

Spectral Properties of Rescaled Sample Correlation Matrices

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Supplementary Material

The Supplementary Material contains detailed proofs of some theoretical results.

1. Some assumptions

Assumption A. Samples satisfy the following independent component structure

$$\mathbf{y}_i = \boldsymbol{\mu} + \boldsymbol{\Gamma}\mathbf{x}_i, i = 1, \dots, n,$$

where $E\mathbf{y}_i = \boldsymbol{\mu}$, $\boldsymbol{\Gamma} = [\text{diag}(\boldsymbol{\Sigma})]^{1/2}\mathbf{R}^{1/2}$ and $\mathbf{x}_i = (x_{1i}, \dots, x_{pi})^T$.

Assumption B. Assume that $\{x_{ji}, j = 1, \dots, p, i = 1, \dots, n\}$ are independent and identically distributed (*i.i.d.*) with

$$E x_{ji} = 0, E x_{ji}^2 = 1, E(|x_{ji}|^4 (\log(|x_{ji}|)^{2+2\epsilon})) < \infty$$

for a small positive number $\epsilon > 0$.

Assumption C. The convergence regime is $\rho_n = p/n \rightarrow \rho \in (0, +\infty)$.

Assumption D. The functions g_1, \dots, g_K are analytic functions in a domain containing the support set $[a_\rho, b_\rho]$ of Marčenko-Pastur law in (2.1).

Assumption E. Assume that $\{x_{ji}, j = 1, \dots, p, i = 1, \dots, n\}$ are independent and identically distributed with

$$\mathbb{E}x_{ji} = 0, \mathbb{E}x_{ji}^2 = 1, \mathbb{E}x_{ji}^4 = \beta_x + 3 + o(1), \mathbb{E}(|x_{ji}|^4(\log(|x_{ji}|)^{2+2\epsilon})) < \infty.$$

Assumption F. Assume

$$\begin{aligned} a_g &= \lim_{p \rightarrow \infty} p^{-1} \sum_{k=1}^p \sum_{h=1}^p g_{kh}^3 \mathbf{e}_h^T \mathbf{R}^{-1/2} \mathbf{e}_k, & a_{\mathbf{R}} &= \lim_{p \rightarrow \infty} p^{-1} \sum_{k,\ell=1}^p \mathbf{e}_k^T \mathbf{R}^{-1} \mathbf{e}_\ell r_{k\ell}^3, \\ c_g &= \lim_{p \rightarrow \infty} p^{-1} \sum_{k=1}^p \sum_{h=1}^p g_{kh}^4, & d_{\mathbf{R}} &= \lim_{p \rightarrow \infty} p^{-1} \text{tr}(\mathbf{R}^2), \\ h_{\mathbf{R}} &= \lim_{p \rightarrow \infty} p^{-1} \sum_{k,\ell=1}^p \mathbf{e}_k^T \mathbf{R}^{-1} \mathbf{e}_\ell r_{k\ell} \sum_{h=1}^p g_{\ell h}^2 g_{kh}^2, \end{aligned}$$

where $\mathbf{R}^{1/2} = (g_{kh})$, $\mathbf{R} = (r_{kh})$ and \mathbf{e}_j is the j th column of $p \times p$ identity matrix for $j = 1, \dots, p$.

2. Limiting spectral distribution

Theorem 2.1. Under Assumptions A-B-C, the empirical spectral distribution $F_n(x)$ of $\mathbf{R}^{-1} \widehat{\mathbf{R}}_n$ converges almost surely to the Marčenko-Pastur law with the index ρ as follows

$$f_\rho(x) = \begin{cases} \frac{1}{2\pi\rho x} \sqrt{(b_\rho - x)(x - a_\rho)}, & \text{if } a_\rho \leq x \leq b_\rho, \\ 0, & \text{otherwise,} \end{cases} \quad (2.1)$$

where $a_\rho = (1 - \sqrt{\rho})^2$, and $b_\rho = (1 + \sqrt{\rho})^2$.

3. Central limit theorem

Theorem 3.1. Under Assumptions A-C-D-E-F, the random vector $(W(g_1), \dots, W(g_K))$ weakly converges to a multivariate Gaussian random vector $(X_{g_1}, \dots, X_{g_K})$ with

$$\mathbb{E}X_{g_\ell} = -\frac{1}{2\pi\mathbf{i}} \oint_{\mathcal{C}} g_\ell(z) \mathbb{E}M(z) dz$$

and

$$\text{Cov}(X_{g_{\ell_1}}, X_{g_{\ell_2}}) = -\frac{1}{4\pi^2} \oint_{\mathcal{C}_1} \oint_{\mathcal{C}_2} g_{\ell_1}(z_1) g_{\ell_2}(z_2) \text{Cov}(M(z_1), M(z_2)) dz_2 dz_1$$

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for $\ell, \ell_1, \ell_2 \in \{1, \dots, K\}$, where $\mathcal{C}, \mathcal{C}_1$, and \mathcal{C}_2 are three contours including $[a_\rho, b_\rho]$, \mathcal{C}_1 and \mathcal{C}_2 are non-overlapping, the contour integral \oint is anticlockwise, and $EM(z)$ and $\text{Cov}(M(z_1), M(z_2))$ are calculated as follows:

$$\begin{aligned} EM(z) &= \frac{\rho \underline{s}^3(z)[1 + \underline{s}(z)]^{-3}}{[1 - \rho \underline{s}^2(z)(1 + \underline{s}(z))^{-2}]^2} + \frac{\beta_x \rho \underline{s}(z) \underline{s}'(z)}{[1 + \underline{s}(z)]^3} \\ &\quad - \frac{\underline{s}'(z)}{[1 + \underline{s}(z)]^2} \frac{[10 - 2a_{\mathbf{R}} + \beta_x(4a_g + c_g - h_{\mathbf{R}})]\rho}{4} \\ &\quad + \frac{\underline{s}'(z)}{[1 + \underline{s}(z)]^3} \frac{[6 - 2a_{\mathbf{R}} + \beta_x(4a_g - c_g - h_{\mathbf{R}})]\rho}{2}, \end{aligned} \quad (3.2)$$

and

$$\begin{aligned} \text{Cov}(M(z_1), M(z_2)) &= 2 \left[\frac{\underline{s}'(z_1) \underline{s}'(z_2)}{(\underline{s}(z_2) - \underline{s}(z_1))^2} - \frac{1}{(z_2 - z_1)^2} \right] \\ &\quad + 2(d_{\mathbf{R}} - 2)\rho \frac{\underline{s}'(z_1) \underline{s}'(z_2)}{(1 + \underline{s}(z_1))^2 (1 + \underline{s}(z_2))^2}, \end{aligned} \quad (3.3)$$

where $'$ is the derivative notation, and $\underline{s}(z)$ is the unique solution to $z = -\underline{s}^{-1}(z) + \rho(1 + \underline{s}(z))^{-1}$, which leads to $\underline{s}'(z) = \underline{s}^2(z) / \{1 - \rho \underline{s}^2(z)[1 + \underline{s}(z)]^{-2}\}$.

4. Proofs of Theorems 2.1 and 3.1

This part is devoted to the proofs of our main theorems. The main strategy of proving Theorems 2.1 and 3.1 relies on the Stieltjes transformation method. Section 4.2 lists notations that will be used below. Section 4.3 provides the proof of Theorem 2.1. Finally, Section 4.4 presents the proof of Theorem 3.1.

4.1 Remove the sample mean

First of all, let

$$\ddot{\mathbf{S}}_n = \frac{1}{n} \sum_{i=1}^n (\mathbf{y}_i - \boldsymbol{\mu})(\mathbf{y}_i - \boldsymbol{\mu})^T,$$

where $\boldsymbol{\mu} = E\mathbf{y}_i$. (1.1) shows that

$$\mathbf{S}_n = (n-1)^{-1} \sum_{i=1}^n (\mathbf{y}_i - \bar{\mathbf{y}})(\mathbf{y}_i - \bar{\mathbf{y}})^T, \quad \hat{\mathbf{R}} = [\text{diag}(\mathbf{S}_n)]^{-1/2} \text{diag}(\mathbf{S}_n) [\text{diag}(\mathbf{S}_n)]^{-1/2}.$$

We will show that

- Studying the limiting spectral distribution (*LSD*) of $\widehat{\mathbf{R}}$ is equivalent to studying the LSD of

$$[\text{diag}(\check{\mathbf{S}}_n)]^{-1/2} \text{diag}(\check{\mathbf{S}}_n) [\text{diag}(\check{\mathbf{S}}_n)]^{-1/2}.$$

- Studying the CLT of LSS of $\widehat{\mathbf{R}}$ is equivalent to studying the CLT of LSS of

$$[\text{diag}(\mathbf{S})]^{-1/2} \text{diag}(\mathbf{S}) [\text{diag}(\mathbf{S})]^{-1/2},$$

where $\mathbf{S} = n^{-1} \sum_{i=1}^n \mathbf{R}^{1/2} \mathbf{x}_i \mathbf{x}_i^T \mathbf{R}^{1/2}$.

- Notice that $\bar{\mathbf{y}} \bar{\mathbf{y}}^T$ has rank one. Combining with Theorem A.43 in Bai and Silverstein (2010), we know that for large n , the difference between ESD of

$$\mathbf{S}_n = (n-1)^{-1} \sum_{i=1}^n (\mathbf{y}_i - \bar{\mathbf{y}})(\mathbf{y}_i - \bar{\mathbf{y}})^T$$

in (1.1) and ESD of $\check{\mathbf{S}}_n$ is neglectable.

Thus, in the proof of Theorem 2.1 in the next section, we only consider $\check{\mathbf{S}}_n$ instead of \mathbf{S}_n .

- Moreover, by the proof of substitution principle for CLT of LSS of sample covariance matrix in Zheng, Bai and Yao (2015), we know that the CLT of LSS of \mathbf{S}_n is obtained only through replacing $y_n = p/n$ by $y_{n-1} = p/(n-1)$ in the CLT of LSS of $\check{\mathbf{S}}_n$.

Due to $E(\mathbf{y}_i - \boldsymbol{\mu}) = \mathbf{0}$, without loss of generality, we assume $\boldsymbol{\mu} = \mathbf{0}$ (that is, $E\mathbf{y}_i = \mathbf{0}$) and use

$$\widehat{\mathbf{S}} = n^{-1} \sum_{i=1}^n \mathbf{y}_i \mathbf{y}_i^T$$

to replace $\check{\mathbf{S}}_n$.

- By Assumption A, we have

$$\hat{\mathbf{S}} = [\text{diag}(\boldsymbol{\Sigma})]^{1/2} n^{-1} \sum_{i=1}^n \mathbf{R}^{1/2} \mathbf{x}_i \mathbf{x}_i^T \mathbf{R}^{1/2} [\text{diag}(\boldsymbol{\Sigma})]^{1/2}.$$

That is, we have

$$\text{diag}(\hat{\mathbf{S}}) = \text{diag} \left(n^{-1} \sum_{i=1}^n \mathbf{y}_i \mathbf{y}_i^T \right) = [\text{diag}(\boldsymbol{\Sigma})]^{1/2} \text{diag} \left(n^{-1} \sum_{i=1}^n \mathbf{R}^{1/2} \mathbf{x}_i \mathbf{x}_i^T \mathbf{R}^{1/2} \right) [\text{diag}(\boldsymbol{\Sigma})]^{1/2},$$

and

$$[\text{diag}(\hat{\mathbf{S}})]^{-1/2} = \text{diag} \left(n^{-1} \sum_{i=1}^n \mathbf{R}^{1/2} \mathbf{x}_i \mathbf{x}_i^T \mathbf{R}^{1/2} \right)^{-1/2} [\text{diag}(\boldsymbol{\Sigma})]^{-1/2}.$$

Let $\mathbf{S} = n^{-1} \sum_{i=1}^n \mathbf{R}^{1/2} \mathbf{x}_i \mathbf{x}_i^T \mathbf{R}^{1/2}$, and we have

$$\begin{aligned} \hat{\mathbf{R}} &= [\text{diag}(\hat{\mathbf{S}})]^{-1/2} \text{diag}(\hat{\mathbf{S}}) [\text{diag}(\hat{\mathbf{S}})]^{-1/2} \\ &= \text{diag} \left(n^{-1} \sum_{i=1}^n \mathbf{R}^{1/2} \mathbf{x}_i \mathbf{x}_i^T \mathbf{R}^{1/2} \right)^{-1/2} n^{-1} \sum_{i=1}^n \mathbf{R}^{1/2} \mathbf{x}_i \mathbf{x}_i^T \mathbf{R}^{1/2} \text{diag} \left(n^{-1} \sum_{i=1}^n \mathbf{R}^{1/2} \mathbf{x}_i \mathbf{x}_i^T \mathbf{R}^{1/2} \right)^{-1/2} \\ &= [\text{diag}(\mathbf{S})]^{-1/2} \mathbf{S} [\text{diag}(\mathbf{S})]^{-1/2}, \end{aligned} \tag{4.1}$$

where $\text{Var}(\mathbf{R}^{1/2} \mathbf{x}_i) = \mathbf{R}$.

4.2 Some notation and preliminary results

Let $\mathbf{G} = \mathbf{R}^{1/2} = (g_{kh})$ and

$$\begin{aligned} \mathbf{r}_j &= \mathbf{x}_j / \sqrt{n}, \quad \mathbf{A}(z) = \sum_{i=1}^n \mathbf{r}_i \mathbf{r}_i^T - z \mathbf{I}_p, \quad \mathbf{A}_j(z) = \mathbf{A}(z) - \mathbf{r}_j \mathbf{r}_j^T, \\ \mathbf{A}_{jk}(z) &= \mathbf{A}_j(z) - \mathbf{r}_k \mathbf{r}_k^T, \quad \mathbf{A}_{jk\ell}(z) = \mathbf{A}_{jk}(z) - \mathbf{r}_\ell \mathbf{r}_\ell^T, \quad \text{for } 1 \leq j, k, \ell \leq n. \end{aligned}$$

Let $\check{\mathbf{A}}_j(z) = \sum_{i < j} \mathbf{r}_i \mathbf{r}_i^T + \sum_{i > j} \check{\mathbf{r}}_i \check{\mathbf{r}}_i^T - z \mathbf{I}$, where $\check{\mathbf{r}}_{j+1}, \dots, \check{\mathbf{r}}_n$ are independent copies of $\mathbf{r}_{j+1}, \dots, \mathbf{r}_n$,

and let \mathbf{E}_j denote the conditional expectation given $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_j$. Moreover, define

$$\begin{aligned} b_j(z) &= [1 + n^{-1} \text{tr}(\mathbf{A}_j^{-1}(z))]^{-1}, \quad b_{ij}(z) = [1 + n^{-1} \text{tr}(\mathbf{A}_{ij}^{-1}(z))]^{-1} \\ b_{ijk}(z) &= [1 + n^{-1} \text{tr}(\mathbf{A}_{ijk}^{-1}(z))]^{-1}, \quad \check{b}_j(z) = [1 + n^{-1} \text{tr}(\check{\mathbf{A}}_j^{-1}(z))]^{-1}, \end{aligned}$$

$$\begin{aligned}
 b(z) &= [1 + n^{-1} \text{Etr}(\mathbf{A}^{-1}(z))]^{-1}, & \beta_{i(j)}(z) &= (1 + \mathbf{r}_i^T \mathbf{A}_{ij}^{-1} \mathbf{r}_i)^{-1}, \\
 \beta_{i(jk)}(z) &= (1 + \mathbf{r}_i^T \mathbf{A}_{ijk}^{-1} \mathbf{r}_i)^{-1}, & \beta_j(z) &= [1 + \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{r}_j]^{-1}, \\
 \widehat{\gamma}_j(z) &= \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{r}_j - n^{-1} \text{tr} \mathbf{A}_j^{-1}(z), & \widehat{\gamma}_{i(j)}(z) &= \mathbf{r}_i^T \mathbf{A}_{ij}^{-1}(z) \mathbf{r}_i - n^{-1} \text{tr} \mathbf{A}_{ij}^{-1}(z), \\
 \widehat{\gamma}_{i(jk)}(z) &= \mathbf{r}_i^T \mathbf{A}_{ijk}^{-1}(z) \mathbf{r}_i - n^{-1} \text{tr} \mathbf{A}_{ijk}^{-1}(z).
 \end{aligned}$$

By eqn.(3.4) of Bai et al. (1998), we have

$$\max(|b_j(z)|, |b_{ij}(z)|, |b(z)|, |\beta_{i(j)}(z)|, |\beta_{i(jk)}(z)|, |\beta_j(z)|) \leq |z| / \text{Im}(z), \quad (4.1)$$

where $\text{Im}(z)$ stands for the imaginary part of z . With the notations above, $\mathbf{A}^{-1}(z)$ can be decomposed as

$$\mathbf{A}^{-1}(z) = \mathbf{A}_j^{-1}(z) - \beta_j(z) \mathbf{A}_j^{-1}(z) \mathbf{r}_j \mathbf{r}_j^T \mathbf{A}_j^{-1}(z), \quad (4.2)$$

see, e.g., the expression above eqn.(2.2) in Bai and Silverstein (2004); and $\beta_j(z)$ can be further written as

$$\beta_j(z) = b_j(z) - \beta_j(z) b_j(z) \widehat{\gamma}_j(z) = b_j(z) - b_j^2(z) \widehat{\gamma}_j(z) + \beta_j(z) b_j^2(z) \widehat{\gamma}_j^2(z), \quad (4.3)$$

and

$$\beta_{i(jk)}(z) = b_{ijk}(z) - \beta_{i(jk)}(z) b_{ijk}(z) \widehat{\gamma}_{i(jk)}(z). \quad (4.4)$$

(see line 9 on Page 569 of Bai and Silverstein (2004)). The matrix $\mathbf{A}_1^{-1}(z)$ can be further decomposed as

$$\mathbf{A}_1^{-1}(z) = \mathbf{B}_1(z) + \mathbf{B}_2(z) + \mathbf{B}_3(z) + \mathbf{B}_4(z), \quad (4.5)$$

where

$$\begin{aligned}
\mathbf{B}_1(z) &= -(z - (n-1)/n \cdot b(z))^{-1} \mathbf{I}_p, \\
\mathbf{B}_2(z) &= b(z) \sum_{i \neq 1} (z - (n-1)/n \cdot b(z))^{-1} (\mathbf{r}_i \mathbf{r}_i^T - n^{-1} \mathbf{I}_p) \mathbf{A}_{i1}^{-1}(z), \\
\mathbf{B}_3(z) &= \sum_{i \neq 1} (\beta_{i(1)}(z) - b(z)) (z - (n-1)/n \cdot b(z))^{-1} \mathbf{r}_i \mathbf{r}_i^T \mathbf{A}_{i1}^{-1}(z), \quad \text{and} \\
\mathbf{B}_4(z) &= n^{-1} b(z) (z - (n-1)/n \cdot b(z))^{-1} \sum_{i \neq 1} (\mathbf{A}_{i1}^{-1}(z) - \mathbf{A}_1^{-1}(z)), \\
&= -n^{-1} b(z) (z - (n-1)/n \cdot b(z))^{-1} \sum_{i \neq 1} \mathbf{A}_{i1}^{-1} \mathbf{r}_i \mathbf{r}_i^T \mathbf{A}_{i1}^{-1} \beta_{i(1)},
\end{aligned}$$

according to eqn.(2.9) of Bai and Silverstein (2004).

