851193259	1.2157379	90
	%					
	gotBu8_o t2	1.5328972	0.068389775	1.3665558493	1.6992386	85
	%					
	gotBu8_o t3	1.8669101	0.059157175	1.7374506710	1.9963695	65
	%					
50	gotBu8_o t4	2.1634166	0.052867643	2.0590930006	2.2677401	16
	%					
	gotBu9_o t1	1.1085327	0.075344119	0.9144716568	1.3025938	92
	%					
	gotBu9_o t2	1.5723952	0.080658357	1.3705164486	1.7742739	89
	%					

Run Models with Saved Amelia Imputations ...

gotBu9_o t3	1.8972073	0.060949151	1.7635665748	2.0308480	65
%					
gotBu9_o t4	2.1744520	0.052956630	2.0700847467	2.2788192	14
%					
alc1_o t1	1.0838649	0.024713178	1.0353598145	1.1323700	7
%					
alc1_o t2	1.8222068	0.039079810	1.7452678606	1.8991457	13
%					
alc1_o t3	2.1559579	0.049407506	2.0590743034	2.2528415	4
%					
alc1_o t4	2.3232203	0.057023186	2.2114559065	2.4349847	1
%					
alc2_o t1	0.9452435	0.023049797	0.9000471012	0.9904399	4
%					
alc2_o t2	1.7927941	0.037263199	1.7196338040	1.8659544	8
%					
alc2_o t3	2.0799246	0.045649850	1.9904424129	2.1694068	2
%					
alc2_o t4	2.4162240	0.063465181	2.2918060063	2.5406419	3
%					
alc3_o t1	1.2920387	0.028085131	1.2367279968	1.3473493	13
%					
alc3_o t2	1.7904809	0.037574521	1.7166399793	1.8643217	10
%					
alc3_o t3	2.0749249	0.046437800	1.9838075236	2.1660423	6
%					

Run Models with Saved Amelia Imputations ...

alc3_o t4 %	2.3576145	0.059155343	2.2416692259	2.4735597	1
alc4_o t1 %	0.9670501	0.022932909	0.9221002399	1.0119999	1
alc4_o t2 %	1.5780912	0.033629754	1.5116963884	1.6444859	16
alc4_o t3 %	1.9339211	0.040759499	1.8539977309	2.0138445	4
alc4_o t4 %	2.2118150	0.051891286	2.1100745672	2.3135553	3
alc5_o t1 %	1.0686177	0.024414241	1.0207193669	1.1165161	6
alc5_o t2 %	1.7269411	0.039685571	1.6475528981	1.8063294	28
alc5_o t3 %	2.0362106	0.044757629	1.9484083280	2.1240128	6
alc5_o t4 %	2.3045093	0.056033477	2.1946835865	2.4143349	1

Amelia imputation results contradict

- Note the estimate here for the alcohol ~ bully linkage is about the same as we got with listwise deletion
- Again, I wondered if there is something wrong in the pooling, check a couple of models from the list

```
summary(ord.amelia.fits[[1]])
```

```
lavaan 0.6-3 ended normally after 17 iterations
```

Optimization method	NLMINB
Number of free parameters	71
Number of observations	4284
Estimator	DWLS
Model Fit Test Statistic	764.863
Degrees of freedom	76
P-value (Chi-square)	0.000

```
Parameter Estimates:
```

Information	Expected
-------------	----------

Amelia imputation results contradict ...

Information saturated (h1)	model	Unstructured		
Standard Errors		Standard		
Latent Variables:				
	Estimate	Std.Err	z-value	P(> z)
gotBully =~				
gotBu1_o	0.715	0.008	92.431	0.000
gotBu2_o	0.737	0.008	94.100	0.000
gotBu3_o	0.737	0.008	89.405	0.000
gotBu4_o	0.750	0.008	97.328	0.000
gotBu5_o	0.772	0.008	93.973	0.000
gotBu6_o	0.797	0.009	93.741	0.000
gotBu7_o	0.718	0.008	88.441	0.000
gotBu8_o	0.776	0.009	84.381	0.000
gotBu9_o	0.764	0.009	81.406	0.000
alcohol =~				
alc1_o	0.815	0.009	92.833	0.000
alc2_o	0.753	0.009	81.492	0.000
alc3_o	0.890	0.008	106.425	0.000
alc4_o	0.832	0.008	100.727	0.000
alc5_o	0.872	0.008	107.576	0.000
Regressions:				
	Estimate	Std.Err	z-value	P(> z)
alcohol ~				
gotBully	0.291	0.007	42.121	0.000

Amelia imutation results contradict ...

