

ESEM, CFA, and Somewhere In-Between: The Effect of Measurement Quality on Model Fit Sensitivity

Tim Konold, Ph.D., University of Virginia

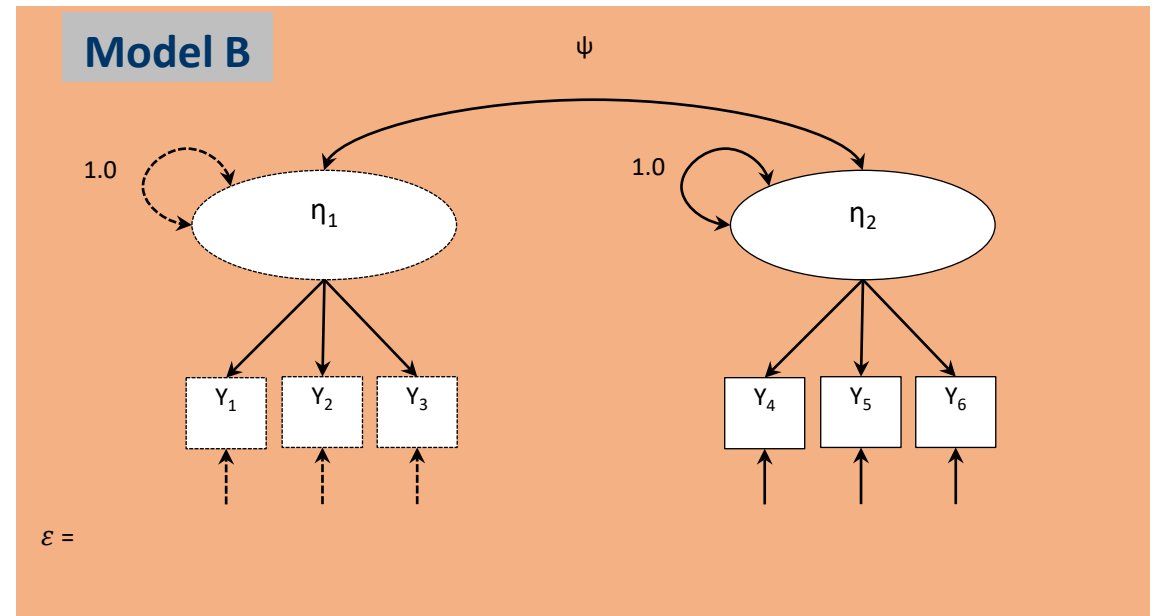
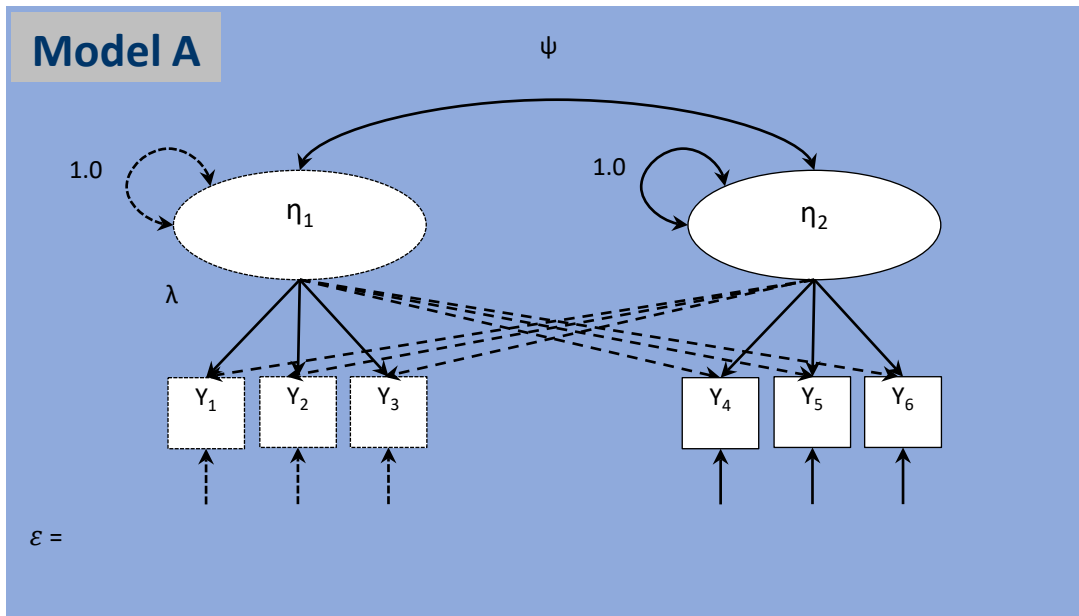
Liz Sanders, Ph.D. University of Washington

ESEM, CFA, and Somewhere In-Between: The Effect of ~~Measurement Quality~~ Cross Loading Sign on Model Fit Sensitivity

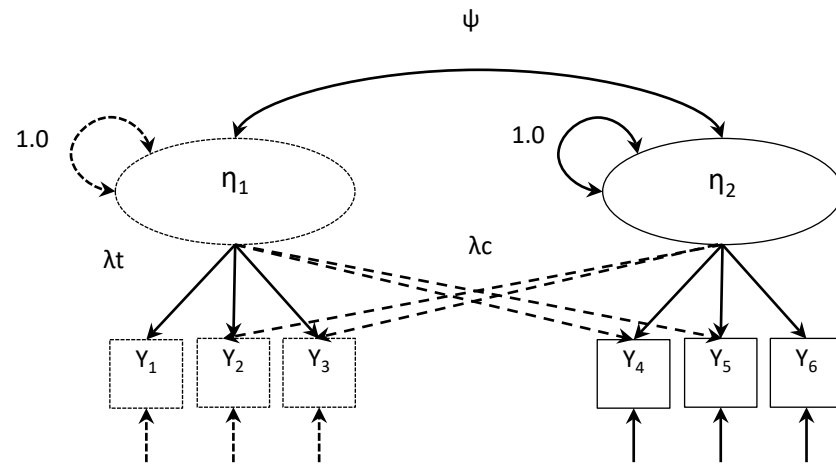
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Introduction

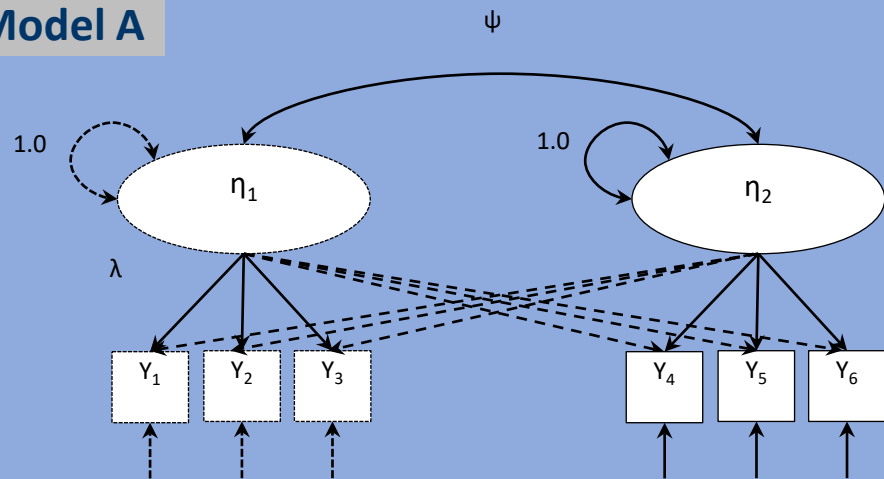


Introduction



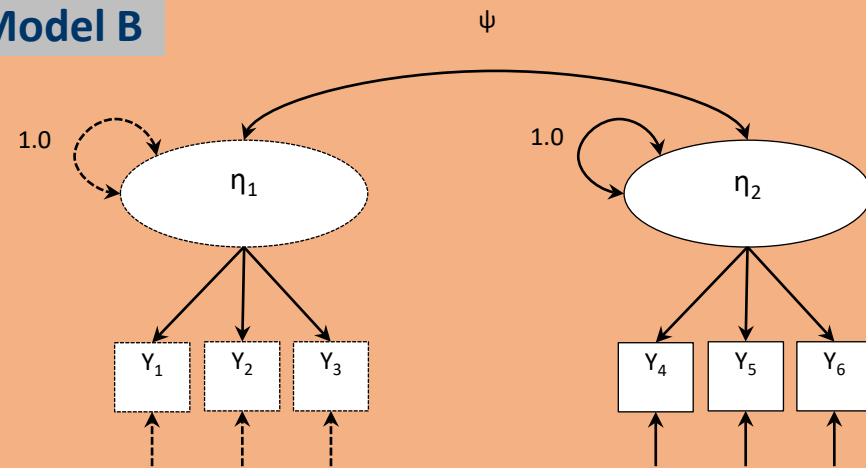
$\varepsilon =$

Model A



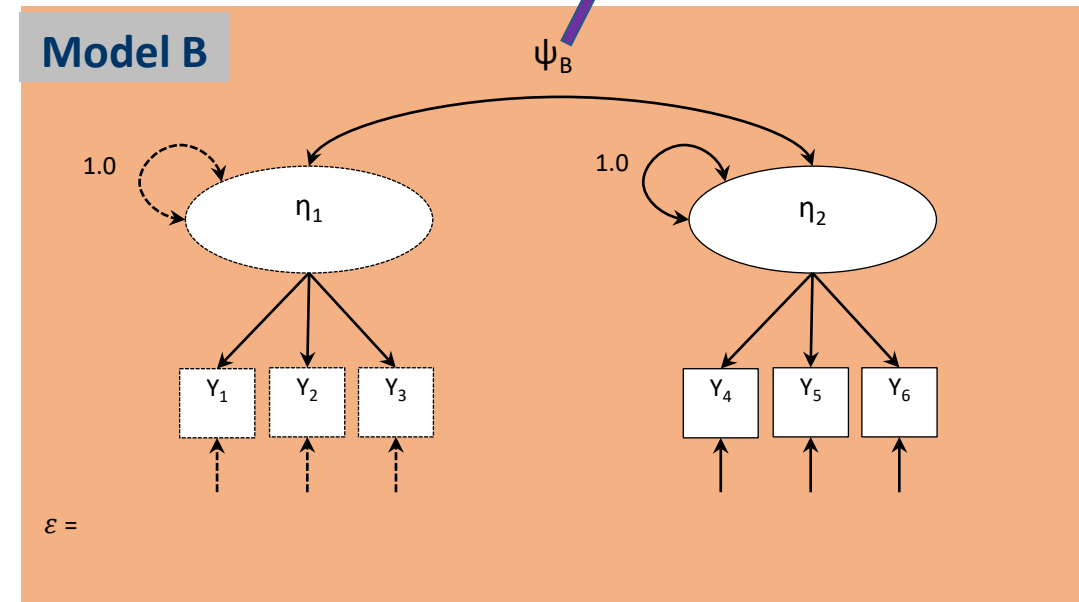
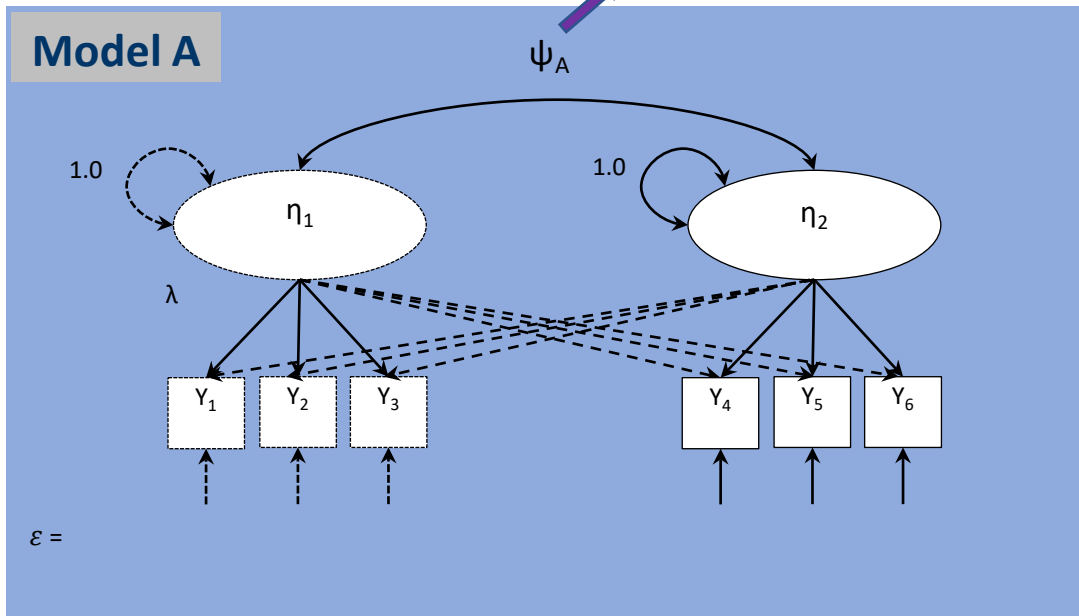
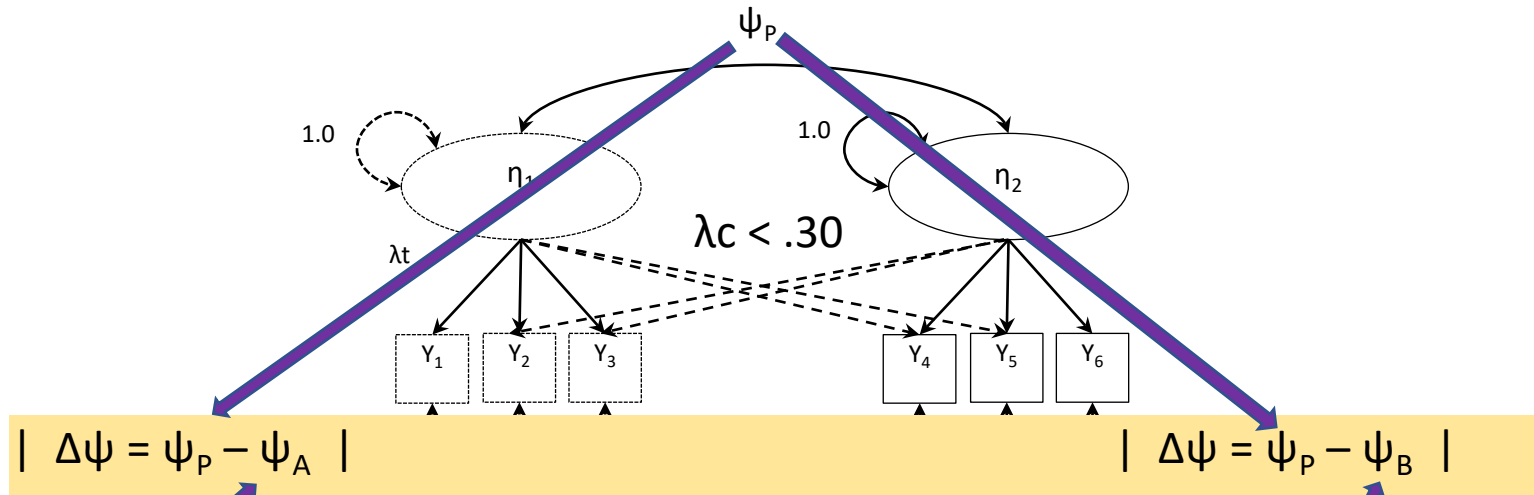
$\varepsilon =$