We also note here that throughout this paper, C and $C_{(\cdot)}$ stand for constants that may take different values from one appearance to another.

4.3 Proof of Theorem 2.1

By Lemma 4 in Karoui (2009), we know that $\|\text{diag}(\mathbf{S}) - \mathbf{I}_p\| \rightarrow 0$ and $\|[\text{diag}(\mathbf{S})]^{-1/2} - \mathbf{I}_p\| \rightarrow 0$ almost surely, where throughout this paper $\|\cdot\|$ denote the spectral norm of a matrix. Let $\mathbf{X} = (\mathbf{x}_1, \dots, \mathbf{x}_n)$. Therefore, follow the same strategy in the proof of Lemma 1 in the same article, we know that the LSD of $\mathbf{R}^{-1} \widehat{\mathbf{R}}$ is the same as that of $n^{-1} \mathbf{X} \mathbf{X}^T$, which we know is the specified Marcenko-Pastur law. In fact, notice that $\|n^{-1} \mathbf{X} \mathbf{X}^T\| \rightarrow (1 + \sqrt{\rho})^2$, *a.s.* as $n \rightarrow \infty$, we obtain

$$\begin{aligned}
& \left\| \mathbf{R}^{-1/2} \widehat{\mathbf{R}} \mathbf{R}^{-1/2} - n^{-1} \mathbf{X} \mathbf{X}^T \right\| \tag{4.1} \\
&= \left\| \mathbf{R}^{-1/2} \left([\text{diag}(\mathbf{S})]^{-1/2} - \mathbf{I}_p + \mathbf{I}_p \right) n^{-1} \mathbf{R}^{1/2} \mathbf{X} \mathbf{X}^T \mathbf{R}^{1/2} \left([\text{diag}(\mathbf{S})]^{-1/2} - \mathbf{I}_p + \mathbf{I}_p \right) \mathbf{R}^{-1/2} - n^{-1} \mathbf{X} \mathbf{X}^T \right\| \\
&\leq \left\| \mathbf{R}^{-1/2} \left([\text{diag}(\mathbf{S})]^{-1/2} - \mathbf{I}_p \right) n^{-1} \mathbf{R}^{1/2} \mathbf{X} \mathbf{X}^T \mathbf{R}^{1/2} \left([\text{diag}(\mathbf{S})]^{-1/2} - \mathbf{I}_p \right) \mathbf{R}^{-1/2} \right\| \\
&\quad + \left\| n^{-1} \mathbf{X} \mathbf{X}^T \mathbf{R}^{1/2} \left([\text{diag}(\mathbf{S})]^{-1/2} - \mathbf{I}_p \right) \mathbf{R}^{-1/2} \right\| \\
&\quad + \left\| \mathbf{R}^{-1/2} \left([\text{diag}(\mathbf{S})]^{-1/2} - \mathbf{I}_p \right) \mathbf{R}^{1/2} n^{-1} \mathbf{X} \mathbf{X}^T \right\| \\
&\leq \left\| \left([\text{diag}(\mathbf{S})]^{-1/2} - \mathbf{I}_p \right) \right\|^2 \left\| n^{-1} \mathbf{X} \mathbf{X}^T \right\| + 2 \left\| \left([\text{diag}(\mathbf{S})]^{-1/2} - \mathbf{I}_p \right) \right\| \left\| n^{-1} \mathbf{X} \mathbf{X}^T \right\| \rightarrow 0 \quad \textit{a.s.}
\end{aligned}$$

4.4 Proof of Theorem 3.1

Observe that by Cauchy integral formula, for any function g that is analytic in a domain containing the interval $[\lambda_1, \lambda_p]$, we have

$$\begin{aligned} W(g) &= \sum_{j=1}^p g(\lambda_j) - p \int g(x) dF_{\rho_n}(x) \\ &= -\frac{1}{2\pi i} \oint_{\mathcal{C}} g(z) M_n(z) dz, \text{ where } M_n(z) := p(s_n(z) - s_{\rho_n}(z)), \end{aligned} \quad (4.1)$$

where $s_{\rho_n}(z)$ is the Stieltjes transformation of the distribution F_{ρ_n} , and \mathcal{C} is any contour inside the domain and surrounding the interval $[\lambda_1, \lambda_p]$. Thus to derive the CLT of the LSS of $\mathbf{R}^{-1}\widehat{\mathbf{R}}$, the analysis of its Stieltjes transformation $s_n(z)$ is required.

Truncation, Centralization and Rescaling the variables

Recall that $\mathbf{S} = n^{-1}\mathbf{G}\mathbf{X}_n\mathbf{X}_n^T\mathbf{G}^T$ and

$$\mathbf{R}^{-1}\widehat{\mathbf{R}} = \mathbf{R}^{-1} [\text{diag}(\mathbf{S})]^{-1/2} n^{-1} \mathbf{R}^{1/2} \mathbf{X}\mathbf{X}^T \mathbf{R}^{1/2} [\text{diag}(\mathbf{S})]^{-1/2}.$$

Denote $\check{\mathbf{X}} = (\check{x}_{ij})$ with $\hat{x}_{ij} = x_{ij} I_{\{|x_{ij}| < \eta_n \sqrt{n}\}}$, $\check{\mathbf{S}} = n^{-1} \mathbf{R}^{1/2} \check{\mathbf{X}}\check{\mathbf{X}}^T \mathbf{R}^{1/2}$,

$$\mathbf{R}^{-1}\check{\mathbf{R}} = \mathbf{R}^{-1} [\text{diag}(\check{\mathbf{S}})]^{-1/2} n^{-1} \mathbf{R}^{1/2} \check{\mathbf{X}}\check{\mathbf{X}}^T \mathbf{R}^{1/2} [\text{diag}(\check{\mathbf{S}})]^{-1/2}$$

and $\check{W}(g)$ be the truncated version of $W(g)$. As have been proved in Karoui (2009), under the moment assumption, we shall select a sequence of $\eta_n = (\log n)^{-(1+\varepsilon)/2} \rightarrow 0$ as $n \rightarrow \infty$ which satisfies that

$$\mathbb{P}(\widehat{\mathbf{R}} \neq \check{\mathbf{R}}, i.o.) = 0.$$

Now define $\tilde{\mathbf{X}} = (\tilde{x}_{ij})$ with $\tilde{x}_{ij} = (\check{x}_{ij} - \mathbb{E}\check{x}_{ij}) / \sqrt{\mathbb{E}(\check{x}_{ij} - \mathbb{E}\check{x}_{ij})^2}$. Also define $\tilde{\mathbf{S}}$, $\tilde{\mathbf{R}}$ and $\tilde{W}(f)$ as the analogues of \mathbf{S} , \mathbf{R} and $W(f)$ with \mathbf{X}_n replaced by $\tilde{\mathbf{X}}$. For large n and any $1 \leq i \leq p, 1 \leq j \leq n$, we have the following estimates

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(1) $|\mathbb{E} \check{x}_{ij}| = \mathbb{E} |x_{ij} I(|x_{ij}| \geq \eta_n \sqrt{n})| \leq \eta_n n^{-3/2} \mathbb{E} |x_{ij}|^4 = o(n^{-3/2}),$

(2)

$$\begin{aligned} \mathbb{E} (\check{x}_{ij} - \mathbb{E} \check{x}_{ij})^2 - 1 &= \mathbb{E} x_{ij}^2 I(|x_{ij}| \geq \eta_n \sqrt{n}) - (\mathbb{E} x_{ij} I(|x_{ij}| \geq \eta_n \sqrt{n}))^2 \\ &\leq \mathbb{E} x_{ij}^2 I(|x_{ij}| \geq \eta_n \sqrt{n}) = o(n^{-1}), \end{aligned} \quad (4.2)$$

(3) $\frac{\sqrt{\mathbb{E}(\check{x}_{ij} - \mathbb{E} \check{x}_{ij})^2} - 1}{\sqrt{\mathbb{E}(\check{x}_{ij} - \mathbb{E} \check{x}_{ij})^2}} \leq 2 (\mathbb{E} (\check{x}_{ij} - \mathbb{E} \check{x}_{ij})^2 - 1) = o(n^{-1}).$

It follows that as $n \rightarrow \infty$,

$$\begin{aligned} \|n^{-1} \check{\mathbf{X}} \check{\mathbf{X}}^T - n^{-1} \tilde{\mathbf{X}} \tilde{\mathbf{X}}^T\| &\leq \|n^{-1} \check{\mathbf{X}} (\check{\mathbf{X}} - \tilde{\mathbf{X}})^T\| + \|n^{-1} (\check{\mathbf{X}} - \tilde{\mathbf{X}}) \tilde{\mathbf{X}}^T\| \\ &\leq 2 \left((1 + \sqrt{\rho_n})^2 \frac{\sqrt{\mathbb{E}(\check{x}_{ij} - \mathbb{E} \check{x}_{ij})^2} - 1}{\sqrt{\mathbb{E}(\check{x}_{ij} - \mathbb{E} \check{x}_{ij})^2}} + 2n^{-1/2} n |\mathbb{E} \check{x}_{ij}| \right) = o(n^{-1}). \end{aligned} \quad (4.3)$$

Let \mathbf{g}_ℓ to be the transpose of the ℓ -th row of the matrix $\mathbf{R}^{1/2}$, it can be verified that

$$\begin{aligned} &\max_{i \leq p} \|\mathbf{g}_i^T \left(\frac{1}{n} \check{\mathbf{X}} \check{\mathbf{X}}^T - \frac{1}{n} \tilde{\mathbf{X}} \tilde{\mathbf{X}}^T \right) \mathbf{g}_i\| \\ &\leq \max_{i \leq p} \left(\left\| \frac{1}{n} \mathbf{g}_i^T \check{\mathbf{X}} (\check{\mathbf{X}} - \tilde{\mathbf{X}})^T \mathbf{g}_i \right\| + \left\| \frac{1}{n} \mathbf{g}_i^T (\check{\mathbf{X}} - \tilde{\mathbf{X}}) \tilde{\mathbf{X}}^T \mathbf{g}_i \right\| \right) = o_{a.s.}(n^{-1}), \end{aligned} \quad (4.4)$$

which implies $\|\text{diag}(\check{\mathbf{S}} - \tilde{\mathbf{S}})\| = o_{a.s.}(n^{-1})$. Then we get

$$\begin{aligned} &\|\check{\mathbf{R}}_n - \tilde{\mathbf{R}}_n\| \\ &\leq \|\mathbf{R}^{-1} (\text{diag}(\check{\mathbf{S}})^{-1/2} - \text{diag}(\tilde{\mathbf{S}})^{-1/2}) n^{-1} \mathbf{R}^{1/2} \check{\mathbf{X}} \check{\mathbf{X}}^T \mathbf{R}^{1/2} \text{diag}(\check{\mathbf{S}})^{-1/2}\| \\ &\quad + \|\mathbf{R}^{-1} \text{diag}(\tilde{\mathbf{S}})^{-1/2} \mathbf{R}^{1/2} (n^{-1} \check{\mathbf{X}} \check{\mathbf{X}}^T - n^{-1} \tilde{\mathbf{X}} \tilde{\mathbf{X}}^T) \mathbf{R}^{1/2} \text{diag}(\check{\mathbf{S}})^{-1/2}\| \\ &\quad + \|\mathbf{R}^{-1} \text{diag}(\tilde{\mathbf{S}})^{-1/2} \mathbf{R}^{1/2} n^{-1} \tilde{\mathbf{X}} \tilde{\mathbf{X}}^T \mathbf{R}^{1/2} (\text{diag}(\check{\mathbf{S}})^{-1/2} - \text{diag}(\tilde{\mathbf{S}})^{-1/2})\| = o_{a.s.}(n^{-1}). \end{aligned} \quad (4.5)$$

Combining the estimates above, we finally obtain for large n

$$\left| \check{W}(g) - \tilde{W}(g) \right| \leq M \sum_{k=1}^p \left| \lambda_k(\check{\mathbf{R}}_n) - \lambda_k(\tilde{\mathbf{R}}_n) \right| \leq Mp \|\check{\mathbf{R}}_n - \tilde{\mathbf{R}}_n\| = o_{a.s.}(1), \quad (4.6)$$

where M is a bound on $|f'(z)|$. Therefore, we shall assume in the following that the underlying variables in the data matrix \mathbf{X} are all truncated at $\eta_n\sqrt{n}$, centralized and rescaled to have unit variances.

Decomposing the Stieltjes transformation of the ESD of $\mathbf{R}^{-1}\widehat{\mathbf{R}}$

We now decompose the Stieltjes transformation of the ESD of $\mathbf{R}^{-1}\widehat{\mathbf{R}}$ into several terms. Denote

$$\begin{aligned} \mathbf{D} &= \mathbf{R}^{-1/2} [\text{diag}(\mathbf{S}) - \mathbf{I}_p] \mathbf{R}^{1/2}, & \mathbf{D}_j &= \mathbf{R}^{-1/2} \left[\text{diag} \left(\frac{1}{n} \sum_{i \neq j} \mathbf{S}_j \right) - \frac{n-1}{n} \mathbf{I}_p \right] \mathbf{R}^{1/2}, \\ \mathbf{L} &= \mathbf{R}^{-1/2} [\text{diag}(\mathbf{S})]^{1/2} \mathbf{R} [\text{diag}(\mathbf{S})]^{1/2} \mathbf{R}^{-1/2}, & \mathbf{H}(z) &= \mathbf{R}^{-1/2} \mathbf{S} \mathbf{R}^{-1/2} - z \mathbf{L}. \end{aligned}$$

Reorganize

$$ps_n(z) = \text{tr} \left(\mathbf{R}^{-1/2} \widehat{\mathbf{R}} \mathbf{R}^{-1/2} - z \mathbf{I}_p \right)^{-1} = \text{tr} (\mathbf{H}^{-1}(z) \mathbf{L}).$$

Moreover, by the identity

$$\mathbf{A}_0^{-1} - \mathbf{B}_0^{-1} = \mathbf{A}_0^{-1} (\mathbf{B}_0 - \mathbf{A}_0) \mathbf{B}_0^{-1} \quad (4.7)$$

for any invertible matrices \mathbf{A}_0 and \mathbf{B}_0 , we have

$$\begin{aligned} \mathbf{H}^{-1}(z) &= \mathbf{A}^{-1}(z) + z \mathbf{H}^{-1}(z) (\mathbf{L} - \mathbf{I}_p) \mathbf{A}^{-1}(z) \\ &= \mathbf{A}^{-1}(z) + z \mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p) \mathbf{A}^{-1}(z) + z^2 \mathbf{H}^{-1}(z) [(\mathbf{L} - \mathbf{I}_p) \mathbf{A}^{-1}(z)]^2. \end{aligned} \quad (4.8)$$

Hence, it follows that

$$\begin{aligned} ps_n(z) &= \text{tr} [\mathbf{H}^{-1}(z)] + \text{tr} [\mathbf{H}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)] \\ &= \text{tr} (\mathbf{A}^{-1}(z)) + \text{tr} [\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)] + z \text{tr} [\mathbf{A}^{-2}(z) (\mathbf{L} - \mathbf{I}_p)] \\ &\quad + z \text{tr} [\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)]^2 + z^2 \text{tr} [\mathbf{A}^{-2}(z) (\mathbf{L} - \mathbf{I}_p) \mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)] \\ &\quad + z^2 \text{tr} \left\{ \mathbf{H}^{-1}(z) (\mathbf{L} - \mathbf{I}_p) [\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)]^2 \right\} \\ &\quad + z^3 \text{tr} \left\{ \mathbf{H}^{-1}(z) [(\mathbf{L} - \mathbf{I}_p) \mathbf{A}^{-1}(z)]^3 \right\}. \end{aligned} \quad (4.9)$$

The following steps of proof will be focused on the analysis of the terms above.