Intercepts:

	Estimate	Std.Err	z-value	P(> z)
.gotBu1_o	0.000			
.gotBu2_o	0.000			
.gotBu3_o	0.000			
.gotBu4_o	0.000			
.gotBu5_o	0.000			
.gotBu6_o	0.000			
.gotBu7_o	0.000			
.gotBu8_o	0.000			
.gotBu9_o	0.000			
.alc1_o	0.000			
.alc2_o	0.000			
.alc3_o	0.000			
.alc4_o	0.000			
.alc5_o	0.000			
gotBully	0.000			
.alcohol	0.000			

Thresholds:

	Estimate	Std.Err	z-value	P(> z)
gotBu1_o t1	0.133	0.019	6.934	0.000
gotBu1_o t2	0.691	0.021	33.067	0.000
gotBu1_o t3	1.017	0.023	43.798	0.000
gotBu1_o t4	1.361	0.027	49.995	0.000

Amelia imputation results contradict ...

gotBu2_o t1	0.283	0.019	14.585	0.000
gotBu2_o t2	0.820	0.022	37.814	0.000
gotBu2_o t3	1.157	0.025	46.984	0.000
gotBu2_o t4	1.569	0.031	51.047	0.000
gotBu3_o t1	0.640	0.021	31.012	0.000
gotBu3_o t2	1.143	0.024	46.713	0.000
gotBu3_o t3	1.500	0.029	50.922	0.000
gotBu3_o t4	1.849	0.037	49.466	0.000
gotBu4_o t1	0.171	0.019	8.856	0.000
gotBu4_o t2	0.760	0.021	35.677	0.000
gotBu4_o t3	1.142	0.024	46.690	0.000
gotBu4_o t4	1.506	0.030	50.940	0.000
gotBu5_o t1	0.731	0.021	34.609	0.000
gotBu5_o t2	1.190	0.025	47.619	0.000
gotBu5_o t3	1.524	0.030	50.990	0.000
gotBu5_o t4	1.893	0.039	48.943	0.000
gotBu6_o t1	0.896	0.022	40.319	0.000
gotBu6_o t2	1.378	0.027	50.165	0.000
gotBu6_o t3	1.779	0.035	50.154	0.000
gotBu6_o t4	2.114	0.047	45.375	0.000
gotBu7_o t1	0.486	0.020	24.304	0.000
gotBu7_o t2	0.961	0.023	42.261	0.000
gotBu7_o t3	1.265	0.026	48.831	0.000
gotBu7_o t4	1.659	0.033	50.893	0.000
gotBu8_o t1	1.062	0.024	44.929	0.000
gotBu8_o t2	1.524	0.030	50.990	0.000

Amelia imputation results contradict ...

gotBu8_o t3	1.883	0.038	49.074	0.000
gotBu8_o t4	2.172	0.049	44.214	0.000
gotBu9_o t1	1.094	0.024	45.660	0.000
gotBu9_o t2	1.591	0.031	51.042	0.000
gotBu9_o t3	1.904	0.039	48.806	0.000
gotBu9_o t4	2.172	0.049	44.214	0.000
alc1_o t1	1.088	0.024	45.539	0.000
alc1_o t2	1.812	0.036	49.859	0.000
alc1_o t3	2.166	0.049	44.341	0.000
alc1_o t4	2.325	0.057	40.814	0.000
alc2_o t1	0.941	0.023	41.674	0.000
alc2_o t2	1.803	0.036	49.945	0.000
alc2_o t3	2.072	0.045	46.159	0.000
alc2_o t4	2.401	0.062	38.979	0.000
alc3_o t1	1.286	0.026	49.125	0.000
alc3_o t2	1.774	0.035	50.202	0.000
alc3_o t3	2.062	0.044	46.338	0.000
alc3_o t4	2.352	0.059	40.173	0.000
alc4_o t1	0.966	0.023	42.420	0.000
alc4_o t2	1.587	0.031	51.044	0.000
alc4_o t3	1.933	0.040	48.406	0.000
alc4_o t4	2.217	0.051	43.252	0.000
alc5_o t1	1.073	0.024	45.175	0.000
alc5_o t2	1.730	0.034	50.526	0.000
alc5_o t3	2.038	0.044	46.759	0.000
alc5_o t4	2.300	0.056	41.408	0.000

Amelia imputation results contradict ...

