
TESTING HOMOGENEITY OF HIGH-DIMENSIONAL COVARIANCE MATRICES

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Supplementary Material

This Supplementary Material contains the proofs of the theorems and simulation results. In particular, Tables 1–4 summarize the simulation results for two samples, and Table 5 provides the simulation results for two samples under the ultra high-dimensional setting. Figure 1 presents the simulation results for three samples with Gaussian populations, and Figure 2 presents the simulation results for three samples with Gamma populations. For ease of exposition, the proof of Theorem 2 is given first and that of Theorem 1 is given later. We first provide two lemmas which are essential to the proofs of Theorems 1 and 2.

S1 Lemmas

Lemma 1. *Under Assumptions (A1)–(A2), we have*

$$\sigma_{A12}^{-1}\{\text{tr}[(\mathbf{S}_1 - \mathbf{S}_2)^2] - \hat{\mu}_{21} - \mu_{A12}\} \rightarrow N(0, 1)$$

where $\hat{\mu}_{21} = \sum_{k=1}^2 (n_k^2 - n_k - 1)n_k^{-1}(n_k - 1)^{-2}(\text{tr}\mathbf{S}_k)^2$,

$$\mu_{A12} = \text{tr}[(\boldsymbol{\Sigma}_1 - \boldsymbol{\Sigma}_2)^2] + \sum_{k=1}^2 \frac{n_k + 1}{(n_k - 1)^2} \text{tr}(\boldsymbol{\Sigma}_k^2) + \sum_{k=1}^2 \frac{\beta_k n_k}{(n_k - 1)^2} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \boldsymbol{\Sigma}_k \mathbf{e}_\ell)^2$$

and

$$\begin{aligned} \sigma_{A12}^2 &= 4[n_1^{-1}\text{tr}(\boldsymbol{\Sigma}_1^2)]^2 + 4[n_2^{-1}\text{tr}(\boldsymbol{\Sigma}_2^2)]^2 + 8n_1^{-1}n_2^{-1}[\text{tr}(\boldsymbol{\Sigma}_1 \boldsymbol{\Sigma}_2)]^2 \\ &\quad + 4[2n_1^{-1}\text{tr}(\boldsymbol{\Sigma}_1^4) + \beta_1 n_1^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^2 \mathbf{e}_\ell)^2] \\ &\quad + 4\{2n_1^{-1}\text{tr}[(\boldsymbol{\Sigma}_1 \boldsymbol{\Sigma}_2)^2] + \beta_1 n_1^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^{1/2} \boldsymbol{\Sigma}_2 \boldsymbol{\Sigma}_1^{1/2} \mathbf{e}_\ell)^2\} \\ &\quad - 8[2n_1^{-1}\text{tr}(\boldsymbol{\Sigma}_1^3 \boldsymbol{\Sigma}_2) + \beta_1 n_1^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^{1/2} \boldsymbol{\Sigma}_2 \boldsymbol{\Sigma}_1^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^2 \mathbf{e}_\ell] \\ &\quad + 4n_2^{-1}[2\text{tr}(\boldsymbol{\Sigma}_2^4) + \beta_2 \sum_{\ell=1}^p (\mathbf{e}_\ell^T \boldsymbol{\Sigma}_2^2 \mathbf{e}_\ell)^2] \\ &\quad + 4\{2n_2^{-1}\text{tr}[(\boldsymbol{\Sigma}_2 \boldsymbol{\Sigma}_1)^2] + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \boldsymbol{\Sigma}_2^{1/2} \boldsymbol{\Sigma}_1 \boldsymbol{\Sigma}_2^{1/2} \mathbf{e}_\ell)^2\} \\ &\quad - 8[2n_2^{-1}\text{tr}(\boldsymbol{\Sigma}_2^3 \boldsymbol{\Sigma}_1) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \boldsymbol{\Sigma}_2^{1/2} \boldsymbol{\Sigma}_1 \boldsymbol{\Sigma}_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \boldsymbol{\Sigma}_2^2 \mathbf{e}_\ell]. \end{aligned}$$

Proof. Define $\mathbf{r}_{ki} = n_k^{-1/2} \mathbf{w}_{ki}$ and $\mathbf{w}_{ki} = (w_{k1i}, \dots, w_{kp_i})^T$ for $i = 1, \dots, n_k$ and $k = 1, 2$. We have

$$(n_k - 1)^2 n_k^{-2} \text{tr}(\mathbf{S}_k^2) = \text{tr}\left[\left(\sum_{i=1}^{n_k} \boldsymbol{\Sigma}_k^{1/2} \mathbf{r}_{ki} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k^{1/2}\right)^2\right] + n_k^2 (\bar{\mathbf{r}}_k^T \boldsymbol{\Sigma}_k \bar{\mathbf{r}}_k)^2 - 2n_k \bar{\mathbf{r}}_k^T \boldsymbol{\Sigma}_k \sum_{i=1}^{n_k} \mathbf{r}_{ki} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \bar{\mathbf{r}}_k,$$

where $\bar{\mathbf{r}}_k = n_k^{-1} \sum_{i=1}^{n_k} \mathbf{r}_{ki}$ for $k = 1, \dots, K$.

First, we will consider the bounded spectral norm case, that is, the maximum eigenvalue of $\boldsymbol{\Sigma}_k$ is bounded for $k = 1, 2$. When $\text{tr}(\boldsymbol{\Sigma}_k^q) = O(p^q)$

holds for $q = 1, 2, 3, 4$ and at least one integer k in the index set $\{1, 2\}$, we only need to consider the CLT of $n^{-2/3}\text{tr}[(\mathbf{S}_1 - \mathbf{S}_2)^2]$, and its proof is almost the same as the proof for the CLT of $\text{tr}[(\mathbf{S}_1 - \mathbf{S}_2)^2]$ when Σ_1 and Σ_2 have the bounded spectral norms.

Specifically, for the bounded spectral norm case, the proof can be completed through the following steps.

Step 1. We will prove $n_k \bar{\mathbf{r}}_k^T \Sigma_k \bar{\mathbf{r}}_k = n_k^{-1} \text{tr} \Sigma_k + o_p(1)$. We have $n_k \bar{\mathbf{r}}_k^T \Sigma_k \bar{\mathbf{r}}_k = 2n_k^{-1} \sum_{i < j} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{kj} + n_k^{-1} \sum_{i=1}^{n_k} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{ki}$. As $E(2n_k^{-1} \sum_{i < j} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{kj}) = 0$ and

$$E(2n_k^{-1} \sum_{i < j} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{kj})^2 \leq 4E(\mathbf{r}_{k1}^T \Sigma_k \mathbf{r}_{k2} \mathbf{r}_{k2}^T \Sigma_k \mathbf{r}_{k1}) = 4n_k^{-2} \text{tr}(\Sigma_k^2) \rightarrow 0,$$

we have $2n_k^{-1} \sum_{i < j} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{kj} = o_p(1)$. Moreover, as $E(n_k^{-1} \sum_{i=1}^{n_k} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{ki}) = n_k^{-1} \text{tr} \Sigma_k$ and $\text{Var}(n_k^{-1} \sum_{i=1}^{n_k} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{ki}) = n_k^{-1} E[(\mathbf{r}_{k1}^T \Sigma_k \mathbf{r}_{k1} - n_k^{-1} \text{tr} \Sigma_k)^2] = n_k^{-2} [2\text{tr}(\Sigma_k^2) + \beta_k \sum_{j=1}^p (\mathbf{e}_j^T \Sigma_k \mathbf{e}_j)^2] \rightarrow 0$ from (1.15) of Bai and Silverstein (2004), we have $n_k^{-1} \sum_{i=1}^{n_k} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{ki} - n_k^{-1} \text{tr} \Sigma_k = o_p(1)$, and thus

$$n_k \bar{\mathbf{r}}_k^T \Sigma_k \bar{\mathbf{r}}_k = n_k^{-1} \text{tr} \Sigma_k + o_p(1).$$

Step 2. We have

$$\begin{aligned} n_k \bar{\mathbf{r}}_k^T \Sigma_k \sum_{i=1}^{n_k} \mathbf{r}_{ki} \mathbf{r}_{ki}^T \Sigma_k \bar{\mathbf{r}}_k &= n_k^{-1} \sum_{i \neq j \neq \ell} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{kj} \mathbf{r}_{kj}^T \Sigma_k \mathbf{r}_{k\ell} + n_k^{-1} \sum_{i \neq j} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{kj} \mathbf{r}_{kj}^T \Sigma_k \mathbf{r}_{ki} \\ &\quad + 2n_k^{-1} \sum_{i \neq j} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{ki} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{kj} + n_k^{-1} \sum_{i=1}^{n_k} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{ki} \mathbf{r}_{ki}^T \Sigma_k \mathbf{r}_{ki}. \end{aligned}$$

Step 2.1. We have

$$\begin{aligned} n_k^{-1} \sum_{i=1}^{n_k} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} &= n_k^{-1} \sum_{i=1}^{n_k} (\mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} - n_k^{-1} \text{tr} \boldsymbol{\Sigma}_k)^2 + (n_k^{-1} \text{tr} \boldsymbol{\Sigma}_k)^2 \\ &\quad + 2n_k^{-1} \sum_{i=1}^{n_k} (n_k^{-1} \text{tr} \boldsymbol{\Sigma}_k) (\mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} - n_k^{-1} \text{tr} \boldsymbol{\Sigma}_k). \end{aligned}$$

As $n_k^{-1} \sum_{i=1}^{n_k} \mathbb{E}[(\mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} - n_k^{-1} \text{tr} \boldsymbol{\Sigma}_k)^2] = n_k^{-2} [2 \text{tr} \boldsymbol{\Sigma}_k^2 + \beta_k \sum_{j=1}^p (\mathbf{e}_j^T \boldsymbol{\Sigma}_k \mathbf{e}_j)^2] \rightarrow 0$, we have $n_k^{-1} \sum_{i=1}^{n_k} (\mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} - n_k^{-1} \text{tr} \boldsymbol{\Sigma}_k)^2 = o_p(1)$. As $n_k^{-1} \sum_{i=1}^{n_k} \mathbb{E}(\mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} - n_k^{-1} \text{tr} \boldsymbol{\Sigma}_k) = 0$ and

$$\text{Var}[n_k^{-1} \sum_{i=1}^{n_k} (\mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} - n_k^{-1} \text{tr} \boldsymbol{\Sigma}_k)] = n_k^{-3} [2 \text{tr} \boldsymbol{\Sigma}_k^2 + \beta_k \sum_{j=1}^p (\mathbf{e}_j^T \boldsymbol{\Sigma}_k \mathbf{e}_j)^2] \rightarrow 0,$$

we have $n_k^{-1} \sum_{i=1}^{n_k} (\mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} - n_k^{-1} \text{tr} \boldsymbol{\Sigma}_k) = o_p(1)$, and thus

$$n_k^{-1} \sum_{i=1}^{n_k} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} = (n_k^{-1} \text{tr} \boldsymbol{\Sigma}_k)^2 + o_p(1). \quad (\text{S1.1})$$