Choosing the contour \mathcal{C}

Recall (4.1), an integral contour should be determined. We choose the contour \mathcal{C} as

$$\begin{aligned} \mathcal{C} &= \mathcal{C}_\ell \cup \mathcal{C}_u \cup \mathcal{C}_b \cup \mathcal{C}_r, \text{ where} \\ \mathcal{C}_u &= \{x + i\nu_0 : x \in [x_\ell, x_r]\}, \quad \mathcal{C}_\ell = \{x_\ell + i\nu : |\nu| \leq \nu_0\} \\ \mathcal{C}_b &= \{x - i\nu_0 : x \in [x_\ell, x_r]\}, \quad \mathcal{C}_r = \{x_r + i\nu : |\nu| \leq \nu_0\} \\ &(x_\ell, x_r) \supset ((1 - \sqrt{\rho})^2, (1 + \sqrt{\rho})^2), \end{aligned}$$

and $\nu_0 > 0$ is to be determined. Let $\mathcal{C}_n = \mathcal{C} \cap \{z : |\Im z| > n^{-2}\}$. Moreover, for an $\varepsilon > 0$ sufficiently small so that

$$x_\ell + \varepsilon \leq (1 - \sqrt{\rho})^2 - \varepsilon \leq (1 + \sqrt{\rho})^2 + \varepsilon \leq x_r - \varepsilon,$$

define

$$\mathcal{B}_n = \{(1 - \sqrt{\rho})^2 - \varepsilon \leq \lambda_{\min}(\mathbf{X}\mathbf{X}^T/n) < \lambda_{\max}(\mathbf{X}\mathbf{X}^T/n) < (1 + \sqrt{\rho})^2 + \varepsilon\}.$$

By Theorem 5.9 in Bai and Silverstein (2010), we have that $P(\mathcal{B}_n) = o(n^{-t})$ for any given

$t > 0$. Let

$$\widehat{M}_n(z) = \begin{cases} M_n(z), & \text{if } z \in \mathcal{C}_n \\ M_n(x_\ell + in^{-2}), & \text{if } \Re z = x_\ell, \Im z \in [0, n^{-2}] \\ M_n(x_\ell - in^{-2}), & \text{if } \Re z = x_\ell, \Im z \in [-n^{-2}, 0) \\ M_n(x_r + in^{-2}), & \text{if } \Re z = x_r, \Im z \in [0, n^{-2}] \\ M_n(x_r - in^{-2}), & \text{if } \Re z = x_r, \Im z \in [-n^{-2}, 0). \end{cases}$$

Observe that on event \mathcal{B}_n , when $\Re z$ equals either x_ℓ or x_r , we have $|M_n(z)| \leq 1/\varepsilon$, hence

$$\begin{aligned} &\left| p \oint_{\mathcal{C}} g(z)(M_n(z) - \widehat{M}_n(z)) dz \right| = \left| p \oint_{\mathcal{C} \setminus \mathcal{C}_n} g(z)(M_n(z) - \widehat{M}_n(z)) dz \right| \\ &\leq K \frac{p}{n^2} \cdot 1/\varepsilon = o(1). \end{aligned}$$

Therefore, in order to establish the limit for $p \oint_{\mathcal{C}} g(z) M_n(z) dz$, it suffices to study $p \oint_{\mathcal{C}} g(z) \widehat{M}_n(z) dz$. Furthermore, since $\Im(z)$ can be chosen to be arbitrarily small, the contribution from the segments C_ℓ and C_r can be made small as well. This allows us to focus only on $z \in \mathcal{C}_u \cup \mathcal{C}_b$ in the following.

CLT of LSS of the sample correlation matrix

We are here in a position to establish the CLT for the LSS of the sample correlation matrix. Indicated by (4.1), the limit of the process $M_n(z) = p(s_n(z) - s_{\rho_n}(z))$ should be derived. Recall that by (4.9), we have

$$\begin{aligned}
 M_n(z) &= \text{tr} \mathbf{A}^{-1}(z) - \text{E}(\text{tr} \mathbf{A}^{-1}(z)) + \text{tr} [\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)] - \text{Etr} [\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)] \\
 &\quad + z \text{tr} [\mathbf{A}^{-2}(z) (\mathbf{L} - \mathbf{I}_p)] - z \text{Etr} [\mathbf{A}^{-2}(z) (\mathbf{L} - \mathbf{I}_p)] \\
 &\quad + \text{E}(\text{tr} \mathbf{A}^{-1}(z)) - p s_{\rho_n}(z) + \text{Etr} [\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)] + z \text{Etr} [\mathbf{A}^{-2}(z) (\mathbf{L} - \mathbf{I}_p)] \\
 &\quad + z \cdot I + z^2 \cdot II + z^2 \cdot III + z^3 \cdot IV.
 \end{aligned} \tag{4.10}$$

The analysis of $M_n(z)$ will be carried out by conducting the following steps.

\mathcal{I} : Derive the limit of $I = \text{tr} [\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)]^2$ and $II = \text{tr} [(\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p))^2 \mathbf{A}^{-1}(z)]$;

\mathcal{II} : Show that

$$III = \text{tr} \left\{ \mathbf{H}^{-1}(z) (\mathbf{L} - \mathbf{I}_p) [\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)]^2 \right\} = o_p(1), \tag{4.11}$$

and

$$IV = \text{tr} \left\{ \mathbf{H}^{-1}(z) [(\mathbf{L} - \mathbf{I}_p) \mathbf{A}^{-1}(z)]^3 \right\} = o_p(1); \tag{4.12}$$

\mathcal{III} : Derive the limit of

$$\begin{aligned}
 V &:= \text{tr} \mathbf{A}^{-1}(z) - \text{E}[\text{tr} \mathbf{A}^{-1}(z)] + \text{tr} [\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)] - \text{Etr} [\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)] \\
 &\quad + z \text{tr} [\mathbf{A}^{-2}(z) (\mathbf{L} - \mathbf{I}_p)] - z \text{Etr} [\mathbf{A}^{-2}(z) (\mathbf{L} - \mathbf{I}_p)];
 \end{aligned} \tag{4.13}$$

\mathcal{IV} : Derive the limit of

$$\begin{aligned} VI := & \text{Etr} [\mathbf{A}^{-1}(z)] - p s_{\rho_n}(z) + \text{Etr} [\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)] \\ & + z \text{Etr} [\mathbf{A}^{-2}(z) (\mathbf{L} - \mathbf{I}_p)]; \end{aligned} \quad (4.14)$$

\mathcal{V} : Combining the steps above to derive the limit of $M_n(z)$ and consequently the CLT of the LSS of $\tilde{\mathbf{R}}$.

We now process these steps one by one. It is worth pointing out that some repeating calculation will be omitted in order to reduce the length of this paper.

Step \mathcal{I} . In this step, we derive the limit of I and II , which is further divided into the following substeps.

$\mathcal{I.1}$: Decompose term I as

$$I = \frac{1+o(1)}{4} I_1 + \frac{1+o(1)}{4} I_2 + \frac{1+o(1)}{2} I_3 + \frac{1+o(1)}{4} I_4 + \frac{1+o(1)}{4} I_5 + \frac{1+o(1)}{16} I_6$$

to be specified below and show that $I_4 + I_5 + I_6 = o_p(1)$;

$\mathcal{I.2}$: Prove that

$$|I_1 - \text{E}(I_1)| + |I_2 - \text{E}(I_2)| + |I_3 - \text{E}(I_3)| = o_p(1);$$

$\mathcal{I.3}$: Prove that $\text{E}(I_1)$, $\text{E}(I_2)$ converge to the same limit

$$\frac{1}{z^2(1+\underline{s}(z))^2} \left(2\rho + \beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{j=1}^p g_{kj}^4 \right)$$

and $\text{E}(I_3)$ converge to

$$\frac{1}{z^2(1+\underline{s}(z))^2} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k,\ell=1}^p \text{E} \left[\mathbf{e}_k^T \mathbf{R}^{-1} \mathbf{e}_\ell r_{kl} \left(2r_{kl}^2 + \beta_x \sum_{h=1}^p g_{kh}^2 g_{\ell h}^2 \right) \right];$$

I.4: Combining the aforementioned results, we obtain that

$$z \cdot I \rightarrow \frac{1}{2z(1 + \underline{s}(z))^2} \left(2\rho + \beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{j=1}^p g_{kj}^4 \right) \quad (4.15)$$

$$+ \frac{1}{2z(1 + \underline{s}(z))^2} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k,\ell=1}^p \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{-1} \mathbf{e}_\ell r_{kl} \left(2r_{kl}^2 + \beta_x \sum_{h=1}^p g_{kh}^2 g_{\ell h}^2 \right) \right];$$

I.5: Repeating the steps as done in I.1-I.4, we similarly get

$$z^2 II \xrightarrow{p} \frac{z^2}{2} \frac{\partial}{\partial z} \left(\frac{1}{z^2(1 + \underline{s}(z))^2} \right) \cdot \left(2\rho + \beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{j=1}^p g_{kj}^4 \right).$$

We now process these substeps one by one.

Step I.1 It is apparent that from Lemma 6.2

$$\|\mathbf{D}\| \xrightarrow{a.s.} 0 \quad \text{and} \quad \max_j \|\mathbf{D}_j\| \xrightarrow{a.s.} 0. \quad (4.16)$$

Using the Taylor formula as $x \rightarrow 1$

$$\sqrt{x} = 1 + \frac{1}{2}(x-1)(1+o(1)) = 1 + \frac{1}{2}(x-1) - \frac{1}{8}(x-1)^2(1+o(1)),$$

one finds that

$$\begin{aligned} \mathbf{L} &= \mathbf{I}_p + \mathbf{R}^{-1/2} \left\{ [\text{diag}(\mathbf{S})]^{1/2} - \mathbf{I}_p \right\} \mathbf{R}^{1/2} + \mathbf{R}^{1/2} \left\{ [\text{diag}(\mathbf{S})]^{1/2} - \mathbf{I}_p \right\} \mathbf{R}^{-1/2} \\ &\quad + \mathbf{R}^{-1/2} \left\{ [\text{diag}(\mathbf{S})]^{1/2} - \mathbf{I}_p \right\} \mathbf{R} \left\{ [\text{diag}(\mathbf{S})]^{1/2} - \mathbf{I}_p \right\} \mathbf{R}^{-1/2} \\ &= \mathbf{I}_p + \frac{1}{2} \mathbf{D} + \frac{1}{2} \mathbf{D}^T - \frac{1+o(1)}{8} \mathbf{D}^2 - \frac{1+o(1)}{8} (\mathbf{D}^T)^2 + \frac{1+o(1)}{4} \mathbf{D} \mathbf{D}^T. \end{aligned} \quad (4.17)$$

This implies

$$\begin{aligned} I &= \frac{1+o(1)}{4} \text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \right)^2 + \frac{1+o(1)}{4} \text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D}^T \right)^2 + \frac{1+o(1)}{2} \text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \mathbf{A}^{-1}(z) \mathbf{D}^T \right) \\ &\quad + \frac{1+o(1)}{4} \text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \mathbf{A}^{-1}(z) \mathbf{D} \mathbf{D}^T \right) + \frac{1+o(1)}{4} \text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D}^T \mathbf{A}^{-1}(z) \mathbf{D} \mathbf{D}^T \right) \\ &\quad + \frac{1+o(1)}{16} \text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \mathbf{D}^T \right)^2 \end{aligned}$$

4. PROOFS OF THEOREMS 2.1 AND 3.1

$$:= \frac{1+o(1)}{4} I_1 + \frac{1+o(1)}{4} I_2 + \frac{1+o(1)}{2} I_3 + \frac{1+o(1)}{4} I_4 + \frac{1+o(1)}{4} I_5 + \frac{1+o(1)}{16} I_6.$$

In the following, we will show that $|I_4| + |I_5| + |I_6| = o_p(1)$. Define \mathbf{g}_ℓ to be the transpose of the ℓ -th row of the matrix $\mathbf{R}^{1/2}$ and \mathbf{e}_ℓ to be the ℓ -th column of the matrix \mathbf{I}_p . It is obvious from (4.2) that

$$\begin{aligned} I_4 &= \frac{1}{n} \sum_{j=1}^n \text{tr} \left[(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}^T \mathbf{A}^{-1}(z) \mathbf{D} \mathbf{D}^T \mathbf{A}^{-1}(z) \mathbf{R}^{-1/2} \right] \\ &= \frac{1}{n} \sum_{j=1}^n \text{tr} \left[(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{D} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \right] \\ &\quad - \frac{1}{n} \sum_{j=1}^n \beta_j(z) \text{tr} \left[(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{D} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{r}_j \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \right] \\ &\quad - \frac{1}{n} \sum_{j=1}^n \beta_j(z) \text{tr} \left[(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{r}_j \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D} \mathbf{D}^T \mathbf{A}^{-1}(z) \mathbf{R}^{-1/2} \right]. \end{aligned}$$

By (4.16) and Lemma 6.3, one finds that

$$\begin{aligned} &\mathbb{E} \left| \frac{1}{n} \sum_{j=1}^n \beta_j(z) \text{tr} \left[(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{D} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{r}_j \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \right] \right. \\ &\quad \left. + \frac{1}{n} \sum_{j=1}^n \beta_j(z) \text{tr} \left[(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{r}_j \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D} \mathbf{D}^T \mathbf{A}^{-1}(z) \mathbf{R}^{-1/2} \right] \right| \\ &\leq \frac{o(1)}{n} \sum_{j=1}^n \mathbb{E} \left[\mathbf{r}_j^T \mathbf{r}_j \|\text{diag}(\mathbf{S}_j) - \mathbf{I}_p\| \right] \\ &\leq \frac{o(1)}{n} \sum_{j=1}^n \mathbb{E}^{1/2} \left(\mathbf{r}_j^T \mathbf{r}_j \right)^2 \mathbb{E}^{1/2} \|\text{diag}(\mathbf{S}_j) - \mathbf{I}_p\|^2 = o(1). \end{aligned}$$

Here

$$\mathbb{E} \|\text{diag}(\mathbf{S}_j) - \mathbf{I}_p\|^2 = \max_{k=1, \dots, p} \mathbb{E} \left(\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1 \right)^2 \leq C. \quad (4.18)$$

Hence, we get from (4.16) and (4.18)

$$\begin{aligned} I_4 &= \frac{1}{n} \sum_{j=1}^n \text{tr} \left[(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{D} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \right] + o_p(1) \\ &= \frac{1}{n} \sum_{j=1}^n \text{tr} \left[(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{D}_j^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \right] \end{aligned}$$

$$\begin{aligned}
& + \frac{1}{n^2} \sum_{j=1}^n \text{tr} \left[(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{R}^{1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{-1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \right] \\
& + \frac{1}{n^2} \sum_{j=1}^n \text{tr} \left[(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}_j^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \right] \\
& + \frac{1}{n^2} \sum_{j=1}^n \text{tr} \left[(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{-1/2} \mathbf{A}_j^{-1}(z) \mathbf{D} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \right] + o_p(1) \\
& = \frac{1}{n} \sum_{j=1}^n \text{tr} \left[(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{D}_j^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \right] + o_p(1).
\end{aligned}$$

Define $\mathbf{M}_j(z) = \mathbf{R}^{1/2} \mathbf{D}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{D}_j^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2}$. It follows from (4.16)

$$\begin{aligned}
\mathbb{E} |I_4|^2 &= \frac{1}{n^2} \sum_{j=1}^n \mathbb{E} |\text{tr} [(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{M}_j(z)]|^2 + o(1) \\
&= \frac{1}{n^2} \sum_{j=1}^n \sum_{k, \ell=1}^p \mathbb{E} \left[\left(\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1 \right) \left(\mathbf{X}_j^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{X}_j - 1 \right) \mathbf{e}_k^T \mathbf{M}_j(z) \mathbf{e}_k \mathbf{e}_\ell^T \overline{\mathbf{M}_j(z)} \mathbf{e}_\ell \right] + o(1) \\
&\leq \frac{o(1)}{n^2} \sum_{j=1}^n \sum_{k, \ell=1}^p \left[2 \left(\mathbf{g}_k^T \mathbf{g}_\ell \right)^2 + \sum_{h=1}^p \left(\mathbf{e}_h^T \mathbf{g}_k \mathbf{g}_\ell^T \mathbf{e}_h \right)^2 \right] = o(1).
\end{aligned}$$

Consequently, we conclude

$$I_4 = o_p(1). \quad (4.19)$$

Using the same method, it can be verified that

$$I_5 + I_6 = o_p(1),$$

which completes this substep.

Step I.2 In this substep, three terms I_1, I_2 and I_3 will be shown to converge in probability to their means, say

$$|I_1 - \mathbb{E}(I_1)| + |I_2 - \mathbb{E}(I_2)| + |I_3 - \mathbb{E}(I_3)| = o_p(1).$$

We start with $I_1 - \mathbb{E}(I_1) = o_p(1)$. By (4.2), we obtain that

$$I_1 - \mathbb{E}(I_1)$$

$$\begin{aligned}
 &= \text{tr}(\mathbf{A}^{-1}(z)\mathbf{D})^2 - \mathbb{E}(\text{tr}\mathbf{A}^{-1}(z)\mathbf{D})^2 \\
 &= \sum_{j=1}^n (\mathbb{E}_j - \mathbb{E}_{j-1}) \left(\text{tr}(\mathbf{A}^{-1}(z)\mathbf{D})^2 - \text{tr}(\mathbf{A}_j^{-1}(z)\mathbf{D}_j)^2 \right) \\
 &= \sum_{j=1}^n (\mathbb{E}_j - \mathbb{E}_{j-1}) \left[-2\beta_j(z)\mathbf{r}_j^T \mathbf{A}_j^{-1}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{r}_j \right. \\
 &\quad + \left(\beta_j(z)\mathbf{r}_j^T \mathbf{A}_j^{-1}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{r}_j \right)^2 \\
 &\quad + \frac{2}{n} \text{tr} \left(\mathbf{A}_j^{-1}(z)\mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z)\mathbf{D}_j \right) \\
 &\quad - \frac{4}{n} \left(\beta_j(z)\mathbf{r}_j^T \mathbf{A}_j^{-1}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z)\mathbf{r}_j \right) \\
 &\quad + \frac{2}{n} \left(\beta_j^2(z)\mathbf{r}_j^T \mathbf{A}_j^{-1}(z)\mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z)\mathbf{r}_j \mathbf{r}_j^T \mathbf{A}_j^{-1}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{r}_j \right) \\
 &\quad + \frac{1}{n^2} \text{tr} \left(\mathbf{A}_j^{-1}(z)\mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \right)^2 \\
 &\quad - \frac{2}{n^2} \left[\beta_j(z)\mathbf{r}_j^T \left[\mathbf{A}_j^{-1}(z)\mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \right]^2 \mathbf{A}_j^{-1}(z)\mathbf{r}_j \right] \\
 &\quad \left. + \frac{1}{n^2} \left(\beta_j(z)\mathbf{r}_j^T \mathbf{A}_j^{-1}(z)\mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z)\mathbf{r}_j \right)^2 \right] \\
 &= \sum_{j=1}^n (-2P_{j1} + P_{j2} + 2P_{j3} - 4P_{j4} + 2P_{j5} + P_{j6} - 2P_{j7} + P_{j8}).
 \end{aligned}$$

We now devote to investigating those terms. Firstly, by (4.3), one gets

$$\begin{aligned}
 \sum_{j=1}^n P_{j1} &= - \sum_{j=1}^n (\mathbb{E}_j - \mathbb{E}_{j-1}) \left(b_j(z)\beta_j(z)\widehat{\gamma}_j(z)\mathbf{r}_j^T \mathbf{A}_j^{-1}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{r}_j \right) \\
 &\quad + \sum_{j=1}^n \mathbb{E}_j \left[b_j(z) \left(\mathbf{r}_j^T \mathbf{A}_j^{-1}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{r}_j - \frac{1}{n} \text{tr}(\mathbf{A}_j^{-2}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{D}_j) \right) \right].
 \end{aligned}$$

