

CRMDA semexamples in Mplus

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Outline

- 1 Example 1: Job Data EFA
- 2 Example 2: CFA
- 3 Example 3: Two Group CFA
- 4 Example 4: Multiple Regression
- 5 Example 5: SEM
- 6 Example: Insomnia SEM
- 7 Example: Growth Curve
- 8 Conclusion

The Repository

- CRMDA created an archive of SEM working examples.
- To the extent possible, the same models are estimated with
 - Mplus
 - R (R Core Team, 2017)
 - Stata
- The repository can be browsed/downloaded:
<https://gitlab.crmda.ku.edu/crmda/semexample>
- The data folder includes the information that is imported into each of the 3 programs.

Runnable examples

- 1 Exploratory Factor Analysis (EFA)
- 2 Confirmatory Factor Analysis (CFA)
- 3 Measurement Invariance (Multi-group CFA)
- 4 Multiple Regression
- 5 Structural Equation Model (SEM)
- 6 Latent Growth Curve (LGC)
- 7 Modeling strategy changes for Ordinal Data

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- 1 Example 1: Job Data EFA
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The job_placement data set

- We obtained data from job seekers (job_placement.csv)
- Six test score variables that we think could measure latent variables
- Work to do
 - 1 Identify a factor structure
 - 2 Determine how our latent variables relate to demographic variables

Excelsior Spring Job Corps center

Items collected from job seekers using the Excelsior Spring Job Corps center

- Variables

id: Subject identification number

wjcalc: Subject's score on the WJ calculation subtest. (Numeric)

wjspl: Subject's score on the WJ spelling subtest. (Numeric)

wratcalc: Subject's score on the WRAT calculation subtest. (Numeric)

wratspl: Subject's score on the WRAT spelling subtest. (Numeric)

waiscalc: Subject's score on the WAIS arithmetic calculations subtest. (Numeric)

waisspl: Subject's score on the WAIS spelling subtest. (Numeric)

edlevel: What is the highest level of education completed by the subject? (Ordinal)

newschl: Did the subject ever change high schools? (Binary: 1=Yes, 0=No)

Excelsior Spring Job Corps center ...

- suspend:** Has the subject ever been suspended from high school?
(Binary: 1=Yes, 0=No)
 - expelled:** Has the subject ever been expelled from high school? (Binary: 1=Yes, 0=No)
 - haveld:** Has the subject been diagnosed with a learning disorder?
(Binary: 1=Yes, 0=No)
 - female:** Gender (Binary: 1=Female, 0=Male)
 - age:** Age in years (Numeric)
- Missing data are coded as “99999”

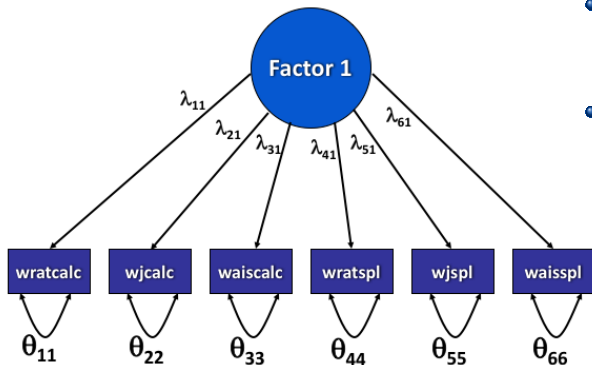
EFA: Questions

- How many latent variables underlie subtest scores (6 items)?
 - Will decide by running various models and comparing the results.
- Mplus can quickly run many EFAs and report the results.
 - Unlike SPSS it does not default to principle components analysis.

The main idea of EFA

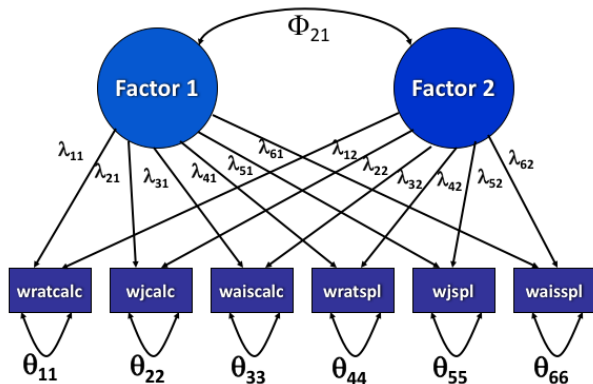
- EFA lets us specify how many factors will be estimated.
- Then it will estimate a loading for *each and every* observed variable on *each* factor.
- Since the 1970s, EFA has been regarded with suspicion by many because it seems to be “atheoretic data-grubbing”.
- Nevertheless, we persist in showing you how to do it.

EFA: One Factor



- λ_{11} is the loading of wrtcalc on Factor 1
- θ_{11} is variance of the residual (unexplained part) of wrtcalc
- if θ_{11} were 0, it would mean that wrtcalc is a perfectly accurate reflection of Factor 1.

EFA: Two Factors



- Φ_{21} is covariance between Factor 1 and 2.
- If $\Phi_{21} = 0$, then the factors are uncorrelated (“orthogonal”)

EFA in Mplus

```
TITLE:
Example 1 - Exploratory Factor Analysis
DATA:
FILE IS "../..//data/job_placement.csv";
VARIABLE:
5  NAMES ARE
id wjcalc wjspl wratspl wratcalc waiscalc waisspl
edlevel newschl suspend expelled haveld female age;
USEVARIABLES ARE
10 wjcalc wjspl wratspl wratcalc waiscalc waisspl;
MISSING ARE all (99999);
ANALYSIS: TYPE = EFA 1 2;
!Default estimator is ML;
!Default rotation is Oblique Geomin;
```

EFA: Beyond Defaults

```

ANALYSIS: TYPE = EFA 1 2;
!Default estimator is ML;
!Default rotation is
    Oblique Geomin;
ESTIMATOR = ML;
MLM;
MLMV
MLR;
MLF;
MUML;
WLS;
WLSM;
WLSMV;
ULS;
ULSMV;

```

5

10

```

ROTATION = GEOMIN;
GEOMIN (OBLIQUE);
GEOMIN (ORTHOGONAL);
QUARTIMIN; OBLIQUE
CF-VARIMAX
CF-QUARTIMAX;
CF-EQUAMAX;
CF-PARSIMAX;
CF-FACPARSIM;
CRAWFER; OBLIQUE 1/p
CRAWFER (OBLIQUE or
    ORTHOGONAL value);
OBLIMIN; OBLIQUE 0
OBLIMIN (OBLIQUE or
    ORTHOGONAL value);
VARIMAX;

```

EFA Results

There are several ways you can access the results.

- 1 In the semexample/Mplus folders
- 2 View online: [efa-01.out](#)
- 3 We will import the verbatim output
Inline Display:

```
Mplus VERSION 7.4 (Linux)
MUTHEN & MUTHEN
05/28/2018 9:31 PM
```

```
INPUT INSTRUCTIONS
```

```
TITLE:
Example 1 - Exploratory Factor Analysis
```

```
DATA:
FILE IS "../..data/job_placement.csv";
```

```
VARIABLE:
NAMES ARE
id wjcalc wjspl wratspl wratcalc waiscalc waisspl
```

EFA Results ...

```
edlevel newschl suspend expelled haveld female age;
```

```
USEVARIABLES ARE
```

```
wjcalc wjspl wratspl wratcalc waiscalc waisspl;
```

```
MISSING ARE all(99999);
```

```
ANALYSIS:
```

```
TYPE = EFA 1 2;
```

```
INPUT READING TERMINATED NORMALLY
```

Example 1 - Exploratory Factor Analysis

SUMMARY OF ANALYSIS

Number of groups	1
Number of observations	322
Number of dependent variables	6
Number of independent variables	0

EFA Results ...

```

Number of continuous latent variables                0

Observed dependent variables

Continuous
WJCALC      WJSPL      WRATSPL      WRATCALC      WAISCALC      WAISSPL

Estimator                ML
Rotation                 GEOMIN
Row standardization     CORRELATION
Type of rotation        OBLIQUE
Epsilon value           Varies
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
Optimization Specifications for the Exploratory Factor Analysis

Rotation Algorithm
Number of random starts                30
Maximum number of iterations           10000
Derivative convergence criterion       0.100D-04

Input data file(s)

```

EFA Results ...

```
../..../data/job_placement.csv
```

```
Input data format  FREE
```

SUMMARY OF DATA

```
Number of missing data patterns          4
```

COVARIANCE COVERAGE OF DATA

```
Minimum covariance coverage value  0.100
```

PROPORTION OF DATA PRESENT

	Covariance Coverage				
	WJCALC	WJSPL	WRATSPL	WRATCALC	WAISCALC
	-----	-----	-----	-----	-----
WJCALC	0.994				
WJSPL	0.994	1.000			
WRATSPL	0.991	0.997	0.997		
WRATCALC	0.994	1.000	0.997	1.000	
WAISCALC	0.988	0.994	0.991	0.994	0.994

EFA Results ...

WAISSPL	0.988	0.994	0.991	0.994	0.994
---------	-------	-------	-------	-------	-------

Covariance Coverage

WAISSPL

WAISSPL	-----
	0.994

SUMMARY OF MODEL FIT INFORMATION

Model	Number of Parameters	Chi-Square	Degrees of Freedom	P-Value
1-factor	18	476.495	9	0.0000
2-factor	23	3.690	4	0.4496

Models Compared	Chi-Square	Degrees of Freedom	P-Value
1-factor against 2-factor	472.805	5	0.0000

EFA Results ...

RESULTS FOR EXPLORATORY FACTOR ANALYSIS

EIGENVALUES FOR SAMPLE CORRELATION MATRIX

	1	2	3	4	5
1	----- 4.041	----- 1.219	----- 0.397	----- 0.142	----- 0.107

EIGENVALUES FOR SAMPLE CORRELATION MATRIX

	6
1	----- 0.095

EXPLORATORY FACTOR ANALYSIS WITH 1 FACTOR(S):

MODEL FIT INFORMATION

Number of Free Parameters 18

Loglikelihood

H0 Value -5361.307
H1 Value -5123.060

EFA Results ...

Information Criteria

Akaike (AIC)	10758.615
Bayesian (BIC)	10826.557
Sample-Size Adjusted BIC	10769.463
(n* = (n + 2) / 24)	

Chi-Square Test of Model Fit

Value	476.495
Degrees of Freedom	9
P-Value	0.0000

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.402	
90 Percent C.I.	0.371	0.433
Probability RMSEA <= .05	0.000	

CFI/TLI

CFI	0.750
TLI	0.583

Chi-Square Test of Model Fit for the Baseline Model

EFA Results ...

Value	1882.335
Degrees of Freedom	15
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value	0.156
-------	-------

MINIMUM ROTATION FUNCTION VALUE	3.49664
---------------------------------	---------

GEOMIN ROTATED LOADINGS (* significant at 5% level)

1

WJCALC	0.516*
WJSPL	0.953*
WRATSPL	0.944*
WRATCALC	0.545*
WAISCALC	0.500*
WAISSPL	0.940*

EFA Results ...

GEOMIN FACTOR CORRELATIONS (* significant at 5% level)

1

1	-----
	1.000

ESTIMATED RESIDUAL VARIANCES

WJCALC

WJSPL

WRATSPL

WRATCALC

WAISCALC

1	-----	-----	-----	-----	-----
	0.734	0.092	0.108	0.703	0.750

ESTIMATED RESIDUAL VARIANCES

WAISSPL

1	-----
	0.117

S.E. GEOMIN ROTATED LOADINGS

1

WJCALC	-----
	0.042
WJSPL	0.007
WRATSPL	0.008
WRATCALC	0.040
WAISCALC	0.043

EFA Results ...

WAISSPL 0.008

S.E. GEOMIN FACTOR CORRELATIONS

1

1 -----
 0.000

S.E. ESTIMATED RESIDUAL VARIANCES

	WJCALC	WJSPL	WRATSPL	WRATCALC	WAISCALC
1	----- 0.043	----- 0.014	----- 0.015	----- 0.044	----- 0.043

S.E. ESTIMATED RESIDUAL VARIANCES

WAISSPL

1 -----
 0.015

Est./S.E. GEOMIN ROTATED LOADINGS

1

WJCALC	----- 12.251
WJSPL	133.177

EFA Results ...

WRATSPL	120.556
WRATCALC	13.487
WAISCALC	11.620
WAISSPL	114.214

Est./S.E. GEOMIN FACTOR CORRELATIONS

	1
1	----- 0.000

Est./S.E. ESTIMATED RESIDUAL VARIANCES

	WJCALC	WJSPL	WRATSPL	WRATCALC	WAISCALC
1	----- 16.879	----- 6.759	----- 7.298	----- 15.974	----- 17.427

Est./S.E. ESTIMATED RESIDUAL VARIANCES

	WAISSPL
1	----- 7.564

EXPLORATORY FACTOR ANALYSIS WITH 2 FACTOR(S):

EFA Results ...

MODEL FIT INFORMATION

Number of Free Parameters	23
---------------------------	----

Loglikelihood

H0 Value	-5124.905
H1 Value	-5123.060

Information Criteria

Akaike (AIC)	10295.810
Bayesian (BIC)	10382.625
Sample-Size Adjusted BIC	10309.672
$(n* = (n + 2) / 24)$	

Chi-Square Test of Model Fit

Value	3.690
Degrees of Freedom	4
P-Value	0.4496

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.000
----------	-------

EFA Results ...

90 Percent C.I.	0.000	0.081
Probability RMSEA <= .05	0.763	

CFI/TLI

CFI	1.000
TLI	1.001

Chi-Square Test of Model Fit for the Baseline Model

Value	1882.335
Degrees of Freedom	15
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value	0.003
-------	-------

MINIMUM ROTATION FUNCTION VALUE	0.12976
---------------------------------	---------

GEOMIN ROTATED LOADINGS (* significant at 5% level)

1	2
---	---

EFA Results ...

WJCALC	----- 0.908*	----- -0.005
WJSPL	-0.001	0.955*
WRATSPL	0.008	0.941*
WRATCALC	0.950*	0.000
WAISCALC	0.653*	0.118*
WAISSPL	-0.006	0.945*

GEMIN FACTOR CORRELATIONS (* significant at 5% level)

	1	2
1	----- 1.000	----- -----
2	0.544*	1.000

ESTIMATED RESIDUAL VARIANCES

	WJCALC	WJSPL	WRATSPL	WRATCALC	WAISCALC
1	----- 0.182	----- 0.089	----- 0.107	----- 0.097	----- 0.476

ESTIMATED RESIDUAL VARIANCES

	WAISSPL
1	----- 0.113

EFA Results ...

S.E. GEOMIN ROTATED LOADINGS

	1	2
WJCALC	0.029	0.027
WJSPL	0.014	0.011
WRATSPL	0.022	0.015
WRATCALC	0.024	0.017
WAISCALC	0.044	0.050
WAISSPL	0.020	0.014

S.E. GEOMIN FACTOR CORRELATIONS

	1	2
1	0.000	
2	0.043	0.000

S.E. ESTIMATED RESIDUAL VARIANCES

	WJCALC	WJSPL	WRATSPL	WRATCALC	WAISCALC
1	0.034	0.014	0.015	0.033	0.041

EFA Results ...

S.E. ESTIMATED RESIDUAL VARIANCES
WAISSPL

1	0.015

Est./S.E. GEOMIN ROTATED LOADINGS
1 2

	-----	-----
WJCALC	31.690	-0.200
WJSPL	-0.052	83.613
WRATSPL	0.367	60.996
WRATCALC	40.076	0.013
WAISCALC	15.009	2.348
WAISSPL	-0.287	65.440

Est./S.E. GEOMIN FACTOR CORRELATIONS
1 2

	-----	-----
1	0.000	
2	12.796	0.000

Est./S.E. ESTIMATED RESIDUAL VARIANCES
WJCALC WJSPL WRATSPL

WRATCALC

WAISCALC

EFA Results ...

1	----- 5.298	----- 6.407	----- 7.148	----- 2.916	----- 11.497
---	----------------	----------------	----------------	----------------	-----------------

Est./S.E. ESTIMATED RESIDUAL VARIANCES
 WAISSPL

1	----- 7.308
---	----------------

FACTOR STRUCTURE

	1	2
WJCALC	----- 0.905	----- 0.488
WJSPL	0.519	0.955
WRATSPL	0.520	0.945
WRATCALC	0.950	0.517
WAISCALC	0.717	0.473
WAISSPL	0.508	0.942

Beginning Time: 21:31:59

Ending Time: 21:31:59

Elapsed Time: 00:00:00

EFA Results ...

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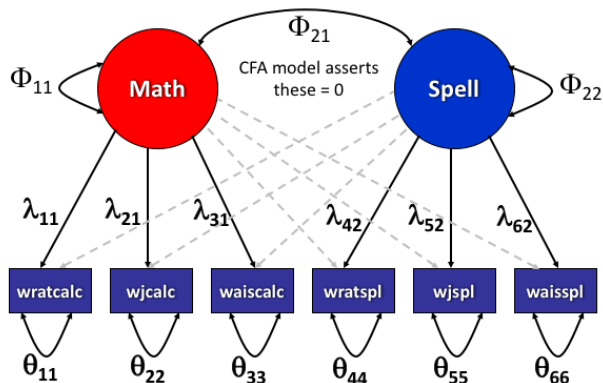
Outline

- 1 Example 1: Job Data EFA
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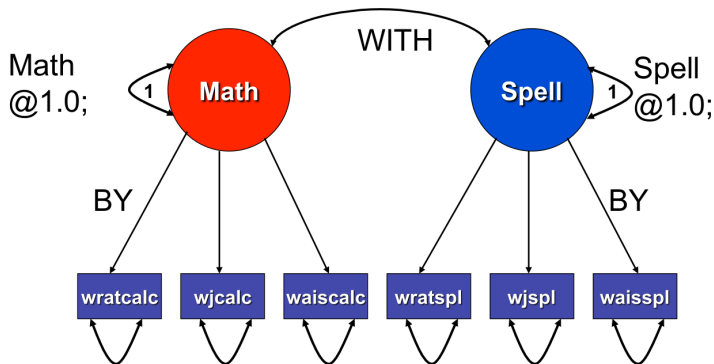
CFA = Confirmatory Factor Analysis

- Some factor loadings are set at 0, on theoretical grounds
 - Put another way, FA where only certain parameters are estimated.
- We believe (on theoretical grounds) that Math is indicated by:
 - wratcalc, wjcalc, waiscalc
- We believe that Spelling is indicated by:
 - wratspl, wjspl, waisspl

CFA: Grayed arrows are fixed at 0



CFA: With Mplus Terminology



About the identification

- All models with latent variables run into the “identification problem”. They have some coefficients that cannot be estimated from the data
 - The usual approach is to fix some coefficients
 - Because some are fixed, the other coefficients “rescale themselves”, absorb the assumed scaling.

Method 1: marker variable approach

- Mplus defaults to use the “**marker variable**” identification.
 - Fix the first indicator for each latent variable to have $\lambda = 1$.
 - The variance of each latent variable is estimated.
 - The latent variable’s scale is thus the scale of the first indicator variable.
 - The loading estimates for other variables are relative to 1.0.
 - A variable that is more sensitive to the latent factor will have a coefficient > 1.0 .
 - This method is preferred by adherents of the LISREL method of factor analysis.

Method 2: standardized latent variable approach

- An alternative identification approach, which is also widely used, is to
 - Fix the variances of each of the latent variables to 1.0 (but usually allow covariance to be estimated)
 - Estimate the λ for each of the indicators.
 - Now the λ estimates are understood as proportions of the factor which are translated into the indicators

CFA Mplus Syntax

```
TITLE:  
  Example 2 - Confirmatory Factor Analysis  
DATA:  
  FILE IS "..../data/job_placement.csv";  
VARIABLE:  
  NAMES ARE  
  id wjcalc wjspl wratspl wratcalc waiscalc waisspl  
  edlevel newschl suspend expelled haveld female  
  age;  
  USEVARIABLES ARE  
  wratspl wjspl waisspl wratcalc wjcalc waiscalc;  
MISSING ARE all(99999);  
MODEL:  
  MATH BY wratcalc* wjcalc waiscalc;  
  SPELL BY wratspl* wjspl waisspl;  
  MATH@1 SPELL@1;!Set variance at 1;
```


CFA Mplus Syntax ...

```
OUTPUT :
```

```
TECH1 ;
```

```
STDYX ;
```

- To alter identification from Mplus defaults, we insert “*” with `wratcalc` and `wratspl`

CFA Results

View online: [cfa-01.out](#)

Inline Display:

```
Mplus VERSION 7.4 (Linux)
MUTHEN & MUTHEN
05/28/2018  9:35 PM
```

INPUT INSTRUCTIONS

TITLE:

```
Example 2 - Confirmatory Factor Analysis
```

DATA:

```
FILE IS "../..data/job_placement.csv";
```

VARIABLE:

NAMES ARE

```
id wjcalc wjspl wratspl wratcalc waiscalc waisspl
edlevel newschl suspend expelled haveld female age;
```

USEVARIABLES ARE

```
wratspl wjspl waisspl wratcalc wjcalc waiscalc;
```

CFA Results ...

MISSING ARE all(99999);

MODEL:

MATH BY wratcalc* wjcalc waiscalc;

SPELL BY wratspl* wjspl waisspl;

MATH@1 SPELL@1;

OUTPUT:

TECH1;

STDYX;

INPUT READING TERMINATED NORMALLY

Example 2 - Confirmatory Factor Analysis

SUMMARY OF ANALYSIS

Number of groups	1
Number of observations	322

CFA Results ...

```

Number of dependent variables           6
Number of independent variables         0
Number of continuous latent variables   2

Observed dependent variables

Continuous
  WRATSPL    WJSPL    WAISSPL    WRATCALC    WJCALC    WAISCALC

Continuous latent variables
  MATH    SPELL

Estimator                               ML
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/job_placement.csv

Input data format  FREE

```

CFA Results ...