Model B

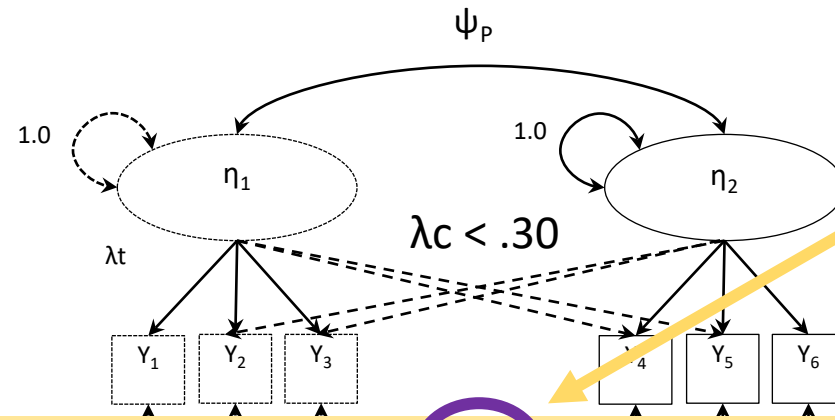


$\varepsilon =$

Introduction



Introduction



Asparouhov & Muthen (2009)

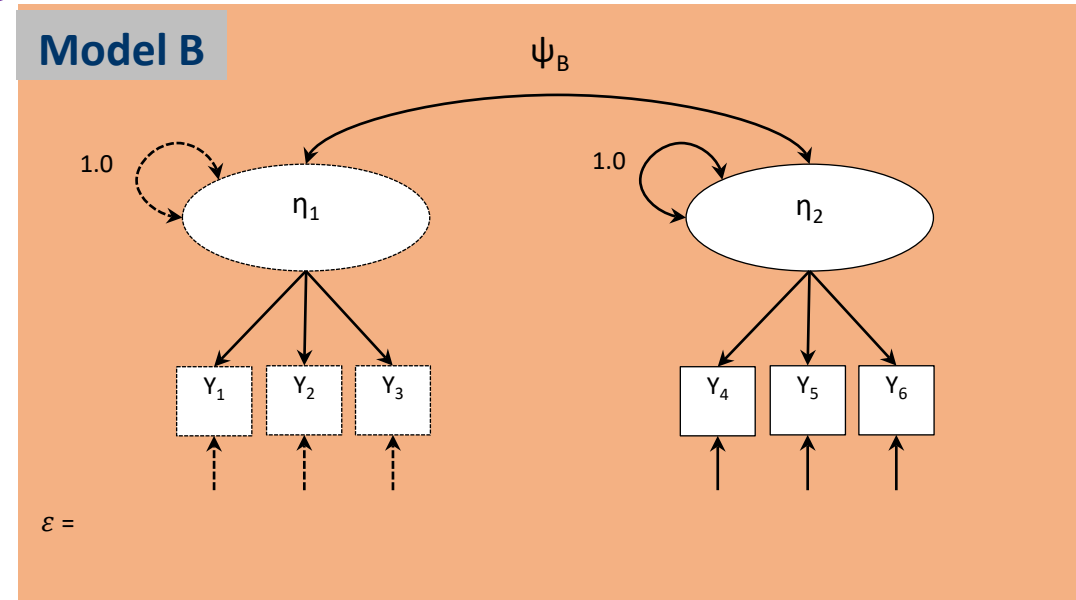
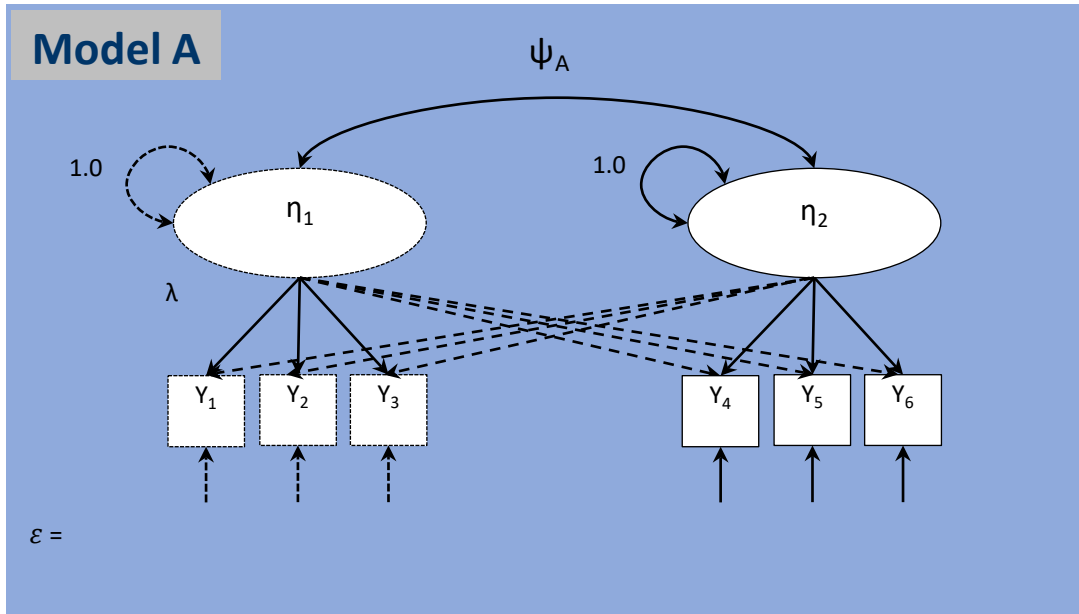
Marsh et al., (2009...2020)

Steenkamp & Maydeu-Olivares, (2023)

$$| \Delta\psi = \psi_P - \psi_A |$$

<

$$| \Delta\psi = \psi_P - \psi_B |$$



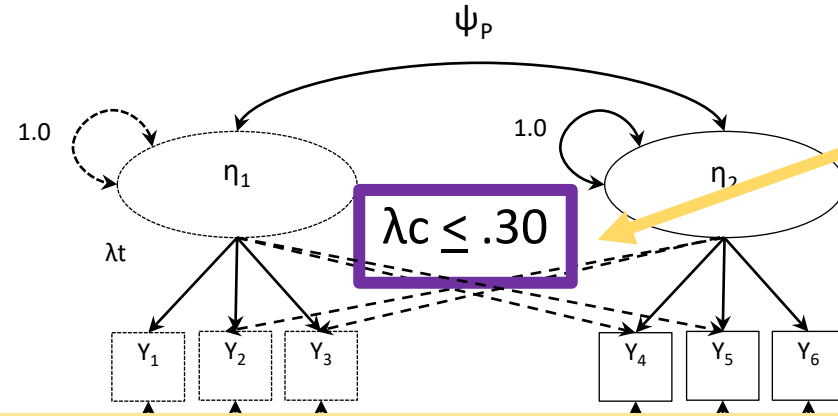
Introduction

Tabachnick & Fidell, 2007

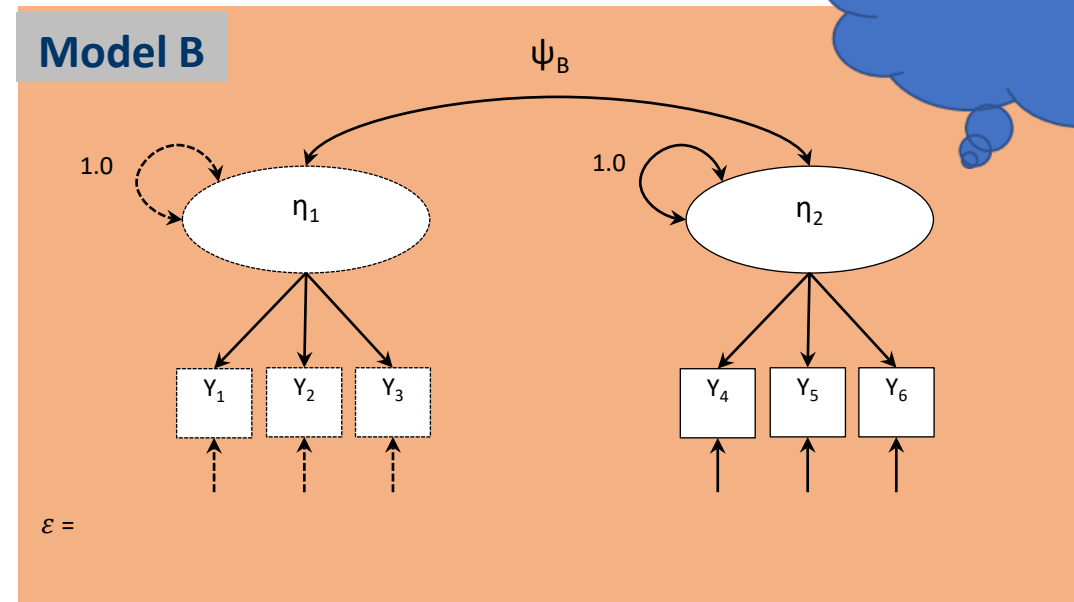
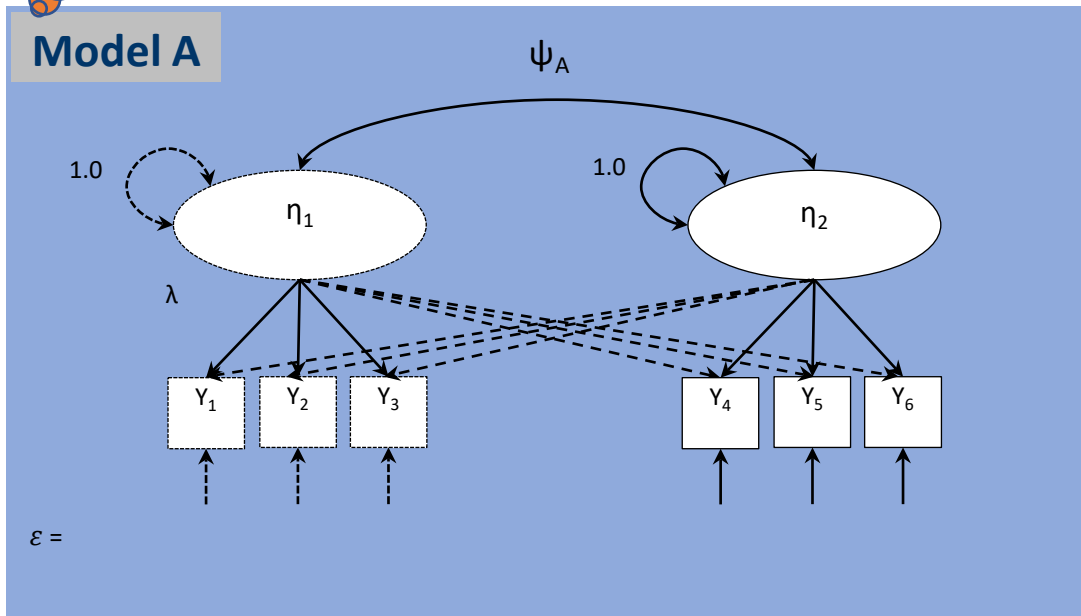
Cudeck & O'Dell, 1994

van Prooijen & van der Kloot, 2001

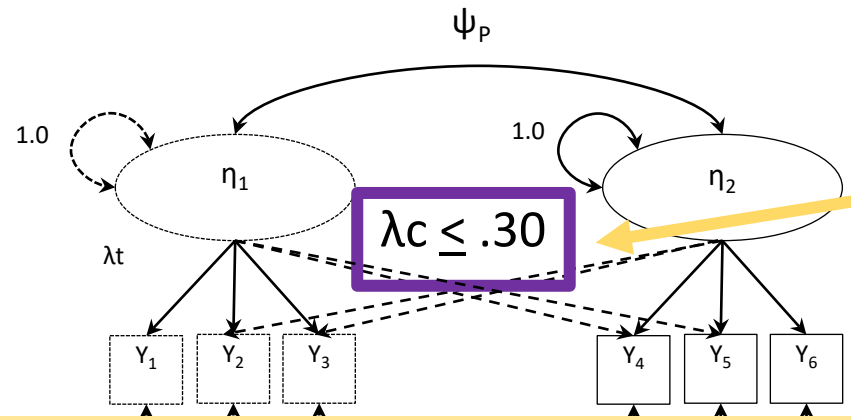
Grice, 2010



$$| \Delta\psi = \psi_P - \psi_A | < | \Delta\psi = \psi_P - \psi_B |$$



Introduction



Hsu, et al. (2014)

λ_c of $|.13| \rightarrow 0$; \rightarrow biased ψ_B

Steenkamp & Maydeu-Olivares (2023)

λ_c of $|.2| \rightarrow 0$;

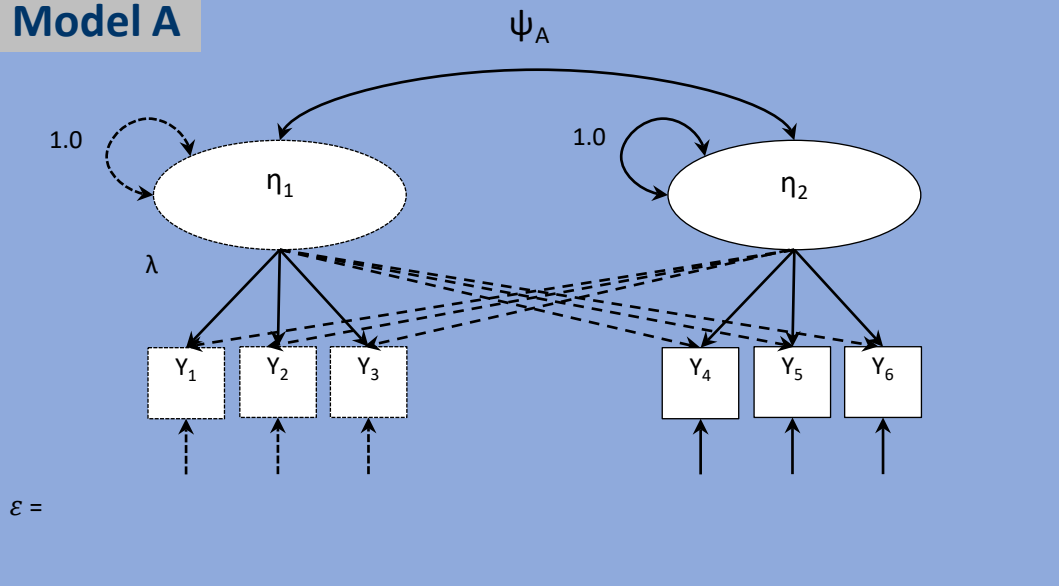
\rightarrow Non-negligible impact on ψ_B

