

Least Favorable Direction Test for Multivariate Analysis of Variance in High Dimension

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Supplementary Material

This supplement contains the proofs of Propositions and Theorems given in the main text.

S1 Technical lemmas

Lemma 1. *Suppose \mathbf{A} is a $p \times r$ matrix with rank r and \mathbf{B} is a $p \times p$ non-zero positive semi-definite matrix. Denote by $\mathbf{A} = \mathbf{U}_\mathbf{A} \mathbf{D}_\mathbf{A} \mathbf{V}_\mathbf{A}^\top$ the singular value decomposition of \mathbf{A} , where $\mathbf{U}_\mathbf{A}$ and $\mathbf{V}_\mathbf{A}$ are $p \times r$ and $r \times r$ column orthogonal matrices, respectively, and $\mathbf{D}_\mathbf{A}$ is a $r \times r$ diagonal matrix. Let $\mathbf{P}_\mathbf{A} = \mathbf{U}_\mathbf{A} \mathbf{U}_\mathbf{A}^\top$ be the projection matrix onto the column space of \mathbf{A} . Then*

$$\max_{a^\top \mathbf{A} = 1, a^\top \mathbf{A} \mathbf{A}^\top a = 0} a^\top \mathbf{B} a = \lambda_1(\mathbf{B}(\mathbf{I}_p - \mathbf{P}_\mathbf{A})).$$

Proof. It can be seen that $a^\top \mathbf{A} \mathbf{A}^\top a = 0$ if and only if $a = (\mathbf{I}_p - \mathbf{P}_\mathbf{A})a$.

Then

$$\max_{a^\top \mathbf{A} = 1, a^\top \mathbf{A} \mathbf{A}^\top a = 0} a^\top \mathbf{B} a = \max_{a^\top \mathbf{A} = 1, \mathbf{P}_\mathbf{A} a = 0} a^\top (\mathbf{I}_p - \mathbf{P}_\mathbf{A}) \mathbf{B} (\mathbf{I}_p - \mathbf{P}_\mathbf{A}) a, \quad (\text{S1.1})$$

which is obviously no greater than $\lambda_1((\mathbf{I} - \mathbf{P}_\mathbf{A}) \mathbf{B} (\mathbf{I} - \mathbf{P}_\mathbf{A}))$. To prove that they are equal, without loss of generality, we can assume $\lambda_1((\mathbf{I} - \mathbf{P}_\mathbf{A}) \mathbf{B} (\mathbf{I} - \mathbf{P}_\mathbf{A})) > 0$. Let α_1 be one eigenvector corresponding to the largest eigenvalue of $(\mathbf{I} - \mathbf{P}_\mathbf{A}) \mathbf{B} (\mathbf{I} - \mathbf{P}_\mathbf{A})$. Since $(\mathbf{I} - \mathbf{P}_\mathbf{A}) \mathbf{B} (\mathbf{I} - \mathbf{P}_\mathbf{A}) \mathbf{P}_\mathbf{A} = (\mathbf{I} - \mathbf{P}_\mathbf{A}) \mathbf{B} (\mathbf{P}_\mathbf{A} - \mathbf{P}_\mathbf{A}) = \mathbf{O}_{p \times p}$ and $\mathbf{P}_\mathbf{A}$ is symmetric, the rows of $\mathbf{P}_\mathbf{A}$ are eigenvectors of $(\mathbf{I} - \mathbf{P}_\mathbf{A}) \mathbf{B} (\mathbf{I} - \mathbf{P}_\mathbf{A})$ corresponding to eigenvalue 0. It follows that $\mathbf{P}_\mathbf{A} \alpha_1 = 0$. Therefore, α_1 satisfies the constraint of (S1.1) and thus (S1.1) is no less than $\lambda_1((\mathbf{I} - \mathbf{P}_\mathbf{A}) \mathbf{B} (\mathbf{I} - \mathbf{P}_\mathbf{A}))$. The conclusion now follows by noting that $\lambda_1((\mathbf{I} - \mathbf{P}_\mathbf{A}) \mathbf{B} (\mathbf{I} - \mathbf{P}_\mathbf{A})) = \lambda_1(\mathbf{B} (\mathbf{I} - \mathbf{P}_\mathbf{A}))$.

□

Lemma 2. Let $\xi_{n,i}$, $i = 1, \dots, n$, $n = 1, 2, \dots$, be iid s -dimensional random vectors with mean zero, covariance matrix \mathbf{M} and finite fourth moment. For $n = 1, 2, \dots$, let $\{a_{n,i}\}_{i=1}^n$ be real random variables which are independent of $\{\xi_{n,i}\}_{i=1}^n$ and satisfy

$$\frac{\max_{1 \leq i \leq n} a_{n,i}^2}{\sum_{i=1}^n a_{n,i}^2} \xrightarrow{P} 0. \quad (\text{S1.2})$$

Then

$$\left(\sum_{i=1}^n a_{n,i}^2 \right)^{-1/2} \sum_{i=1}^n a_{n,i} \xi_{n,i} \xrightarrow{\mathcal{L}} \mathcal{N}_s(\mathbf{0}_s, \mathbf{M}).$$

Proof. First we observe that if $\{a_{n,i}\}_{i=1}^n$ are fixed numbers satisfying (S1.2), then Lyapunov central limit theorem and continuity theorem imply that for any $t \in \mathbb{R}^s$,

$$\mathbb{E} \left[\exp \left(\left(\sum_{i=1}^n a_{n,i}^2 \right)^{-1/2} \sum_{i=1}^n a_{n,i} i t^\top \xi_{n,i} \right) \right] \rightarrow \exp \left(-\frac{1}{2} t^\top \mathbf{M} t \right).$$

We only need to prove that for every subsequence of $\{n\}$, there is a further subsequence along which the conclusion holds. Let $\{m(n)\}$ be a subsequence of $\{n\}$. We can find a further subsequence of $\{m(n)\}$ along which (S1.2) holds almost surely. Then along this subsequence, our previous argument implies that for any $t \in \mathbb{R}^s$,

$$\mathbb{E} \left[\exp \left(\left(\sum_{i=1}^n a_{n,i}^2 \right)^{-1/2} \sum_{i=1}^n a_{n,i} i t^\top \xi_{n,i} \right) \middle| a_{n,1}, \dots, a_{n,n} \right] \rightarrow \exp \left(-\frac{1}{2} t^\top \mathbf{M} t \right)$$

almost surely. Then by dominated convergence theorem, we have

$$\mathbb{E} \left[\exp \left(\left(\sum_{i=1}^n a_{n,i}^2 \right)^{-1/2} \sum_{i=1}^n a_{n,i} i t^\top \xi_{n,i} \right) \right] \rightarrow \exp \left(-\frac{1}{2} t^\top \mathbf{M} t \right)$$

along this further subsequence. This implies the conclusion holds along this further subsequence, which completes the proof.

□

Lemma 3 (Weyl's inequality). *Let \mathbf{A} and \mathbf{B} be two symmetric $n \times n$ ma-*

trices. If $r + s - 1 \leq i \leq j + k - n$, we have

$$\lambda_j(\mathbf{A}) + \lambda_k(\mathbf{B}) \leq \lambda_i(\mathbf{A} + \mathbf{B}) \leq \lambda_r(\mathbf{A}) + \lambda_s(\mathbf{B}).$$

See, for example, Horn and Johnson (2012) Theorem 4.3.1.

Lemma 4 (von Neumann's trace theorem). *Let \mathbf{A} and \mathbf{B} be two $m \times n$ matrices. Let $\sigma_1(\mathbf{A}) \geq \dots \geq \sigma_q(\mathbf{A})$ and $\sigma_1(\mathbf{B}) \geq \dots \geq \sigma_q(\mathbf{B})$ denote the non-increasingly ordered singular values of \mathbf{A} and \mathbf{B} , respectively. Then*

$$\text{tr}(\mathbf{A}\mathbf{B}^\top) \leq \sum_{i=1}^{\min(m,n)} \sigma_i(\mathbf{A})\sigma_i(\mathbf{B}).$$

See, for example, Horn and Johnson (2012) Theorem 7.4.1.1.

Lemma 5. *Let $\{Z_i\}_{i=1}^n$ be iid m -dimensional random vectors with common distribution $\mathcal{N}_m(\mathbf{0}_m, \mathbf{I}_m)$. Then for any n -dimensional vector $\omega = (\omega_1, \dots, \omega_n)^\top$, we have*

$$\left\| \sum_{i=1}^n \omega_i (Z_i Z_i^\top - \mathbf{I}_m) \right\| = O_P(|\omega|_2 \sqrt{m} + |\omega|_\infty m),$$

where $|\omega|_2 = \sqrt{\sum_{i=1}^n \omega_i^2}$ and $|\omega|_\infty = \max_{1 \leq i \leq n} |\omega_i|$.

Remark 1. Our proof implies that the conclusion is still valid if ω is random and is independent of $\{Z_i\}_{i=1}^n$.

Proof. Our proof is adapted from the proof of Theorem 5.39 in Vershynin (2010). By Lemma 5.2 and Lemma 5.4 of Vershynin (2010), there exists a

set $\mathcal{C} \subset \{x \in \mathbb{R}^m : |x|_2 = 1\}$ satisfying $\text{Card}(\mathcal{C}) \leq 9^m$ such that for any $m \times m$ symmetric matrix \mathbf{A} ,

$$\|A\| \leq 2 \max_{x \in \mathcal{C}} |x^\top \mathbf{A} x|. \quad (\text{S1.3})$$

Then for $t > 4$,

$$\begin{aligned} & \Pr \left(\left\| \sum_{i=1}^n \omega_i (Z_i Z_i^\top - \mathbf{I}_m) \right\| > t(|\omega|_2 \sqrt{m} + |\omega|_\infty m) \right) \\ & \leq \Pr \left(2 \max_{x \in \mathcal{C}} \left| \sum_{i=1}^n \omega_i (x^\top Z_i Z_i^\top x - 1) \right| > t(|\omega|_2 \sqrt{m} + |\omega|_\infty m) \right) \\ & \leq \sum_{x \in \mathcal{C}} \Pr \left(\left| \sum_{i=1}^n \omega_i (x^\top Z_i Z_i^\top x - 1) \right| > 2|\omega|_2 \sqrt{\frac{mt}{4}} + 2|\omega|_\infty \frac{mt}{4} \right) \\ & \leq 2 \cdot 9^m \exp \left(-\frac{mt}{4} \right) = 2 \exp((2 \log 3 - t/4)m), \end{aligned}$$

where the first inequality follows from (S1.3), the second inequality follows from the union bound and the third inequality follows Lemma 1 of Laurent and Massart (2000). The upper bound $2 \exp((2 \log 3 - t/4)m)$ can be arbitrarily small as long as t is large enough. This completes the proof. \square

S2 Proofs of Propositions 1-4

Proof of Proposition 1. We only need to deal with the matrix $n^{-1} \mathbf{Z}^\top \mathbf{\Lambda} \mathbf{Z}$

since it shares the same non-zero eigenvalues as $\hat{\Sigma}$. Write

$$\begin{aligned} n^{-1} \mathbf{Z}^\top \mathbf{\Lambda} \mathbf{Z} &= n^{-1} \mathbf{Z}_1^\top \mathbf{\Lambda}_1 \mathbf{Z}_1 + n^{-1} \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \\ &= n^{-1} \mathbf{Z}_1^\top \mathbf{\Lambda}_1 \mathbf{Z}_1 + n^{-1} \text{tr}(\mathbf{\Lambda}_2) \mathbf{I}_n + n^{-1} (\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \text{tr}(\mathbf{\Lambda}_2) \mathbf{I}_n). \end{aligned}$$

Then Weyl's inequality implies that for $i = 1, \dots, r$,

$$\begin{aligned} & \left| \lambda_i \left(n^{-1} \mathbf{Z}^\top \mathbf{\Lambda} \mathbf{Z} \right) - \lambda_i \left(n^{-1} \mathbf{Z}_1^\top \mathbf{\Lambda}_1 \mathbf{Z}_1 \right) - n^{-1} \text{tr}(\mathbf{\Lambda}_2) \right| \\ & \leq n^{-1} \left\| \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \text{tr}(\mathbf{\Lambda}_2) \mathbf{I}_n \right\|. \end{aligned} \quad (\text{S2.1})$$

Using Weyl's inequality, we can derive the following lower bound for $\lambda_i(\mathbf{Z}_1^\top \mathbf{\Lambda}_1 \mathbf{Z}_1)$, $i = 1, \dots, r$.

$$\begin{aligned} \lambda_i(\mathbf{Z}_1^\top \mathbf{\Lambda}_1 \mathbf{Z}_1) & \geq \lambda_i(\mathbf{Z}_1^\top \text{diag}(\boldsymbol{\lambda}_i \mathbf{I}_i, \mathbf{O}_{(r-i) \times (r-i)}) \mathbf{Z}_1) \\ & = \lambda_i \left(\boldsymbol{\lambda}_i \mathbf{Z}_1^\top \mathbf{Z}_1 - \boldsymbol{\lambda}_i \mathbf{Z}_1^\top \text{diag}(\mathbf{O}_{i \times i}, \mathbf{I}_{r-i}) \mathbf{Z}_1 \right) \\ & \geq \lambda_r \left(\boldsymbol{\lambda}_i \mathbf{Z}_1^\top \mathbf{Z}_1 \right) + \lambda_{n+i-r} \left(- \boldsymbol{\lambda}_i \mathbf{Z}_1^\top \text{diag}(\mathbf{O}_{i \times i}, \mathbf{I}_{r-i}) \mathbf{Z}_1 \right) \\ & = \boldsymbol{\lambda}_i \lambda_r (\mathbf{Z}_1 \mathbf{Z}_1^\top). \end{aligned}$$

Similarly, we can derive the following upper bound for $\lambda_i(\mathbf{Z}_1^\top \mathbf{\Lambda}_1 \mathbf{Z}_1)$, $i = 1, \dots, r$.

$$\begin{aligned} & \lambda_i(\mathbf{Z}_1^\top \mathbf{\Lambda}_1 \mathbf{Z}_1) \\ & = \lambda_i \left(\mathbf{Z}_1^\top \text{diag}(\boldsymbol{\lambda}_1, \dots, \boldsymbol{\lambda}_{i-1}, \mathbf{O}_{(r-i+1) \times (r-i+1)}) \mathbf{Z}_1 \right. \\ & \quad \left. + \mathbf{Z}_1^\top \text{diag}(\mathbf{O}_{(i-1) \times (i-1)}, \boldsymbol{\lambda}_i, \dots, \boldsymbol{\lambda}_r) \mathbf{Z}_1 \right) \\ & \leq \lambda_i \left(\mathbf{Z}_1^\top \text{diag}(\boldsymbol{\lambda}_1, \dots, \boldsymbol{\lambda}_{i-1}, \mathbf{O}_{(r-i+1) \times (r-i+1)}) \mathbf{Z}_1 \right) \\ & \quad + \lambda_1 \left(\mathbf{Z}_1^\top \text{diag}(\mathbf{O}_{(i-1) \times (i-1)}, \boldsymbol{\lambda}_i, \dots, \boldsymbol{\lambda}_r) \mathbf{Z}_1 \right) \\ & \leq \lambda_1(\mathbf{Z}_1^\top \text{diag}(\mathbf{O}_{(i-1) \times (i-1)}, \boldsymbol{\lambda}_i \mathbf{I}_{r-i+1}) \mathbf{Z}_1) \\ & \leq \boldsymbol{\lambda}_i \lambda_1 (\mathbf{Z}_1 \mathbf{Z}_1^\top). \end{aligned}$$

The above lower bound and upper bound imply

$$\begin{aligned}
& \left| \lambda_i(n^{-1} \mathbf{Z}_1^\top \mathbf{\Lambda}_1 \mathbf{Z}_1) - \lambda_i \right| \\
& \leq \lambda_i \max(|\lambda_1(n^{-1} \mathbf{Z}_1 \mathbf{Z}_1^\top) - 1|, |\lambda_r(n^{-1} \mathbf{Z}_1 \mathbf{Z}_1^\top) - 1|) \\
& = \lambda_i \|n^{-1} \mathbf{Z}_1 \mathbf{Z}_1^\top - \mathbf{I}_r\|.
\end{aligned} \tag{S2.2}$$

Combining the bounds (S2.1) and (S2.2) gives that for $i = 1, \dots, r$,

$$\begin{aligned}
& \left| \lambda_i(n^{-1} \mathbf{Z}^\top \mathbf{\Lambda} \mathbf{Z}) - \lambda_i - n^{-1} \text{tr}(\mathbf{\Lambda}_2) \right| \\
& \leq n^{-1} \|\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \text{tr}(\mathbf{\Lambda}_2) \mathbf{I}_n\| + \lambda_i \|n^{-1} \mathbf{Z}_1 \mathbf{Z}_1^\top - \mathbf{I}_r\|.
\end{aligned}$$

From Lemma 5, we have

$$\|n^{-1} \mathbf{Z}_1 \mathbf{Z}_1^\top - \mathbf{I}_r\| = O_P\left(\sqrt{\frac{r}{n}}\right), \tag{S2.3}$$

$$n^{-1} \|\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \text{tr}(\mathbf{\Lambda}_2) \mathbf{I}_n\| = O_P\left(\sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2^2)}{n}} + \lambda_{r+1}\right). \tag{S2.4}$$

This proves the first statement.