SUMMARY OF DATA

Number of missing data patterns 4

COVARIANCE COVERAGE OF DATA

Minimum covariance coverage value 0.100

PROPORTION OF DATA PRESENT

	Covariance Coverage				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	0.997				
WJSPL	0.997	1.000			
WAISSPL	0.991	0.994	0.994		
WRATCALC	0.997	1.000	0.994	1.000	
WJCALC	0.991	0.994	0.988	0.994	0.994
WAISCALC	0.991	0.994	0.994	0.994	0.988

Covariance Coverage

CFA Results ...

```

                WAISCALC
                -----
WAISCALC      0.994

```

THE MODEL ESTIMATION TERMINATED NORMALLY

MODEL FIT INFORMATION

Number of Free Parameters 19

Loglikelihood

H0 Value -5127.830

H1 Value -5123.060

Information Criteria

Akaike (AIC) 10293.661

Bayesian (BIC) 10365.377

Sample-Size Adjusted BIC 10305.112

(n* = (n + 2) / 24)

CFA Results ...

Chi-Square Test of Model Fit

Value	9.540
Degrees of Freedom	8
P-Value	0.2988

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.024	
90 Percent C.I.	0.000	0.073
Probability RMSEA \leq .05	0.761	

CFI/TLI

CFI	0.999
TLI	0.998

Chi-Square Test of Model Fit for the Baseline Model

Value	1882.335
Degrees of Freedom	15
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value	0.024
-------	-------

CFA Results ...

MODEL RESULTS

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
MATH BY				
WRATCALC	6.041	0.276	21.921	0.000
WJCALC	4.144	0.203	20.370	0.000
WAISCALC	2.410	0.165	14.636	0.000
SPELL BY				
WRATSPL	6.532	0.288	22.645	0.000
WJSPL	6.809	0.296	23.025	0.000
WAISSPL	6.354	0.283	22.463	0.000
SPELL WITH				
MATH	0.553	0.042	13.191	0.000
Intercepts				
WRATSPL	36.484	0.385	94.751	0.000
WJSPL	41.674	0.398	104.808	0.000
WAISSPL	37.163	0.376	98.788	0.000
WRATCALC	38.922	0.355	109.514	0.000

CFA Results ...

WJCALC	23.812	0.255	93.298	0.000
WAISCALC	11.022	0.186	59.230	0.000
Variances				
MATH	1.000	0.000	999.000	999.000
SPELL	1.000	0.000	999.000	999.000
Residual Variances				
WRATSPL	5.053	0.612	8.256	0.000
WJSPL	4.541	0.616	7.369	0.000
WAISSPL	5.156	0.599	8.605	0.000
WRATCALC	4.179	1.014	4.122	0.000
WJCALC	3.769	0.537	7.017	0.000
WAISCALC	5.304	0.457	11.596	0.000

STANDARDIZED MODEL RESULTS

STDYX Standardization

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
MATH BY WRATCALC	0.947	0.014	69.150	0.000

CFA Results ...

WJCALC	0.906	0.015	58.479	0.000
WAISCALC	0.723	0.029	24.919	0.000
SPELL BY				
WRATSPL	0.946	0.008	120.751	0.000
WJSPL	0.954	0.007	133.161	0.000
WAISSPL	0.942	0.008	115.722	0.000
SPELL WITH				
MATH	0.553	0.042	13.191	0.000
Intercepts				
WRATSPL	5.282	0.215	24.513	0.000
WJSPL	5.841	0.237	24.664	0.000
WAISSPL	5.508	0.224	24.555	0.000
WRATCALC	6.103	0.247	24.722	0.000
WJCALC	5.203	0.213	24.447	0.000
WAISCALC	3.306	0.142	23.269	0.000
Variances				
MATH	1.000	0.000	999.000	999.000
SPELL	1.000	0.000	999.000	999.000
Residual Variances				
WRATSPL	0.106	0.015	7.150	0.000
WJSPL	0.089	0.014	6.521	0.000

CFA Results ...

PARAMETER SPECIFICATION

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
1	1	2	3	4	5

	NU
	WAISCALC

1	6

	LAMBDA	
	MATH	SPELL
	-----	-----
WRATSPL	0	7
WJSPL	0	8
WAISSPL	0	9
WRATCALC	10	0
WJCALC	11	0
WAISCALC	12	0

CFA Results ...

THETA					
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	13				
WJSPL	0	14			
WAISSPL	0	0	15		
WRATCALC	0	0	0	16	
WJCALC	0	0	0	0	17
WAISCALC	0	0	0	0	0

THETA	
	WAISCALC

WAISCALC	18

ALPHA		
	MATH	SPELL
	-----	-----
1	0	0

BETA

CFA Results ...

	MATH	SPELL
MATH	0	0
SPELL	0	0

	PSI MATH	SPELL
MATH	0	
SPELL	19	0

STARTING VALUES

	NU WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
1	36.483	41.674	37.197	38.922	23.822

	NU WAISCALC
1	11.016

CFA Results ...

	LAMBDA	
	MATH	SPELL
	-----	-----
WRATSPL	0.000	1.000
WJSPL	0.000	1.000
WAISSPL	0.000	1.000
WRATCALC	1.000	0.000
WJCALC	1.000	0.000
WAISCALC	1.000	0.000

	THETA				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	23.922				
WJSPL	0.000	25.455			
WAISSPL	0.000	0.000	22.785		
WRATCALC	0.000	0.000	0.000	20.337	
WJCALC	0.000	0.000	0.000	0.000	10.464
WAISCALC	0.000	0.000	0.000	0.000	0.000

THETA
WAISCALC

CFA Results ...

WAISCALC	-----	
	5.545	

	ALPHA	
	MATH	SPELL
1	-----	-----
	0.000	0.000

	BETA	
	MATH	SPELL
MATH	-----	-----
	0.000	0.000
SPELL	0.000	0.000

	PSI	
	MATH	SPELL
MATH	-----	-----
	1.000	
SPELL	0.000	1.000

Beginning Time: 21:35:18

Ending Time: 21:35:18

CFA Results ...

Elapsed Time: 00:00:00

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Outline

- 1 Example 1: Job Data EFA
- 2 Example 2: CFA
- 3 Example 3: Two Group CFA
- 4 Example 4: Multiple Regression
- 5 Example 5: SEM
- 6 Example: Insomnia SEM
- 7 Example: Growth Curve
- 8 Conclusion

Should 2 groups be included in same model?

- Goal is to compare males and female on Math and Spelling
- If groups cannot be included in same model, then formal comparison is impossible.
- Focus on measurement parameters.
 - loadings λ , which indicate an indicator depends on the latent ability (Math or Spelling)
 - item intercepts, which we will call τ

This is called:

Measurement Invariance Analysis is the attempt to determine if the measurement parameters (λ_j , τ_j) are equal for both groups. If they are, then the values of the latent variables might be compared between groups.

Measurement Invariance Analysis

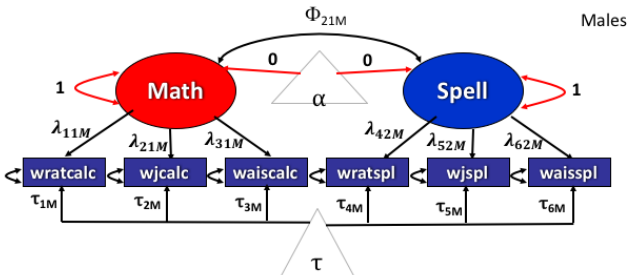
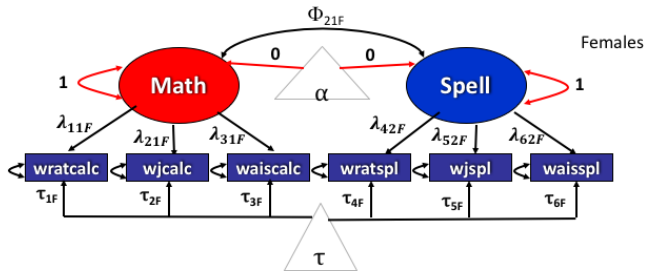
- Quantitative psychologists developed a sequence of steps where we can test for the similarity of the models for the two groups.
 - Step 1. Configural invariance (same variables, same number of factors)
 - Step 2. Metric invariance (equal factor loadings, λ)
 - Step 3. Scalar invariance (equal factor loadings λ and item intercepts τ)

These three models are treating the groups in increasingly similar ways.

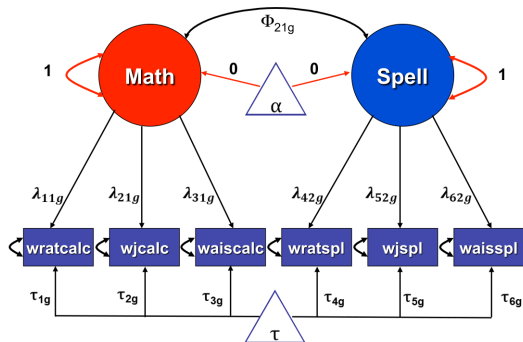
Configural Invariance Model

Configural: The same model applies for $g = \text{Males}$ and $g = \text{Females}$, but...
We allow the loadings λ_{ijg} and intercepts τ_{ig} to differ between the groups.

Configural Invariance Model ...



Another Graphic, perhaps more common type



Mplus code for Configural Analysis

```
TITLE:
  Example 3 - Multiple Group Confirmatory Factor
    Analysis (Configural Invariance)
DATA:
  FILE IS "../..//data/job_placement.csv";
5 VARIABLE:
  NAMES ARE
  id wjcalc wjspl wratspl wratcalc waiscalc waisspl
  edlevel newschl suspend expelled haveld female
  age;
  USEVARIABLES ARE
10 wratspl wjspl waisspl wratcalc wjcalc waiscalc;
  MISSING ARE all(99999);
  GROUPING IS female (0=male 1 =female);
MODEL:
  MATH BY wratcalc* wjcalc waiscalc;
```


Mplus code for Configural Analysis ...

```
15  SPELL BY wratspl* wjspl waisspl;
    [wratspl wjspl waisspl wratcalc wjcalc waiscalc];
    MATH@1 SPELL@1;
    [MATH@0 SPELL@0];
MODEL female:
20  MATH BY wratcalc* wjcalc waiscalc;
    SPELL BY wratspl* wjspl waisspl;
    [wratspl wjspl waisspl wratcalc wjcalc waiscalc];
OUTPUT:
25  TECH1;
    STDYX;
```

Configural Model Results

View Online: [cfa-01-1-configural.out](#)

Inline Display:

```
Mplus VERSION 7.4 (Linux)
MUTHEN & MUTHEN
05/28/2018  9:42 PM
```

INPUT INSTRUCTIONS

TITLE:

```
Example 3 - Multiple Group Confirmatory Factor Analysis (Configural Invariance)
```

DATA:

```
FILE IS "../..data/job_placement.csv";
```

VARIABLE:

NAMES ARE

```
id wjcalc wjspl wratspl wratcalc waiscalc waisspl
edlevel newschl suspend expelled haveld female age;
```

USEVARIABLES ARE

```
wratspl wjspl waisspl wratcalc wjcalc waiscalc;
MISSING ARE all(99999);
```

Configural Model Results ...

```
GROUPING IS female(0=male 1=female);
```

```
MODEL:
```

```
MATH BY wratcalc* wjcalc waiscalc;  
SPELL BY wratspl* wjspl waisspl;  
[wratspl wjspl waisspl wratcalc wjcalc waiscalc];  
MATH@1 SPELL@1;  
[MATH@0 SPELL@0];
```

```
MODEL female:
```

```
MATH BY wratcalc* wjcalc waiscalc;  
SPELL BY wratspl* wjspl waisspl;  
[wratspl wjspl waisspl wratcalc wjcalc waiscalc];
```

```
OUTPUT:
```

```
TECH1;  
STDYX;
```

```
INPUT READING TERMINATED NORMALLY
```

Configural Model Results ...

Example 3 - Multiple Group Confirmatory Factor Analysis (Configural Invariance)

SUMMARY OF ANALYSIS

Number of groups						2
Number of observations						
Group MALE						221
Group FEMALE						101
Total sample size						322
Number of dependent variables						6
Number of independent variables						0
Number of continuous latent variables						2
Observed dependent variables						
Continuous						
WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC	WAISSCALC	
Continuous latent variables						
MATH	SPELL					
Variables with special functions						
Grouping variable	FEMALE					

Configural Model Results ...

Estimator	ML
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/job_placement.csv

Input data format FREE

SUMMARY OF DATA

Group MALE	
Number of missing data patterns	2
Group FEMALE	
Number of missing data patterns	4

COVARIANCE COVERAGE OF DATA

Minimum covariance coverage value 0.100

Configural Model Results ...

PROPORTION OF DATA PRESENT FOR MALE

	Covariance Coverage				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	1.000				
WJSPL	1.000	1.000			
WAISSPL	0.995	0.995	0.995		
WRATCALC	1.000	1.000	0.995	1.000	
WJCALC	1.000	1.000	0.995	1.000	1.000
WAISCALC	0.995	0.995	0.995	0.995	0.995

	Covariance Coverage
	WAISCALC

WAISCALC	0.995

PROPORTION OF DATA PRESENT FOR FEMALE

Covariance Coverage

Configural Model Results ...

	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
WRATSPL	0.990				
WJSPL	0.990	1.000			
WAISSPL	0.980	0.990	0.990		
WRATCALC	0.990	1.000	0.990	1.000	
WJCALC	0.970	0.980	0.970	0.980	0.980
WAISCALC	0.980	0.990	0.990	0.990	0.970

Covariance Coverage

WAISCALC

WAISCALC	0.990
----------	-------

THE MODEL ESTIMATION TERMINATED NORMALLY

MODEL FIT INFORMATION

Number of Free Parameters

38

Loglikelihood

Configural Model Results ...

H0 Value	-5112.407
H1 Value	-5102.800

Information Criteria

Akaike (AIC)	10300.815
Bayesian (BIC)	10444.248
Sample-Size Adjusted BIC	10323.717
(n* = (n + 2) / 24)	

Chi-Square Test of Model Fit

Value	19.215
Degrees of Freedom	16
P-Value	0.2577

Chi-Square Contribution From Each Group

MALE	7.660
FEMALE	11.555

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.035
90 Percent C.I.	0.000 0.085

Configural Model Results ...

Probability RMSEA \leq .05 0.631

CFI/TLI

CFI 0.998

TLI 0.997

Chi-Square Test of Model Fit for the Baseline Model

Value 1905.113

Degrees of Freedom 30

P-Value 0.0000

SRMR (Standardized Root Mean Square Residual)

Value 0.029

MODEL RESULTS

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

Group MALE

Configural Model Results ...

MATH	BY				
WRATCALC		5.764	0.330	17.492	0.000
WJCALC		4.123	0.244	16.926	0.000
WAISCALC		2.446	0.200	12.212	0.000
SPELL	BY				
WRATSPL		6.278	0.337	18.646	0.000
WJSPL		6.788	0.359	18.929	0.000
WAISSPL		6.177	0.335	18.441	0.000
SPELL	WITH				
MATH		0.520	0.053	9.761	0.000
Means					
MATH		0.000	0.000	999.000	999.000
SPELL		0.000	0.000	999.000	999.000
Intercepts					
WRATSPL		36.172	0.448	80.762	0.000
WJSPL		41.371	0.480	86.157	0.000
WAISSPL		36.767	0.444	82.899	0.000
WRATCALC		39.222	0.417	93.961	0.000
WJCALC		23.910	0.305	78.419	0.000
WAISCALC		11.367	0.226	50.321	0.000
Variances					

Configural Model Results ...

MATH	1.000	0.000	999.000	999.000	
SPELL	1.000	0.000	999.000	999.000	
Residual Variances					
WRATSPL	4.915	0.726	6.772	0.000	
WJSPL	4.881	0.795	6.138	0.000	
WAISSPL	5.285	0.736	7.176	0.000	
WRATCALC	5.281	1.232	4.286	0.000	
WJCALC	3.548	0.664	5.341	0.000	
WAISCALC	5.269	0.557	9.463	0.000	
Group FEMALE					
MATH	BY				
WRATCALC		6.545	0.496	13.187	0.000
WJCALC		4.215	0.366	11.530	0.000
WAISCALC		2.290	0.276	8.306	0.000
SPELL	BY				
WRATSPL		7.010	0.547	12.817	0.000
WJSPL		6.833	0.518	13.182	0.000
WAISSPL		6.638	0.520	12.763	0.000
SPELL	WITH				
MATH		0.634	0.064	9.914	0.000

Configural Model Results ...

Means

MATH	0.000	0.000	999.000	999.000
SPELL	0.000	0.000	999.000	999.000

Intercepts

WRATSPL	37.171	0.735	50.589	0.000
WJSPL	42.337	0.705	60.073	0.000
WAISSPL	38.030	0.697	54.578	0.000
WRATCALC	38.267	0.666	57.464	0.000
WJCALC	23.603	0.465	50.781	0.000
WAISCALC	10.266	0.316	32.528	0.000

Variances

MATH	1.000	0.000	999.000	999.000
SPELL	1.000	0.000	999.000	999.000

Residual Variances

WRATSPL	5.307	1.105	4.804	0.000
WJSPL	3.474	0.910	3.818	0.000
WAISSPL	4.900	1.007	4.865	0.000
WRATCALC	1.956	1.621	1.207	0.227
WJCALC	3.960	0.848	4.672	0.000
WAISCALC	4.766	0.714	6.678	0.000

STANDARDIZED MODEL RESULTS

Configural Model Results ...

STDYX Standardization

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
Group MALE				
MATH BY				
WRATCALC	0.929	0.018	50.941	0.000
WJCALC	0.910	0.019	47.330	0.000
WAISCALC	0.729	0.035	20.921	0.000
SPELL BY				
WRATSPL	0.943	0.010	94.427	0.000
WJSPL	0.951	0.009	102.619	0.000
WAISSPL	0.937	0.011	89.110	0.000
SPELL WITH				
MATH	0.520	0.053	9.761	0.000
Means				
MATH	0.000	0.000	999.000	999.000
SPELL	0.000	0.000	999.000	999.000

Configural Model Results ...

Intercepts

WRATSPL	5.433	0.267	20.346	0.000
WJSPL	5.796	0.284	20.425	0.000
WAISSPL	5.579	0.274	20.368	0.000
WRATCALC	6.320	0.308	20.516	0.000
WJCALC	5.275	0.260	20.306	0.000
WAISCALC	3.389	0.175	19.365	0.000

Variances

MATH	1.000	0.000	999.000	999.000
SPELL	1.000	0.000	999.000	999.000

Residual Variances

WRATSPL	0.111	0.019	5.887	0.000
WJSPL	0.096	0.018	5.435	0.000
WAISSPL	0.122	0.020	6.171	0.000
WRATCALC	0.137	0.034	4.048	0.000
WJCALC	0.173	0.035	4.939	0.000
WAISCALC	0.468	0.051	9.214	0.000

Group FEMALE

MATH BY

WRATCALC	0.978	0.019	52.179	0.000
WJCALC	0.904	0.024	37.149	0.000
WAISCALC	0.724	0.050	14.423	0.000

Configural Model Results ...

SPELL	BY				
WRATSPL		0.950	0.012	76.115	0.000
WJSPL		0.965	0.011	91.847	0.000
WAISSPL		0.949	0.013	74.596	0.000
SPELL	WITH				
MATH		0.634	0.064	9.914	0.000
Means					
MATH		0.000	0.000	999.000	999.000
SPELL		0.000	0.000	999.000	999.000
Intercepts					
WRATSPL		5.037	0.368	13.682	0.000
WJSPL		5.977	0.432	13.831	0.000
WAISSPL		5.434	0.396	13.728	0.000
WRATCALC		5.718	0.414	13.797	0.000
WJCALC		5.064	0.371	13.633	0.000
WAISCALC		3.245	0.250	12.985	0.000
Variances					
MATH		1.000	0.000	999.000	999.000
SPELL		1.000	0.000	999.000	999.000
Residual Variances					

Configural Model Results ...

WRATSPL	0.097	0.024	4.110	0.000
WJSPL	0.069	0.020	3.417	0.001
WAISSPL	0.100	0.024	4.147	0.000
WRATCALC	0.044	0.037	1.191	0.234
WJCALC	0.182	0.044	4.141	0.000
WAISSCALC	0.476	0.073	6.552	0.000

R-SQUARE

Group MALE

Observed Variable	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
WRATSPL	0.889	0.019	47.213	0.000
WJSPL	0.904	0.018	51.310	0.000
WAISSPL	0.878	0.020	44.555	0.000
WRATCALC	0.863	0.034	25.471	0.000
WJCALC	0.827	0.035	23.665	0.000
WAISSCALC	0.532	0.051	10.460	0.000

Group FEMALE

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

Configural Model Results ...

NU						
WAISCALC						
1	6					
		LAMBDA				
		MATH	SPELL			
		-----	-----			
WRATSPL	0	7				
WJSPL	0	8				
WAISSPL	0	9				
WRATCALC	10	0				
WJCALC	11	0				
WAISCALC	12	0				
		THETA				
		WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
		-----	-----	-----	-----	-----
WRATSPL	13					
WJSPL	0	14				
WAISSPL	0	0	15			
WRATCALC	0	0	0	16		
WJCALC	0	0	0	0	17	

Configural Model Results ...