λ_c of $|.3| \rightarrow 0$;

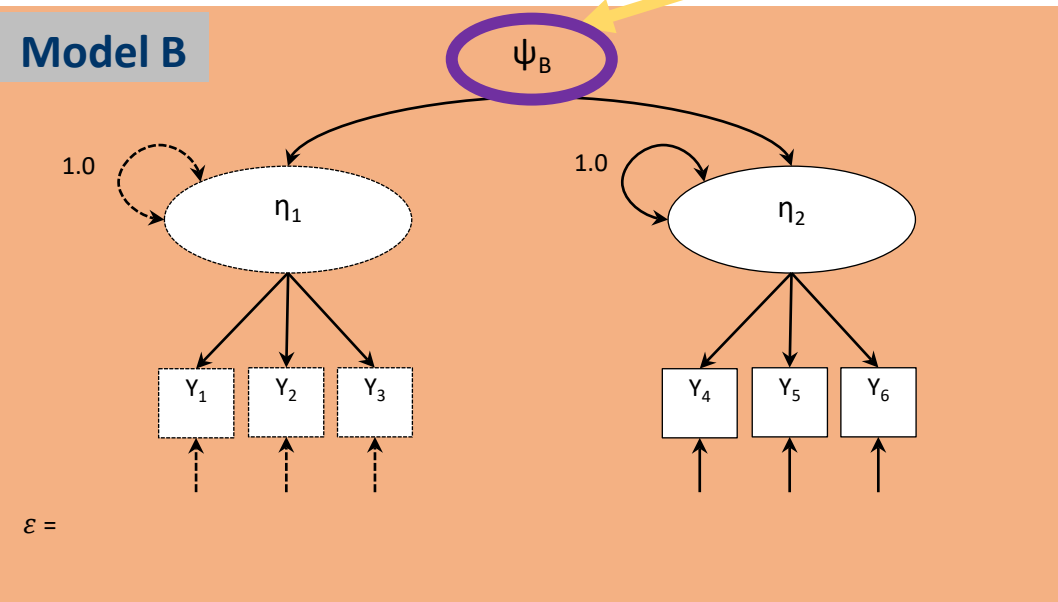
\rightarrow "appreciable" bias in ψ_B

$$|\Delta\psi = \psi_P - \psi_A| < |\Delta\psi = \psi_B - \psi_A|$$

Model A



Model B



Introduction

What to do? How do we balance these advantages?

ESEM / UFA

Better “Fit”, less likely to reject ‘trivially’ mis-specified models

More realistic evaluation of simple structure

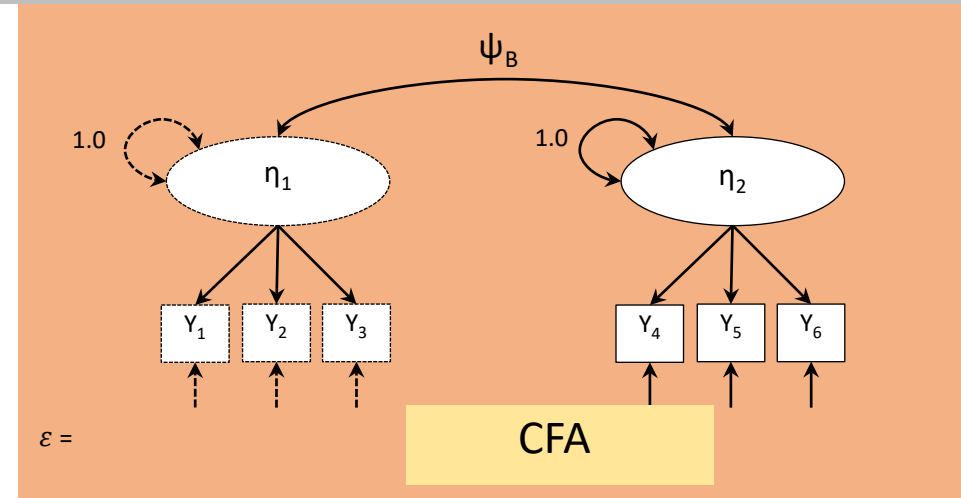
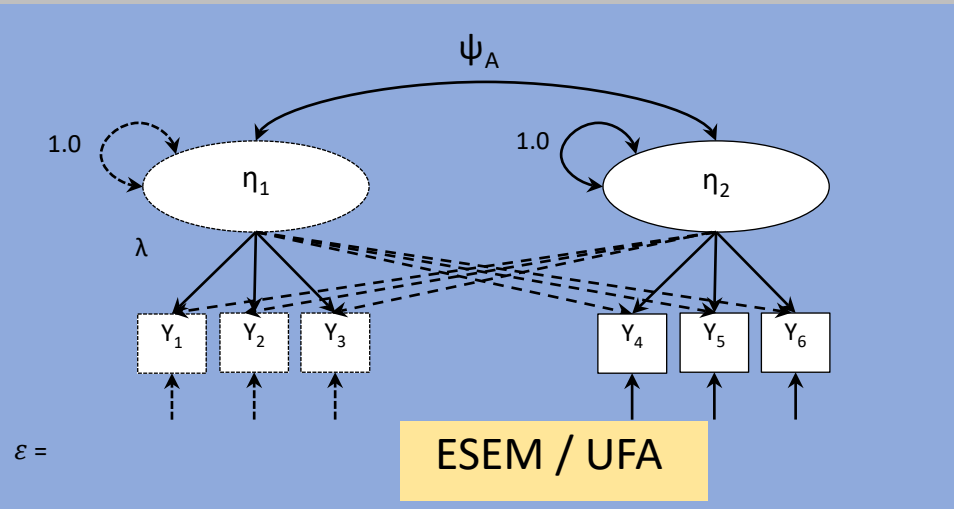
Less biased structural relationships

Allows for a more approximate hypothesis of the measurement structure

~~Single solution that is invariant to type of rotation~~

CFA

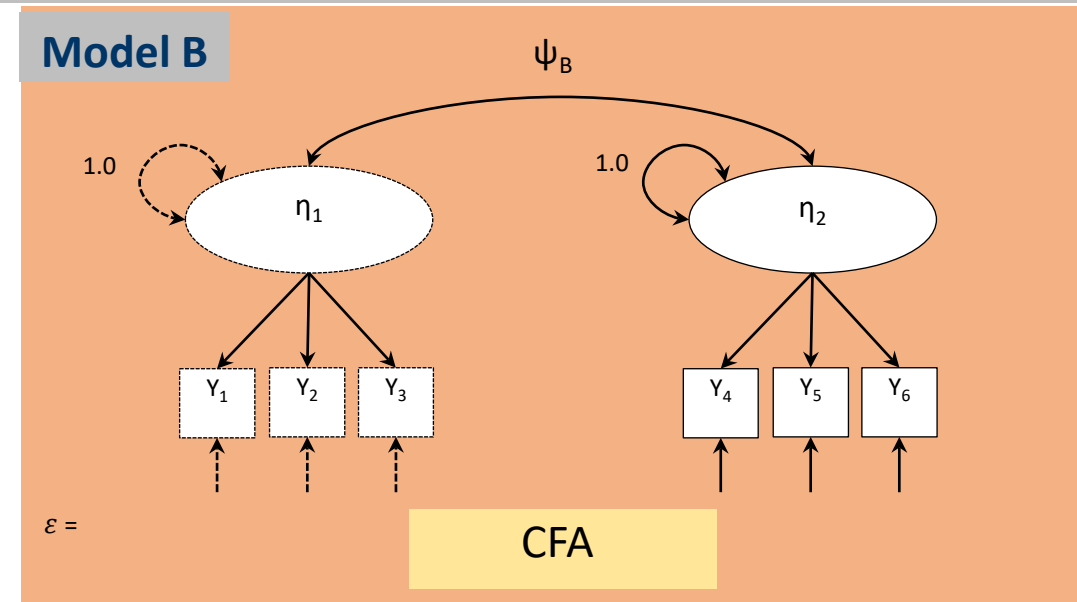
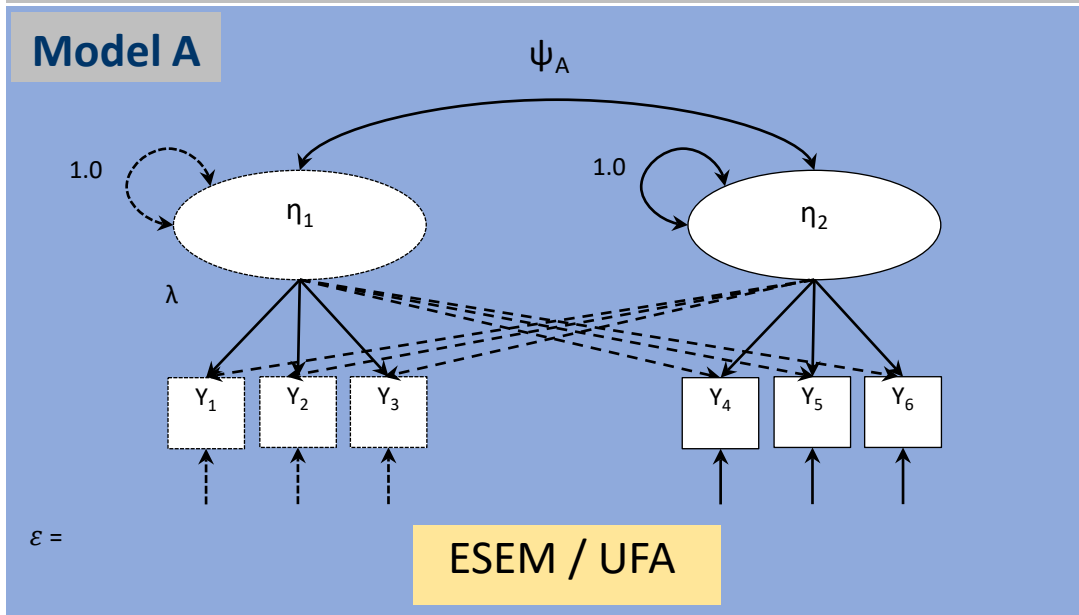
Parsimony



Introduction

A strategy (Steenkamp & Maydeu-Olivares, 2023)

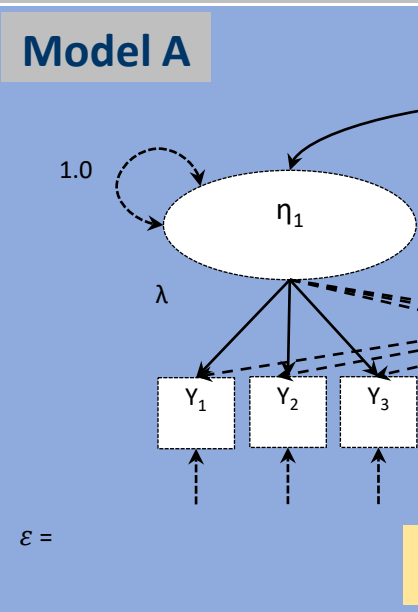
- ~Estimate both UFA and CFA models
- ~Compare model estimates (e.g., factor correlations, factor loadings)
- ~IFF LRT rejects the CFA model...
- ~Evaluate overall fit of each (LRT, CFI, RMSEA)
- ~Evaluate nested model comparisons (Δ LRT, Δ CFI, Δ RMSEA)



Introduction

A strategy (Steenkamp & Maydeu-Olivares, 2023)

- ~Estimate both UFA and CFA models
- ~Compare model estimates (e.g., factor correlations, factor loadings)
- ~Evaluate overall fit of each (LRT, CFI, RMSEA)
- ~Evaluate nested model comparisons (Δ LRT, Δ CFI, Δ RMSEA)



Remain the most frequently reported measures of fit in SEM (Jackson et al., 2009)

Have been emphasized in recent methodological evaluations of UFA (Steenkamp & Maydeu-Olivares, 2023; Marsh, et al., 2020)

with van Zyl and Klooster (2022) noting that in evaluations of UFA “the CFI, TLI, and RMSEA should always be reported and used as the primary criterion for both establishing model fit and to discriminate between models” (p. 9)

-
- **Briefly revisit structural bias when non-zero cross loadings in the population are constrained to zero**
 - **Examine the degree to which GFI and Δ GFI measures are sensitive to what has been considered ignorable cross loadings**
 - **Examine GFI and Δ GFI (in comparison to ESEM/UFA specification) measures relative to the CFA models that give rise to structural relationship bias and those that do not**

Evaluating Fit

Stand-Alone Models

$$LR = (N - 1)Fml \sim \chi_{df_H}^2$$

$$CFI = 1 - \frac{\max[0, (\chi_H^2 - df_H)]}{\max[0, (\chi_B^2 - df_B)]}$$

$$RMSEA = \sqrt{\max\left[0, \frac{\chi_H^2 - df_H}{df_H \times (N)}\right]}$$

Nested Models

$$\Delta LR = LR_B - LR_A \sim \chi_{df_B - df_A}^2$$

$$\Delta CFI = CFI_A - CFI_B$$

$$\Delta RMSEA = RMSEA_A - RMSEA_B$$

$$RMSEA_D = \sqrt{\frac{\Delta LR - df_D}{df_D (N - 1)}}$$

RMSEA_D (Browne & Du, 1992), recently re-introduced (Savalei et al., 2023)

- ~Integration of ΔLR and $\Delta RMSEA$ (replaces NCP in RMSEA with $\Delta LR - df_D$; and model df with df_D)
- ~Purports to overcome limitations of ΔLR (too restrictive) and $\Delta RMSEA$ (too forgiving)
- ~Said to better focus on changes in the two nested models (relative to df) rather than common aspects
- ~As $n \rightarrow N$, RMSEA_D \rightarrow population value; rather than being overpowered for trivial differences
- ~On same scale as RMSEA, and provides for CIs

$$LR = (N - 1)Fml \sim \chi_{df_H}^2$$

$$\Delta LR = LR_B - LR_A \sim \chi_{df_B - df_A}^2$$

$$CFI = 1 - \frac{\max[0, (\chi_H^2 - df_H)]}{\max[0, (\chi_B^2 - df_B)]}$$

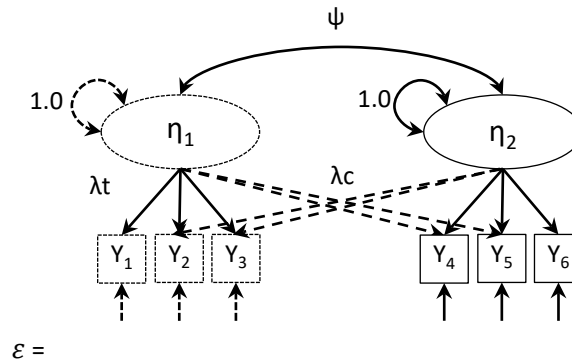
$$\Delta CFI = CFI_A - CFI_B$$

$$RMSEA = \sqrt{\max\left[0, \frac{\chi_H^2 - df_H}{df_H \times (N)}\right]}$$

$$\Delta RMSEA = RMSEA_A - RMSEA_B$$

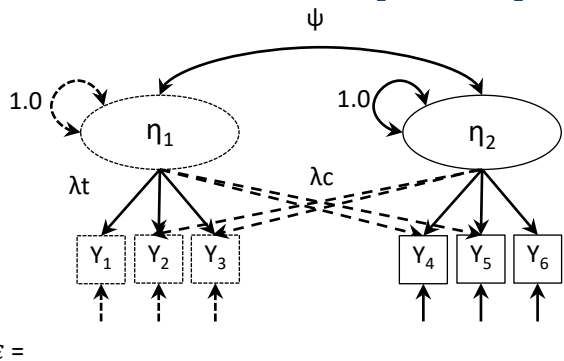
$$RMSEA_D = \sqrt{\frac{\Delta LR - df_D}{df_D (N - 1)}}$$

