

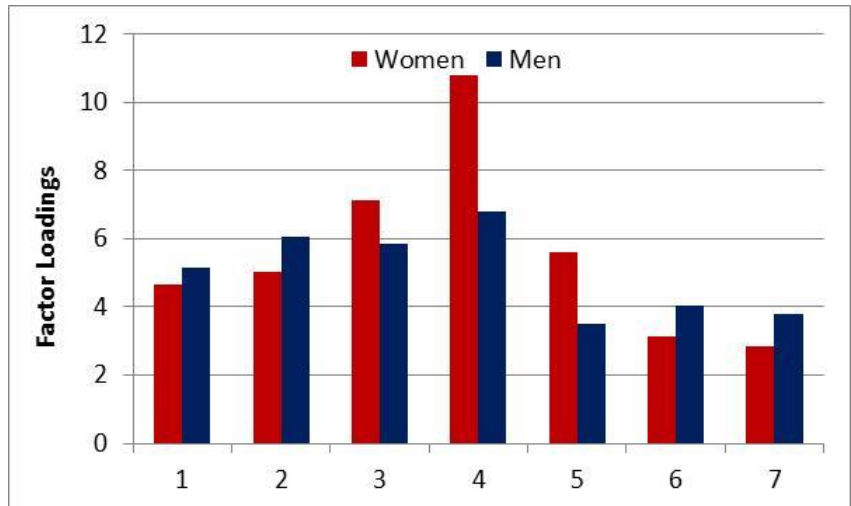
### Testing Multiple-Group Measurement Invariance in Item Factor Models using ML in Mplus v. 8.4

Example data: 635 older adults (age 80–100) self-reporting on 7 items assessing the Instrumental Activities of Daily Living (IADL) as follows. We are examining differences between men ( $N=214$ ) and women ( $N=420$ ). Each item has **two** response options (0 = needs help, 1 = does not need help), and on the right are the means by gender.

Item		Mean	
		Men	Women
1	Housework	0.596	0.657
2	Bedmaking	0.816	0.853
3	Cooking	0.687	0.814
4	Everyday shopping	0.743	0.622
5	Walking to places	0.751	0.603
6	Handling banking	0.767	0.716
7	Using a telephone	0.934	0.948

### Multiple Group IFA Model Syntax and Truncated Output:

<pre> <b>TITLE:</b> Assess invariance for binary IADL items using ML <b>DATA:</b> FILE = Example7cd.csv; ! Don't need path if in same folder         FORMAT = free;           ! Default         TYPE = INDIVIDUAL;       ! Default  <b>VARIABLE:</b> NAMES = case female dial-dia7 cial-cia7; ! All vars in data         USEVARIABLES = dial-dia7; ! All vars in model         CATEGORICAL = dial-dia7; ! All ordinal outcomes         CLASSES = group(2);      ! 2 fake latent classes         KNOWNCLASS = group(female=0 1); ! Latent classes = gender         MISSING = ALL (9999);    ! Missing value code         IDVARIABLE = case;       ! Person ID variable  <b>ANALYSIS:</b> ESTIMATOR = ML; LINK = LOGIT; ! Full-info ML in logits         CONVERGENCE = 0.0000001; ! For OS comparability         TYPE = MIXTURE;          ! For latent classes         ALGORITHM = INTEGRATION; ! Required for latent classes  <b>PLOT:</b> TYPE = PLOT1 PLOT2 PLOT3; ! Get all the plots  <b>OUTPUT:</b> STDYX RESIDUAL; ! No voo-doo for IFA in ML  <b>MODEL:</b>  ! MEN REFERENCE GROUP CONFIGURAL MODEL (will stay the same) %OVERALL% ! Needed for fake latent class model ! Factor loadings all estimated, just labeled IADL BY dial-dia7* (L1-L7); ! Item thresholds all free, just labeled [dial\$1-dia7\$1*] (T1-T7); ! Factor mean=0 and variance=1 for identification [IADL@0]; IADL@1;  ! WOMEN ALTERNATIVE GROUP 2 CONFIGURAL MODEL 1 %group#2% ! Needed for fake latent class model ! Factor loadings all estimated IADL BY dial-dia7*; ! Item thresholds all free [dial\$1-dia7\$1*]; ! Factor mean=0 and variance=1 for identification [IADL@0]; IADL@1; </pre>	<p><b>MODEL FIT INFORMATION</b></p> <p>Number of Free Parameters 29</p> <p>Loglikelihood H0 Value -1796.106</p> <p>Information Criteria Akaike (AIC) 3650.212 Bayesian (BIC) 3779.367 Sample-Size Adjusted BIC 3687.295 (n* = (n + 2) / 24)</p> <p>This will be our baseline configural model. 29 parameters estimated = 2*[7 loadings + 7 thresholds] = 28 1 extra parameter is estimated as the logit of the proportion of the sample in group#1 (the female=0 group).</p>
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**Model 1. Configural Invariance Model**  
**(Everything separate across groups \*except\* for parameters needed to be constrained for identification)**