Variances:

	Estimate	Std.Err	z-value	P(> z)
gotBully	1.000			
.alcohol	1.000			
.gotBu1_o	0.489			
.gotBu2_o	0.457			
.gotBu3_o	0.457			
.gotBu4_o	0.437			
.gotBu5_o	0.404			
.gotBu6_o	0.364			
.gotBu7_o	0.484			
.gotBu8_o	0.398			
.gotBu9_o	0.417			
.alc1_o	0.280			
.alc2_o	0.385			
.alc3_o	0.141			
.alc4_o	0.249			
.alc5_o	0.176			

Scales y*:

	Estimate	Std.Err	z-value	P(> z)
gotBu1_o	1.000			
gotBu2_o	1.000			
gotBu3_o	1.000			
gotBu4_o	1.000			

Amelia imputation results contradict ...

```

gotBu5_o      1.000
gotBu6_o      1.000
gotBu7_o      1.000
gotBu8_o      1.000
gotBu9_o      1.000
alc1_o        1.000
alc2_o        1.000
alc3_o        1.000
alc4_o        1.000
alc5_o        1.000

```

```
summary(ord.amelia.fits[[4]])
```

```
lavaan 0.6-3 ended normally after 18 iterations
```

Optimization method	NLMINB
Number of free parameters	71
Number of observations	4284
Estimator	DWLS
Model Fit Test Statistic	903.987
Degrees of freedom	76
P-value (Chi-square)	0.000

Amelia imputation results contradict ...

Parameter Estimates:

Information	Expected
Information saturated (h1) model	Unstructured
Standard Errors	Standard

Latent Variables:

	Estimate	Std.Err	z-value	P(> z)
gotBully =~				
gotBu1_o	0.696	0.008	85.674	0.000
gotBu2_o	0.716	0.008	87.057	0.000
gotBu3_o	0.732	0.009	84.147	0.000
gotBu4_o	0.720	0.008	87.517	0.000
gotBu5_o	0.777	0.009	88.781	0.000
gotBu6_o	0.802	0.009	87.388	0.000
gotBu7_o	0.713	0.009	82.949	0.000
gotBu8_o	0.780	0.010	79.846	0.000
gotBu9_o	0.785	0.010	78.526	0.000
alcohol =~				
alc1_o	0.815	0.009	92.809	0.000
alc2_o	0.748	0.009	80.726	0.000
alc3_o	0.893	0.008	106.294	0.000
alc4_o	0.829	0.008	99.381	0.000
alc5_o	0.865	0.008	105.867	0.000

Amelia imputation results contradict ...

Regressions:

	Estimate	Std.Err	z-value	P(> z)
alcohol ~ gotBully	0.298	0.007	41.621	0.000

Intercepts:

	Estimate	Std.Err	z-value	P(> z)
.gotBu1_o	0.000			
.gotBu2_o	0.000			
.gotBu3_o	0.000			
.gotBu4_o	0.000			
.gotBu5_o	0.000			
.gotBu6_o	0.000			
.gotBu7_o	0.000			
.gotBu8_o	0.000			
.gotBu9_o	0.000			
.alc1_o	0.000			
.alc2_o	0.000			
.alc3_o	0.000			
.alc4_o	0.000			
.alc5_o	0.000			
gotBully	0.000			
.alcohol	0.000			

Thresholds:

	Estimate	Std.Err	z-value	P(> z)
--	----------	---------	---------	---------

Amelia imputation results contradict ...