WAISCALC 0 0 0 0 0

THETA

 WAISCALC

WAISCALC -----
 18

ALPHA

 MATH

 SPELL

1 -----
 0

 0

BETA

 MATH

 SPELL

MATH -----
 0

 0

SPELL -----
 0

 0

PSI

 MATH

 SPELL

MATH -----
 0

Configural Model Results ...

SPELL 19 0

PARAMETER SPECIFICATION FOR FEMALE

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
1	----- 20	----- 21	----- 22	----- 23	----- 24

	NU
	WAISCALC
1	----- 25

	LAMBDA	
	MATH	SPELL
WRATSPL	----- 0	----- 26
WJSPL	0	27
WAISSPL	0	28
WRATCALC	29	0
WJCALC	30	0

Configural Model Results ...

WAISCALC	31	0			
	THETA				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	32				
WJSPL	0	33			
WAISSPL	0	0	34		
WRATCALC	0	0	0	35	
WJCALC	0	0	0	0	36
WAISCALC	0	0	0	0	0

	THETA	
	WAISCALC	

WAISCALC	37	

	ALPHA	
	MATH	SPELL
	-----	-----
1	0	0

Configural Model Results ...

		BETA	
		MATH	SPELL
MATH	-----	0	-----
SPELL		0	0

		PSI	
		MATH	SPELL
MATH	-----	0	-----
SPELL		38	0

STARTING VALUES FOR MALE

		NU								
		WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC				
1	-----	36.172	-----	41.371	-----	36.786	-----	39.222	-----	23.910

		NU
		WAISCALC

Configural Model Results ...

1 11.350

LAMBDA

	MATH	SPELL
	-----	-----
WRATSPL	0.000	1.000
WJSPL	0.000	1.000
WAISSPL	0.000	1.000
WRATCALC	1.000	0.000
WJCALC	1.000	0.000
WAISCALC	1.000	0.000

THETA

	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	22.166				
WJSPL	0.000	25.479			
WAISSPL	0.000	0.000	21.761		
WRATCALC	0.000	0.000	0.000	19.254	
WJCALC	0.000	0.000	0.000	0.000	10.272
WAISCALC	0.000	0.000	0.000	0.000	0.000

THETA

Configural Model Results ...

	WAISCALC
WAISCALC	5.605

	ALPHA	
	MATH	SPELL
1	0.000	0.000

	BETA	
	MATH	SPELL
MATH	0.000	0.000
SPELL	0.000	0.000

	PSI	
	MATH	SPELL
MATH	1.000	
SPELL	0.000	1.000

STARTING VALUES FOR FEMALE

Configural Model Results ...

	NU WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
1	----- 36.172	----- 41.371	----- 36.786	----- 39.222	----- 23.910

	NU WAISCALC
1	----- 11.350

	LAMBDA MATH	SPELL
WRATSPL	----- 0.000	----- 1.000
WJSPL	0.000	1.000
WAISSPL	0.000	1.000
WRATCALC	1.000	0.000
WJCALC	1.000	0.000
WAISCALC	1.000	0.000

THETA

Configural Model Results ...

	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
WRATSPL	22.166				
WJSPL	0.000	25.479			
WAISSPL	0.000	0.000	21.761		
WRATCALC	0.000	0.000	0.000	19.254	
WJCALC	0.000	0.000	0.000	0.000	10.272
WAISCALC	0.000	0.000	0.000	0.000	0.000

THETA

	WAISCALC
WAISCALC	5.605

ALPHA

	MATH	SPELL
1	0.000	0.000

BETA

	MATH	SPELL
MATH	0.000	0.000

Configural Model Results ...

SPELL 0.000 0.000

PSI

MATH

SPELL

MATH 1.000

SPELL 0.000

1.000

Beginning Time: 21:42:53

Ending Time: 21:42:53

Elapsed Time: 00:00:00

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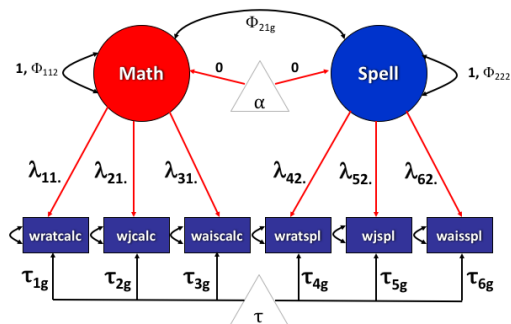
Support: Support@StatModel.com

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Step 2: Metric Invariance

- We fit the Configural Model because we need something to compare against.
- The first model we will compare is the Metric Invariance model.
- In the Metric Invariance analysis, we remove the subscript g from the **loadings**, to assert that loadings are same for both groups

Metric Invariance Illustrated



- The λ coefficients have a "." in place of "g", a reminder of the fact they are not differentiated between groups
- still differing among groups:
 - τ_{ig}
 - Φ

Metric Model: Syntax

```
TITLE:
Example 3 - Multiple Group Confirmatory Factor
          Analysis (Weak Invariance)
DATA:
  FILE IS "../..//data/job_placement.csv";
5 VARIABLE:
  NAMES ARE
  id wjcalc wjspl wratspl wratcalc waiscalc waisspl
  edlevel newschl suspend expelled haveld female
  age;
  USEVARIABLES ARE
10 wratspl wjspl waisspl wratcalc wjcalc waiscalc;
  MISSING ARE all(99999);
  GROUPING IS female (0=male 1 =female);
MODEL:
  MATH BY wratcalc* wjcalc waiscalc;
```

Metric Model: Syntax ...

```
15 SPELL BY wratspl* wjspl waisspl;  
   [wratspl wjspl waisspl wratcalc wjcalc waiscalc];  
MATH@1 SPELL@1;  
   [MATH@0 SPELL@0];  
20 MODEL female:  
   [wratspl wjspl waisspl wratcalc wjcalc waiscalc];  
   MATH SPELL;  
OUTPUT:  
   TECH1;  
   STDYX;
```

Metric Model Results

View Online: [cfa-01-2-metric.out](#)

Inline Display:

Mplus VERSION 7.4 (Linux)
MUTHEN & MUTHEN
05/28/2018 9:42 PM

INPUT INSTRUCTIONS

TITLE:

Example 3 - Multiple Group Confirmatory Factor Analysis (Metric Invariance)

DATA:

FILE IS "../..../data/job_placement.csv";

VARIABLE:

NAMES ARE

id wjcalc wjspl wratspl wratcalc waiscalc waisspl
edlevel newschl suspend expelled haveld female age;

USEVARIABLES ARE

wratspl wjspl waisspl wratcalc wjcalc waiscalc;

Metric Model Results ...

```
MISSING ARE all(99999);
```

```
GROUPING IS female(0=male 1=female);
```

```
MODEL:
```

```
MATH BY wratcalc* wjcalc waiscalc;
```

```
SPELL BY wratspl* wjspl waisspl;
```

```
[wratspl wjspl waisspl wratcalc wjcalc waiscalc];
```

```
MATH@1 SPELL@1;
```

```
[MATH@0 SPELL@0];
```

```
MODEL female:
```

```
[wratspl wjspl waisspl wratcalc wjcalc waiscalc];
```

```
MATH SPELL;
```

```
OUTPUT:
```

```
TECH1;
```

```
STDYX;
```

```
INPUT READING TERMINATED NORMALLY
```

Metric Model Results ...

Example 3 - Multiple Group Confirmatory Factor Analysis (Metric Invariance)

SUMMARY OF ANALYSIS

Number of groups	2
Number of observations	
Group MALE	221
Group FEMALE	101
Total sample size	322

Number of dependent variables	6
Number of independent variables	0
Number of continuous latent variables	2

Observed dependent variables

Continuous

WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC	WAISSCALC
---------	-------	---------	----------	--------	-----------

Continuous latent variables

MATH	SPELL
------	-------

Variables with special functions

Grouping variable	FEMALE
-------------------	--------

Metric Model Results ...

Estimator	ML
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/job_placement.csv

Input data format FREE

SUMMARY OF DATA

Group MALE	
Number of missing data patterns	2
Group FEMALE	
Number of missing data patterns	4

COVARIANCE COVERAGE OF DATA

Minimum covariance coverage value 0.100

Metric Model Results ...

PROPORTION OF DATA PRESENT FOR MALE

	Covariance Coverage				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	1.000				
WJSPL	1.000	1.000			
WAISSPL	0.995	0.995	0.995		
WRATCALC	1.000	1.000	0.995	1.000	
WJCALC	1.000	1.000	0.995	1.000	1.000
WAISCALC	0.995	0.995	0.995	0.995	0.995

	Covariance Coverage
	WAISCALC

WAISCALC	0.995

PROPORTION OF DATA PRESENT FOR FEMALE

Covariance Coverage

Metric Model Results ...

	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
WRATSPL	0.990				
WJSPL	0.990	1.000			
WAISSPL	0.980	0.990	0.990		
WRATCALC	0.990	1.000	0.990	1.000	
WJCALC	0.970	0.980	0.970	0.980	0.980
WAISCALC	0.980	0.990	0.990	0.990	0.970

Covariance Coverage

WAISCALC

WAISCALC	0.990
----------	-------

THE MODEL ESTIMATION TERMINATED NORMALLY

MODEL FIT INFORMATION

Number of Free Parameters 34

Loglikelihood

Metric Model Results ...

H0 Value	-5115.735
H1 Value	-5102.800

Information Criteria

Akaike (AIC)	10299.471
Bayesian (BIC)	10427.805
Sample-Size Adjusted BIC	10319.962
(n* = (n + 2) / 24)	

Chi-Square Test of Model Fit

Value	25.871
Degrees of Freedom	20
P-Value	0.1701

Chi-Square Contribution From Each Group

MALE	10.290
FEMALE	15.581

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.043
90 Percent C.I.	0.000 0.085

Metric Model Results ...

Probability RMSEA \leq .05 0.566

CFI/TLI

CFI 0.997

TLI 0.995

Chi-Square Test of Model Fit for the Baseline Model

Value 1905.113

Degrees of Freedom 30

P-Value 0.0000

SRMR (Standardized Root Mean Square Residual)

Value 0.054

MODEL RESULTS

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

Group MALE

Metric Model Results ...

MATH	BY				
WRATCALC		5.898	0.317	18.614	0.000
WJCALC		4.039	0.232	17.414	0.000
WAISCALC		2.321	0.173	13.389	0.000
SPELL	BY				
WRATSPL		6.387	0.332	19.245	0.000
WJSPL		6.640	0.346	19.186	0.000
WAISSPL		6.213	0.326	19.054	0.000
SPELL	WITH				
MATH		0.519	0.053	9.733	0.000
Means					
MATH		0.000	0.000	999.000	999.000
SPELL		0.000	0.000	999.000	999.000
Intercepts					
WRATSPL		36.172	0.454	79.673	0.000
WJSPL		41.371	0.472	87.637	0.000
WAISSPL		36.767	0.446	82.522	0.000
WRATCALC		39.222	0.422	92.880	0.000
WJCALC		23.910	0.302	79.192	0.000
WAISCALC		11.367	0.221	51.462	0.000

Variances

Metric Model Results ...

MATH	1.000	0.000	999.000	999.000	
SPELL	1.000	0.000	999.000	999.000	
Residual Variances					
WRATSPL	4.764	0.721	6.607	0.000	
WJSPL	5.154	0.790	6.526	0.000	
WAISSPL	5.233	0.732	7.153	0.000	
WRATCALC	4.619	1.205	3.834	0.000	
WJCALC	3.835	0.654	5.864	0.000	
WAISCALC	5.370	0.564	9.525	0.000	
Group FEMALE					
MATH	BY				
WRATCALC		5.898	0.317	18.614	0.000
WJCALC		4.039	0.232	17.414	0.000
WAISCALC		2.321	0.173	13.389	0.000
SPELL	BY				
WRATSPL		6.387	0.332	19.245	0.000
WJSPL		6.640	0.346	19.186	0.000
WAISSPL		6.213	0.326	19.054	0.000
SPELL	WITH				
MATH		0.731	0.150	4.857	0.000

Metric Model Results ...

Means

MATH	0.000	0.000	999.000	999.000
SPELL	0.000	0.000	999.000	999.000

Intercepts

WRATSPL	37.172	0.713	52.136	0.000
WJSPL	42.337	0.721	58.719	0.000
WAISSPL	38.031	0.691	55.052	0.000
WRATCALC	38.267	0.656	58.319	0.000
WJCALC	23.602	0.474	49.832	0.000
WAISCALC	10.263	0.330	31.085	0.000

Variances

MATH	1.162	0.211	5.501	0.000
SPELL	1.117	0.197	5.676	0.000

Residual Variances

WRATSPL	5.686	1.123	5.061	0.000
WJSPL	3.230	0.929	3.478	0.001
WAISSPL	4.995	1.023	4.883	0.000
WRATCALC	3.047	1.413	2.157	0.031
WJCALC	3.603	0.778	4.632	0.000
WAISCALC	4.700	0.714	6.587	0.000

STANDARDIZED MODEL RESULTS

Metric Model Results ...

STDYX Standardization

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
Group MALE				
MATH BY				
WRATCALC	0.940	0.016	57.271	0.000
WJCALC	0.900	0.019	46.925	0.000
WAISCALC	0.708	0.034	20.796	0.000
SPELL BY				
WRATSPL	0.946	0.009	102.287	0.000
WJSPL	0.946	0.010	97.916	0.000
WAISSPL	0.938	0.010	93.327	0.000
SPELL WITH				
MATH	0.519	0.053	9.733	0.000
Means				
MATH	0.000	0.000	999.000	999.000
SPELL	0.000	0.000	999.000	999.000

Metric Model Results ...

Intercepts

WRATSPL	5.359	0.260	20.603	0.000
WJSPL	5.895	0.282	20.922	0.000
WAISSPL	5.553	0.267	20.809	0.000
WRATCALC	6.248	0.304	20.582	0.000
WJCALC	5.327	0.254	20.933	0.000
WAISCALC	3.466	0.163	21.210	0.000

Variances

MATH	1.000	0.000	999.000	999.000
SPELL	1.000	0.000	999.000	999.000

Residual Variances

WRATSPL	0.105	0.018	5.973	0.000
WJSPL	0.105	0.018	5.723	0.000
WAISSPL	0.119	0.019	6.325	0.000
WRATCALC	0.117	0.031	3.802	0.000
WJCALC	0.190	0.035	5.517	0.000
WAISCALC	0.499	0.048	10.368	0.000

Group FEMALE

MATH BY

WRATCALC	0.964	0.018	54.629	0.000
WJCALC	0.917	0.020	46.823	0.000
WAISCALC	0.756	0.038	19.875	0.000

Metric Model Results ...

SPELL	BY				
WRATSPL		0.943	0.014	69.743	0.000
WJSPL		0.969	0.010	101.347	0.000
WAISSPL		0.947	0.013	73.865	0.000
SPELL	WITH				
MATH		0.641	0.063	10.165	0.000
Means					
MATH		0.000	0.000	999.000	999.000
SPELL		0.000	0.000	999.000	999.000
Intercepts					
WRATSPL		5.192	0.361	14.385	0.000
WJSPL		5.843	0.420	13.926	0.000
WAISSPL		5.482	0.383	14.308	0.000
WRATCALC		5.803	0.416	13.950	0.000
WJCALC		4.969	0.353	14.068	0.000
WAISSCALC		3.100	0.218	14.229	0.000
Variances					
MATH		1.000	0.000	999.000	999.000
SPELL		1.000	0.000	999.000	999.000
Residual Variances					

Metric Model Results ...

WRATSPL	0.111	0.025	4.350	0.000
WJSPL	0.062	0.019	3.322	0.001
WAISSPL	0.104	0.024	4.277	0.000
WRATCALC	0.070	0.034	2.058	0.040
WJCALC	0.160	0.036	4.449	0.000
WAISSCALC	0.429	0.057	7.461	0.000

R-SQUARE

Group MALE

Observed Variable	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
WRATSPL	0.895	0.018	51.143	0.000
WJSPL	0.895	0.018	48.958	0.000
WAISSPL	0.881	0.019	46.663	0.000
WRATCALC	0.883	0.031	28.635	0.000
WJCALC	0.810	0.035	23.462	0.000
WAISSCALC	0.501	0.048	10.398	0.000

Group FEMALE

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

Metric Model Results ...

WRATSPL	0.889	0.025	34.872	0.000
WJSPL	0.938	0.019	50.673	0.000
WAISSPL	0.896	0.024	36.933	0.000
WRATCALC	0.930	0.034	27.314	0.000
WJCALC	0.840	0.036	23.411	0.000
WAISCALC	0.571	0.057	9.938	0.000

QUALITY OF NUMERICAL RESULTS

Condition Number for the Information Matrix 0.626E-04
 (ratio of smallest to largest eigenvalue)

TECHNICAL 1 OUTPUT

PARAMETER SPECIFICATION FOR MALE

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
1	1	2	3	4	5

Metric Model Results ...

NU
 WAISCALC

 1 6

	LAMBDA	
	MATH	SPELL
	-----	-----
WRATSPL	0	7
WJSPL	0	8
WAISSPL	0	9
WRATCALC	10	0
WJCALC	11	0
WAISCALC	12	0

	THETA				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	13				
WJSPL	0	14			
WAISSPL	0	0	15		
WRATCALC	0	0	0	16	
WJCALC	0	0	0	0	17

Metric Model Results ...

WAISCALC	0	0	0	0	0
----------	---	---	---	---	---

THETA

WAISCALC

WAISCALC	-----
	18

ALPHA

MATH

SPELL

1

0

0

BETA

MATH

SPELL

MATH	-----	0
------	-------	---

SPELL	0
-------	---

-----	0
-------	---

0

PSI

MATH

SPELL

MATH	-----	0
------	-------	---

Metric Model Results ...

SPELL 19 0

PARAMETER SPECIFICATION FOR FEMALE

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
1	----- 20	----- 21	----- 22	----- 23	----- 24

	NU
	WAISCALC
1	----- 25

	LAMBDA	
	MATH	SPELL
WRATSPL	----- 0	----- 7
WJSPL	0	8
WAISSPL	0	9
WRATCALC	10	0
WJCALC	11	0

Metric Model Results ...

WAISCALC	12	0			
	THETA				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	26				
WJSPL	0	27			
WAISSPL	0	0	28		
WRATCALC	0	0	0	29	
WJCALC	0	0	0	0	30
WAISCALC	0	0	0	0	0

	THETA	
	WAISCALC	

WAISCALC	31	

	ALPHA	
	MATH	SPELL
	-----	-----
1	0	0

Metric Model Results ...

	BETA	
	MATH	SPELL
MATH	----- 0	----- 0
SPELL	0	0

	PSI	
	MATH	SPELL
MATH	----- 32	-----
SPELL	33	34

STARTING VALUES FOR MALE

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
1	----- 36.172	----- 41.371	----- 36.786	----- 39.222	----- 23.910

	NU
	WAISCALC

Metric Model Results ...

1 11.350

LAMBDA

MATH

SPELL

	-----	-----
WRATSPL	0.000	1.000
WJSPL	0.000	1.000
WAISSPL	0.000	1.000
WRATCALC	1.000	0.000
WJCALC	1.000	0.000
WAISCALC	1.000	0.000

THETA

WRATSPL

WJSPL

WAISSPL

WRATCALC

WJCALC

	-----	-----	-----	-----	-----
WRATSPL	22.166				
WJSPL	0.000	25.479			
WAISSPL	0.000	0.000	21.761		
WRATCALC	0.000	0.000	0.000	19.254	
WJCALC	0.000	0.000	0.000	0.000	10.272
WAISCALC	0.000	0.000	0.000	0.000	0.000

THETA

Metric Model Results ...

	WAISCALC

WAISCALC	5.605

	ALPHA	
	MATH	SPELL
	-----	-----
1	0.000	0.000

	BETA	
	MATH	SPELL
	-----	-----
MATH	0.000	0.000
SPELL	0.000	0.000

	PSI	
	MATH	SPELL
	-----	-----
MATH	1.000	
SPELL	0.000	1.000

STARTING VALUES FOR FEMALE

Metric Model Results ...

	NU WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
1	----- 36.172	----- 41.371	----- 36.786	----- 39.222	----- 23.910

	NU WAISCALC
1	----- 11.350

	LAMBDA MATH	SPELL
WRATSPL	----- 0.000	----- 7.010
WJSPL	0.000	6.828
WAISSPL	0.000	6.653
WRATCALC	6.633	0.000
WJCALC	4.167	0.000
WAISCALC	2.259	0.000

THETA

Metric Model Results ...

	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
WRATSPL	22.166				
WJSPL	0.000	25.479			
WAISSPL	0.000	0.000	21.761		
WRATCALC	0.000	0.000	0.000	19.254	
WJCALC	0.000	0.000	0.000	0.000	10.272
WAISCALC	0.000	0.000	0.000	0.000	0.000

THETA

WAISCALC

WAISCALC	5.605
----------	-------

ALPHA

MATH

SPELL

1	0.000	0.000
---	-------	-------

BETA

MATH

SPELL

MATH	0.000	0.000
------	-------	-------

Metric Model Results ...