Step 2.2. We have $n^{-1} \sum_{i \neq j \neq \ell} \mathbb{E}(\mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{kj} \mathbf{r}_{kj}^T \boldsymbol{\Sigma}_k \mathbf{r}_{k\ell}) = 0$ and

$$\begin{aligned} &\mathbb{E}(n^{-1} \sum_{i \neq j \neq \ell} \mathbf{r}_i^T \boldsymbol{\Sigma} \mathbf{r}_j \mathbf{r}_j^T \boldsymbol{\Sigma} \mathbf{r}_\ell)^2 \\ &\leq 10n^{-2} \text{tr}(\boldsymbol{\Sigma}^4) + (2n^{-3} + 24n^{-4}) [2 \text{tr}(\boldsymbol{\Sigma}^4) + \beta_k \sum_{j=1}^p (\mathbf{e}_j^T \boldsymbol{\Sigma}^2 \mathbf{e}_j)^2 + (\text{tr} \boldsymbol{\Sigma}^2)^2] \\ &\quad + 24n^{-4} \beta_k \sum_{j=1}^p (\mathbf{e}_j^T \boldsymbol{\Sigma}^2 \mathbf{e}_j)^2 + 8n^{-4} \beta_k^2 \sum_{j=1}^p \sum_{\ell=1}^p (\mathbf{e}_j^T \boldsymbol{\Sigma} \mathbf{e}_\ell)^4 \rightarrow 0. \end{aligned}$$

This leads to $n_k^{-1} \sum_{i \neq j \neq \ell} (\mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{kj} \mathbf{r}_{kj}^T \boldsymbol{\Sigma}_k \mathbf{r}_{k\ell}) = o_p(1)$.

Step 2.3. We have $n_k^{-1} \sum_{i \neq j} \text{E} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{kj} \mathbf{r}_{kj}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} = (n_k - 1) n_k^{-2} \text{tr}(\boldsymbol{\Sigma}_k^2)$ and

$$\begin{aligned} & n_k^{-2} \text{E} \left(\sum_{i \neq j} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{kj} \mathbf{r}_{kj}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} \right)^2 \\ & \leq 6n_k^{-5}(n_k - 1)\beta_k \sum_{j=1}^p (\mathbf{e}_j^T \boldsymbol{\Sigma}_k^2 \mathbf{e}_j)^2 + 2n_k^{-5}(n_k - 1)\beta_k^2 \sum_{j=1}^p \sum_{\ell=1}^p (\mathbf{e}_j^T \boldsymbol{\Sigma}_k \mathbf{e}_\ell)^4 \\ & \quad + n_k^{-5}(n_k - 1)(n_k - 2)(n_k - 3)[\text{tr}(\boldsymbol{\Sigma}_k^2)]^2 \\ & \quad + 2n_k^{-5}(n_k - 1)(2n_k - 1)[2\text{tr}(\boldsymbol{\Sigma}_k^4) + \beta_k \sum_{j=1}^p (\mathbf{e}_j^T \boldsymbol{\Sigma}_k^2 \mathbf{e}_j)^2 + (\text{tr} \boldsymbol{\Sigma}_k^2)^2]. \end{aligned}$$

As a result, we have $\text{Var}(n_k^{-1} \sum_{i \neq j} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{kj} \mathbf{r}_{kj}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki}) \rightarrow 0$; that is,

$$n_k^{-1} \sum_{i \neq j} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{kj} \mathbf{r}_{kj}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} - (n_k - 1) n_k^{-2} \text{tr}(\boldsymbol{\Sigma}_k^2) = o_p(1).$$

Step 2.4. We have

$$\begin{aligned} & n_k^{-1} \sum_{i \neq j} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{kj} \\ & = n_k^{-1} \sum_{i \neq j} (\mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} - n_k^{-1} \text{tr} \boldsymbol{\Sigma}_k) \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{kj} + (n_k^{-2} \text{tr} \boldsymbol{\Sigma}_k) \sum_{i \neq j} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{kj}, \end{aligned}$$

and further,

$$\begin{aligned} & n_k^{-1} \left| \sum_{i \neq j} (\mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} - n_k^{-1} \text{tr} \boldsymbol{\Sigma}_k) \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{kj} \right| \\ & \leq 2n_k^{-1} \sum_i (\mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} - n_k^{-1} \text{tr} \boldsymbol{\Sigma}_k)^2 + 2n_k^{-1} \sum_{i \neq j} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{kj} \mathbf{r}_{kj}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki}. \end{aligned}$$

Then, we have $n_k^{-1} \sum_{i \neq j} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{kj} = o_p(1)$.

Thus, we have

$$\text{tr}(\mathbf{S}_k^2) = \frac{n_k^2}{(n_k - 1)^2} \text{tr} \left[\left(\sum_{i=1}^{n_k} \boldsymbol{\Sigma}_k^{1/2} \mathbf{r}_{ki} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k^{1/2} \right)^2 \right] - \frac{n_k + 1}{n_k(n_k - 1)^2} (\text{tr} \boldsymbol{\Sigma}_k)^2 - \frac{2}{n_k - 1} \text{tr}(\boldsymbol{\Sigma}_k^2) + o_p(1).$$

As $\text{tr}\mathbf{S}_k = n_k(n_k - 1)^{-1}(\sum_{i=1}^{n_k} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} - n_k \bar{\mathbf{r}}_k^T \boldsymbol{\Sigma}_k \bar{\mathbf{r}}_k)$, we have

$$\text{tr}\mathbf{S}_k = n_k(n_k - 1)^{-1} \sum_{i=1}^{n_k} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k \mathbf{r}_{ki} - (n_k - 1)^{-1} \text{tr}\boldsymbol{\Sigma}_k + o_p(1).$$

As shown in Bai and Silverstein (2004, pp. 559–560),

$$\text{tr}\left[\left(\sum_{i=1}^{n_k} \boldsymbol{\Sigma}_k^{1/2} \mathbf{r}_{ki} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k^{1/2}\right)^q\right] - \text{tr}\left[\left(\sum_{i=1}^{n_k} \boldsymbol{\Sigma}_k^{1/2} \tilde{\mathbf{r}}_{ki} \tilde{\mathbf{r}}_{ki}^T \boldsymbol{\Sigma}_k^{1/2}\right)^q\right] = o_p(1), \quad q = 1, 2$$

where $\tilde{\mathbf{r}}_{ki} = n_k^{-1/2} \tilde{\mathbf{w}}_{ki}$, $\tilde{\mathbf{w}}_{ki} = (\tilde{w}_{k1i}, \dots, \tilde{w}_{kp_i})^T$,

$$\tilde{w}_{k\ell i} = [\text{Var}(w_{k\ell i} \delta_{\{|w_{k\ell i}| \leq \sqrt{n_k} \eta_{n_k}\}})]^{-1/2} [w_{k\ell i} \delta_{\{|w_{k\ell i}| \leq \sqrt{n_k} \eta_{n_k}\}} - \text{E}(w_{k\ell i} \delta_{\{|w_{k\ell i}| \leq \sqrt{n_k} \eta_{n_k}\}})],$$

$$|\tilde{w}_{k\ell i}| \leq c\sqrt{n_k} \eta_{n_k}, \quad \text{E}\tilde{w}_{k\ell i} = 0, \quad \text{E}(\tilde{w}_{k\ell i}^2) = 1 \text{ and } \text{E}(\tilde{w}_{k\ell i}^4) < \infty \text{ for } \ell = 1, \dots, p$$

and $i = 1, \dots, n_k$ with $\eta_n \downarrow 0$, $\sqrt{n}\eta_n \rightarrow \infty$ and c being a positive constant.

For simplicity, we rename the variable $\tilde{w}_{k\ell i}$ simply as $w_{k\ell i}$ and proceed by

assuming that $|w_{k\ell i}| \leq \sqrt{n_k} \eta_{n_k}$, $\text{E}w_{k\ell i} = 0$, $\text{E}(w_{k\ell i}^2) = 1$ and $\text{E}(w_{k\ell i}^4) < \infty$

with $\eta_{n_k} \downarrow 0$ and $\sqrt{n_k} \eta_{n_k} \rightarrow \infty$. Let $\mathbf{B}_k = \sum_{i=1}^{n_k} \boldsymbol{\Sigma}_k^{1/2} \mathbf{r}_{ki} \mathbf{r}_{ki}^T \boldsymbol{\Sigma}_k^{1/2}$, then

$$\text{tr}\mathbf{S}_k = n_k(n_k - 1)^{-1} \text{tr}\mathbf{B}_k - (n_k - 1)^{-1} \text{tr}\boldsymbol{\Sigma}_k + o_p(1) \text{ and}$$

$$\begin{aligned} \text{tr}(\mathbf{S}_k^2) &= \frac{n_k^2}{(n_k - 1)^2} \text{tr}(\mathbf{B}_k^2) - \frac{n_k + 1}{n_k(n_k - 1)^2} (\text{tr}\boldsymbol{\Sigma}_k)^2 - \frac{2}{n_k - 1} \text{tr}(\boldsymbol{\Sigma}_k^2) + o_p(1). \end{aligned} \tag{S1.2}$$

Step 3. In this step, we show that

$$\begin{aligned} &\text{tr}[(\mathbf{S}_1 - \mathbf{S}_2)^2] - \text{Etr}[(\mathbf{S}_1 - \mathbf{S}_2)^2] \\ &= n_1^2(n_1 - 1)^{-2} [\text{tr}(\mathbf{B}_1^2) - \text{Etr}(\mathbf{B}_1^2)] + n_2^2(n_2 - 1)^{-2} [\text{tr}(\mathbf{B}_2^2) - \text{Etr}(\mathbf{B}_2^2)] \\ &\quad - 2n_1n_2(n_1 - 1)^{-1}(n_2 - 1)^{-1} [\text{tr}(\mathbf{B}_1 \mathbf{B}_2) - \text{Etr}(\boldsymbol{\Sigma}_1 \boldsymbol{\Sigma}_2)] + o_p(1) \end{aligned}$$

is asymptotically normal. When $\Sigma_1 = \Sigma_2 = \Sigma$, we have

$$\begin{aligned} & \text{Etr}(\mathbf{S}_1^2) + \text{Etr}(\Sigma_2^2) - 2\text{Etr}(\mathbf{S}_1 \mathbf{S}_2) \\ &= \sum_{k=1}^2 \frac{n_k + 1}{(n_k - 1)^2} \text{tr}(\Sigma^2) + \sum_{k=1}^2 \frac{n_k^2 - n_k - 1}{n_k(n_k - 1)^2} (\text{tr}\Sigma)^2 + \sum_{k=1}^2 \frac{\beta_k n_k}{(n_k - 1)^2} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_k \mathbf{e}_\ell)^2, \end{aligned}$$

where

$$\begin{aligned} & p^{-1} \text{Etr}(\mathbf{S}_k^2) \\ &= \frac{n_k(n_k - 1) + 2}{p(n_k - 1)^2} \text{tr}(\Sigma_k^2) + \frac{n_k^2 - n_k - 1}{pn_k(n_k - 1)^2} (\text{tr}\Sigma_k)^2 + \frac{\beta_w n_k}{p(n_k - 1)^2} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_k \mathbf{e}_\ell)^2 \\ &= p^{-1} \text{tr}(\Sigma_k^2) + \frac{n_k + 1}{p(n_k - 1)^2} \text{tr}(\Sigma_k^2) + \frac{n_k^2 - n_k - 1}{pn_k(n_k - 1)^2} (\text{tr}\Sigma_k)^2 + \frac{\beta_w n_k}{p(n_k - 1)^2} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_k \mathbf{e}_\ell)^2. \end{aligned}$$