Population Model

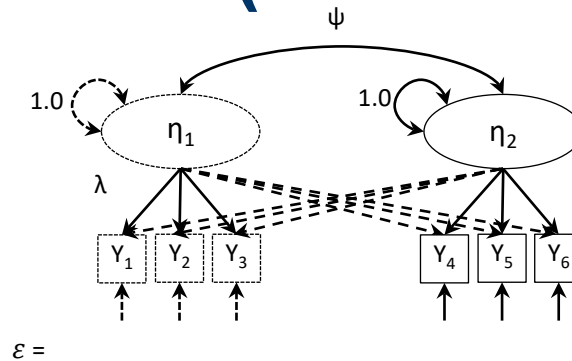


Two-Factor Models with unit normal factors and indicator scales

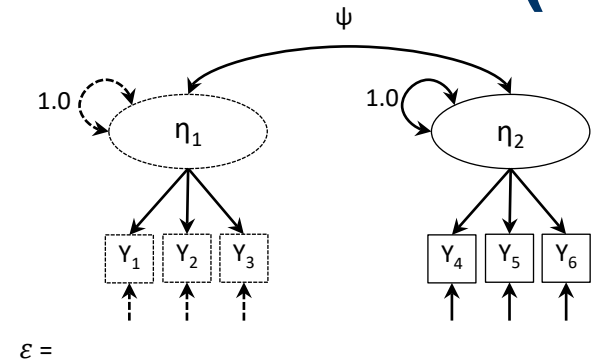
Correct (CS)



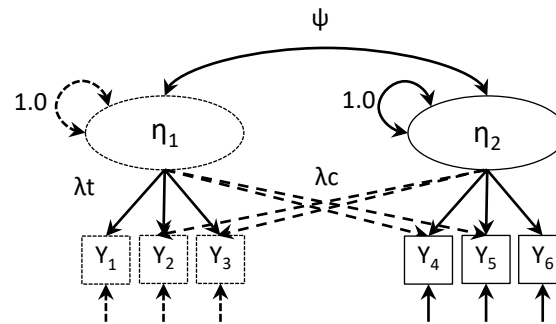
UFA (aka ESEM)



ICM-CFA (IS)



Population Model



With an Emphasis on Measurement Quality of the Target Indicators

6-, 12- and 24-indicator models;

Target Loadings: $\lambda_t = .40, .50, .60, .70, .80$;

Cross Loadings: $\lambda_c = .00, .15, .30$;

Cross Loading Saturation: 33%, 67%

Cross Loading Sign: Positive, Negative, Mixed¹

Factor Correlations: $\psi_{1,2} = 0, .1, .2, .3, .4, \text{ and } .5$;

Sample Sizes: $N = 200, 500, 1000$

crossed with

crossed with

nearly crossed with

crossed with

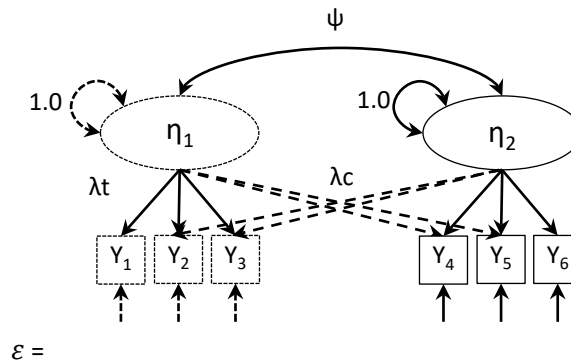
nearly crossed with

crossed with

($N = 1,000$ data sets were generated for the 4,860 conditions)

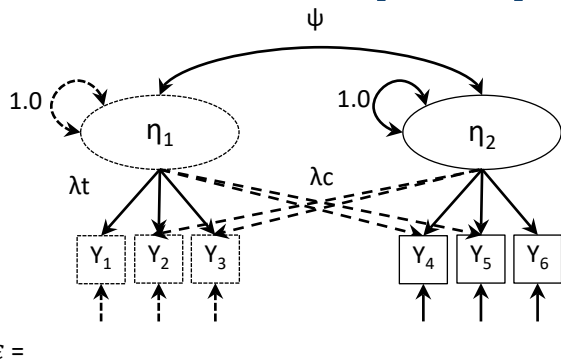
1. Absolute cross-loading magnitudes were held constant across signs (e.g., $\pm .15$ was only paired with $\pm .15$).

Population Model

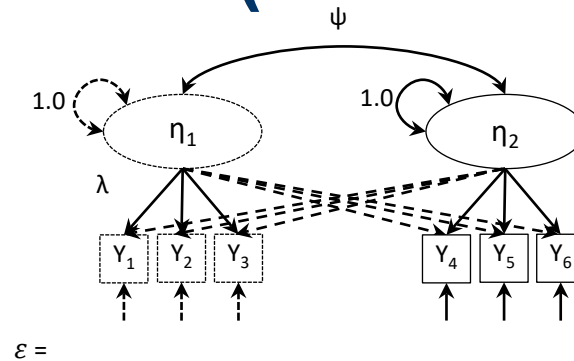


- Data were generated and analyzed using *Mplus 8* within the *MplusAutomation R* package
- All models were estimated with ML
- Results today based on UFA with Target rotation

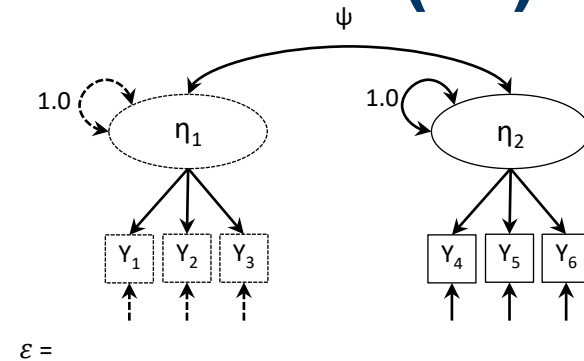
Correct (CS)



UFA (aka ESEM)



CFA (IS)



Results: Factor Correlation Bias

$$= \Psi_{\text{Population}} - \Psi_{\text{estimated model}}$$

Factor Correlation Raw Bias across Models for Selected Conditions: N = 1,000 Sample Size

Sample Size	Num Indic	F-F Cor	TL Mag	CL Mag	Pop Cor	33% of Indicators Cross-Load									67% of Indicators Cross-Load								
						All Positive CLs			All Negative CLs			Mix Pos/Neg CLs			All Positive CLs			All Negative CLs			Mix Pos/Neg CLs		
						CFA (CS)	UFA	CFA	CFA (CS)	UFA	CFA	CFA (CS)	UFA	CFA	CFA (CS)	UFA	CFA	CFA (CS)	UFA	CFA	CFA (CS)	UFA	CFA
1000	6	.0	.4	.15	.0	.00	.22	.27	.00	-.22	-.27	.00	.00	.00	.02	.39	.52	-.02	-.38	-.52	.00	.00	-.01
				.30	.0	.01	.38	.53	-.01	-.38	-.52	.00	.00	.00	.05	.44	.89	-.04	-.44	-.89	.00	.00	-.02
				.8	.15	.0	.00	.12	.13	.00	-.12	-.13	.00	.00	.00	.00	.24	.27	.00	-.24	-.26	.00	.00
		.5	.4	.15	.5	.00	.03	.17	.00	-.27	-.19	.00	-.09	.06	-.02	.07	.31	-.01	-.45	-.43	.01	-.11	.12
				.30	.5	-.01	.07	.29	.00	-.50	-.65	.00	-.13	.14	-.06	.03	.46	-.06	-.69	-1.19	.01	-.19	.33
				.8	.15	.5	.00	.08	.10	.00	-.11	-.08	.00	-.01	.06	.00	.16	.19	.00	-.21	-.17	.00	-.02
	24	.0	.4	.15	.0	.00	.23	.26	.00	-.23	-.26	.00	.00	.00	.00	.42	.50	.00	-.42	-.50	.00	.00	.00
				.30	.0	.00	.40	.51	.00	-.40	-.51	.00	.00	.00	.00	.59	.89	.00	-.59	-.88	.00	.00	.00
				.8	.15	.0	.00	.12	.13	.00	-.12	-.13	.00	.00	.00	.00	.24	.26	.00	-.24	-.26	.00	.00
		.5	.4	.15	.5	.00	.12	.17	.00	-.24	-.20	.00	-.05	.07	.00	.23	.31	.00	-.45	-.43	.00	-.07	.13
				.30	.5	.00	.19	.29	.00	-.50	-.51	.00	-.10	.23	.01	.28	.46	.00	-.76	-1.11	.00	-.16	.40
				.8	.15	.5	.00	.08	.10	.00	-.11	-.09	.00	-.01	.04	.00	.16	.19	.00	-.21	-.19	.00	-.02
			.30	.5	.00	.14	.21	.00	-.23	.32	.00	-.03	.19	.00	.26	.37	.00	-.45	-.38	.00	-.06	.35	

Results: Factor Correlation Bias

$$= \Psi_{\text{Population}} - \Psi_{\text{estimated model}}$$

Factor Correlation Raw Bias across Models for Selected Conditions: N = 1,000 Sample Size

Sample Size	Num Indic	F-F Cor	TL Mag	CL Mag	Pop Cor	33% of Indicators Cross-Load									67% of Indicators Cross-Load								
						All Positive CLs			All Negative CLs			Mix Pos/Neg CLs			All Positive CLs			All Negative CLs			Mix Pos/Neg CLs		
						CFA (CS)	UFA	CFA	CFA (CS)	UFA	CFA	CFA (CS)	UFA	CFA	CFA (CS)	UFA	CFA	CFA (CS)	UFA	CFA	CFA (CS)	UFA	CFA
1000	6	.0	.4	.15	.0	.00	.22	.27	.00	-.22	-.27	.00	.00	.00	.02	.39	.52	-.02	-.38	-.52	.00	.00	-.01
				.30	.0	.01	.38	.53	-.01	-.38	-.52	.00	.00	.00	.05	.44	.89	-.04	-.44	-.89	.00	.00	-.02
				.8	.15	.0	.00	.12	.13	.00	-.12	-.13	.00	.00	.00	.00	.24	.27	.00	-.24	-.26	.00	.00
		.5	.4	.15	.5	.00	.03	.17	.00	-.27	-.19	.00	-.09	.06	-.02	.07	.31	-.01	-.45	-.43	.01	-.11	.12
				.30	.5	-.01	.07	.29	.00	-.50	-.65	.00	-.13	.14	-.06	.03	.46	-.06	-.69	-1.19	.01	-.19	.33
				.8	.15	.5	.00	.08	.10	.00	-.11	-.08	.00	-.01	.06	.00	.16	.19	.00	-.21	-.17	.00	-.02
	24	.0	.4	.15	.0	.00	.23	.26	.00	-.23	-.26	.00	.00	.00	.00	.42	.50	.00	-.42	-.50	.00	.00	.00
				.30	.0	.00	.40	.51	.00	-.40	-.51	.00	.00	.00	.00	.59	.89	.00	-.59	-.88	.00	.00	.00
				.8	.15	.0	.00	.12	.13	.00	-.12	-.13	.00	.00	.00	.00	.24	.26	.00	-.24	-.26	.00	.00
		.5	.4	.15	.5	.00	.12	.17	.00	-.24	-.20	.00	-.05	.07	.00	.23	.31	.00	-.45	-.43	.00	-.07	.13
				.30	.5	.00	.19	.29	.00	-.50	-.51	.00	-.10	.23	.01	.28	.46	.00	-.76	-1.11	.00	-.16	.40
				.8	.15	.5	.00	.08	.10	.00	-.11	-.09	.00	-.01	.04	.00	.16	.19	.00	-.21	-.19	.00	-.02
			.30	.5	.00	.14	.21	.00	-.23	.32	.00	-.03	.19	.00	.26	.37	.00	-.45	-.38	.00	-.06	.35	

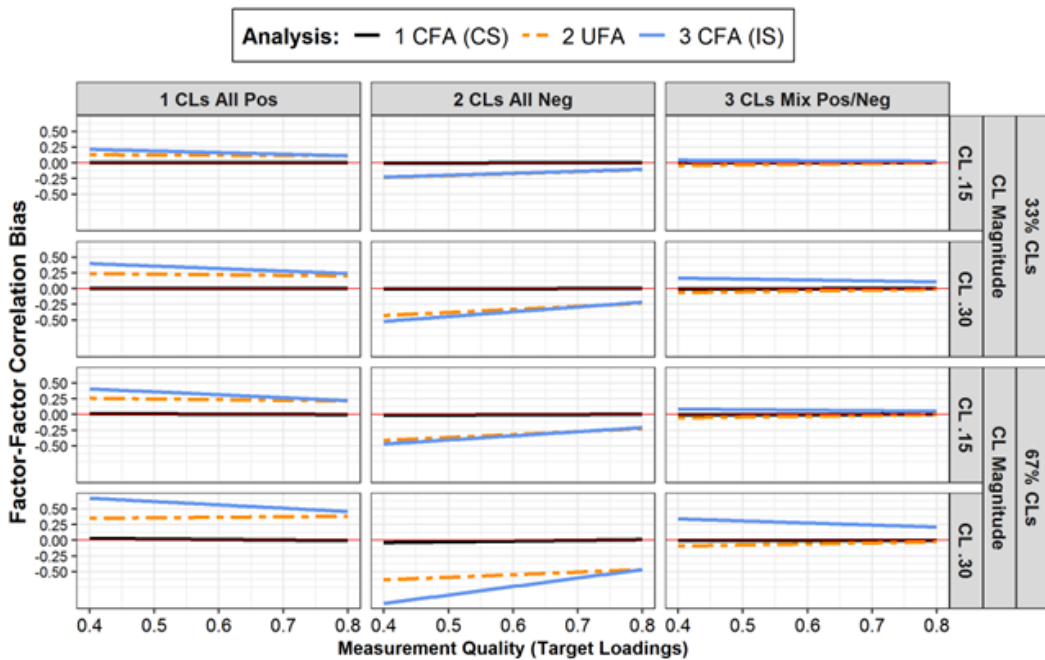
$$\text{Bias} = \Psi_{\text{Population}} - \Psi_{\text{estimated model}}$$