SPELL 0.000 0.000

PSI

MATH

SPELL

MATH 1.000

SPELL 0.000

1.000

Beginning Time: 21:42:53

Ending Time: 21:42:53

Elapsed Time: 00:00:00

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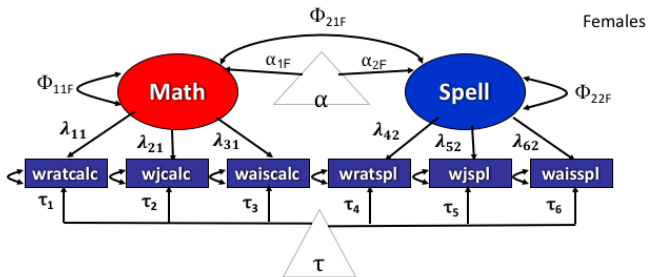
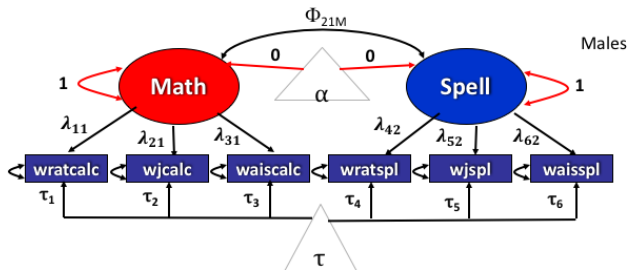
Support: Support@StatModel.com

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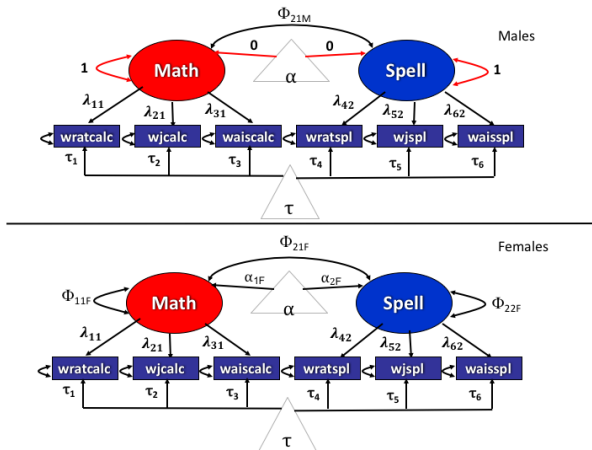
Step 3: Scalar Invariance Model

- Assume not only that the loadings are the same between groups, but also
- the intercepts on the measurement indicators (τ_i) are the same

Scalar Invariance Illustrated



Scalar Invariance Model



- The last subscript disappeared on λ and τ . No more group differentiation
- Only remaining group differences are the latent variable's
 - intercept, 0 for M, α for F, reflecting the average group difference
 - And variances, Φ_{11g} , Φ_{22g} , and covariance Φ_{21g}

Scalar Invariance Model Syntax

```
TITLE:
Example 3 - Multiple Group Confirmatory Factor
      Analysis (Strong Invariance)
DATA:
  FILE IS "../..//data/job_placement.csv";
5 VARIABLE:
  NAMES ARE
  id wjcalc wjspl wratspl wratcalc waiscalc waisspl
  edlevel newschl suspend expelled haveld female
  age;
  USEVARIABLES ARE
10 wratspl wjspl waisspl wratcalc wjcalc waiscalc;
  MISSING ARE all(99999);
  GROUPING IS female (0=male 1 =female);
MODEL:
  MATH BY wratcalc* wjcalc waiscalc;
```

Scalar Invariance Model Syntax ...

```
15 SPELL BY wratspl* wjspl waisspl;  
[wratspl wjspl waisspl wratcalc wjcalc waiscalc];  
MATH@1 SPELL@1;  
[MATH@0 SPELL@0];
```

```
20 MODEL female:  
MATH SPELL;  
[MATH SPELL];
```

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

Scalar Model Estimates

View Online: [cfa-01-3-scalar.out](#)

Inline Display:

```
Mplus VERSION 7.4 (Linux)
MUTHEN & MUTHEN
05/28/2018  9:42 PM
```

INPUT INSTRUCTIONS

TITLE:

```
Example 3 - Multiple Group Confirmatory Factor Analysis (Scalar Invariance)
```

DATA:

```
FILE IS "../..data/job_placement.csv";
```

VARIABLE:

NAMES ARE

```
id wjcalc wjspl wratspl wratcalc waiscalc waisspl
edlevel newschl suspend expelled haveld female age;
```

USEVARIABLES ARE

```
wratspl wjspl waisspl wratcalc wjcalc waiscalc;
```

Scalar Model Estimates ...

```
MISSING ARE all(99999);
```

```
GROUPING IS female(0=male 1=female);
```

```
MODEL:
```

```
MATH BY wratcalc* wjcalc waiscalc;
```

```
SPELL BY wratspl* wjspl waisspl;
```

```
[wratspl wjspl waisspl wratcalc wjcalc waiscalc];
```

```
MATH@1 SPELL@1;
```

```
[MATH@0 SPELL@0];
```

```
MODEL female:
```

```
MATH SPELL;
```

```
[MATH SPELL];
```

```
OUTPUT:
```

```
TECH1;
```

```
STDYX;
```

```
INPUT READING TERMINATED NORMALLY
```


Scalar Model Estimates ...

Example 3 - Multiple Group Confirmatory Factor Analysis (Scalar Invariance)

SUMMARY OF ANALYSIS

Number of groups	2
Number of observations	
Group MALE	221
Group FEMALE	101
Total sample size	322

Number of dependent variables	6
Number of independent variables	0
Number of continuous latent variables	2

Observed dependent variables

Continuous

WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC	WAISSCALC
---------	-------	---------	----------	--------	-----------

Continuous latent variables

MATH	SPELL
------	-------

Variables with special functions

Grouping variable	FEMALE
-------------------	--------

Scalar Model Estimates ...

Estimator	ML
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/job_placement.csv

Input data format FREE

SUMMARY OF DATA

Group MALE	
Number of missing data patterns	2
Group FEMALE	
Number of missing data patterns	4

COVARIANCE COVERAGE OF DATA

Minimum covariance coverage value 0.100

Scalar Model Estimates ...

PROPORTION OF DATA PRESENT FOR MALE

	Covariance Coverage				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	1.000				
WJSPL	1.000	1.000			
WAISSPL	0.995	0.995	0.995		
WRATCALC	1.000	1.000	0.995	1.000	
WJCALC	1.000	1.000	0.995	1.000	1.000
WAISCALC	0.995	0.995	0.995	0.995	0.995

	Covariance Coverage
	WAISCALC

WAISCALC	0.995

PROPORTION OF DATA PRESENT FOR FEMALE

Covariance Coverage

Scalar Model Estimates ...

	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
WRATSPL	0.990				
WJSPL	0.990	1.000			
WAISSPL	0.980	0.990	0.990		
WRATCALC	0.990	1.000	0.990	1.000	
WJCALC	0.970	0.980	0.970	0.980	0.980
WAISCALC	0.980	0.990	0.990	0.990	0.970

Covariance Coverage

WAISCALC

WAISCALC	0.990
----------	-------

THE MODEL ESTIMATION TERMINATED NORMALLY

MODEL FIT INFORMATION

Number of Free Parameters 30

Loglikelihood

Scalar Model Estimates ...

H0 Value	-5120.981
H1 Value	-5102.800

Information Criteria

Akaike (AIC)	10301.961
Bayesian (BIC)	10415.198
Sample-Size Adjusted BIC	10320.042
(n* = (n + 2) / 24)	

Chi-Square Test of Model Fit

Value	36.362
Degrees of Freedom	24
P-Value	0.0506

Chi-Square Contribution From Each Group

MALE	13.816
FEMALE	22.546

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.057
90 Percent C.I.	0.000 0.092

Scalar Model Estimates ...

Probability RMSEA \leq .05 0.356

CFI/TLI

CFI 0.993

TLI 0.992

Chi-Square Test of Model Fit for the Baseline Model

Value 1905.113

Degrees of Freedom 30

P-Value 0.0000

SRMR (Standardized Root Mean Square Residual)

Value 0.053

MODEL RESULTS

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

Group MALE

Scalar Model Estimates ...

MATH	BY				
WRATCALC		5.907	0.317	18.658	0.000
WJCALC		4.017	0.232	17.352	0.000
WAISCALC		2.341	0.175	13.388	0.000
SPELL	BY				
WRATSPL		6.385	0.332	19.249	0.000
WJSPL		6.634	0.346	19.183	0.000
WAISSPL		6.221	0.326	19.064	0.000
SPELL	WITH				
MATH		0.519	0.053	9.732	0.000
Means					
MATH		0.000	0.000	999.000	999.000
SPELL		0.000	0.000	999.000	999.000
Intercepts					
WRATSPL		36.157	0.449	80.484	0.000
WJSPL		41.317	0.466	88.688	0.000
WAISSPL		36.843	0.440	83.721	0.000
WRATCALC		39.217	0.419	93.584	0.000
WJCALC		24.016	0.295	81.470	0.000
WAISCALC		11.125	0.207	53.742	0.000
Variances					

Scalar Model Estimates ...

MATH	1.000	0.000	999.000	999.000	
SPELL	1.000	0.000	999.000	999.000	
Residual Variances					
WRATSPL	4.761	0.721	6.603	0.000	
WJSPL	5.173	0.791	6.541	0.000	
WAISSPL	5.235	0.734	7.134	0.000	
WRATCALC	4.518	1.204	3.753	0.000	
WJCALC	3.911	0.655	5.970	0.000	
WAISCALC	5.431	0.576	9.435	0.000	
Group FEMALE					
MATH	BY				
WRATCALC		5.907	0.317	18.658	0.000
WJCALC		4.017	0.232	17.352	0.000
WAISCALC		2.341	0.175	13.388	0.000
SPELL	BY				
WRATSPL		6.385	0.332	19.249	0.000
WJSPL		6.634	0.346	19.183	0.000
WAISSPL		6.221	0.326	19.064	0.000
SPELL	WITH				
MATH		0.731	0.151	4.855	0.000

Scalar Model Estimates ...

Means

MATH	-0.160	0.131	-1.223	0.221
SPELL	0.165	0.127	1.296	0.195

Intercepts

WRATSPL	36.157	0.449	80.484	0.000
WJSPL	41.317	0.466	88.688	0.000
WAISSPL	36.843	0.440	83.721	0.000
WRATCALC	39.217	0.419	93.584	0.000
WJCALC	24.016	0.295	81.470	0.000
WAISCALC	11.125	0.207	53.742	0.000

Variances

MATH	1.167	0.212	5.502	0.000
SPELL	1.118	0.197	5.676	0.000

Residual Variances

WRATSPL	5.681	1.123	5.056	0.000
WJSPL	3.237	0.930	3.481	0.000
WAISSPL	5.035	1.032	4.880	0.000
WRATCALC	2.841	1.424	1.995	0.046
WJCALC	3.743	0.793	4.723	0.000
WAISCALC	4.977	0.766	6.496	0.000

STANDARDIZED MODEL RESULTS

Scalar Model Estimates ...

STDYX Standardization

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
Group MALE				
MATH BY				
WRATCALC	0.941	0.016	57.567	0.000
WJCALC	0.897	0.019	46.272	0.000
WAISCALC	0.709	0.034	20.811	0.000
SPELL BY				
WRATSPL	0.946	0.009	102.225	0.000
WJSPL	0.946	0.010	97.559	0.000
WAISSPL	0.939	0.010	93.435	0.000
SPELL WITH				
MATH	0.519	0.053	9.732	0.000
Means				
MATH	0.000	0.000	999.000	999.000
SPELL	0.000	0.000	999.000	999.000

Scalar Model Estimates ...

Intercepts

WRATSPL	5.359	0.260	20.603	0.000
WJSPL	5.892	0.282	20.925	0.000
WAISSPL	5.558	0.267	20.819	0.000
WRATCALC	6.247	0.304	20.582	0.000
WJCALC	5.363	0.254	21.086	0.000
WAISCALC	3.368	0.160	21.063	0.000

Variances

MATH	1.000	0.000	999.000	999.000
SPELL	1.000	0.000	999.000	999.000

Residual Variances

WRATSPL	0.105	0.018	5.970	0.000
WJSPL	0.105	0.018	5.733	0.000
WAISSPL	0.119	0.019	6.319	0.000
WRATCALC	0.115	0.031	3.727	0.000
WJCALC	0.195	0.035	5.607	0.000
WAISCALC	0.498	0.048	10.312	0.000

Group FEMALE

MATH BY

WRATCALC	0.967	0.018	54.795	0.000
WJCALC	0.913	0.020	45.824	0.000
WAISCALC	0.750	0.039	19.366	0.000

Scalar Model Estimates ...

SPELL	BY				
WRATSPL		0.943	0.014	69.796	0.000
WJSPL		0.969	0.010	101.046	0.000
WAISSPL		0.946	0.013	73.563	0.000
SPELL	WITH				
MATH		0.640	0.063	10.128	0.000
Means					
MATH		-0.148	0.121	-1.220	0.222
SPELL		0.156	0.121	1.293	0.196
Intercepts					
WRATSPL		5.050	0.343	14.709	0.000
WJSPL		5.705	0.403	14.156	0.000
WAISSPL		5.301	0.363	14.585	0.000
WRATCALC		5.942	0.419	14.183	0.000
WJCALC		5.054	0.350	14.428	0.000
WAISSCALC		3.299	0.210	15.730	0.000
Variances					
MATH		1.000	0.000	999.000	999.000
SPELL		1.000	0.000	999.000	999.000
Residual Variances					

Scalar Model Estimates ...

WRATSPL	0.111	0.025	4.349	0.000
WJSPL	0.062	0.019	3.324	0.001
WAISSPL	0.104	0.024	4.280	0.000
WRATCALC	0.065	0.034	1.912	0.056
WJCALC	0.166	0.036	4.553	0.000
WAISSCALC	0.438	0.058	7.536	0.000

R-SQUARE

Group MALE

Observed Variable	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
WRATSPL	0.895	0.018	51.112	0.000
WJSPL	0.895	0.018	48.779	0.000
WAISSPL	0.881	0.019	46.717	0.000
WRATCALC	0.885	0.031	28.784	0.000
WJCALC	0.805	0.035	23.136	0.000
WAISSCALC	0.502	0.048	10.406	0.000

Group FEMALE

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

Scalar Model Estimates ...

WRATSPL	0.889	0.025	34.898	0.000
WJSPL	0.938	0.019	50.523	0.000
WAISSPL	0.896	0.024	36.781	0.000
WRATCALC	0.935	0.034	27.398	0.000
WJCALC	0.834	0.036	22.912	0.000
WAISCALC	0.562	0.058	9.683	0.000

QUALITY OF NUMERICAL RESULTS

Condition Number for the Information Matrix 0.345E-04
 (ratio of smallest to largest eigenvalue)

TECHNICAL 1 OUTPUT

PARAMETER SPECIFICATION FOR MALE

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
1	1	2	3	4	5

Scalar Model Estimates ...

NU
 WAISCALC

 1 6

	LAMBDA	
	MATH	SPELL
	-----	-----
WRATSPL	0	7
WJSPL	0	8
WAISSPL	0	9
WRATCALC	10	0
WJCALC	11	0
WAISCALC	12	0

	THETA				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	13				
WJSPL	0	14			
WAISSPL	0	0	15		
WRATCALC	0	0	0	16	
WJCALC	0	0	0	0	17

Scalar Model Estimates ...

WAISCALC 0 0 0 0 0

THETA

 WAISCALC

WAISCALC -----
 18

ALPHA

 MATH

 SPELL

1

 0

 0

BETA

 MATH

 SPELL

MATH -----
 0

 0

SPELL -----
 0

 0

PSI

 MATH

 SPELL

MATH -----
 0

Scalar Model Estimates ...

SPELL 19 0

PARAMETER SPECIFICATION FOR FEMALE

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
1	1	2	3	4	5

	NU
	WAISCALC

1	6

	LAMBDA	
	MATH	SPELL
	-----	-----
WRATSPL	0	7
WJSPL	0	8
WAISSPL	0	9
WRATCALC	10	0
WJCALC	11	0

Scalar Model Estimates ...

WAISCALC	12	0			
	THETA				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	20				
WJSPL	0	21			
WAISSPL	0	0	22		
WRATCALC	0	0	0	23	
WJCALC	0	0	0	0	24
WAISCALC	0	0	0	0	0

	THETA	
	WAISCALC	

WAISCALC	25	

	ALPHA	
	MATH	SPELL
	-----	-----
1	26	27

Scalar Model Estimates ...

	BETA	
	MATH	SPELL
MATH	----- 0	----- 0
SPELL	0	0

	PSI	
	MATH	SPELL
MATH	----- 28	----- 30
SPELL	29	30

STARTING VALUES FOR MALE

	NU				
	WRATSPL	WJSPL	WAISPL	WRATCALC	WJCALC
1	----- 36.172	----- 41.371	----- 36.786	----- 39.222	----- 23.910

	NU
	WAISCALC

Scalar Model Estimates ...

1 11.350

LAMBDA

MATH

SPELL

	-----	-----
WRATSPL	0.000	1.000
WJSPL	0.000	1.000
WAISSPL	0.000	1.000
WRATCALC	1.000	0.000
WJCALC	1.000	0.000
WAISCALC	1.000	0.000

THETA

WRATSPL

WJSPL

WAISSPL

WRATCALC

WJCALC

	-----	-----	-----	-----	-----
WRATSPL	22.166				
WJSPL	0.000	25.479			
WAISSPL	0.000	0.000	21.761		
WRATCALC	0.000	0.000	0.000	19.254	
WJCALC	0.000	0.000	0.000	0.000	10.272
WAISCALC	0.000	0.000	0.000	0.000	0.000

THETA

Scalar Model Estimates ...

	WAISCALC
WAISCALC	5.605

	ALPHA	
	MATH	SPELL
1	0.000	0.000

	BETA	
	MATH	SPELL
MATH	0.000	0.000
SPELL	0.000	0.000

	PSI	
	MATH	SPELL
MATH	1.000	
SPELL	0.000	1.000

STARTING VALUES FOR FEMALE

Scalar Model Estimates ...

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
1	36.172	41.371	36.786	39.222	23.910

	NU
	WAISCALC
1	11.350

	LAMBDA	
	MATH	SPELL
WRATSPL	0.000	7.010
WJSPL	0.000	6.828
WAISSPL	0.000	6.653
WRATCALC	6.633	0.000
WJCALC	4.167	0.000
WAISCALC	2.259	0.000

THETA

Scalar Model Estimates ...

	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
WRATSPL	22.166				
WJSPL	0.000	25.479			
WAISSPL	0.000	0.000	21.761		
WRATCALC	0.000	0.000	0.000	19.254	
WJCALC	0.000	0.000	0.000	0.000	10.272
WAISCALC	0.000	0.000	0.000	0.000	0.000

THETA

WAISCALC

WAISCALC 5.605

ALPHA

MATH

SPELL

1 0.000

0.000

BETA

MATH

SPELL

MATH 0.000

0.000

Nested Model Comparisons

- The most general model is the Configural Model, the two groups are treated differently
- We want to know if the “same loadings” requirement caused the quality of the fitted model to be reduced.
 - This is called a “chi square difference test”. The estimated χ^2 stats from the 2 models are used

$$\chi_{diff}^2 = \chi_{metric}^2 - \chi_{config}^2$$

The degrees of freedom for

$$\chi_{diff}^2 = df_{metric} - df_{config}$$

- If that test says “there’s no difference”, then we proceed to compare the scalar model.

This is a special Mplus feature

- Mplus 7.1 introduced an automated way to conduct the difference of models test
- The Measurement Invariance Testing offered by Mplus, is considered to be one of the value-added components in the program.
- Related to the “difftest”, another way Mplus offers to test the difference between two models.

Request a Nested Model Comparison

- Modify the MODEL and ANALYSIS stanzas in the Mplus program.
- The first part of program is the same

```
TITLE:  
Example 3 - Multiple Group Confirmatory Factor  
Analysis (Strong Invariance)  
DATA:  
FILE IS "../..//data/job_placement.csv";  
VARIABLE:  
NAMES ARE  
id wjcalc wjspl wratspl wratcalc waiscalc waisspl  
edlevel newschl suspend expelled haveld female  
age;  
USEVARIABLES ARE  
wratspl wjspl waisspl wratcalc wjcalc waiscalc;  
MISSING ARE all(99999);  
GROUPING IS female (0=male 1 =female);
```

Request a Nested Model Comparison ...

- Here is the part that is different:

```
ANALYSIS:
```

```
    Model = configural metric scalar;
```

```
MODEL:
```

```
    MATH BY wratcalc wjcalc waiscalc;
```

```
    SPELL BY wratspl wjspl waisspl;
```

- Same ending stanza

```
OUTPUT:
```

```
    TECH1;
```

```
    STDYX;
```

Scalar Invariance result

View Online: [cfa-01-4-shortcut.out](#)

Inline Display:

```
Mplus VERSION 7.4 (Linux)
MUTHEN & MUTHEN
05/28/2018  9:42 PM
```

INPUT INSTRUCTIONS

TITLE:

```
Example 3 - Multiple Group Confirmatory Factor Analysis
```

DATA:

```
FILE IS "../..../data/job_placement.csv";
```

VARIABLE:

NAMES ARE

```
id wjcalc wjspl wratspl wratcalc waiscalc waisspl
edlevel newschl suspend expelled haveld female age;
```

USEVARIABLES ARE

```
wratspl wjspl waisspl wratcalc wjcalc waiscalc;
```

Scalar Invariance result ...

```
MISSING ARE all(99999);  
  
GROUPING IS female(0=male 1=female);
```

ANALYSIS:

```
Model = configural metric scalar;
```

MODEL:

```
MATH BY wratcalc wjcalc waiscalc;  
SPELL BY wratspl wjspl waisspl;
```

OUTPUT:

```
TECH1;
```

INPUT READING TERMINATED NORMALLY

Example 3 - Multiple Group Confirmatory Factor Analysis

SUMMARY OF ANALYSIS

Number of groups

2

Scalar Invariance result ...

```

Number of observations
  Group MALE                                221
  Group FEMALE                              101
  Total sample size                          322

Number of dependent variables                6
Number of independent variables              0
Number of continuous latent variables        2

Observed dependent variables

Continuous
  WRATSPL      WJSPL      WAISSPL      WRATCALC      WJCALC      WAISCALC

Continuous latent variables
  MATH          SPELL

Variables with special functions

  Grouping variable      FEMALE

Estimator                                ML
Information matrix        OBSERVED
Maximum number of iterations          1000
Convergence criterion        0.500D-04
Maximum number of steepest descent iterations      20

```

Scalar Invariance result ...

```
Maximum number of iterations for H1          2000
Convergence criterion for H1                0.100D-03
```

```
Input data file(s)
  ../../data/job_placement.csv
```

```
Input data format  FREE
```

SUMMARY OF DATA

```
Group MALE
  Number of missing data patterns          2
```

```
Group FEMALE
  Number of missing data patterns          4
```

COVARIANCE COVERAGE OF DATA

```
Minimum covariance coverage value  0.100
```

```
PROPORTION OF DATA PRESENT FOR MALE
```


Scalar Invariance result ...

