## Summary of test statistics studied in "Corrected Goodness-of-fit Test in Covariance Structure Analysis"

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In a structural equation modeling, a model-implied covariance matrix  $\Sigma(\theta)$ ,  $(p \times p)$  is fitted to the sample covariance matrix  $S_N$  which is obtained from N observations  $(\mathbf{z}_1, ..., \mathbf{z}_N)$ ,  $(p \times 1)$  in order to minimize a discrepancy function. In practice, unknown parameter  $\theta$  is typically obtained by minimizing the Wishart likelihood  $F_{ML}(\theta)$ :

$$\begin{split} \widehat{\boldsymbol{\theta}}_{ML} &= \underset{\boldsymbol{\theta}}{\operatorname{argmin}} F_{ML}\left(\boldsymbol{\theta}\right), \\ F_{ML}\left(\boldsymbol{\theta}\right) &= \log\left|\boldsymbol{\Sigma}\left(\boldsymbol{\theta}\right)\right| - \log\left|\mathbf{S}_{N}\right| + \operatorname{tr}\left(\mathbf{S}_{N}\boldsymbol{\Sigma}\left(\boldsymbol{\theta}\right)^{-1}\right) - p. \end{split}$$

The conventional goodness of fit test or a likelihood ratio test is defined as

$$T_{ML} = n \cdot \hat{F}_{ML} \tag{1}$$

where  $\widehat{F}_{ML} = F_{ML}(\widehat{\boldsymbol{\theta}}_{ML})$  and n = N - 1.

The reweighted least squares (RLS) goodness-of-fit test due to Browne (1974) is defined as

$$T_{RLS} = \frac{n}{2} \operatorname{tr} \left\{ \left( \mathbf{S}_N - \widehat{\boldsymbol{\Sigma}} \right) \widehat{\boldsymbol{\Sigma}}^{-1} \right\}^2 = \frac{n}{2} \left\{ \mathbf{s}_N - \widehat{\boldsymbol{\sigma}} \right\}' \mathbf{D}_p' \left( \widehat{\boldsymbol{\Sigma}}^{-1} \otimes \widehat{\boldsymbol{\Sigma}}^{-1} \right) \mathbf{D}_p \left\{ \mathbf{s}_N - \widehat{\boldsymbol{\sigma}} \right\}$$
(2)

where  $\widehat{\mathbf{\Sigma}} = \mathbf{\Sigma}(\widehat{\boldsymbol{\theta}}_{ML})$ ,  $\widehat{\boldsymbol{\sigma}} = \operatorname{vech}(\widehat{\mathbf{\Sigma}})$ ,  $\mathbf{s}_N = \operatorname{vech}(\mathbf{S}_N)$ , and  $\mathbf{D}_p$  is a  $p^2 \times \frac{p(p+1)}{2}$  duplication matrix such that  $\operatorname{vec}(\mathbf{A}) = \mathbf{D}_p \operatorname{vech}(\mathbf{A})$ . Browne (1974) has demonstrated that two test statistics  $T_{ML}$  and  $T_{RLS}$  are related as follows:

$$T_{ML} = T_{RLS} + B \tag{3}$$

where  $B = n \sum_{k=3}^{\infty} \frac{1}{k} \operatorname{tr} \left\{ \mathbf{I}_p - \mathbf{S}_N \widehat{\boldsymbol{\Sigma}}^{-1} \right\}^k$ .

Under multivariate normality and correct specification for  $\Sigma(\theta)$ ,  $T_{ML}$  and  $T_{RLS}$  asymptotically follows the chi-square distribution as  $N \to \infty$  with p fixed (Browne, 1974, Proposition 7):

$$T_{ML}, T_{RLS} \xrightarrow{d} \chi_{df}^2$$

where  $df = p^* - q$  with  $p^* = p(p+1)/2$  and q is the number of free parameters in  $\theta$ .

