University of British Columbia, Department of Psychology Looking Beyond Fit Indices: A Bifactor Model Best Practice Study

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Introduction

The bifactor model (see Figure 1) was developed 80 years ago¹, and received huge popularity recently. Popular applications in psychology include: general psychopathology model, general personality factor model, general mental ability model, etc.

Figure 1. Standard Bifactor Model



Statistical Pitfalls of Bifactor Model

- Tendency to overfit²
- Propensity to produce anomalous results including small loadings, unexpected negative loadings, and negative factor variances³
- Instability across samples⁴
- Under-identification problems⁵

Best Practice Recommendations

- Ensuring solution interpretability; considering alternative models (i.e., hierarchical model, bifactor S-1 model)⁶
- Reliability and dimensionality indices, including omegaH, omegaHS, ECV, PUC, H, and FD⁷

Method

With keyword "bifactor", from *PsycInfo*, we found 109 articles published in 2020, in which N = 97 fit one or more confirmatory bifactor model(s). From 81 articles with model structural details reported, we coded $N_m = 107$ bifactor models:

- Model statistics: fit indices, omega reliabilities, dimensionality measures.
- <u>Model structure</u>: factor loadings, scale length, # of group factors, model type (i.e., S-1 model).
- <u>Data disclosure</u>: correlation matrix, raw dataset, factor loadings.





General Factor Loadings



odels (22%), post-hoc bifactor S-1 model (73%)	F
actor Loading (out of 57 retained models)	g
Standardized factor loading (see Figure 2):	
general factor ($M = 0.547$, $SD = 0.200$), group	
factor ($M = 0.386$, $SD = 0.250$).	
<u>Negative loading</u> : general factor ($M = -0.088$),	
group factor ($M = -0.103$); at least one negative	
loading (anywhere: 46%, general factor: 9%,	
group factor: 42%).	
<u>Small loading</u> : at least one $\lambda_{gen} \leq 0.5 $ (88%), at	
least one $\lambda_{gen} \leq 0.3 $ (51%), at least one $\lambda_{grp} \leq$	
0.3 (81%).	
mega Reliabilities (out of 49 retained models)	
<u>Total scale</u> : ω_H (<i>M</i> = 0.768, <i>SD</i> = 0.142), ω_{HS} (<i>M</i>	
= 0.148, SD = 0.116); ω_H > 0.7 (71%), ω_H < 0.5	
(6%). Negative Residual Variances (13%).	
<u>Subscale</u> : $\omega_{H,j}$ (<i>M</i> = 0.507, <i>SD</i> = 0.239), $\omega_{HS,j}$ (<i>M</i>	
= 0.320, SD = 0.220), at least one $\omega_{HS,j} > \omega_{H,j}$	
(69%), at least one $\omega_{HS,j} > 0.5$ (45%).	
Replication: omega reliabilities replicated in 13	
articles (out of 22); average absolute deviation	
was 0.113 (<i>SD</i> = 0.099).	

Discussion

Figure 4. OmegaHS (% of variances explained by roup factors): Min, Max, Mean, Range Omega Reliability (Subscale)



References

41(5), 407-422.

Reliability/Dimensionality measures: not considered by 53% of the articles; reported values cannot be replicated in 41% of studies. • At least one anomalous result: retained bifactor models (100%), rejected models (100%). Rejected models: weaker subscales ($\omega_{HS,i}$: M =0.241), smaller negative loadings (general factor: M = -0.345, group factor: M = -0.201). Limitation: small sample size.

• Group factors are weaker and more problematic: less reliable, smaller factor loadings, more likely to have unexpected negative loadings and small loadings.

Group factors are expected to be less reliable than general factor. We suggest an alternative statistics "partial or conditional reliability":

$$c_{HS} = \frac{\sum_{j=1}^{k} (\sum_{t=1}^{m_j} \lambda_{g,t})^2}{\sum_{j=1}^{k} (\sum_{t=1}^{m_j} \lambda_{g,t})^2 + \sum_{i=1}^{p} U_i}$$

• (k: # of group factors, m_i : # of items in subscale j, $\lambda_{a,t}$: group factor loading, U_i : ith residual variance)

Holzinger, K.J., Swineford, F. (1937). The Bi-factor method. *Psychometrika*, 2, 41–54. 2. Murray, A. L., & Johnson, W. (2013). The limitations of model fit in comparing the bifactor versus higher-order models of human cognitive ability structure. Intelligence,

Eid, M., Geiser, C., Koch, T., & Heene, M. (2017). Anomalous results in G-factor models: Explanations and alternatives. Psychological Methods, 22(3), 541–562. Bornovalova, M. A., Choate, A. M., Fatimah, H., Petersen, K. J., & Wiernik, B. M.

(2020). Appropriate use of bifactor analysis in psychopathology research: Appreciating benefits and limitations. Biological Psychiatry, 88(1), 18-27.

Green, S., & Yang, Y. (2017). Empirical underidentification with the BIFACTOR model: A case study. Educational and Psychological Measurement, 78(5), 717-736.

Watts, A. L., Poore, H., & Waldman, I. (2019). Riskier tests of the validity of Bifactor models of Psychopathology. Clinical *Psychological Science*, 7(6), 1285–1303. Rodriguez, A., Reise, S. P., & Haviland, M. G. (2016). Evaluating bifactor models: Calculating and interpreting statistical indices. *Psychological Methods*, 21(2), 137–150.