Since $\sum_{k=1}^2 (n_k^2 - n_k - 1)n_k^{-1}(n_k - 1)^{-2}(\text{tr}\Sigma)^2 \rightarrow \infty$, we need to establish

the CLT of

$$\text{tr}(\mathbf{S}_1 - \mathbf{S}_2)^2 - \hat{\mu}_{21},$$

where $\hat{\mu}_{21} = \sum_{k=1}^2 (n_k^2 - n_k - 1)n_k^{-1}(n_k - 1)^{-2}(\text{tr}\mathbf{S}_k)^2$. Let E_ℓ be the conditional expectation given $\{\mathbf{x}_{11}, \dots, \mathbf{x}_{1\ell}\}$ and E_{ℓ, \mathbf{S}_2} be the conditional expectation given $\{\mathbf{x}_{11}, \dots, \mathbf{x}_{1\ell}, \mathbf{S}_2\}$. Based on the martingale difference central limit theory, it can be derived that conditional on \mathbf{S}_2 ,

$$\begin{aligned} & \sigma_{11A}^{-1/2} [\text{tr}(\mathbf{S}_1^2) - \text{Etr}(\mathbf{S}_1^2) - 2\text{tr}(\mathbf{S}_1 \mathbf{S}_2) + 2\text{tr}(\Sigma_1 \mathbf{S}_2) \\ & \quad - \frac{n_1^2 - n_1 - 1}{n_1(n_1 - 1)^2} \{(\text{tr}\mathbf{S}_1)^2 - (\text{tr}\Sigma_1)^2\}] \xrightarrow{d} N(0, 1), \end{aligned} \quad (\text{S1.3})$$

where $\sigma_{11A} = \sigma_{110A} + 4\sigma_{220A} - 4\sigma_{120A} + 4n_1^{-2}(\text{tr}\Sigma_1)^2\sigma_{330A} - 4n_1^{-1}(\text{tr}\Sigma_1)\sigma_{130A} +$

$8n_1^{-1}(\text{tr}\Sigma_1)\sigma_{230A}$, with

$$\begin{aligned}\sigma_{110A} &= \sum_{\ell=1}^{n_1}[(E_\ell - E_{\ell-1})\text{tr}(\mathbf{B}_1^2)]^2, \\ \sigma_{220A} &= \sum_{\ell=1}^{n_1}[(E_{\ell,\mathbf{B}_2} - E_{\ell-1,\mathbf{B}_2})\text{tr}(\mathbf{B}_1\mathbf{B}_2)]^2, \\ \sigma_{330A} &= \sum_{\ell=1}^{n_1}[(E_{\ell,\mathbf{B}_2} - E_{\ell-1,\mathbf{B}_2})\text{tr}(\mathbf{B}_1)]^2, \\ \sigma_{120A} &= \sum_{\ell=1}^{n_1}[(E_\ell - E_{\ell-1})\text{tr}(\mathbf{B}_1^2)][(E_{\ell,\mathbf{B}_2} - E_{\ell-1,\mathbf{B}_2})\text{tr}(\mathbf{B}_1\mathbf{B}_2)], \\ \sigma_{130A} &= \sum_{\ell=1}^{n_1}[(E_\ell - E_{\ell-1})\text{tr}(\mathbf{B}_1^2)][(E_{\ell,\mathbf{B}_2} - E_{\ell-1,\mathbf{B}_2})\text{tr}\mathbf{B}_1], \\ \sigma_{230A} &= \sum_{\ell=1}^{n_1}[(E_{\ell,\mathbf{B}_2} - E_{\ell-1,\mathbf{B}_2})(\text{tr}\mathbf{B}_1)][(E_{\ell,\mathbf{B}_2} - E_{\ell-1,\mathbf{B}_2})\text{tr}(\mathbf{B}_1\mathbf{B}_2)].\end{aligned}$$

Moreover, we have

$$\begin{aligned}\sigma_{22A}^{-1/2} &[\text{tr}(\mathbf{S}_2^2) - \text{Etr}(\mathbf{S}_2^2) - 2[\text{tr}(\Sigma_1\mathbf{S}_2) - \text{tr}(\Sigma_1\Sigma_2)] \\ &- \frac{n_2^2 - n_2 - 1}{n_2(n_2 - 1)^2} \{ (\text{tr}\mathbf{S}_2)^2 - (\text{tr}\Sigma_2)^2 \}] \xrightarrow{d} N(0, 1),\end{aligned}\quad (\text{S1.4})$$

where

$$\sigma_{22A} = \sigma_{440A} + 4\sigma_{550A} + 4n_2^{-2}(\text{tr}\Sigma_2)^2\sigma_{660A} - 4\sigma_{450A} - 4n_2^{-1}(\text{tr}\Sigma_2)\sigma_{460A} + 8n_2^{-1}(\text{tr}\Sigma_2)\sigma_{560A},$$

with

$$\begin{aligned}\sigma_{440A} &= \sum_{\ell=1}^{n_2}[(E_\ell - E_{\ell-1})\text{tr}(\mathbf{B}_2^2)]^2, \\ \sigma_{550A} &= \sum_{\ell=1}^{n_2}[(E_\ell - E_{\ell-1})\text{tr}(\mathbf{B}_2\Sigma_1)]^2, \\ \sigma_{660A} &= \sum_{\ell=1}^{n_2}[(E_\ell - E_{\ell-1})\text{tr}\mathbf{B}_2]^2,\end{aligned}$$

$$\begin{aligned}
 \sigma_{450A} &= \sum_{\ell=1}^{n_2} [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2^2)] [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2 \boldsymbol{\Sigma}_1)], \\
 \sigma_{460A} &= \sum_{\ell=1}^{n_2} [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2^2)] [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2)], \\
 \sigma_{560A} &= \sum_{\ell=1}^{n_2} [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2 \boldsymbol{\Sigma}_1)] [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2)].
 \end{aligned}$$

Step 4. Next, we show that

$$\begin{aligned}
 \mu_{A2} &= \text{Etr}(\mathbf{S}_2^2) + \text{Etr}(\mathbf{S}_1^2) - 2\text{tr}(\boldsymbol{\Sigma}_1 \boldsymbol{\Sigma}_2) - \sum_{k=1}^2 \frac{n_k^2 - n_k - 1}{n_k(n_k - 1)^2} (\text{tr} \boldsymbol{\Sigma}_k)^2 \\
 &= \text{tr}[(\boldsymbol{\Sigma}_1 - \boldsymbol{\Sigma}_2)^2] + \frac{n_1 + 1}{(n_1 - 1)^2} \text{tr}(\boldsymbol{\Sigma}_1^2) + \frac{n_2 + 1}{(n_2 - 1)^2} \text{tr}(\boldsymbol{\Sigma}_2^2) \\
 &\quad + \frac{\beta_1 n_1}{(n_1 - 1)^2} \sum_{k=1}^p (\mathbf{e}_k^T \boldsymbol{\Sigma}_1 \mathbf{e}_k)^2 + \frac{\beta_2 n_2}{(n_2 - 1)^2} \sum_{k=1}^p (\mathbf{e}_k^T \boldsymbol{\Sigma}_2 \mathbf{e}_k)^2.
 \end{aligned}$$

Moreover, we have

$$\begin{aligned}
 \sigma_{110A} &= \sum_{\ell=1}^{n_1} [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_1^2)]^2 \\
 &= 4n_1^{-1} [2\text{tr}(\boldsymbol{\Sigma}_1^4) + \beta_1 \sum_{\ell=1}^p (\mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^2 \mathbf{e}_\ell)^2] \\
 &\quad + 2(n_1^{-1} \text{tr} \boldsymbol{\Sigma}_1)^2 n_1^{-1} [2\text{tr}(\boldsymbol{\Sigma}_1^2) + \beta_1 \sum_{\ell=1}^p (\mathbf{e}_\ell^T \boldsymbol{\Sigma}_1 \mathbf{e}_\ell)^2] \\
 &\quad + 4[n_1^{-1} \text{tr}(\boldsymbol{\Sigma}_1^2)]^2 + 8(n_1^{-1} \text{tr} \boldsymbol{\Sigma}_1) n_1^{-1} [2\text{tr}(\boldsymbol{\Sigma}_1^3) + \beta_1 \sum_{\ell=1}^p \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^2 \mathbf{e}_\ell \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1 \mathbf{e}_\ell] \\
 &\quad + 2(n_1^{-1} \text{tr} \boldsymbol{\Sigma}_1)^2 n_1^{-1} [2\text{tr}(\boldsymbol{\Sigma}_1^2) + \beta_1 \sum_{\ell=1}^p (\mathbf{e}_\ell^T \boldsymbol{\Sigma}_1 \mathbf{e}_\ell)^2],
 \end{aligned}$$