Next we prove the second statement. Note that

$$\begin{aligned}
\sum_{i=r+1}^n \lambda_i(\hat{\Sigma}) &= \sum_{i=r+1}^n \lambda_i(n^{-1} \mathbf{Z}^\top \mathbf{\Lambda} \mathbf{Z}) \\
&= \text{tr}(n^{-1} \mathbf{Z}^\top \mathbf{\Lambda} \mathbf{Z}) - \sum_{i=1}^r \lambda_i(n^{-1} \mathbf{Z}^\top \mathbf{\Lambda} \mathbf{Z}) \\
&= \text{tr}(n^{-1} \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2) - \frac{r}{n} \text{tr}(\mathbf{\Lambda}_2) \\
&\quad - \left(\sum_{i=1}^r \lambda_i(n^{-1} \mathbf{Z}^\top \mathbf{\Lambda} \mathbf{Z}) - \text{tr}(n^{-1} \mathbf{Z}_1^\top \mathbf{\Lambda}_1 \mathbf{Z}_1) - \frac{r}{n} \text{tr}(\mathbf{\Lambda}_2) \right).
\end{aligned}$$

It follows from inequalities (S2.1) and (S2.4) that

$$\begin{aligned} & \left| \sum_{i=1}^r \lambda_i(n^{-1} \mathbf{Z}^\top \mathbf{\Lambda} \mathbf{Z}) - \text{tr}(n^{-1} \mathbf{Z}_1^\top \mathbf{\Lambda}_1 \mathbf{Z}_1) - \frac{r}{n} \text{tr}(\mathbf{\Lambda}_2) \right| \\ & \leq \frac{r}{n} \|\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \text{tr}(\mathbf{\Lambda}_2) \mathbf{I}_n\| = O_P \left(r \sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2^2)}{n}} + r \boldsymbol{\lambda}_{r+1} \right). \end{aligned}$$

Thus,

$$\sum_{i=r+1}^n \lambda_i(\hat{\boldsymbol{\Sigma}}) = \text{tr}(n^{-1} \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2) - \frac{r}{n} \text{tr}(\mathbf{\Lambda}_2) + O_P \left(r \sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2^2)}{n}} + r \boldsymbol{\lambda}_{r+1} \right).$$

It is straightforward to show that

$$\mathbb{E} \text{tr}(n^{-1} \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2) = \text{tr}(\mathbf{\Lambda}_2), \quad \text{Var}(\text{tr}(n^{-1} \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2)) = \frac{2}{n} \text{tr}(\mathbf{\Lambda}_2^2).$$

Hence

$$\begin{aligned} & \sum_{i=r+1}^n \lambda_i(\hat{\boldsymbol{\Sigma}}) \\ & = \text{tr}(\mathbf{\Lambda}_2) + O_P \left(\sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2^2)}{n}} \right) - \frac{r}{n} \text{tr}(\mathbf{\Lambda}_2) + O_P \left(r \sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2^2)}{n}} + r \boldsymbol{\lambda}_{r+1} \right) \\ & = \text{tr}(\mathbf{\Lambda}_2) - \frac{r}{n} \text{tr}(\mathbf{\Lambda}_2) + O_P \left(r \sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2^2)}{n}} + r \boldsymbol{\lambda}_{r+1} \right). \end{aligned}$$

This completes the proof of the second statement. \square

Proof of Proposition 2. The first two statements are direct consequences

of Proposition 1 and the condition $r = o(n)$. Next we prove the third state-

ment. We have $\widehat{\text{tr}(\mathbf{\Lambda}_2^2)} = n^{-2} \sum_{i=r+1}^n \lambda_i^2(\mathbf{Y}^\top \mathbf{Y} - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n)$. Note that Weyl's

inequality implies that for $i = r+1, \dots, n$,

$$\lambda_i(\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n) \leq \lambda_i(\mathbf{Y}^\top \mathbf{Y} - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n) \leq \lambda_{i-r}(\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n).$$

Define

$$\begin{aligned}\mathcal{C}_1 &= \left\{ i : 1 \leq i \leq n, \lambda_i \left(\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n \right) > 0 \right\}, \\ \mathcal{C}_2 &= \left\{ i : r+1 \leq i \leq n, \lambda_{i-r} \left(\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n \right) \leq 0 \right\}.\end{aligned}$$

It can be seen that $\mathcal{C}_1 \cap \mathcal{C}_2 = \emptyset$ and $\text{Card}(\mathcal{C}_1 \cup \mathcal{C}_2) \geq n - r$. For $i \geq r+1$ and $i \in \mathcal{C}_1$,

$$\lambda_i^2(\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n) \leq \lambda_i^2(\mathbf{Y}^\top \mathbf{Y} - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n) \leq \lambda_{i-r}^2(\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n);$$

for $i \in \mathcal{C}_2$,

$$\lambda_{i-r}^2(\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n) \leq \lambda_i^2(\mathbf{Y}^\top \mathbf{Y} - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n) \leq \lambda_i^2(\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n);$$

for $i \geq r+1$ and $i \notin \mathcal{C}_1 \cup \mathcal{C}_2$,

$$\begin{aligned}& \lambda_i^2(\mathbf{Y}^\top \mathbf{Y} - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n) \\ & \leq \max \left(\lambda_{i-r}^2(\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n), \lambda_i^2(\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n) \right).\end{aligned}$$

Therefore,

$$\begin{aligned}
& \left| \sum_{i=r+1}^n \lambda_i^2 \left(\mathbf{Y}^\top \mathbf{Y} - \widehat{\text{tr}(\Lambda_2)} \mathbf{I}_n \right) - \text{tr}(\mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 - \widehat{\text{tr}(\Lambda_2)} \mathbf{I}_n)^2 \right| \\
& \leq \left| \sum_{i>r, i \in \mathcal{C}_1} \lambda_i^2 \left(\mathbf{Y}^\top \mathbf{Y} - \widehat{\text{tr}(\Lambda_2)} \mathbf{I}_n \right) - \sum_{i \in \mathcal{C}_1} \lambda_i^2 \left(\mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 - \widehat{\text{tr}(\Lambda_2)} \mathbf{I}_n \right) \right| \\
& \quad + \left| \sum_{i>r, i \in \mathcal{C}_2} \lambda_i^2 \left(\mathbf{Y}^\top \mathbf{Y} - \widehat{\text{tr}(\Lambda_2)} \mathbf{I}_n \right) - \sum_{i \notin \mathcal{C}_1} \lambda_i^2 \left(\mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 - \widehat{\text{tr}(\Lambda_2)} \mathbf{I}_n \right) \right| \\
& \quad + \left| \sum_{i>r, i \notin \mathcal{C}_1 \cup \mathcal{C}_2} \lambda_i^2 \left(\mathbf{Y}^\top \mathbf{Y} - \widehat{\text{tr}(\Lambda_2)} \mathbf{I}_n \right) \right| \\
& \leq 3r \|\mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 - \widehat{\text{tr}(\Lambda_2)} \mathbf{I}_n\|^2 \\
& \leq 3r \left(\|\mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 - \text{tr}(\Lambda_2) \mathbf{I}_n\| + \left| \text{tr}(\Lambda_2) - \widehat{\text{tr}(\Lambda_2)} \right| \right)^2 \\
& = O_P \left(rn \text{tr}(\Lambda_2^2) + rn^2 \lambda_{r+1}^2 \right).
\end{aligned} \tag{S2.5}$$

where the last equality follows from (S2.4) and the second statement of the proposition.

Now we deal with $\text{tr}(\mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 - \widehat{\text{tr}(\Lambda_2)} \mathbf{I}_n)^2$. Let $Z_{2,i}$ be the i th column of \mathbf{Z}_2 , $i = 1, \dots, n$. Then

$$\text{tr}(\mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 - \widehat{\text{tr}(\Lambda_2)} \mathbf{I}_n)^2 = \sum_{i=1}^n (Z_{2,i}^\top \Lambda_2 Z_{2,i} - \widehat{\text{tr}(\Lambda_2)})^2 + 2 \sum_{1 \leq i < j \leq n} (Z_{2,i}^\top \Lambda_2 Z_{2,j})^2.$$

For the first term, we have

$$\sum_{i=1}^n (Z_{2,i}^\top \Lambda_2 Z_{2,i} - \widehat{\text{tr}(\Lambda_2)})^2 \leq 2 \sum_{i=1}^n (Z_{2,i}^\top \Lambda_2 Z_{2,i} - \text{tr}(\Lambda_2))^2 + 2n (\widehat{\text{tr}(\Lambda_2)} - \text{tr}(\Lambda_2))^2.$$

Then it follows from the second statement of the proposition and the fact

$E \sum_{i=1}^n (Z_{2,i}^\top \mathbf{\Lambda}_2 Z_{2,i} - \text{tr}(\mathbf{\Lambda}_2))^2 = 2n \text{tr}(\mathbf{\Lambda}_2^2)$ that

$$\sum_{i=1}^n (Z_{2,i}^\top \mathbf{\Lambda}_2 Z_{2,i} - \widehat{\text{tr}(\mathbf{\Lambda}_2)})^2 = O_P((n + r^2) \text{tr}(\mathbf{\Lambda}_2^2) + r^2 n \boldsymbol{\lambda}_{r+1}^2). \quad (\text{S2.6})$$

For the second term, it is straightforward to show that $E 2 \sum_{1 \leq i < j \leq n} (Z_{2,i}^\top \mathbf{\Lambda}_2 Z_{2,j})^2 = n(n-1) \text{tr}(\mathbf{\Lambda}_2^2)$. Furthermore, Chen et al. (2010), Proposition A.2 implies that

$$\begin{aligned} \text{Var} \left(2 \sum_{1 \leq i < j \leq n} (Z_{2,i}^\top \mathbf{\Lambda}_2 Z_{2,j})^2 \right) &= O(n^2 \text{tr}^2(\mathbf{\Lambda}_2^2) + n^3 \text{tr}(\mathbf{\Lambda}_2^4)) \\ &= O(n^2 \text{tr}^2(\mathbf{\Lambda}_2^2) + n \text{tr}(\mathbf{\Lambda}_2^2) n^2 \boldsymbol{\lambda}_{r+1}^2) \\ &= O(n^2 \text{tr}^2(\mathbf{\Lambda}_2^2) + n^4 \boldsymbol{\lambda}_{r+1}^4). \end{aligned}$$

Thus,

$$2 \sum_{1 \leq i < j \leq n} (Z_{2,i}^\top \mathbf{\Lambda}_2 Z_{2,j})^2 = n^2 \text{tr}(\mathbf{\Lambda}_2^2) + O_P(n \text{tr}(\mathbf{\Lambda}_2^2) + n^2 \boldsymbol{\lambda}_{r+1}^2).$$

Combining the last display and (S2.6) yields

$$\text{tr}(\mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n)^2 = n^2 \text{tr}(\mathbf{\Lambda}_2^2) + O_P((n + r^2) \text{tr}(\mathbf{\Lambda}_2^2) + (n + r^2) n \boldsymbol{\lambda}_{r+1}^2).$$

Combine the last display and (S2.5), we have

$$\sum_{i=r+1}^n \lambda_i^2 \left(\mathbf{Y}^\top \mathbf{Y} - \widehat{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_n \right) = O_P(rn \text{tr}(\mathbf{\Lambda}_2^2) + rn^2 \boldsymbol{\lambda}_{r+1}^2).$$

This completes the proof. □

Proposition 6. *Suppose that $r = o(n)$ and $r\lambda_{r+1}/\text{tr}(\Lambda_2) \rightarrow 0$. Then*

$$\|\mathbf{P}_{\mathbf{Y},1} - \mathbf{P}_{\mathbf{Y},1}^*\| = O_P\left(\frac{\lambda_{r+1} + n^{-1}\text{tr}(\Lambda_2)}{\lambda_r + n^{-1}\text{tr}(\Lambda_2)}\right),$$

where

$$\mathbf{P}_{\mathbf{Y},1}^* = \mathbf{U} \begin{pmatrix} \mathbf{I}_r \\ \mathbf{Q} \end{pmatrix} (\mathbf{I}_r + \mathbf{Q}^\top \mathbf{Q})^{-1} \begin{pmatrix} \mathbf{I}_r & \mathbf{Q}^\top \end{pmatrix} \mathbf{U}^\top.$$

Proof. The following intermediate matrix

$$\begin{aligned} \hat{\Sigma}_0 = & n^{-1} \mathbf{U}_1 \Lambda_1^{1/2} \mathbf{Z}_1 \mathbf{Z}_1^\top \Lambda_1^{1/2} \mathbf{U}_1^\top + n^{-1} \mathbf{U}_1 \Lambda_1^{1/2} \mathbf{Z}_1 \mathbf{Z}_2^\top \Lambda_2^{1/2} \mathbf{U}_2^\top \\ & + n^{-1} \mathbf{U}_2 \Lambda_2^{1/2} \mathbf{Z}_2 \mathbf{Z}_1^\top \Lambda_1^{1/2} \mathbf{U}_1^\top + n^{-1} \mathbf{U}_2 \Lambda_2^{1/2} \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2^{1/2} \mathbf{U}_2^\top \end{aligned}$$

plays a key role in the proof. It can be seen that

$$\hat{\Sigma}_0 = n^{-1} \mathbf{U} \begin{pmatrix} \mathbf{I}_r \\ \mathbf{Q} \end{pmatrix} \Lambda_1^{1/2} \mathbf{Z}_1 \mathbf{Z}_1^\top \Lambda_1^{1/2} \begin{pmatrix} \mathbf{I}_r & \mathbf{Q}^\top \end{pmatrix} \mathbf{U}^\top.$$

Consequently, $\hat{\Sigma}_0$ is a positive semi-definite matrix with rank r , and $\mathbf{P}_{\mathbf{Y},1}^*$ is the projection matrix onto the rank r principal subspace of $\hat{\Sigma}_0$.

From Cai et al. (2015), Proposition 1, we have

$$\|\mathbf{P}_{\mathbf{Y},1} - \mathbf{P}_{\mathbf{Y},1}^*\| \leq \frac{2\|\hat{\Sigma} - \hat{\Sigma}_0\|}{\lambda_r(\hat{\Sigma}_0)}. \quad (\text{S2.7})$$

We have the following upper bound for $\|\hat{\Sigma} - \hat{\Sigma}_0\|$.

$$\begin{aligned}
& \|\hat{\Sigma} - \hat{\Sigma}_0\| \\
&= n^{-1} \left\| \mathbf{U}_2 \Lambda_2^{1/2} \mathbf{Z}_2 \mathbf{Z}_2^\top \Lambda_2^{1/2} \mathbf{U}_2^\top - \mathbf{U}_2 \Lambda_2^{1/2} \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2^{1/2} \mathbf{U}_2^\top \right\| \\
&= n^{-1} \left\| \Lambda_2^{1/2} \mathbf{Z}_2 (\mathbf{I}_n - \mathbf{V}_{\mathbf{Z}_1} \mathbf{V}_{\mathbf{Z}_1}^\top) \mathbf{Z}_2^\top \Lambda_2^{1/2} \right\| \\
&\leq n^{-1} \left\| \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \right\| \tag{S2.8} \\
&\leq n^{-1} \left\| \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 - \text{tr}(\Lambda_2) \mathbf{I}_n \right\| + n^{-1} \text{tr}(\Lambda_2) \\
&= O_P \left(\sqrt{\frac{\text{tr}(\Lambda_2^2)}{n}} + \lambda_{r+1} + n^{-1} \text{tr}(\Lambda_2) \right) \\
&= O_P \left(\lambda_{r+1} + n^{-1} \text{tr}(\Lambda_2) \right),
\end{aligned}$$

where the second last equality follows from (S2.4) and the last equality follows from

$$\sqrt{\frac{\text{tr}(\Lambda_2^2)}{n}} \leq \sqrt{\frac{\lambda_{r+1} \text{tr}(\Lambda_2)}{n}} \leq \frac{1}{2} (\lambda_{r+1} + n^{-1} \text{tr}(\Lambda_2)).$$

Now we deal with $\lambda_r(\hat{\Sigma}_0)$. We have

$$\begin{aligned}
\lambda_r(\hat{\Sigma}_0) &= \lambda_r \left(n^{-1} (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{1/2} \Lambda_1^{1/2} (\mathbf{I}_r + \mathbf{Q}^\top \mathbf{Q}) \Lambda_1^{1/2} (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{1/2} \right) \\
&= \lambda_r \left(n^{-1} (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{1/2} \Lambda_1 (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{1/2} + n^{-1} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \right).
\end{aligned}$$

