Example 15 — Higher-order CFA	
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Description Remarks and examples Reference Also see

# Description

sem can be used to estimate higher-order confirmatory factor analysis models.

```
. use https://www.stata-press.com/data/r19/sem_hcfa1 (Higher-order CFA)
```

. ssd describe

Summary statistics d	ata from	
https://www.stata-pr	ess.com/data/1	19/sem_hcfa1.dta
Observations:	251	Higher-order CFA
Variables:	16	25 May 2024 11:26
		(_dta has notes)

Variable name	Variable label
phyab1	Physical ability 1
phyab2	Physical ability 2
phyab3	Physical ability 3
phyab4	Physical ability 4
appear1	Appearance 1
appear2	Appearance 2
appear3	Appearance 3
appear4	Appearance 4
peerrel1	Relationship w/ peers 1
peerrel2	Relationship w/ peers 2
peerrel3	Relationship w/ peers 3
peerrel4	Relationship w/ peers 4
parrel1	Relationship w/ parent 1
parrel2	Relationship w/ parent 2
parrel3	Relationship w/ parent 3
parrel4	Relationship w/ parent 4

. notes

\_dta:

- Summary statistics data from Marsh, H. W. and Hocevar, D., 1985, "Application of confirmatory factor analysis to the study of self-concept: First- and higher order factor models and their invariance across groups", \_Psychological Bulletin\_, 97: 562-582.
- Summary statistics based on 251 students from Sydney, Australia in Grade 5.
- 3. Data collected using the Self-Description Questionnaire and includes sixteen subscales designed to measure nonacademic traits: four intended to measure physical ability, four intended to measure physical appearance, four intended to measure relations with peers, and four intended to measure relations with parents.

See Higher-order CFA models in [SEM] Intro 5 for background.

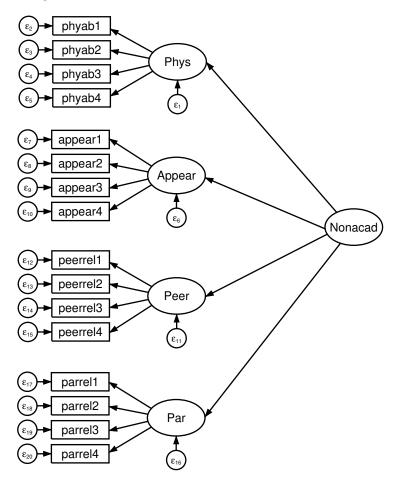
# **Remarks and examples**

Remarks are presented under the following headings:

Fitting the model Fitting the model with the Builder

### Fitting the model

We fit the following model:



```
. sem (Phys -> phyab1 phyab2 phyab3 phyab4)
>
      (Appear -> appear1 appear2 appear3 appear4)
>
      (Peer -> peerrel1 peerrel2 peerrel3 peerrel4)
>
      (Par -> parrel1 parrel2 parrel3 parrel4)
>
      (Nonacad -> Phys Appear Peer Par)
Endogenous variables
  Measurement: phyab1 phyab2 phyab3 phyab4 appear1 appear2 appear3 appear4
               peerrel1 peerrel2 peerrel3 peerrel4 parrel1 parrel2 parrel3
               parrel4
  Latent:
               Phys Appear Peer Par
Exogenous variables
  Latent: Nonacad
Fitting target model:
Iteration 0: Log likelihood = -7686.6699
                                          (not concave)
Iteration 1: Log likelihood = -7643.7387
                                           (not concave)
Iteration 2: Log likelihood = -7616.2966 (not concave)
Iteration 3: Log likelihood = -7597.6133
Iteration 4: Log likelihood = -7588.9515
Iteration 5: Log likelihood = -7585.3162
Iteration 6: Log likelihood = -7584.8125
Iteration 7: Log likelihood = -7584.7885
Iteration 8: Log likelihood = -7584.7881
Structural equation model
                                                           Number of obs = 251
Estimation method: ml
Log likelihood = -7584.7881
 (1)
       [phyab1]Phys = 1
 (2)
       [appear1] Appear = 1
 (3)
       [peerrel1]Peer = 1
 (4)
       [parrel1]Par = 1
 (5)
      [Phys]Nonacad = 1
```