$$\begin{aligned}
\sigma_{220A} &= \sum_{\ell=1}^{n_1} [(\mathbf{E}_{\ell,\mathbf{B}_2} - \mathbf{E}_{\ell-1,\mathbf{B}_2}) \text{tr}(\mathbf{B}_1 \mathbf{B}_2)]^2 \\
&= 2n_1^{-1} \text{tr}[(\boldsymbol{\Sigma}_1 \mathbf{B}_2)^2] + \beta_1 n_1^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^{1/2} \mathbf{B}_2 \boldsymbol{\Sigma}_1^{1/2} \mathbf{e}_\ell)^2 \\
&= \frac{2}{n_1 n_2} [\text{tr}(\boldsymbol{\Sigma}_1 \boldsymbol{\Sigma}_2)]^2 + [2n_1^{-1} \text{tr}[(\boldsymbol{\Sigma}_1 \boldsymbol{\Sigma}_2)^2] + \beta_1 n_1^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^{1/2} \boldsymbol{\Sigma}_2 \boldsymbol{\Sigma}_1^{1/2} \mathbf{e}_\ell)^2] + o_p(1), \\
\sigma_{330A} &= \sum_{\ell=1}^{n_1} [(\mathbf{E}_{\ell,\mathbf{B}_2} - \mathbf{E}_{\ell-1,\mathbf{B}_2}) \text{tr} \mathbf{B}_1]^2 \\
&= \sum_{\ell=1}^{n_1} \mathbf{E}(\mathbf{r}_\ell^T \boldsymbol{\Sigma}_1 \mathbf{r}_\ell)^2 = 2n_1^{-1} \text{tr}(\boldsymbol{\Sigma}_1^2) + \beta_1 n_1^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \boldsymbol{\Sigma}_1 \mathbf{e}_\ell)^2, \\
\sigma_{120A} &= \sum_{\ell=1}^{n_1} [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1}) \text{tr}(\mathbf{B}_1^2)] [(\mathbf{E}_{\ell,\mathbf{B}_2} - \mathbf{E}_{\ell-1,\mathbf{B}_2}) \text{tr}(\mathbf{B}_1 \mathbf{B}_2)] \\
&= 2n_1^{-1} [2 \text{tr}(\boldsymbol{\Sigma}_1^3 \mathbf{B}_2) + \beta_1 \sum_{\ell=1}^p \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^{1/2} \mathbf{B}_2 \boldsymbol{\Sigma}_1^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^2 \mathbf{e}_\ell] \\
&\quad + 2(n_1^{-1} \text{tr} \boldsymbol{\Sigma}_1) n_1^{-1} \left[2 \text{tr}(\boldsymbol{\Sigma}_1^2 \mathbf{B}_2) + \beta_1 \sum_{\ell=1}^p \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1 \mathbf{e}_\ell \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^{1/2} \mathbf{B}_2 \boldsymbol{\Sigma}_1^{1/2} \mathbf{e}_\ell \right] \\
&= 2[2n_1^{-1} \text{tr}(\boldsymbol{\Sigma}_1^3 \boldsymbol{\Sigma}_2) + \beta_1 n_1^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^{1/2} \boldsymbol{\Sigma}_2 \boldsymbol{\Sigma}_1^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^2 \mathbf{e}_\ell] \\
&\quad + 2(n_1^{-1} \text{tr} \boldsymbol{\Sigma}_1) n_1^{-1} \left[2 \text{tr}(\boldsymbol{\Sigma}_1^2 \boldsymbol{\Sigma}_2) + \beta_1 \sum_{\ell=1}^p \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1 \mathbf{e}_\ell \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^{1/2} \boldsymbol{\Sigma}_2 \boldsymbol{\Sigma}_1^{1/2} \mathbf{e}_\ell \right] + o_p(1), \\
\sigma_{130A} &= \sum_{\ell=1}^{n_1} [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1}) \text{tr}(\mathbf{B}_1^2)] [(\mathbf{E}_{\ell,\mathbf{B}_2} - \mathbf{E}_{\ell-1,\mathbf{B}_2}) \text{tr}(\text{tr} \mathbf{B}_1)] \\
&= \sum_{\ell=1}^{n_1} [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1}) \text{tr}(\mathbf{B}_1^2)] [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1}) \text{tr} \mathbf{B}_1] \\
&= 2(2n_1^{-1} \text{tr}(\boldsymbol{\Sigma}_1^3) + \beta_1 n_1^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1 \mathbf{e}_\ell \mathbf{e}_\ell^T \boldsymbol{\Sigma}_1^2 \mathbf{e}_\ell) \\
&\quad + 2(n_1^{-1} \text{tr} \boldsymbol{\Sigma}_1) [2n_1^{-1} \text{tr}(\boldsymbol{\Sigma}_1^2) + \beta_1 n_1^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \boldsymbol{\Sigma}_1 \mathbf{e}_\ell)^2],
\end{aligned}$$

$$\begin{aligned}
 \sigma_{230A} &= \sum_{\ell=1}^{n_1} [(\mathbf{E}_{\ell, \mathbf{B}_2} - \mathbf{E}_{\ell-1, \mathbf{B}_2})(\text{tr} \mathbf{B}_1)][(\mathbf{E}_{\ell, \mathbf{B}_2} - \mathbf{E}_{\ell-1, \mathbf{B}_2})\text{tr}(\mathbf{B}_1 \mathbf{B}_2)] \\
 &= \sum_{\ell=1}^{n_1} [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1})\text{tr} \mathbf{B}_1][(\mathbf{E}_{\ell, \mathbf{B}_2} - \mathbf{E}_{\ell-1, \mathbf{B}_2})\text{tr}(\mathbf{B}_1 \mathbf{B}_2)] \\
 &= \sum_{\ell=1}^{n_1} \mathbf{E}_{\ell-1, \mathbf{B}_2} (\mathbf{r}_\ell^T \Sigma_1 \mathbf{r}_\ell - n_1^{-1} \text{tr} \Sigma_1) [\mathbf{r}_\ell^T \Sigma_1^{1/2} \mathbf{B}_2 \Sigma_1^{1/2} \mathbf{r}_\ell - n_1^{-1} \text{tr}(\Sigma_1 \mathbf{B}_2)] \\
 &= [2n_1^{-1} \text{tr}(\Sigma_1^2 \Sigma_2) + \beta_1 n_1^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_1 \mathbf{e}_\ell)(\mathbf{e}_\ell^T \Sigma_1^{1/2} \Sigma_2 \Sigma_1^{1/2} \mathbf{e}_\ell)], \\
 \sigma_{440A} &= \sum_{\ell=1}^{n_2} [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1})\text{tr}(\mathbf{B}_2^2)]^2 \\
 &= 4n_2^{-1} [2\text{tr}(\Sigma_2^4) + \beta_2 \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell)^2] + 2(n_2^{-1} \text{tr} \Sigma_2)^2 n_2^{-1} [2\text{tr}(\Sigma_2^2) + \beta_2 \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell)^2] \\
 &\quad + 4[n_2^{-1} \text{tr}(\Sigma_2^2)]^2 + 8(n_2^{-1} \text{tr} \Sigma_2) n_2^{-1} [2\text{tr}(\Sigma_2^3) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell] \\
 &\quad + 2(n_2^{-1} \text{tr} \Sigma_2)^2 n_2^{-1} [2\text{tr}(\Sigma_2^2) + \beta_2 \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell)^2], \\
 \sigma_{550A} &= \sum_{\ell=1}^{n_2} [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1})\text{tr}(\mathbf{B}_2 \Sigma_1)]^2 \\
 &= [2n_2^{-1} \text{tr}[(\Sigma_2 \Sigma_1)^2] + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell)^2], \\
 \sigma_{660A} &= \sum_{\ell=1}^{n_2} [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1})\text{tr} \mathbf{B}_2]^2 = [2n_2^{-1} \text{tr}(\Sigma_2^2) + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell)^2], \\
 \sigma_{450A} &= \sum_{\ell=1}^{n_2} [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1})\text{tr}(\mathbf{B}_2^2)][(\mathbf{E}_\ell - \mathbf{E}_{\ell-1})\text{tr} \mathbf{B}_2 \Sigma_1] \\
 &= 2n_2^{-1} [2\text{tr}(\Sigma_2^3 \Sigma_1) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell] \\
 &\quad + 2(n_2^{-1} \text{tr} \Sigma_2) n_2^{-1} \left[2\text{tr}(\Sigma_2^2 \Sigma_1) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell \right],
 \end{aligned}$$

$$\begin{aligned}
\sigma_{460A} &= \sum_{\ell=1}^{n_2} [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2^2)] [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2)] \\
&= 2[2n_2^{-1} \text{tr}(\Sigma_2^3) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell] \\
&\quad + 2(n_2^{-1} \text{tr}(\Sigma_2)) \left[2n_2^{-1} \text{tr}(\Sigma_2^2) + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell)^2 \right], \\
\sigma_{560A} &= \sum_{\ell=1}^{n_2} [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2 \Sigma_1)] [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2)] \\
&= [2n_2^{-1} \text{tr}[(\Sigma_1 \Sigma_2^2)] + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell)(\mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell)].
\end{aligned}$$

Thus, we have

$$\begin{aligned}
\sigma_{11A} &= \sigma_{110A} + 4\sigma_{220A} + 4n_1^{-2} (\text{tr}(\Sigma_1))^2 \sigma_{330A} - 4\sigma_{120A} + 4n_1^{-1} (\text{tr}(\Sigma_1))(-\sigma_{130A} + 2\sigma_{230A}) \\
&= 4[2n_1^{-1} \text{tr}(\Sigma_1^4) + \beta_1 n_1^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_1^2 \mathbf{e}_\ell)^2] + 4[n_1^{-1} \text{tr}(\Sigma_1^2)]^2 + \frac{8n_2}{n_1} (n_2^{-1} \text{tr}(\Sigma_1 \Sigma_2))^2 \\
&\quad + 4[2n_1^{-1} \text{tr}[(\Sigma_1 \Sigma_2)^2] + \beta_1 n_1^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_1^{1/2} \Sigma_2 \Sigma_1^{1/2} \mathbf{e}_\ell)^2] \\
&\quad - 8[2n_1^{-1} \text{tr}(\Sigma_1^3 \Sigma_2) + \beta_1 n_1^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_1^{1/2} \Sigma_2 \Sigma_1^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_1^2 \mathbf{e}_\ell],
\end{aligned}$$

and

$$\begin{aligned}
\sigma_{22A} &= \sigma_{440A} + 4\sigma_{550A} + 4n_2^{-2} (\text{tr}(\Sigma_2))^2 \sigma_{660A} - 4\sigma_{450A} + 4n_2^{-1} (\text{tr}(\Sigma_2))(-\sigma_{460A} + 2\sigma_{560A}) \\
&= 4n_2^{-1} [2\text{tr}(\Sigma_2^4) + \beta_2 \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell)^2] + 4[n_2^{-1} \text{tr}(\Sigma_2^2)]^2 \\
&\quad + 4[2n_2^{-1} \text{tr}[(\Sigma_2 \Sigma_1)^2] + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell)^2] \\
&\quad - 8(2n_2^{-1} \text{tr}(\Sigma_2^3 \Sigma_1) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell).
\end{aligned}$$

Step 5. According to (S1.3) and (S1.4), we have

$$\sigma_{A2}^{-1}\{\text{tr}[(\mathbf{S}_1 - \mathbf{S}_2)^2] - \hat{\mu}_{21} - \mu_{A2}\} \rightarrow N(0, 1),$$

where

$$\mu_{A2} = \text{tr}[(\Sigma_1 - \Sigma_2)^2] + \sum_{k=1}^2 \frac{n_k + 1}{(n_k - 1)^2} \text{tr}(\Sigma_k^2) + \sum_{k=1}^2 \frac{\beta_k n_k}{(n_k - 1)^2} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_k \mathbf{e}_\ell)^2$$

and

$$\begin{aligned} \sigma_{A2}^2 &= \sigma_{11A} + \sigma_{22A} \\ &= 4[n_1^{-1} \text{tr}(\Sigma_1^2)]^2 + 4[n_2^{-1} \text{tr}(\Sigma_2^2)]^2 + 8n_1^{-1}n_2^{-1}[\text{tr}(\Sigma_1 \Sigma_2)]^2 \\ &\quad + 4 \left[2n_1^{-1} \text{tr}(\Sigma_1^4) + \beta_1 n_1^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_1^2 \mathbf{e}_\ell)^2 \right] \\ &\quad + 4 \{ 2n_1^{-1} \text{tr}[(\Sigma_1 \Sigma_2)^2] + \beta_1 n_1^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_1^{1/2} \Sigma_2 \Sigma_1^{1/2} \mathbf{e}_\ell)^2 \} \\ &\quad - 8[2n_1^{-1} \text{tr}(\Sigma_1^3 \Sigma_2) + \beta_1 n_1^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_1^{1/2} \Sigma_2 \Sigma_1^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_1^2 \mathbf{e}_\ell] \\ &\quad + 4n_2^{-1} \left[2\text{tr}(\Sigma_2^4) + \beta_2 \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell)^2 \right] \\ &\quad + 4 \{ 2n_2^{-1} \text{tr}[(\Sigma_2 \Sigma_1)^2] + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell)^2 \} \\ &\quad - 8[2n_2^{-1} \text{tr}(\Sigma_2^3 \Sigma_1) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell]. \end{aligned}$$

As a result, the proof of Lemma 1 is completed.