It can be seen that $\mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1}$ is a $(p-r) \times r$ random matrix with iid $\mathcal{N}(0, 1)$

entries. Then Lemma 5 implies that

$$\begin{aligned}
& \left\| n^{-1} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \boldsymbol{\Lambda}_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} - n^{-1} \text{tr}(\boldsymbol{\Lambda}_2) \mathbf{I}_r \right\| \\
&= O_P \left(n^{-1} \sqrt{r \text{tr}(\boldsymbol{\Lambda}_2^2)} + r n^{-1} \boldsymbol{\lambda}_{r+1} \right) \\
&= O_P \left(n^{-1} \sqrt{r \boldsymbol{\lambda}_{r+1} \text{tr}(\boldsymbol{\Lambda}_2)} + r n^{-1} \boldsymbol{\lambda}_{r+1} \right) \\
&= o_P \left(n^{-1} \text{tr}(\boldsymbol{\Lambda}_2) \right),
\end{aligned} \tag{S2.9}$$

where the last equality follows from the condition $r \boldsymbol{\lambda}_{r+1} / \text{tr}(\boldsymbol{\Lambda}_2) \rightarrow 0$. Then

it follows from Weyl's inequality that

$$\begin{aligned}
& \left| \lambda_r(\hat{\boldsymbol{\Sigma}}_0) - \lambda_r \left(n^{-1} (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{1/2} \boldsymbol{\Lambda}_1 (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{1/2} + n^{-1} \text{tr}(\boldsymbol{\Lambda}_2) \mathbf{I}_r \right) \right| \\
&\leq \left\| n^{-1} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \boldsymbol{\Lambda}_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} - n^{-1} \text{tr}(\boldsymbol{\Lambda}_2) \mathbf{I}_r \right\| \\
&= o_P \left(n^{-1} \text{tr}(\boldsymbol{\Lambda}_2) \right).
\end{aligned}$$

On the other hand, (S2.2) and (S2.3) imply that

$$\begin{aligned}
& \lambda_r \left(n^{-1} (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{1/2} \boldsymbol{\Lambda}_1 (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{1/2} + n^{-1} \text{tr}(\boldsymbol{\Lambda}_2) \mathbf{I}_r \right) \\
&= \lambda_r \left(n^{-1} \mathbf{Z}_1^\top \boldsymbol{\Lambda}_1 \mathbf{Z}_1 \right) + n^{-1} \text{tr}(\boldsymbol{\Lambda}_2) \\
&= \boldsymbol{\lambda}_r + o_P(\boldsymbol{\lambda}_r) + n^{-1} \text{tr}(\boldsymbol{\Lambda}_2).
\end{aligned}$$

Hence we have

$$\lambda_r(\hat{\boldsymbol{\Sigma}}_0) = (1 + o_P(1))(\boldsymbol{\lambda}_r + n^{-1} \text{tr}(\boldsymbol{\Lambda}_2)). \tag{S2.10}$$

Then the conclusion follows from (S2.7), (S2.8) and (S2.10). \square

Proof of Proposition 3. Note that

$$\left\| \mathbf{P}_{\mathbf{Y},1} - \mathbf{P}_{\mathbf{Y},1}^\dagger \right\| \leq \left\| \mathbf{P}_{\mathbf{Y},1} - \mathbf{P}_{\mathbf{Y},1}^* \right\| + \left\| \mathbf{P}_{\mathbf{Y},1}^* - \mathbf{P}_{\mathbf{Y},1}^\dagger \right\|.$$

Under the condition $\text{tr}(\mathbf{\Lambda}_2)/(n\lambda_r) \rightarrow 0$, Proposition 6 implies that

$$\|\mathbf{P}_{\mathbf{Y},1} - \mathbf{P}_{\mathbf{Y},1}^*\| = O_P\left(\frac{\lambda_{r+1}}{\lambda_r} + \frac{\text{tr}(\mathbf{\Lambda}_2)}{n\lambda_r}\right).$$

So we only need to deal with $\|\mathbf{P}_{\mathbf{Y},1}^* - \mathbf{P}_{\mathbf{Y},1}^\dagger\|$. We have

$$\begin{aligned} & \|\mathbf{P}_{\mathbf{Y},1}^* - \mathbf{P}_{\mathbf{Y},1}^\dagger\| \\ & \leq \left\| \mathbf{P}_{\mathbf{Y},1}^* - \mathbf{U} \begin{pmatrix} \mathbf{I}_r \\ \mathbf{Q} \end{pmatrix} \begin{pmatrix} \mathbf{I}_r & \mathbf{Q}^\top \end{pmatrix} \mathbf{U}^\top \right\| + \left\| \mathbf{U} \begin{pmatrix} \mathbf{I}_r \\ \mathbf{Q} \end{pmatrix} \begin{pmatrix} \mathbf{I}_r & \mathbf{Q}^\top \end{pmatrix} \mathbf{U}^\top - \mathbf{P}_{\mathbf{Y},1}^\dagger \right\| \\ & = \left\| \begin{pmatrix} \mathbf{I}_r \\ \mathbf{Q} \end{pmatrix} \left((\mathbf{I}_r + \mathbf{Q}^\top \mathbf{Q})^{-1} - \mathbf{I}_r \right) \begin{pmatrix} \mathbf{I}_r & \mathbf{Q}^\top \end{pmatrix} \right\| + \|\mathbf{U}_2 \mathbf{Q} \mathbf{Q}^\top \mathbf{U}_2^\top\| \\ & = \left\| \left((\mathbf{I}_r + \mathbf{Q}^\top \mathbf{Q})^{-1} - \mathbf{I}_r \right) (\mathbf{I}_r + \mathbf{Q}^\top \mathbf{Q}) \right\| + \|\mathbf{U}_2 \mathbf{Q} \mathbf{Q}^\top \mathbf{U}_2^\top\| \\ & = 2 \|\mathbf{Q}^\top \mathbf{Q}\|. \end{aligned}$$

Note that

$$\begin{aligned} \|\mathbf{Q}^\top \mathbf{Q}\| &= \left\| \mathbf{\Lambda}_1^{-1/2} (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{-1/2} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{-1/2} \mathbf{\Lambda}_1^{-1/2} \right\| \\ &\leq \lambda_r^{-1} \|(\mathbf{Z}_1 \mathbf{Z}_1^\top)^{-1}\| \|\mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1}\| \\ &= O_P\left(\frac{\text{tr}(\mathbf{\Lambda}_2)}{n\lambda_r}\right), \end{aligned} \tag{S2.11}$$

where the second last equality follows from the fact $\|(\mathbf{Z}_1 \mathbf{Z}_1^\top)^{-1}\| = \lambda_r (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{-1}$,

(S2.3), (S2.9) and Weyl's inequality. Therefore, we have

$$\|\mathbf{P}_{\mathbf{Y},1}^* - \mathbf{P}_{\mathbf{Y},1}^\dagger\| = O_P\left(\frac{\text{tr}(\mathbf{\Lambda}_2)}{n\lambda_r}\right).$$

This completes the proof.

□

Proposition 7. *Suppose that $r = o(n)$ and $n\lambda_{r+1}/\text{tr}(\Lambda_2) \rightarrow 0$. Then*

$$\|\mathbf{P}_{\mathbf{Y},2} - \mathbf{P}_{\mathbf{Y},2}^*\| = O_P \left(\min \left(\sqrt{\frac{\text{tr}(\Lambda_2)\lambda_1}{n\lambda_r^2}}, 1 \right) \right).$$

where $\mathbf{P}_{\mathbf{Y},2}^* = \mathbf{U}_2 \Lambda_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \left(\tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \right)^{-1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2^{1/2} \mathbf{U}_2^\top$.

Proof. We only need to prove that for any subsequence of $\{n\}$, there is a further subsequence along which the conclusion holds. Thus, without loss of generality, we assume $\text{tr}(\Lambda_2)\lambda_1/(n\lambda_r^2) \rightarrow c \in [0, +\infty]$. Since $\mathbf{P}_{\mathbf{Y},2}$ and $\mathbf{P}_{\mathbf{Y},2}^*$ are both projection matrices, we have $\|\mathbf{P}_{\mathbf{Y},2} - \mathbf{P}_{\mathbf{Y},2}^*\| \leq 2$. Therefore, the conclusion holds if $c > 0$. In the rest of the proof, we assume $c = 0$, that is $\text{tr}(\Lambda_2)\lambda_1/(n\lambda_r^2) \rightarrow 0$.

Note that $\mathbf{U}_{\mathbf{Y},2}$ is in fact the leading $n-r$ eigenvectors of $(\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1})\hat{\Sigma}(\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1})$. Under the condition $n\lambda_{r+1}/\text{tr}(\Lambda_2) \rightarrow 0$, Proposition 3 implies that

$$\|\mathbf{P}_{\mathbf{Y},1} - \mathbf{P}_{\mathbf{Y},1}^\dagger\| = O_P \left(\frac{\text{tr}(\Lambda_2)}{n\lambda_r} \right).$$

It can be seen that

$$\begin{aligned} & \left\| (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1})\hat{\Sigma}(\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}) - (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger)\hat{\Sigma}(\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \right\| \\ & \leq \left\| (\mathbf{P}_{\mathbf{Y},1}^\dagger - \mathbf{P}_{\mathbf{Y},1})\hat{\Sigma}(\mathbf{P}_{\mathbf{Y},1}^\dagger - \mathbf{P}_{\mathbf{Y},1}) \right\| + 2 \left\| (\mathbf{P}_{\mathbf{Y},1}^\dagger - \mathbf{P}_{\mathbf{Y},1})\hat{\Sigma}(\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \right\|. \end{aligned}$$

Under the condition $n\lambda_{r+1}/\text{tr}(\mathbf{\Lambda}_2) \rightarrow 0$, Proposition 1 implies that

$$\begin{aligned}\|\hat{\Sigma}\| &= \lambda_1 \left(1 + \frac{\text{tr}(\mathbf{\Lambda}_2)}{n\lambda_1} + O_P \left(\sqrt{\frac{r}{n}} + \sqrt{\frac{\lambda_{r+1}}{\lambda_1} \frac{\text{tr}(\mathbf{\Lambda}_2)}{n\lambda_1}} + \frac{\lambda_{r+1}}{\lambda_1} \right) \right) \\ &= \lambda_1(1 + o_P(1)).\end{aligned}$$

Then

$$\begin{aligned}\left\| (\mathbf{P}_{\mathbf{Y},1}^\dagger - \mathbf{P}_{\mathbf{Y},1}) \hat{\Sigma} (\mathbf{P}_{\mathbf{Y},1}^\dagger - \mathbf{P}_{\mathbf{Y},1}) \right\| &\leq \|\hat{\Sigma}\| \left\| \mathbf{P}_{\mathbf{Y},1}^\dagger - \mathbf{P}_{\mathbf{Y},1} \right\|^2 \\ &= O_P \left(\frac{\text{tr}^2(\mathbf{\Lambda}_2) \lambda_1}{n^2 \lambda_r^2} \right).\end{aligned}\tag{S2.12}$$

On the other hand, we have

$$\begin{aligned}& \left\| (\mathbf{P}_{\mathbf{Y},1}^\dagger - \mathbf{P}_{\mathbf{Y},1}) \hat{\Sigma} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \right\| \\ & \leq \left\| \mathbf{P}_{\mathbf{Y},1}^\dagger - \mathbf{P}_{\mathbf{Y},1} \right\| \left\| n^{-1} \mathbf{U} \mathbf{\Lambda}^{1/2} \mathbf{Z} \right\| \left\| \mathbf{Z}^\top \mathbf{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \right\| \\ & = n^{-1/2} \left\| \mathbf{P}_{\mathbf{Y},1}^\dagger - \mathbf{P}_{\mathbf{Y},1} \right\| \left\| \hat{\Sigma} \right\|^{1/2} \left\| \mathbf{Z}^\top \mathbf{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \right\| \\ & = O_P \left(\frac{\text{tr}(\mathbf{\Lambda}_2) \lambda_1^{1/2}}{n^{3/2} \lambda_r} \right) \left\| \mathbf{Z}^\top \mathbf{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \right\|.\end{aligned}$$

It is straightforward to show that

$$\begin{aligned}& \mathbf{Z}^\top \mathbf{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \\ & = \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^{1/2} \mathbf{U}_2^\top - \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \mathbf{Z}_1^\top (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{-1} \mathbf{\Lambda}_1^{-1/2} \mathbf{U}_1^\top.\end{aligned}\tag{S2.13}$$

Then

$$\begin{aligned}& \left\| \mathbf{Z}^\top \mathbf{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \right\| \\ & \leq \left\| \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \right\|^{1/2} + \lambda_r^{-1/2} \left\| \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \right\| \left\| (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{-1} \right\|^{1/2}.\end{aligned}$$

It follows from (S2.4) and the condition $n\lambda_{r+1}/\text{tr}(\mathbf{\Lambda}_2) \rightarrow 0$ that

$$\left\| \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \right\| = (1 + o_P(1)) \text{tr}(\mathbf{\Lambda}_2).\tag{S2.14}$$

Consequently,

$$\begin{aligned} \left\| \mathbf{Z}^\top \boldsymbol{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \right\| &= O_P(\text{tr}^{1/2}(\boldsymbol{\Lambda}_2)) + O_P\left(\frac{\text{tr}(\boldsymbol{\Lambda}_2)}{\sqrt{n\boldsymbol{\lambda}_r}}\right) \\ &= O_P(\text{tr}^{1/2}(\boldsymbol{\Lambda}_2)). \end{aligned}$$

Thus,

$$\left\| (\mathbf{P}_{\mathbf{Y},1}^\dagger - \mathbf{P}_{\mathbf{Y},1}) \hat{\boldsymbol{\Sigma}} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \right\| = O_P\left(\frac{\text{tr}^{3/2}(\boldsymbol{\Lambda}_2) \boldsymbol{\lambda}_1^{1/2}}{n^{3/2} \boldsymbol{\lambda}_r}\right). \quad (\text{S2.15})$$

Combine (S2.12) and (S2.15), we obtain

$$\begin{aligned} &\left\| (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}) \hat{\boldsymbol{\Sigma}} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}) - (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \hat{\boldsymbol{\Sigma}} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \right\| \\ &= O_P\left(\frac{\text{tr}^2(\boldsymbol{\Lambda}_2) \boldsymbol{\lambda}_1}{n^2 \boldsymbol{\lambda}_r^2} + \frac{\text{tr}^{3/2}(\boldsymbol{\Lambda}_2) \boldsymbol{\lambda}_1^{1/2}}{n^{3/2} \boldsymbol{\lambda}_r}\right). \end{aligned}$$

Now we deal with $(\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \hat{\boldsymbol{\Sigma}} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger)$. In view of (S2.13), we have

$$\begin{aligned} &(\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \hat{\boldsymbol{\Sigma}} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \\ &= n^{-1} \mathbf{U}_2 \boldsymbol{\Lambda}_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \boldsymbol{\Lambda}_2^{1/2} \mathbf{U}_2^\top - n^{-1} \mathbf{U}_2 \boldsymbol{\Lambda}_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \boldsymbol{\Lambda}_2^{1/2} \mathbf{Q} \mathbf{U}_1^\top \\ &\quad - n^{-1} \mathbf{U}_1 \mathbf{Q}^\top \boldsymbol{\Lambda}_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \boldsymbol{\Lambda}_2^{1/2} \mathbf{U}_2^\top + n^{-1} \mathbf{U}_1 \mathbf{Q}^\top \boldsymbol{\Lambda}_2^{1/2} \mathbf{Z}_2 \mathbf{Z}_2^\top \boldsymbol{\Lambda}_2^{1/2} \mathbf{Q} \mathbf{U}_1^\top. \end{aligned}$$

Then

$$\begin{aligned} &\left\| (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \hat{\boldsymbol{\Sigma}} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) - n^{-1} \mathbf{U}_2 \boldsymbol{\Lambda}_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \boldsymbol{\Lambda}_2^{1/2} \mathbf{U}_2^\top \right\| \\ &\leq n^{-1} \left\| \boldsymbol{\Lambda}_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \boldsymbol{\Lambda}_2^{1/2} \mathbf{Q} \right\| + n^{-1} \left\| \mathbf{Q}^\top \boldsymbol{\Lambda}_2^{1/2} \mathbf{Z}_2 \mathbf{Z}_2^\top \boldsymbol{\Lambda}_2^{1/2} \mathbf{Q} \right\| \\ &\leq n^{-1} \left\| \mathbf{Z}_2^\top \boldsymbol{\Lambda}_2 \mathbf{Z}_2 \right\| \left\| \mathbf{Q}^\top \mathbf{Q} \right\|^{1/2} + n^{-1} \left\| \mathbf{Z}_2^\top \boldsymbol{\Lambda}_2 \mathbf{Z}_2 \right\| \left\| \mathbf{Q}^\top \mathbf{Q} \right\| \\ &= O_P\left(\frac{\text{tr}^{3/2}(\boldsymbol{\Lambda}_2)}{n^{3/2} \boldsymbol{\lambda}_r^{1/2}}\right), \end{aligned}$$

where the last equality follows from (S2.11) and (S2.14).