UNSTANDARDIZED MODEL RESULTS (IFA MODEL SOLUTION)						UNSTANDARDIZED MODEL RESULTS (IFA MODEL SOLUTION)					
	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value			Estimate	S.E.	Est./S.E.	Two-Tailed P-Value	
<b>Latent Class 1 (0) - refers to female=0 from KNOWNCLASS</b>						<b>Latent Class 2 (1) - refers to female=1 from KNOWNCLASS</b>					
IADL	BY -- FACTOR LOADINGS					IADL	BY -- FACTOR LOADINGS				
DIA1	5.138	1.321	3.890	0.000		DIA1	4.664	0.731	6.377	0.000	
DIA2	6.066	1.830	3.315	0.001		DIA2	5.014	0.998	5.026	0.000	
DIA3	5.828	1.566	3.722	0.000		DIA3	7.108	1.670	4.257	0.000	
DIA4	6.792	2.086	3.256	0.001		<b>DIA4</b>	<b>10.774</b>	<b>8.194</b>	<b>1.315</b>	<b>0.189 !!!</b>	
DIA5	3.520	0.717	4.907	0.000		DIA5	5.611	1.145	4.902	0.000	
DIA6	4.026	0.890	4.525	0.000		DIA6	3.143	0.422	7.440	0.000	
DIA7	3.789	1.211	3.129	0.002		DIA7	2.857	0.610	4.681	0.000	
<b>Means: MEAN OF THETA FIXED=0 FOR IDENTIFICATION</b>						<b>Means: MEAN OF THETA FIXED=0 FOR IDENTIFICATION</b>					
IADL	0.000	0.000	999.000	999.000		IADL	0.000	0.000	999.000	999.000	
<b>Thresholds: EXPECTED LOGIT OF Y=0 IF THETA=0</b>						<b>Thresholds</b>					
DIA1\$1	-1.372	0.622	-2.204	0.028		DIA1\$1	-2.188	0.423	-5.177	0.000	
DIA2\$1	-5.632	1.614	-3.489	0.000		DIA2\$1	-5.694	1.020	-5.583	0.000	
DIA3\$1	-3.005	0.919	-3.270	0.001		DIA3\$1	-6.715	1.494	-4.496	0.000	
DIA4\$1	-4.378	1.337	-3.274	0.001		DIA4\$1	-3.791	2.941	-1.289	0.197	
DIA5\$1	-2.722	0.552	-4.928	0.000		DIA5\$1	-1.798	0.491	-3.659	0.000	
DIA6\$1	-3.311	0.707	-4.684	0.000		DIA6\$1	-2.184	0.313	-6.976	0.000	
DIA7\$1	-6.398	1.701	-3.762	0.000		DIA7\$1	-5.577	0.886	-6.298	0.000	
<b>Variances: VARIANCE OF THETA FIXED=1 FOR IDENTIFICATION</b>						<b>Variances: VARIANCE OF THETA FIXED=1 FOR IDENTIFICATION</b>					
IADL	1.000	0.000	999.000	999.000		IADL	1.000	0.000	999.000	999.000	
<b>Although Mplus does give IRT a-parameters and b-parameters for binary items, it rescales them to assume a theta mean=0 and variance=1 in both groups. Thus, they will not be invariant even when the loadings and thresholds are invariant. For this reason, they are not shown here (but one can calculate the non-invariant versions using MODEL CONSTRAINT or excel).</b>						<b>Categorical Latent Variables</b>					
						<b>Means: logit of proportion of men (group#1) = .34</b>					
						GROUP#1            -0.670            0.084            -7.985            0.000					

**Model 2a. Metric Invariance Model (IFA loadings held equal across groups – Mplus IRT discriminations still vary via factor variances)**