Lemma 2. *Under Assumptions (A1)–(A2), we have*

$$\begin{pmatrix} \sigma_{Ak_1k_2}^2 & \sigma_{Ak_1k_2k_2k_3} \\ \sigma_{Ak_1k_2k_2k_3} & \sigma_{Ak_2k_3j}^2 \end{pmatrix}^{-1/2} \begin{pmatrix} \text{tr}(\mathbf{S}_{k_1} - \mathbf{S}_{k_2})^2 - \hat{\mu}_{k_1k_2} - \mu_{Ak_1k_2} \\ \text{tr}(\mathbf{S}_{k_2} - \mathbf{S}_{k_3})^2 - \hat{\mu}_{k_2k_3} - \mu_{Ak_2k_3} \end{pmatrix} \rightarrow N(\mathbf{0}_2, \mathbf{I}_2),$$

where $\mathbf{0}_2 = (0, 0)^T$ and \mathbf{I}_2 is the 2×2 identity matrix with

$$\begin{aligned} \hat{\mu}_{ij} &= \sum_{k=i,j} (n_k^2 - n_k - 1)n_k^{-1}(n_k - 1)^{-2}(\text{tr}\mathbf{S}_k)^2, \\ \mu_{Aij} &= \text{tr}[(\Sigma_i - \Sigma_j)^2] + \sum_{k=i,j} \frac{n_k + 1}{(n_k - 1)^2} \text{tr}(\Sigma_k^2) + \sum_{k=i,j} \frac{\beta_k n_k}{(n_k - 1)^2} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_k \mathbf{e}_\ell)^2, \\ \sigma_{Aij}^2 &= 4[n_i^{-1} \text{tr}(\Sigma_i^2)]^2 + 4[n_j^{-1} \text{tr}(\Sigma_j^2)]^2 + 8n_i^{-1}n_j^{-1}[\text{tr}(\Sigma_i \Sigma_j)]^2 \\ &\quad + 4[2n_i^{-1} \text{tr}(\Sigma_i^4) + \beta_i n_i^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_i^2 \mathbf{e}_\ell)^2] \\ &\quad + 4\{2n_i^{-1} \text{tr}[(\Sigma_i \Sigma_j)^2] + \beta_i n_i^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_i^{1/2} \Sigma_j \Sigma_i^{1/2} \mathbf{e}_\ell)^2\} \\ &\quad - 8[2n_i^{-1} \text{tr}(\Sigma_i^3 \Sigma_j) + \beta_i n_i^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_i^{1/2} \Sigma_j \Sigma_i^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_i^2 \mathbf{e}_\ell] \\ &\quad + 4[2n_j^{-1} \text{tr}(\Sigma_j^4) + \beta_j n_j^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_j^2 \mathbf{e}_\ell)^2] \\ &\quad + 4\{2n_j^{-1} \text{tr}[(\Sigma_j \Sigma_i)^2] + \beta_j n_j^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_j^{1/2} \Sigma_i \Sigma_j^{1/2} \mathbf{e}_\ell)^2\} \\ &\quad - 8(2n_j^{-1} \text{tr}(\Sigma_j^3 \Sigma_i) + \beta_j n_j^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_j^{1/2} \Sigma_i \Sigma_j^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_j^2 \mathbf{e}_\ell)], \end{aligned}$$

$$\begin{aligned}
 & \sigma_{Ak_1k_2k_2k_3} \\
 = & 4[n_{k_2}^{-1}\text{tr}(\Sigma_{k_2}^2)]^2 + 4[2n_{k_2}^{-1}\text{tr}(\Sigma_{k_2}^4) + \beta_{k_2}n_{k_2}^{-1}\sum_{\ell=1}^p(\mathbf{e}_\ell^T\Sigma_{k_2}^2\mathbf{e}_\ell)^2] \\
 & -4[2n_{k_2}^{-1}\text{tr}(\Sigma_{k_2}^3\Sigma_{k_3}) + \beta_{k_2}n_{k_2}^{-1}\sum_{\ell=1}^p\mathbf{e}_\ell^T\Sigma_{k_2}^{1/2}\Sigma_{k_3}\Sigma_{k_2}^{1/2}\mathbf{e}_\ell\mathbf{e}_\ell^T\Sigma_{k_2}^2\mathbf{e}_\ell] \\
 & -4[2n_{k_2}^{-1}\text{tr}(\Sigma_{k_2}^3\Sigma_{k_1}) + \beta_{k_2}n_{k_2}^{-1}\sum_{\ell=1}^p\mathbf{e}_\ell^T\Sigma_{k_2}^{1/2}\Sigma_{k_1}\Sigma_{k_2}^{1/2}\mathbf{e}_\ell\mathbf{e}_\ell^T\Sigma_{k_2}^2\mathbf{e}_\ell] \\
 & +4[2n_{k_2}^{-1}\text{tr}(\Sigma_{k_1}\Sigma_{k_2}\Sigma_{k_3}\Sigma_{k_2}) + \beta_{k_2}n_{k_2}^{-1}\sum_{\ell=1}^p\mathbf{e}_\ell^T\Sigma_{k_2}^{1/2}\Sigma_{k_1}\Sigma_{k_2}^{1/2}\mathbf{e}_\ell\mathbf{e}_\ell^T\Sigma_{k_2}^{1/2}\Sigma_{k_3}\Sigma_{k_2}^{1/2}\mathbf{e}_\ell].
 \end{aligned}$$

Proof. Similar to Lemma 1, we first consider Σ_k with the bounded spectral norm for all $k = k_1, k_2, k_3$. When $\text{tr}(\Sigma_k^q) = O(p^q)$ for $q = 1, 2, 3, 4$ and at least one k in the index set $\{k_1, k_2, k_3\}$, the proof mimics that in the bounded spectral norm case. Similar to the proof of Lemma 1, it can be shown that $\text{tr}(\mathbf{S}_{k_1} - \mathbf{S}_{k_2})^2 - \hat{\mu}_{k_1k_2} - \mu_{Ak_1k_2}$ and $\text{tr}(\mathbf{S}_{k_2} - \mathbf{S}_{k_3})^2 - \hat{\mu}_{k_2k_3} - \mu_{Ak_2k_3}$ are asymptotically distributed as a bivariate normal distribution with the asymptotic variances $\sigma_{Ak_1k_2}^2$, $\sigma_{Ak_2k_3}^2$ and asymptotic covariance $\sigma_{Ak_1k_2k_2k_3}$. For simplicity of presentation, we consider the computation of σ_{A1223} , which is given by

$$\begin{aligned}
 \sigma_{A1223} = & \tilde{\sigma}_{110A} - 2\tilde{\sigma}_{130A} - 2\tilde{\sigma}_{120A} + 4\tilde{\sigma}_{230A} + 4n_2^{-2}(\text{tr}\Sigma_2)^2\sigma_{660A} - 2(n_2^{-1}\text{tr}\Sigma_2)\sigma_{460A} \\
 & +4(n_2^{-1}\text{tr}\Sigma_2)\sigma_{560A} - 2(n_2^{-1}\text{tr}\Sigma_2)\sigma_{460A} + 4(n_2^{-1}\text{tr}\Sigma_2)\sigma_{760A},
 \end{aligned}$$

where

$$\begin{aligned}
\sigma_{760A} &= \sum_{\ell=1}^{n_2} [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2 \mathbf{B}_3)][(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2)], \\
\tilde{\sigma}_{110A} &= \sum_{\ell=1}^{n_2} [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2^2)]^2, \\
\tilde{\sigma}_{120A} &= \sum_{\ell=1}^{n_2} [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2^2)][(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2 \mathbf{B}_1)], \\
\tilde{\sigma}_{130A} &= \sum_{\ell=1}^{n_2} [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2^2)][(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2 \mathbf{B}_3)], \\
\tilde{\sigma}_{230A} &= \sum_{\ell=1}^{n_2} [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2 \mathbf{B}_1)][(E_{\ell, \mathbf{B}_3} - E_{\ell-1, \mathbf{B}_3}) \text{tr}(\mathbf{B}_2 \mathbf{B}_3)].
\end{aligned}$$

Then, we have

$$\begin{aligned}
\sigma_{760A} &= \sum_{\ell=1}^{n_2} [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2 \mathbf{B}_3)][(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2)] \\
&= [2n_2^{-1} \text{tr}(\Sigma_3 \Sigma_2^2) + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell)(\mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_3 \Sigma_2^{1/2} \mathbf{e}_\ell)], \\
\tilde{\sigma}_{110A} &= \sum_{\ell=1}^{n_2} [(E_\ell - E_{\ell-1}) \text{tr}(\mathbf{B}_2^2)]^2 \\
&= 4n_2^{-1} [2\text{tr}(\Sigma_2^4) + \beta_2 \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell)^2] + 4(n_2^{-1} \text{tr} \Sigma_2)^2 n_2^{-1} [2\text{tr}(\Sigma_2^2) + \beta_2 \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell)^2] \\
&\quad + 4[n_2^{-1} \text{tr}(\Sigma_2^2)]^2 + 8(n_2^{-1} \text{tr} \Sigma_2) n_2^{-1} [2\text{tr}(\Sigma_2^3) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell],
\end{aligned}$$