Combine the above bounds, we obtain

$$\begin{aligned} & \left\| (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}) \hat{\Sigma} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}) - n^{-1} \mathbf{U}_2 \mathbf{\Lambda}_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^{1/2} \mathbf{U}_2^\top \right\| \\ &= O_P \left(\frac{\text{tr}^2(\mathbf{\Lambda}_2) \boldsymbol{\lambda}_1}{n^2 \boldsymbol{\lambda}_r^2} + \frac{\text{tr}^{3/2}(\mathbf{\Lambda}_2) \boldsymbol{\lambda}_1^{1/2}}{n^{3/2} \boldsymbol{\lambda}_r} \right). \end{aligned} \quad (\text{S2.16})$$

The matrix $n^{-1} \mathbf{U}_2 \mathbf{\Lambda}_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^{1/2} \mathbf{U}_2^\top$ shares the same non-zero eigenvalues as $n^{-1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1}$. Note that $\mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1}$ is a $p \times (n-r)$ random matrix with iid $\mathcal{N}(0,1)$ entries. Then it follows from Lemma 5 and the condition $n \boldsymbol{\lambda}_{r+1} / \text{tr}(\mathbf{\Lambda}_2) \rightarrow 0$ that

$$\begin{aligned} & \left\| n^{-1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} - n^{-1} \text{tr}(\mathbf{\Lambda}_2) \mathbf{I}_{n-r} \right\| \\ &= O_P \left(n^{-1/2} \sqrt{\text{tr}(\mathbf{\Lambda}_2^2) + \boldsymbol{\lambda}_{r+1}} \right) \\ &= O_P \left(n^{-1/2} \sqrt{\boldsymbol{\lambda}_{r+1} \text{tr}(\mathbf{\Lambda}_2) + \boldsymbol{\lambda}_{r+1}} \right) \\ &= o_P \left(n^{-1} \text{tr}(\mathbf{\Lambda}_2) \right). \end{aligned} \quad (\text{S2.17})$$

This bound, combined with Weyl's inequality, leads to

$$\lambda_{n-r} \left(n^{-1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \right) = (1 + o_P(1)) n^{-1} \text{tr}(\mathbf{\Lambda}_1). \quad (\text{S2.18})$$

It can be seen that the matrix $\mathbf{P}_{\mathbf{Y},2}^*$ is the projection matrix onto the rank $n-r$ principal subspace of $n^{-1} \mathbf{U}_2 \mathbf{\Lambda}_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^{1/2} \mathbf{U}_2^\top$. Therefore, Cai

et al. (2015), Proposition 1 implies that

$$\begin{aligned}
& \|\mathbf{P}_{\mathbf{Y},2} - \mathbf{P}_{\mathbf{Y},2}^*\| \\
& \leq \frac{2 \left\| (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}) \hat{\Sigma} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}) - n^{-1} \mathbf{U}_2 \Lambda_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2^{1/2} \mathbf{U}_2^\top \right\|}{\lambda_{n-r} \left(n^{-1} \mathbf{U}_2 \Lambda_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2^{1/2} \mathbf{U}_2^\top \right)} \\
& = O_P \left(\frac{\text{tr}(\Lambda_2) \lambda_1}{n \lambda_r^2} + \sqrt{\frac{\text{tr}(\Lambda_2) \lambda_1}{n \lambda_r^2}} \right) \\
& = O_P \left(\sqrt{\frac{\text{tr}(\Lambda_2) \lambda_1}{n \lambda_r^2}} \right),
\end{aligned}$$

where the second last equality follows from (S2.16) and (S2.18). This completes the proof. \square

Proof of Proposition 4. By some algebra, it can be seen that

$$\begin{aligned}
\left\| \mathbf{P}_{\mathbf{Y},2}^* - \mathbf{P}_{\mathbf{Y},2}^\dagger \right\| &= (\text{tr}(\Lambda_2))^{-1} \left\| \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} - \text{tr}(\Lambda_2) \mathbf{I}_{n-r} \right\| \\
&= O_P \left(\frac{\sqrt{n \text{tr}(\Lambda_2^2)}}{\text{tr}(\Lambda_2)} + \frac{n \lambda_{r+1}}{\text{tr}(\Lambda_2)} \right) \\
&= O_P \left(\sqrt{\frac{n \lambda_{r+1}}{\text{tr}(\Lambda_2)}} \right),
\end{aligned}$$

where the second last equality follows from (S2.17) and the last equality follows from the fact $\sqrt{n \text{tr}(\Lambda_2^2)}/\text{tr}(\Lambda_2) \leq \sqrt{n \lambda_{r+1}/\text{tr}(\Lambda_2)}$ and the condition $\sqrt{n \lambda_{r+1}/\text{tr}(\Lambda_2)} \rightarrow 0$. Then the conclusion follows from the last display and Proposition 7. \square

S3 Proofs of Theorems 1 and 2

It can be seen that \mathbf{XJC} is independent of \mathbf{Y} . We write $\mathbf{XJC} = \boldsymbol{\Theta}\mathbf{C} + \mathbf{U}\boldsymbol{\Lambda}^{1/2}\mathbf{Z}^\dagger$, where \mathbf{Z}^\dagger is a $p \times (k-1)$ matrix with iid $\mathcal{N}(0, 1)$ entries and is independent of \mathbf{Z} . Then

$$\begin{aligned} & \mathbf{C}^\top \mathbf{J}^\top \mathbf{X}^\top (\mathbf{I}_p - \mathbf{P}_\mathbf{Y}) \mathbf{XJC} \\ &= \mathbf{Z}^{\dagger\top} \boldsymbol{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_\mathbf{Y}) \mathbf{U} \boldsymbol{\Lambda}^{1/2} \mathbf{Z}^\dagger + \mathbf{C}^\top \boldsymbol{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_\mathbf{Y}) \boldsymbol{\Theta} \mathbf{C} \\ & \quad + \mathbf{C}^\top \boldsymbol{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_\mathbf{Y}) \mathbf{U} \boldsymbol{\Lambda}^{1/2} \mathbf{Z}^\dagger + \mathbf{Z}^{\dagger\top} \boldsymbol{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_\mathbf{Y}) \boldsymbol{\Theta} \mathbf{C}. \end{aligned} \quad (\text{S3.1})$$

It can be seen that the first term of (S3.1) can be written as

$$\mathbf{Z}^{\dagger\top} \boldsymbol{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_\mathbf{Y}) \mathbf{U} \boldsymbol{\Lambda}^{1/2} \mathbf{Z}^\dagger = \sum_{i=1}^p \lambda_i ((\mathbf{I}_p - \mathbf{P}_\mathbf{Y}) \boldsymbol{\Sigma} (\mathbf{I}_p - \mathbf{P}_\mathbf{Y})) \eta_i \eta_i^\top,$$

where η_1, \dots, η_p are independent $\mathcal{N}_{k-1}(\mathbf{0}_{k-1}, \mathbf{I}_{k-1})$ random vectors and are independent of $\mathbf{P}_\mathbf{Y}$.

Lemma 6. *Suppose that $n\lambda_1/\text{tr}(\boldsymbol{\Sigma}) \rightarrow 0$. Then*

$$\begin{aligned} \text{tr}((\mathbf{I}_p - \mathbf{P}_\mathbf{Y}) \boldsymbol{\Sigma} (\mathbf{I}_p - \mathbf{P}_\mathbf{Y})) &= \text{tr}(\boldsymbol{\Sigma}) - \frac{n \text{tr}(\boldsymbol{\Sigma}^2)}{\text{tr}(\boldsymbol{\Sigma})} + O_P \left(n(\lambda_1 - \lambda_p) \sqrt{\frac{n\lambda_1}{\text{tr}(\boldsymbol{\Sigma})}} \right), \\ \text{tr}((\mathbf{I}_p - \mathbf{P}_\mathbf{Y}) \boldsymbol{\Sigma} (\mathbf{I}_p - \mathbf{P}_\mathbf{Y}))^2 &= \text{tr}(\boldsymbol{\Sigma}^2) - \frac{n \text{tr}^2(\boldsymbol{\Sigma}^2)}{\text{tr}^2(\boldsymbol{\Sigma})} + O_P(n\lambda_1(\lambda_1 - \lambda_p)). \end{aligned}$$

Proof. First we approximate $\mathbf{P}_\mathbf{Y}$ by a simple expression. We have

$$\begin{aligned} \|\mathbf{P}_\mathbf{Y} - (\text{tr}(\boldsymbol{\Sigma}))^{-1} \mathbf{Y} \mathbf{Y}^\top\| &= \|\mathbf{Y} (\mathbf{Y}^\top \mathbf{Y})^{-1} \mathbf{Y}^\top - (\text{tr}(\boldsymbol{\Sigma}))^{-1} \mathbf{Y} \mathbf{Y}^\top\| \\ &= (\text{tr}(\boldsymbol{\Sigma}))^{-1} \|\mathbf{Y}^\top \mathbf{Y} - \text{tr}(\boldsymbol{\Sigma}) \mathbf{I}_n\|. \end{aligned}$$

Then from Lemma 5, we have

$$\begin{aligned}
\|\mathbf{P}_Y - (\text{tr}(\Sigma))^{-1} \mathbf{Y} \mathbf{Y}^\top\| &= (\text{tr}(\Sigma))^{-1} \|\mathbf{Z}^\top \Sigma \mathbf{Z} - \text{tr}(\Sigma) \mathbf{I}_n\| \\
&= O_P \left(\frac{\sqrt{n \text{tr}(\Sigma^2)}}{\text{tr}(\Sigma)} + \frac{n \lambda_1}{\text{tr}(\Sigma)} \right) \\
&= O_P \left(\frac{\sqrt{n \lambda_1 \text{tr}(\Sigma)}}{\text{tr}(\Sigma)} + \frac{n \lambda_1}{\text{tr}(\Sigma)} \right) \\
&= O_P \left(\sqrt{\frac{n \lambda_1}{\text{tr}(\Sigma)}} \right).
\end{aligned} \tag{S3.2}$$

Now we deal with $\text{tr}((\mathbf{I}_p - \mathbf{P}_Y) \Sigma (\mathbf{I}_p - \mathbf{P}_Y))$. It can be seen that

$$\begin{aligned}
&\text{tr}((\mathbf{I}_p - \mathbf{P}_Y) \Sigma (\mathbf{I}_p - \mathbf{P}_Y)) \\
&= \text{tr}(\Sigma) - \text{tr}(\Sigma \mathbf{P}_Y) \\
&= \text{tr}(\Sigma) - \text{tr} \left(\left(\Sigma - \frac{\text{tr}(\Sigma^2)}{\text{tr}(\Sigma)} \mathbf{I}_p \right) \mathbf{P}_Y \right) - \frac{n \text{tr}(\Sigma^2)}{\text{tr}(\Sigma)}.
\end{aligned} \tag{S3.3}$$

For the second term, we have

$$\begin{aligned}
&\left| \text{tr} \left(\left(\Sigma - \frac{\text{tr}(\Sigma^2)}{\text{tr}(\Sigma)} \mathbf{I}_p \right) \mathbf{P}_Y \right) - (\text{tr}(\Sigma))^{-1} \text{tr} \left(\left(\Sigma - \frac{\text{tr}(\Sigma^2)}{\text{tr}(\Sigma)} \mathbf{I}_p \right) \mathbf{Y} \mathbf{Y}^\top \right) \right| \\
&= \left| \text{tr} \left(\left(\Sigma - \frac{\text{tr}(\Sigma^2)}{\text{tr}(\Sigma)} \mathbf{I}_p \right) (\mathbf{P}_Y - (\text{tr}(\Sigma))^{-1} \mathbf{Y} \mathbf{Y}^\top) \right) \right| \\
&\leq 2n \left\| \Sigma - \frac{\text{tr}(\Sigma^2)}{\text{tr}(\Sigma)} \mathbf{I}_p \right\| \|\mathbf{P}_Y - (\text{tr}(\Sigma))^{-1} \mathbf{Y} \mathbf{Y}^\top\| \\
&= O_P \left(n(\lambda_1 - \lambda_p) \sqrt{\frac{n \lambda_1}{\text{tr}(\Sigma)}} \right),
\end{aligned}$$

where the last inequality follows from von Neumann's trace theorem and the fact $\text{Rank}(\mathbf{P}_Y - (\text{tr}(\Sigma))^{-1} \mathbf{Y} \mathbf{Y}^\top) \leq 2n$, and the last equality follows from (S3.2) and the fact $\text{tr}(\Sigma^2)/\text{tr}(\Sigma) \in [\lambda_p, \lambda_1]$. On the other hand, it is

straightforward to show that

$$\mathbb{E} \left((\text{tr}(\boldsymbol{\Sigma}))^{-1} \text{tr} \left(\left(\boldsymbol{\Sigma} - \frac{\text{tr}(\boldsymbol{\Sigma}^2)}{\text{tr}(\boldsymbol{\Sigma})} \mathbf{I}_p \right) \mathbf{Y} \mathbf{Y}^\top \right) \right) = 0,$$

and

$$\begin{aligned} & \text{Var} \left((\text{tr}(\boldsymbol{\Sigma}))^{-1} \text{tr} \left(\left(\boldsymbol{\Sigma} - \frac{\text{tr}(\boldsymbol{\Sigma}^2)}{\text{tr}(\boldsymbol{\Sigma})} \mathbf{I}_p \right) \mathbf{Y} \mathbf{Y}^\top \right) \right) \\ &= \frac{2n}{\text{tr}^2(\boldsymbol{\Sigma})} \text{tr} \left(\boldsymbol{\Sigma}^2 - \frac{\text{tr}(\boldsymbol{\Sigma}^2)}{\text{tr}(\boldsymbol{\Sigma})} \boldsymbol{\Sigma} \right)^2 \\ &= \frac{2n}{\text{tr}^2(\boldsymbol{\Sigma})} \sum_{i=1}^p \lambda_i^2 \left(\lambda_i - \frac{\text{tr}(\boldsymbol{\Sigma}^2)}{\text{tr}(\boldsymbol{\Sigma})} \right)^2 \\ &\leq \frac{2n\lambda_1(\lambda_1 - \lambda_p)^2}{\text{tr}(\boldsymbol{\Sigma})}. \end{aligned}$$

Thus,

$$\text{tr} \left(\left(\boldsymbol{\Sigma} - \frac{\text{tr}(\boldsymbol{\Sigma}^2)}{\text{tr}(\boldsymbol{\Sigma})} \mathbf{I}_p \right) \mathbf{P}_\mathbf{Y} \right) = O_P \left(n(\lambda_1 - \lambda_p) \sqrt{\frac{n\lambda_1}{\text{tr}(\boldsymbol{\Sigma})}} \right).$$

Then the first statement follows from the last display and (S3.3).

Next we deal with $\text{tr}((\mathbf{I}_p - \mathbf{P}_\mathbf{Y})\boldsymbol{\Sigma}(\mathbf{I}_p - \mathbf{P}_\mathbf{Y}))^2$. We have

$$\text{tr}((\mathbf{I}_p - \mathbf{P}_\mathbf{Y})\boldsymbol{\Sigma}(\mathbf{I}_p - \mathbf{P}_\mathbf{Y}))^2 = \text{tr}(\boldsymbol{\Sigma}^2) - 2\text{tr}(\boldsymbol{\Sigma}^2\mathbf{P}_\mathbf{Y}) + \text{tr}((\boldsymbol{\Sigma}\mathbf{P}_\mathbf{Y})^2).$$

From von Neumann's trace theorem, the second term satisfies

$$\left| \text{tr}(\boldsymbol{\Sigma}^2\mathbf{P}_\mathbf{Y}) - \frac{n\text{tr}^2(\boldsymbol{\Sigma}^2)}{\text{tr}^2(\boldsymbol{\Sigma})} \right| = \left| \text{tr} \left(\left(\boldsymbol{\Sigma}^2 - \frac{\text{tr}^2(\boldsymbol{\Sigma}^2)}{\text{tr}^2(\boldsymbol{\Sigma})} \mathbf{I}_p \right) \mathbf{P}_\mathbf{Y} \right) \right| \leq n\lambda_1(\lambda_1 - \lambda_p),$$

and the third term satisfies

$$\begin{aligned}
& \left| \text{tr}((\mathbf{\Sigma} \mathbf{P}_Y)^2) - \frac{n \text{tr}^2(\mathbf{\Sigma}^2)}{\text{tr}^2(\mathbf{\Sigma})} \right| \\
&= \left| \text{tr} \left(\left(\mathbf{\Sigma} + \frac{\text{tr}(\mathbf{\Sigma}^2)}{\text{tr}(\mathbf{\Sigma})} \mathbf{I}_p \right) \mathbf{P}_Y \left(\mathbf{\Sigma} - \frac{\text{tr}(\mathbf{\Sigma}^2)}{\text{tr}(\mathbf{\Sigma})} \mathbf{I}_p \right) \mathbf{P}_Y \right) \right| \\
&\leq 2n\lambda_1(\lambda_1 - \lambda_p).
\end{aligned}$$

This completes the proof of the second statement. \square

Proof of Theorem 1. In the current context, Lemma 6 implies that

$$\text{tr}((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{\Sigma}(\mathbf{I}_p - \mathbf{P}_Y)) = \text{tr}(\mathbf{\Sigma}) - \frac{n \text{tr}(\mathbf{\Sigma}^2)}{\text{tr}(\mathbf{\Sigma})} + o_P(\sqrt{\text{tr}(\mathbf{\Sigma}^2)}), \quad (\text{S3.4})$$

$$\text{tr}((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{\Sigma}(\mathbf{I}_p - \mathbf{P}_Y))^2 = (1 + o_P(1)) \text{tr}(\mathbf{\Sigma}^2). \quad (\text{S3.5})$$

The fact $\lambda_1((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{\Sigma}(\mathbf{I}_p - \mathbf{P}_Y)) \leq \lambda_1$ and (S3.5) imply that the first term of (S3.1) satisfies the Lyapunov condition

$$\frac{\lambda_1((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{\Sigma}(\mathbf{I}_p - \mathbf{P}_Y))}{\sqrt{\text{tr}((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{\Sigma}(\mathbf{I}_p - \mathbf{P}_Y))^2}} \leq \frac{\lambda_1}{\sqrt{(1 + o_P(1)) \text{tr}^2(\mathbf{\Sigma})}} \xrightarrow{P} 0.$$

