Description Remarks and examples Also see

# **Description**

This example continues where [SEM] **Example 7** left off, where we typed

# Remarks and examples

Remarks are presented under the following headings:

```
Using test to evaluate adding constraints
Refitting the model with added constraints
Using estat scoretests to test whether constraints can be relaxed
```

We want to show you how to evaluate potential constraints after estimation, how to fit a model with constraints, and how to evaluate enforced constraints after estimation.

Obviously, in a real analysis, if you evaluated potential constraints after estimation, there would be no reason to evaluate enforced constraints after estimation, and vice versa.

## Using test to evaluate adding constraints

In this model of respondents and corresponding friends, it would be surprising if the coefficients relating friends' characteristics to respondents' occupational aspirations and vice versa were not equal. It would also be surprising if coefficients relating a respondent's characteristics to his occupational aspirations were not equal to those of his friends' characteristics to his occupational aspirations. The paths that we suspect should be equal are

You are about to learn that to test whether those paths have equal coefficients, you type

In Stata, \_b[] is how one accesses the estimated parameters. It is difficult to remember what the names are. To determine the names of the parameters, replay the sem results with the coeflegend option:

```
. sem, coeflegend
Structural equation model
                                                             Number of obs = 329
Estimation method: ml
Log likelihood = -2617.0489
               Coefficient Legend
Structural
  r_occasp
   f_occasp
                 .2773441
                            _b[r_occasp:f_occasp]
     r_intel
                 .2854766
                           _b[r_occasp:r_intel]
       r_ses
                            _b[r_occasp:r_ses]
                 .1570082
       f_ses
                  .0973327
                            _b[r_occasp:f_ses]
  f_occasp
                           _b[f_occasp:r_occasp]
   r_occasp
                 .2118102
       r_ses
                 .0794194
                           _b[f_occasp:r_ses]
       f_ses
                 .1681772
                           b[f occasp:f ses]
     f intel
                 .3693682
                           _b[f_occasp:f_intel]
var(e.r oc~p)
                 .6868304
                            b[/var(e.r occasp)]
var(e.f_oc~p)
                 .6359151
                            b[/var(e.f occasp)]
cov(e.r oc~p,
```

b[/cov(e.r occasp,e.f occasp)]

LR test of model vs. saturated: chi2(0) = 0.00

-.1536992

Prob > chi2 = .

With the parameter names at hand, to perform the test, we can type

```
. test ( b[r occasp:r intel ] == b[f occasp:f intel ])
                             ]==_b[f_occasp:f_ses
       ( b[r occasp:r ses
                                                    1)
>
       (_b[r_occasp:f_ses
                             ]==_b[f_occasp:r_ses
                                                    1)
       (_b[r_occasp:f_occasp] == _b[f_occasp:r_occasp])
 (1)
       [r_occasp]r_intel - [f_occasp]f_intel = 0
       [r_{occasp}]r_{ses} - [f_{occasp}]f_{ses} = 0
 (3)
       [r_occasp]f_ses - [f_occasp]r_ses = 0
 (4)
       [r_occasp]f_occasp - [f_occasp]r_occasp = 0
           chi2(4) =
                          1.61
         Prob > chi2 =
                          0.8062
```

We cannot reject the constraint, just as we expected.

e.f occasp)

### Refitting the model with added constraints

cov(e.r oc~p,

e.f occasp)