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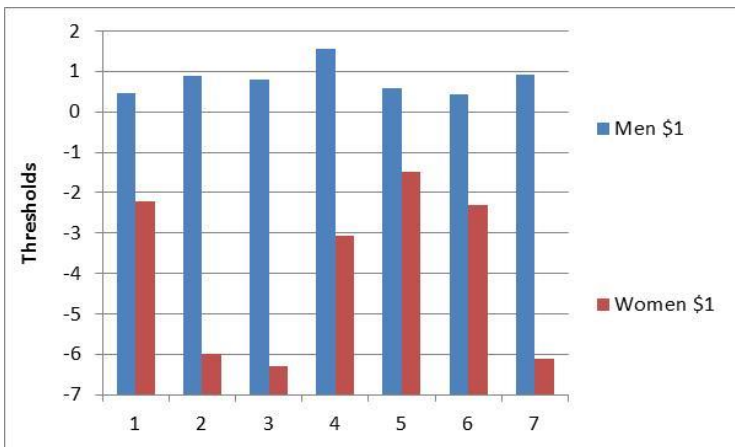
MODEL:
! MEN REFERENCE GROUP CONFIGURAL MODEL (will stay the same)
%OVERALL% ! Needed for fake latent class model
! Factor loadings all estimated, just labeled
IADL BY dial-dia7* (L1-L7);
! Item thresholds all free, just labeled
[dial$1-dia7$1*] (T1-T7);
! Factor mean=0 and variance=1 for identification
[IADL@0]; IADL@1;

! WOMEN ALTERNATIVE GROUP 2 METRIC MODEL 2
%group#2% ! Needed for fake latent class model
! Factor loadings NOW CONSTRAINED EQUAL TO MEN
IADL BY dial-dia7* (L1-L7);
! Item thresholds all free
[dial$1-dia7$1*];
! Factor mean=0 for identification
! Factor variance NOW ESTIMATED
[IADL@0]; IADL*;
    
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MODEL FIT INFORMATION

Number of Free Parameters	23
Loglikelihood	
H0 Value	-1799.928
Information Criteria	
Akaike (AIC)	3645.855
Bayesian (BIC)	3748.288
Sample-Size Adjusted BIC	3675.266
(n* = (n + 2) / 24)	

Does the full metric invariance model (2a) fit significantly worse than the configural model (1)? Nope,  $-2\Delta LL(df=6) = 7.64, p = .27$



	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
Latent Class 1 (0)				
IADL BY				
DIA1	4.691	0.714	6.570	0.000
DIA2	5.209	0.938	5.550	0.000
DIA3	6.517	1.258	5.179	0.000
DIA4	8.631	2.433	3.548	0.000
DIA5	4.611	0.669	6.893	0.000
DIA6	3.337	0.465	7.181	0.000
DIA7	3.195	0.629	5.082	0.000
Means: Factor mean fixed=0 for identification				
IADL	0.000	0.000	999.000	999.000
Thresholds				
DIA1\$1	-1.267	0.473	-2.681	0.007
DIA2\$1	-4.872	0.876	-5.563	0.000
DIA3\$1	-3.308	0.811	-4.078	0.000
DIA4\$1	-5.408	1.559	-3.468	0.001
DIA5\$1	-3.387	0.580	-5.838	0.000
DIA6\$1	-2.828	0.441	-6.409	0.000
DIA7\$1	-5.626	0.924	-6.091	0.000
Variances: Factor variance fixed=1 for identification				
IADL	1.000	0.000	999.000	999.000
Latent Class 2 (1)				
IADL BY				
DIA1	4.691	0.714	6.570	0.000
DIA2	5.209	0.938	5.550	0.000
DIA3	6.517	1.258	5.179	0.000
DIA4	8.631	2.433	3.548	0.000
DIA5	4.611	0.669	6.893	0.000
DIA6	3.337	0.465	7.181	0.000
DIA7	3.195	0.629	5.082	0.000
Means: Factor mean fixed=0 for identification				
IADL	0.000	0.000	999.000	999.000
Thresholds				
DIA1\$1	-2.220	0.402	-5.520	0.000
DIA2\$1	-5.983	0.913	-6.555	0.000
DIA3\$1	-6.310	1.089	-5.794	0.000
DIA4\$1	-3.079	0.995	-3.095	0.002
DIA5\$1	-1.495	0.358	-4.179	0.000
DIA6\$1	-2.317	0.315	-7.348	0.000
DIA7\$1	-6.127	0.882	-6.949	0.000
Variances: Factor variance now estimated				
IADL	1.056	0.240	4.409	0.000