From Lemma 2, we have

$$\frac{\mathbf{Z}^\dagger \mathbf{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U} \mathbf{\Lambda}^{1/2} \mathbf{Z}^\dagger - \text{tr}((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{\Sigma}(\mathbf{I}_p - \mathbf{P}_Y)) \mathbf{I}_{k-1}}{\sqrt{\text{tr}((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{\Sigma}(\mathbf{I}_p - \mathbf{P}_Y))^2}} \xrightarrow{\mathcal{L}} \mathbf{W}_{k-1}.$$

Then it follows from (S3.4), (S3.5) and Slutsky's theorem that

$$\frac{\mathbf{Z}^\dagger \mathbf{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U} \mathbf{\Lambda}^{1/2} \mathbf{Z}^\dagger - (\text{tr}(\mathbf{\Sigma}) - n \text{tr}(\mathbf{\Sigma}^2)/\text{tr}(\mathbf{\Sigma})) \mathbf{I}_{k-1}}{\sqrt{\text{tr}(\mathbf{\Sigma}^2)}} \xrightarrow{\mathcal{L}} \mathbf{W}_{k-1}. \quad (\text{S3.6})$$

Next we consider the second term of (S3.1). Note that

$$\|\mathbf{C}^\top \boldsymbol{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_Y) \boldsymbol{\Theta} \mathbf{C} - \mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Theta} \mathbf{C}\| = \|\mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{Y} (\mathbf{Y}^\top \mathbf{Y})^{-1} \mathbf{Y}^\top \boldsymbol{\Theta} \mathbf{C}\|.$$

We have

$$\begin{aligned} & \|\mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{Y} (\mathbf{Y}^\top \mathbf{Y})^{-1} \mathbf{Y}^\top \boldsymbol{\Theta} \mathbf{C} - \text{tr}(\boldsymbol{\Sigma})^{-1} \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{Y} \mathbf{Y}^\top \boldsymbol{\Theta} \mathbf{C}\| \\ & \leq \|\mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{Y} \mathbf{Y}^\top \boldsymbol{\Theta} \mathbf{C}\| \|(\mathbf{Y}^\top \mathbf{Y})^{-1} - \text{tr}(\boldsymbol{\Sigma})^{-1} \mathbf{I}_n\| \\ & \leq \|\text{tr}(\boldsymbol{\Sigma})^{-1} \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{Y} \mathbf{Y}^\top \boldsymbol{\Theta} \mathbf{C}\| \|(\mathbf{Y}^\top \mathbf{Y})^{-1}\| \|\mathbf{Y}^\top \mathbf{Y} - \text{tr}(\boldsymbol{\Sigma}) \mathbf{I}_n\|. \end{aligned}$$

From Lemma 5, we have

$$\begin{aligned} \|\mathbf{Y}^\top \mathbf{Y} - \text{tr}(\boldsymbol{\Sigma}) \mathbf{I}_n\| &= \|\mathbf{Z}^\top \boldsymbol{\Lambda} \mathbf{Z} - \text{tr}(\boldsymbol{\Sigma}) \mathbf{I}_n\| \\ &= O_P(\sqrt{n \text{tr}(\boldsymbol{\Sigma}^2)} + n \lambda_1) \\ &= o_P(\text{tr}(\boldsymbol{\Sigma})). \end{aligned}$$

Then $\|(\mathbf{Y}^\top \mathbf{Y})^{-1}\| = \lambda_n^{-1}(\mathbf{Z}^\top \boldsymbol{\Lambda} \mathbf{Z}) = (1 + o_P(1)) \text{tr}(\boldsymbol{\Sigma})$. Therefore,

$$\begin{aligned} & \|\mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{Y} (\mathbf{Y}^\top \mathbf{Y})^{-1} \mathbf{Y}^\top \boldsymbol{\Theta} \mathbf{C} - \text{tr}(\boldsymbol{\Sigma})^{-1} \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{Y} \mathbf{Y}^\top \boldsymbol{\Theta} \mathbf{C}\| \\ &= o_P(\|\text{tr}(\boldsymbol{\Sigma})^{-1} \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{Y} \mathbf{Y}^\top \boldsymbol{\Theta} \mathbf{C}\|). \end{aligned}$$

Note that the columns of $\mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{Y} = \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{U} \boldsymbol{\Lambda}^{1/2} \mathbf{Z}$ are iid $\mathcal{N}_{k-1}(\mathbf{0}_{k-1}, \mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Sigma} \boldsymbol{\Theta} \mathbf{C})$

random vectors. Hence we can write $\mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{Y} = (\mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Sigma} \boldsymbol{\Theta} \mathbf{C})^{1/2} \mathbf{Z}^*$, where

\mathbf{Z}^* is a $(k-1) \times n$ random matrix with iid $\mathcal{N}(0, 1)$ entries. Then

$$\begin{aligned} & \left\| \text{tr}(\boldsymbol{\Sigma})^{-1} \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{Y} \mathbf{Y}^\top \boldsymbol{\Theta} \mathbf{C} - \frac{n}{\text{tr}(\boldsymbol{\Sigma})} \mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Sigma} \boldsymbol{\Theta} \mathbf{C} \right\| \\ & \leq \frac{n}{\text{tr}(\boldsymbol{\Sigma})} \|\mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Sigma} \boldsymbol{\Theta} \mathbf{C}\| \|n^{-1} \mathbf{Z}^* \mathbf{Z}^{*\top} - \mathbf{I}_{k-1}\| \\ &= o_P\left(\frac{n}{\text{tr}(\boldsymbol{\Sigma})} \|\mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Sigma} \boldsymbol{\Theta} \mathbf{C}\|\right), \end{aligned}$$

where the last equality follows from the law of large numbers. Combine the above arguments, we have

$$\begin{aligned}
\|\mathbf{C}^\top \boldsymbol{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_Y) \boldsymbol{\Theta} \mathbf{C} - \mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Theta} \mathbf{C}\| &= (1 + o_P(1)) \frac{n}{\text{tr}(\boldsymbol{\Sigma})} \|\mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Sigma} \boldsymbol{\Theta} \mathbf{C}\| \\
&\leq (1 + o_P(1)) \frac{n\lambda_1}{\text{tr}(\boldsymbol{\Sigma})} \|\mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Theta} \mathbf{C}\| \\
&= o_P\left(\sqrt{\text{tr}(\boldsymbol{\Sigma}^2)}\right).
\end{aligned} \tag{S3.7}$$

Now we deal with the cross term of (S3.1). Note that

$$\begin{aligned}
&\mathbb{E}[\|\mathbf{C}^\top \boldsymbol{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U} \boldsymbol{\Lambda}^{1/2} \mathbf{Z}^\dagger\|_F^2 | \mathbf{Y}] \\
&= (k-1) \text{tr}(\mathbf{C}^\top \boldsymbol{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_Y) \boldsymbol{\Sigma} (\mathbf{I}_p - \mathbf{P}_Y) \boldsymbol{\Theta} \mathbf{C}) \\
&\leq (k-1) \lambda_1 \text{tr}(\mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Theta} \mathbf{C}).
\end{aligned}$$

Therefore,

$$\begin{aligned}
\|\mathbf{C}^\top \boldsymbol{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U} \boldsymbol{\Lambda}^{1/2} \mathbf{Z}^\dagger\| &= o_P\left(\sqrt{\lambda_1 \text{tr}(\mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Theta} \mathbf{C})}\right) \\
&= o_P\left(\sqrt{\text{tr}(\boldsymbol{\Sigma}^2)}\right),
\end{aligned} \tag{S3.8}$$

where the last equality follows from the conditions $\lambda_1/\sqrt{\text{tr}(\boldsymbol{\Sigma}^2)} \rightarrow 0$ and $\text{tr}(\mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Theta} \mathbf{C}) \leq (k-1) \|\mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Theta} \mathbf{C}\| = O(\sqrt{\text{tr}(\boldsymbol{\Sigma}^2)})$.

It follows from (S3.7), (S3.8) and Weyl's inequality that

$$\begin{aligned}
 & |T(\mathbf{X}) - (\lambda_1 (\mathbf{Z}^{\dagger\top} \mathbf{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U} \mathbf{\Lambda}^{1/2} \mathbf{Z}^\dagger + \mathbf{C}^\top \mathbf{\Theta}^\top \mathbf{\Theta} \mathbf{C}))| \\
 & \leq \|\mathbf{C}^\top \mathbf{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{\Theta} \mathbf{C} - \mathbf{C}^\top \mathbf{\Theta}^\top \mathbf{\Theta} \mathbf{C} \\
 & \quad + \mathbf{C}^\top \mathbf{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U} \mathbf{\Lambda}^{1/2} \mathbf{Z}^\dagger + \mathbf{Z}^{\dagger\top} \mathbf{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{\Theta} \mathbf{C}\| \\
 & \leq \|\mathbf{C}^\top \mathbf{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{\Theta} \mathbf{C} - \mathbf{C}^\top \mathbf{\Theta}^\top \mathbf{\Theta} \mathbf{C}\| + 2 \|\mathbf{C}^\top \mathbf{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U} \mathbf{\Lambda}^{1/2} \mathbf{Z}^\dagger\| \\
 & = o_P \left(\sqrt{\text{tr}(\mathbf{\Sigma}^2)} \right).
 \end{aligned}$$

But (S3.6) implies that

$$\begin{aligned}
 & \frac{1}{\sqrt{\text{tr}(\mathbf{\Sigma}^2)}} (\lambda_1 (\mathbf{Z}^{\dagger\top} \mathbf{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U} \mathbf{\Lambda}^{1/2} \mathbf{Z}^\dagger + \mathbf{C}^\top \mathbf{\Theta}^\top \mathbf{\Theta} \mathbf{C}) \\
 & \quad - (\text{tr}(\mathbf{\Sigma}) - n \text{tr}(\mathbf{\Sigma}^2)/\text{tr}(\mathbf{\Sigma}))) \\
 & = \lambda_1 \left(\frac{\mathbf{Z}^{\dagger\top} \mathbf{\Lambda}^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U} \mathbf{\Lambda}^{1/2} \mathbf{Z}^\dagger - (\text{tr}(\mathbf{\Sigma}) - n \text{tr}(\mathbf{\Sigma}^2)/\text{tr}(\mathbf{\Sigma})) \mathbf{I}_{k-1}}{\sqrt{\text{tr}(\mathbf{\Sigma}^2)}} \right. \\
 & \quad \left. + \frac{\mathbf{C}^\top \mathbf{\Theta}^\top \mathbf{\Theta} \mathbf{C}}{\sqrt{\text{tr}(\mathbf{\Sigma}^2)}} \right) \\
 & \sim \lambda_1 \left(\mathbf{W}_{k-1} + \frac{\mathbf{C}^\top \mathbf{\Theta}^\top \mathbf{\Theta} \mathbf{C}}{\sqrt{\text{tr}(\mathbf{\Sigma}^2)}} \right) + o_P(1).
 \end{aligned}$$

This completes the proof.

□

Proof of Corollary 1. It is straightforward to show that $\widehat{\text{E tr}(\mathbf{\Sigma})} = \text{tr}(\mathbf{\Sigma})$ and $\widehat{\text{Var}(\text{tr}(\mathbf{\Sigma}))} = 2n^{-1} \text{tr}(\mathbf{\Sigma}^2)$. Then $\widehat{\text{tr}(\mathbf{\Sigma})} = \text{tr}(\mathbf{\Sigma}) + O_P(\sqrt{n^{-1} \text{tr}(\mathbf{\Sigma}^2)})$.

Let Z_1, \dots, Z_n be the columns of \mathbf{Z} . Then we have

$$\begin{aligned} \widehat{\text{tr}(\Sigma^2)} &= n^{-2} \text{tr}(\mathbf{Z}^\top \mathbf{\Lambda} \mathbf{Z} - n^{-1} \text{tr}(\mathbf{Z}^\top \mathbf{\Lambda} \mathbf{Z}) \mathbf{I}_n)^2 \\ &= n^{-2} \sum_{i=1}^n (Z_i^\top \mathbf{\Lambda} Z_i - n^{-1} \sum_{i=1}^n Z_i^\top \mathbf{\Lambda} Z_i)^2 + 2n^{-2} \sum_{1 \leq i < j \leq n} (Z_i^\top \mathbf{\Lambda} Z_i)^2. \end{aligned}$$

It can be seen that $n^{-2} \sum_{i=1}^n (Z_i^\top \mathbf{\Lambda} Z_i - n^{-1} \sum_{i=1}^n Z_i^\top \mathbf{\Lambda} Z_i)^2 = O_P(n^{-1} \text{tr}(\Sigma^2))$.

On the other hand, we have $E 2 \sum_{1 \leq i < j \leq n} (Z_i^\top \mathbf{\Lambda} Z_i)^2 = n(n-1) \text{tr}(\Sigma^2)$. Furthermore, Chen et al. (2010), Proposition A.2 implies that

$$\text{Var} \left(2 \sum_{1 \leq i < j \leq n} (Z_i^\top \mathbf{\Lambda} Z_j)^2 \right) = O(n^2 \text{tr}^2(\Sigma^2) + n^3 \text{tr}(\Sigma^4)) = O(n^3 \text{tr}^2(\Sigma^2)).$$

Hence $\widehat{\text{tr}(\Sigma^2)} = (1 + O_P(n^{-1/2})) \text{tr}(\Sigma^2)$.

Thus, we have

$$\begin{aligned} & \widehat{\text{tr}(\Sigma)} - n \widehat{\text{tr}(\Sigma^2)} / \widehat{\text{tr}(\Sigma)} \\ &= \text{tr}(\Sigma) + O_P(\sqrt{n^{-1} \text{tr}(\Sigma^2)}) - \frac{n \text{tr}(\Sigma^2)(1 + O_P(n^{-1/2}))}{\text{tr}(\Sigma)(1 + O_P(\sqrt{n^{-1} \text{tr}(\Sigma^2)} / \text{tr}^2(\Sigma)))} \\ &= \text{tr}(\Sigma) + O_P(\sqrt{n^{-1} \text{tr}(\Sigma^2)}) - \frac{n \text{tr}(\Sigma^2)}{\text{tr}(\Sigma)} \left(1 + O_P \left(\frac{1}{\sqrt{n}} + \sqrt{\frac{\text{tr}(\Sigma^2)}{n \text{tr}^2(\Sigma)}} \right) \right) \\ &= \text{tr}(\Sigma) - \frac{n \text{tr}(\Sigma^2)}{\text{tr}(\Sigma)} + o_P(\sqrt{\text{tr}(\Sigma^2)}). \end{aligned}$$

Therefore,

$$Q_1 = \frac{T(\mathbf{X}) - (\text{tr}(\Sigma) - n \text{tr}(\Sigma^2) / \text{tr}(\Sigma))}{\sqrt{\text{tr}(\Sigma^2)}} + o_P(1).$$

Then the conclusion follows from Theorem 1. \square

Lemma 7. Suppose that $r = o(n)$, $\text{tr}(\Lambda_2)\lambda_1/(n\lambda_r^2) \rightarrow 0$, $n\lambda_{r+1}/\text{tr}(\Lambda_2) \rightarrow$

0. Then uniformly for $i = 1, \dots, r$,

$$\begin{aligned} & \lambda_i((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y)) \\ &= n^{-1} \text{tr}(\Lambda_2) \left(1 + O_P \left(\sqrt{\frac{\text{tr}(\Lambda_2)\lambda_1}{n\lambda_r^2}} + \sqrt{\frac{n\lambda_{r+1}}{\text{tr}(\Lambda_2)}} + \sqrt{\frac{r}{n}} \right) \right). \end{aligned}$$

Proof. Note that

$$(\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y) = (\mathbf{I}_p - \mathbf{P}_{Y,2})(\mathbf{I}_p - \mathbf{P}_{Y,1})\Sigma(\mathbf{I}_p - \mathbf{P}_{Y,1})(\mathbf{I}_p - \mathbf{P}_{Y,2}). \quad (\text{S3.9})$$

We first deal with $(\mathbf{I}_p - \mathbf{P}_{Y,1})\Sigma(\mathbf{I}_p - \mathbf{P}_{Y,1})$. Under the condition $n\lambda_{r+1}/\text{tr}(\Lambda_2) \rightarrow$

0, Proposition 3 implies that

$$\|\mathbf{U}_{Y,1}\mathbf{U}_{Y,1}^\top - \mathbf{P}_{Y,1}^\dagger\| = O_P\left(\frac{\text{tr}(\Lambda_2)}{n\lambda_r}\right).$$

From the decomposition

$$\begin{aligned} & (\mathbf{I}_p - \mathbf{P}_{Y,1})\Sigma(\mathbf{I}_p - \mathbf{P}_{Y,1}) \\ &= (\mathbf{I}_p - \mathbf{P}_{Y,1}^\dagger)\Sigma(\mathbf{I}_p - \mathbf{P}_{Y,1}^\dagger) + (\mathbf{P}_{Y,1}^\dagger - \mathbf{P}_{Y,1})\Sigma(\mathbf{I}_p - \mathbf{P}_{Y,1}^\dagger) \\ & \quad + (\mathbf{I}_p - \mathbf{P}_{Y,1}^\dagger)\Sigma(\mathbf{P}_{Y,1}^\dagger - \mathbf{P}_{Y,1}) + (\mathbf{P}_{Y,1}^\dagger - \mathbf{P}_{Y,1})\Sigma(\mathbf{P}_{Y,1}^\dagger - \mathbf{P}_{Y,1}), \end{aligned}$$

we have

$$\begin{aligned} & \left\| (\mathbf{I}_p - \mathbf{P}_{Y,1})\Sigma(\mathbf{I}_p - \mathbf{P}_{Y,1}) - (\mathbf{I}_p - \mathbf{P}_{Y,1}^\dagger)\Sigma(\mathbf{I}_p - \mathbf{P}_{Y,1}^\dagger) \right\| \\ & \leq 2 \left\| \mathbf{P}_{Y,1}^\dagger - \mathbf{P}_{Y,1} \right\| \left\| \Sigma(\mathbf{I}_p - \mathbf{P}_{Y,1}^\dagger) \right\| + \lambda_1 \left\| \mathbf{P}_{Y,1}^\dagger - \mathbf{P}_{Y,1} \right\|^2. \\ & = O_P\left(\frac{\text{tr}(\Lambda_2)}{n\lambda_r}\right) \left\| \Sigma(\mathbf{I}_p - \mathbf{P}_{Y,1}^\dagger) \right\| + O_P\left(\frac{\text{tr}^2(\Lambda_2)\lambda_1}{n^2\lambda_r^2}\right). \end{aligned}$$