	Covariance Coverage				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	1.000				
WJSPL	1.000	1.000			
WAISSPL	0.995	0.995	0.995		
WRATCALC	1.000	1.000	0.995	1.000	
WJCALC	1.000	1.000	0.995	1.000	1.000
WAISCALC	0.995	0.995	0.995	0.995	0.995

	Covariance Coverage
	WAISCALC

WAISCALC	0.995

PROPORTION OF DATA PRESENT FOR FEMALE

	Covariance Coverage				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	0.990				
WJSPL	0.990	1.000			
WAISSPL	0.980	0.990	0.990		

Scalar Invariance result ...

WRATCALC	0.990	1.000	0.990	1.000	
WJCALC	0.970	0.980	0.970	0.980	0.980
WAISCALC	0.980	0.990	0.990	0.990	0.970

Covariance Coverage

WAISCALC

WAISCALC	-----
	0.990

MODEL FIT INFORMATION

Invariance Testing

Model	Number of Parameters	Chi-Square	Degrees of Freedom	P-Value
Configural	38	19.215	16	0.2577
Metric	34	25.871	20	0.1701
Scalar	30	36.362	24	0.0506

Models Compared	Chi-Square	Degrees of Freedom	P-Value
Metric against Configural	6.656	4	0.1552

Scalar Invariance result ...

Scalar against Configural	17.146	8	0.0286
Scalar against Metric	10.490	4	0.0329

MODEL FIT INFORMATION FOR THE CONFIGURAL MODEL

Number of Free Parameters 38

Loglikelihood

H0 Value	-5112.407
H1 Value	-5102.800

Information Criteria

Akaike (AIC)	10300.815
Bayesian (BIC)	10444.248
Sample-Size Adjusted BIC	10323.717
(n* = (n + 2) / 24)	

Chi-Square Test of Model Fit

Value	19.215
Degrees of Freedom	16
P-Value	0.2577

Scalar Invariance result ...

Chi-Square Contribution and P-Value From Each Group (degrees of freedom = 8)

MALE	7.660	0.467
FEMALE	11.555	0.172

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.035	
90 Percent C.I.	0.000	0.085
Probability RMSEA <= .05	0.631	

CFI/TLI

CFI	0.998
TLI	0.997

Chi-Square Test of Model Fit for the Baseline Model

Value	1905.113
Degrees of Freedom	30
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

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

Scalar Invariance result ...

MODEL FIT INFORMATION FOR THE METRIC MODEL

Number of Free Parameters 34

Loglikelihood

H0 Value -5115.735

H1 Value -5102.800

Information Criteria

Akaike (AIC) 10299.471

Bayesian (BIC) 10427.805

Sample-Size Adjusted BIC 10319.962

($n^* = (n + 2) / 24$)

Chi-Square Test of Model Fit

Value 25.871

Degrees of Freedom 20

P-Value 0.1701

Chi-Square Contribution From Each Group

MALE 10.290

Scalar Invariance result ...

FEMALE 15.581

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.043	
90 Percent C.I.	0.000	0.085
Probability RMSEA <= .05	0.566	

CFI/TLI

CFI	0.997
TLI	0.995

Chi-Square Test of Model Fit for the Baseline Model

Value	1905.113
Degrees of Freedom	30
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value	0.054
-------	-------

MODEL FIT INFORMATION FOR THE SCALAR MODEL

Scalar Invariance result ...

Number of Free Parameters 30

Loglikelihood

H0 Value	-5120.981
H1 Value	-5102.800

Information Criteria

Akaike (AIC)	10301.961
Bayesian (BIC)	10415.198
Sample-Size Adjusted BIC	10320.042
(n* = (n + 2) / 24)	

Chi-Square Test of Model Fit

Value	36.362
Degrees of Freedom	24
P-Value	0.0506

Chi-Square Contribution From Each Group

MALE	13.815
FEMALE	22.547

RMSEA (Root Mean Square Error Of Approximation)

Scalar Invariance result ...

Estimate	0.057	
90 Percent C.I.	0.000	0.092
Probability RMSEA <= .05	0.356	

CFI/TLI

CFI	0.993
TLI	0.992

Chi-Square Test of Model Fit for the Baseline Model

Value	1905.113
Degrees of Freedom	30
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value	0.053
-------	-------

MODEL RESULTS FOR THE CONFIGURAL MODEL

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

Scalar Invariance result ...

Group MALE

MATH	BY				
WRATCALC		1.000	0.000	999.000	999.000
WJCALC		0.715	0.037	19.444	0.000
WAISCALC		0.424	0.032	13.334	0.000
SPELL	BY				
WRATSPL		1.000	0.000	999.000	999.000
WJSPL		1.081	0.037	29.078	0.000
WAISSPL		0.984	0.036	27.585	0.000
SPELL	WITH				
MATH		18.836	2.940	6.407	0.000
Means					
MATH		0.000	0.000	999.000	999.000
SPELL		0.000	0.000	999.000	999.000
Intercepts					
WRATSPL		36.172	0.448	80.762	0.000
WJSPL		41.371	0.480	86.157	0.000
WAISSPL		36.767	0.444	82.899	0.000
WRATCALC		39.222	0.417	93.962	0.000
WJCALC		23.910	0.305	78.420	0.000

Scalar Invariance result ...

WAISCALC	11.367	0.226	50.321	0.000
Variances				
MATH	33.227	3.799	8.746	0.000
SPELL	39.417	4.228	9.323	0.000
Residual Variances				
WRATSPL	4.915	0.726	6.772	0.000
WJSPL	4.881	0.795	6.138	0.000
WAISSPL	5.285	0.736	7.176	0.000
WRATCALC	5.280	1.232	4.285	0.000
WJCALC	3.549	0.664	5.342	0.000
WAISCALC	5.269	0.557	9.463	0.000
Group FEMALE				
MATH	BY			
WRATCALC		1.000	0.000	999.000
WJCALC		0.644	0.040	15.980
WAISCALC		0.350	0.036	9.607
SPELL	BY			
WRATSPL		1.000	0.000	999.000
WJSPL		0.975	0.044	22.359
WAISSPL		0.947	0.046	20.668

Scalar Invariance result ...

SPELL	WITH				
MATH		29.101	5.608	5.189	0.000
Means					
MATH		0.000	0.000	999.000	999.000
SPELL		0.000	0.000	999.000	999.000
Intercepts					
WRATSPL		37.171	0.735	50.589	0.000
WJSPL		42.337	0.705	60.073	0.000
WAISSPL		38.030	0.697	54.578	0.000
WRATCALC		38.267	0.666	57.464	0.000
WJCALC		23.603	0.465	50.781	0.000
WAISCALC		10.266	0.316	32.528	0.000
Variances					
MATH		42.833	6.496	6.594	0.000
SPELL		49.143	7.669	6.408	0.000
Residual Variances					
WRATSPL		5.307	1.105	4.804	0.000
WJSPL		3.474	0.910	3.818	0.000
WAISSPL		4.900	1.007	4.865	0.000
WRATCALC		1.957	1.621	1.208	0.227
WJCALC		3.960	0.848	4.672	0.000
WAISCALC		4.766	0.714	6.678	0.000

Scalar Invariance result ...

MODEL RESULTS FOR THE METRIC MODEL

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
Group MALE				
MATH BY				
WRATCALC	1.000	0.000	999.000	999.000
WJCALC	0.685	0.027	24.915	0.000
WAISCALC	0.393	0.024	16.236	0.000
SPELL BY				
WRATSPL	1.000	0.000	999.000	999.000
WJSPL	1.040	0.029	35.953	0.000
WAISSPL	0.973	0.028	34.330	0.000
SPELL WITH				
MATH	19.557	3.018	6.479	0.000
Means				
MATH	0.000	0.000	999.000	999.000
SPELL	0.000	0.000	999.000	999.000

Scalar Invariance result ...

Intercepts				
WRATSPL	36.172	0.454	79.673	0.000
WJSPL	41.371	0.472	87.637	0.000
WAISSPL	36.767	0.446	82.522	0.000
WRATCALC	39.222	0.422	92.879	0.000
WJCALC	23.910	0.302	79.191	0.000
WAISCALC	11.367	0.221	51.462	0.000
Variances				
MATH	34.791	3.738	9.307	0.000
SPELL	40.789	4.239	9.622	0.000
Residual Variances				
WRATSPL	4.764	0.721	6.607	0.000
WJSPL	5.154	0.790	6.526	0.000
WAISSPL	5.233	0.732	7.153	0.000
WRATCALC	4.620	1.205	3.834	0.000
WJCALC	3.835	0.654	5.864	0.000
WAISCALC	5.370	0.564	9.525	0.000
Group FEMALE				
MATH	BY			
WRATCALC		1.000	0.000	999.000
WJCALC		0.685	0.027	24.915
WAISCALC		0.393	0.024	16.236

Scalar Invariance result ...

SPELL BY					
WRATSPL	1.000	0.000	999.000	999.000	
WJSPL	1.040	0.029	35.953	0.000	
WAISSPL	0.973	0.028	34.330	0.000	
SPELL WITH					
MATH	27.515	5.266	5.225	0.000	
Means					
MATH	0.000	0.000	999.000	999.000	
SPELL	0.000	0.000	999.000	999.000	
Intercepts					
WRATSPL	37.172	0.713	52.138	0.000	
WJSPL	42.337	0.721	58.721	0.000	
WAISSPL	38.031	0.691	55.053	0.000	
WRATCALC	38.267	0.656	58.322	0.000	
WJCALC	23.602	0.474	49.834	0.000	
WAISCALC	10.263	0.330	31.086	0.000	
Variances					
MATH	40.435	6.233	6.487	0.000	
SPELL	45.576	6.856	6.648	0.000	
Residual Variances					

Scalar Invariance result ...

WRATSPL	5.686	1.124	5.061	0.000
WJSPL	3.230	0.929	3.478	0.001
WAISSPL	4.995	1.023	4.883	0.000
WRATCALC	3.048	1.413	2.157	0.031
WJCALC	3.603	0.778	4.631	0.000
WAISCALC	4.700	0.714	6.586	0.000

MODEL RESULTS FOR THE SCALAR MODEL

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
Group MALE				
MATH BY				
WRATCALC	1.000	0.000	999.000	999.000
WJCALC	0.680	0.027	24.897	0.000
WAISCALC	0.396	0.024	16.225	0.000
SPELL BY				
WRATSPL	1.000	0.000	999.000	999.000
WJSPL	1.039	0.029	36.045	0.000
WAISSPL	0.974	0.028	34.431	0.000
SPELL WITH				

Scalar Invariance result ...

MATH	19.579	3.020	6.482	0.000
Means				
MATH	0.000	0.000	999.000	999.000
SPELL	0.000	0.000	999.000	999.000
Intercepts				
WRATSPL	36.157	0.449	80.484	0.000
WJSPL	41.316	0.466	88.687	0.000
WAISSPL	36.843	0.440	83.720	0.000
WRATCALC	39.217	0.419	93.583	0.000
WJCALC	24.016	0.295	81.469	0.000
WAISCALC	11.125	0.207	53.742	0.000
Variances				
MATH	34.893	3.740	9.329	0.000
SPELL	40.764	4.236	9.624	0.000
Residual Variances				
WRATSPL	4.762	0.721	6.603	0.000
WJSPL	5.173	0.791	6.541	0.000
WAISSPL	5.235	0.734	7.134	0.000
WRATCALC	4.518	1.204	3.753	0.000
WJCALC	3.911	0.655	5.970	0.000
WAISCALC	5.431	0.576	9.435	0.000

Scalar Invariance result ...

Group FEMALE

MATH	BY				
WRATCALC		1.000	0.000	999.000	999.000
WJCALC		0.680	0.027	24.897	0.000
WAISCALC		0.396	0.024	16.225	0.000
SPELL	BY				
WRATSPL		1.000	0.000	999.000	999.000
WJSPL		1.039	0.029	36.045	0.000
WAISSPL		0.974	0.028	34.431	0.000
SPELL	WITH				
MATH		27.567	5.276	5.225	0.000
Means					
MATH		-0.944	0.771	-1.225	0.221
SPELL		1.054	0.811	1.298	0.194
Intercepts					
WRATSPL		36.157	0.449	80.484	0.000
WJSPL		41.316	0.466	88.687	0.000
WAISSPL		36.843	0.440	83.720	0.000
WRATCALC		39.217	0.419	93.583	0.000
WJCALC		24.016	0.295	81.469	0.000
WAISCALC		11.125	0.207	53.742	0.000

Scalar Invariance result ...

QUALITY OF NUMERICAL RESULTS FOR THE SCALAR MODEL

Condition Number for the Information Matrix
(ratio of smallest to largest eigenvalue)

0.342E-04

TECHNICAL 1 OUTPUT FOR THE CONFIGURAL MODEL

PARAMETER SPECIFICATION FOR MALE

```

      NU
      WRATSPL      WJSPL      WAISSPL      WRATCALC      WJCALC
1      -----      -----      -----      -----      -----
      1              2              3              4              5

```

```

      NU
      WAISCALC
1      -----
      6

```

```

      LAMBDA
      MATH      SPELL

```

Scalar Invariance result ...

WRATSPL	0	0
WJSPL	0	7
WAISSPL	0	8
WRATCALC	0	0
WJCALC	9	0
WAISCALC	10	0

THETA

	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
WRATSPL	11				
WJSPL	0	12			
WAISSPL	0	0	13		
WRATCALC	0	0	0	14	
WJCALC	0	0	0	0	15
WAISCALC	0	0	0	0	0

THETA

WAISCALC

WAISCALC	16
----------	----

Scalar Invariance result ...

ALPHA

	MATH	SPELL
1	----- 0	----- 0

BETA

	MATH	SPELL
MATH	----- 0	----- 0
SPELL	0	----- 0

PSI

	MATH	SPELL
MATH	----- 17	
SPELL	18	----- 19

PARAMETER SPECIFICATION FOR FEMALE

NU	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----

Scalar Invariance result ...

1 20 21 22 23 24

NU

WAISCALC

1 -----
 25

LAMBDA

MATH

SPELL

	-----	-----
WRATSPL	0	0
WJSPL	0	26
WAISSPL	0	27
WRATCALC	0	0
WJCALC	28	0
WAISCALC	29	0

THETA

WRATSPL

WJSPL

WAISSPL

WRATCALC

WJCALC

	-----	-----	-----	-----	-----
WRATSPL	30				
WJSPL	0	31			
WAISSPL	0	0	32		

Scalar Invariance result ...

WRATCALC	0	0	0	33	
WJCALC	0	0	0	0	34
WAISCALC	0	0	0	0	0

	THETA	
	WAISCALC	
WAISCALC	-----	
	35	

	ALPHA	
	MATH	SPELL
1	-----	-----
	0	0

	BETA	
	MATH	SPELL
MATH	-----	-----
	0	0
SPELL	0	0

	PSI	
	MATH	SPELL

Scalar Invariance result ...

MATH	-----	-----
	36	
SPELL	37	38

STARTING VALUES FOR MALE

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
1	36.172	41.371	36.786	39.222	23.910

	NU
	WAISCALC

1	11.350

	LAMBDA	
	MATH	SPELL
	-----	-----
WRATSPL	0.000	1.000
WJSPL	0.000	1.082
WAISSPL	0.000	0.984

Scalar Invariance result ...

WRATCALC	1.000	0.000
WJCALC	0.719	0.000
WAISCALC	0.421	0.000

THETA					
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	22.166				
WJSPL	0.000	25.479			
WAISSPL	0.000	0.000	21.761		
WRATCALC	0.000	0.000	0.000	19.254	
WJCALC	0.000	0.000	0.000	0.000	10.272
WAISCALC	0.000	0.000	0.000	0.000	0.000

THETA	
	WAISCALC

WAISCALC	5.605

ALPHA		
	MATH	SPELL
	-----	-----
1	0.000	0.000

Scalar Invariance result ...

	BETA	
	MATH	SPELL
MATH	0.000	0.000
SPELL	0.000	0.000

	PSI	
	MATH	SPELL
MATH	0.050	
SPELL	0.000	0.050

STARTING VALUES FOR FEMALE

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
1	36.172	41.371	36.786	39.222	23.910

NU

Scalar Invariance result ...

	WAISCALC
1	----- 11.350

	LAMBDA	
	MATH	SPELL
	-----	-----
WRATSPL	0.000	1.000
WJSPL	0.000	0.974
WAISSPL	0.000	0.949
WRATCALC	1.000	0.000
WJCALC	0.628	0.000
WAISCALC	0.341	0.000

	THETA				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	22.166				
WJSPL	0.000	25.479			
WAISSPL	0.000	0.000	21.761		
WRATCALC	0.000	0.000	0.000	19.254	
WJCALC	0.000	0.000	0.000	0.000	10.272
WAISCALC	0.000	0.000	0.000	0.000	0.000

Scalar Invariance result ...

THETA	
WAISCALC	
WAISCALC	5.605

ALPHA		
	MATH	SPELL
1	0.000	0.000

BETA		
	MATH	SPELL
MATH	0.000	0.000
SPELL	0.000	0.000

PSI		
	MATH	SPELL
MATH	0.050	
SPELL	0.000	0.050

Scalar Invariance result ...

TECHNICAL 1 OUTPUT FOR THE METRIC MODEL

PARAMETER SPECIFICATION FOR MALE

```

      NU
      WRATSPL      WJSPL      WAISSPL      WRATCALC      WJCALC
      -----      -----      -----      -----      -----
1      1            2            3            4            5

```

```

      NU
      WAISCALC
      -----
1      6

```

```

      LAMBDA
      MATH      SPELL
      -----      -----
WRATSPL      0            0
WJSPL        0            7
WAISSPL      0            8
WRATCALC     0            0

```

Scalar Invariance result ...

WJCALC	9	0
WAISCALC	10	0

	THETA				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	11				
WJSPL	0	12			
WAISSPL	0	0	13		
WRATCALC	0	0	0	14	
WJCALC	0	0	0	0	15
WAISCALC	0	0	0	0	0

	THETA
	WAISCALC

WAISCALC	16

	ALPHA	
	MATH	SPELL
	-----	-----
1	0	0

Scalar Invariance result ...

		BETA	
		MATH	SPELL
MATH		----- 0	----- 0
SPELL		0	0

		PSI	
		MATH	SPELL
MATH		----- 17	----- 19
SPELL		18	19

PARAMETER SPECIFICATION FOR FEMALE

		NU				
		WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
1		----- 20	----- 21	----- 22	----- 23	----- 24

		NU
		WAISCALC

Scalar Invariance result ...

1 -----
 25

	LAMBDA	
	MATH	SPELL
	-----	-----
WRATSPL	0	0
WJSPL	0	7
WAISSPL	0	8
WRATCALC	0	0
WJCALC	9	0
WAISCALC	10	0

	THETA				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	26				
WJSPL	0	27			
WAISSPL	0	0	28		
WRATCALC	0	0	0	29	
WJCALC	0	0	0	0	30
WAISCALC	0	0	0	0	0

Scalar Invariance result ...

	THETA	
	WAISCALC	

WAISCALC		31

	ALPHA		
	MATH		SPELL
	-----		-----
1		0	0

	BETA		
	MATH		SPELL
	-----		-----
MATH		0	0
SPELL		0	0

	PSI		
	MATH		SPELL
	-----		-----
MATH		32	
SPELL		33	34

Scalar Invariance result ...

STARTING VALUES FOR MALE

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
1	36.172	41.371	36.786	39.222	23.910

	NU
	WAISCALC
1	11.350

	LAMBDA	
	MATH	SPELL
WRATSPL	0.000	1.000
WJSPL	0.000	1.082
WAISSPL	0.000	0.984
WRATCALC	1.000	0.000
WJCALC	0.719	0.000
WAISCALC	0.421	0.000

Scalar Invariance result ...

THETA					
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	22.166				
WJSPL	0.000	25.479			
WAISSPL	0.000	0.000	21.761		
WRATCALC	0.000	0.000	0.000	19.254	
WJCALC	0.000	0.000	0.000	0.000	10.272
WAISCALC	0.000	0.000	0.000	0.000	0.000

THETA	
	WAISCALC

WAISCALC	5.605

ALPHA		
	MATH	SPELL
	-----	-----
1	0.000	0.000

BETA	
	MATH

Scalar Invariance result ...

MATH	0.000	0.000
SPELL	0.000	0.000

	PSI	
	MATH	SPELL
	-----	-----
MATH	0.050	
SPELL	0.000	0.050

STARTING VALUES FOR FEMALE

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
1	36.172	41.371	36.786	39.222	23.910

	NU
	WAISCALC

1	11.350

Scalar Invariance result ...