**Model 3a. Full Threshold Invariance Model (all IFA thresholds held equal across groups – Mplus IRT difficulties can still vary)**

<pre> MODEL: ! MEN REFERENCE GROUP CONFIGURAL MODEL (will stay the same) %OVERALL% ! Needed for fake latent class model ! Factor loadings all estimated, just labeled IADL BY dial-dia7* (L1-L7); ! Item thresholds all free, just labeled [dial\$1-dia7\$1*] (T1-T7); ! Factor mean=0 and variance=1 for identification [IADL@0]; IADL@1;  ! WOMEN ALTERNATIVE GROUP 2 SCALAR MODEL 3 %group#2% ! Needed for fake latent class model ! Factor loadings NOW CONSTRAINED EQUAL TO MEN IADL BY dial-dia7* (L1-L7); ! Item thresholds NOW CONSTRAINED EQUAL TO MEN [dial\$1-dia7\$1*] (T1-T7); ! Factor mean NOW ESTIMATED ! Factor variance NOW ESTIMATED [IADL*]; IADL*;  MODEL FIT INFORMATION Number of Free Parameters          17 Loglikelihood   H0 Value                        -1860.402 Information Criteria   Akaike (AIC)                    3754.804   Bayesian (BIC)                  3830.515   Sample-Size Adjusted BIC       3776.542         </pre> <p><b>Does the full scalar invariance model (3a) fit significantly worse than the full metric model (2a)?</b>  <b>Yep, <math>-2\Delta LL(df=6) = 120.95, p &lt; .01</math></b></p> <p>Based on the local misfit of the univariate distributions (from TECH10 output, see right), it looks like item 3 is the biggest problem... so the same steps would be followed as in our other examples. That is, we would try to free problematic thresholds until the scalar model fits not worse than the metric model (or the configural model, if you choose to constrain both loadings and thresholds at the same time).</p> <p>(See example 7c for a sample results section using WLSMV and polytomous versions of these items)</p>	<pre> RESIDUAL OUTPUT  UNIVARIATE DISTRIBUTION FIT FOR CLASS 1  Variable      Observed   Estimated Residual (Obs.-Est.) Stand. Residual DIA1   Category 1   0.404     0.352     0.052     1.591   Category 2   0.596     0.648    -0.052    -1.591 DIA2   Category 1   0.184     0.168     0.017     0.664   Category 2   0.816     0.832    -0.017    -0.664 DIA3   Category 1   0.313     0.229     0.084     <b>2.939</b>   Category 2   0.687     0.771    -0.084    -<b>2.939</b> DIA4   Category 1   0.257     0.333    -0.076    -2.362   Category 2   0.743     0.667     0.076     2.362 DIA5   Category 1   0.249     0.335    -0.086    -2.683   Category 2   0.751     0.665     0.086     2.683 DIA6   Category 1   0.233     0.261    -0.028    -0.946   Category 2   0.767     0.739     0.028     0.946 DIA7   Category 1   0.066     0.062     0.004     0.233   Category 2   0.934     0.938    -0.004    -0.233  UNIVARIATE DISTRIBUTION FIT FOR CLASS 2  Variable      Observed   Estimated Residual (Obs.-Est.) Stand. Residual DIA1   Category 1   0.343     0.370    -0.027    -1.132   Category 2   0.657     0.630     0.027     1.132 DIA2   Category 1   0.147     0.162    -0.015    -0.813   Category 2   0.853     0.838     0.015     0.813 DIA3   Category 1   0.186     0.230    -0.044    -<b>2.165</b>   Category 2   0.814     0.770     0.044     <b>2.165</b> DIA4   Category 1   0.378     0.348     0.030     1.298   Category 2   0.622     0.652    -0.030    -1.298 DIA5   Category 1   0.397     0.350     0.046     1.985   Category 2   0.603     0.650    -0.046    -1.985 DIA6   Category 1   0.284     0.267     0.017     0.785   Category 2   0.716     0.733    -0.017    -0.785 DIA7   Category 1   0.052     0.053    -0.001    -0.078   Category 2   0.948     0.947     0.001     0.078         </pre>
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