Note that

$$\begin{aligned}
& \left\| \Sigma(\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \right\| \\
&= \left\| \mathbf{U}_2 \mathbf{\Lambda}_2 \mathbf{U}_2^\top - \mathbf{U}_1 \mathbf{\Lambda}_1 \mathbf{Q}^\top \mathbf{U}_2^\top - \mathbf{U}_2 \mathbf{\Lambda}_2 \mathbf{Q} \mathbf{U}_1^\top \right\| \\
&\leq \lambda_{r+1} + \left\| \mathbf{\Lambda}_1 \mathbf{Q}^\top \right\| + \lambda_{r+1} \left\| \mathbf{Q} \right\| \\
&= \lambda_{r+1} + \left\| \mathbf{\Lambda}_1^{1/2} (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{-1/2} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^{1/2} \right\| + \lambda_{r+1} \left\| \mathbf{Q}^\top \mathbf{Q} \right\|^{1/2} \\
&\leq \lambda_{r+1} + \lambda_1^{1/2} \left\| (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{-1/2} \right\| \left\| \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \right\|^{1/2} + \lambda_{r+1} \left\| \mathbf{Q}^\top \mathbf{Q} \right\|^{1/2} \\
&= O_P \left(\sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2) \lambda_1}{n}} \right),
\end{aligned}$$

where the last equality follows from (S2.9), (S2.11) and the condition $n\lambda_{r+1}/\text{tr}(\mathbf{\Lambda}_2) \rightarrow$

0. Thus,

$$\begin{aligned}
& \left\| (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}) \Sigma(\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}) - (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \Sigma(\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \right\| \\
&= O_P \left(\frac{\text{tr}^{3/2}(\mathbf{\Lambda}_2) \lambda_1^{1/2}}{n^{3/2} \lambda_r} \right).
\end{aligned} \tag{S3.10}$$

From the decomposition

$$\begin{aligned}
& (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \Sigma(\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \\
&= \mathbf{U}_2 \mathbf{Q} \mathbf{\Lambda}_1 \mathbf{Q}^\top \mathbf{U}_2^\top + \mathbf{U}_2 \mathbf{\Lambda}_2 \mathbf{U}_2^\top - \mathbf{U}_2 \mathbf{\Lambda}_2 \mathbf{Q} \mathbf{U}_1^\top - \mathbf{U}_1 \mathbf{Q}^\top \mathbf{\Lambda}_2 \mathbf{U}_2^\top + \mathbf{U}_1 \mathbf{Q}^\top \mathbf{\Lambda}_2 \mathbf{Q} \mathbf{U}_1^\top,
\end{aligned}$$

we have

$$\begin{aligned}
& \left\| (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) \Sigma(\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},1}^\dagger) - \mathbf{U}_2 \mathbf{Q} \mathbf{\Lambda}_1 \mathbf{Q}^\top \mathbf{U}_2^\top \right\| \\
&\leq \lambda_{r+1} (1 + 2 \left\| \mathbf{Q}^\top \mathbf{Q} \right\|^{1/2} + \left\| \mathbf{Q}^\top \mathbf{Q} \right\|) \\
&= O_P(\lambda_{r+1}),
\end{aligned} \tag{S3.11}$$

where the last equality follows from (S2.11). Note that $\mathbf{U}_2 \mathbf{Q} \mathbf{\Lambda}_1 \mathbf{Q}^\top \mathbf{U}_2^\top = \mathbf{U}_2 \mathbf{\Lambda}_2^{1/2} \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{-1} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^{1/2} \mathbf{U}_2^\top$. We have

$$\begin{aligned}
 & \left\| \mathbf{U}_2 \mathbf{Q} \mathbf{\Lambda}_1 \mathbf{Q}^\top \mathbf{U}_2^\top - n^{-1} \mathbf{U}_2 \mathbf{\Lambda}_2^{1/2} \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^{1/2} \mathbf{U}_2^\top \right\| \\
 & \leq \left\| \mathbf{U}_2 \mathbf{\Lambda}_2^{1/2} \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^{1/2} \mathbf{U}_2^\top \right\| \left\| (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{-1} - n^{-1} \mathbf{I}_r \right\| \\
 & \leq \left\| \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \right\| \left\| (\mathbf{Z}_1 \mathbf{Z}_1^\top)^{-1} \right\| \left\| n^{-1} \mathbf{Z}_1 \mathbf{Z}_1^\top - \mathbf{I}_r \right\| \\
 & = O_P \left(\frac{r^{1/2} \text{tr}(\mathbf{\Lambda}_2)}{n^{3/2}} \right),
 \end{aligned} \tag{S3.12}$$

where the last equality follows from (S2.3) and (S2.9). From (S3.9), (S3.10), (S3.11) and (S3.12), we obtain that

$$\begin{aligned}
 & \|(\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \mathbf{\Sigma} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \\
 & - n^{-1} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},2}) \mathbf{U}_2 \mathbf{\Lambda}_2^{1/2} \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^{1/2} \mathbf{U}_2^\top (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y},2}) \| \\
 & = O_P \left(\left(\sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2) \lambda_1}{n \lambda_r^2}} + \frac{n \lambda_{r+1}}{\text{tr}(\mathbf{\Lambda}_2)} + \sqrt{\frac{r}{n}} \right) \frac{\text{tr}(\mathbf{\Lambda}_2)}{n} \right).
 \end{aligned}$$

Thus, the last display, together with Weyl's inequality, implies that uniformly for $i = 1, \dots, r$,

$$\begin{aligned}
 & \lambda_i ((\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \mathbf{\Sigma} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}})) \\
 & = n^{-1} \lambda_i \left(\mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^{1/2} \mathbf{U}_2^\top (\mathbf{I} - \mathbf{P}_{\mathbf{Y},2}) \mathbf{U}_2 \mathbf{\Lambda}_2^{1/2} \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \right) \\
 & + O_P \left(\left(\sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2) \lambda_1}{n \lambda_r^2}} + \frac{n \lambda_{r+1}}{\text{tr}(\mathbf{\Lambda}_2)} + \sqrt{\frac{r}{n}} \right) \frac{\text{tr}(\mathbf{\Lambda}_2)}{n} \right).
 \end{aligned}$$

Note that

$$\begin{aligned}
& \left\| n^{-1} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2^{1/2} \mathbf{U}_2^\top (\mathbf{I} - \mathbf{P}_{\mathbf{Y},2}) \mathbf{U}_2 \Lambda_2^{1/2} \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \right. \\
& \quad \left. - \left(n^{-1} \text{tr}(\Lambda_2) \mathbf{I}_r - (n \text{tr}(\Lambda_2))^{-1} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \right) \right\| \\
& \leq \left\| n^{-1} \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} - n^{-1} \text{tr}(\Lambda_2) \mathbf{I}_r \right\| \\
& \quad + n^{-1} \left\| \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \right\| \left\| \mathbf{P}_{\mathbf{Y},2} - (\text{tr}(\Lambda_2))^{-1} \Lambda_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2^{1/2} \right\| \\
& = O_P \left(\left(\sqrt{\frac{\text{tr}(\Lambda_2) \lambda_1}{n \lambda_r^2}} + \sqrt{\frac{n \lambda_{r+1}}{\text{tr}(\Lambda_2)}} \right) \frac{\text{tr}(\Lambda_2)}{n} \right),
\end{aligned}$$

where the last equality follows from (S2.9) and Proposition 4. Then it follows from Weyl's inequality that uniformly for $i = 1, \dots, r$,

$$\begin{aligned}
& \lambda_i((\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \Sigma (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}})) \\
& = n^{-1} \text{tr}(\Lambda_2) \\
& \quad - (n \text{tr}(\Lambda_2))^{-1} \lambda_{r+1-i} \left(\mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \right) \\
& \quad + O_P \left(\left(\sqrt{\frac{\text{tr}(\Lambda_2) \lambda_1}{n \lambda_r^2}} + \sqrt{\frac{n \lambda_{r+1}}{\text{tr}(\Lambda_2)}} + \sqrt{\frac{r}{n}} \right) \frac{\text{tr}(\Lambda_2)}{n} \right).
\end{aligned} \tag{S3.13}$$

Now we deal with the matrix $\mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1}$. Note that $\mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1}$ and $\mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1}$ both have iid $\mathcal{N}(0, 1)$ entries and they are mutually independent. Then Lemma 5 implies that

$$\begin{aligned}
& \left\| \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} - \text{tr}(\Lambda_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2) \mathbf{I}_r \right\| \\
& = O_P \left(\sqrt{r \text{tr}(\Lambda_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2)^2} + r \left\| \Lambda_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \Lambda_2 \right\| \right).
\end{aligned}$$

By some algebra, we have

$$\begin{aligned} & \left\| \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} - \text{tr}(\tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1}) \mathbf{I}_r \right\| \\ &= O_P \left(\sqrt{r} \left\| \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \right\| \text{tr}(\tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1}) \right. \\ & \quad \left. + r \left\| \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \right\| \right). \end{aligned}$$

Since $E \text{tr}(\tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1}) = (n - r) \text{tr}(\mathbf{\Lambda}_2^2)$, we have

$$\text{tr}(\tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1}) = O_P(n \text{tr}(\mathbf{\Lambda}_2^2)) = O_P(n \boldsymbol{\lambda}_{r+1} \text{tr}(\mathbf{\Lambda}_2)).$$

On the other hand, Lemma 5 implies that

$$\left\| \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2^2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \right\| = O_P(\text{tr}(\mathbf{\Lambda}_2^2) + n \boldsymbol{\lambda}_{r+1}^2) = O_P(\boldsymbol{\lambda}_{r+1} \text{tr}(\mathbf{\Lambda}_2)).$$

Combine these bounds, we have

$$\left\| \mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} - n \text{tr}(\mathbf{\Lambda}_2^2) \mathbf{I}_r \right\| = O_P(\sqrt{rn} \boldsymbol{\lambda}_{r+1} \text{tr}(\mathbf{\Lambda}_2)).$$

The last display, combined with Weyl's inequality, implies that uniformly for $i = 1, \dots, r$,

$$(n \text{tr}(\mathbf{\Lambda}_2))^{-1} \lambda_i \left(\mathbf{V}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \mathbf{\Lambda}_2 \mathbf{Z}_2 \mathbf{V}_{\mathbf{Z}_1} \right) = O_P(\boldsymbol{\lambda}_{r+1}).$$

Then (S3.13) and the last display implies that uniformly for $i = 1, \dots, r$,

$$\begin{aligned} & \lambda_i((\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \boldsymbol{\Sigma} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}})) \\ &= n^{-1} \text{tr}(\mathbf{\Lambda}_2) + O_P \left(\left(\sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2) \boldsymbol{\lambda}_1}{n \boldsymbol{\lambda}_r^2}} + \sqrt{\frac{n \boldsymbol{\lambda}_{r+1}}{\text{tr}(\mathbf{\Lambda}_2)}} + \sqrt{\frac{r}{n}} \right) \frac{\text{tr}(\mathbf{\Lambda}_2)}{n} \right). \end{aligned}$$

This completes the proof. □

Lemma 8. *Suppose that $r = o(n)$, $\text{tr}(\Lambda_2)\lambda_1/(n\lambda_r^2) \rightarrow 0$, $n\lambda_{r+1}/\text{tr}(\Lambda_2) \rightarrow$*

0. Then

$$\begin{aligned} & \sum_{i=r+1}^p \lambda_i ((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y)) \\ &= \text{tr}(\Lambda_2) - \frac{n \text{tr}(\Lambda_2^2)}{\text{tr}(\Lambda_2)} \\ & \quad + O_P \left(n(\lambda_{r+1} - \lambda_p) \left(\sqrt{\frac{\text{tr}(\Lambda_2)\lambda_1}{n\lambda_r^2}} + \sqrt{\frac{n\lambda_{r+1}}{\text{tr}(\Lambda_2)}} \right) + r\lambda_{r+1} \right). \end{aligned}$$

Proof. Write $\Sigma = \mathbf{U}_1\Lambda_1\mathbf{U}_1^\top + \mathbf{U}_2\Lambda_2\mathbf{U}_2^\top$. Note that $\mathbf{U}_1\Lambda_1\mathbf{U}_1^\top$ is of rank r .

Then Weyl's inequality implies that for $i = r+1, \dots, p$,

$$\lambda_i ((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y)) \geq \lambda_i ((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{U}_2\Lambda_2\mathbf{U}_2^\top(\mathbf{I}_p - \mathbf{P}_Y)), \quad (\text{S3.14})$$

$$\begin{aligned} \lambda_i ((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y)) &\leq \lambda_{i-r} ((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{U}_2\Lambda_2\mathbf{U}_2^\top(\mathbf{I}_p - \mathbf{P}_Y)). \\ & \hspace{15em} (\text{S3.15}) \end{aligned}$$

Hence we have

$$\begin{aligned} & \left| \sum_{i=r+1}^p \lambda_i ((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y)) - \text{tr}((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{U}_2\Lambda_2\mathbf{U}_2^\top(\mathbf{I}_p - \mathbf{P}_Y)) \right| \\ & \leq r\lambda_1 ((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{U}_2\Lambda_2\mathbf{U}_2^\top(\mathbf{I}_p - \mathbf{P}_Y)) \\ & \leq r\lambda_{r+1}. \end{aligned} \quad (\text{S3.16})$$

Write

$$\begin{aligned}
& \text{tr} \left((\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U}_2 \boldsymbol{\Lambda}_2 \mathbf{U}_2^\top (\mathbf{I}_p - \mathbf{P}_Y) \right) \\
&= \text{tr} \left(\boldsymbol{\Lambda}_2 \mathbf{U}_2^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U}_2 \right) \\
&= \text{tr}(\boldsymbol{\Lambda}_2) - \text{tr} \left(\left(\boldsymbol{\Lambda}_2 - \frac{\text{tr}(\boldsymbol{\Lambda}_2^2)}{\text{tr}(\boldsymbol{\Lambda}_2)} \mathbf{I}_{p-r} \right) \mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2 \right) - \frac{\text{tr}(\boldsymbol{\Lambda}_2^2)}{\text{tr}(\boldsymbol{\Lambda}_2)} \text{tr}(\mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2).
\end{aligned} \tag{S3.17}$$

For the third term, note that $\text{tr}(\mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2) = \text{tr}(\mathbf{P}_Y) - \text{tr}(\mathbf{P}_Y \mathbf{U}_1 \mathbf{U}_1^\top)$.