Denote $\mathbb{E}_{(j)} = \mathbb{E}(\cdot | \mathcal{F}_j)$ and

$$\mathcal{F}_j = \sigma\{\mathbf{r}_1, \dots, \mathbf{r}_{j-1}, \mathbf{r}_{j+1}, \dots, \mathbf{r}_n\}.$$

It follows that from (4.16) and Lemma 6.1

$$\begin{aligned}
 &\mathbb{E} \left| \sum_{j=1}^n (\mathbb{E}_j - \mathbb{E}_{j-1}) \left(b_j(z)\beta_j(z)\widehat{\gamma}_j(z)\mathbf{r}_j^T \mathbf{A}_j^{-1}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{r}_j \right) \right|^2 \\
 &\leq C \sum_{j=1}^n \mathbb{E} \left| \widehat{\gamma}_j(z) \left(\mathbf{r}_j^T \mathbf{A}_j^{-1}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{r}_j - \frac{1}{n} \text{tr}(\mathbf{A}_j^{-2}(z)\mathbf{D}_j \mathbf{A}_j^{-1}(z)\mathbf{D}_j) \right) \right|^2
 \end{aligned}$$

$$\begin{aligned}
 & + \frac{C}{n^2} \sum_{j=1}^n \mathbb{E} \left| \widehat{\gamma}_j(z) \operatorname{tr} \left(\mathbf{A}_j^{-2}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{D}_j \right) \right|^2 \\
 & \leq C \sum_{j=1}^n \mathbb{E} \left[\mathbb{E}_{(j)}^{1/2} \left| \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j - \frac{1}{n} \operatorname{tr} \left(\mathbf{A}_j^{-2}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{D}_j \right) \right|^4 \right. \\
 & \quad \left. \times \mathbb{E}_{(j)}^{1/2} |\widehat{\gamma}_j(z)|^4 \right] + C \sum_{j=1}^n \mathbb{E} \left[\|\mathbf{D}_j\|^4 \mathbb{E}_{(j)} |\widehat{\gamma}_j(z)|^2 \right] \leq \frac{C}{n} \sum_{j=1}^n \mathbb{E} \|\mathbf{D}_j\|^4 \rightarrow 0,
 \end{aligned}$$

and

$$\begin{aligned}
 & \mathbb{E} \left| \sum_{j=1}^n \mathbb{E}_j \left[b_j(z) \left(\mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j - \frac{1}{n} \operatorname{tr} \left(\mathbf{A}_j^{-2}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{D}_j \right) \right) \right] \right|^2 \\
 & \leq C \sum_{j=1}^n \mathbb{E} \left| \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j - \frac{1}{n} \operatorname{tr} \left(\mathbf{A}_j^{-2}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{D}_j \right) \right|^2 \\
 & \leq \frac{C}{n} \sum_{j=1}^n \mathbb{E} \|\mathbf{D}_j\|^4 \rightarrow 0.
 \end{aligned}$$

Hence, we have

$$\sum_{j=1}^n P_{j1} = o_p(1). \tag{4.20}$$

Secondly, we deal with $\sum_{j=1}^n P_{j2}$. By (4.3), we have

$$\begin{aligned}
 & \sum_{j=1}^n P_{j2} \\
 = & \sum_{j=1}^n (\mathbb{E}_j - \mathbb{E}_{j-1}) \left[b_j(z) \left(\mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j - \frac{1}{n} \operatorname{tr} \left(\mathbf{A}_j^{-2}(z) \mathbf{D}_j \right) \right) \right]^2 \\
 & + \frac{2}{n} \sum_{j=1}^n (\mathbb{E}_j - \mathbb{E}_{j-1}) \left[b_j^2(z) \left(\mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j - \frac{1}{n} \operatorname{tr} \left(\mathbf{A}_j^{-2}(z) \mathbf{D}_j \right) \right) \operatorname{tr} \left(\mathbf{A}_j^{-2}(z) \mathbf{D}_j \right) \right] \\
 & + 2 \sum_{j=1}^n (\mathbb{E}_j - \mathbb{E}_{j-1}) \left(b_j(z) \sqrt{\beta_j(z)} \sqrt{\widehat{\gamma}_j(z)} \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j \right)^2 \\
 & + \sum_{j=1}^n (\mathbb{E}_j - \mathbb{E}_{j-1}) \left(b_j(z) \beta_j(z) \widehat{\gamma}_j(z) \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j \right)^2.
 \end{aligned}$$

For the first term of the righthand side of the above equality, it yields that from Lemma 6.1

$$\begin{aligned}
 & \mathbb{E} \left| \sum_{j=1}^n (\mathbb{E}_j - \mathbb{E}_{j-1}) \left[b_j(z) \left(\mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j - \frac{1}{n} \operatorname{tr} \left(\mathbf{A}_j^{-2}(z) \mathbf{D}_j \right) \right) \right] \right|^2 \\
 & \leq C \sum_{j=1}^n \mathbb{E} \left| \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j - \frac{1}{n} \operatorname{tr} \left(\mathbf{A}_j^{-2}(z) \mathbf{D}_j \right) \right|^4 \leq \frac{C n_n^4}{n} \sum_{j=1}^n \mathbb{E} \|\mathbf{D}_j\|^4 \rightarrow 0.
 \end{aligned}$$

4. PROOFS OF THEOREMS 2.1 AND 3.1

Other terms are similar. Hence, one gets

$$\sum_{j=1}^n P_{j2} = o_p(1). \quad (4.21)$$

Thirdly, we prove that $\sum_{j=1}^n P_{j3} = o_p(1)$. Define \mathbf{g}_ℓ to be the transpose of the ℓ -th row of the matrix \mathbf{G} . It follows from (4.16) that

$$\begin{aligned} \mathbb{E} \left| \sum_{j=1}^n P_{j3} \right|^2 &\leq C \sum_{j=1}^n \mathbb{E} \left| \sum_{\ell=1}^p \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_\ell \left(\mathbf{r}_j^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{r}_j - \frac{1}{n} \right) \right|^2 \\ &\leq C \sum_{j=1}^n \sum_{\ell_1=1}^p \sum_{\ell_2=1}^p \mathbb{E} \left[\mathbf{e}_{\ell_1}^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_{\ell_1} \mathbf{e}_{\ell_2}^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(\bar{z}) \mathbf{D}_j \right. \\ &\quad \left. \times \mathbf{A}_j^{-1}(\bar{z}) \mathbf{R}^{-1/2} \mathbf{e}_{\ell_2} \left(\mathbf{r}_j^T \mathbf{g}_{\ell_1} \mathbf{g}_{\ell_1}^T \mathbf{r}_j - \frac{1}{n} \right) \left(\mathbf{r}_j^T \mathbf{g}_{\ell_2} \mathbf{g}_{\ell_2}^T \mathbf{r}_j - \frac{1}{n} \right) \right] \\ &\leq \frac{C}{n^2} \sum_{j=1}^n \sum_{\ell_1=1}^p \sum_{\ell_2=1}^p \mathbb{E} \left[\mathbf{e}_{\ell_1}^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_{\ell_1} \mathbf{e}_{\ell_2}^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(\bar{z}) \mathbf{D}_j \right. \\ &\quad \left. \times \mathbf{A}_j^{-1}(\bar{z}) \mathbf{R}^{-1/2} \mathbf{e}_{\ell_2} \left(2 \left(\mathbf{g}_{\ell_1}^T \mathbf{g}_{\ell_2} \right)^2 + \beta_x \sum_{h=1}^p \mathbf{e}_h^T \mathbf{g}_{\ell_1} \mathbf{g}_{\ell_1}^T \mathbf{e}_h \mathbf{e}_h^T \mathbf{g}_{\ell_2} \mathbf{g}_{\ell_2}^T \mathbf{e}_h \right) \right] \\ &\leq \frac{C}{n} \sum_{j=1}^n \mathbb{E} \|\mathbf{D}_j\|^2 \rightarrow 0. \end{aligned}$$

This shows that

$$\sum_{j=1}^n P_{j3} = o_p(1). \quad (4.22)$$

Next, combining (4.16), Lemma 6.1, and Lemma 6.3, we have

$$\begin{aligned} \mathbb{E} \left| \sum_{j=1}^n P_{j4} \right| &\leq \frac{C}{n} \sum_{j=1}^n \mathbb{E} \left| \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{r}_j \right| \\ &\leq \frac{C}{n} \sum_{j=1}^n \max_{\ell=1, \dots, p} \mathbb{E} \left[\|\mathbf{D}_j\| \mathbf{r}_j^T \mathbf{r}_j \left| \mathbf{X}_j^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{X}_j - 1 \right| \right] \\ &\leq \frac{C}{n} \sum_{j=1}^n \max_{\ell=1, \dots, p} \mathbb{E} \left[\|\mathbf{D}_j\| \mathbb{E}_{(j)}^{1/2} \left| \mathbf{r}_j^T \mathbf{r}_j \right|^2 \mathbb{E}_{(j)}^{1/2} \left| \mathbf{X}_j^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{X}_j - 1 \right|^2 \right] \\ &\leq \frac{C}{n} \sum_{j=1}^n \mathbb{E} \|\mathbf{D}_j\| \rightarrow 0. \end{aligned}$$

Therefore, it yields that

$$\sum_{j=1}^n P_{j4} = o_p(1). \quad (4.23)$$

Using the same methods, we can get

$$\sum_{j=1}^n (P_{j5} + P_{j7} + P_{j8}) = o_p(1). \quad (4.24)$$

Now we deal with P_{j6} . Denote

$$\mathbf{M}_q = \mathbf{g}_q \mathbf{g}_q^T = (m_{kj}^q), q = 1, 2, 3, 4,$$

then one finds that for any positive number t

$$\mathbf{e}_k^T \mathbf{M}_q \mathbf{e}_k \geq 0 \quad \text{and} \quad \text{tr}(\mathbf{M}_q)^t = 1.$$

By the proof of Lemma 6.1 [see Bai and Silverstein (2010)], we can get

$$\begin{aligned} & \mathbb{E} \left[\prod_{q=1}^4 \left(\mathbf{x}_j^T \mathbf{M}_q \mathbf{x}_j - \text{tr}(\mathbf{M}_q) \right) \right] \\ &= \mathbb{E} \left[\prod_{q=1}^4 \sum_{l_q=1}^p (x_{l_q j}^2 - 1) m_{l_q l_q}^q \right] + \mathbb{E} \left[\prod_{q=1}^4 \sum_{l_q \neq k_q=1}^p x_{l_q j} x_{k_q j} m_{l_q k_q}^q \right] \\ &\leq C \left| \mathbb{E} \left[\sum_{l=1}^p (x_{lj}^2 - 1)^4 m_{ll}^1 m_{ll}^2 m_{ll}^3 m_{ll}^4 \right] + \eta_n^2 n \sum_{q_1 \neq q_2=1}^4 \sum_{l_{q_1}=1}^p \sum_{l_{q_2}=1}^p (m_{l_{q_1} l_{q_1}}^{q_1})^3 m_{l_{q_2} l_{q_2}}^{q_2} \right. \\ &\quad \left. + \sum_{q_1 \neq q_2=1}^4 \sum_{l_{q_1}=1}^p \sum_{l_{q_2}=1}^p (m_{l_{q_1} l_{q_1}}^{q_1})^2 (m_{l_{q_2} l_{q_2}}^{q_2})^2 \right| + C \prod_{q=1}^4 \mathbb{E}^{1/4} \left| \sum_{l_q \neq k_q=1}^p x_{l_q j} x_{k_q j} m_{l_q k_q}^q \right|^4 \\ &\leq C \eta_n^4 n^2 \sum_{l=1}^p m_{ll}^1 m_{ll}^2 m_{ll}^3 m_{ll}^4 + C \eta_n^2 n + \prod_{q=1}^4 \left| \text{tr}(\mathbf{M}_q \mathbf{M}_q^T) \right|^{1/4} \\ &\leq C \eta_n^4 n^2 \sum_{l=1}^p m_{ll}^1 m_{ll}^2 m_{ll}^3 m_{ll}^4 + C \eta_n^2 n. \end{aligned}$$

By Cauchy-Schwarz inequality and the above inequality, it yields that

$$\begin{aligned} & \mathbb{E} \left| \sum_{j=1}^n P_{j6} \right|^2 \\ &= \sum_{j=1}^n \mathbb{E} \left| \sum_{\ell_1=1}^p \sum_{\ell_2=1}^p \left(\mathbf{r}_j^T \mathbf{g}_{\ell_1} \mathbf{g}_{\ell_1}^T \mathbf{r}_j - \frac{1}{n} \right) \left(\mathbf{r}_j^T \mathbf{g}_{\ell_2} \mathbf{g}_{\ell_2}^T \mathbf{r}_j - \frac{1}{n} \right) \mathbf{e}_{\ell_1}^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_{\ell_2} \right. \\ &\quad \left. \times \mathbf{e}_{\ell_2}^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_{\ell_1} \right|^2 \end{aligned}$$

$$\begin{aligned}
 &= \sum_{j=1}^n \sum_{\ell_1, \ell_2, \ell_3, \ell_4=1}^p \mathbb{E} \left[\left(\mathbf{r}_j^T \mathbf{g}_{\ell_1} \mathbf{g}_{\ell_1}^T \mathbf{r}_j - \frac{1}{n} \right) \left(\mathbf{r}_j^T \mathbf{g}_{\ell_2} \mathbf{g}_{\ell_2}^T \mathbf{r}_j - \frac{1}{n} \right) \left(\mathbf{r}_j^T \mathbf{g}_{\ell_3} \mathbf{g}_{\ell_3}^T \mathbf{r}_j - \frac{1}{n} \right) \right. \\
 &\quad \times \left(\mathbf{r}_j^T \mathbf{g}_{\ell_4} \mathbf{g}_{\ell_4}^T \mathbf{r}_j - \frac{1}{n} \right) \mathbf{e}_{\ell_1}^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_{\ell_2} \mathbf{e}_{\ell_2}^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_{\ell_1} \\
 &\quad \left. \times \mathbf{e}_{\ell_3}^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(\bar{z}) \mathbf{R}^{-1/2} \mathbf{e}_{\ell_4} \mathbf{e}_{\ell_4}^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(\bar{z}) \mathbf{R}^{-1/2} \mathbf{e}_{\ell_3} \right] \\
 &\leq \frac{C\eta_n^4}{n^2} \sum_{j=1}^n \sum_{\ell_1, \ell_2, \ell_3, \ell_4=1}^p \sum_{k=1}^p m_{kk}^{\ell_1} m_{kk}^{\ell_2} m_{kk}^{\ell_3} m_{kk}^{\ell_4} \\
 &\quad + \frac{C\eta_n^2}{n^3} \sum_{j=1}^n \mathbb{E} \left[\sum_{\ell_1, \ell_2=1}^p \mathbf{e}_{\ell_1}^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_{\ell_2} \mathbf{e}_{\ell_2}^T \mathbf{R}^{-1/2} \mathbf{A}_j^{-1}(\bar{z}) \mathbf{R}^{1/2} \mathbf{e}_{\ell_1} \right]^2 \\
 &\leq C\eta_n^2 \rightarrow 0.
 \end{aligned}$$

Consequently, it follows that

$$\sum_{j=1}^n P_{j6} = o_p(1). \quad (4.25)$$

Together with (4.20)-(4.25), we conclude that

$$I_1 - \mathbb{E}(I_1) = o_p(1). \quad (4.26)$$

Similar to the proof of (4.26), we shall prove $I_2 - \mathbb{E}(I_2) = o_p(1)$ and $I_3 - \mathbb{E}(I_3) = o_p(1)$.

Step $\mathcal{I}.3$ We now deduce the limit of the expectation of I_1 . Rewrite

$$\begin{aligned}
 I_1 &= \text{tr}(\mathbf{A}^{-1}(z)\mathbf{D})^2 = \frac{1}{n} \sum_{j=1}^n \text{tr} \left[\mathbf{A}^{-1}(z)\mathbf{D}\mathbf{A}^{-1}(z)\mathbf{R}^{-1/2}(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p)\mathbf{R}^{1/2} \right] \\
 &= \frac{1}{n} \sum_{j=1}^n \text{tr} \left[\mathbf{A}_j^{-1}(z)\mathbf{D}\mathbf{A}_j^{-1}(z)\mathbf{R}^{-1/2}(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p)\mathbf{R}^{1/2} \right] \\
 &\quad - \frac{2}{n} \sum_{j=1}^n \beta_j(z) \mathbf{r}_j^T \mathbf{A}_j^{-1}(z)\mathbf{D}\mathbf{A}_j^{-1}(z)\mathbf{R}^{-1/2}(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p)\mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z)\mathbf{r}_j \\
 &\quad + \frac{1}{n} \sum_{j=1}^n \beta_j^2(z) \mathbf{r}_j^T \mathbf{A}_j^{-1}(z)\mathbf{D}\mathbf{A}_j^{-1}(z)\mathbf{r}_j \mathbf{r}_j^T \mathbf{A}_j^{-1}(z)\mathbf{R}^{-1/2}(\text{diag}(\mathbf{S}_j) - \mathbf{I}_p)\mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z)\mathbf{r}_j \\
 &:= I_{1.1} - 2 \times I_{1.2} + I_{1.3}, \quad (4.27)
 \end{aligned}$$

where the fourth equation holds due to the decomposition of $\mathbf{A}^{-1}(z)$ in (4.2).