	LAMBDA	
	MATH	SPELL
WRATSPL	0.000	1.000
WJSPL	0.000	0.974
WAISSPL	0.000	0.949
WRATCALC	1.000	0.000
WJCALC	0.628	0.000
WAISCALC	0.341	0.000

	THETA				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
WRATSPL	22.166				
WJSPL	0.000	25.479			
WAISSPL	0.000	0.000	21.761		
WRATCALC	0.000	0.000	0.000	19.254	
WJCALC	0.000	0.000	0.000	0.000	10.272
WAISCALC	0.000	0.000	0.000	0.000	0.000

	THETA
	WAISCALC
WAISCALC	5.605

Scalar Invariance result ...

ALPHA	
	MATH SPELL
1	----- 0.000 0.000

BETA	
	MATH SPELL
MATH	----- 0.000 0.000
SPELL	0.000 0.000

PSI	
	MATH SPELL
MATH	----- 0.050 0.050
SPELL	0.000 0.050

TECHNICAL 1 OUTPUT FOR THE SCALAR MODEL

PARAMETER SPECIFICATION FOR MALE

Scalar Invariance result ...

NU		WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
		-----	-----	-----	-----	-----
1		1	2	3	4	5
NU		WAISCALC				

1		6				
LAMBDA		MATH	SPELL			
		-----	-----			
WRATSPL		0	0			
WJSPL		0	7			
WAISSPL		0	8			
WRATCALC		0	0			
WJCALC		9	0			
WAISCALC		10	0			
THETA						

Scalar Invariance result ...

	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
WRATSPL	11				
WJSPL	0	12			
WAISSPL	0	0	13		
WRATCALC	0	0	0	14	
WJCALC	0	0	0	0	15
WAISCALC	0	0	0	0	0

THETA

WAISCALC

WAISCALC

16

ALPHA

MATH

SPELL

1

0

0

BETA

MATH

SPELL

MATH

0

0

Scalar Invariance result ...

SPELL	0	0
	PSI	
	MATH	SPELL
	-----	-----
MATH	17	
SPELL	18	19

PARAMETER SPECIFICATION FOR FEMALE

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
1	1	2	3	4	5

	NU
	WAISCALC

1	6

LAMBDA

Scalar Invariance result ...

	MATH	SPELL
WRATSPL	0	0
WJSPL	0	7
WAISSPL	0	8
WRATCALC	0	0
WJCALC	9	0
WAISCALC	10	0

	THETA				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
WRATSPL	20				
WJSPL	0	21			
WAISSPL	0	0	22		
WRATCALC	0	0	0	23	
WJCALC	0	0	0	0	24
WAISCALC	0	0	0	0	0

	THETA
	WAISCALC
WAISCALC	25

Scalar Invariance result ...

ALPHA

	MATH	SPELL
1	26	27

BETA

	MATH	SPELL
MATH	0	0
SPELL	0	0

PSI

	MATH	SPELL
MATH	28	
SPELL	29	30

STARTING VALUES FOR MALE

NU	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
----	---------	-------	---------	----------	--------

Scalar Invariance result ...

1	-----	-----	-----	-----	-----
	36.172	41.371	36.786	39.222	23.910
	NU				
	WAISCALC				
1	-----				
	11.350				
	LAMBDA				
	MATH	SPELL			
	-----	-----			
WRATSPL	0.000	1.000			
WJSPL	0.000	1.082			
WAISSPL	0.000	0.984			
WRATCALC	1.000	0.000			
WJCALC	0.719	0.000			
WAISCALC	0.421	0.000			
	THETA				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	22.166				
WJSPL	0.000	25.479			

Scalar Invariance result ...

WAISSPL	0.000	0.000	21.761		
WRATCALC	0.000	0.000	0.000	19.254	
WJCALC	0.000	0.000	0.000	0.000	10.272
WAISCALC	0.000	0.000	0.000	0.000	0.000

THETA

WAISCALC

WAISCALC -----
 5.605

ALPHA

MATH

1

0.000

SPELL

0.000

BETA

MATH

MATH -----
 0.000

SPELL -----
 0.000

SPELL

0.000

0.000

PSI

Scalar Invariance result ...

	MATH	SPELL
MATH	0.050	
SPELL	0.000	0.050

STARTING VALUES FOR FEMALE

NU	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
1	36.172	41.371	36.786	39.222	23.910

NU	WAISCALC
1	11.350

LAMBDA	MATH	SPELL
WRATSPL	0.000	1.000
WJSPL	0.000	0.974

Scalar Invariance result ...

WAISSPL	0.000	0.949
WRATCALC	1.000	0.000
WJCALC	0.628	0.000
WAISCALC	0.341	0.000

THETA					
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	22.166				
WJSPL	0.000	25.479			
WAISSPL	0.000	0.000	21.761		
WRATCALC	0.000	0.000	0.000	19.254	
WJCALC	0.000	0.000	0.000	0.000	10.272
WAISCALC	0.000	0.000	0.000	0.000	0.000

THETA	
	WAISCALC

WAISCALC	5.605

ALPHA	
	MATH

	SPELL

Scalar Invariance result ...

1 0.000 0.000

BETA

MATH SPELL

	-----	-----
MATH	0.000	0.000
SPELL	0.000	0.000

PSI

MATH SPELL

	-----	-----
MATH	0.050	
SPELL	0.000	0.050

TECHNICAL 9 OUTPUT

Error messages for the Configural Model:

THE MODEL ESTIMATION TERMINATED NORMALLY

Error messages for the Metric Model:

THE MODEL ESTIMATION TERMINATED NORMALLY

Scalar Invariance result ...

Error messages for the Scalar Model:

THE MODEL ESTIMATION TERMINATED NORMALLY

Beginning Time: 21:42:53
Ending Time: 21:42:53
Elapsed Time: 00:00:00

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Outline

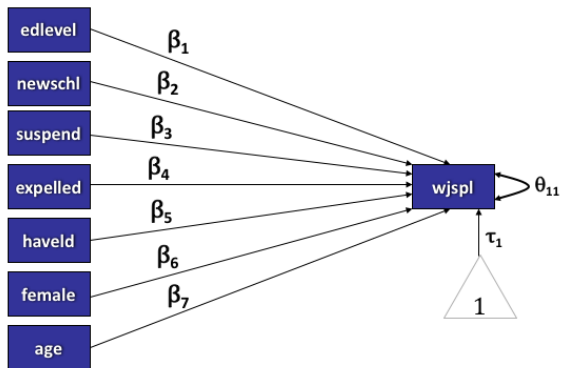
- 1 Example 1: Job Data EFA
- 2 Example 2: CFA
- 3 Example 3: Two Group CFA
- 4 Example 4: Multiple Regression**
- 5 Example 5: SEM
- 6 Example: Insomnia SEM
- 7 Example: Growth Curve
- 8 Conclusion

Incorporate predictors

- Mplus does same ordinary regression model as other programs might do.
- We can start from common building blocks as regression, and then add features
- Predict WJ spelling scores (wjspl) using predictor variables

Multiple Regression Diagram

Multiple Regression Diagram ...



Mplus Regression Syntax

- Note the difference is in the "ON" usage of the MODEL stanza

```
TITLE:  
  Example 4 - Multiple Linear Regression  
DATA:  
  FILE IS "../..//data/job_placement.csv";  
VARIABLE:  
  NAMES ARE  
  id wjcalc wjspl wratspl wratcalc waiscalc waisspl  
  edlevel newschl suspend expelled haveld female  
  age;  
  USEVARIABLES ARE  
  wjspl edlevel newschl suspend expelled haveld  
  female age;  
  MISSING ARE all(99999);  
MODEL:
```

Mplus Regression Syntax ...

```
wjspl ON edlevel newschl suspend expelled haveld  
      female age;
```

Regression Results

View Online: [reg-01.out](#)

Inline Display:

```
Mplus VERSION 7.4 (Linux)
MUTHEN & MUTHEN
05/28/2018  9:45 PM
```

INPUT INSTRUCTIONS

TITLE:

```
Example 4 - Multiple Linear Regression
```

DATA:

```
FILE IS "../..data/job_placement.csv";
```

VARIABLE:

NAMES ARE

```
id wjcalc wjspl wratspl wratcalc waiscalc waisspl
edlevel newschl suspend expelled haveld female age;
```

USEVARIABLES ARE

```
wjspl edlevel newschl suspend expelled haveld female age;
```


Regression Results ...

```
MISSING ARE all(99999);
```

```
MODEL:
```

```
wjspl ON edlevel newschl suspend expelled haveld female age;
```

```
*** WARNING
```

```
Data set contains cases with missing on x-variables.
```

```
These cases were not included in the analysis.
```

```
Number of cases with missing on x-variables: 9
```

```
1 WARNING(S) FOUND IN THE INPUT INSTRUCTIONS
```

Example 4 - Multiple Linear Regression

SUMMARY OF ANALYSIS

Number of groups	1
Number of observations	313
Number of dependent variables	1
Number of independent variables	7
Number of continuous latent variables	0

Regression Results ...

Observed dependent variables

Continuous

WJSPL

Observed independent variables

EDLEVEL	NEWSCHL	SUSPEND	EXPELLED	HAVELD	FEMALE
AGE					

Estimator	ML
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/job_placement.csv

Input data format FREE

SUMMARY OF DATA

Regression Results ...

Number of missing data patterns 1

COVARIANCE COVERAGE OF DATA

Minimum covariance coverage value 0.100

PROPORTION OF DATA PRESENT

	Covariance Coverage				
	WJSPL	EDLEVEL	NEWSCHL	SUSPEND	EXPELLED
	-----	-----	-----	-----	-----
WJSPL	1.000				
EDLEVEL	1.000	1.000			
NEWSCHL	1.000	1.000	1.000		
SUSPEND	1.000	1.000	1.000	1.000	
EXPELLED	1.000	1.000	1.000	1.000	1.000
HAVELD	1.000	1.000	1.000	1.000	1.000
FEMALE	1.000	1.000	1.000	1.000	1.000
AGE	1.000	1.000	1.000	1.000	1.000

Covariance Coverage

Regression Results ...

	HAVELD	FEMALE	AGE
	-----	-----	-----
HAVELD	1.000		
FEMALE	1.000	1.000	
AGE	1.000	1.000	1.000

THE MODEL ESTIMATION TERMINATED NORMALLY

MODEL FIT INFORMATION

Number of Free Parameters 9

Loglikelihood

H0 Value -1021.060

H1 Value -1021.060

Information Criteria

Akaike (AIC) 2060.120

Bayesian (BIC) 2093.836

Sample-Size Adjusted BIC 2065.291

Regression Results ...

$$(n^* = (n + 2) / 24)$$

Chi-Square Test of Model Fit

Value	0.000
Degrees of Freedom	0
P-Value	0.0000

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.000	
90 Percent C.I.	0.000	0.000
Probability RMSEA <= .05	0.000	

CFI/TLI

CFI	1.000
TLI	1.000

Chi-Square Test of Model Fit for the Baseline Model

Value	78.031
Degrees of Freedom	7
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Regression Results ...

Value 0.000

MODEL RESULTS

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
WJSPL ON				
EDLEVEL	1.162	0.324	3.584	0.000
NEWSCHL	0.063	0.747	0.085	0.932
SUSPEND	-0.052	0.773	-0.067	0.947
EXPELLED	-2.758	1.098	-2.510	0.012
HAVELD	-6.974	0.988	-7.063	0.000
FEMALE	0.720	0.792	0.909	0.363
AGE	0.412	0.207	1.994	0.046
Intercepts				
WJSPL	21.951	4.414	4.973	0.000
Residual Variances				
WJSPL	39.905	3.190	12.510	0.000

Regression Results ...

QUALITY OF NUMERICAL RESULTS

Condition Number for the Information Matrix 0.257E-03
(ratio of smallest to largest eigenvalue)

Beginning Time: 21:45:53
Ending Time: 21:45:53
Elapsed Time: 00:00:00

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Outline

- 1 Example 1: Job Data EFA
- 2 Example 2: CFA
- 3 Example 3: Two Group CFA
- 4 Example 4: Multiple Regression
- 5 Example 5: SEM**
- 6 Example: Insomnia SEM
- 7 Example: Growth Curve
- 8 Conclusion

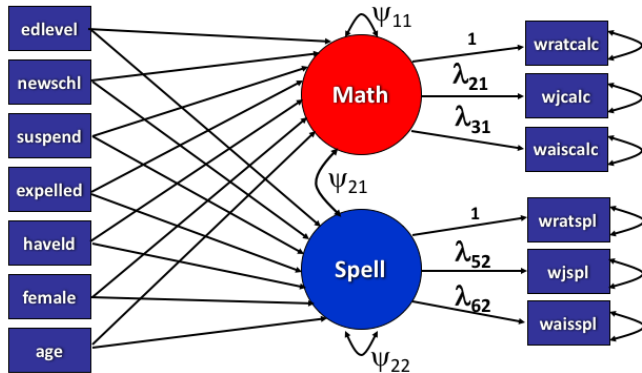
SEM = Structural Equation Model

- SEM Combines CFA and Regression Concepts
 - A latent variable can be a predictor or an outcome variable
- Observed variables can be either
 - indicators (used to identify the latent constructs)
 - exogenous predictors or outcomes that are treated as “correctly measured” variables (e.g., no measurement model)

SEM terminology

- Originally dubbed LISREL (linear structural relations) by creator Karl Joreskog
- Two components
 - 1 A measurement model, the CFA we did previously
 - 2 A regression component that treats latent variables as predictors or outcomes

SEM Diagram



SEM Syntax

```
TITLE:
Example 5 - Structural Equation Model;

DATA:
5   FILE IS "../..//data/job_placement.csv";
VARIABLE:
   NAMES ARE
   id wjcalc wjspl wratspl wratcalc waiscalc waisspl
   edlevel newschl suspend expelled haveld female
   age;
10  USEVARIABLES ARE
   wratspl wjspl waisspl wratcalc wjcalc waiscalc
   edlevel newschl suspend expelled haveld female
   age;
   MISSING ARE all(99999);
MODEL:
```

SEM Syntax ...

```
15 MATH BY wratcalc wjcalc waiscalc;  
SPELL BY wratspl wjspl waisspl;  
MATH ON edlevel newschl suspend expelled haveld  
      female age;  
SPELL ON edlevel newschl suspend expelled haveld  
      female age;  
MATH WITH SPELL;
```

```
20 OUTPUT:
```

```
SAMPSTAT;  
TECH1;  
STDYX;
```

SEM Results

View Online: [sem-01.out](#)

Inline Display:

```
Mplus VERSION 7.4 (Linux)
MUTHEN & MUTHEN
05/28/2018  9:48 PM
```

INPUT INSTRUCTIONS

TITLE:

```
Example 5 - Structural Equation Model
```

DATA:

```
FILE IS "../..data/job_placement.csv";
```

VARIABLE:

NAMES ARE

```
id wjcalc wjspl wratspl wratcalc waiscalc waisspl
edlevel newschl suspend expelled haveld female age;
```

USEVARIABLES ARE

```
wratspl wjspl waisspl wratcalc wjcalc waiscalc
edlevel newschl suspend expelled haveld female age;
```

SEM Results ...

```
MISSING ARE all(99999);
```

```
MODEL:
```

```
  MATH BY wratcalc wjcalc waiscalc;
```

```
  SPELL BY wratspl wjspl waisspl;
```

```
  MATH ON edlevel newschl suspend expelled haveld female age;
```

```
  SPELL ON edlevel newschl suspend expelled haveld female age;
```

```
  MATH WITH SPELL;
```

```
OUTPUT:
```

```
  SAMPSTAT;
```

```
  TECH1;
```

```
  STDYX;
```

```
*** WARNING
```

```
  Data set contains cases with missing on x-variables.
```

```
  These cases were not included in the analysis.
```

```
  Number of cases with missing on x-variables: 9
```

```
  1 WARNING(S) FOUND IN THE INPUT INSTRUCTIONS
```

SEM Results ...

Example 5 - Structural Equation Model

SUMMARY OF ANALYSIS

Number of groups	1
Number of observations	313
Number of dependent variables	6
Number of independent variables	7
Number of continuous latent variables	2

Observed dependent variables

Continuous

WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC	WAISSCALC
---------	-------	---------	----------	--------	-----------

Observed independent variables

EDLEVEL	NEWSCHL	SUSPEND	EXPELLED	HAVELD	FEMALE
AGE					

Continuous latent variables

MATH	SPELL
------	-------

Estimator

ML

Information matrix

OBSERVED

SEM Results ...

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/job_placement.csv

Input data format FREE

SUMMARY OF DATA

Number of missing data patterns	4
---------------------------------	---

COVARIANCE COVERAGE OF DATA

Minimum covariance coverage value	0.100
-----------------------------------	-------

PROPORTION OF DATA PRESENT

Covariance Coverage

SEM Results ...

	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	0.997				
WJSPL	0.997	1.000			
WAISSPL	0.990	0.994	0.994		
WRATCALC	0.997	1.000	0.994	1.000	
WJCALC	0.990	0.994	0.987	0.994	0.994
WAISCALC	0.990	0.994	0.994	0.994	0.987
EDLEVEL	0.997	1.000	0.994	1.000	0.994
NEWSCHL	0.997	1.000	0.994	1.000	0.994
SUSPEND	0.997	1.000	0.994	1.000	0.994
EXPELLED	0.997	1.000	0.994	1.000	0.994
HAVELD	0.997	1.000	0.994	1.000	0.994
FEMALE	0.997	1.000	0.994	1.000	0.994
AGE	0.997	1.000	0.994	1.000	0.994

Covariance Coverage

	WAISCALC	EDLEVEL	NEWSCHL	SUSPEND	EXPELLED
	-----	-----	-----	-----	-----
WAISCALC	0.994				
EDLEVEL	0.994	1.000			
NEWSCHL	0.994	1.000	1.000		
SUSPEND	0.994	1.000	1.000	1.000	
EXPELLED	0.994	1.000	1.000	1.000	1.000
HAVELD	0.994	1.000	1.000	1.000	1.000

SEM Results ...

FEMALE	0.994	1.000	1.000	1.000	1.000
AGE	0.994	1.000	1.000	1.000	1.000

	Covariance Coverage		
	HAVELD	FEMALE	AGE
	-----	-----	-----
HAVELD	1.000		
FEMALE	1.000	1.000	
AGE	1.000	1.000	1.000

SAMPLE STATISTICS

ESTIMATED SAMPLE STATISTICS

	Means				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
1	36.575	41.789	37.310	38.930	23.844

	Means				
	WAISCALC	EDLEVEL	NEWSCHL	SUSPEND	EXPELLED

SEM Results ...

1	----- 11.022	----- 11.153	----- 0.546	----- 0.518	----- 0.141
	Means				
	HAVELD	FEMALE	AGE		
1	----- 0.157	----- 0.316	----- 19.719		
	Covariances				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
WRATSPL	----- 48.101				
WJSPL	44.763	----- 51.202			
WAISSPL	41.698	43.272	----- 45.292		
WRATCALC	22.641	22.816	21.201	----- 41.292	
WJCALC	14.816	15.467	14.420	25.536	----- 21.338
WAISSCALC	10.784	11.343	10.170	14.964	10.276
EDLEVEL	2.139	2.182	2.081	3.190	2.390
NEWSCHL	-0.135	-0.003	-0.072	-0.022	-0.033
SUSPEND	-0.355	-0.277	-0.239	-0.689	-0.374
EXPELLED	-0.356	-0.370	-0.353	-0.201	-0.119
HAVELD	-0.854	-0.932	-0.868	-0.177	-0.168
FEMALE	0.198	0.191	0.236	-0.185	-0.059
AGE	2.638	2.650	2.021	4.051	2.962

SEM Results ...

	Covariances				
	WAISCALC	EDLEVEL	NEWSCHL	SUSPEND	EXPELLED
	-----	-----	-----	-----	-----
WAISCALC	11.278				
EDLEVEL	1.166	1.472			
NEWSCHL	0.064	-0.026	0.248		
SUSPEND	-0.031	-0.057	0.046	0.250	
EXPELLED	-0.006	-0.031	0.035	0.052	0.121
HAVELD	-0.131	-0.008	-0.015	-0.004	-0.003
FEMALE	-0.237	-0.064	0.019	-0.033	-0.019
AGE	1.404	0.912	-0.022	-0.158	-0.015

	Covariances		
	HAVELD	FEMALE	AGE
	-----	-----	-----
HAVELD	0.132		
FEMALE	-0.008	0.216	
AGE	-0.007	-0.001	3.627

	Correlations				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----

SEM Results ...

WRATSPL	1.000				
WJSPL	0.902	1.000			
WAISSPL	0.893	0.899	1.000		
WRATCALC	0.508	0.496	0.490	1.000	
WJCALC	0.462	0.468	0.464	0.860	1.000
WAISCALC	0.463	0.472	0.450	0.693	0.662
EDLEVEL	0.254	0.251	0.255	0.409	0.427
NEWSCHL	-0.039	-0.001	-0.022	-0.007	-0.014
SUSPEND	-0.102	-0.078	-0.071	-0.215	-0.162
EXPELLED	-0.148	-0.149	-0.151	-0.090	-0.074
HAVELD	-0.339	-0.358	-0.355	-0.076	-0.100
FEMALE	0.061	0.057	0.075	-0.062	-0.027
AGE	0.200	0.194	0.158	0.331	0.337

	Correlations				
	WAISCALC	EDLEVEL	NEWSCHL	SUSPEND	EXPELLED
	-----	-----	-----	-----	-----
WAISCALC	1.000				
EDLEVEL	0.286	1.000			
NEWSCHL	0.038	-0.044	1.000		
SUSPEND	-0.018	-0.094	0.186	1.000	
EXPELLED	-0.005	-0.074	0.202	0.298	1.000
HAVELD	-0.108	-0.018	-0.084	-0.024	-0.022
FEMALE	-0.152	-0.114	0.082	-0.141	-0.117
AGE	0.219	0.395	-0.023	-0.166	-0.022

SEM Results ...