Since \mathbf{P}_Y is of rank n and \mathbf{U}_1 is of rank r , we have

$$|\text{tr}(\mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2) - n| \leq r. \tag{S3.18}$$

Next we deal with the second term. We have

$$\begin{aligned}
& \left| \text{tr} \left(\left(\boldsymbol{\Lambda}_2 - \frac{\text{tr}(\boldsymbol{\Lambda}_2^2)}{\text{tr}(\boldsymbol{\Lambda}_2)} \mathbf{I}_{p-r} \right) \mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2 \right) \right. \\
& \quad \left. - \text{tr} \left(\left(\boldsymbol{\Lambda}_2 - \frac{\text{tr}(\boldsymbol{\Lambda}_2^2)}{\text{tr}(\boldsymbol{\Lambda}_2)} \mathbf{I}_{p-r} \right) \mathbf{U}_2^\top (\mathbf{P}_{Y,1}^\dagger + \mathbf{P}_{Y,2}^\dagger) \mathbf{U}_2 \right) \right| \\
&= \left| \text{tr} \left(\left(\boldsymbol{\Lambda}_2 - \frac{\text{tr}(\boldsymbol{\Lambda}_2^2)}{\text{tr}(\boldsymbol{\Lambda}_2)} \mathbf{I}_{p-r} \right) \mathbf{U}_2^\top (\mathbf{P}_Y - \mathbf{P}_{Y,1}^\dagger - \mathbf{P}_{Y,2}^\dagger) \mathbf{U}_2 \right) \right|.
\end{aligned}$$

Since $\text{tr}(\boldsymbol{\Lambda}_2^2)/\text{tr}(\boldsymbol{\Lambda}_2) \in [\boldsymbol{\lambda}_p, \boldsymbol{\lambda}_{r+1}]$, we have $\|\boldsymbol{\Lambda}_2 - (\text{tr}(\boldsymbol{\Lambda}_2^2)/\text{tr}(\boldsymbol{\Lambda}_2))\mathbf{I}_{p-r}\| \leq$

$\boldsymbol{\lambda}_{r+1} - \boldsymbol{\lambda}_p$. Also note that the rank of the matrix $\mathbf{P}_Y - \mathbf{P}_{Y,1}^\dagger - \mathbf{P}_{Y,2}^\dagger$ is at

most $2n$. Therefore, von Neumann's trace theorem implies that

$$\begin{aligned}
& \left| \text{tr} \left(\left(\Lambda_2 - \frac{\text{tr}(\Lambda_2^2)}{\text{tr}(\Lambda_2)} \mathbf{I}_{p-r} \right) \mathbf{U}_2^\top \left(\mathbf{P}_Y - \mathbf{P}_{Y,1}^\dagger - \mathbf{P}_{Y,2}^\dagger \right) \mathbf{U}_2 \right) \right| \\
& \leq 2n(\lambda_{r+1} - \lambda_p) \left\| \mathbf{P}_Y - \mathbf{P}_{Y,1}^\dagger - \mathbf{P}_{Y,2}^\dagger \right\| \\
& \leq 2n(\lambda_{r+1} - \lambda_p) \left(\left\| \mathbf{P}_{Y,1} - \mathbf{P}_{Y,1}^\dagger \right\| + \left\| \mathbf{P}_{Y,2} - \mathbf{P}_{Y,2}^\dagger \right\| \right) \\
& = O_P \left(n(\lambda_{r+1} - \lambda_p) \left(\sqrt{\frac{\text{tr}(\Lambda_2) \lambda_1}{n \lambda_r^2}} + \sqrt{\frac{n \lambda_{r+1}}{\text{tr}(\Lambda_2)}} \right) \right),
\end{aligned} \tag{S3.19}$$

where the last equality follows from Proposition 3 and Proposition 4. Note that

$$\begin{aligned}
& \text{tr} \left(\left(\Lambda_2 - \frac{\text{tr}(\Lambda_2^2)}{\text{tr}(\Lambda_2)} \mathbf{I}_{p-r} \right) \mathbf{U}_2^\top \left(\mathbf{P}_{Y,1}^\dagger + \mathbf{P}_{Y,2}^\dagger \right) \mathbf{U}_2 \right) \\
& = \text{tr} \left(\left(\Lambda_2 - \frac{\text{tr}(\Lambda_2^2)}{\text{tr}(\Lambda_2)} \mathbf{I}_{p-r} \right) \mathbf{U}_2^\top \mathbf{P}_{Y,2}^\dagger \mathbf{U}_2 \right) \\
& = \frac{1}{\text{tr}(\Lambda_2)} \text{tr} \left(\tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \left(\Lambda_2^2 - \frac{\text{tr}(\Lambda_2^2)}{\text{tr}(\Lambda_2)} \Lambda_2 \right) \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \right)
\end{aligned}$$

It is straightforward to show that

$$\mathbb{E} \text{tr} \left(\tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \left(\Lambda_2^2 - \frac{\text{tr}(\Lambda_2^2)}{\text{tr}(\Lambda_2)} \Lambda_2 \right) \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \right) = 0,$$

and

$$\begin{aligned}
& \text{Var} \left(\text{tr} \left(\tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \left(\Lambda_2^2 - \frac{\text{tr}(\Lambda_2^2)}{\text{tr}(\Lambda_2)} \Lambda_2 \right) \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \right) \right) \\
& = 2(n-r) \text{tr} \left(\Lambda_2^2 - \frac{\text{tr}(\Lambda_2^2)}{\text{tr}(\Lambda_2)} \Lambda_2 \right)^2 \\
& \leq 2n \text{tr}(\Lambda_2^2) (\lambda_{r+1} - \lambda_p)^2 \\
& \leq 2n \lambda_{r+1} \text{tr}(\Lambda_2) (\lambda_{r+1} - \lambda_p)^2.
\end{aligned}$$

Thus,

$$\begin{aligned} & \text{tr} \left(\left(\mathbf{\Lambda}_2 - \frac{\text{tr}(\mathbf{\Lambda}_2^2)}{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_{p-r} \right) \mathbf{U}_2^\top \left(\mathbf{P}_{\mathbf{Y},1}^\dagger + \mathbf{P}_{\mathbf{Y},2}^\dagger \right) \mathbf{U}_2 \right) \\ &= O_P \left((\boldsymbol{\lambda}_{r+1} - \boldsymbol{\lambda}_p) \sqrt{\frac{n\boldsymbol{\lambda}_{r+1}}{\text{tr}(\mathbf{\Lambda}_2)}} \right). \end{aligned}$$

The last display, combined with (S3.19), leads to

$$\begin{aligned} & \text{tr} \left(\left(\mathbf{\Lambda}_2 - \frac{\text{tr}(\mathbf{\Lambda}_2^2)}{\text{tr}(\mathbf{\Lambda}_2)} \mathbf{I}_{p-r} \right) \mathbf{U}_2^\top \mathbf{P}_{\mathbf{Y}} \mathbf{U}_2 \right) \\ &= O_P \left(n(\boldsymbol{\lambda}_{r+1} - \boldsymbol{\lambda}_p) \left(\sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2)\boldsymbol{\lambda}_1}{n\boldsymbol{\lambda}_r^2}} + \sqrt{\frac{n\boldsymbol{\lambda}_{r+1}}{\text{tr}(\mathbf{\Lambda}_2)}} \right) \right). \end{aligned}$$

It then follows from (S3.17), (S3.18) and the last display that

$$\begin{aligned} & \text{tr} \left((\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \mathbf{U}_2 \mathbf{\Lambda}_2 \mathbf{U}_2^\top (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \right) \\ &= \text{tr}(\mathbf{\Lambda}_2) - \frac{n \text{tr}(\mathbf{\Lambda}_2^2)}{\text{tr}(\mathbf{\Lambda}_2)} \\ &+ O_P \left(n(\boldsymbol{\lambda}_{r+1} - \boldsymbol{\lambda}_p) \left(\sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2)\boldsymbol{\lambda}_1}{n\boldsymbol{\lambda}_r^2}} + \sqrt{\frac{n\boldsymbol{\lambda}_{r+1}}{\text{tr}(\mathbf{\Lambda}_2)}} \right) + r\boldsymbol{\lambda}_{r+1} \right). \end{aligned}$$

Then the conclusion follows from (S3.16) and the last display. \square

Lemma 9. *Suppose $p > n$, we have*

$$\begin{aligned} & \sum_{i=r+1}^p \lambda_i^2 ((\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \boldsymbol{\Sigma} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}})) \\ &= \text{tr}(\mathbf{\Lambda}_2^2) - \frac{n \text{tr}^2(\mathbf{\Lambda}_2^2)}{\text{tr}^2(\mathbf{\Lambda}_2)} + O_P \left(n\boldsymbol{\lambda}_{r+1}(\boldsymbol{\lambda}_{r+1} - \boldsymbol{\lambda}_p) + r\boldsymbol{\lambda}_{r+1}^2 \right). \end{aligned}$$

Proof. From (S3.14) and (S3.15), we have

$$\begin{aligned}
& \left| \sum_{i=r+1}^p \lambda_i^2 ((\mathbf{I}_p - \mathbf{P}_Y) \Sigma (\mathbf{I}_p - \mathbf{P}_Y)) - \text{tr} ((\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U}_2 \mathbf{\Lambda}_2 \mathbf{U}_2^\top (\mathbf{I}_p - \mathbf{P}_Y)) \right|^2 \\
& \leq r \lambda_1^2 ((\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U}_2 \mathbf{\Lambda}_2 \mathbf{U}_2^\top (\mathbf{I}_p - \mathbf{P}_Y)) \\
& \leq r \lambda_{r+1}^2.
\end{aligned} \tag{S3.20}$$

It is straightforward to show that

$$\begin{aligned}
& \text{tr} ((\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U}_2 \mathbf{\Lambda}_2 \mathbf{U}_2^\top (\mathbf{I}_p - \mathbf{P}_Y))^2 \\
& = \text{tr}(\mathbf{\Lambda}_2^2) - 2 \text{tr}(\mathbf{\Lambda}_2^2 \mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2) + \text{tr}(\mathbf{\Lambda}_2 \mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2)^2.
\end{aligned}$$

For the second term, we have

$$\begin{aligned}
& \left| \text{tr}(\mathbf{\Lambda}_2^2 \mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2) - \frac{\text{tr}^2(\mathbf{\Lambda}_2^2)}{\text{tr}^2(\mathbf{\Lambda}_2)} \text{tr}(\mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2) \right| \\
& = \left| \text{tr} \left(\left(\mathbf{\Lambda}_2^2 - \frac{\text{tr}^2(\mathbf{\Lambda}_2^2)}{\text{tr}^2(\mathbf{\Lambda}_2)} \mathbf{I}_{p-r} \right) \mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2 \right) \right| \\
& \leq n \lambda_{r+1} (\lambda_{r+1} - \lambda_p),
\end{aligned}$$

where the last equality follows from von Neumann's trace theorem. The

last display, combined with (S3.18), implies that

$$\text{tr}(\mathbf{\Lambda}_2^2 \mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2) = \frac{n \text{tr}^2(\mathbf{\Lambda}_2^2)}{\text{tr}^2(\mathbf{\Lambda}_2)} + O_P(n \lambda_{r+1} (\lambda_{r+1} - \lambda_p) + r \lambda_{r+1}^2).$$

For the third term, von Neumann's trace theorem implies that

$$\begin{aligned}
 & \left| \text{tr}(\Lambda_2 \mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2)^2 - \frac{\text{tr}^2(\Lambda_2^2)}{\text{tr}^2(\Lambda_2)} \text{tr}(\mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2)^2 \right| \\
 &= \left| \text{tr} \left(\left(\Lambda_2 - \frac{\text{tr}(\Lambda_2^2)}{\text{tr}(\Lambda_2)} \mathbf{I}_{p-r} \right) \mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2 \left(\Lambda_2 + \frac{\text{tr}(\Lambda_2^2)}{\text{tr}(\Lambda_2)} \mathbf{I}_{p-r} \right) \mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2 \right) \right| \\
 &\leq 2n\lambda_{r+1}(\lambda_{r+1} - \lambda_p).
 \end{aligned}$$

Note that

$$\begin{aligned}
 \text{tr}(\mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2)^2 &= \text{tr}(\mathbf{P}_Y - \mathbf{P}_Y \mathbf{U}_1 \mathbf{U}_1^\top)^2 \\
 &= n - 2 \text{tr}(\mathbf{P}_Y \mathbf{U}_1 \mathbf{U}_1^\top) + \text{tr}(\mathbf{P}_Y \mathbf{U}_1 \mathbf{U}_1^\top)^2 \\
 &= n + O_P(r).
 \end{aligned}$$

Therefore, the third term satisfies

$$\text{tr}(\Lambda_2 \mathbf{U}_2^\top \mathbf{P}_Y \mathbf{U}_2)^2 = \frac{n \text{tr}^2(\Lambda_2^2)}{\text{tr}^2(\Lambda_2)} + O_P(n\lambda_{r+1}(\lambda_{r+1} - \lambda_p) + r\lambda_{r+1}^2).$$

Thus,

$$\begin{aligned}
 & \text{tr}((\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U}_2 \Lambda_2 \mathbf{U}_2^\top (\mathbf{I}_p - \mathbf{P}_Y))^2 \\
 &= \text{tr}(\Lambda_2^2) - \frac{n \text{tr}^2(\Lambda_2^2)}{\text{tr}^2(\Lambda_2)} + O_P(n\lambda_{r+1}(\lambda_{r+1} - \lambda_p) + r\lambda_{r+1}^2).
 \end{aligned}$$

Then the conclusion follows from the last display and (S3.20). \square

Proof of Theorem 2. We have

$$\begin{aligned}
 & \mathbf{Z}^{\dagger\top} \Lambda^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U} \Lambda^{1/2} \mathbf{Z}^\dagger \\
 &= \sum_{i=1}^r \lambda_i ((\mathbf{I}_p - \mathbf{P}_Y) \Sigma (\mathbf{I}_p - \mathbf{P}_Y)) \eta_i \eta_i^\top + \sum_{i=r+1}^p \lambda_i ((\mathbf{I}_p - \mathbf{P}_Y) \Sigma (\mathbf{I}_p - \mathbf{P}_Y)) \eta_i \eta_i^\top.
 \end{aligned}$$

From Lemma 7, the first term satisfies

$$\sum_{i=1}^r \lambda_i((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y))\eta_i\eta_i^\top = (1 + o_P(r^{-1/2}))n^{-1} \text{tr}(\Lambda_2) \sum_{i=1}^r \eta_i\eta_i^\top.$$

Then

$$\begin{aligned} & \frac{\sum_{i=1}^r \lambda_i((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y))\eta_i\eta_i^\top - rn^{-1} \text{tr}(\Lambda_2)\mathbf{I}_{k-1}}{\sqrt{rn^{-1} \text{tr}(\Lambda_2)}} \\ &= \frac{\sum_{i=1}^r \eta_i\eta_i^\top - r\mathbf{I}_{k-1}}{\sqrt{r}} + o_P(1). \end{aligned} \quad (\text{S3.21})$$

Next we deal with the term $\sum_{i=r+1}^p \lambda_i((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y))\eta_i\eta_i^\top$. In

the current context, Lemma 8 and Lemma 9 imply that

$$\sum_{i=r+1}^p \lambda_i((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y)) = \text{tr}(\Lambda_2) - \frac{n \text{tr}(\Lambda_2^2)}{\text{tr}(\Lambda_2)} + o_P\left(\sqrt{\text{tr}(\Lambda_2^2)}\right), \quad (\text{S3.22})$$

$$\sum_{i=r+1}^p \lambda_i^2((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y)) = (1 + o_P(1)) \text{tr}(\Lambda_2^2). \quad (\text{S3.23})$$

By Weyl's inequality, we have

$$\begin{aligned} & \lambda_{r+1}((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y)) \\ &= \lambda_{r+1}\left((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{U}_1\Lambda_1\mathbf{U}_1^\top(\mathbf{I}_p - \mathbf{P}_Y) + (\mathbf{I}_p - \mathbf{P}_Y)\mathbf{U}_2\Lambda_2\mathbf{U}_2^\top(\mathbf{I}_p - \mathbf{P}_Y)\right) \\ &\leq \lambda_1\left((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{U}_2\Lambda_2\mathbf{U}_2^\top(\mathbf{I}_p - \mathbf{P}_Y)\right) \\ &\leq \lambda_{r+1}. \end{aligned}$$

The last display and (S3.22) imply that

$$\frac{\lambda_{r+1}^2\left((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{U}_2\Lambda_2\mathbf{U}_2^\top(\mathbf{I}_p - \mathbf{P}_Y)\right)}{\sum_{i=r+1}^p \lambda_i^2\left((\mathbf{I}_p - \mathbf{P}_Y)\mathbf{U}_2\Lambda_2\mathbf{U}_2^\top(\mathbf{I}_p - \mathbf{P}_Y)\right)} \leq \frac{\lambda_{r+1}^2}{(1 + o_P(1)) \text{tr}(\Lambda_2^2)} \xrightarrow{P} 0.$$

Then Lemma 2 implies that

$$\frac{1}{\sqrt{\sum_{i=r+1}^p \lambda_i^2((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y))}} \left(\sum_{i=r+1}^p \lambda_i((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y))\eta_i\eta_i^\top - \sum_{i=r+1}^p \lambda_i((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y))\mathbf{I}_{k-1} \right) \xrightarrow{\mathcal{L}} \mathbf{W}_{k-1}.$$

The last display, combined with (S3.22) and (S3.23), leads to

$$\frac{1}{\sqrt{\text{tr}(\Lambda_2^2)}} \left(\sum_{i=r+1}^p \lambda_i((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y))\eta_i\eta_i^\top - (\text{tr}(\Lambda_2) - n \text{tr}(\Lambda_2^2)/\text{tr}(\Lambda_2)) \mathbf{I}_{k-1} \right) \xrightarrow{\mathcal{L}} \mathbf{W}_{k-1}. \quad (\text{S3.24})$$

Note that $\sum_{i=1}^r \eta_i\eta_i^\top$ is independent of $\sum_{i=r+1}^p \lambda_i((\mathbf{I}_p - \mathbf{P}_Y)\Sigma(\mathbf{I}_p - \mathbf{P}_Y))\eta_i\eta_i^\top$.

Then (S3.21) and (S3.24) implies that

$$\begin{aligned} & \frac{\mathbf{Z}^\dagger \Lambda^{1/2} \mathbf{U}^\top (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U} \Lambda^{1/2} \mathbf{Z}^\dagger - ((1 + r/n) \text{tr}(\Lambda_2) - n \text{tr}(\Lambda_2^2)/\text{tr}(\Lambda_2)) \mathbf{I}_{k-1}}{\sqrt{rn^{-2} \text{tr}^2(\Lambda_2) + \text{tr}(\Lambda_2^2)}} \\ & \sim \frac{n^{-1} \text{tr}(\Lambda_2)}{\sqrt{rn^{-2} \text{tr}^2(\Lambda_2) + \text{tr}(\Lambda_2^2)}} (\mathbf{W}_{k-1}^* - r \mathbf{I}_{k-1}) + \frac{\sqrt{\text{tr}(\Lambda_2^2)}}{\sqrt{rn^{-2} \text{tr}^2(\Lambda_2) + \text{tr}(\Lambda_2^2)}} \mathbf{W}_{k-1} \\ & + o_P(1). \end{aligned} \quad (\text{S3.25})$$

This completes the proof of the first statement.