gotBu1_o t1	0.047	0.019	2.444	0.015
gotBu1_o t2	0.575	0.020	28.255	0.000
gotBu1_o t3	0.891	0.022	40.154	0.000
gotBu1_o t4	1.261	0.026	48.774	0.000
gotBu2_o t1	0.190	0.019	9.833	0.000
gotBu2_o t2	0.711	0.021	33.825	0.000
gotBu2_o t3	1.050	0.024	44.632	0.000
gotBu2_o t4	1.491	0.029	50.890	0.000
gotBu3_o t1	0.695	0.021	33.214	0.000
gotBu3_o t2	1.169	0.025	47.228	0.000
gotBu3_o t3	1.541	0.030	51.022	0.000
gotBu3_o t4	1.879	0.038	49.116	0.000
gotBu4_o t1	0.255	0.019	13.155	0.000
gotBu4_o t2	0.856	0.022	39.020	0.000
gotBu4_o t3	1.208	0.025	47.936	0.000
gotBu4_o t4	1.565	0.031	51.045	0.000
gotBu5_o t1	0.807	0.022	37.361	0.000
gotBu5_o t2	1.255	0.026	48.678	0.000
gotBu5_o t3	1.569	0.031	51.047	0.000
gotBu5_o t4	1.907	0.039	48.759	0.000
gotBu6_o t1	1.029	0.023	44.104	0.000
gotBu6_o t2	1.486	0.029	50.869	0.000
gotBu6_o t3	1.840	0.037	49.572	0.000
gotBu6_o t4	2.159	0.049	44.465	0.000
gotBu7_o t1	0.565	0.020	27.838	0.000
gotBu7_o t2	1.047	0.024	44.557	0.000

Amelia imputation results contradict ...

00	gotBu7_o t3	1.344	0.027	49.830	0.000
	gotBu7_o t4	1.707	0.034	50.668	0.000
	gotBu8_o t1	1.140	0.024	46.644	0.000
	gotBu8_o t2	1.623	0.032	50.997	0.000
	gotBu8_o t3	1.929	0.040	48.459	0.000
95	gotBu8_o t4	2.191	0.050	43.819	0.000
	gotBu9_o t1	1.212	0.025	47.998	0.000
	gotBu9_o t2	1.675	0.033	50.827	0.000
	gotBu9_o t3	1.960	0.041	48.011	0.000
	gotBu9_o t4	2.191	0.050	43.819	0.000
00	alc1_o t1	1.084	0.024	45.443	0.000
	alc1_o t2	1.809	0.036	49.888	0.000
	alc1_o t3	2.142	0.048	44.823	0.000
	alc1_o t4	2.325	0.057	40.814	0.000
	alc2_o t1	0.946	0.023	41.835	0.000
05	alc2_o t2	1.800	0.036	49.973	0.000
	alc2_o t3	2.082	0.045	45.974	0.000
	alc2_o t4	2.423	0.063	38.450	0.000
	alc3_o t1	1.305	0.026	49.372	0.000
	alc3_o t2	1.794	0.036	50.027	0.000
10	alc3_o t3	2.082	0.045	45.974	0.000
	alc3_o t4	2.361	0.059	39.948	0.000
	alc4_o t1	0.967	0.023	42.446	0.000
	alc4_o t2	1.575	0.031	51.048	0.000
	alc4_o t3	1.926	0.040	48.511	0.000
15	alc4_o t4	2.197	0.050	43.682	0.000

Amelia imputation results contradict ...

alc5_o t1	1.062	0.024	44.929	0.000
alc5_o t2	1.717	0.034	50.608	0.000
alc5_o t3	2.038	0.044	46.759	0.000
alc5_o t4	2.300	0.056	41.408	0.000

Variances:

	Estimate	Std.Err	z-value	P(> z)
gotBully	1.000			
.alcohol	1.000			
.gotBu1_o	0.515			
.gotBu2_o	0.487			
.gotBu3_o	0.464			
.gotBu4_o	0.482			
.gotBu5_o	0.396			
.gotBu6_o	0.357			
.gotBu7_o	0.492			
.gotBu8_o	0.392			
.gotBu9_o	0.383			
.alc1_o	0.277			
.alc2_o	0.391			
.alc3_o	0.133			
.alc4_o	0.252			
.alc5_o	0.185			