To begin with, we prove that $E(I_{1.2}) = o(1)$. Note that

$$\begin{aligned} E(I_{1.2}) &= E\left(\beta_1(z)\mathbf{r}_1^T\mathbf{A}_1^{-1}(z)\mathbf{D}_1\mathbf{A}_1^{-1}(z)\mathbf{R}^{-1/2}(\text{diag}(\mathbf{S}_1) - \mathbf{I}_p)\mathbf{R}^{1/2}\mathbf{A}_1^{-1}(z)\mathbf{r}_1\right) + \frac{1}{n}E\left(\beta_1(z)\right. \\ &\quad \left.\times \mathbf{r}_1^T\mathbf{A}_1^{-1}(z)\mathbf{R}^{-1/2}(\text{diag}(\mathbf{S}_1) - \mathbf{I}_p)\mathbf{R}^{1/2}\mathbf{A}_1^{-1}(z)\mathbf{R}^{-1/2}(\text{diag}(\mathbf{S}_1) - \mathbf{I}_p)\mathbf{R}^{1/2}\mathbf{A}_1^{-1}(z)\mathbf{r}_1\right) \\ &:= E(I_{1.2.1}) + E(I_{1.2.2}). \end{aligned}$$

Using (4.16) and Lemma 6.1, we get

$$\begin{aligned} &|E(I_{1.2.1})| \\ &\leq Cn \sum_{\ell=1}^p E\left|\mathbf{r}_1^T\mathbf{A}_1^{-1}(z)\mathbf{D}_1\mathbf{A}_1^{-1}(z)\mathbf{R}^{-1/2}\mathbf{e}_\ell\mathbf{e}_\ell^T\mathbf{R}^{1/2}\mathbf{A}_1^{-1}(z)\mathbf{r}_1\left(\mathbf{r}_1^T\mathbf{g}_\ell\mathbf{g}_\ell^T\mathbf{r}_1 - \frac{1}{n}\right)\right| \\ &\leq Cn \sum_{\ell=1}^p E\left|\left(\mathbf{r}_1^T\mathbf{A}_1^{-1}(z)\mathbf{D}_1\mathbf{A}_1^{-1}(z)\mathbf{R}^{-1/2}\mathbf{e}_\ell\mathbf{e}_\ell^T\mathbf{R}^{1/2}\mathbf{A}_1^{-1}(z)\mathbf{r}_1 - \frac{1}{n}\mathbf{e}_\ell^T\mathbf{R}^{1/2}\mathbf{A}_1^{-2}(z)\mathbf{D}_1\right.\right. \\ &\quad \left.\left.\times \mathbf{A}_1^{-1}(z)\mathbf{R}^{-1/2}\mathbf{e}_\ell\right)\left(\mathbf{r}_1^T\mathbf{g}_\ell\mathbf{g}_\ell^T\mathbf{r}_1 - \frac{1}{n}\right)\right| + C \sum_{\ell=1}^p E\left[\|\mathbf{D}_1\|\left|\mathbf{r}_1^T\mathbf{g}_\ell\mathbf{g}_\ell^T\mathbf{r}_1 - \frac{1}{n}\right|\right] \\ &\leq Cn \sum_{\ell=1}^p E\left[E_{(1)}^{1/2}\left|\mathbf{r}_1^T\mathbf{A}_1^{-1}(z)\mathbf{D}_1\mathbf{A}_1^{-1}(z)\mathbf{R}^{-1/2}\mathbf{e}_\ell\mathbf{e}_\ell^T\mathbf{R}^{1/2}\mathbf{A}_1^{-1}(z)\mathbf{r}_1 - \frac{1}{n}\mathbf{e}_\ell^T\mathbf{R}^{1/2}\mathbf{A}_1^{-2}(z)\mathbf{D}_1\right.\right. \\ &\quad \left.\left.\times \mathbf{A}_1^{-1}(z)\mathbf{R}^{-1/2}\mathbf{e}_\ell\right|^2 E_{(1)}^{1/2}\left|\mathbf{r}_1^T\mathbf{g}_\ell\mathbf{g}_\ell^T\mathbf{r}_1 - \frac{1}{n}\right|^2\right] + C \sum_{\ell=1}^p E\|\mathbf{D}_1\|E\left|\mathbf{r}_1^T\mathbf{g}_\ell\mathbf{g}_\ell^T\mathbf{r}_1 - \frac{1}{n}\right| \\ &\leq CE\|\mathbf{D}_1\| \rightarrow 0. \end{aligned}$$

By Lemma 6.1 and Lemma 6.3, it follows that

$$\begin{aligned} |E(I_{1.2.2})| &\leq \frac{C}{n}E\left(\mathbf{r}_1^T\mathbf{r}_1\|\text{diag}(\mathbf{S}_1) - \mathbf{I}_p\|^2\right) \\ &\leq \frac{C}{n}E^{1/2}\left(\mathbf{r}_1^T\mathbf{r}_1\right)^2 E^{1/2}\|\text{diag}(\mathbf{S}_1) - \mathbf{I}_p\|^4 \\ &\leq \frac{C}{n}\max_{k=1,\dots,p}E^{1/2}\left(\mathbf{X}_j^T\mathbf{g}_k\mathbf{g}_k^T\mathbf{X}_j - 1\right)^4 \leq C\eta_n^2 \rightarrow 0. \end{aligned}$$

These two inequalities show that

$$E(I_{1.2}) = o(1).$$

Next we show that $E(I_{1.3}) = o(1)$. Note that

$$E(I_{1.3})$$

$$\begin{aligned}
 &= \mathbb{E} \left(\beta_1^2(z) \mathbf{r}_1^T \mathbf{A}_1^{-1}(z) \mathbf{D}_1 \mathbf{A}_1^{-1}(z) \mathbf{r}_1 \mathbf{r}_1^T \mathbf{A}_1^{-1}(z) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_1) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{A}_1^{-1}(z) \mathbf{r}_1 \right) \\
 &\quad + \frac{1}{n} \mathbb{E} \left(\beta_1(z) \mathbf{r}_1^T \mathbf{A}_1^{-1}(z) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_1) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{A}_1^{-1}(z) \mathbf{r}_1 \right)^2 \\
 &:= \mathbb{E}(I_{1.3.1}) + \mathbb{E}(I_{1.3.2}).
 \end{aligned}$$

Due to (4.16) and Lemma 6.1, it implies that

$$\begin{aligned}
 |\mathbb{E}(I_{1.3.1})| &\leq C \mathbb{E} \left[\left(\mathbf{r}_1^T \mathbf{r}_1 \right)^2 \|\mathbf{D}_1 (\text{diag}(\mathbf{S}_1) - \mathbf{I}_p)\| \right] \\
 &\leq C \max_{\ell=1, \dots, p} \mathbb{E} \left[\|\mathbf{D}_1\| \mathbb{E}_{(1)}^{1/2} \left| \mathbf{r}_1^T \mathbf{r}_1 \right|^4 \mathbb{E}_{(1)}^{1/2} \left| \mathbf{X}_1^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{X}_1 - 1 \right|^2 \right] \\
 &\leq C \mathbb{E} \|\mathbf{D}_1\| \rightarrow 0
 \end{aligned}$$

and

$$\begin{aligned}
 |\mathbb{E}(I_{1.3.2})| &\leq \frac{C}{n} \mathbb{E} \left[\left(\mathbf{r}_1^T \mathbf{r}_1 \right)^2 \|(\text{diag}(\mathbf{S}_1) - \mathbf{I}_p)\|^2 \right] \\
 &\leq \frac{C}{n} \max_{\ell=1, \dots, p} \mathbb{E} \left[\mathbb{E}_{(1)}^{1/2} \left| \mathbf{r}_1^T \mathbf{r}_1 \right|^4 \mathbb{E}_{(1)}^{1/2} \left| \mathbf{X}_1^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{X}_1 - 1 \right|^4 \right] \\
 &\leq C \eta_n^2 \rightarrow 0.
 \end{aligned}$$

Hence, we have $\mathbb{E}(I_{1.3}) = o(1)$.

Now we analyze $\mathbb{E}(I_{1.1})$. Noting that $g_k^T g_k = 1$, we have that

$$\begin{aligned}
 \mathbb{E}(I_{1.1}) &= \frac{1}{n^2} \sum_{j=1}^n \text{Etr} \left((\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \right)^2 \\
 &= \frac{1}{n^2} \sum_{j=1}^n \sum_{k, \ell=1}^p \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right. \\
 &\quad \left. \times (\mathbf{X}_j^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{X}_j - 1) (\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1) \right] \\
 &= \frac{1}{n} \sum_{k, \ell=1}^p \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_1^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_1^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right. \\
 &\quad \left. \times (\mathbf{X}_j^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{X}_j - 1) (\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1) \right].
 \end{aligned}$$

Further note that

$$\mathbb{E} \left((\mathbf{X}_1^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{X}_1 - 1) (\mathbf{X}_1^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_1 - 1) \right) = \beta_x \sum_{j=1}^p g_{\ell j}^2 g_{k j}^2 + 2r_{k\ell}^2, \quad (4.28)$$

where $\beta_x = \mathbb{E}(X_{11}^4) - 3$ and $r_{k\ell} = \mathbf{g}_k^T \mathbf{g}_\ell$. Therefore

$$\begin{aligned} \mathbb{E}(I_{1.1}) &= \frac{1}{n} \sum_{k=1}^p \sum_{\ell=1}^p \left(\beta_x \sum_{j=1}^p g_{\ell j}^2 g_{kj}^2 + 2r_{k\ell}^2 \right) \\ &\quad \times \mathbb{E} \left(\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_1^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_1^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right). \end{aligned}$$

Using the decomposition of $\mathbf{A}_1^{-1}(z)$ in (4.5), we then get

$$\begin{aligned} \mathbb{E}(I_{1.1}) &= \frac{1}{n} \left(z - \frac{n-1}{n} b(z) \right)^{-2} \left(2p + \beta_x \sum_{k=1}^p \sum_{j=1}^p g_{kj}^4 \right) \\ &\quad - \frac{2}{n(z - (n-1)/n \cdot b(z))} \sum_{k=1}^p \sum_{m=2}^4 \left(2 + \beta_x \sum_{j=1}^p g_{kj}^4 \right) \mathbb{E} \left(\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{B}_m(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right) \\ &\quad + \frac{2r_{k\ell}^2 + \beta_x \sum_{j=1}^p g_{kj}^2 g_{\ell j}^2}{n} \sum_{k,\ell=1}^p \sum_{u=2}^4 \sum_{v=2}^4 \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{B}_u(z) \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{B}_v(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right] \\ &:= \mathbb{E}(I_{1.1.1}) + \mathbb{E}(I_{1.1.2}) + \mathbb{E}(I_{1.1.3}). \end{aligned}$$

We first have that

$$\mathbb{E}(I_{1.1.1}) = \frac{2\rho_n + \beta_x n^{-1} \sum_{k=1}^p \sum_{j=1}^p g_{kj}^4}{z^2(1 + \underline{g}(z))^2} + o(1)$$

by (2.17) of Bai and Silverstein (2004).

Next we show that both $\mathbb{E}(I_{1.1.2})$ and $\mathbb{E}(I_{1.1.3})$ are $o(1)$. By conditioning on \mathcal{F}_1 we get

$\mathbb{E}(B_2(z)) = 0$. Furthermore, by (4.4) and Lemma 6.1, it implies that

$$\begin{aligned} &\left| \frac{1}{n} \mathbb{E} \sum_{k=1}^p \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{B}_3(z) \mathbf{R}^{-1/2} \mathbf{e}_k \cdot \left(2 + \beta_x \sum_{j=1}^p g_{kj}^4 \right) \right| \\ &\leq K \sum_{k=1}^p \left| \mathbb{E} \left[(\beta_{2(1)}(z) - b(z)) \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{21}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right] \right| \\ &\leq K \sum_{k=1}^p \left| \mathbb{E} \left[(\beta_{2(1)}(z) - b_{21}(z)) \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{21}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right] \right| \\ &\quad + K \sum_{k=1}^p \mathbb{E} \left[|b_{21}(z) - b(z)| \left| \mathbb{E}_{(2)} \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{21}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right| \right] \\ &\leq K \sum_{k=1}^p \left| \mathbb{E} \left[-b_{21}^2(z) \widehat{\gamma}_{2(1)}(z) + \beta_{2(1)}(z) b_{21}^2(z) \widehat{\gamma}_{2(1)}^2(z) \right] \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{21}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right| + o(1) \end{aligned}$$

4. PROOFS OF THEOREMS 2.1 AND 3.1

$$\begin{aligned}
&\leq K \sum_{k=1}^p \left[\mathbb{E} \left| \mathbb{E}_{(2)} \widehat{\gamma}_{2(1)}(z) \mathbf{r}_2^T \mathbf{A}_{21}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{r}_2 \right| + \mathbb{E} \left| \widehat{\gamma}_{2(1)}^2(z) \mathbf{r}_2^T \mathbf{A}_{21}^{-1}(z) \mathbf{e}_k \mathbf{e}_k^T \mathbf{r}_2 \right| \right] + o(1) \\
&\leq K \sum_{k=1}^p \mathbb{E} \left| \mathbb{E}_{(2)} \left[\widehat{\gamma}_{2(1)}(z) (\mathbf{r}_2^T \mathbf{A}_{21}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{r}_2 - n^{-1} \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{21}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k) \right] \right| \\
&\quad + K \sum_{k=1}^p \mathbb{E} \left| \widehat{\gamma}_{2(1)}^2(z) (\mathbf{r}_2^T \mathbf{A}_{21}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{r}_2 - n^{-1} \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{21}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k) \right| \\
&\quad + \frac{K}{n} \sum_{k=1}^p \mathbb{E} \left| [\widehat{\gamma}_{2(1)}(z)]^2 \right| + o(1) \\
&= o(1),
\end{aligned}$$

where K is a constant. Finally, by a similar decomposition to (4.2) we get that

$$\begin{aligned}
&\left| \frac{1}{n} \sum_{k=1}^p \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{B}_4(z) \mathbf{R}^{-1/2} \mathbf{e}_k \cdot \left(2 + \beta_x \sum_{j=1}^p g_{kj}^4 \right) \right] \right| \\
&\leq \frac{K}{n^2} \sum_{k=1}^p \left| \sum_{i \geq 2} \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{A}_{i1}^{-1}(z) - \mathbf{A}_1^{-1}(z)) \mathbf{R}^{-1/2} \mathbf{e}_k \right] \right| \\
&\leq \frac{K}{n^2} \sum_{k=1}^p \left| \sum_{i \geq 2} \mathbb{E} \left[\beta_{i(1)}(z) \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{i1}^{-1}(z) \mathbf{r}_i \mathbf{r}_i^T \mathbf{A}_{i1}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right] \right| \\
&\leq \frac{K}{n} \sum_{k=1}^p \left| \mathbb{E} \left[\beta_{2(1)}(z) \mathbf{r}_2^T \mathbf{A}_{21}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{21}^{-1}(z) \mathbf{r}_2 \right] \right| \\
&\leq \frac{K}{n} \sum_{k=1}^p \left| \mathbb{E} b_{2(1)}(z) \beta_{2(1)}(z) \widehat{\gamma}_{2(1)}(z) \mathbf{r}_2^T \mathbf{A}_{21}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{21}^{-1}(z) \mathbf{r}_2 \right| + o(1) \\
&\leq \frac{K}{n} \sum_{k=1}^p \mathbb{E} \left| \widehat{\gamma}_{2(1)}(z) (\mathbf{r}_2^T \mathbf{A}_{21}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{21}^{-1}(z) \mathbf{r}_2 - n^{-1} \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{21}^{-2}(z) \mathbf{R}^{-1/2} \mathbf{e}_k) \right| \\
&\quad + \frac{K}{n} \mathbb{E} \left| \widehat{\gamma}_{2(1)}(z) \right| + o(1) \\
&\leq \frac{K}{n} \sum_{k=1}^p \mathbb{E}^{1/2} \left| \mathbf{r}_2^T \mathbf{A}_{21}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{21}^{-1}(z) \mathbf{r}_2 - \frac{1}{n} \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{21}^{-2}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right| \\
&\quad \times \mathbb{E}^{1/2} \left| \widehat{\gamma}_{2(1)}(z) \right|^2 + o(1) \\
&= o(1).
\end{aligned}$$

To sum up, we proved that $E(I_{1.1.2}) = o(1)$.

We now show that $E(I_{1.1.3}) = o(1)$. Denoting $d_{kl} = 2r_{k\ell}^2 + \beta_x \sum_{h=1}^p g_{kh}^2 g_{\ell h}^2$, we have

$$\left| \frac{1}{n} \sum_{k, \ell=1}^p d_{kl} \mathbb{E} \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{B}_2(z) \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{B}_2(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right|$$

$$\begin{aligned}
 &\leq \frac{K}{n} \left| \sum_{k,\ell=1}^p \sum_{i,j=2}^n d_{kl} \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_i \mathbf{r}_i^T - n^{-1} \mathbf{I}) \mathbf{A}_{i1}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} (\mathbf{r}_j \mathbf{r}_j^T - n^{-1} \mathbf{I}) \mathbf{A}_{j1}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \right] \right| \\
 &\leq K \left| \sum_{k,\ell=1}^p d_{kl} \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{21}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{21}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \right] \right| \\
 &\quad + Kn \left| \sum_{k,\ell=1}^p d_{kl} \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{21}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} (\mathbf{r}_3 \mathbf{r}_3^T - n^{-1} \mathbf{I}) \mathbf{A}_{31}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \right] \right| \\
 &\leq \frac{K}{n^2} \sum_{k,\ell=1}^p d_{kl} + Kn \left| \sum_{k,\ell=1}^p d_{kl} \mathbb{E} \left[\beta_{2(31)} \beta_{3(21)} \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-1} \mathbf{r}_3 \mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \right. \right. \\
 &\quad \left. \left. \times (\mathbf{r}_3 \mathbf{r}_3^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-1} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \right] \right| \\
 &\leq Kn \left| \sum_{k,\ell=1}^p d_{kl} \mathbb{E} \left[\beta_{2(31)} \beta_{3(21)} \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-1} \mathbf{r}_3 \mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \right. \right. \\
 &\quad \left. \left. \times (\mathbf{r}_3 \mathbf{r}_3^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-1} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \right] \right| + o(1).
 \end{aligned}$$

Note that

$$\beta_{2(31)}(z) = b_{321}(z) - \beta_{2(31)}(z) b_{321}(z) \widehat{\gamma}_{2(31)}(z)$$

and

$$\beta_{3(21)}(z) = b_{321}(z) - \beta_{3(21)}(z) b_{321}(z) \widehat{\gamma}_{3(21)}(z).$$

Using (1.15) of Bai and Silverstein (2004) many times, we have for each $k, l = 1, \dots, p$,

$$\begin{aligned}
 &\mathbb{E} \left[b_{321}^2(z) \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-1} \mathbf{r}_3 \mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} (\mathbf{r}_3 \mathbf{r}_3^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-1} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \right] \\
 &= \mathbb{E} \left[b_{321}^2(z) \mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-1} \mathbf{r}_3 \mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{r}_3 \right] \\
 &\quad - \frac{1}{n^2} \mathbb{E} \left[b_{321}^2(z) \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{r}_2 (\mathbf{r}_2^T \mathbf{A}_{321}^{-2} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{r}_2 - n^{-1} \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-2} \mathbf{R}^{-1/2} \mathbf{e}_\ell) \right] \\
 &= \mathbb{E} \left[b_{321}^2(z) \left(\mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-1} \mathbf{r}_3 - \frac{1}{n} \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \right. \right. \\
 &\quad \left. \left. \cdot \mathbf{A}_{321}^{-2} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \right) \cdot \left(\mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{r}_3 - \frac{1}{n} \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \right) \right] \\
 &\quad + \frac{1}{n^2} \mathbb{E} \left[b_{321}^2(z) \left(\mathbf{r}_2^T \mathbf{A}_{321}^{-2} \mathbf{r}_2 - \frac{1}{n} \text{tr} \left(\mathbf{A}_{321}^{-2} \right) \right) \left(\mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{r}_2 - \frac{1}{n} \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \right) \right. \\
 &\quad \left. \cdot \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \right] + \frac{1}{n^4} \mathbb{E} \left[b_{321}^2(z) \text{tr} \left(\mathbf{A}_{321}^{-2} \right) \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \right] \\
 &\quad - \frac{1}{n^4} \mathbb{E} \left[b_{321}^2(z) \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-3} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \right]
 \end{aligned}$$