Factor Correlation Raw Bias across Models for Selected Conditions: N = 1,000 Sample Size

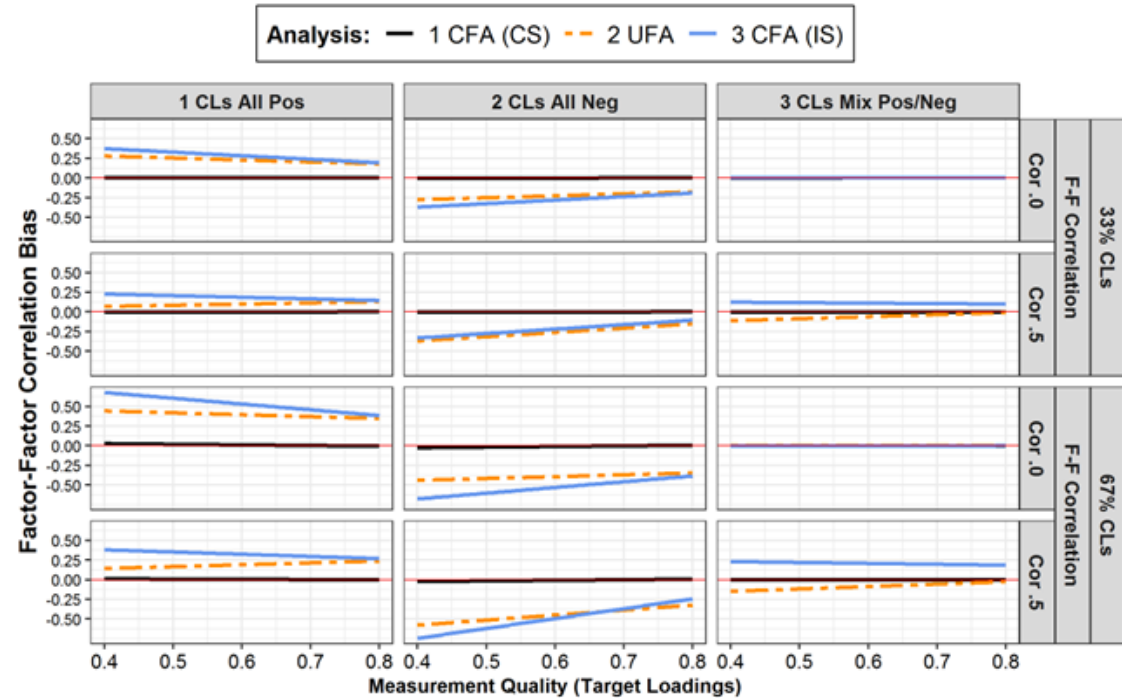
Sample Size	Num Indic	F-F Cor	TL Mag	CL Mag	Pop Cor	33% of Indicators Cross-Load									67% of Indicators Cross-Load										
						All Positive CLs			All Negative CLs			Mix Pos/Neg CLs			All Positive CLs			All Negative CLs			Mix Pos/Neg CLs				
						CFA (CS)	CFA UFA	CFA (IS)	CFA (CS)	CFA UFA	CFA (IS)	CFA (CS)	CFA UFA	CFA (IS)	CFA (CS)	CFA UFA	CFA (IS)	CFA (CS)	CFA UFA	CFA (IS)	CFA (CS)	CFA UFA	CFA (IS)		
1000	6	.0	.4	.15	.0	.00	.22	.27	.00	-.22	-.27	.00	.00	.00	.02	.39	.52	-.02	-.38	-.52	.00	.00	.00		
				.30	.0	.01	.38	.53	-.01	-.38	-.52	.00	.00	.00	.05	.44	.89	-.04	-.44	-.89	.00	.00	.01		
				.8	.15	.00	.12	.13	.00	-.12	-.13	.00	.00	.00	.00	.24	.27	.00	-.24	-.26	.00	.00	.00		
		.5	.4	.15	.5	.00	.03	.67	.00	-.27	.31	.00	-.09	.56	-.02	.07	.81	-.01	-.45	.07	.01	-.11	.62		
				.30	.5	-.01	.07	.79	.00	-.50	-.15	.00	-.13	.64	-.06	.03	.96	-.06	-.69	-.69	.01	-.19	.83		
				.8	.15	.5	.00	.08	.60	.00	-.11	.42	.00	-.01	.56	.00	.16	.69	.00	-.21	.33	.00	-.02	.61	
	.30		.5	.00	.14	.71	.00	-.23	.36	.00	-.03	.66	.00	.26	.87	.00	-.45	.10	.00	-.06	.77				
			24	.0	.4	.15	.0	.00	.23	.26	.00	-.23	-.26	.00	.00	.00	.00	.42	.50	.00	-.42	-.50	.00	.00	.00
						.30	.0	.00	.40	.51	.00	-.40	-.51	.00	.00	.00	.00	.59	.89	.00	-.59	-.88	.00	.00	.00
	.8	.15				.0	.00	.12	.13	.00	-.12	-.13	.00	.00	.00	.00	.24	.26	.00	-.24	-.26	.00	.00	.00	
	.5	.4		.15	.5	.00	.12	.67	.00	-.24	.30	.00	-.05	.57	.00	.23	.81	.00	-.45	.07	.00	-.07	.63		
				.30	.5	.00	.19	.79	.00	-.50	-.01	.00	-.10	.73	.01	.28	.96	.00	-.76	-.61	.00	-.16	.90		
.8				.15	.5	.00	.08	.60	.00	-.11	.41	.00	-.01	.54	.00	.16	.69	.00	-.21	.31	.00	-.02	.59		
.30	.5	.00	.14	.71	.00	-.23	.32	.00	-.03	.69	.00	.26	.87	.00	-.45	.12	.00	-.06	.85						

Factor-Factor Correlation Bias across Analysis

CL Magnitude








F-F Correlation



$$= \psi_{\text{Population}} - \psi_{\text{CFA}}$$

Partial Eta-Squared Values
for Raw CFA Factor-Factor
Correlation Bias

Population Design Conditions	Raw FF R Bias	
Main Effects		
N	0.00	
Model Size	0.01	
Factor Correlation (FC)	0.21	
Target Loadings (TL)	0.20	
Cross Loading (CL) Value	0.20	
CL Sign 	<u>1.00</u>	
CL Saturation	0.01	
Two-Way Interactions		
N * Model Size	0.00	
N * FC	0.00	
N * TL	0.00	
N * CL Value	0.00	
N * CL Sign	0.00	
N * CL Saturation	0.00	
Model Size * FC	0.01	
Model Size * TL	0.02	
Model Size * CL Value	0.03	
Model Size * CL Sign	0.04	
Model Size * CL Saturation	0.00	
FC * TL	<u>0.26</u>	
FC * CL Value	0.12	
FC * CL Sign 	<u>0.85</u>	
FC * CL Saturation	0.07	
TL * CL Value	0.11	
TL * CL Sign 	<u>0.92</u>	
TL * CL Saturation	0.04	
CL Value * CL Sign 	<u>0.96</u>	
CL Value * CL Saturation	0.00	
CL Sign * CL Saturation 	<u>0.95</u>	

Sample Size	Num Indic	F-F Cor	TL Mag	CL Mag	All Positive Cross-Loadings				All Negative Cross-Loadings				Mix of Pos/Neg Cross-Loadings			
					Analysis		UFA vs.		Analysis		UFA vs.		Analysis		UFA vs.	
					CFA (CS)	UFA	CFA (IS)	CFA (IS)	CFA (CS)	UFA	CFA (IS)	CFA (IS)	CFA (CS)	UFA	CFA (IS)	CFA (IS)
1000	6	.0	.4	.15	.008	.008	.026	-.018	.008	.008	.025	-.017	.009	.009	.066	-.057
				.30	.007	.007	.029	-.022	.006	.006	.028	-.022	.010	.010	.122	-.113
			.8	.010	.010	.090	-.080	.009	.009	.089	-.080	.010	.010	.162	-.153	
		.5	.4	.15	.010	.010	.164	-.154	.009	.009	.163	-.154	.010	.010	.326	-.316
				.30	.006	.006	.014	-.008	.008	.008	.030	-.022	.009	.009	.037	-.028
			.8	.004	.004	.015	-.011	.007	.007	.043	-.036	.009	.009	.065	-.056	
	24	.0	.4	.15	.010	.010	.071	-.061	.009	.009	.076	-.067	.010	.010	.123	-.113
				.30	.010	.010	.158	-.148	.009	.009	.155	-.146	.010	.010	.211	-.201
			.8	.004	.004	.018	-.014	.004	.004	.018	-.014	.004	.004	.035	-.031	
		.5	.4	.15	.004	.004	.025	-.021	.004	.004	.025	-.021	.004	.004	.072	-.068
				.30	.004	.004	.035	-.031	.004	.004	.035	-.031	.004	.004	.063	-.059
			.8	.004	.004	.069	-.065	.004	.004	.069	-.065	.004	.004	.131	-.127	
.8	.4	.15	.004	.004	.011	-.007	.004	.004	.018	-.014	.004	.004	.026	-.022		
		.30	.004	.004	.014	-.010	.004	.004	.033	-.029	.004	.004	.044	-.040		
	.8	.004	.004	.030	-.026	.004	.004	.031	-.027	.004	.004	.053	-.049			
			.30	.004	.004	.059	-.055	.004	.004	.062	-.058	.004	.004	.110	-.106	

Sample Size	Num Indic	F-F Cor	TL Mag	CL Mag	All Positive Cross-Loadings				All Negative Cross-Loadings				Mix of Pos/Neg Cross-Loadings			
					Analysis		UFA vs.		Analysis		UFA vs.		Analysis		UFA vs.	
					CFA (CS)	UFA	CFA (IS)	CFA (IS)	CFA (CS)	UFA	CFA (IS)	CFA (IS)	CFA (CS)	UFA	CFA (IS)	CFA (IS)
1000	6	.0	.4	.15	.008	.008	.026	-.018	.008	.008	.025	-.017	.009	.009	.066	-.057
				.30	.007	.007	.029	-.022	.006	.006	.028	-.022	.010	.010	.122	-.113
			.8	.010	.010	.090	-.080	.009	.009	.089	-.080	.010	.010	.162	-.153	
		.5	.4	.15	.010	.010	.164	-.154	.009	.009	.163	-.154	.010	.010	.326	-.316
				.30	.006	.006	.014	-.008	.008	.008	.030	-.022	.009	.009	.037	-.028
			.8	.004	.004	.015	-.011	.007	.007	.043	-.036	.009	.009	.065	-.056	
	24	.0	.4	.15	.010	.010	.071	-.061	.009	.009	.076	-.067	.010	.010	.123	-.113
				.30	.010	.010	.158	-.148	.009	.009	.155	-.146	.010	.010	.211	-.201
			.8	.004	.004	.018	-.014	.004	.004	.018	-.014	.004	.004	.035	-.031	
		.5	.4	.15	.004	.004	.025	-.021	.004	.004	.025	-.021	.004	.004	.072	-.068
				.30	.004	.004	.035	-.031	.004	.004	.035	-.031	.004	.004	.063	-.059
			.8	.004	.004	.069	-.065	.004	.004	.069	-.065	.004	.004	.131	-.127	
.5	.4	.15	.004	.004	.011	-.007	.004	.004	.018	-.014	.004	.004	.026	-.022		
		.30	.004	.004	.014	-.010	.004	.004	.033	-.029	.004	.004	.044	-.040		
	.8	.004	.004	.030	-.026	.004	.004	.031	-.027	.004	.004	.053	-.049			
				.30	.004	.004	.059	-.055	.004	.004	.062	-.058	.004	.004	.110	-.106

Across the 4,320 design conditions, average (across 1,000 replications) RMSEA values for all (100%) UFA models were $\leq .05$.

RMSEA: ($M = .04, SD = .03$) ($M = .05, SD = .03$), ($M = .08, SD = .05$)

Sample Size	Num Indic	F-F Cor	TL Mag	CL Mag	All Positive Cross-Loadings				All Negative Cross-Loadings				Mix of Pos/Neg Cross-Loadings			
					Analysis		UFA vs.	CFA (IS)	Analysis		UFA vs.	Analysis		UFA vs.		
					CFA (CS)	UFA	CFA (IS)		CFA (CS)	UFA	CFA (IS)	CFA (IS)	CFA (CS)	UFA	CFA (IS)	CFA (IS)
1000	6	.0	.4	.15	.008	.008	.026	-.018	.008	.008	.025	-.017	.009	.009	.066	-.057
				.30	.007	.007	.029	-.022	.006	.006	.028	-.022	.010	.010	.122	-.113
				.8	.010	.010	.090	-.080	.009	.009	.089	-.080	.010	.010	.162	-.153
		.5	.4	.15	.010	.010	.164	-.154	.009	.009	.163	-.154	.010	.010	.326	-.316
				.30	.006	.006	.014	-.008	.008	.008	.030	-.022	.009	.009	.037	-.028
				.8	.004	.004	.015	-.011	.007	.007	.043	-.036	.009	.009	.065	-.056
	24	.0	.4	.15	.010	.010	.071	-.061	.009	.009	.076	-.067	.010	.010	.123	-.113
				.30	.010	.010	.158	-.148	.009	.009	.155	-.146	.010	.010	.211	-.201
				.8	.004	.004	.018	-.014	.004	.004	.018	-.014	.004	.004	.035	-.031
		.5	.4	.15	.004	.004	.025	-.021	.004	.004	.025	-.021	.004	.004	.072	-.068
				.30	.004	.004	.035	-.031	.004	.004	.035	-.031	.004	.004	.063	-.059
				.8	.004	.004	.069	-.065	.004	.004	.069	-.065	.004	.004	.131	-.127
.8	.4	.15	.004	.004	.011	-.007	.004	.004	.018	-.014	.004	.004	.026	-.022		
		.30	.004	.004	.014	-.010	.004	.004	.033	-.029	.004	.004	.044	-.040		
		.8	.004	.004	.030	-.026	.004	.004	.031	-.027	.004	.004	.053	-.049		
				.30	.004	.004	.059	-.055	.004	.004	.062	-.058	.004	.004	.110	-.106

Across the 4,320 design conditions, average (across 1,000 replications) RMSEA values for 51% of the CFA models were $\leq .05$.