Scales y*:

	Estimate	Std.Err	z-value	P(> z)
--	----------	---------	---------	---------

Amelia imputation results contradict ...

```
gotBu1_o      1.000
gotBu2_o      1.000
gotBu3_o      1.000
45 gotBu4_o      1.000
gotBu5_o      1.000
gotBu6_o      1.000
gotBu7_o      1.000
gotBu8_o      1.000
50 gotBu9_o      1.000
alc1_o        1.000
alc2_o        1.000
alc3_o        1.000
alc4_o        1.000
55 alc5_o        1.000
```

Changes I would make

- The alcohol variables have empty or very small cells. All of the lavaan runs for ordinal models warn us about that. I'd probably combine levels 3, 4, 5
- Raise number of imputations to 20
- Include more auxiliary variables in the imputation model
- Fix semTable to
 - show the number of cases used in the results

References

- Iannotti, R. J. (2005-200). *Health behavior in school-aged children (hbsc), 2005-2006 [Data file and code book]*. Retrieved from <https://www.icpsr.umich.edu/icpsrweb/ICPSR/studies/28241?q=HBSC>
- Muthén, L. K., & Muthén, B. O. (2017). *Mplus User's Guide* (8th ed.). Los Angeles, CA: Muthén & Muthén.
- Rosseel, Y. (2012). lavaan: An r package for structural equation modeling. *Journal of Statistical Software*, 48(2), 1-36. Retrieved from <http://www.jstatsoft.org/v48/i02/>

Session

sessionInfo()

```

R version 3.6.0 (2019-04-26)
Platform: x86_64-pc-linux-gnu (64-bit)
Running under: Ubuntu 19.04

Matrix products: default
BLAS: /usr/lib/x86_64-linux-gnu/atlas/libblas.so.3.10.3
LAPACK: /usr/lib/x86_64-linux-gnu/atlas/liblapack.so.3.10.3

locale:
 [1] LC_CTYPE=en_US.UTF-8      LC_NUMERIC=C               LC_TIME=en_US.UTF-8
 [4] LC_COLLATE=en_US.UTF-8   LC_MONETARY=en_US.UTF-8  LC_MESSAGES=en_US.UTF-8
 [7] LC_PAPER=en_US.UTF-8     LC_NAME=C                 LC_ADDRESS=C
[10] LC_TELEPHONE=C          LC_MEASUREMENT=en_US.UTF-8 LC_IDENTIFICATION=C

attached base packages:
[1] stats      graphics  grDevices  utils      datasets  methods    base

other attached packages:
[1] mitools_2.4      mice_3.5.0      lattice_0.20-38 semTable_1.6    lavaan_0.6-3

loaded via a namespace (and not attached):
 [1] stationery_0.98.6 tidyselect_0.2.5 xfun_0.7          purrr_0.3.2      splines_3.6.0
 [6] generics_0.0.2   htmltools_0.3.6  stats4_3.6.0     pan_1.6           survival_2.44-1.1
[11] rlang_0.3.4      jomo_2.6-8       pillar_1.4.0     nloptr_1.2.1     foreign_0.8-71
[16] glue_1.3.1       DBI_1.0.0        plyr_1.8.4       zip_2.0.2         kutils_1.69
[21] evaluate_0.13    knitr_1.22       parallel_3.6.0   broom_0.5.2      Rcpp_1.0.1
[26] xtable_1.8-4     backports_1.1.4  lme4_1.1-21     mnormt_1.5-5     digest_0.6.18
[31] openxlsx_4.1.0  dplyr_0.8.1     grid_3.6.0       tools_3.6.0      magrittr_1.5
[36] tibble_2.1.1     crayon_1.3.4     tidyr_0.8.3      pbivnorm_0.6.0   pkgconfig_2.0.2
[41] MASS_7.3-51.4    Matrix_1.2-17    assertthat_0.2.1 minqa_1.2.4      rmarkdown_1.12

```

Session ...

```
[46] mitml_0.3-7      R6_2.4.0          boot_1.3-22       rpart_4.1-15      nnet_7.3-12
[51] nlme_3.1-140     compiler_3.6.0
```