Under the violation of multivariate normality,  $T_{ML}$  and  $T_{RLS}$  do not follow the standard chi-square distribution. Corrected tests that approximately follow the standard chi-square distribution are defined as

$$\dot{T}_{ML} = \frac{df}{\widehat{a}_1} T_{ML}, \qquad \ddot{T}_{ML} = \frac{\widehat{a}_1}{\widehat{a}_2} T_{ML}, \qquad \ddot{T}_{ML}^c = \frac{\widehat{a}_1}{\widehat{a}_2^c} T_{ML}, \tag{4}$$

$$\dot{T}_{RLS} = \frac{df}{\widehat{a}_1} T_{RLS}, \qquad \ddot{T}_{RLS} = \frac{\widehat{a}_1}{\widehat{a}_2} T_{RLS}, \qquad \ddot{T}_{RLS}^c = \frac{\widehat{a}_1}{\widehat{a}_2^c} T_{RLS},$$
(5)

where

$$\widehat{a}_{1} = \operatorname{tr}(\widehat{\mathbf{V}}), \qquad \widehat{a}_{2} = \operatorname{tr}(\widehat{\mathbf{V}}^{2}),$$

$$\widehat{a}_{2}^{c} = \frac{(N-2)(N-1)\operatorname{tr}(\mathbf{H}^{2}) - N(N-1)\operatorname{tr}(\mathbf{D}^{2}) + [\operatorname{tr}(\mathbf{H})]^{2}}{N(N-1)(N-2)(N-3)}$$

$$\widehat{\mathbf{V}} = \widehat{\mathbf{U}}^{1/2} \widehat{\mathbf{\Omega}} \widehat{\mathbf{U}}^{1/2}, \qquad \widehat{\mathbf{U}} = \widehat{\mathbf{W}} - \widehat{\mathbf{W}} \widehat{\mathbf{G}} \left( \widehat{\mathbf{G}}' \widehat{\mathbf{W}} \widehat{\mathbf{G}} \right)^{-1} \widehat{\mathbf{G}}' \widehat{\mathbf{W}}, \qquad \widehat{\mathbf{W}} = \frac{1}{2} \mathbf{D}'_p \left( \widehat{\mathbf{\Sigma}}^{-1} \otimes \widehat{\mathbf{\Sigma}}^{-1} \right) \mathbf{D}_p,$$

$$\widehat{\mathbf{G}} = \frac{\partial \boldsymbol{\sigma}(\widehat{\boldsymbol{\theta}}_{ML})}{\partial \boldsymbol{\theta}'}, \quad \widehat{\mathbf{\Omega}} = \frac{1}{N-1} \sum_{i=1}^{N} (\mathbf{s}_i - \bar{\mathbf{s}})(\mathbf{s}_i - \bar{\mathbf{s}})', \quad \mathbf{s}_i = \text{vech}\{(\mathbf{z}_i - \bar{\mathbf{z}})(\mathbf{z}_i - \bar{\mathbf{z}})'\}, \quad \bar{\mathbf{s}} = N^{-1} \sum_{i=1}^{N} \mathbf{s}_i,$$

$$\mathbf{H} = \sum_{i=1}^{N} \mathbf{y}_i \mathbf{y}'_i, \quad \mathbf{D} = \text{diag}\left(\mathbf{y}'_1 \mathbf{y}_1, ..., \mathbf{y}'_N \mathbf{y}_N\right), \quad \mathbf{y}_i = \widehat{\mathbf{w}}_i - \overline{\widehat{\mathbf{w}}}, \quad \overline{\widehat{\mathbf{w}}} = \frac{1}{N} \sum_{i=1}^{N} \widehat{\mathbf{w}}_i, \quad \widehat{\mathbf{w}}_i = \widehat{\mathbf{U}}^{1/2} \mathbf{s}_i.$$

The notations used in R code and in the paper are summarized in the following table.

Table 1: Notations in R code and paper

| R code      | Paper              |
|-------------|--------------------|
| $T_ML$      | $T_{ML}$           |
| T_RLS       | $T_{RLS}$          |
| T_SB1_ML    | $\dot{T}_{ML}$     |
| T_SB1_RLS   | $\dot{T}_{RLS}$    |
| T_SB2_ML    | $\ddot{T}_{ML}$    |
| T_SB2_RLS   | $\ddot{T}_{RLS}$   |
| T_SB2_ML_c  | $\ddot{T}^c_{ML}$  |
| T_SB2_RLS_c | $\ddot{T}^c_{RLS}$ |

## References

Browne, M. W. (1974) "Generalized Least Squares Estimators in the Analysis of Covariances," South African Statistical Journal, 8, p. 24.