-.1582175

We could refit the model with these constraints by typing

```
. sem (r occasp <- f occasp@b1 r intel@b2 r ses@b3 f ses@b4)</pre>
      (f_occasp <- r_occasp@b1 f_intel@b2 f_ses@b3 r_ses@b4),
                               cov(e.r_occasp*e.f_occasp)
Endogenous variables
  Observed: r_occasp f_occasp
Exogenous variables
  Observed: r_intel r_ses f_ses f_intel
Fitting target model:
Iteration 0: Log likelihood = -2617.8735
Iteration 1: Log likelihood = -2617.8705
Iteration 2: Log likelihood = -2617.8705
Structural equation model
                                                            Number of obs = 329
Estimation method: ml
Log likelihood = -2617.8705
       [r occasp]f occasp - [f occasp]r occasp = 0
       [r_occasp]r_intel - [f_occasp]f_intel = 0
 (2)
       [r_{occasp}]r_{ses} - [f_{occasp}]f_{ses} = 0
 (3)
 (4)
       [r_occasp]f_ses - [f_occasp]r_ses = 0
                               MIO
               Coefficient std. err.
                                                 P>|z|
                                                           [95% conf. interval]
Structural
 r_occasp
   f_occasp
                 .2471578
                            .1024504
                                         2.41
                                                 0.016
                                                           .0463588
                                                                        .4479568
                                         8.02
     r_{intel}
                 .3271847
                             .0407973
                                                 0.000
                                                           .2472234
                                                                        .4071459
                 .1635056
                            .0380582
                                         4.30
                                                 0.000
                                                           .0889129
                                                                        .2380984
      r_ses
       f ses
                  .088364
                            .0427106
                                         2.07
                                                 0.039
                                                           .0046529
                                                                        .1720752
  f_occasp
   r_occasp
                 .2471578
                             .1024504
                                          2.41
                                                 0.016
                                                           .0463588
                                                                        .4479568
      r_ses
                  .088364
                             .0427106
                                          2.07
                                                 0.039
                                                           .0046529
                                                                        .1720752
                                                                        .2380984
      f_ses
                 .1635056
                             .0380582
                                         4.30
                                                 0.000
                                                           .0889129
     f_{intel}
                 .3271847
                             .0407973
                                          8.02
                                                 0.000
                                                           .2472234
                                                                        .4071459
var(e.r oc~p)
                 .6884513
                             .0538641
                                                           .5905757
                                                                        .8025477
var(e.f oc~p)
                 .6364713
                            .0496867
                                                           .5461715
                                                                        .7417005
```

#### LR test of model vs. saturated: chi2(4) = 1.64Prob > chi2 = 0.8010

-1.12 0.262

-.4345942

.1181592

```
Using estat scoretests to test whether constraints can be relaxed
```

.1410111

```
. estat scoretests
(no score tests to report; all chi2 values less than 3.841458820694123)
```

. estat scoretests, min(0)

Score tests for linear constraints

- (1)  $[r\_occasp]f\_occasp [f\_occasp]r\_occasp = 0$
- (2) [r\_occasp]r\_intel [f\_occasp]f\_intel = 0
- ( 3) [r\_occasp]r\_ses [f\_occasp]f\_ses = 0
  ( 4) [r\_occasp]f\_ses [f\_occasp]r\_ses = 0

	chi2	df	P>chi2
(1)	0.014	1	0.91
(2)	1.225	1	0.27
(3)	0.055	1	0.81
(4)	0.136	1	0.71

### Notes:

- 1. When we began this example, we used test to evaluate potential constraints that we were considering. We obtained an overall  $\chi^2(4)$  statistic of 1.61 and thus could not reject the constraints at any reasonable level.
- 2. We then refit the model with those constraints.
- 3. For pedantic reasons, now we use estat scoretests to evaluate relaxing constraints included in the model. estat scoretests does not report a joint test. You cannot sum the  $\chi^2$  values to obtain a joint test statistic. Thus we learn only that the individual constraints should not be relaxed at reasonable confidence levels.
- 4. Thus when evaluating multiple constraints, it is better to fit the model without the constraints and use test to evaluate them jointly.

## Also see

[SEM] Example 7 — Nonrecursive structural model

[SEM] sem — Structural equation model estimation command

[SEM] sem and gsem path notation — Command syntax for path diagrams

[SEM] estat scoretests — Score tests

[SEM] test — Wald test of linear hypotheses

Stata, Stata Press, and Mata are registered trademarks of StataCorp LLC. Stata and Stata Press are registered trademarks with the World Intellectual Property Organization of the United Nations. StataNow and NetCourseNow are trademarks of StataCorp LLC. Other brand and product names are registered trademarks or trademarks of their respective companies. Copyright © 1985–2025 StataCorp LLC, College Station, TX, USA. All rights reserved



For suggested citations, see the FAQ on citing Stata documentation.