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$$\begin{aligned}
& -\frac{1}{n^2} \mathbb{E} \left[b_{321}^2(z) \left(\mathbf{r}_2^T \mathbf{A}_{321}^{-2} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{r}_2 - n^{-1} \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-2} \mathbf{R}^{-1/2} \mathbf{e}_\ell \right)^2 \right] \\
= & \mathbb{E} \left[b_{321}^2(z) \left(\mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-1} \mathbf{r}_3 - \frac{1}{n} \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \right. \right. \\
& \quad \left. \left. \cdot \mathbf{A}_{321}^{-2} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \right) \cdot \left(\mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{r}_3 - \frac{1}{n} \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \right) \right] \\
& + O(n^{-3}) \\
= & \frac{1}{n^2} \mathbb{E} \left[b_{321}^2(z) \left(\mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-2} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{r}_2 \right. \right. \\
& \quad \left. \left. + \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-2} \mathbf{r}_2 \right. \right. \\
& \quad \left. \left. + \beta_x \sum_{h=1}^p \mathbf{e}_h^T \mathbf{A}_{321}^{-1} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-1} \mathbf{e}_h \mathbf{e}_h^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{e}_h \right) \right] \\
& + O(n^{-3}) \\
= & \frac{1}{n^2} \mathbb{E} \left[b_{321}^2(z) \left(2 \left(\mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{r}_2 - \frac{1}{n} \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \right) \right. \right. \\
& \quad \left. \left. \cdot \left(\mathbf{r}_2^T \mathbf{A}_{321}^{-2} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{r}_2 - \frac{1}{n} \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-3} \mathbf{R}^{-1/2} \mathbf{e}_\ell \right) \right. \right. \\
& \quad \left. \left. + \beta_x \sum_{h=1}^p \mathbf{e}_h^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{e}_h \left(\mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{r}_2 - \frac{1}{n} \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \right) \right. \right. \\
& \quad \left. \left. \cdot \left(\mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{e}_h \mathbf{e}_h^T \mathbf{A}_{321}^{-1} \mathbf{r}_2 - \frac{1}{n} \mathbf{e}_h^T \mathbf{A}_{321}^{-2} \mathbf{e}_h \right) \right) \right] + O(n^{-3}) \\
= & O(n^{-3}).
\end{aligned}$$

By Lemma 6.1, it follows that for each $k, l = 1, \dots, p$,

$$\begin{aligned}
& \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}_p) \mathbf{A}_{321}^{-1} \mathbf{r}_3 \mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} (\mathbf{r}_3 \mathbf{r}_3^T - n^{-1} \mathbf{I}_p) \mathbf{A}_{321}^{-1} \mathbf{r}_2 \right. \\
& \quad \left. \cdot \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \beta_{231}(z) b_{321}(z) \widehat{\gamma}_{2(31)}(z) \right] \\
\leq & K \mathbb{E} \left| \widehat{\gamma}_{2(31)}(z) \left(\mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{r}_3 - \frac{1}{n} \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \right) \right. \\
& \quad \left. \cdot \left(\mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - \frac{1}{n} \mathbf{I}_p) \mathbf{A}_{321}^{-1} \mathbf{r}_3 \right. \right. \\
& \quad \left. \left. - \frac{1}{n} \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - \frac{1}{n} \mathbf{I}_p) \mathbf{A}_{321}^{-2} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \right) \right| \\
& + \frac{K}{n} \mathbb{E} \left| \widehat{\gamma}_{2(31)}(z) \left(\mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - \frac{1}{n} \mathbf{I}_p) \mathbf{A}_{321}^{-1} \mathbf{r}_3 \right. \right. \\
& \quad \left. \left. - \frac{1}{n} \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - \frac{1}{n} \mathbf{I}_p) \mathbf{A}_{321}^{-2} \mathbf{r}_2 \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \right) \right| \\
& + \frac{K}{n^2} \mathbb{E} \left| \widehat{\gamma}_{2(31)}(z) \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-2} \mathbf{r}_2 \right|
\end{aligned}$$

$$\begin{aligned}
 & + \frac{K}{n} \mathbb{E} \left| \widehat{\gamma}_{2(31)}(z) \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{r}_2 \left(\mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \right. \right. \\
 & \quad \left. \left. \cdot \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-1} \mathbf{r}_3 - \frac{1}{n} \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-1} \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \right) \right| \\
 & + \frac{K}{n^2} \mathbb{E} \left| \widehat{\gamma}_{2(31)}(z) \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_{321}^{-1} \mathbf{r}_2 \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}) \mathbf{A}_{321}^{-2} \mathbf{R}^{-1/2} \mathbf{e}_\ell \right| \\
 & \leq \frac{K}{n^2} \mathbb{E} \left| \widehat{\gamma}_{2(31)}(z) \mathbf{r}_2^T \mathbf{r}_2 \mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - \frac{1}{n} \mathbf{I}_p)^2 \mathbf{R}^{-1/2} \mathbf{e}_k \right| \\
 & + \frac{K}{n^2} \mathbb{E} \left| \widehat{\gamma}_{2(31)}(z) \mathbf{r}_2^T \mathbf{r}_2 \sqrt{\mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I})^2 \mathbf{R}^{-1/2} \mathbf{e}_k} \right| \\
 & \leq \frac{K}{n^2} \mathbb{E}^{1/2} |\widehat{\gamma}_{2(31)}(z)|^2 = O(n^{-5/2}).
 \end{aligned}$$

Furthermore, for each $k, l = 1, \dots, p$,

$$\begin{aligned}
 & \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}_p) \mathbf{A}_{321}^{-1} \mathbf{r}_3 \mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} (\mathbf{r}_3 \mathbf{r}_3^T - n^{-1} \mathbf{I}_p) \mathbf{A}_{321}^{-1} \mathbf{r}_2 \right. \\
 & \quad \left. \cdot \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \beta_{321}(z) b_{321}(z) \widehat{\gamma}_{3(21)}(z) \right] = O(n^{-5/2})
 \end{aligned}$$

and

$$\begin{aligned}
 & \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} (\mathbf{r}_2 \mathbf{r}_2^T - n^{-1} \mathbf{I}_p) \mathbf{A}_{321}^{-1} \mathbf{r}_3 \mathbf{r}_3^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} (\mathbf{r}_3 \mathbf{r}_3^T - n^{-1} \mathbf{I}_p) \mathbf{A}_{321}^{-1} \mathbf{r}_2 \right. \\
 & \quad \left. \cdot \mathbf{r}_2^T \mathbf{A}_{321}^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \beta_{321}(z) \beta_{231}(z) b_{321}^2(z) \widehat{\gamma}_{3(21)}(z) \widehat{\gamma}_{2(31)}(z) \right] = O(n^{-5/2}).
 \end{aligned}$$

Hence, we conclude that

$$\left| \frac{1}{n} \sum_{k, \ell=1}^p d_{kl} \mathbb{E} \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{B}_2(z) \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{B}_2(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right| \leq Kn \sum_{k, \ell=1}^p d_{kl} + o(1) = o(1).$$

Similarly, we have

$$\frac{1}{n} \sum_{k, \ell=1}^p d_{kl} \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{B}_u(z) \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{B}_v(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right] = o(1), \quad \text{for } u, v = 2, 3, 4.$$

Combining the estimates above we get the desired convergence that

$$\mathbb{E}(I_1) = \frac{2\rho_n + \beta_x \frac{1}{n} \sum_{k=1}^p \sum_{j=1}^p g_{kj}^4}{z^2(1 + \underline{s}(z))^2} + o(1) \tag{4.29}$$

$$\rightarrow \frac{1}{z^2(1 + \underline{s}(z))^2} \left(2\rho + \beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{j=1}^p g_{kj}^4 \right).$$

Repeating the proof of (4.29), it implies that

$$\mathbb{E}(I_2) \rightarrow \frac{1}{z^2(1 + \underline{s}(z))^2} \left(2\rho + \beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{j=1}^p g_{kj}^4 \right)$$

and

$$\begin{aligned} & \mathbb{E}(I_3) \\ &= \frac{1}{n} \sum_{j=1}^n \text{Etr} \left[\mathbf{A}_j^{-1}(z) \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \right] + o(1) \\ &= \frac{1}{n^2} \sum_{j=1}^n \sum_{k, \ell=1}^p \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{-1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{1/2} \mathbf{e}_k \right. \\ & \quad \left. \cdot \left(\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1 \right) \left(\mathbf{X}_j^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{X}_j - 1 \right) \right] + o(1) \\ &= \frac{1}{n^2} \sum_{j=1}^n \sum_{k, \ell=1}^p \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{-1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_\ell \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{1/2} \mathbf{e}_k \right. \\ & \quad \left. \cdot \left(2r_{kl}^2 + \beta_x \sum_{h=1}^p g_{kh}^2 g_{\ell h}^2 \right) \right] + o(1) \\ & \rightarrow \frac{1}{z^2(1 + \underline{s}(z))^2} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k, \ell=1}^p \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{-1} \mathbf{e}_\ell r_{kl} \left(2r_{kl}^2 + \beta_x \sum_{h=1}^p g_{kh}^2 g_{\ell h}^2 \right) \right]. \end{aligned}$$

Step $\mathcal{I}.4$ By Step $\mathcal{I}.1$ - $\mathcal{I}.3$, we conclude that

$$\begin{aligned} z \cdot I & \rightarrow \frac{1}{2z(1 + \underline{s}(z))^2} \left(2\rho + \beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{j=1}^p g_{kj}^4 \right) \\ & + \frac{1}{2z(1 + \underline{s}(z))^2} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k, \ell=1}^p \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{-1} \mathbf{e}_\ell r_{kl} \left(2r_{kl}^2 + \beta_x \sum_{h=1}^p g_{kh}^2 g_{\ell h}^2 \right) \right]. \end{aligned} \tag{4.30}$$

Step $\mathcal{I}.5$ Since

$$II = \frac{1}{2} \cdot \frac{\partial I}{\partial z},$$

it is apparent that

$$\begin{aligned}
 & z^2 \cdot II \\
 & \xrightarrow{p} \frac{z^2}{4} \frac{\partial}{\partial z} \left(\frac{1}{z^2(1+\underline{s}(z))^2} \right) \cdot \left(2\rho + \beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{j=1}^p g_{kj}^A \right) \\
 & \quad + \frac{z^2}{4} \frac{\partial}{\partial z} \left(\frac{1}{z^2(1+\underline{s}(z))^2} \right) \cdot \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k,\ell=1}^p \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{-1} \mathbf{e}_\ell r_{kl} \left(2r_{kl}^2 + \beta_x \sum_{h=1}^p g_{kh}^2 g_{\ell h}^2 \right) \right],
 \end{aligned} \tag{4.31}$$

which is similar to the proof of (4.30).

Step II In this step, we are going to showing that

$$III + IV = o_p(1).$$

Firstly, it can be verified that from Lemma 6.2

$$\|\mathbf{H}^{-1}(z)\| \leq \frac{1}{\Im z} \quad \text{a.s.}$$

Using (4.15) and Lemma 6.4, one has

$$\begin{aligned}
 |III + IV| & \leq C \|\mathbf{H}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)\| \operatorname{tr} (\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p))^2 \\
 & \leq C \|\mathbf{L} - \mathbf{I}_p\| \leq C \|\mathbf{D}\| \rightarrow 0 \quad \text{in probability.}
 \end{aligned}$$

Therefore, we conclude that

$$III + IV = o_p(1). \tag{4.32}$$

Step III . In this step we derive the CLT of

$$\begin{aligned}
 V & = [\operatorname{tr} \mathbf{A}^{-1}(z) - \mathbb{E}(\operatorname{tr} \mathbf{A}^{-1}(z))] + [\operatorname{tr} (\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)) - \mathbb{E} \operatorname{tr} (\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p))] \\
 & \quad + z [\operatorname{tr} (\mathbf{A}^{-2}(z) (\mathbf{L} - \mathbf{I}_p)) - \mathbb{E} \operatorname{tr} (\mathbf{A}^{-2}(z) (\mathbf{L} - \mathbf{I}_p))] \\
 & := V.1 + V.2 + V.3.
 \end{aligned}$$

The outline of this step is as follows:

4. PROOFS OF THEOREMS 2.1 AND 3.1

III.1: Prove $V.1 = \sum_{j=1}^n W_j(z) + o_p(1)$ where $W_j(z) = -E_j \left(\frac{d}{dz} b_j(z) \hat{\gamma}_j(z) \right)$.

III.2: Show that

$$V.2 = \sum_{j=1}^n Q_{j1}(z) + o_p(1)$$

and

$$V.3 = z \cdot \frac{\partial}{\partial z} \sum_{j=1}^n Q_{j1}(z) + o_p(1)$$

where $Q_{j1}(z) = \frac{1}{n} E_j \text{tr} \left(\mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \right)$.

III.3: Show that the process indexed by z

$$\left(\sum_{j=1}^n \left(W_j(z) + Q_{j1}(z) + z \cdot \frac{\partial}{\partial z} Q_{j1}(z) \right) \right)$$

converges weakly to a Gaussian process.

Step III.1 According to Bai and Silverstein (2004) (Page 562), we obtain

$$V.1 = \text{tr} \mathbf{A}^{-1}(z) - \text{Etr} \left(\mathbf{A}^{-1}(z) \right) = \sum_{j=1}^n W_j(z) + o_p(1), \quad (4.33)$$

where $W_j(z) = -E_j \left(\frac{d}{dz} b_j(z) \hat{\gamma}_j(z) \right)$. We note here that Zheng, Bai and Yao (2015) also obtain

that $\text{tr} \mathbf{A}^{-1}(z) - p s_{\rho_n}(z)$ converges to a Gaussian process with the following mean function

$$\frac{\rho \underline{s}^3(z) (1 + \underline{s}(z))^{-3}}{(1 - \rho \underline{s}^2(z) (1 + \underline{s}(z))^{-2})^2} + \frac{\beta_x \cdot \rho \underline{s}^3(z) (1 + \underline{s}(z))^{-3}}{1 - \rho \underline{s}^2(z) (1 + \underline{s}(z))^{-2}}$$

and the covariance function

$$2 \left[\frac{\frac{\partial \underline{s}(z_1)}{\partial z_1} \frac{\partial \underline{s}(z_2)}{\partial z_2}}{(\underline{s}(z_2) - \underline{s}(z_1))^2} - \frac{1}{(z_1 - z_2)^2} \right] + \frac{\beta_x \rho \frac{\partial \underline{s}(z_1)}{\partial z_1} \frac{\partial \underline{s}(z_2)}{\partial z_2}}{(1 + \underline{s}(z_1))^2 (1 + \underline{s}(z_2))^2}.$$

Step III.2 It follows that from (4.17) that

$$\begin{aligned} V.2 &= \frac{1 + o(1)}{2} \left[\text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \right) - \text{Etr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \right) \right] \\ &\quad + \frac{1 + o(1)}{2} \left[\text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D}^T \right) - \text{Etr} \left(\mathbf{A}^{-1}(z) \mathbf{D}^T \right) \right] \end{aligned}$$

$$+ \frac{1+o(1)}{4} \left[\text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \mathbf{D}^T \right) - \text{Etr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \mathbf{D}^T \right) \right].$$

Similar to the proof of (4.26), we see that

$$\frac{1+o(1)}{4} \left[\text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \mathbf{D}^T \right) - \text{Etr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \mathbf{D}^T \right) \right] = o_p(1).$$

Hence, one finds that

$$\begin{aligned} V.2 &= \frac{1+o(1)}{2} \left[\text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \right) - \text{Etr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \right) \right] \\ &\quad + \frac{1+o(1)}{2} \left[\text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D}^T \right) - \text{Etr} \left(\mathbf{A}^{-1}(z) \mathbf{D}^T \right) \right] + o_p(1) \\ &:= \frac{1+o(1)}{2} V.2.1 + \frac{1+o(1)}{2} V.2.2 + o_p(1). \end{aligned}$$

Using (4.2), we obtain that

$$\begin{aligned} V.2.1 &= \text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \right) - \text{E} \left(\text{tr} \mathbf{A}^{-1}(z) \mathbf{D} \right) \\ &= \sum_{j=1}^n (\mathbf{E}_j - \mathbf{E}_{j-1}) \left(\text{tr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \right) - \text{tr} \left(\mathbf{A}_j^{-1}(z) \mathbf{D}_j \right) \right) \\ &= \frac{1}{n} \sum_{j=1}^n (\mathbf{E}_j - \mathbf{E}_{j-1}) \text{tr} \left(\mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \left(\text{diag} \left(\mathbf{S}_j \right) - \mathbf{I}_p \right) \mathbf{R}^{1/2} \right) \\ &\quad - \sum_{j=1}^n (\mathbf{E}_j - \mathbf{E}_{j-1}) \left(\beta_j(z) \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j \right) \\ &\quad - \frac{1}{n} \sum_{j=1}^n (\mathbf{E}_j - \mathbf{E}_{j-1}) \left(\beta_j(z) \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \left(\text{diag} \left(\mathbf{S}_j \right) - \mathbf{I}_p \right) \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{r}_j \right) \\ &= \sum_{j=1}^n (Q_{j1}(z) - Q_{j2}(z) - Q_{j3}(z)). \end{aligned} \tag{4.34}$$

We shall show that both $\sum_{j=1}^n Q_{j2}(z)$ and $\sum_{j=1}^n Q_{j3}(z)$ are $o_p(1)$.