		Coefficient	OIM std. err.	Z	P> z	[95% conf.	interval]
St	tructural Phys Nonacad	1	(constraine	ed)			
	Appear Nonacad	2.202491	.3975476	5.54	0.000	1.423312	2.98167
	Peer Nonacad	1.448035	.2921383	4.96	0.000	.8754549	2.020616
	Par Nonacad	.569956	.1382741	4.12	0.000	.2989437	.8409683
Me	easurement phyab1 Phys	1	(constraine	ed)			
	_cons	8.2	.1159065	70.75	0.000	7.972827	8.427173
	phyab2 Phys _cons	.9332477 8.23	.1285726	7.26 67.34	0.000	.68125 7.990479	1.185245 8.469521

phyab3						
Phys	1.529936	.1573845	9.72	0.000	1.221468	1.83840
_cons	8.17	.1303953	62.66	0.000	7.91443	8.4255
phyab4 Phys	1.325641	.1338053	9.91	0.000	1.063387	1.58789
cons	8.56	.1146471	74.66	0.000	8.335296	8.78470
appear1		(	1)			
Appear	1 7.41	(constraine .1474041	50.27	0.000	7.121093	7.69890
cons	/.41	.1474041	50.27	0.000	7.121093	1.09090
appear2						
Appear	1.0719	.0821893	13.04	0.000	.9108121	1.23298
_cons	7	.1644123	42.58	0.000	6.677758	7.32224
appear3						
Appear	1.035198	.0893075	11.59	0.000	.8601581	1.21023
_cons	7.17	.1562231	45.90	0.000	6.863808	7.47619
appear4						
Appear	.9424492	.0860848	10.95	0.000	.7737262	1.11117
_cons	7.4	.1474041	50.20	0.000	7.111093	7.68890
neerre <sup>11</sup>						
peerrel1 Peer	1	(constraine	ed)			
_cons	8.81	.1077186	81.79	0.000	8.598875	9.02112
peerrel2 Peer	1.214379	.1556051	7.80	0.000	.9093989	1.5193
cons	7.94	.1215769	65.31	0.000	7.701714	8.17828
peerrel3	4 4 4 7 9 9 9	100701	0.74	0 000	1 000044	0 04474
Peer	1.667829 7.52	.190761 .1373248	8.74 54.76	0.000	1.293944	2.04171
_cons	1.52	.1373240	54.70	0.000	7.250848	7.78915
peerrel4						
Peer	1.363627	.159982	8.52	0.000	1.050068	1.67718
cons	8.29	.1222066	67.84	0.000	8.050479	8.52952
parrel1						
- Par	1	(constraine	ed)			
_cons	9.35	.0825215	113.30	0.000	9.188261	9.51173
parrel2						
Par	1.159754	.184581	6.28	0.000	.7979822	1.52152
_cons	9.13	.0988998	92.32	0.000	8.93616	9.3238
	<u> </u>					
parrel3 Par	2.035143	.2623826	7.76	0.000	1.520882	2.54940
_cons	8.67	.1114983	77.76	0.000	8.451467	8.88853
parrel4	1 651000	0116151	7 01	0 000	1 027044	2 0665
Par	1.651802 9	.2116151 .0926003	7.81	0.000 0.000	1.237044	2.0665
_cons	9	.0920003	97.19	0.000	8.818507	9.18149
ar(e.phyab1)	2.07466	.2075636			1.705244	2.52410
ar(e.phyab2)	2.618638	.252693			2.167386	3.16384
ar(e.phyab3) ar(e.phyab4)	1.231013 1.019261	.2062531 .1600644			.8864333 .7492262	1.7095 1.38662

var(e.appe~1)	1.986955	.2711164		1.520699	2.596169
var(e.appe~2)	2.801673	.3526427		2.189162	3.585561
var(e.appe~3)	2.41072	.300262		1.888545	3.077276
var(e.appe~4)	2.374508	.2872554		1.873267	3.009868
var(e.peer~1)	1.866632	.18965		1.529595	2.277933
var(e.peer~2)	2.167766	.2288099		1.762654	2.665984
var(e.peer~3)	1.824346	.2516762		1.392131	2.390749
var(e.peer~4)	1.803918	.212599		1.431856	2.272659
var(e.parr~1)	1.214141	.1195921		1.000982	1.472692
var(e.parr~2)	1.789125	.1748043		1.477322	2.166738
var(e.parr~3)	1.069717	.1767086		.7738511	1.478702
var(e.parr~4)	.8013735	.121231		.5957527	1.077963
var(e.Phys)	.911538	.1933432		.6014913	1.381403
var(e.Appear)	1.59518	.3704939		1.011838	2.514828
var(e.Peer)	.2368108	.1193956		.0881539	.6361528
var(e.Par)	.3697854	.0915049		.2276755	.600597
var(Nonacad)	.3858166	.1237638		.2057449	.7234903
LR test of mod	lel vs. satura	ated: chi2(100	) = 219.48	Prob > chi	2 = 0.0000

Notes:

- 1. The idea behind this model is that physical ability, appearance, and relationships with peers and parents may be determined by a latent variable containing nonacademic traits. This model was suggested by Bollen (1989, 315).
- 2. sem automatically provided normalization constraints for the first-order factors Phys, Appear, Peer, and Par. Their path coefficients were set to 1.
- 3. sem automatically provided a normalization constraint for the second-order factor Nonacad. Its path coefficient was set to 1.