RMSEA: ($M = .04, SD = .03$) ($M = .05, SD = .03$), ($M = .08, SD = .05$)

Sample Size	Num Indic	F-F Cor	TL Mag	CL Mag	All Positive Cross-Loadings				All Negative Cross-Loadings				Mix of Pos/Neg Cross-Loadings					
					Analysis			UFA vs.	Analysis			UFA vs.	Analysis			UFA vs.		
					CFA (CS)	UFA	CFA (IS)	CFA (IS)	CFA (CS)	UFA	CFA (IS)	CFA (IS)	CFA (CS)	UFA	CFA (IS)	CFA (IS)		
1000	6	.0	.4	.15	.008	.008	.026	-.018	.008	.008	.025	-.017	.009	.009	.066	-.057		
				.30	.007	.007	.029	-.022	.006	.006	.028	-.022	.010	.010	.122	-.113		
				.8	.010	.010	.090	-.080	.009	.009	.089	-.080	.010	.010	.162	-.153		
				.15	.010	.010	.164	-.154	.009	.009	.163	-.154	.010	.010	.326	-.316		
			.5	.4	.15	.15	.006	.006	.014	-.008	.008	.008	.030	-.022	.009	.009	.037	-.028
						.30	.004	.004	.015	-.011	.007	.007	.043	-.036	.009	.009	.065	-.056
						.8	.010	.010	.071	-.061	.009	.009	.076	-.067	.010	.010	.123	-.113
						.15	.010	.010	.071	-.061	.009	.009	.076	-.067	.010	.010	.123	-.113

Many applied researchers might stop here, if they had begun with a CFA model and found an $RMSEA \leq .05$

.5	.4	.15	.15	.004	.004	.011	-.007	.004	.004	.018	-.014	.004	.004	.026	-.022
			.30	.004	.004	.014	-.010	.004	.004	.033	-.029	.004	.004	.044	-.040
			.8	.004	.004	.030	-.026	.004	.004	.031	-.027	.004	.004	.053	-.049
			.15	.004	.004	.059	-.055	.004	.004	.062	-.058	.004	.004	.110	-.106

Across the 4,320 design conditions, average (across 1,000 replications) RMSEA values for 51% of the CFA models were $\leq .05$.

RMSEA

RMSEA Results for Selected Conditions with 67% Cross-Loading: All Values Represent the Mean of 1,000 Replications

Sample Size	Num Indic	F-F Cor	TL Mag	CL Mag	All Positive Cross-Loadings				All Negative Cross-Loadings				Mix of Pos/Neg Cross-Loadings			
					Analysis			UFA vs. CFA (IS)	Analysis			UFA vs. CFA (IS)	Analysis			UFA vs. CFA (IS)
					CFA (CS)	UFA	CFA (IS)	CFA (IS)	CFA (CS)	UFA	CFA (IS)	CFA (IS)	CFA (CS)	UFA	CFA (IS)	CFA (IS)
1000	6	.0	.4	.15	.008	.008	.026	-.018	.008	.008	.025	-.017	.009	.009	.066	-.057
				.30	.007	.007	.029	-.022	.006	.006	.028	-.022	.010	.010	.122	-.113
			.8	.010	.010	.090	-.080	.009	.009	.089	-.080	.010	.010	.162	-.153	
		.5	.4	.15	.010	.010	.164	-.154	.009	.009	.163	-.154	.010	.010	.326	-.316
				.30	.006	.006	.014	-.008	.008	.008	.030	-.022	.009	.009	.037	-.028
			.8	.004	.004	.015	-.011	.007	.007	.043	-.036	.009	.009	.065	-.056	
	24	.0	.4	.15	.010	.010	.071	-.061	.009	.009	.076	-.067	.010	.010	.123	-.113
				.30	.010	.010	.158	-.148	.009	.009	.155	-.146	.010	.010	.211	-.201
			.8	.004	.004	.018	-.014	.004	.004	.018	-.014	.004	.004	.035	-.031	
		.5	.4	.15	.004	.004	.025	-.021	.004	.004	.025	-.021	.004	.004	.072	-.068
				.30	.004	.004	.035	-.031	.004	.004	.035	-.031	.004	.004	.063	-.059
			.8	.004	.004	.069	-.065	.004	.004	.069	-.065	.004	.004	.131	-.127	
.5	.4	.15	.004	.004	.011	-.007	.004	.004	.018	-.014	.004	.004	.026	-.022		

If they had not, and decided to conduct UFA anyway, only 15% (of the 4,320 different design conditions) would have found them equivalent ($\Delta RMSEA \geq -.015$)

DISCLAIMER

We do not advocate for the use of cutoffs when gauging the quality of models through GFI and Δ GFI metrics in applied work

We do so as a way of organizing and framing our results.

Results

*CFI Results for Selected Conditions with 67% Cross-Loading:
All Values Represent the Mean of 1,000 Replications*

CFA RMSEA: $M = .98, SD = .01$

$M = .95, SD = .03$

$M = .88, SD = .09$

Sample Size	Num Indic	F-F Cor	TL Mag	CL Mag	All Positive Cross-Loadings				All Negative Cross-Loadings				Mix of Pos/Neg Cross-Loadings				
					Analysis			UFA vs.	Analysis			UFA vs.	Analysis			UFA vs.	
					CFA (CS)	UFA	CFA (IS)	CFA (IS)	CFA (CS)	UFA	CFA (IS)	CFA (IS)	CFA (CS)	UFA	CFA (IS)	CFA (IS)	
1000	6	.0	.4	.15	.995	.995	.964	.031	.996	.996	.966	.030	.994	.994	.791	.203	
				.30	.998	.998	.978	.020	.998	.998	.979	.019	.996	.996	.541	.455	
			.8	.15	1.000	1.000	.975	.025	1.000	1.000	.975	.025	1.000	1.000	.919	.081	
				.30	1.000	1.000	.935	.065	1.000	1.000	.935	.065	1.000	1.000	.723	.277	
			.5	.4	.15	.998	.998	.992	.006	.991	.991	.915	.076	.996	.996	.942	.054
					.30	1.000	.999	.996	.003	.994	.994	.869	.125	.997	.997	.893	.104
		.8	.15	.15	1.000	1.000	.990	.010	.999	.999	.976	.023	1.000	1.000	.958	.042	
				.30	1.000	1.000	.977	.023	.999	.999	.882	.117	1.000	1.000	.916	.084	

We see similar patterns with the CFI and Δ CFI...

Across the 4,320 design conditions, over 99% of the average (across 1,000 replications) CFI values from UFA models were $\geq .95$

48% of the CFI values for the CFA models were $\geq .95$

Only 6% of the UFA vs. CFA contrasts would have found them equivalent (Δ CFI $\leq .01$)

Results

RMSEA_D Results Comparing UFA with Incorrectly Specified CFA for Selected Conditions: All Values Represent the Mean of 1,000 Replications

RMSEA _D Rejections:					83%			91%			99%			
					67% of Indicators Cross-Load									
					All Positive CLs			All Negative CLs			Mix of Pos/Neg CLs			
					90% CI			90% CI			90% CI			
Sample Num	F-F	TL	CL		M	LB	UB	M	LB	UB	M	LB	UB	
Size	Indic	Cor	Mag	Mag										
1000	6	.0	.4	.15	.042	[.028]	.041	[.027]	.094	[.073	, .124]	
				.30	.047	[.031]	.046	[.031]	.174	[.150	, .202]	
				.8	.15	.128	[.105	, .156]	.127	[.105	, .156]	.229	[.205	, .257]
				.30	.232	[.208	, .259]	.231	[.207	, .259]	.462	[.437	, .489]	
		.5	.4	.15	.027	[.019]	.047	[.031]	.056	[.038]	.087]
				.30	.030	[.020]	.064	[.045]	.093	[.072	, .123]	
			.8	.15	.102	[.080	, .131]	.109	[.086	, .138]	.174	[.150	, .202]	

As well with the RMSEA_D...

Across the 4,320 design conditions, 93% of the CFA models would be rejected (RMSEA_D > .05; Sareveli et al., 2023)



Should we be rejecting (CFA) models that constrain minor cross-loadings ($< .30$; Marsh, et al., 2020) to zero as some suggest for:

CFA evaluations of measurement structure (van Prooijen & van der Kloot, 2001)

(or) calculating factor scores (Grice, 2010)

(or) evaluating simple structure (Tabachnick & Fidell, 2007)

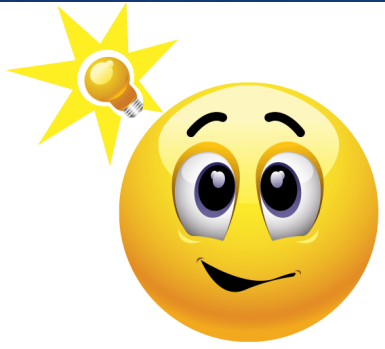


OR,

Should we consider that low cross loadings ($< |.30|$) when fixed to zero can produce meaningful structural bias (shown here and elsewhere in the literature)?

How might we quantify meaningful structural bias?

Examples of elsewhere: Hsu, et al., 2014; Marsh et al., 2013, 2014; Steenkamp & Maydeu-Olivares, 2023



How about this?

Drawing from recent methodological research in the context of UFA and ESEM, we follow Steenkamp & Maydeu-Olivares (2023) in asserting, on the basis of Cohen's (1988) recommendations, that factor correlation bias $\leq |0.1|$ “are unlikely to be meaningful”

We extend this here for purposes of framing our results around potentially meaningful values, to include bias values $\geq |.30|$ that have been characterized as a ‘medium effect’ (Cohen, 1988, p. 80) as nonignorable levels of factor correlation bias.

"Good Fit"

Bias	CFI CFI ≥ .95 (48%)	RMSEA RMSEA ≤ .05 (51%)	RMSEA _D
> .10	77%	76%	NA
> .30	30%	34%	NA
≤ .10	23%	24%	NA
≤ .30	70%	67%	NA

"Poor Fit"

Bias	CFI < .95 (52%)	RMSEA > .05 (49%)	RMSEA _D
> .10	66%	67%	NA
> .30	36%	33%	NA
≤ .10	34%	33%	NA
≤ .30	64%	66%	NA

CFA ~ UFA

Bias	ΔCFI ≤ .01 (6%)	ΔRMSEA ≥ -.015 (15%)	RMSEA _D ≤ .05 (7%)
> .10	95%	91%	97%
> .30	10%	32%	27%
≤ .10	5%	9%	3%
≤ .30	90%	68%	73%

CFA ≠ CFA

Bias	ΔCFI > .01 (94%)	ΔRMSEA < -.015 (85%)	RMSEA _D > .05 (93%)
> .10	70%	68%	70%
> .30	35%	33%	34%
≤ .10	30%	32%	30%
≤ .30	65%	67%	66%

- ~ 50% of the CFA models were found to provide a reasonable fit to the data
- Of these, a little over 20% had ignorable levels of bias (≤ .10)
- And ~ 70% had bias values at or below nonignorable levels (≤ .30)
- However, ~ 30% had nonignorable bias values (> .30)

Based on averages across the 4,320 different design conditions (each with 1,000 replications)

"Good Fit"

Bias	CFI CFI ≥ .95 (48%)	RMSEA RMSEA ≤ .05 (51%)	RMSEA _D
> .10	77%	76%	NA
> .30	30%	34%	NA
≤ .10	23%	24%	NA
≤ .30	70%	67%	NA

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Bias	CFI < .95 (52%)	RMSEA > .05 (49%)	RMSEA _D
> .10	66%	67%	NA
> .30	36%	33%	NA
≤ .10	34%	33%	NA
≤ .30	64%	66%	NA

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≤ .10	5%	9%	3%
≤ .30	90%	68%	73%

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≤ .10	30%	32%	30%
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> .10	77%	76%	NA
> .30	30%	34%	NA
≤ .10	23%	24%	NA
≤ .30	70%	67%	NA

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Bias	CFI < .95 (52%)	RMSEA > .05 (49%)	RMSEA _D
> .10	66%	67%	NA
> .30	36%	33%	NA
≤ .10	34%	33%	NA
≤ .30	64%	66%	NA

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> .10	95%	91%	97%
> .30	10%	32%	27%
≤ .10	5%	9%	3%
≤ .30	90%	68%	73%

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> .10	70%	68%	70%
> .30	35%	33%	34%
≤ .10	30%	32%	30%
≤ .30	65%	67%	66%

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- However, ~ 30% had nonignorable bias values (> .30)

Based on averages across the 4,320 different design conditions (each with 1,000 replications)

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Bias	CFI CFI ≥ .95 (48%)	RMSEA RMSEA ≤ .05 (51%)	RMSEA _D
> .10	77%	76%	NA
> .30	30%	34%	NA
≤ .10	23%	24%	NA
≤ .30	70%	67%	NA

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Bias	CFI < .95 (52%)	RMSEA > .05 (49%)	RMSEA _D
> .10	66%	67%	NA
> .30	36%	33%	NA
≤ .10	34%	33%	NA
≤ .30	64%	66%	NA

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> .10	95%	91%	97%
> .30	10%	32%	27%
≤ .10	5%	9%	3%
≤ .30	90%	68%	73%

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> .10	70%	68%	70%
> .30	35%	33%	34%
≤ .10	30%	32%	30%
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- However, ~ 30% had nonignorable bias values (> .30)

Based on averages across the 4,320 different design conditions (each with 1,000 replications)

	Bias	CFI	RMSEA	RMSEA _D
"Good Fit"		CFI ≥ .95 (48%)	RMSEA ≤ .05 (51%)	
	> .10	77%	76%	NA
	> .30	30%	34%	NA
	≤ .10	23%	24%	NA
	≤ .30	70%	67%	NA
"Poor Fit"		CFI < .95 (52%)	RMSEA > .05 (49%)	
	> .10	66%	67%	NA
	> .30	36%	33%	NA
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CFA ~ UFA		ΔCFI ≤ .01 (6%)	ΔRMSEA ≥ -.015 (15%)	RMSEA _D ≤ .05 (7%)
	> .10	95%	91%	97%
	> .30	10%	32%	27%
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CFA ≠ CFA		ΔCFI > .01 (94%)	ΔRMSEA < -.015 (85%)	RMSEA _D > .05 (93%)
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Note. Δ CFI did well here

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Based on averages across the 4,320 different design conditions (each with 1,000 replications)

CFA models deemed similar to their UFA counterparts by way of Δ GFI criteria were more likely to have the following design conditions:

Δ CFI:











Positive CL condition (75%)...and none of the mixed CL conditions
Lower (33%) CL saturations (76%)
Low (.15) CL values (91%)
High TL ($\geq .7$) conditions (54%)

Δ RMSEA:

Positive CL condition (55%)...with some mixed CL conditions (8%)
Lower CL saturations (69%)
Low CL values (87%)
Low target loadings (73%)

RMSEA_D

Positive CL condition (63%)...with some mixed CL conditions (3%)
Lower CL saturations (75%)
Low CL values (88%)
Low target loadings (66%)
Smaller models (50%)