	Correlations		
	HAVELD	FEMALE	AGE
	-----	-----	-----
HAVELD	1.000		
FEMALE	-0.047	1.000	
AGE	-0.010	-0.001	1.000

MAXIMUM LOG-LIKELIHOOD VALUE FOR THE UNRESTRICTED (H1) MODEL IS -6866.105

UNIVARIATE SAMPLE STATISTICS

UNIVARIATE HIGHER-ORDER MOMENT DESCRIPTIVE STATISTICS

Variable/ Sample Size	Mean/ Variance	Skewness/ Kurtosis	Minimum/ Maximum	% with Min/Max	20%/60%	Percentiles 40%/80%	M
WRATSPL	36.574	-0.048	19.000	0.32%	30.000	35.000	3
312.000	48.232	-0.499	57.000	0.32%	38.000	43.000	
WJSPL	41.789	-0.209	19.000	0.32%	36.000	40.000	4
313.000	51.202	-0.290	57.000	0.64%	44.000	48.000	
WAISSPL	37.344	-0.438	16.000	0.32%	31.000	36.000	

SEM Results ...

	311.000	45.345	-0.128	52.000	0.32%	39.000	43.000	
WRATCALC		38.930	0.191	23.000	0.32%	33.000	37.000	3
	313.000	41.292	-0.593	54.000	0.32%	40.000	45.000	
WJCALC		23.855	0.085	13.000	0.64%	20.000	23.000	2
	311.000	21.301	-0.521	36.000	0.32%	25.000	28.000	
WAISCALC		11.019	0.749	6.000	1.29%	8.000	9.000	1
	311.000	11.241	-0.146	21.000	0.64%	11.000	14.000	
EDLEVEL		11.153	-0.639	8.000	1.60%	10.000	11.000	1
	313.000	1.472	-0.525	13.000	7.35%	12.000	12.000	
NEWSCHL		0.546	-0.186	0.000	45.37%	0.000	0.000	
	313.000	0.248	-1.965	1.000	54.63%	1.000	1.000	
SUSPEND		0.518	-0.070	0.000	48.24%	0.000	0.000	
	313.000	0.250	-1.995	1.000	51.76%	1.000	1.000	
EXPELLED		0.141	2.068	0.000	85.94%	0.000	0.000	
	313.000	0.121	2.277	1.000	14.06%	0.000	0.000	
HAVELD		0.157	1.890	0.000	84.35%	0.000	0.000	
	313.000	0.132	1.573	1.000	15.65%	0.000	0.000	
FEMALE		0.316	0.790	0.000	68.37%	0.000	0.000	
	313.000	0.216	-1.376	1.000	31.63%	0.000	1.000	
AGE		19.719	0.802	16.000	0.96%	18.000	19.000	1
	313.000	3.627	0.530	27.000	0.32%	20.000	21.000	

THE MODEL ESTIMATION TERMINATED NORMALLY

SEM Results ...

MODEL FIT INFORMATION

Number of Free Parameters	33
---------------------------	----

Loglikelihood

H0 Value	-4909.694
H1 Value	-4884.805

Information Criteria

Akaike (AIC)	9885.389
Bayesian (BIC)	10009.014
Sample-Size Adjusted BIC	9904.348
$(n* = (n + 2) / 24)$	

Chi-Square Test of Model Fit

Value	49.780
Degrees of Freedom	36
P-Value	0.0630

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.035
----------	-------

SEM Results ...

90 Percent C.I.	0.000	0.057
Probability RMSEA <= .05	0.859	

CFI/TLI

CFI	0.993
TLI	0.989

Chi-Square Test of Model Fit for the Baseline Model

Value	2035.834
Degrees of Freedom	57
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value	0.024
-------	-------

MODEL RESULTS

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

SEM Results ...

WRATCALC	1.000	0.000	999.000	999.000
WJCALC	0.684	0.027	25.116	0.000
WAISCALC	0.403	0.024	16.552	0.000
SPELL BY				
WRATSPL	1.000	0.000	999.000	999.000
WJSPL	1.038	0.029	35.933	0.000
WAISSPL	0.967	0.028	34.549	0.000
MATH ON				
EDLEVEL	1.720	0.282	6.097	0.000
NEWSCHL	0.535	0.649	0.824	0.410
SUSPEND	-1.727	0.677	-2.550	0.011
EXPELLED	-0.524	0.954	-0.549	0.583
HAVELD	-1.495	0.858	-1.743	0.081
FEMALE	-0.757	0.689	-1.098	0.272
AGE	0.617	0.179	3.441	0.001
SPELL ON				
EDLEVEL	1.169	0.302	3.874	0.000
NEWSCHL	-0.212	0.694	-0.305	0.760
SUSPEND	-0.111	0.718	-0.155	0.877
EXPELLED	-2.540	1.021	-2.488	0.013
HAVELD	-6.630	0.923	-7.182	0.000
FEMALE	0.843	0.736	1.145	0.252
AGE	0.349	0.192	1.819	0.069

SEM Results ...

MATH	WITH				
SPELL		15.401	2.048	7.520	0.000
Intercepts					
WRATSPL		17.949	4.113	4.364	0.000
WJSPL		22.456	4.267	5.263	0.000
WAISSPL		19.289	3.978	4.849	0.000
WRATCALC		8.718	3.847	2.266	0.023
WJCALC		3.169	2.710	1.169	0.242
WAISCALC		-1.141	1.686	-0.677	0.498
Residual Variances					
WRATSPL		4.985	0.612	8.151	0.000
WJSPL		4.751	0.626	7.587	0.000
WAISSPL		4.960	0.591	8.385	0.000
WRATCALC		4.035	0.966	4.175	0.000
WJCALC		3.876	0.528	7.335	0.000
WAISCALC		5.221	0.457	11.426	0.000
MATH		27.561	2.600	10.601	0.000
SPELL		32.786	2.961	11.074	0.000

STANDARDIZED MODEL RESULTS

SEM Results ...

STDYX Standardization

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
MATH BY				
WRATCALC	0.950	0.013	73.867	0.000
WJCALC	0.905	0.015	59.784	0.000
WAISCALC	0.732	0.029	25.661	0.000
SPELL BY				
WRATSPL	0.947	0.008	121.996	0.000
WJSPL	0.952	0.007	130.144	0.000
WAISSPL	0.944	0.008	117.377	0.000
MATH ON				
EDLEVEL	0.342	0.054	6.381	0.000
NEWSCHL	0.044	0.053	0.825	0.409
SUSPEND	-0.141	0.055	-2.576	0.010
EXPELLED	-0.030	0.054	-0.550	0.583
HAVELD	-0.089	0.051	-1.746	0.081
FEMALE	-0.058	0.052	-1.100	0.271
AGE	0.193	0.055	3.487	0.000
SPELL ON				
EDLEVEL	0.216	0.055	3.942	0.000

SEM Results ...

NEWSCHL	-0.016	0.053	-0.305	0.760
SUSPEND	-0.008	0.055	-0.155	0.877
EXPELLED	-0.134	0.054	-2.505	0.012
HAVELD	-0.367	0.048	-7.689	0.000
FEMALE	0.060	0.052	1.147	0.251
AGE	0.101	0.055	1.826	0.068
MATH WITH				
SPELL	0.512	0.046	11.178	0.000
Intercepts				
WRATSPL	2.588	0.630	4.106	0.000
WJSPL	3.138	0.643	4.879	0.000
WAISSPL	2.866	0.633	4.525	0.000
WRATCALC	1.357	0.627	2.165	0.030
WJCALC	0.686	0.601	1.143	0.253
WAISCALC	-0.340	0.496	-0.686	0.493
Residual Variances				
WRATSPL	0.104	0.015	7.052	0.000
WJSPL	0.093	0.014	6.656	0.000
WAISSPL	0.109	0.015	7.214	0.000
WRATCALC	0.098	0.024	3.999	0.000
WJCALC	0.182	0.027	6.641	0.000
WAISCALC	0.464	0.042	11.088	0.000
MATH	0.740	0.045	16.551	0.000

SEM Results ...

SPELL	0.760	0.043	17.632	0.000
-------	-------	-------	--------	-------

R-SQUARE

Observed Variable	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
WRATSPL	0.896	0.015	60.998	0.000
WJSPL	0.907	0.014	65.072	0.000
WAISSPL	0.891	0.015	58.689	0.000
WRATCALC	0.902	0.024	36.933	0.000
WJCALC	0.818	0.027	29.892	0.000
WAISCALC	0.536	0.042	12.830	0.000

Latent Variable	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
MATH	0.260	0.045	5.823	0.000
SPELL	0.240	0.043	5.556	0.000

QUALITY OF NUMERICAL RESULTS

Condition Number for the Information Matrix (ratio of smallest to largest eigenvalue)	0.247E-05
--	-----------

SEM Results ...

TECHNICAL 1 OUTPUT

PARAMETER SPECIFICATION

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
1	1	2	3	4	5
	NU				
	WAISSCALC	EDLEVEL	NEWSCHL	SUSPEND	EXPELLED
	-----	-----	-----	-----	-----
1	6	0	0	0	0
	NU				
	HAVELD	FEMALE	AGE		
	-----	-----	-----		
1	0	0	0		

SEM Results ...

	LAMBDA				
	MATH	SPELL	EDLEVEL	NEWSCHL	SUSPEND
	-----	-----	-----	-----	-----
WRATSPL	0	0	0	0	0
WJSPL	0	7	0	0	0
WAISSPL	0	8	0	0	0
WRATCALC	0	0	0	0	0
WJCALC	9	0	0	0	0
WAISSCALC	10	0	0	0	0
EDLEVEL	0	0	0	0	0
NEWSCHL	0	0	0	0	0
SUSPEND	0	0	0	0	0
EXPELLED	0	0	0	0	0
HAVELD	0	0	0	0	0
FEMALE	0	0	0	0	0
AGE	0	0	0	0	0

	LAMBDA			
	EXPELLED	HAVELD	FEMALE	AGE
	-----	-----	-----	-----
WRATSPL	0	0	0	0
WJSPL	0	0	0	0
WAISSPL	0	0	0	0
WRATCALC	0	0	0	0
WJCALC	0	0	0	0

SEM Results ...

WAISCALC	0	0	0	0
EDLEVEL	0	0	0	0
NEWSCHL	0	0	0	0
SUSPEND	0	0	0	0
EXPELLED	0	0	0	0
HAVELD	0	0	0	0
FEMALE	0	0	0	0
AGE	0	0	0	0

	THETA				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
WRATSPL	11				
WJSPL	0	12			
WAISSPL	0	0	13		
WRATCALC	0	0	0	14	
WJCALC	0	0	0	0	15
WAISCALC	0	0	0	0	0
EDLEVEL	0	0	0	0	0
NEWSCHL	0	0	0	0	0
SUSPEND	0	0	0	0	0
EXPELLED	0	0	0	0	0
HAVELD	0	0	0	0	0
FEMALE	0	0	0	0	0
AGE	0	0	0	0	0

SEM Results ...

THETA					
	WAISCALC	EDLEVEL	NEWSCHL	SUSPEND	EXPELLED
	-----	-----	-----	-----	-----
WAISCALC	16				
EDLEVEL	0	0			
NEWSCHL	0	0	0		
SUSPEND	0	0	0	0	
EXPELLED	0	0	0	0	0
HAVELD	0	0	0	0	0
FEMALE	0	0	0	0	0
AGE	0	0	0	0	0

THETA			
	HAVELD	FEMALE	AGE
	-----	-----	-----
HAVELD	0		
FEMALE	0	0	
AGE	0	0	0

ALPHA					
	MATH	SPELL	EDLEVEL	NEWSCHL	SUSPEND
	-----	-----	-----	-----	-----

SEM Results ...

1 0 0 0 0 0

ALPHA

EXPELLED

HAVELD

FEMALE

AGE

1

0

0

0

0

BETA

MATH

SPELL

EDLEVEL

NEWSCHL

SUSPEND

MATH

0

0

17

18

19

SPELL

0

0

24

25

26

EDLEVEL

0

0

0

0

0

NEWSCHL

0

0

0

0

0

SUSPEND

0

0

0

0

0

EXPELLED

0

0

0

0

0

HAVELD

0

0

0

0

0

FEMALE

0

0

0

0

0

AGE

0

0

0

0

0

BETA

EXPELLED

HAVELD

FEMALE

AGE

SEM Results ...

MATH	20	21	22	23
SPELL	27	28	29	30
EDLEVEL	0	0	0	0
NEWSCHL	0	0	0	0
SUSPEND	0	0	0	0
EXPELLED	0	0	0	0
HAVELD	0	0	0	0
FEMALE	0	0	0	0
AGE	0	0	0	0

PSI

	MATH	SPELL	EDLEVEL	NEWSCHL	SUSPEND
MATH	31				
SPELL	32	33			
EDLEVEL	0	0	0		
NEWSCHL	0	0	0	0	
SUSPEND	0	0	0	0	0
EXPELLED	0	0	0	0	0
HAVELD	0	0	0	0	0
FEMALE	0	0	0	0	0
AGE	0	0	0	0	0

PSI

SEM Results ...

	EXPELLED	HAVELD	FEMALE	AGE
	-----	-----	-----	-----
EXPELLED	0			
HAVELD	0	0		
FEMALE	0	0	0	
AGE	0	0	0	0

STARTING VALUES

	NU				
	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
	-----	-----	-----	-----	-----
1	36.574	41.789	37.344	38.930	23.855

	NU				
	WAISCALC	EDLEVEL	NEWSCHL	SUSPEND	EXPELLED
	-----	-----	-----	-----	-----
1	11.019	0.000	0.000	0.000	0.000

	NU		
	HAVELD	FEMALE	AGE
	-----	-----	-----

SEM Results ...

1 0.000 0.000 0.000

	LAMBDA				
	MATH	SPELL	EDLEVEL	NEWSCHL	SUSPEND
	-----	-----	-----	-----	-----
WRATSPL	0.000	1.000	0.000	0.000	0.000
WJSPL	0.000	1.038	0.000	0.000	0.000
WAISSPL	0.000	0.967	0.000	0.000	0.000
WRATCALC	1.000	0.000	0.000	0.000	0.000
WJCALC	0.687	0.000	0.000	0.000	0.000
WAISCALC	0.402	0.000	0.000	0.000	0.000
EDLEVEL	0.000	0.000	1.000	0.000	0.000
NEWSCHL	0.000	0.000	0.000	1.000	0.000
SUSPEND	0.000	0.000	0.000	0.000	1.000
EXPELLED	0.000	0.000	0.000	0.000	0.000
HAVELD	0.000	0.000	0.000	0.000	0.000
FEMALE	0.000	0.000	0.000	0.000	0.000
AGE	0.000	0.000	0.000	0.000	0.000

	LAMBDA			
	EXPELLED	HAVELD	FEMALE	AGE
	-----	-----	-----	-----
WRATSPL	0.000	0.000	0.000	0.000
WJSPL	0.000	0.000	0.000	0.000

SEM Results ...

WAISSPL	0.000	0.000	0.000	0.000
WRATCALC	0.000	0.000	0.000	0.000
WJCALC	0.000	0.000	0.000	0.000
WAISSCALC	0.000	0.000	0.000	0.000
EDLEVEL	0.000	0.000	0.000	0.000
NEWSCHL	0.000	0.000	0.000	0.000
SUSPEND	0.000	0.000	0.000	0.000
EXPELLED	1.000	0.000	0.000	0.000
HAVELD	0.000	1.000	0.000	0.000
FEMALE	0.000	0.000	1.000	0.000
AGE	0.000	0.000	0.000	1.000

THETA

	WRATSPL	WJSPL	WAISSPL	WRATCALC	WJCALC
WRATSPL	24.116				
WJSPL	0.000	25.601			
WAISSPL	0.000	0.000	22.672		
WRATCALC	0.000	0.000	0.000	20.646	
WJCALC	0.000	0.000	0.000	0.000	10.650
WAISSCALC	0.000	0.000	0.000	0.000	0.000
EDLEVEL	0.000	0.000	0.000	0.000	0.000
NEWSCHL	0.000	0.000	0.000	0.000	0.000
SUSPEND	0.000	0.000	0.000	0.000	0.000
EXPELLED	0.000	0.000	0.000	0.000	0.000

SEM Results ...

HAVELD	0.000	0.000	0.000	0.000	0.000
FEMALE	0.000	0.000	0.000	0.000	0.000
AGE	0.000	0.000	0.000	0.000	0.000

	THETA				
	WAISCALC	EDLEVEL	NEWSCHL	SUSPEND	EXPELLED
	-----	-----	-----	-----	-----
WAISCALC	5.620				
EDLEVEL	0.000	0.000			
NEWSCHL	0.000	0.000	0.000		
SUSPEND	0.000	0.000	0.000	0.000	
EXPELLED	0.000	0.000	0.000	0.000	0.000
HAVELD	0.000	0.000	0.000	0.000	0.000
FEMALE	0.000	0.000	0.000	0.000	0.000
AGE	0.000	0.000	0.000	0.000	0.000

	THETA		
	HAVELD	FEMALE	AGE
	-----	-----	-----
HAVELD	0.000		
FEMALE	0.000	0.000	
AGE	0.000	0.000	0.000

SEM Results ...

	ALPHA MATH	SPELL	EDLEVEL	NEWSCHL	SUSPEND
	-----	-----	-----	-----	-----
1	0.000	0.000	11.153	0.546	0.518

	ALPHA EXPELLED	HAVELD	FEMALE	AGE
	-----	-----	-----	-----
1	0.141	0.157	0.316	19.719

	BETA MATH	SPELL	EDLEVEL	NEWSCHL	SUSPEND
	-----	-----	-----	-----	-----
MATH	0.000	0.000	0.000	0.000	0.000
SPELL	0.000	0.000	0.000	0.000	0.000
EDLEVEL	0.000	0.000	0.000	0.000	0.000
NEWSCHL	0.000	0.000	0.000	0.000	0.000
SUSPEND	0.000	0.000	0.000	0.000	0.000
EXPELLED	0.000	0.000	0.000	0.000	0.000
HAVELD	0.000	0.000	0.000	0.000	0.000
FEMALE	0.000	0.000	0.000	0.000	0.000
AGE	0.000	0.000	0.000	0.000	0.000

SEM Results ...

	BETA			
	EXPELLED	HAVELD	FEMALE	AGE
	-----	-----	-----	-----
MATH	0.000	0.000	0.000	0.000
SPELL	0.000	0.000	0.000	0.000
EDLEVEL	0.000	0.000	0.000	0.000
NEWSCHL	0.000	0.000	0.000	0.000
SUSPEND	0.000	0.000	0.000	0.000
EXPELLED	0.000	0.000	0.000	0.000
HAVELD	0.000	0.000	0.000	0.000
FEMALE	0.000	0.000	0.000	0.000
AGE	0.000	0.000	0.000	0.000

	PSI				
	MATH	SPELL	EDLEVEL	NEWSCHL	SUSPEND
	-----	-----	-----	-----	-----
MATH	0.050				
SPELL	0.000	0.050			
EDLEVEL	0.000	0.000	1.472		
NEWSCHL	0.000	0.000	-0.026	0.248	
SUSPEND	0.000	0.000	-0.057	0.046	0.250
EXPELLED	0.000	0.000	-0.031	0.035	0.052
HAVELD	0.000	0.000	-0.008	-0.015	-0.004
FEMALE	0.000	0.000	-0.064	0.019	-0.033
AGE	0.000	0.000	0.912	-0.022	-0.158

SEM Results ...

	PSI			
	EXPELLED	HAVELD	FEMALE	AGE
	-----	-----	-----	-----
EXPELLED	0.121			
HAVELD	-0.003	0.132		
FEMALE	-0.019	-0.008	0.216	
AGE	-0.015	-0.007	-0.001	3.627

Beginning Time: 21:48:08

Ending Time: 21:48:08

Elapsed Time: 00:00:00

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SEM Results ...