#### Fitting the model with the Builder

Use the diagram above for reference.

1. Open the dataset.

In the Command window, type

. use https://www.stata-press.com/data/r19/sem\_hcfa1

2. Open a new Builder diagram.

#### Select menu item Statistics > SEM (structural equation modeling) > Model building and estimation.

3. Enlarge the size of the canvas to accommodate the length of the diagram.

Click on the Adjust canvas size button,  $\Phi$ , in the Standard Toolbar, change the second size to 7 (inches), and then click on **OK**.

- 4. Change the size of the observed variables' rectangles.
  - a. In the SEM Builder menu, select Settings > Variables > All observed....
  - b. In the resulting dialog box, change the second size to . 25 and click on OK.

5. Create the measurement component for physical ability.

Select the Add measurement component tool,  $\mathcal{W}$ . Then using the darker one-inch grid lines in the background as a guide, click in the diagram about two inches in from the left and one inch down from the top.

In the resulting dialog box,

- a. change the Latent variable name to Phys;
- b. select phyab1, phyab2, phyab3, and phyab4 by using the Measurement variables control;
- c. select Left in the Measurement direction control;
- d. click on OK.

If you wish, move the component by clicking on any variable and dragging it.

- 6. Create the remaining first-order measurement components.
  - a. Repeat the process from item 5, but place the measurement component on the grid line two inches in from the left and about two and one-half inches down from the top. Label the latent variable Appear, and select measurement variables appear1, appear2, appear3, and appear4.
  - b. Repeat the process from item 5, but place the measurement component on the grid line two inches in from the left and about four inches down from the top. Label the latent variable Peer, and select measurement variables peerrel1, peerrel2, peerrel3, and peerrel4.
  - c. Repeat the process from item 5, but place the measurement component on the grid line two inches in from the left and about five and one-half inches down from the top. Label the latent variable Par, and select measurement variables parrel1, parrel2, parrel3, and parrel4.
- 7. Create the second-order latent variable.
  - a. Select the Add latent variable tool,  $^{\bigcirc}$ , and then click in the diagram about two inches in from the right and vertically centered between the Appear and Peer latent variables.
  - b. In the Contextual Toolbar, type Nonacad in the Name control and press Enter.
- 8. Create paths from Nonacad to each of the first-order latent variables.
  - a. Select the Add path tool, -.
  - b. Click in the upper-left quadrant of the Nonacad oval (it will highlight when you hover over it), and drag a path to the lower-left quadrant of the Phys oval (it will highlight when you can release to connect the path).
  - c. Continuing with the  $\rightarrow$  tool, create the following paths by clicking first on the left side of the Nonacad variable and dragging to the right side of the first-order latent variable.

```
Nonacad -> Appear
Nonacad -> Peer
Nonacad -> Par
```

9. Clean up the direction of the errors.

We want the errors for each of the latent variables to be below the latent variable. The errors for Phys, Appear, and Peer are likely to have been created in other directions.

a. Choose the Select tool,

- b. Click in the Phys oval.
- c. Click on one of the Error rotation buttons, 2, in the Contextual Toolbar until the error is below the latent variable.

Repeat this for all errors on latent variables that are not below the latent variable.

10. Clean up the paths.

If you do not like where a path has been connected to its variable, use the Select tool,  $\blacktriangleright$ , to click on the path, and then simply click on where it connects to an oval and drag the endpoint.

11. Estimate.

Click on the **Estimate** button,  $\mathbb{P}$ , in the Standard Toolbar, and then click on **OK** in the resulting SEM estimation options dialog box.

Tip: See the tips of [SEM] **Example 9** to make creating paths somewhat easier than described above.

You can open a completed diagram in the Builder by typing

. webgetsem sem\_hcfa1

### Reference

Bollen, K. A. 1989. Structural Equations with Latent Variables. New York: Wiley. https://doi.org/10.1002/9781118619179.

## Also see

[SEM] sem — Structural equation model estimation command

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