Population Design Conditions	Raw FF R Bias	Δ CFI	Δ RMSEA	RMSEA _D
Main Effects				
N	0.00	0.04	<u>0.71</u>	0.04
Model Size	0.01	0.11	<u>0.96</u>	<u>0.73</u>
Factor Correlation (FC)	0.21	<u>0.61</u>	<u>0.75</u>	<u>0.88</u>
Target Loadings (TL)	0.20	<u>0.61</u>	<u>0.98</u>	<u>0.99</u>
Cross Loading (CL) Value	0.20	<u>0.92</u>	<u>0.98</u>	<u>0.99</u>
CL Sign 	<u>1.00</u> 	<u>0.96</u>	<u>0.98</u>	<u>0.99</u>
CL Saturation	0.01	<u>0.74</u>	<u>0.91</u>	<u>0.97</u>
Two-Way Interactions				
N * Model Size	0.00	0.01	0.11	0.00
N * FC	0.00	0.00	0.01	0.01
N * TL	0.00	0.02	0.08	0.03
N * CL Value	0.00	0.01	0.01	0.00
N * CL Sign	0.00	0.04	0.06	0.05
N * CL Saturation	0.00	0.00	0.00	0.00
Model Size * FC	0.01	0.08	<u>0.40</u>	0.08
Model Size * TL	0.02	0.12	<u>0.91</u>	<u>0.61</u>
Model Size * CL Value	0.03	0.01	<u>0.73</u>	<u>0.51</u>
Model Size * CL Sign	0.04	0.05	<u>0.81</u>	<u>0.47</u>
Model Size * CL Saturation	0.00	0.00	<u>0.43</u>	0.20
FC * TL	<u>0.26</u>	0.08	0.16	0.24
FC * CL Value	0.12	<u>0.27</u>	<u>0.24</u>	<u>0.44</u>
FC * CL Sign 	<u>0.85</u> 	<u>0.84</u>	<u>0.72</u>	<u>0.85</u>
FC * CL Saturation	0.07	0.06	0.05	0.12
TL * CL Value	0.11	0.05	<u>0.87</u>	<u>0.94</u>
TL * CL Sign 	<u>0.92</u> 	<u>0.63</u>	<u>0.80</u>	<u>0.91</u>
TL * CL Saturation	0.04	0.01	<u>0.67</u>	<u>0.83</u>
CL Value * CL Sign 	<u>0.96</u> 	<u>0.88</u>	<u>0.87</u>	<u>0.95</u>
CL Value * CL Saturation	0.00	0.25	<u>0.38</u>	<u>0.64</u>
CL Sign * CL Saturation 	<u>0.95</u> 	<u>0.70</u>	<u>0.70</u>	<u>0.86</u>

Partial Eta-Squared Values
for Raw Factor-Factor
Correlation Bias
and
UFA vs. CFA GFI Change
by Population Design
Condition

Population Design Conditions	Raw FF R Bias	Δ CFI	Δ RMSEA	RMSEA _D
Main Effects				
N	0.00	0.04	0.71	0.04
Model Size	0.01	0.11	0.96	0.73
Factor Correlation (FC)	0.21	0.61	0.75	0.88
Target Loadings (TL)	0.20	0.61	0.98	0.99
Cross Loading (CL) Value	0.20	0.92	0.98	0.99
CL Sign	1.00	0.96	0.98	0.99
CL Saturation	0.01	0.74	0.91	0.97
Two-Way Interactions				
N * Model Size	0.00	0.01	0.11	0.00
N * FC	0.00	0.00	0.01	0.01
N * TL	0.00	0.02	0.08	0.03
N * CL Value	0.00	0.01	0.01	0.00
N * CL Sign	0.00	0.04	0.06	0.05
N * CL Saturation	0.00	0.00	0.00	0.00
Model Size * FC	0.01	0.08	0.40	0.08
Model Size * TL	0.02	0.12	0.91	0.61
Model Size * CL Value	0.03	0.01	0.73	0.51
Model Size * CL Sign	0.04	0.05	0.81	0.47
Model Size * CL Saturation	0.00	0.00	0.43	0.20
FC * TL	0.26	0.08	0.16	0.24
FC * CL Value	0.12	0.27	0.24	0.44
FC * CL Sign	0.85	0.84	0.72	0.85
FC * CL Saturation	0.07	0.06	0.05	0.12
TL * CL Value	0.11	0.05	0.87	0.94
TL * CL Sign	0.92	0.63	0.80	0.91
TL * CL Saturation	0.04	0.01	0.67	0.83
CL Value * CL Sign	0.96	0.88	0.87	0.95
CL Value * CL Saturation	0.00	0.25	0.38	0.64
CL Sign * CL Saturation	0.95	0.70	0.70	0.86

Partial Eta-Squared Values
for Raw Factor-Factor
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Closing Thoughts

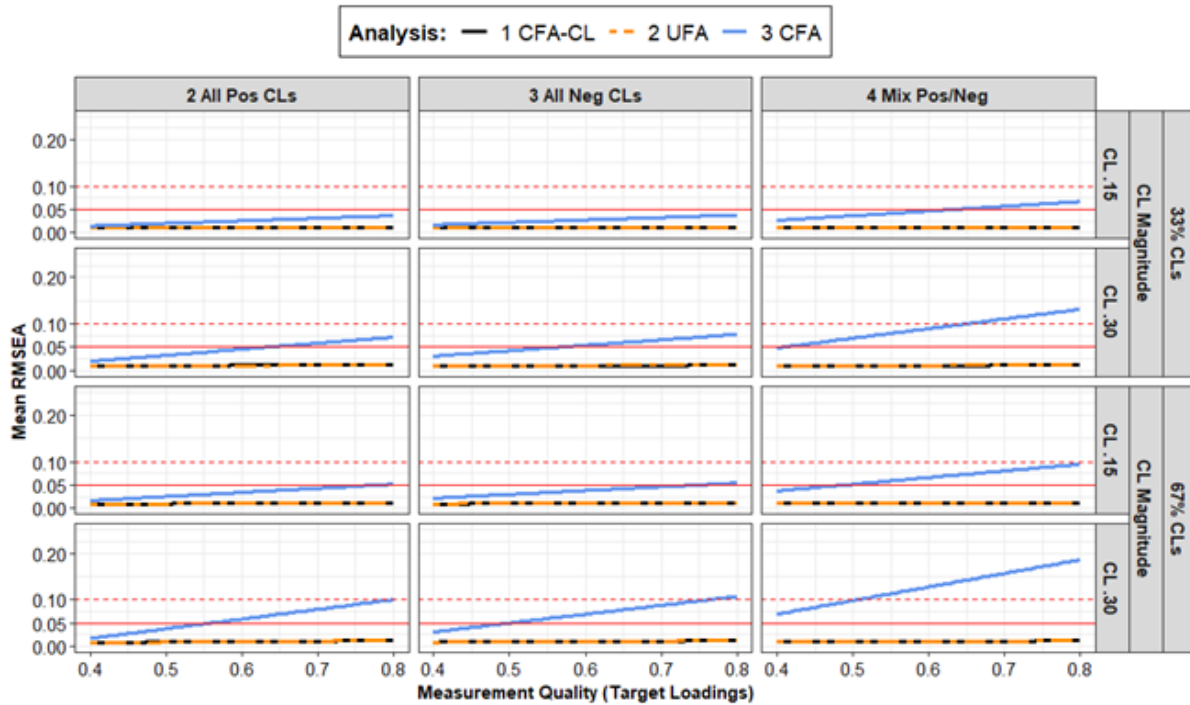
- GFI measures provide a convenient tool for gauging model fit, however, there seems to be many false negatives and false positives (in relation to FF correlation bias) when used to evaluate CFA models (relative to UFA models)
- It may have come for us to reconsider what constitutes a ‘trivial misspecification’ and to more fully embrace the idea that unidimensional indicators are “an inconvenient fiction” (Marsh et al., 2013, p. 258)
- We encourage users of CFA to not solely rely on stand-alone GFI measures, even when they are ‘good’ in relation to historical thresholds, but to also examine and consider modification indices that might point to misspecifications in CFA models that could lead to less biased parameter estimates among structural components
- We encourage users of UFA/ESEM to also include examination their estimated model parameters when evaluating their models, regardless of “strong” GFI values that may accompany their models.
 - In the current study, nearly all UFA models produced GFI values suggestive of good fit.
 - We are not particularly pleased with the amount of factor correlation bias present in the UFA models.
 - This has been shown elsewhere (e.g., Morin et al., 2013).
 - There remains room for improvement in UFA bias that might come from inspection model results that reveal (along with theory) that some aspects of the UFA measurement model could be constrained to zero

Appendix

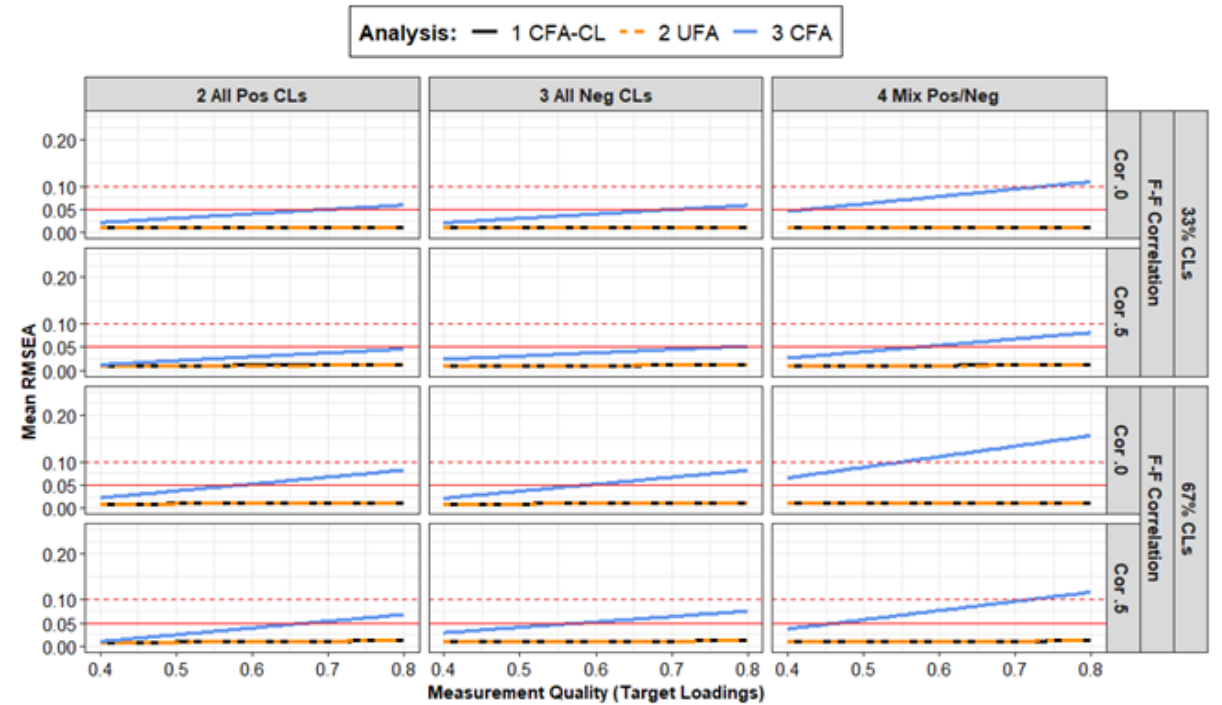


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CL Magnitude



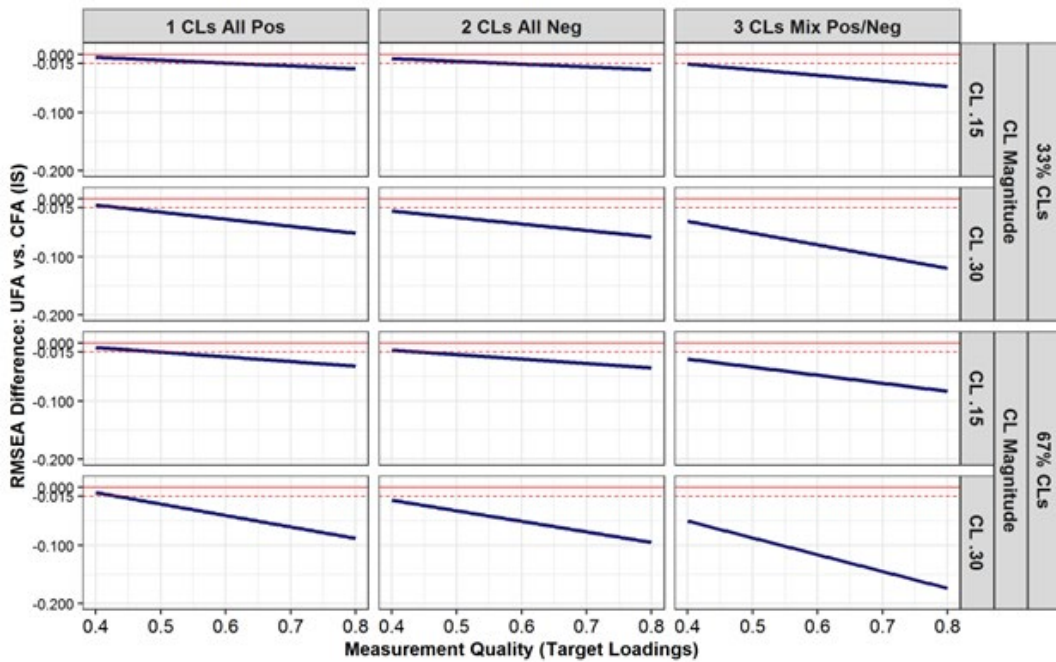
F-F Correlation



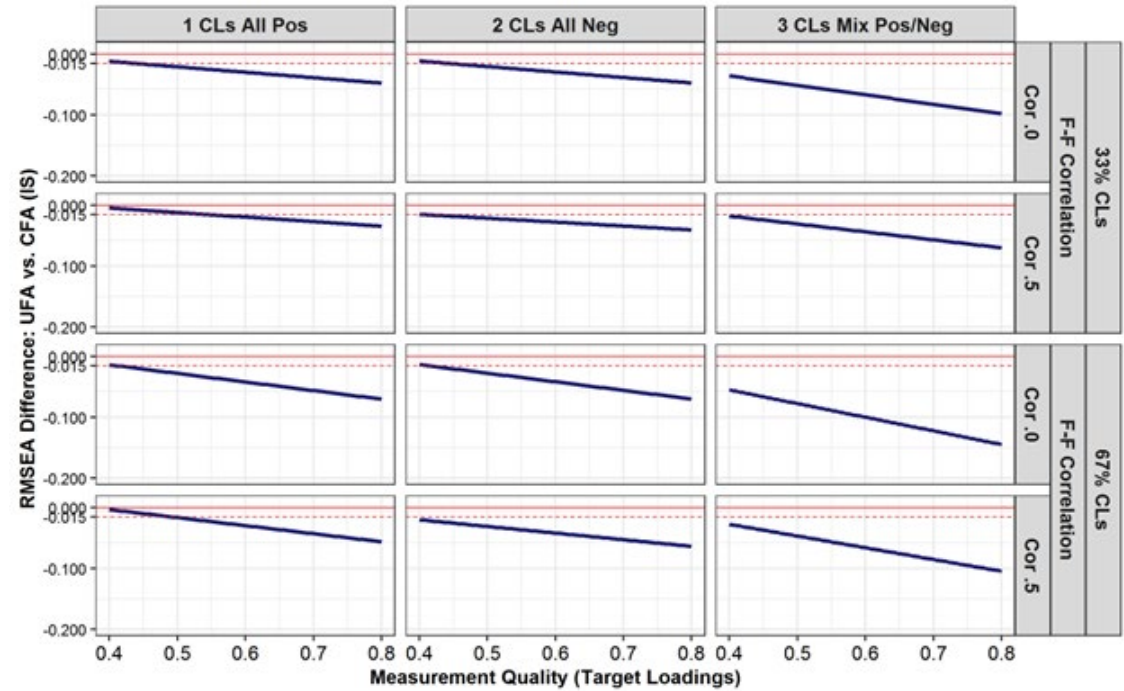
Results

Results for RMSEA Differences between UFA and CFA (IS)