$$\begin{aligned}
 \tilde{\sigma}_{120A} &= \sum_{\ell=1}^{n_2} [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1}) \text{tr}(\mathbf{B}_2^2)] [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1}) \text{tr}(\mathbf{B}_2 \mathbf{B}_1)] \\
 &= 2n_2^{-1} [2\text{tr}(\Sigma_2^3 \mathbf{B}_1) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \mathbf{B}_1 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell] \\
 &\quad + 2(n_2^{-1} \text{tr} \Sigma_2) n_2^{-1} [2\text{tr}(\Sigma_2^2 \mathbf{B}_1) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^{1/2} \mathbf{B}_1 \Sigma_2^{1/2} \mathbf{e}_\ell] \\
 &= 2n_2^{-1} [2\text{tr}(\Sigma_2^3 \Sigma_1) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell] \\
 &\quad + 2(n_2^{-1} \text{tr} \Sigma_2) n_2^{-1} [2\text{tr}(\Sigma_2^2 \Sigma_1) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell] + o_p(1), \\
 \tilde{\sigma}_{130A} &= \sum_{\ell=1}^{n_2} [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1}) \text{tr}(\mathbf{B}_2^2)] [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1}) \text{tr}(\mathbf{B}_2 \mathbf{B}_3)] \\
 &= 2n_2^{-1} [2\text{tr}(\Sigma_2^3 \mathbf{B}_3) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \mathbf{B}_3 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell] \\
 &\quad + 2(n_2^{-1} \text{tr} \Sigma_2) n_2^{-1} [2\text{tr}(\Sigma_2^2 \mathbf{B}_3) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^{1/2} \mathbf{B}_3 \Sigma_2^{1/2} \mathbf{e}_\ell] \\
 &= 2n_2^{-1} [2\text{tr}(\Sigma_2^3 \Sigma_3) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_3 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell] \\
 &\quad + 2(n_2^{-1} \text{tr} \Sigma_2) n_2^{-1} [2\text{tr}(\Sigma_2^2 \Sigma_3) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_3 \Sigma_2^{1/2} \mathbf{e}_\ell] + o_p(1), \\
 \tilde{\sigma}_{230A} &= \sum_{\ell=1}^{n_2} [(\mathbf{E}_\ell - \mathbf{E}_{\ell-1}) \text{tr}(\mathbf{B}_2 \mathbf{B}_1)] [(\mathbf{E}_{\ell, \mathbf{B}_3} - \mathbf{E}_{\ell-1, \mathbf{B}_3}) \text{tr}(\mathbf{B}_2 \mathbf{B}_3)] \\
 &= n_2^{-1} [2\text{tr}(\mathbf{B}_1 \Sigma_2 \mathbf{B}_3 \Sigma_2) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \mathbf{B}_1 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^{1/2} \mathbf{B}_3 \Sigma_2^{1/2} \mathbf{e}_\ell] \\
 &= [2n_2^{-1} \text{tr}(\Sigma_1 \Sigma_2 \Sigma_3 \Sigma_2) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_3 \Sigma_2^{1/2} \mathbf{e}_\ell] + o_p(1).
 \end{aligned}$$

Thus, we obtain

$$\begin{aligned}
\sigma_{A1223} = & 4[n_2^{-1}\text{tr}(\Sigma_2^2)]^2 + 4n_2^{-1}[2\text{tr}(\Sigma_2^4) + \beta_2 \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell)^2] \\
& + 4(n_2^{-1}\text{tr}\Sigma_2)^2 n_2^{-1}[2\text{tr}(\Sigma_2^2) + \beta_2 \sum_{\ell} (\mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell)^2] \\
& + 8(n_2^{-1}\text{tr}\Sigma_2)n_2^{-1}[2\text{tr}(\Sigma_2^3) + \beta_2 \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell] \\
& - 4[2n_2^{-1}\text{tr}(\Sigma_2^3 \Sigma_3) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_3 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell] \\
& - 4(n_2^{-1}\text{tr}\Sigma_2)[2n_2^{-1}\text{tr}(\Sigma_2^2 \Sigma_3) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_3 \Sigma_2^{1/2} \mathbf{e}_\ell] \\
& - 4[2n_2^{-1}\text{tr}(\Sigma_2^3 \Sigma_1) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell] \\
& - 4(n_2^{-1}\text{tr}\Sigma_2)[2n_2^{-1}\text{tr}(\Sigma_2^2 \Sigma_1) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell] \\
& + 4[2n_2^{-1}\text{tr}(\Sigma_1 \Sigma_2 \Sigma_3 \Sigma_2) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_3 \Sigma_2^{1/2} \mathbf{e}_\ell] \\
& + 4(n_2^{-1}\text{tr}\Sigma_2)^2[2n_2^{-1}\text{tr}(\Sigma_2^2) + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell)^2] \\
& - 4(n_2^{-1}\text{tr}\Sigma_2)[2n_2^{-1}\text{tr}(\Sigma_2^3) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell] \\
& - 4(n_2^{-1}\text{tr}\Sigma_2)(n_2^{-1}\text{tr}\Sigma_2)[2n_2^{-1}\text{tr}(\Sigma_2^2) + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell)^2] \\
& + 4(n_2^{-1}\text{tr}\Sigma_2)[2n_2^{-1}\text{tr}(\Sigma_1 \Sigma_2^2) + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell)(\mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell)] \\
& - 4(n_2^{-1}\text{tr}\Sigma_2)[2n_2^{-1}\text{tr}(\Sigma_2^3) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell] \\
& - 4(n_2^{-1}\text{tr}\Sigma_2)(n_2^{-1}\text{tr}\Sigma_2)[2n_2^{-1}\text{tr}(\Sigma_2^2) + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell)^2] \\
& + 4(n_2^{-1}\text{tr}\Sigma_2)[2n_2^{-1}\text{tr}(\Sigma_3 \Sigma_2^2) + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2 \mathbf{e}_\ell)(\mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_3 \Sigma_2^{1/2} \mathbf{e}_\ell)]
\end{aligned}$$

$$\begin{aligned}
 &= 4[n_2^{-1}\text{tr}(\Sigma_2^2)]^2 + 4[2n_2^{-1}\text{tr}(\Sigma_2^4) + \beta_2 n_2^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell)^2] \\
 &\quad - 4[2n_2^{-1}\text{tr}(\Sigma_2^3 \Sigma_3) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_3 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell] \\
 &\quad - 4[2n_2^{-1}\text{tr}(\Sigma_2^3 \Sigma_1) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^2 \mathbf{e}_\ell] \\
 &\quad + 4[2n_2^{-1}\text{tr}(\Sigma_1 \Sigma_2 \Sigma_3 \Sigma_2) + \beta_2 n_2^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_1 \Sigma_2^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_2^{1/2} \Sigma_3 \Sigma_2^{1/2} \mathbf{e}_\ell].
 \end{aligned}$$

Generally, $\sigma_{Ak_1k_2k_2k_3}$ is obtained by replacing n_i , Σ_i and β_i by n_{k_i} , Σ_{k_i} and β_{k_i} in σ_{A1223} . That is,

$$\begin{aligned}
 \sigma_{Ak_1k_2k_2k_3} &= 4[n_{k_2}^{-1}\text{tr}(\Sigma_{k_2}^2)]^2 + 4[2n_{k_2}^{-1}\text{tr}(\Sigma_{k_2}^4) + \beta_{k_2} n_{k_2}^{-1} \sum_{\ell=1}^p (\mathbf{e}_\ell^T \Sigma_{k_2}^2 \mathbf{e}_\ell)^2] \\
 &\quad - 4[2n_{k_2}^{-1}\text{tr}(\Sigma_{k_2}^3 \Sigma_{k_3}) + \beta_{k_2} n_{k_2}^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_{k_2}^{1/2} \Sigma_{k_3} \Sigma_{k_2}^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_{k_2}^2 \mathbf{e}_\ell] \\
 &\quad - 4[2n_{k_2}^{-1}\text{tr}(\Sigma_{k_2}^3 \Sigma_{k_1}) + \beta_{k_2} n_{k_2}^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_{k_2}^{1/2} \Sigma_{k_1} \Sigma_{k_2}^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_{k_2}^2 \mathbf{e}_\ell] \\
 &\quad + 4[2n_{k_2}^{-1}\text{tr}(\Sigma_{k_1} \Sigma_{k_2} \Sigma_{k_3} \Sigma_{k_2}) + \beta_{k_2} n_{k_2}^{-1} \sum_{\ell=1}^p \mathbf{e}_\ell^T \Sigma_{k_2}^{1/2} \Sigma_{k_1} \Sigma_{k_2}^{1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \Sigma_{k_2}^{1/2} \Sigma_{k_3} \Sigma_{k_2}^{1/2} \mathbf{e}_\ell].
 \end{aligned}$$

The proof of Lemma 2 is completed.

S2 Proofs of Theorem 2 and Theorem 1

By Lemma 2 and the delta method, under the conditions of Lemma 2, we have

$$\sigma_{AK}^{-1}(T_{K1} - \hat{\mu}_{K1} - \mu_{AK}) \xrightarrow{d} N(0, 1),$$

where

$$\hat{\mu}_{K1} = \sum_{1 \leq k_1 < k_2 \leq K} \omega_{k_1 k_2} \hat{\mu}_{k_1 k_2} = \sum_{1 \leq k_1 < k_2 \leq K} \omega_{k_1 k_2} \left[\sum_{k=k_1, k_2} (n_k^2 - n_k - 1) n_k^{-1} (n_k - 1)^{-2} (\text{tr} \mathbf{S}_k)^2 \right],$$

and $\mu_{AK} = \sum_{1 \leq i < j \leq K} \omega_{ij} \mu_{Aij}$ and

$$\begin{aligned} \sigma_{AK}^2 &= \sum_{1 \leq i < j \leq K} \omega_{ij}^2 \sigma_{Aij}^2 + 2 \sum_{i < j < k} \omega_{ij} \omega_{jk} \sigma_{Aijjk} \\ &\quad + 2 \sum_{i < k < j} \omega_{ij} \omega_{jk} \sigma_{Aijjk} + 2 \sum_{j < i < k} \omega_{ij} \omega_{jk} \sigma_{Aijjk}, \end{aligned}$$

with the weights $\{\omega_{ij}, 1 \leq i, j \leq K\}$ and $\omega_{ij} = \omega_{ji}$. By (32) and (33) in

Cai, Liu and Xia (2013), we have

$$\begin{aligned} &\max_{1 \leq \ell_1 \leq \ell_2 \leq p} \delta_{k_1 k_2 \ell_1 \ell_2} - s(n_{k_1}, n_{k_2}, p) \\ &\stackrel{a.s.}{\geq} 0.5 \max_{1 \leq i \leq j \leq p} \frac{(\sigma_{k_1 ij} - \sigma_{k_2 ij})^2}{\hat{\theta}_{k_1 ij}/n_{k_1} + \hat{\theta}_{k_2 ij}/n_{k_2}} - s(n_{k_1}, n_{k_2}, p) - 4 \log p + 0.5 \log \log p \\ &\stackrel{a.s.}{\geq} 0.5 \max_{1 \leq i \leq j \leq p} \frac{(\sigma_{k_1 ij} - \sigma_{k_2 ij})^2}{\theta_{k_1 ij}/n_{k_1} + \theta_{k_2 ij}/n_{k_2}} - s(n_{k_1}, n_{k_2}, p) - 4 \log p + 0.5 \log \log p \end{aligned}$$

Then, there exists a pair of $1 \leq k_1, k_2 \leq K$ satisfying

$$0.5 \max_{1 \leq i \leq j \leq p} \frac{(\sigma_{k_1 ij} - \sigma_{k_2 ij})^2}{\theta_{k_1 ij}/n_{k_1} + \theta_{k_2 ij}/n_{k_2}} \geq s(n_1, n_2, p) + 4 \log p.$$