We start with $\sum_{j=1}^n Q_{j2}(z)$. By (4.3), we have

$$\begin{aligned} \sum_{j=1}^n Q_{j2}(z) &= \sum_{j=1}^n \mathbf{E}_j \left[b_j(z) \left(\mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j - \frac{1}{n} \text{tr} \left(\mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \right) \right) \right] \\ &\quad - \sum_{j=1}^n (\mathbf{E}_j - \mathbf{E}_{j-1}) \left(b_j^2(z) \widehat{\gamma}_j(z) \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j \right) \end{aligned}$$

$$+ \sum_{j=1}^n (\mathbf{E}_j - \mathbf{E}_{j-1}) \left(b_j^2(z) \beta_j(z) \widehat{\gamma}_j^2(z) \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j \right).$$

Using (4.16) and Lemma 6.1, it follows that

$$\begin{aligned} & \mathbf{E} \left| \sum_{j=1}^n (\mathbf{E}_j - \mathbf{E}_{j-1}) \left(b_j^2(z) \widehat{\gamma}_j(z) \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j \right) \right|^2 \\ & \leq C \sum_{j=1}^n \mathbf{E} \left| \widehat{\gamma}_j(z) \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j \right|^2 \\ & \leq C \sum_{j=1}^n \mathbf{E} \left| \widehat{\gamma}_j(z) \left(\mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j - \frac{1}{n} \text{tr}(\mathbf{A}_j^{-2}(z) \mathbf{D}_j) \right) \right|^2 \\ & \quad + \frac{C}{n^2} \sum_{j=1}^n \mathbf{E} \left| \widehat{\gamma}_j(z) \text{tr}(\mathbf{A}_j^{-2}(z) \mathbf{D}_j) \right|^2 \\ & \leq C \sum_{j=1}^n \mathbf{E} \left[\mathbf{E}_{(j)}^{1/2} |\widehat{\gamma}_j(z)|^4 \mathbf{E}_{(j)}^{1/2} \left| \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j - \frac{1}{n} \text{tr}(\mathbf{A}_j^{-2}(z) \mathbf{D}_j) \right|^4 \right] \\ & \quad + \frac{C}{n^2} \sum_{j=1}^n \mathbf{E} \|\mathbf{D}_j\|^2 \leq \frac{C\eta_n^4}{n} \sum_{j=1}^n \mathbf{E} \|\mathbf{D}_j\|^2 + \frac{C}{n^2} \sum_{j=1}^n \mathbf{E} \|\mathbf{D}_j\|^2 \rightarrow 0, \end{aligned}$$

and

$$\mathbf{E} \left| \sum_{j=1}^n (\mathbf{E}_j - \mathbf{E}_{j-1}) \left(b_j^2(z) \beta_j(z) \widehat{\gamma}_j^2(z) \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j \right) \right| \rightarrow 0.$$

Furthermore,

$$\begin{aligned} & \mathbf{E} \left| \sum_{j=1}^n \mathbf{E}_j \left[b_j(z) \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j - \frac{1}{n} \text{tr}(\mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z)) \right] \right|^2 \\ & \leq C \sum_{j=1}^n \mathbf{E} \left| \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z) \mathbf{r}_j - \frac{1}{n} \text{tr}(\mathbf{A}_j^{-1}(z) \mathbf{D}_j \mathbf{A}_j^{-1}(z)) \right|^2 \\ & \leq \frac{C}{n} \sum_{j=1}^n \mathbf{E} \|\mathbf{D}_j\|^2 \rightarrow 0. \end{aligned}$$

That is, $\sum_{j=1}^n Q_{j2}(z) = o_p(1)$.

Next we prove that $\sum_{j=1}^n Q_{j3}(z) = o_p(1)$. Applying Lemma 6.1 and Lemma 6.3, one finds

that

$$\mathbf{E} \left| \sum_{j=1}^n Q_{j3}(z) \right|^2 \leq \frac{C}{n^2} \sum_{j=1}^n \mathbf{E} \left[\mathbf{r}_j^T \mathbf{r}_j \|\text{diag}(\mathbf{S}_j) - \mathbf{I}_p\|^2 \right]$$

$$\begin{aligned} &\leq \frac{C}{n^2} \sum_{j=1}^n \mathbb{E}^{1/2} \left| \mathbf{r}_j^T \mathbf{r}_j \right|^4 \mathbb{E}^{1/2} \|\text{diag}(\mathbf{S}_j) - \mathbf{I}_p\|^4 \\ &\leq C\eta_n^2 \rightarrow 0. \end{aligned}$$

Consequently, it follows that

$$\sum_{j=1}^n Q_{j3}(z) = o_p(1).$$

Combining the two estimates with (4.34), we then obtain that

$$V.2.1 = \sum_{j=1}^n Q_{j1}(z) + o_p(1).$$

Using the same method of the proof of the above inequality, it yields that

$$\begin{aligned} V.2.2 &= \frac{1}{n} \sum_{j=1}^n (\mathbb{E}_j - \mathbb{E}_{j-1}) \text{tr} \left(\mathbf{A}_j^{-1}(z) \mathbf{R}^{1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{-1/2} \right) + o_p(1) \\ &= \sum_{j=1}^n Q_{j1}(z) + o_p(1). \end{aligned}$$

Consequently, it follows that

$$V.2 = (1 + o(1)) \sum_{j=1}^n Q_{j1}(z) + o_p(1).$$

Note that from (4.28)

$$\begin{aligned} \mathbb{E} \left| \sum_{j=1}^n Q_{j1}(z) \right|^2 &\leq \frac{C}{n^2} \sum_{j=1}^n \mathbb{E} \left| \sum_{k=1}^p \left(\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1 \right) \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right|^2 \\ &= \frac{C}{n^2} \sum_{j=1}^n \sum_{k, \ell=1}^p \mathbb{E} \left[\left(\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1 \right) \left(\mathbf{X}_j^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{X}_j - 1 \right) \right] \\ &\quad \cdot \mathbb{E} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_\ell \right] \\ &\leq \frac{C}{n^2} \sum_{j=1}^n \sum_{k, \ell=1}^p \mathbb{E} \left[\left(\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1 \right) \left(\mathbf{X}_j^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{X}_j - 1 \right) \right] \\ &= \frac{C}{n^2} \sum_{j=1}^n \sum_{k, \ell=1}^p \mathbb{E} \left[2r_{k\ell}^2 + \beta_x \sum_{h=1}^p g_{\ell h}^2 g_{kh}^2 \right] \leq C. \end{aligned}$$

Hence, we conclude that

$$V.2 = \sum_{j=1}^n Q_{j1}(z) + o_p(1). \quad (4.35)$$

Because $V.3 = z \frac{\partial}{\partial z} V.2$, we have

$$V.3 = z \sum_{j=1}^n \frac{\partial}{\partial z} Q_{j1}(z) + o_p(1)$$

similar to the proof of (4.35). Combining (4.33) and (4.35) yields

$$V = \sum_{j=1}^n \left[W_j(z) + Q_{j1}(z) + z \cdot \frac{\partial}{\partial z} Q_{j1}(z) \right] + o_p(1).$$

Step III.3 Next we show that

$$\sum_{j=1}^n \left[W_j(z) + Q_{j1}(z) + z \cdot \frac{\partial}{\partial z} Q_{j1}(z) \right]$$

converges weakly to a Gaussian process in z .

For any $z_1, \dots, z_r \in \mathbf{C}_+$, $\alpha_1, \dots, \alpha_r \in \mathbb{R}$ and any $\varepsilon > 0$, we have

$$\begin{aligned} & \sum_{j=1}^n E \left(\left| \sum_{\ell=1}^r \alpha_\ell Q_{j1}(z_\ell) \right|^2 I \left(\left| \sum_{\ell=1}^r \alpha_\ell Q_{j1}(z_\ell) \right| \geq \varepsilon \right) \right) \\ & \leq \frac{1}{\varepsilon^2} \sum_{j=1}^n E \left| \sum_{\ell=1}^r \alpha_\ell Q_{j1}(z_\ell) \right|^4 \\ & = \frac{1}{n^4 \varepsilon^2} \sum_{j=1}^n E \left| \sum_{\ell=1}^r \alpha_\ell E_j \operatorname{tr} \left(\mathbf{A}_j^{-1}(z_\ell) \mathbf{R}^{-1/2} (\operatorname{diag}(\mathbf{S}_j) - \mathbf{I}) \mathbf{R}^{1/2} \right) \right|^4 \\ & = \frac{1}{n^4 \varepsilon^2} \sum_{j=1}^n E \left| \sum_{\ell=1}^r \alpha_\ell E_j \sum_{k=1}^p \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_\ell) \mathbf{R}^{-1/2} \mathbf{e}_k \cdot (\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1) \right|^4 \\ & \leq \frac{C}{n^4 \varepsilon^2} \sum_{j=1}^n \sum_{\ell=1}^r \alpha_\ell^4 \cdot E \left| E_j \sum_{k=1}^p \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_\ell) \mathbf{R}^{-1/2} \mathbf{e}_k \cdot (\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1) \right|^4 \\ & \leq \frac{C}{n^4 \varepsilon^2} \sum_{j=1}^n \sum_{\ell=1}^r \alpha_\ell^4 \cdot E \left| \mathbf{X}_j^T \left(\sum_{k=1}^p \mathbf{g}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_\ell) \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{g}_k^T \right) \mathbf{X}_j - \operatorname{tr}(\mathbf{A}_j^{-1}(z_\ell)) \right|^4 \\ & \leq \frac{C}{\varepsilon^2} \sum_{\ell=1}^r \alpha_\ell^4 \cdot \eta_n^4 = o(1), \end{aligned}$$

where the second to the last inequality hold due to Jensen's inequality, and in the last inequality

we used the bound (9.9.6) on Page 271 of Bai and Silverstein (2010) applied with $q = 4$ and

the fact that

$$\begin{aligned} \text{tr} \left(\sum_{k=1}^p \mathbf{g}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_\ell) \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{g}_k^T \right) &= \sum_{k=1}^p \mathbf{g}_k^T \mathbf{g}_k \cdot \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_\ell) \mathbf{R}^{-1/2} \mathbf{e}_k \\ &= \sum_{k=1}^p \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_\ell) \mathbf{R}^{-1/2} \mathbf{e}_k = \text{tr} \left(\mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_\ell) \mathbf{R}^{-1/2} \right) = \text{tr}(\mathbf{A}_j^{-1}(z_\ell)). \end{aligned}$$

The other terms have the similar results.

Next we derive the limit of the quadratic variation process, which is a sum involving the following six processes:

$$\sum_{j=1}^n \mathbb{E}_{j-1} (W_j(z_1) W_j(z_2)), \quad \sum_{j=1}^n \mathbb{E}_{j-1} (Q_{j1}(z_1) Q_{j1}(z_2)), \quad \sum_{j=1}^n \mathbb{E}_{j-1} \left(\frac{\partial Q_{j1}(z_1)}{\partial z_1} \frac{\partial Q_{j1}(z_2)}{\partial z_2} \right),$$

and

$$\sum_{j=1}^n \mathbb{E}_{j-1} (W_j(z_1) Q_{j1}(z_2)), \quad \sum_{j=1}^n \mathbb{E}_{j-1} \left(W_j(z_1) \frac{\partial Q_{j1}(z_2)}{\partial z_2} \right), \quad \sum_{j=1}^n \mathbb{E}_{j-1} \left(Q_{j1}(z_1) \frac{\partial Q_{j1}(z_2)}{\partial z_2} \right).$$

In fact we only need to derive the limits of the following terms

$$\sum_{j=1}^n \mathbb{E}_{j-1} (Q_{j1}(z_1) W_j(z_2)), \quad \sum_{j=1}^n \mathbb{E}_{j-1} (Q_{j1}(z_1) Q_{j1}(z_2)).$$

First we have

$$\begin{aligned} & \sum_{j=1}^n \mathbb{E}_{j-1} (Q_{j1}(z_1) Q_{j1}(z_2)) \\ &= \sum_{j=1}^n \mathbb{E}_{j-1} \left[\sum_{k=1}^p \mathbb{E}_j \left(\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_1) \mathbf{R}^{-1/2} \mathbf{e}_k \left(\mathbf{r}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{r}_j - \frac{1}{n} \right) \right) \right. \\ & \quad \left. \cdot \sum_{\ell=1}^p \mathbb{E}_j \left(\mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_2) \mathbf{R}^{-1/2} \mathbf{e}_\ell \left(\mathbf{r}_j^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{r}_j - \frac{1}{n} \right) \right) \right] \\ &= \sum_{j=1}^n \sum_{k,\ell=1}^p \mathbb{E}_j \left(\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_1) \mathbf{R}^{-1/2} \mathbf{e}_k \right) \mathbb{E}_j \left(\mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_2) \mathbf{R}^{-1/2} \mathbf{e}_\ell \right) \\ & \quad \cdot \mathbb{E}_{j-1} \left[\left(\mathbf{r}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{r}_j - \frac{1}{n} \right) \left(\mathbf{r}_j^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{r}_j - \frac{1}{n} \right) \right] \\ &= \frac{1}{n^2} \sum_{j=1}^n \sum_{k,\ell=1}^p \mathbb{E}_j \left(\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_1) \mathbf{R}^{-1/2} \mathbf{e}_k \right) \mathbb{E}_j \left(\mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_2) \mathbf{R}^{-1/2} \mathbf{e}_\ell \right) \\ & \quad \cdot \left(\beta_x \sum_{h=1}^p \left(\mathbf{e}_h^T \mathbf{R}^{1/2} \mathbf{e}_k \mathbf{e}_\ell^T \mathbf{R}^{1/2} \mathbf{e}_h \right)^2 + 2r_{k\ell}^2 \right) \end{aligned}$$

$$\xrightarrow{P_2} \frac{1}{z_1 z_2 (1 + \underline{s}(z_1))(1 + \underline{s}(z_2))} \lim_{n \rightarrow +\infty} \left(\frac{2}{n} \text{tr} \mathbf{R}^2 + \frac{\beta_x}{n} \sum_{h=1}^p (\mathbf{e}_h^T \mathbf{R} \mathbf{e}_h)^2 \right) \quad (4.36)$$

where in the last equation we used (4.28), and the last convergence is due to Lemma 6.2. of Zheng (2012).

Secondly,

$$\begin{aligned} & \sum_{j=1}^n \mathbf{E}_{j-1} (Q_{j1}(z_1) W_j(z_2)) \\ &= -\frac{\partial}{\partial z_2} \left\{ \sum_{j=1}^n \mathbf{E}_{j-1} \left[\mathbf{E}_j \left(\text{tr} \mathbf{A}_j^{-1}(z_1) \frac{1}{n} \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \right) \right. \right. \\ & \quad \left. \left. \cdot \mathbf{E}_j \left(b_j(z_2) \left(\mathbf{r}_j^T \mathbf{A}_j^{-1}(z_2) \mathbf{r}_j - \frac{\text{tr} \mathbf{A}_j^{-1}(z_2)}{n} \right) \right) \right] \right\} \\ &= -\frac{\partial}{\partial z_2} \left\{ \sum_{j=1}^n \sum_{k=1}^p \mathbf{E}_{j-1} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_1) \mathbf{R}^{-1/2} \mathbf{e}_k \left(\mathbf{r}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{r}_j - \frac{1}{n} \right) \right. \right. \\ & \quad \left. \left. \cdot \check{b}_j(z_2) \left(\mathbf{r}_j^T \check{\mathbf{A}}_j^{-1}(z_2) \mathbf{r}_j - \frac{1}{n} \text{tr} \check{\mathbf{A}}_j^{-1}(z_2) \right) \right] \right\}. \end{aligned}$$

Observe that

$$\begin{aligned} & E \left(\left(\mathbf{r}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{r}_j - \frac{1}{n} \right) \left(\mathbf{r}_j^T \check{\mathbf{A}}_j^{-1}(z_2) \mathbf{r}_j - \frac{1}{n} \text{tr} \check{\mathbf{A}}_j^{-1}(z_2) \right) \middle| \mathbf{r}_1, \dots, \mathbf{r}_{j-1}, \check{\mathbf{r}}_{j+1}, \dots, \check{\mathbf{r}}_n \right) \\ &= \frac{2}{n^2} \mathbf{g}_k^T \check{\mathbf{A}}_j^{-1}(z_2) \mathbf{g}_k + \frac{\beta_x}{n^2} \sum_{h=1}^p (\mathbf{e}_h^T \mathbf{R}^{1/2} \mathbf{e}_k)^2 \mathbf{e}_h^T \check{\mathbf{A}}_j^{-1}(z_2) \mathbf{e}_h. \end{aligned}$$

Then we see that

$$\begin{aligned} & \sum_{j=1}^n \mathbf{E}_{j-1} (Q_{j1}(z_1) W_j(z_2)) \\ &= -\frac{1}{n^2} \cdot \frac{\partial}{\partial z_2} \left\{ b(z_2) \sum_{j=1}^n \sum_{k=1}^p \mathbf{E}_{j-1} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_1) \mathbf{R}^{-1/2} \mathbf{e}_k \cdot \left(2 \mathbf{g}_k^T \check{\mathbf{A}}_j^{-1}(z_2) \mathbf{g}_k \right. \right. \right. \\ & \quad \left. \left. \left. + \beta_x \sum_{h=1}^p (\mathbf{e}_h^T \mathbf{R}^{1/2} \mathbf{e}_k)^2 \mathbf{e}_h^T \check{\mathbf{A}}_j^{-1}(z_2) \mathbf{e}_h \right) \right] \right\} + o(1) \\ &= \frac{1}{n^2} \cdot \frac{\partial}{\partial z_2} \left\{ z_2 \underline{s}(z_2) \sum_{j=1}^n \sum_{k=1}^p \mathbf{E}_{j-1} \left[\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z_1) \mathbf{R}^{-1/2} \mathbf{e}_k \cdot \left(2 \mathbf{g}_k^T \check{\mathbf{A}}_j^{-1}(z_2) \mathbf{g}_k \right. \right. \right. \end{aligned}$$

$$\begin{aligned}
 & \left. + \beta_x \sum_{h=1}^p \left(\mathbf{e}_h^T \mathbf{R}^{1/2} \mathbf{e}_k \right)^2 \mathbf{e}_h^T \check{\mathbf{A}}_j^{-1}(z_2) \mathbf{e}_h \right) \Bigg] \Bigg\} + o(1) \\
 \xrightarrow{p} & \frac{\partial}{\partial z_2} \left(\frac{z_2 \underline{s}(z_2)}{z_1 z_2 (1 + \underline{s}(z_1)) (1 + \underline{s}(z_2))} \right) \\
 & \quad \cdot \left(2\rho + \lim_{n \rightarrow \infty} \frac{\beta_x}{n} \sum_{h=1}^p \mathbf{e}_h^T \mathbf{R} \mathbf{e}_h \right) \\
 = & -\frac{1}{z_1 (1 + \underline{s}(z_1))} \frac{\partial}{\partial z_2} \left(\frac{1}{1 + \underline{s}(z_2)} \right) \cdot (2 + \beta_x) \rho,
 \end{aligned}$$

where by (4.3) and some computation, we have $|b_j(z) - b(z)| \leq Kn^{-1}$ and $|b(z) - z\underline{s}(z)| \leq Kn^{-1}$.