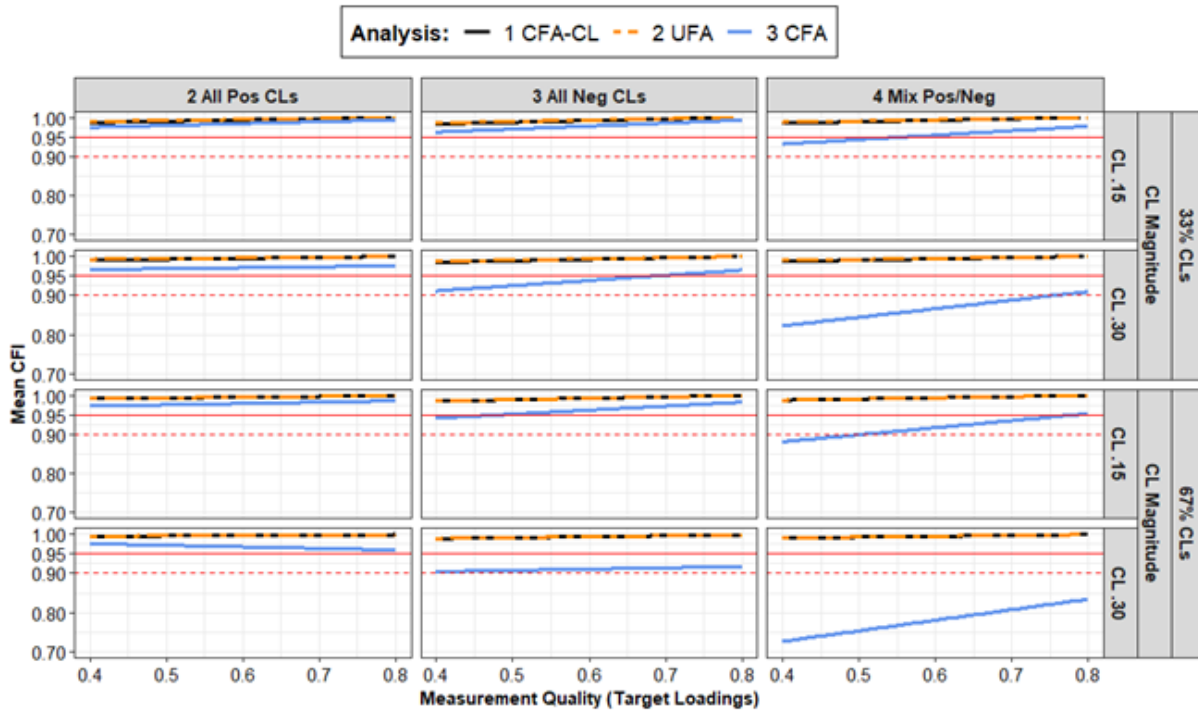
CL Magnitude



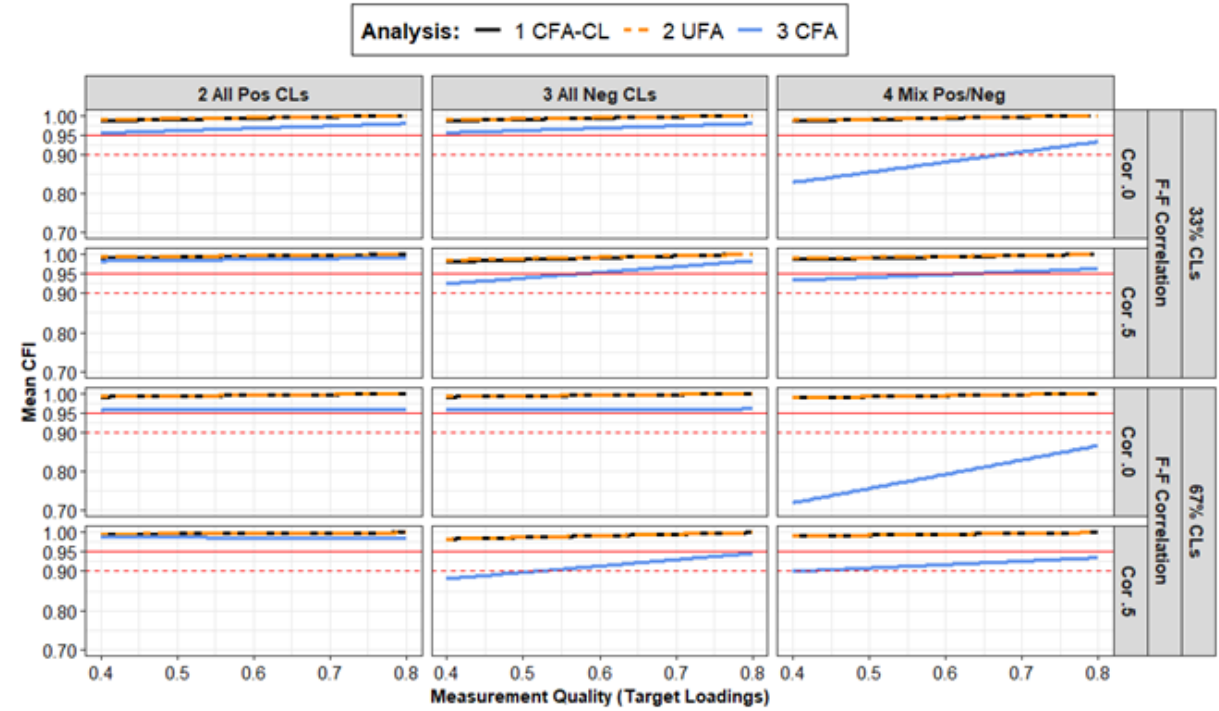
F-F Correlation



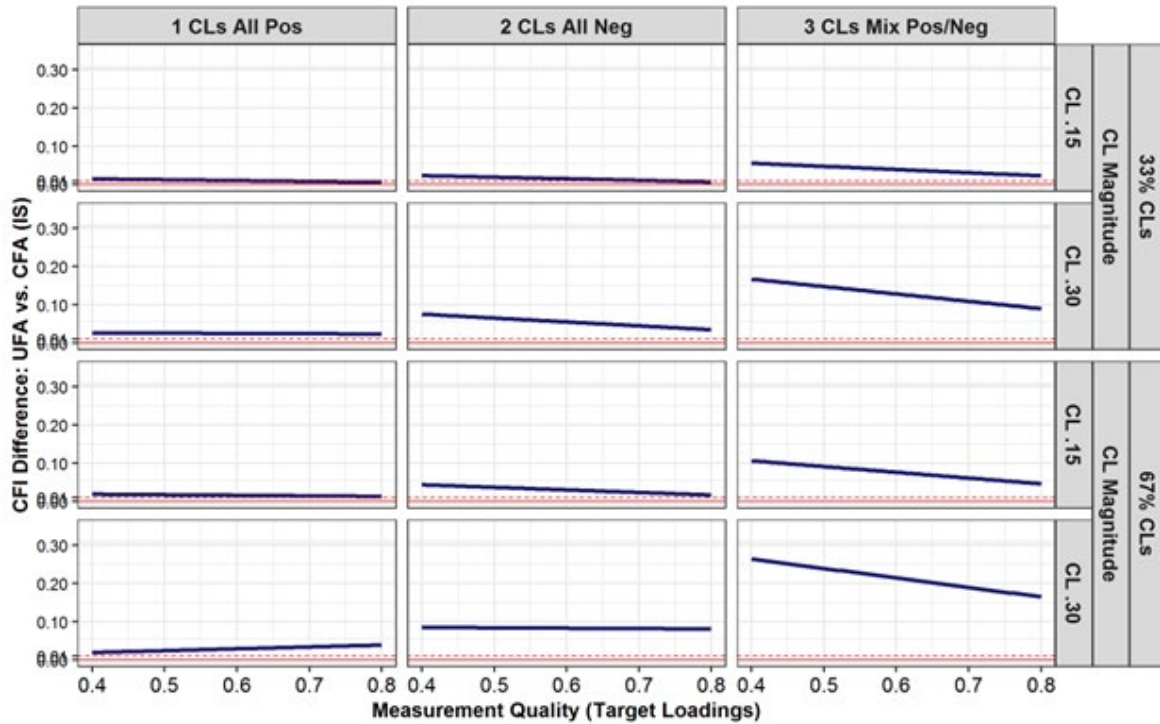
CL Magnitude



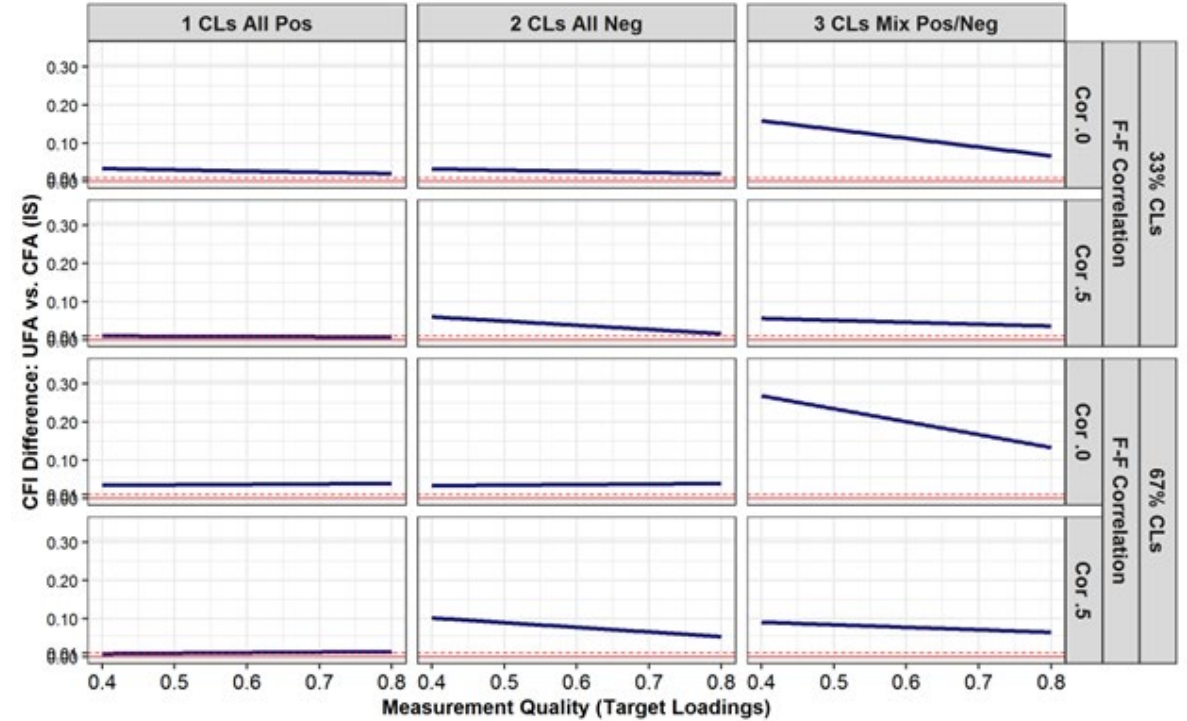
F-F Correlation



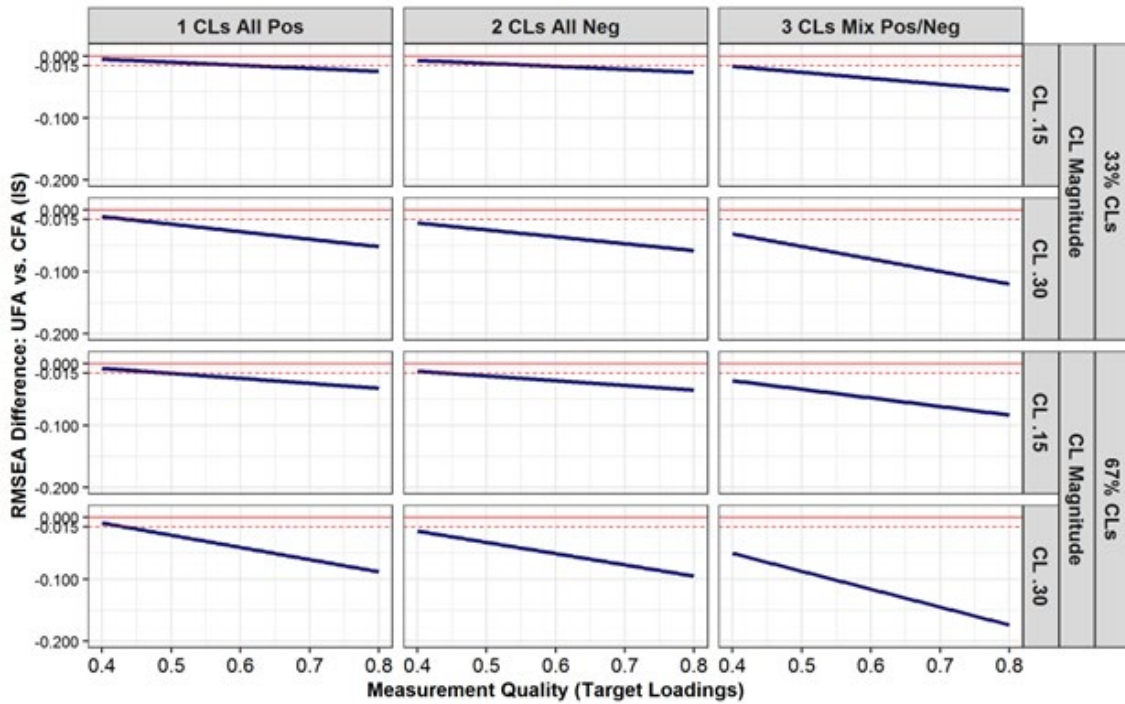
CL Magnitude



F-F Correlation



CL Magnitude



F-F Correlation