Then we have

$$\max_{1 \leq \ell_1 \leq \ell_2 \leq p} \delta_{k_1 k_2 \ell_1 \ell_2} - s(n_{k_1}, n_{k_2}, p) \stackrel{a.s.}{\geq} 0.5 \log \log p$$

and

$$\max_{1 \leq \ell_1 \leq \ell_2 \leq p} \delta_{k_1 k_2 \ell_1 \ell_2} - s(n_{k_1}, n_{k_2}, p) \stackrel{a.s.}{>} 0$$

Therefore, we have

$$\max_{1 \leq i < j \leq K} \left\{ I \left\{ \max_{1 \leq \ell_1 \leq \ell_2 \leq p} \delta_{ij\ell_1\ell_2} > s(n_i, n_j, p) \right\} \right\} \stackrel{a.s.}{=} 1.$$

That is, $T_{K2} \stackrel{a.s.}{=} K_0$. Then, we have

$$\sigma_{AK}^{-1}(T_K - K_0 - \hat{\mu}_{K1} - \mu_{AK}) \xrightarrow{d} N(0, 1).$$

We first focus on the proof of the case with $K = 2$, and that of $K \geq 3$ can be shown in a similar way. When $K = 2$, the power function is

$$\begin{aligned} g_2(\Sigma_1, \Sigma_2) &= P_{H_{A2}}(T_2 - \hat{\mu}_{21} > \hat{\mu}_2 + z_{1-\alpha}\hat{\sigma}_2) \\ &= P_{H_{A2}} \left(\frac{T_2 - \hat{\mu}_{21} - \mu_{A2}}{\sigma_{A2}} > \frac{\hat{\mu}_2 - \mu_{A2} + z_{1-\alpha}\hat{\sigma}_2}{\sigma_{A2}} \right) \\ &> P_{H_{A2}} \left(\frac{T_{21} - \hat{\mu}_{21} - \mu_{A2}}{\sigma_{A2}} > \frac{z_{1-\alpha}\hat{\sigma}_2}{\sigma_{A2}} \right). \end{aligned}$$

Step 1. When $p^{-1}\text{tr}\mathbf{A}_{12}^2 \rightarrow 0$ and $\text{tr}\mathbf{A}_{12}^2 > c_0$ for some positive constant c_0 , $\sigma_{A2}/\hat{\sigma}_2 \rightarrow 1$. That is, $(z_{1-\alpha}\hat{\sigma}_2)/\sigma_{A2} - z_{1-\alpha} = o_p(1)$. Then, when n_1, n_2 are large enough, we have $g_2(\Sigma_1, \Sigma_2) > \alpha$.

Step 2. We have $g_2(\Sigma_1, \Sigma_2) > P_{H_{A2}}((T_{21} - \hat{\mu}_{21} - \mu_{A2})\sigma_{A2}^{-1} > (\hat{\mu}_2 - \mu_{A2} + z_{1-\alpha}\hat{\sigma}_2)\sigma_{A2}^{-1})$. When $\text{tr}(\mathbf{A}_{12}^2) \rightarrow \infty$, we have $(\hat{\mu}_2 - \mu_{A2})\sigma_{A2}^{-1} \rightarrow -\infty$ in probability. Then, the power function satisfies $g_2(\Sigma_1, \Sigma_2) \rightarrow 1$.

Step 3. We have

$$\begin{aligned}
 g_2(\Sigma_1, \Sigma_2) &= P_{H_{A2}}(T_2 - \hat{\mu}_{21} > \hat{\mu}_2 + z_{1-\alpha}\hat{\sigma}_2) \\
 &= P_{H_{A2}}\left(\frac{T_2 - \hat{\mu}_{21} - \mu_{A2}}{\sigma_{A2}} > \frac{\hat{\mu}_2 - \mu_{A2} + z_{1-\alpha}\hat{\sigma}_2}{\sigma_{A2}}\right) \\
 &> P_{H_{A2}}\left(\frac{T_2 - \hat{\mu}_{21} - \mu_{A2}}{\sigma_{A2}} > \frac{z_{1-\alpha}\hat{\sigma}_2}{\sigma_{A2}}\right) \\
 &> P_{H_{A2}}\left(\frac{T_{21} - \hat{\mu}_{21} - \mu_{A2}}{\sigma_{A2}} + \frac{T_{K2}}{\sigma_{A2}} > \frac{z_{1-\alpha}\hat{\sigma}_2}{\sigma_{A2}}\right).
 \end{aligned}$$

Because $\frac{T_{21} - \hat{\mu}_{21} - \mu_{A2}}{\sigma_{A2}}$ is asymptotically normal under H_1 and $T_{K2} \xrightarrow{a.s.} K_0$, if

$$0.5 \max_{1 \leq i \leq j \leq p} \frac{(\sigma_{k_1ij} - \sigma_{k_2ij})^2}{\theta_{k_1ij}/n_{k_1} + \theta_{k_2ij}/n_{k_2}} \geq s(n_1, n_2, p) + 4 \log p,$$

then the power function $g_2(\Sigma_1, \Sigma_2)$ tends to one.

The proof of Theorem 2 is completed. Moreover, Theorem 1 is a special case of Theorem 2.

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Table 1: Comparison of empirical sizes and power (in percentage) of the proposed two-sample test T_2 with four existing methods under Scenario 1.

p	n_1	n_2	Gaussian				Gamma				
			T_2	CLX	LC	SC	YP	T_2	CLX	LC	YP
			Size (%)								
100	60	60	6.1	5.3	4.9	5.2	7.1	5.8	3.3	6.3	21.4
	100	200	5.8	5.5	4.2	5.0	—	5.4	3.6	6.6	—
	200	200	6.3	4.9	3.8	5.5	5.0	5.7	3.4	4.2	13.5
	300	300	5.9	4.6	5.4	5.4	6.0	5.9	3.8	5.2	11.9
300	60	60	5.9	6.0	4.2	4.9	10.5	5.0	3.4	4.6	22.5
	100	200	5.2	4.5	7.8	5.2	—	4.9	3.2	5.2	—
	200	200	5.6	4.4	4.2	5.2	4.1	5.4	2.8	6.0	13.9
	300	300	5.6	4.2	5.6	5.4	6.7	5.4	3.2	4.0	13.4
Power (%)											
100	60	60	63.8	6.9	62.2	66.2	94.6	63.3	4.7	62.8	95.6
	100	200	68.7	8.7	68.8	70.2	—	67.1	6.2	66.4	—
	200	200	32.1	6.2	30.8	32.0	36.8	31.9	4.4	29.8	44.0
	300	300	26.1	5.6	27.2	26.1	24.2	26.1	4.7	26.8	36.0
300	60	60	99.9	7.5	100.0	100.0	99.8	99.9	4.9	100.0	97.5
	100	200	100.0	9.9	100.0	100.0	—	100.0	7.4	100.0	—
	200	200	96.4	5.6	95.8	96.6	100.0	95.8	3.5	96.0	100.0
	300	300	87.4	5.8	88.6	87.7	99.9	87.6	3.8	86.8	100.0

Note: Four existing tests include Yang and Pan (2017) (YP), Li and Chen (2012) (LC), Cai, Liu and Xia (2013) (CLX) and Schott (2007) (SC), and “—” denotes “not applicable”.

Table 2: Comparison of empirical sizes and power (in percentage) of the proposed two-sample test T_2 with four existing methods under Scenario 2.

p	n_1	n_2	Gaussian				Gamma				
			T_2	CLX	LC	SC	YP	T_2	CLX	LC	YP
Size (%)											
100	60	60	6.6	5.7	5.0	5.7	100.0	6.0	3.8	6.1	100.0
	100	200	5.5	5.9	4.8	4.4	—	5.8	4.4	5.6	—
	200	200	6.3	4.9	4.0	5.7	99.8	6.1	3.7	6.6	97.7
	300	300	5.6	4.7	2.6	5.3	70.6	5.4	3.3	5.4	96.3
300	60	60	5.8	5.4	5.6	4.9	100.0	5.1	3.4	4.9	100.0
	100	200	5.8	4.9	5.6	5.1	—	6.5	3.8	4.4	—
	200	200	6.1	4.3	4.4	6.0	100.0	5.8	3.2	4.8	100.0
	300	300	5.4	4.6	6.8	5.1	53.5	5.0	3.7	4.4	99.9
Power (%)											
100	60	60	37.9	8.4	38.5	38.9	100.0	38.0	6.5	40.4	100.0
	100	200	90.0	56.3	89.8	88.7	—	88.8	53.5	86.2	—
	200	200	99.3	53.7	99.6	99.3	100.0	99.1	42.8	99.2	99.9
	300	300	100.0	91.8	100.0	100.0	99.8	100.0	81.0	100.0	95.3
300	60	60	36.2	6.4	37.8	39.0	100.0	36.8	4.8	38.4	100.0
	100	200	91.5	53.7	85.0	90.5	—	91.4	47.9	89.4	—
	200	200	99.7	44.0	99.8	99.7	100.0	99.7	31.4	99.2	100.0
	300	300	100.0	90.3	100.0	100.0	100.0	100.0	75.7	100.0	99.7

Table 3: Comparison of empirical sizes and power (in percentage) of the proposed two-sample test T_2 with four existing methods under Scenario 3.

p	n_1	n_2	Gaussian				Gamma				
			T_2	CLX	LC	SC	YP	T_2	CLX	LC	YP
Size (%)											
100	60	60	6.3	5.2	5.6	5.2	100.0	5.5	3.3	4.5	100.0
	100	200	5.9	4.5	5.0	5.3	—	5.7	3.5	7.0	—
	200	200	5.8	4.6	6.8	5.6	50.0	5.8	3.1	5.4	99.5
300	300	300	5.6	4.4	5.4	5.1	9.4	5.5	4.0	3.6	30.8
	60	60	5.4	5.4	5.1	4.8	99.7	4.6	3.1	5.2	100.0
	100	200	5.6	5.4	6.2	5.0	—	5.0	3.5	5.8	—
300	200	200	5.5	4.4	4.4	4.9	77.3	5.2	2.6	6.4	9.5
	300	300	5.2	4.7	5.6	5.0	23.9	5.5	3.5	4.6	80.6
Power (%)											
100	60	60	43.5	53.8	7.9	11.5	100.0	37.2	42.9	16.0	100.0
	100	200	99.0	99.6	13.4	24.5	—	96.1	98.3	29.8	—
	200	200	100.0	100.0	43.2	41.2	100.0	99.8	99.9	36.6	100.0
300	300	300	100.0	100.0	85.4	66.2	100.0	100.0	100.0	88.6	100.0
	60	60	24.2	33.3	5.9	6.5	100.0	20.0	25.7	6.6	100.0
	100	200	95.8	98.5	13.6	8.9	—	89.9	95.4	11.6	—
300	200	200	99.9	100.0	6.8	12.0	99.9	99.1	99.7	11.0	100.0
	300	300	100.0	100.0	27.0	18.2	100.0	100.0	100.0	12.6	100.0