Now we prove the second statement. For the second term of (S3.1), we have $\mathbf{C}^\top \Theta^\top (\mathbf{I}_p - \mathbf{P}_Y) \Theta \mathbf{C} = \mathbf{C}^\top \Theta^\top \Theta \mathbf{C} - \mathbf{C}^\top \Theta^\top \mathbf{P}_Y \Theta \mathbf{C}$. We need to deal with $\mathbf{C}^\top \Theta^\top \mathbf{P}_Y \Theta \mathbf{C}$. Note that Proposition 3 implies that

$$\|\mathbf{P}_{Y,1} - \mathbf{U}_1 \mathbf{U}_1^\top\| \leq \|\mathbf{P}_{Y,1} - \mathbf{P}_{Y,1}^\dagger\| + 2\|\mathbf{Q}\| = o_P(1).$$

It follows from the last display and Proposition 4 that

$$\begin{aligned}
& \left\| \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{P}_Y \boldsymbol{\Theta} \mathbf{C} - \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{U}_1 \mathbf{U}_1^\top \boldsymbol{\Theta} \mathbf{C} - \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{P}_{Y,2}^\dagger \boldsymbol{\Theta} \mathbf{C} \right\| \\
& \leq \left\| \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{P}_{Y,1} \boldsymbol{\Theta} \mathbf{C} - \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{U}_1 \mathbf{U}_1^\top \boldsymbol{\Theta} \mathbf{C} \right\| \\
& \quad + \left\| \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{P}_{Y,2} \boldsymbol{\Theta} \mathbf{C} - \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{P}_{Y,2}^\dagger \boldsymbol{\Theta} \mathbf{C} \right\| \\
& \leq \left\| \mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Theta} \mathbf{C} \right\| \left(\left\| \mathbf{P}_{Y,1} - \mathbf{U}_1 \mathbf{U}_1^\top \right\| + \left\| \mathbf{P}_{Y,2} - \mathbf{P}_{Y,2}^\dagger \right\| \right) \\
& =_{o_P} \left(\sqrt{rn^{-2} \text{tr}^2(\boldsymbol{\Lambda}_2) + \text{tr}(\boldsymbol{\Lambda}_2^2)} \right).
\end{aligned}$$

We have

$$\mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{P}_{Y,2}^\dagger \boldsymbol{\Theta} \mathbf{C} = (\text{tr}(\boldsymbol{\Lambda}_2))^{-1} \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{U}_2 \boldsymbol{\Lambda}_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} \tilde{\mathbf{V}}_{\mathbf{Z}_1}^\top \mathbf{Z}_2^\top \boldsymbol{\Lambda}_2^{1/2} \mathbf{U}_2^\top \boldsymbol{\Theta} \mathbf{C}.$$

Note that $\mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1}$ is a $(p-r) \times (n-r)$ matrix with iid $\mathcal{N}(0, 1)$ entries. Then the columns of $\mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{U}_2 \boldsymbol{\Lambda}_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1}$ are iid $\mathcal{N}_{k-1}(\mathbf{0}_{k-1}, \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{U}_2 \boldsymbol{\Lambda}_2 \mathbf{U}_2^\top \boldsymbol{\Theta} \mathbf{C})$ random vectors. Write $\mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{U}_2 \boldsymbol{\Lambda}_2^{1/2} \mathbf{Z}_2 \tilde{\mathbf{V}}_{\mathbf{Z}_1} = (\mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{U}_2 \boldsymbol{\Lambda}_2 \mathbf{U}_2^\top \boldsymbol{\Theta} \mathbf{C})^{1/2} \mathbf{Z}^*$, where \mathbf{Z}^* is a $(k-1) \times (n-r)$ random matrix with iid $\mathcal{N}(0, 1)$ entries. Then

$$\begin{aligned}
& \left\| \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{P}_{Y,2}^\dagger \boldsymbol{\Theta} \mathbf{C} - \frac{n}{\text{tr}(\boldsymbol{\Lambda}_2)} \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{U}_2 \boldsymbol{\Lambda}_2 \mathbf{U}_2^\top \boldsymbol{\Theta} \mathbf{C} \right\| \\
& \leq \frac{n}{\text{tr}(\boldsymbol{\Lambda}_2)} \left\| \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{U}_2 \boldsymbol{\Lambda}_2 \mathbf{U}_2^\top \boldsymbol{\Theta} \mathbf{C} \right\| \left\| n^{-1} \mathbf{Z}^* \mathbf{Z}^{*\top} - \mathbf{I}_{k-1} \right\| \\
& =_{o_P} \left(\sqrt{rn^{-2} \text{tr}^2(\boldsymbol{\Lambda}_2) + \text{tr}(\boldsymbol{\Lambda}_2^2)} \right),
\end{aligned}$$

where the last equality follows from the law of large numbers, the local

alternative condition and the condition $n\lambda_{r+1}/\text{tr}(\Lambda_2) \rightarrow 0$. But

$$\begin{aligned} \frac{n}{\text{tr}(\Lambda_2)} \|\mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{U}_2 \Lambda_2 \mathbf{U}_2^\top \boldsymbol{\Theta} \mathbf{C}\| &\leq \frac{n\lambda_2}{\text{tr}(\Lambda_2)} \|\mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Theta} \mathbf{C}\| \\ &= o_P \left(\sqrt{rn^{-2} \text{tr}^2(\Lambda_2) + \text{tr}(\Lambda_2^2)} \right). \end{aligned}$$

Hence $\|\mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{P}_{\mathbf{Y},2}^\dagger \boldsymbol{\Theta} \mathbf{C}\| = o_P \left(\sqrt{rn^{-2} \text{tr}^2(\Lambda_2) + \text{tr}(\Lambda_2^2)} \right)$. Consequently,
 $\|\mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{P}_{\mathbf{Y}} \boldsymbol{\Theta} \mathbf{C} - \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{U}_1 \mathbf{U}_1^\top \boldsymbol{\Theta} \mathbf{C}\| = o_P \left(\sqrt{rn^{-2} \text{tr}^2(\Lambda_2) + \text{tr}(\Lambda_2^2)} \right)$. Thus,

the second term of (S3.1) satisfies

$$\begin{aligned} &\|\mathbf{C}^\top \boldsymbol{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \boldsymbol{\Theta} \mathbf{C} - \mathbf{C}^\top \boldsymbol{\Theta}^\top \mathbf{U}_2 \mathbf{U}_2^\top \boldsymbol{\Theta} \mathbf{C}\| \\ &= o_P \left(\sqrt{rn^{-2} \text{tr}^2(\Lambda_2) + \text{tr}(\Lambda_2^2)} \right). \end{aligned} \tag{S3.26}$$

Next we consider the cross term of (S3.1). Note that

$$\begin{aligned} &\mathbb{E}[\|\mathbf{C}^\top \boldsymbol{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \mathbf{U} \Lambda^{1/2} \mathbf{Z}^\dagger\|_F^2 | \mathbf{Y}] \\ &= (k-1) \text{tr}(\mathbf{C}^\top \boldsymbol{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \boldsymbol{\Sigma} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \boldsymbol{\Theta} \mathbf{C}) \\ &\leq (k-1) \lambda_1((\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \boldsymbol{\Sigma} (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}})) \text{tr}(\mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Theta} \mathbf{C}) \\ &= O_P(n^{-1} \text{tr}(\Lambda_2) \|\mathbf{C}^\top \boldsymbol{\Theta}^\top \boldsymbol{\Theta} \mathbf{C}\|), \end{aligned}$$

where the last equality follows from Lemma 7. Under the condition $r \rightarrow \infty$

or $\text{tr}(\Lambda_2)/(n\sqrt{\text{tr}(\Lambda_2^2)}) \rightarrow 0$, we have $n^{-1} \text{tr}(\Lambda_2) = o_P \left(\sqrt{rn^{-2} \text{tr}^2(\Lambda_2) + \text{tr}(\Lambda_2^2)} \right)$.

Therefore,

$$\|\mathbf{C}^\top \boldsymbol{\Theta}^\top (\mathbf{I}_p - \mathbf{P}_{\mathbf{Y}}) \mathbf{U} \Lambda^{1/2} \mathbf{Z}^\dagger\| = o_P \left(\sqrt{rn^{-2} \text{tr}^2(\Lambda_2) + \text{tr}(\Lambda_2^2)} \right).$$

It follows from the last display, (S3.26) and Weyl's inequality that

$$\begin{aligned} & |T(\mathbf{X}) - \lambda_1 (\mathbf{Z}^\dagger{}^\top \mathbf{\Lambda}^{1/2} \mathbf{U}^\dagger (\mathbf{I}_p - \mathbf{P}_Y) \mathbf{U} \mathbf{\Lambda}^{1/2} \mathbf{Z}^\dagger + \mathbf{C}^\top \mathbf{\Theta}^\top \mathbf{U}_2 \mathbf{U}_2^\top \mathbf{\Theta} \mathbf{C})| \\ &= o_P \left(\sqrt{rn^{-2} \text{tr}^2(\mathbf{\Lambda}_2) + \text{tr}(\mathbf{\Lambda}_2^2)} \right). \end{aligned}$$

Then the second statement follows from the last display and (S3.25). \square

Proof of Corollary 2. From Proposition 2, we have

$$rn^{-2}(\widehat{\text{tr}(\mathbf{\Lambda}_2)})^2 + \widehat{\text{tr}(\mathbf{\Lambda}_2^2)} = (1 + o_P(1))(rn^{-2} \text{tr}^2(\mathbf{\Lambda}_2) + \text{tr}(\mathbf{\Lambda}_2^2)),$$

and

$$\begin{aligned} & (1 + r/n) \widehat{\text{tr}(\mathbf{\Lambda}_2)} - n \widehat{\text{tr}(\mathbf{\Lambda}_2^2)} / \widehat{\text{tr}(\mathbf{\Lambda}_2)} \\ &= (1 + r/n) \text{tr}(\mathbf{\Lambda}_2) + O_P \left(r \sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2^2)}{n}} + r \lambda_{r+1} \right) \\ & \quad - \frac{n \text{tr}(\mathbf{\Lambda}_2^2) (1 + O_P(r/n + r \lambda_{r+1}^2 / \text{tr}(\mathbf{\Lambda}_2^2)))}{\text{tr}(\mathbf{\Lambda}_2) (1 + O_P(r \sqrt{\text{tr}(\mathbf{\Lambda}_2^2)/n \text{tr}^2(\mathbf{\Lambda}_2)} + r \lambda_{r+1} / \text{tr}(\mathbf{\Lambda}_2)))} \\ &= (1 + r/n) \text{tr}(\mathbf{\Lambda}_2) + O_P \left(r \sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2^2)}{n}} + r \lambda_{r+1} \right) \\ & \quad - \frac{n \text{tr}(\mathbf{\Lambda}_2^2)}{\text{tr}(\mathbf{\Lambda}_2)} \left(1 + O_P \left(\frac{r}{n} + \frac{r \lambda_{r+1}^2}{\text{tr}(\mathbf{\Lambda}_2^2)} + r \sqrt{\frac{\text{tr}(\mathbf{\Lambda}_2^2)}{n \text{tr}^2(\mathbf{\Lambda}_2)}} + \frac{r \lambda_{r+1}}{\text{tr}(\mathbf{\Lambda}_2)} \right) \right) \\ &= (1 + r/n) \text{tr}(\mathbf{\Lambda}_2) - \frac{n \text{tr}(\mathbf{\Lambda}_2^2)}{\text{tr}(\mathbf{\Lambda}_2)} + o_P \left(\sqrt{\text{tr}(\mathbf{\Lambda}_2^2)} \right). \end{aligned}$$

Therefore,

$$Q_2 = \frac{T(\mathbf{X}) - ((1 + r/n) \text{tr}(\mathbf{\Lambda}_2) - n \text{tr}(\mathbf{\Lambda}_2^2) / \text{tr}(\mathbf{\Lambda}_2))}{\sqrt{rn^{-2} \text{tr}^2(\mathbf{\Lambda}_2) + \text{tr}(\mathbf{\Lambda}_2^2)}} + o_P(1).$$

On the other hand, it is not hard to see that the ratio consistency of $\widehat{\text{tr}(\boldsymbol{\Lambda}_2)}$ and $\widehat{\text{tr}(\boldsymbol{\Lambda}_2^2)}$ imply $F_2^{-1}(1 - \alpha; \widehat{\text{tr}(\boldsymbol{\Lambda}_2)}, \widehat{\text{tr}(\boldsymbol{\Lambda}_2^2)}) = F_2^{-1}(1 - \alpha; \text{tr}(\boldsymbol{\Lambda}_2), \text{tr}(\boldsymbol{\Lambda}_2^2)) + o_P(1)$. Then the conclusion follows from Theorem 2 and Slutsky's theorem.

□

Proof of Proposition 5. Under the conditions of Theorem 1, we have $n\boldsymbol{\lambda}_1 / \text{tr}(\boldsymbol{\Sigma}) \rightarrow 0$. From Lemma 5 and Weyl's inequality, we have

$$\begin{aligned} \lambda_1(\hat{\boldsymbol{\Sigma}}) &= n^{-1} \lambda_1(\mathbf{Z}^\top \boldsymbol{\Lambda} \mathbf{Z}) \\ &= n^{-1} \text{tr}(\boldsymbol{\Sigma}) + O_P\left(\sqrt{\frac{\text{tr}(\boldsymbol{\Sigma}^2)}{n}} + \boldsymbol{\lambda}_1\right) \\ &= (1 + o_P(1))n^{-1} \text{tr}(\boldsymbol{\Sigma}). \end{aligned}$$

From the proof of Corollary 1, we have $\text{tr}(\hat{\boldsymbol{\Sigma}}) = (1 + o_P(1)) \text{tr}(\boldsymbol{\Sigma})$. Therefore,

$$\frac{n\lambda_1(\hat{\boldsymbol{\Sigma}})}{\text{tr}(\hat{\boldsymbol{\Sigma}})} \xrightarrow{P} 1.$$

This completes the proof of (i).

Now we prove (ii). Under the conditions of Theorem 2, Proposition 1 implies that

$$\begin{aligned} \frac{n\lambda_1(\hat{\boldsymbol{\Sigma}})}{\text{tr}(\hat{\boldsymbol{\Sigma}})} &= \frac{n\lambda_1(\hat{\boldsymbol{\Sigma}})}{\sum_{i=1}^r \lambda_i(\hat{\boldsymbol{\Sigma}}) + \sum_{i=r+1}^n \lambda_i(\hat{\boldsymbol{\Sigma}})} \\ &= (1 + o_P(1)) \frac{n\boldsymbol{\lambda}_1 + \text{tr}(\boldsymbol{\Lambda}_2)}{\sum_{i=1}^r \boldsymbol{\lambda}_i + \text{tr}(\boldsymbol{\Lambda}_2)} \\ &\geq (1 + o_P(1)) \frac{n\boldsymbol{\lambda}_1}{r\boldsymbol{\lambda}_1 + \text{tr}(\boldsymbol{\Lambda}_2)} \xrightarrow{P} \infty. \end{aligned}$$

It follows that

$$\Pr \left(\frac{n\lambda_1(\hat{\Sigma})}{\text{tr}(\hat{\Sigma})} < \tau \right) \rightarrow 0.$$

Next we consider the consistency of \hat{r} . Note that

$$\{\hat{r} = r\} = \left\{ \frac{n\lambda_{i+1}(\hat{\Sigma})}{\sum_{j=i+1}^n \lambda_j(\hat{\Sigma})} \geq \tau, i = 1, \dots, r-1 \right\} \cap \left\{ \frac{n\lambda_{r+1}(\hat{\Sigma})}{\sum_{j=r+1}^n \lambda_j(\hat{\Sigma})} < \tau \right\}.$$

But Proposition 1 implies that uniformly for $i = 1, \dots, r-1$,

$$\begin{aligned} \frac{n\lambda_{i+1}(\hat{\Sigma})}{\sum_{j=i+1}^n \lambda_j(\hat{\Sigma})} &\geq \frac{n\lambda_{i+1}(\hat{\Sigma})}{(r-i)\lambda_{i+1}(\hat{\Sigma}) + \sum_{j=r+1}^n \lambda_j(\hat{\Sigma})} \\ &= (1 + o_P(1)) \frac{n\lambda_{i+1} + \text{tr}(\Lambda_2)}{(r-i)\lambda_{i+1} + (1-i/n)\text{tr}(\Lambda_2)} \xrightarrow{P} \infty. \end{aligned}$$

Thus, we only need to prove that

$$\Pr \left(\frac{n\lambda_{r+1}(\hat{\Sigma})}{\sum_{j=r+1}^n \lambda_j(\hat{\Sigma})} < \tau \right) \rightarrow 1.$$

Weyl' inequality implies that $n\lambda_{r+1}(\hat{\Sigma}) = \lambda_{r+1}(\mathbf{Z}_1^\top \Lambda_1 \mathbf{Z}_1 + \mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2) \leq \lambda_1(\mathbf{Z}_2^\top \Lambda_2 \mathbf{Z}_2)$. Then using Lemma 5, we have $n\lambda_{r+1}(\hat{\Sigma}) \leq (1 + o_P(1)) \text{tr}(\Lambda_2)$.

Thus,

$$\frac{n\lambda_{r+1}(\hat{\Sigma})}{\sum_{j=r+1}^n \lambda_j(\hat{\Sigma})} \leq (1 + o_P(1)).$$

This completes the proof. □

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