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Outline

- 1 Example 1: Job Data EFA
- 2 Example 2: CFA
- 3 Example 3: Two Group CFA
- 4 Example 4: Multiple Regression
- 5 Example 5: SEM
- 6 Example: Insomnia SEM**
- 7 Example: Growth Curve
- 8 Conclusion

Insomnia Project

- Data file is “insomnia.dat” has 7 indicators of sleep difficulty
- Question: How do the separate dimensions of insomnia predict the quality of life
- Here we are focused on the regression model, not on the measurement model (which, we assume, has been investigated previously).
 - Estimate regression coefficients for 3 insomnia factors predicting Quality of life

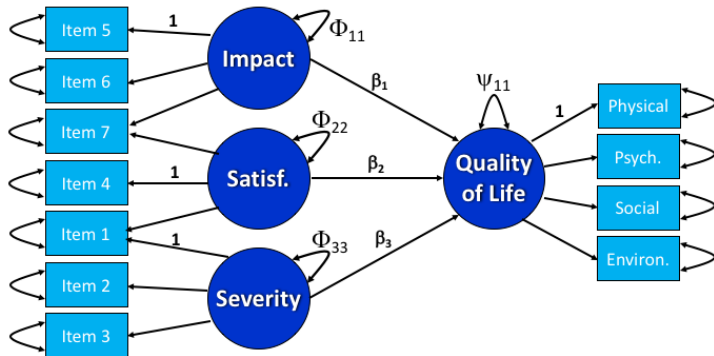
Insomnia SEM

- Three latent variables about insomnia (7 observed indicators altogether)
 - 1 Impact of insomnia
 - 2 Severity of insomnia
 - 3 Satisfaction with sleep
- Four items that are indicators of Quality of Life

About the Model

- We used the “marker variable” approach for identification
 - the loading (λ) of the FIRST indicator of each factor is fixed at 1
 - This means the other estimates for that factor are scaled relative to the first item

Insomnia SEM Diagram



About the Model

- We used the “marker variable” approach for identification
 - the loading (λ) of the FIRST indicator of each factor is fixed at 1
 - This means the other estimates for that factor are scaled relative to the first item

Mplus Syntax

```
TITLE: Example 5B
DATA:
  FILE IS "../..//data/insomnia.dat";
VARIABLE:
  names are insom1-insom7 cesd1-cesd20 phy psy soc
    env;
  usevariables are insom1-insom7 phy psy soc env;
  missing all (999);
MODEL:
  Impact by insom5 insom6 insom7;
  Severity by insom1 insom2 insom3;
  Satisf by insom4 insom1 insom7;
  Qol by phy psy soc env;
  Qol on Impact Severity Satisf;
```

Results

View Online: [sem-02.out](#)

Inline Display:

```
Mplus VERSION 7.4 (Linux)
MUTHEN & MUTHEN
05/28/2018  9:48 PM
```

INPUT INSTRUCTIONS

TITLE:

```
Example 5b
```

DATA:

```
FILE IS "../..data/insomnia.dat";
```

VARIABLE:

```
names are insom1-insom7 cesd1-cesd20 phy psy soc env;
usevariables are insom1-insom7 phy psy soc env;
missing all (999);
```

MODEL:

```
Impact by insom5 insom6 insom7;
Severity by insom1 insom2 insom3;
```

Results ...

```
Satisf by insom4 insom1 insom7;
Qol by phy psy soc env;
Qol on Impact Severity Satisf;
```

INPUT READING TERMINATED NORMALLY

Example 5b

SUMMARY OF ANALYSIS

Number of groups	1
Number of observations	103
Number of dependent variables	11
Number of independent variables	0
Number of continuous latent variables	4

Observed dependent variables

Continuous

INSOM1	INSOM2	INSOM3	INSOM4	INSOM5	INSOM6
--------	--------	--------	--------	--------	--------

Results ...

```

INSOM7      PHY      PSY      SOC      ENV

Continuous latent variables
  IMPACT      SEVERITY  SATISF      QOL

Estimator                                ML
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/insomnia.dat

Input data format  FREE

SUMMARY OF DATA

  Number of missing data patterns      3

COVARIANCE COVERAGE OF DATA

```

Results ...

Minimum covariance coverage value 0.100

PROPORTION OF DATA PRESENT

	Covariance Coverage				
	INSOM1	INSOM2	INSOM3	INSOM4	INSOM5
	-----	-----	-----	-----	-----
INSOM1	1.000				
INSOM2	1.000	1.000			
INSOM3	1.000	1.000	1.000		
INSOM4	1.000	1.000	1.000	1.000	
INSOM5	1.000	1.000	1.000	1.000	1.000
INSOM6	1.000	1.000	1.000	1.000	1.000
INSOM7	1.000	1.000	1.000	1.000	1.000
PHY	1.000	1.000	1.000	1.000	1.000
PSY	0.990	0.990	0.990	0.990	0.990
SOC	0.990	0.990	0.990	0.990	0.990
ENV	1.000	1.000	1.000	1.000	1.000

	Covariance Coverage				
	INSOM6	INSOM7	PHY	PSY	SOC
	-----	-----	-----	-----	-----

Results ...

INSOM6	1.000					
INSOM7	1.000	1.000				
PHY	1.000	1.000	1.000			
PSY	0.990	0.990	0.990	0.990		
SOC	0.990	0.990	0.990	0.990	0.981	0.990
ENV	1.000	1.000	1.000	0.990	0.990	0.990

Covariance Coverage

ENV

ENV	----- 1.000
-----	----------------

THE MODEL ESTIMATION TERMINATED NORMALLY

MODEL FIT INFORMATION

Number of Free Parameters 41

Loglikelihood

H0 Value -1044.951

Results ...

H1 Value	-1023.210
----------	-----------

Information Criteria

Akaike (AIC)	2171.901
Bayesian (BIC)	2279.925
Sample-Size Adjusted BIC	2150.413
(n* = (n + 2) / 24)	

Chi-Square Test of Model Fit

Value	43.481
Degrees of Freedom	36
P-Value	0.1829

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.045	
90 Percent C.I.	0.000	0.087
Probability RMSEA <= .05	0.539	

CFI/TLI

CFI	0.986
TLI	0.979

Results ...

Chi-Square Test of Model Fit for the Baseline Model

Value	602.272
Degrees of Freedom	55
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value	0.045
-------	-------

MODEL RESULTS

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
IMPACT BY				
INSOM5	1.000	0.000	999.000	999.000
INSOM6	0.650	0.110	5.923	0.000
INSOM7	0.317	0.118	2.676	0.007
SEVERITY BY				
INSOM1	1.000	0.000	999.000	999.000
INSOM2	6.966	6.248	1.115	0.265
INSOM3	5.562	5.055	1.100	0.271

Results ...

SATISF	BY				
INSOM4		1.000	0.000	999.000	999.000
INSOM1		0.756	0.174	4.354	0.000
INSOM7		0.801	0.187	4.275	0.000
QOL	BY				
PHY		1.000	0.000	999.000	999.000
PSY		1.176	0.115	10.196	0.000
SOC		0.803	0.118	6.778	0.000
ENV		0.942	0.101	9.301	0.000
QOL	ON				
IMPACT		-0.041	0.065	-0.629	0.530
SEVERITY		-0.189	0.595	-0.318	0.750
SATISF		-0.302	0.140	-2.156	0.031
SEVERITY WITH					
IMPACT		0.045	0.044	1.035	0.301
SATISF WITH					
IMPACT		0.492	0.107	4.612	0.000
SEVERITY		0.058	0.052	1.120	0.263
Intercepts					
INSOM1		0.718	0.082	8.792	0.000

Results ...

INSOM2	0.670	0.090	7.407	0.000
INSOM3	0.738	0.091	8.105	0.000
INSOM4	1.738	0.087	20.023	0.000
INSOM5	1.621	0.112	14.523	0.000
INSOM6	0.951	0.100	9.485	0.000
INSOM7	1.184	0.103	11.554	0.000
PHY	3.659	0.050	72.783	0.000
PSY	3.413	0.060	56.813	0.000
SOC	3.523	0.054	64.839	0.000
ENV	3.520	0.050	69.751	0.000

Variances

IMPACT	1.109	0.228	4.857	0.000
SEVERITY	0.015	0.026	0.556	0.578
SATISF	0.509	0.114	4.482	0.000

Residual Variances

INSOM1	0.295	0.054	5.485	0.000
INSOM2	0.136	0.095	1.428	0.153
INSOM3	0.403	0.082	4.924	0.000
INSOM4	0.267	0.063	4.228	0.000
INSOM5	0.174	0.146	1.195	0.232
INSOM6	0.568	0.100	5.676	0.000
INSOM7	0.394	0.071	5.591	0.000
PHY	0.068	0.015	4.443	0.000
PSY	0.104	0.021	4.830	0.000

Results ...

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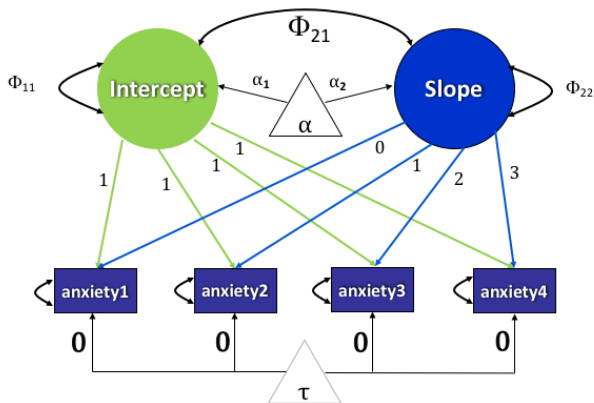
Outline

- 1 Example 1: Job Data EFA
- 2 Example 2: CFA
- 3 Example 3: Two Group CFA
- 4 Example 4: Multiple Regression
- 5 Example 5: SEM
- 6 Example: Insomnia SEM
- 7 Example: Growth Curve
- 8 Conclusion

Anxiety data

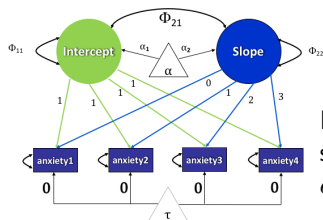
- “anxiety.dat” includes anxiety measures at 4 time points named anxiety1 , anxiety2 , anxiety3 , anxiety4
- Some people (eg current CRMDA director) find the graphic representation of the Latent Growth Curve Model to be quite confusing and difficult to understand. You’ve been warned.

LGCM Diagram



LGCM Diagram

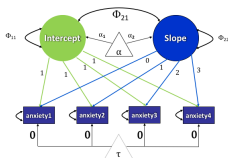
This amounts to a “random effects” multilevel model



- time, is coded 0, 1, 2, 3. Time values are written on the graph as if they were “slope coefficients”.
- In the Mplus code, this is “LINEAR” latent variable

LGCM is based on idea of “wide” data, predicts for separate outcomes *anxiety*₁, *anxiety*₂, *anxiety*₃, *anxiety*₄

LGCM Diagram



- LGCM: Think of intercept and slope as random effects, where the center points are α_1 and α_2 .
- MLM view: Effect of time passing for case i on anxiety at time t

$$anxiety_{it} = intercept_i + slope_i \cdot time + error_{it}$$

- MLM typically asserts $intercept_i \sim N(\alpha_1, \sigma_{b0}^2)$ and $slope_i \sim N(\alpha_2, \sigma_{b1}^2)$, using the LGCM notation

LGCM Syntax

```
TITLE:
Example 6 - Linear Latent Growth Curve Model with
    4 Time Points
DATA:
FILE IS "../.../data/anxiety.dat";
VARIABLE:
NAMES ARE anxiety1 anxiety2 anxiety3 anxiety4;
MODEL:
INTCEPT BY anxiety1@1 anxiety2@1 anxiety3@1
    anxiety4@1;
LINEAR BY anxiety1@0 anxiety2@1 anxiety3@2
    anxiety4@3;
anxiety1-anxiety4;
[anxiety1-anxiety4@0];
INTCEPT LINEAR;
[INTCEPT LINEAR];
```

LGCM Syntax ...

```
INTCEPT WITH LINEAR;
```

LGCM Results

View Online: [lgc-01.out](#)

Inline Display:

```
Mplus VERSION 7.4 (Linux)
MUTHEN & MUTHEN
05/28/2018  9:50 PM
```

INPUT INSTRUCTIONS

TITLE:

```
Example 6 - Linear Latent Growth Curve Model with 4 Time Points
```

DATA:

```
FILE IS "../..data/anxiety.dat";
```

VARIABLE:

NAMES ARE

```
anxiety1 anxiety2 anxiety3 anxiety4;
```

MODEL:

```
INTCEPT BY anxiety1@1 anxiety2@1 anxiety3@1 anxiety4@1;
```

```
LINEAR BY anxiety1@0 anxiety2@1 anxiety3@2 anxiety4@3;
```

LGCM Results ...

```
anxiety1-anxiety4;
[anxiety1-anxiety4@0];
```

```
INTCEPT LINEAR;
[INTCEPT LINEAR];
INTCEPT WITH LINEAR;
```

INPUT READING TERMINATED NORMALLY

Example 6 - Linear Latent Growth Curve Model with 4 Time Points

SUMMARY OF ANALYSIS

Number of groups	1
Number of observations	485
Number of dependent variables	4
Number of independent variables	0
Number of continuous latent variables	2

Observed dependent variables

LGCM Results ...

```
Continuous
  ANXIETY1    ANXIETY2    ANXIETY3    ANXIETY4
```

```
Continuous latent variables
  INTCEPT    LINEAR
```

```
Estimator                                ML
Information matrix                        OBSERVED
Maximum number of iterations              1000
Convergence criterion                     0.500D-04
Maximum number of steepest descent iterations 20
```

```
Input data file(s)
  ../../data/anxiety.dat
```

```
Input data format  FREE
```

```
THE MODEL ESTIMATION TERMINATED NORMALLY
```

```
MODEL FIT INFORMATION
```

LGCM Results ...

Number of Free Parameters 9

Loglikelihood

H0 Value	-567.499
H1 Value	-553.855

Information Criteria

Akaike (AIC)	1152.997
Bayesian (BIC)	1190.654
Sample-Size Adjusted BIC	1162.089
(n* = (n + 2) / 24)	

Chi-Square Test of Model Fit

Value	27.288
Degrees of Freedom	5
P-Value	0.0001

RMSEA (Root Mean Square Error Of Approximation)

Estimate	0.096
90 Percent C.I.	0.063 0.133
Probability RMSEA <= .05	0.013

LGCM Results ...

CFI/TLI

CFI	0.981
TLI	0.977

Chi-Square Test of Model Fit for the Baseline Model

Value	1182.102
Degrees of Freedom	6
P-Value	0.0000

SRMR (Standardized Root Mean Square Residual)

Value	0.061
-------	-------

MODEL RESULTS

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
INTCEPT BY				
ANXIETY1	1.000	0.000	999.000	999.000
ANXIETY2	1.000	0.000	999.000	999.000

LGCM Results ...

ANXIETY3	1.000	0.000	999.000	999.000
ANXIETY4	1.000	0.000	999.000	999.000
LINEAR BY				
ANXIETY1	0.000	0.000	999.000	999.000
ANXIETY2	1.000	0.000	999.000	999.000
ANXIETY3	2.000	0.000	999.000	999.000
ANXIETY4	3.000	0.000	999.000	999.000
INTCEPT WITH				
LINEAR	-0.011	0.003	-3.485	0.000
Means				
INTCEPT	0.698	0.020	34.866	0.000
LINEAR	-0.062	0.006	-10.405	0.000
Intercepts				
ANXIETY1	0.000	0.000	999.000	999.000
ANXIETY2	0.000	0.000	999.000	999.000
ANXIETY3	0.000	0.000	999.000	999.000
ANXIETY4	0.000	0.000	999.000	999.000
Variances				
INTCEPT	0.151	0.013	11.996	0.000
LINEAR	0.007	0.001	4.667	0.000

LGCM Results ...

Residual Variances

ANXIETY1	0.067	0.008	8.675	0.000
ANXIETY2	0.048	0.004	10.965	0.000
ANXIETY3	0.048	0.004	11.180	0.000
ANXIETY4	0.040	0.006	6.545	0.000

QUALITY OF NUMERICAL RESULTS

Condition Number for the Information Matrix 0.150E-02
(ratio of smallest to largest eigenvalue)

Beginning Time: 21:50:25
Ending Time: 21:50:25
Elapsed Time: 00:00:00

MUTHEN & MUTHEN
3463 Stoner Ave.
Los Angeles, CA 90066

Tel: (310) 391-9971
Fax: (310) 391-8971
Web: www.StatModel.com

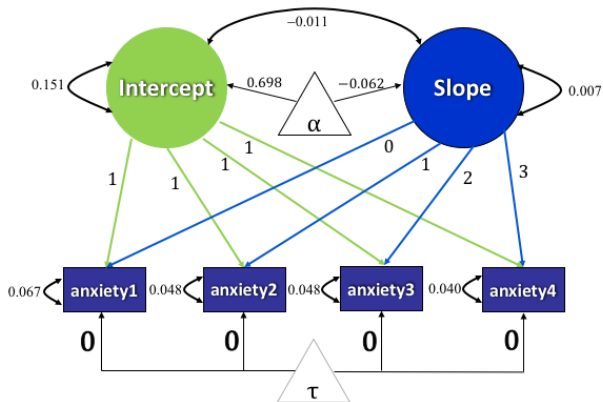
LGCM Results ...

Support: Support@StatModel.com

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LGCM Diagram w/Estimates

LGCM Diagram w/Estimates ...



Compare with a multi-level model

- R's lme4 package is a state of the art framework for estimating models with random effects
- Because this Mplus example has observed indicators of anxiety, we can estimate the same model with lme4 and compare the results.

R data import

```
ddir <- "data"
anxiety <- read.table(file.path(ddir,
  "anxiety.dat"), sep = "\t", header = FALSE,
  col.names = paste0("anxiety", 1:4))
## rockchalk::summarize(anxiety)
```

- The data must be reshaped, to one long column for anxiety.

```
anxiety2 <- reshape(anxiety,
  varying=c("anxiety1", "anxiety2", "anxiety3",
  "anxiety4"),
  v.names = "anxiety", direction = "long")
anxiety2 <- anxiety2[order(anxiety2$id,
  anxiety2$time), ]
head(anxiety2, 15)
```

R data import ...

```
time anxiety id
1.1 1 1.714286 1
1.2 2 1.500000 1
1.3 3 1.714286 1
1.4 4 1.928571 1
2.1 1 1.642857 2
2.2 2 1.642857 2
2.3 3 1.214286 2
2.4 4 1.928571 2
3.1 1 1.642857 3
3.2 2 2.000000 3
3.3 3 1.928571 3
3.4 4 1.857143 3
4.1 1 1.428571 4
4.2 2 1.857143 4
4.3 3 1.928571 4
```

R data import ...

```
library(lme4)
## set time to begin at 0 to match the Mplus
  calculations
anxiety2$time <- anxiety2$time -1
m1 <- lmer(anxiety ~ time + (time | id), data =
  anxiety2)
summary(m1)
```

```
Linear mixed model fit by REML ['lmerMod']
Formula: anxiety ~ time + (time | id)
Data: anxiety2
```

```
REML criterion at convergence: 1161.2
```

```
Scaled residuals:
```

Min	1Q	Median	3Q	Max
-3.0428	-0.5191	-0.0757	0.4761	3.2378

```
Random effects:
```

Groups	Name	Variance	Std.Dev.	Corr
id	(Intercept)	0.154950	0.39364	

R data import ...

```
15      time      0.007544 0.08686 -0.39
   Residual      0.049801 0.22316
Number of obs: 1940, groups: id, 485

Fixed effects:
20      Estimate Std. Error t value
(Intercept)  0.702800   0.019783   35.53
time        -0.065068   0.006008  -10.83

Correlation of Fixed Effects:
25      (Intr)
time -0.489
```

R data import ...

- Estimated coefficients are very similar

	Mplus LGCM	lme4
Intercept Variance	0.151	0.1549
Slope Variance	0.007	0.0075
Correlation Intercept/Slope	*	-0.39
* Mplus reports covariance, not correlation, as	-0.011	
Intercept Mean	0.698	0.702
Time (Mean of random slope)	-0.062	-0.0650
Anxiety Residual variance time 1	0.067	
time 2	0.048	
time 3	0.048	
time 4	0.050	+
lme4 did not separately estimate variances		0.0498

R data import ...

Why do I need Mplus, if I can estimate that with lme4?

- lme4 works because the outcome variable is observed, not latent.
- If the outcome variable were latent, with indicators, then there's almost no hope we could code this up with lme4
- lme4 has not (yet, maybe ever) implemented convenient method to allow residual variance to differ by time point.

Why did I wish to try lme4?

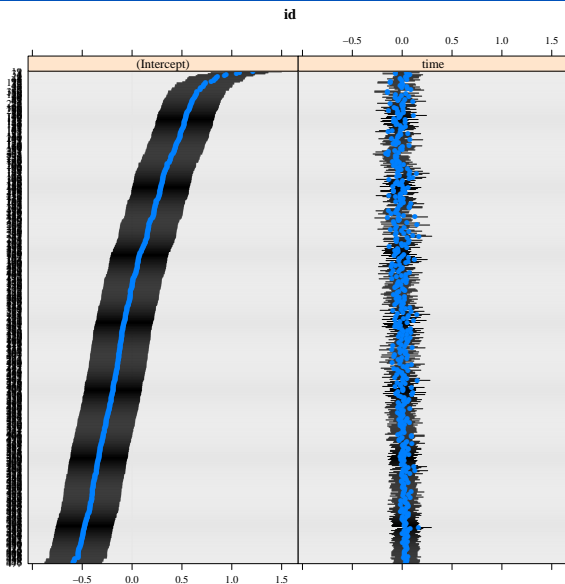
- The model is more understandable for me to think of this as data in the long format
- It is “just” a time regression with intercepts and slopes that vary among people

$$anxiety_{it} = \beta_{0i} + \beta_{1i} \cdot time_t + \varepsilon_{it}$$

$$\beta_{0i} \sim N(0, \sigma_{\beta_0}^2) \text{ and } \beta_{1i} \sim N(0, \sigma_{\beta_1}^2)$$

- lme4 (R more broadly) offers many convenient tools to visualize the output.

R data import ...



Outline

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We will keep making examples

- It is difficult to keep all of these details straight, but we are dedicated to try
- We should have full working input and output from Mplus (Muthén & Muthén, 2017) , R (R Core Team, 2017), and Stata

References

Muthén, L. K. & Muthén, B. O. (2017). *Mplus User's Guide*. Los Angeles, CA: Muthén & Muthén, 8th edition.

R Core Team (2017). *R: A Language and Environment for Statistical Computing*. R Foundation for Statistical Computing, Vienna, Austria.

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] lattice_0.20-38 lme4_1.1-21      Matrix_1.2-17
```

Session ...

loaded via a namespace (and not attached):

```
[1] minqa_1.2.4      MASS_7.3-51.4   compiler_3.6.0  tools_3.6.0  
     Rcpp_1.0.1      splines_3.6.0  
[7] nlme_3.1-140    grid_3.6.0      nloptr_1.2.1    boot_1.3-22
```