Table 4: Comparison of empirical sizes and power (in percentage) of the proposed two-sample test T_2 with four existing methods under Scenario 4.

p	n_1	n_2	Gaussian				Gamma				
			T_2	CLX	LC	SC	YP	T_2	CLX	LC	YP
Size (%)											
100	60	60	6.0	5.4	4.5	5.3	8.0	5.0	3.8	5.5	17.8
	100	200	5.5	4.5	5.0	5.2	—	6.2	3.9	6.0	—
	200	200	5.8	4.2	7.2	5.2	6.5	6.1	3.3	6.0	11.6
	300	300	5.1	4.7	4.4	4.8	5.5	5.5	3.8	5.0	12.8
300	60	60	6.0	5.4	4.7	4.9	7.7	4.9	3.0	6.6	20.1
	100	200	5.2	5.5	4.4	4.6	—	5.3	3.5	4.6	—
	200	200	4.8	5.1	4.8	4.6	5.8	5.4	2.7	5.2	11.3
	300	300	5.3	4.4	5.4	5.0	5.7	5.4	3.3	4.8	13.7
Power (%)											
100	60	60	36.2	17.6	31.0	33.8	86.7	32.9	8.7	34.4	84.9
	100	200	96.1	98.1	47.2	48.7	—	71.7	69.4	48.3	—
	200	200	99.2	99.8	67.2	70.3	99.7	85.5	79.6	65.2	97.5
	300	300	100.0	100.0	87.2	89.5	99.9	98.1	97.3	83.0	99.2
300	60	60	65.1	16.1	62.0	67.2	96.7	65.0	7.7	65.4	79.3
	100	200	99.3	99.7	57.0	55.5	—	81.9	76.5	80.3	—
	200	200	99.8	99.9	68.0	68.0	100.0	89.2	83.8	67.2	99.8
	300	300	100.0	100.0	78.5	79.7	100.0	98.6	98.1	75.6	100.0

Table 5: Empirical sizes and empirical power (in percentage) of the proposed two-sample test T_2 for ultra high-dimensional cases under Scenarios 1–4.

p	n_1	n_2	Size (%)				Power (%)			
			1	2	3	4	1	2	3	4
500	60	60	5.5	5.8	5.9	5.5	100.0	35.5	18.8	84.7
	100	200	5.0	5.2	5.5	5.6	100.0	91.8	93.7	99.7
	200	200	4.9	5.5	5.4	5.0	100.0	99.8	99.8	99.9
	300	300	5.5	5.2	5.4	5.1	100.0	100.0	100.0	100.0
1000	60	60	4.8	5.2	5.3	4.5	100.0	34.5	14.1	98.1
	100	200	5.1	5.3	4.9	5.3	100.0	92.4	90.0	100.0
	200	200	5.7	5.4	4.3	4.7	100.0	99.5	99.4	100.0
	300	300	4.9	5.2	5.4	4.9	100.0	100.0	100.0	100.0

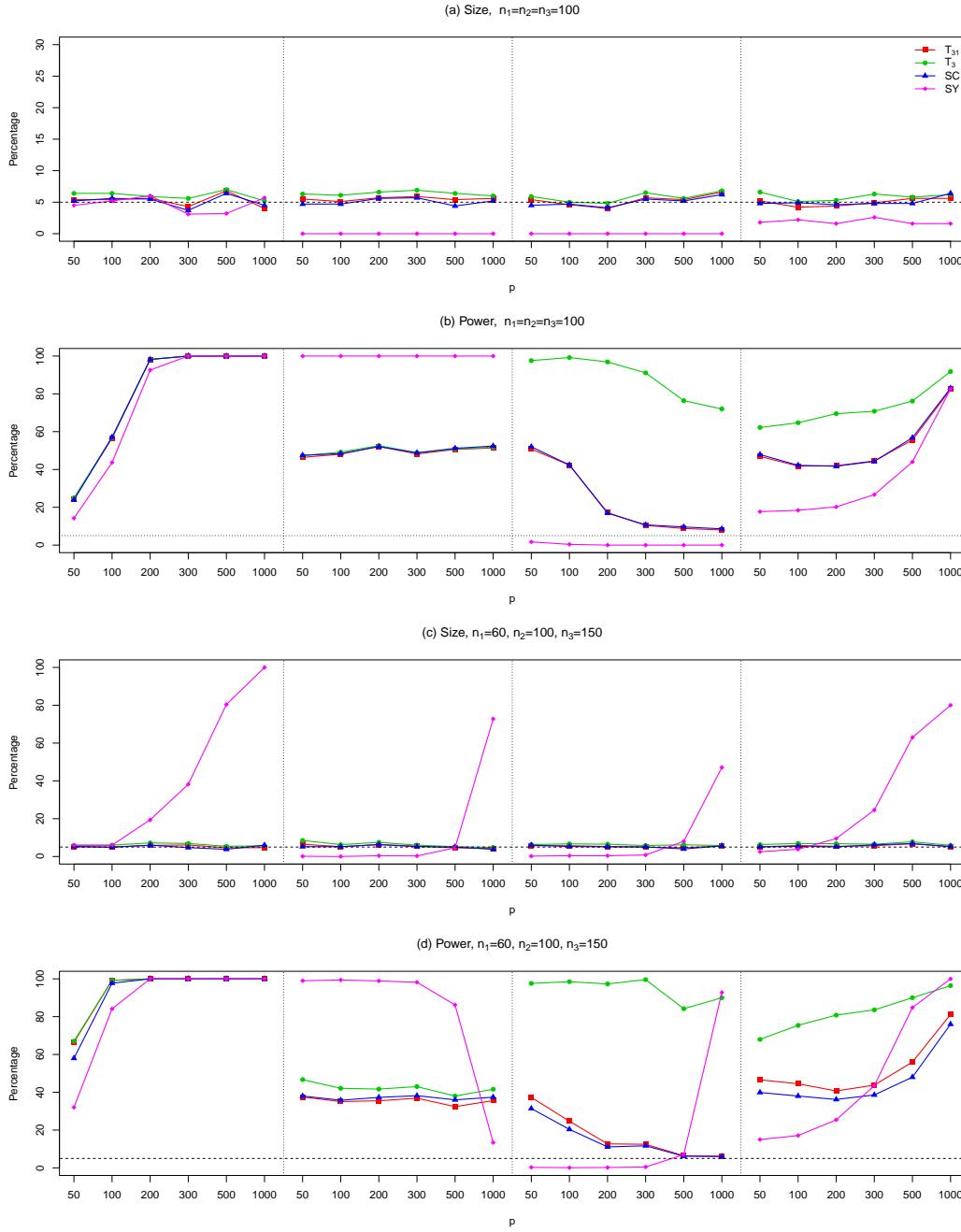


Figure 1: Simulation results for testing the equality of three covariance matrices with Gaussian populations under Scenarios 1–4 in comparison with two existing tests of Schott (2007) (SC) and Srivastava and Yanagihara (2010) (SY).

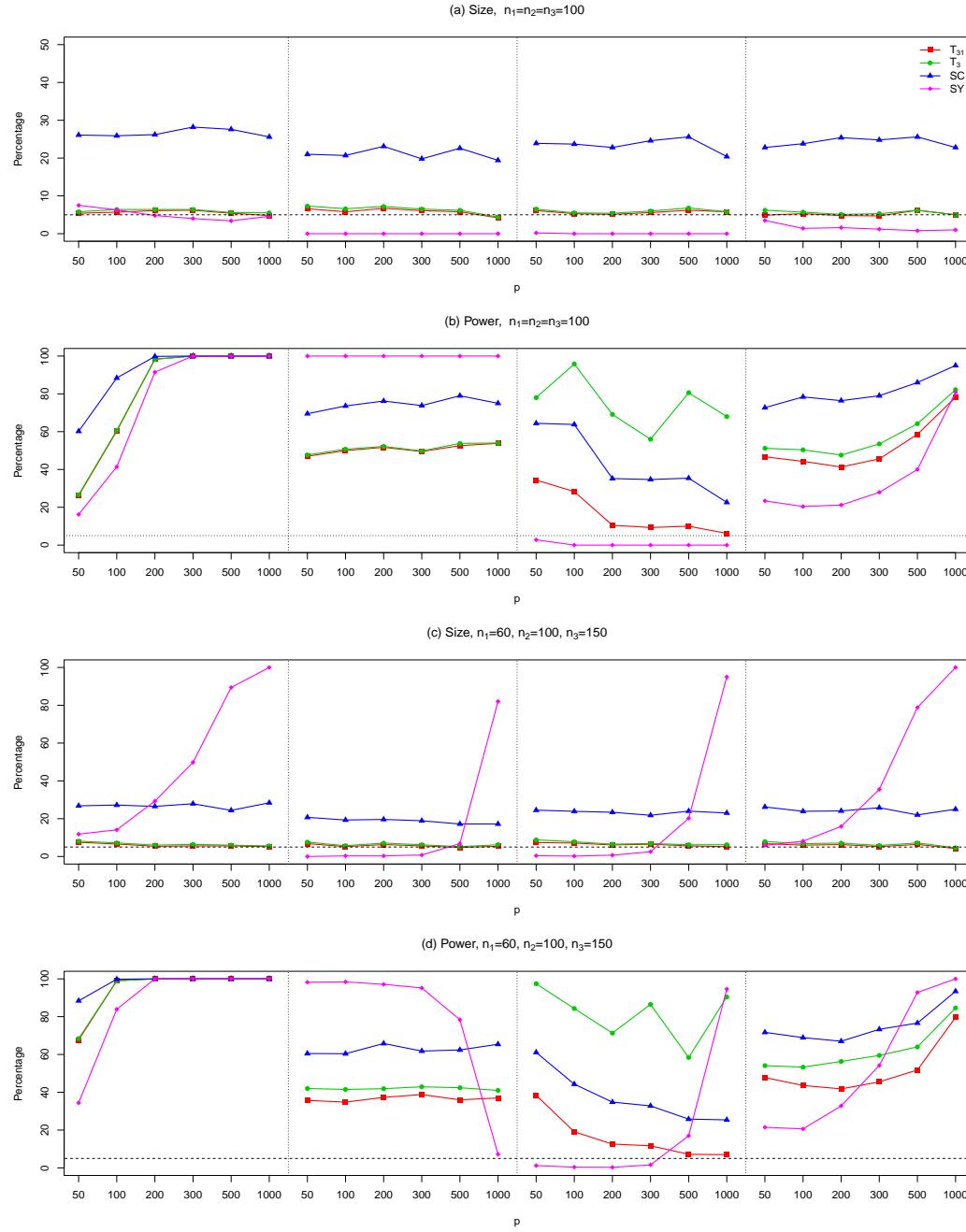


Figure 2: Simulation results for testing the equality of three covariance matrices with Gamma populations under Scenarios 1–4 in comparison with two existing tests of Schott (2007) (SC) and Srivastava and Yanagihara (2010) (SY).