Moreover, $s(z)$ is the Stieltjes transformation of the M̄cenko-Pastur law, $\underline{s}(z) = -(1 - \rho)z^{-1} + \rho s(z)$ and $z = -\underline{s}^{-1}(z) + \rho(1 + \underline{s}(z))^{-1}$.

Combining the results above and by the martingale CLT and Lemma 2.14 of Bai and Sil-verstein (2010) we then obtain that the process

$$\sum_{j=1}^n \left[W_j(z) + Q_{j1}(z) + z \cdot \frac{\partial}{\partial z} Q_{j1}(z) \right]$$

converges weakly to a Gaussian process with mean zero and covariance functions

$$\begin{aligned}
 & v(z_1, z_2) \\
 = & 2 \left(\frac{\underline{s}'(z_1) \underline{s}'(z_2)}{(\underline{s}(z_2) - \underline{s}(z_1))^2} - \frac{1}{(z_2 - z_1)^2} \right) + \rho \beta_x \cdot \frac{\partial}{\partial z_1} \left(\frac{1}{(1 + \underline{s}(z_1))} \right) \cdot \frac{\partial}{\partial z_2} \left(\frac{1}{(1 + \underline{s}(z_2))} \right) \\
 & + \frac{(\rho \beta_x t \mathbf{G} + \lim_{n \rightarrow \infty} \frac{2}{n} \text{tr} \mathbf{R}^2)}{z_1 z_2 (1 + \underline{s}(z_1)) (1 + \underline{s}(z_2))} \\
 & + z_1 z_2 \left(\rho \beta_x t \mathbf{G} + \lim_{n \rightarrow \infty} \frac{2}{n} \text{tr} \mathbf{R}^2 \right) \frac{\partial}{\partial z_1} \left(\frac{1}{z_1 (1 + \underline{s}(z_1))} \right) \cdot \frac{\partial}{\partial z_2} \left(\frac{1}{z_2 (1 + \underline{s}(z_2))} \right) \\
 & - (\rho \beta_x + 2\rho) \frac{1}{z_1 (1 + \underline{s}(z_1))} \cdot \frac{\partial}{\partial z_2} \left(\frac{1}{(1 + \underline{s}(z_2))} \right) \\
 & - (\rho \beta_x + 2\rho) \frac{1}{z_2 (1 + \underline{s}(z_2))} \cdot \frac{\partial}{\partial z_1} \left(\frac{1}{(1 + \underline{s}(z_1))} \right) \\
 & - (\rho \beta_x + 2\rho) z_1 \cdot \frac{\partial}{\partial z_1} \left(\frac{1}{z_1 (1 + \underline{s}(z_1))} \right) \cdot \frac{\partial}{\partial z_2} \left(\frac{1}{(1 + \underline{s}(z_2))} \right) \\
 & - (\rho \beta_x + 2\rho) z_2 \cdot \frac{\partial}{\partial z_1} \left(\frac{1}{(1 + \underline{s}(z_1))} \right) \cdot \frac{\partial}{\partial z_2} \left(\frac{1}{z_2 (1 + \underline{s}(z_2))} \right) \\
 & + (\rho \beta_x t \mathbf{G} + \lim_{n \rightarrow \infty} \frac{2}{n} \text{tr} \mathbf{R}^2) z_1 \cdot \frac{\partial}{\partial z_1} \left(\frac{1}{z_1 (1 + \underline{s}(z_1))} \right) \cdot \frac{1}{z_2 (1 + \underline{s}(z_2))}
 \end{aligned}$$

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$$\begin{aligned}
& +(\rho\beta_x t_{\mathbf{G}} + \lim_{n \rightarrow \infty} \frac{2}{n} \text{tr} \mathbf{R}^2) z_2 \cdot \frac{1}{z_1(1 + \underline{s}(z_1))} \cdot \frac{\partial}{\partial z_2} \left(\frac{1}{z_2(1 + \underline{s}(z_2))} \right) \\
= & 2 \left(\frac{\underline{s}'(z_1) \underline{s}'(z_2)}{(\underline{s}(z_2) - \underline{s}(z_1))^2} - \frac{1}{(z_2 - z_1)^2} \right) + \frac{\underline{s}'(z_1) \underline{s}'(z_2)}{(1 + \underline{s}(z_1))^2 (1 + \underline{s}(z_2))^2} \\
& \cdot \left(\rho\beta_x t_{\mathbf{G}} + \lim_{n \rightarrow \infty} \frac{2}{n} \text{tr} \mathbf{R}^2 - \rho\beta_x - 4\rho \right). \tag{4.37}
\end{aligned}$$

Step \mathcal{IV} Now we derive the limit of the term VI defined in (4.14). Firstly, Zheng, Bai and Yao (2015) proved that

$$\text{Etr} \mathbf{A}^{-1}(z) - p s_{\rho_n}(z) \rightarrow \frac{\rho \underline{s}^3(z)(1 + \underline{s}(z))^{-3}}{(1 - \rho \underline{s}^2(z)(1 + \underline{s}(z))^{-2})^2} + \frac{\beta_x \rho \underline{s}^3(z)(1 + \underline{s}(z))^{-3}}{1 - \rho \underline{s}^2(z)(1 + \underline{s}(z))^{-2}}. \tag{4.38}$$

Then it follows that from (4.17)

$$\begin{aligned}
\text{Etr} (\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p)) &= \frac{1}{2} \text{Etr} (\mathbf{A}^{-1}(z) \mathbf{D}) + \frac{1}{2} \text{Etr} (\mathbf{A}^{-1}(z) \mathbf{D}^T) \\
&+ \frac{1 + o(1)}{4} \text{Etr} (\mathbf{A}^{-1}(z) \mathbf{D} \mathbf{D}^T) - \frac{1 + o(1)}{8} \text{Etr} (\mathbf{A}^{-1}(z) \mathbf{D}^2) \\
&- \frac{1 + o(1)}{8} \text{Etr} (\mathbf{A}^{-1}(z) (\mathbf{D}^T)^2). \tag{4.39}
\end{aligned}$$

We begin with the third term of the righthand side of (4.39). It is apparent that

$$\begin{aligned}
\text{Etr} (\mathbf{A}^{-1}(z) \mathbf{D} \mathbf{D}^T) &= \frac{1}{n} \sum_{j=1}^n \text{Etr} (\mathbf{A}^{-1}(z) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}^T) \\
&= \frac{1}{n^2} \sum_{j=1}^n \text{Etr} (\mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{-1/2}) \\
&+ \frac{1}{n} \sum_{j=1}^n \text{Etr} (\mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}_j^T) \\
&- \frac{1}{n} \sum_{j=1}^n \text{E} (\beta_j(z) \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{r}_j).
\end{aligned}$$

By conditioning on \mathcal{F}_j one sees that the second term equals 0. Applying (4.16) and (4.18), we get

$$\begin{aligned}
& \left| \frac{1}{n} \sum_{j=1}^n \text{E} (\beta_j(z) \mathbf{r}_j^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \mathbf{D}^T \mathbf{A}_j^{-1}(z) \mathbf{r}_j) \right| \\
& \leq \frac{o(1)}{n} \sum_{j=1}^n \text{E} (\mathbf{r}_j^T \mathbf{r}_j \|\text{diag}(\mathbf{S}_j) - \mathbf{I}_p\|) \leq \frac{o(1)}{n} \sum_{j=1}^n \text{E}^{1/2} (\mathbf{r}_j^T \mathbf{r}_j)^2 = o(1).
\end{aligned}$$

Therefore, we see that from (4.28)

$$\begin{aligned}
 & \text{Etr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \mathbf{D}^T \right) \\
 &= \frac{1}{n^2} \sum_{j=1}^n \text{Etr} \left(\mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{-1/2} \right) + o(1) \\
 &= \frac{1}{n^2} \sum_{j=1}^n \sum_{k, \ell=1}^p \text{E} \left(\left(\mathbf{X}_j \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{X}_j - 1 \right) \left(\mathbf{X}_j \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1 \right) \mathbf{e}_\ell^T \mathbf{R} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{-1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_\ell \right) + o(1) \\
 &= \frac{1}{n^2} \sum_{j=1}^n \sum_{k, \ell=1}^p \left(\beta_x \sum_{h=1}^p g_{\ell h}^2 g_{kh}^2 + 2r_{k\ell}^2 \right) \text{E} \left(\mathbf{e}_\ell^T \mathbf{R} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{-1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_\ell \right) + o(1) \\
 &\rightarrow -\frac{1}{z(1+\underline{s}(z))} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k, \ell=1}^p \left(\beta_x \sum_{h=1}^p g_{\ell h}^2 g_{kh}^2 + 2r_{k\ell}^2 \right) \left(\mathbf{e}_\ell^T \mathbf{R} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{-1} \mathbf{e}_\ell \right), \tag{4.40}
 \end{aligned}$$

where the proof of (4.40) is similar to those of (4.36). It can also be verified that

$$\text{Etr} \left(\mathbf{A}^{-1}(z) \mathbf{D}^2 \right) \rightarrow -\frac{1}{z(1+\underline{s}(z))} \left(\beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{h=1}^p g_{kh}^4 + 2\rho \right) \tag{4.41}$$

and

$$\text{Etr} \left(\mathbf{A}^{-1}(z) \left(\mathbf{D}^T \right)^2 \right) \rightarrow -\frac{1}{z(1+\underline{s}(z))} \left(\beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{h=1}^p g_{kh}^4 + 2\rho \right). \tag{4.42}$$

Next, using (4.2) we have

$$\begin{aligned}
 \text{Etr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \right) &= \frac{1}{n} \sum_{j=1}^n \text{Etr} \left(\mathbf{A}^{-1}(z) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \right) \\
 &= \frac{1}{n} \sum_{k=1}^p \sum_{j=1}^n \text{E} \left(\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k (\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1) \right) \\
 &= \frac{1}{n} \sum_{k=1}^p \sum_{j=1}^n \text{E} \left(\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k (\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1) \right) \\
 &\quad - \frac{1}{n} \sum_{k=1}^p \sum_{j=1}^n \text{E} \left(\beta_j(z) \mathbf{r}_j^T \mathbf{A}_j^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{r}_j (\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1) \right) \\
 &= -\frac{1}{n} \sum_{k=1}^p \sum_{j=1}^n \text{E} \left(\beta_j(z) \mathbf{r}_j^T \mathbf{A}_j^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{r}_j (\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1) \right).
 \end{aligned}$$

By the decomposition (4.3) of $\beta_j(z)$ and (9.9.6) of Bai and Silverstein (2010), we have

$$\text{Etr} \left(\mathbf{A}^{-1}(z) \mathbf{D} \right)$$

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$$\begin{aligned}
&= -\frac{1}{n} \sum_{k=1}^p \sum_{j=1}^n \mathbb{E} \left(b_j(z) \mathbf{r}_j^T \mathbf{A}_j^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{r}_j (\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1) \right) + o(1) \\
&= -\frac{1}{n} \sum_{k=1}^p \sum_{j=1}^n \mathbb{E} \left[b_j(z) \left(\mathbf{r}_j^T \mathbf{A}_j^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{r}_j - \frac{1}{n} \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-2}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right) (\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1) \right] \\
&\quad - \frac{1}{n^2} \sum_{k=1}^p \sum_{j=1}^n \mathbb{E} \left(b_j(z) \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-2}(z) \mathbf{R}^{-1/2} \mathbf{e}_k (\mathbf{X}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{X}_j - 1) \right) + o(1). \\
&:= VI.2.1 + VI.2.2 + o(1). \tag{4.43}
\end{aligned}$$

By conditioning on F_j one sees that the second term equals 0. Furthermore, one has

$$\begin{aligned}
VI.2.1 &= -\frac{1}{n^2} \sum_{k=1}^p \sum_{j=1}^n \mathbb{E} \left[b_j(z) \left(2\mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{g}_k \mathbf{g}_k^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{-1/2} \mathbf{e}_k \right. \right. \\
&\quad \left. \left. + \beta_x \sum_{h=1}^p g_{kh}^2 \mathbf{e}_h^T \mathbf{A}_j^{-1} \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbf{A}_j^{-1}(z) \mathbf{e}_h \right) \right] + o(1) \\
&\rightarrow \frac{\underline{s}(z)}{z(1+\underline{s}(z))^2} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \left(2\mathbf{e}_k^T \mathbf{R} \mathbf{e}_k + \beta_x \sum_{h=1}^p g_{kh}^3 \mathbf{e}_h^T \mathbf{R}^{-1/2} \mathbf{e}_k \right), \tag{4.44}
\end{aligned}$$

where the proof of (4.44) is similar to those of (4.36). Similarly, we get

$$\begin{aligned}
&\text{Etr} \left(\mathbf{A}^{-1}(z) \mathbf{D}^T \right) \\
&= -\frac{1}{n^2} \sum_{k=1}^p \sum_{j=1}^n \mathbb{E} \left[b_j(z) \left(2\mathbf{e}_k^T \mathbf{R}^{-1/2} \mathbf{A}_j^{-1}(z) \mathbf{g}_k \mathbf{g}_k^T \mathbf{A}_j^{-1}(z) \mathbf{R}^{1/2} \mathbf{e}_k \right. \right. \\
&\quad \left. \left. + \beta_x \sum_{h=1}^p g_{kh}^2 \mathbf{e}_h^T \mathbf{A}_j^{-1} \mathbf{R}^{1/2} \mathbf{e}_k \mathbf{e}_k^T \mathbf{R}^{-1/2} \mathbf{A}_j^{-1}(z) \mathbf{e}_h \right) \right] + o(1) \\
&\rightarrow \frac{\underline{s}(z)}{z(1+\underline{s}(z))^2} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \left(2\mathbf{e}_k^T \mathbf{R} \mathbf{e}_k + \beta_x \sum_{h=1}^p g_{kh}^3 \mathbf{e}_h^T \mathbf{R}^{-1/2} \mathbf{e}_k \right). \tag{4.45}
\end{aligned}$$

Combining (4.39)-(4.45), we see that

$$\begin{aligned}
&\text{Etr} \left(\mathbf{A}^{-1}(z) (\mathbf{L} - \mathbf{I}_p) \right) \\
&\rightarrow \frac{\underline{s}(z)}{z(1+\underline{s}(z))^2} \left(2\rho + \beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{h=1}^p g_{kh}^3 \mathbf{e}_h^T \mathbf{R}^{-1/2} \mathbf{e}_k \right) \\
&\quad - \frac{1}{4z(1+\underline{s}(z))} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k,\ell=1}^p \mathbf{e}_k^T \mathbf{R}^{-1} \mathbf{e}_\ell r_{k\ell} \cdot \left(\beta_x \sum_{h=1}^p g_{\ell h}^2 g_{kh}^2 + 2r_{k\ell}^2 \right) \\
&\quad + \frac{1}{4z(1+\underline{s}(z))} \left(\beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{h=1}^p g_{kh}^4 + 2\rho \right). \tag{4.46}
\end{aligned}$$

It follows that the limit of the third term of the term VI defined in (4.14)

$$\begin{aligned}
 & z \text{Etr}(\mathbf{A}^{-2}(z)(\mathbf{L} - \mathbf{I}_p)) \\
 \rightarrow & z \frac{\partial}{\partial z} \left(\frac{\underline{s}(z)}{z(1 + \underline{s}(z))^2} \right) \left(2\rho + \beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{h=1}^p g_{kh}^3 \mathbf{e}_h^T \mathbf{R}^{-1/2} \mathbf{e}_k \right) \\
 & - z \frac{\partial}{\partial z} \left(\frac{1}{4z(1 + \underline{s}(z))} \right) \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k,\ell=1}^p \mathbf{e}_k^T \mathbf{R}^{-1} \mathbf{e}_\ell r_{k\ell} \cdot \left(\beta_x \sum_{h=1}^p g_{\ell h}^2 g_{kh}^2 + 2r_{k\ell}^2 \right) \\
 & + z \frac{\partial}{\partial z} \left(\frac{1}{4z(1 + \underline{s}(z))} \right) \left(\beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{h=1}^p g_{kh}^4 + 2\rho \right). \tag{4.47}
 \end{aligned}$$

Combining (4.13), (4.38), (4.46) and (4.47), the limit of the term VI defined in (4.14) is

$$\begin{aligned}
 & \frac{\rho \underline{s}^3(z)(1 + \underline{s}(z))^{-3}}{(1 - \rho \underline{s}^2(z)(1 + \underline{s}(z))^{-2})^2} + \frac{\beta_x \cdot \rho \underline{s}^3(z)(1 + \underline{s}(z))^{-3}}{1 - \rho \underline{s}^2(z)(1 + \underline{s}(z))^{-2}} \\
 & + \left[\frac{\underline{s}(z)}{z(1 + \underline{s}(z))^2} + z \frac{\partial}{\partial z} \left(\frac{\underline{s}(z)}{z(1 + \underline{s}(z))^2} \right) \right] \left(2\rho + \beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{h=1}^p g_{kh}^3 \mathbf{e}_h^T \mathbf{R}^{-1/2} \mathbf{e}_k \right) \\
 & - \left[\frac{1}{4z(1 + \underline{s}(z))} + z \frac{\partial}{\partial z} \left(\frac{1}{4z(1 + \underline{s}(z))} \right) \right] \left[\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k,\ell=1}^p \mathbf{e}_k^T \mathbf{R}^{-1} \mathbf{e}_\ell r_{k\ell} \right. \\
 & \quad \left. \cdot \left(\beta_x \sum_{h=1}^p g_{\ell h}^2 g_{kh}^2 + 2r_{k\ell}^2 \right) - \left(\beta_x \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^p \sum_{h=1}^p g_{kh}^4 + 2\rho \right) \right]. \tag{4.48}
 \end{aligned}$$

Step IV We are finally ready to establish the CLT of LSS of the matrix $\mathbf{R}^{-1} \widehat{\mathbf{R}}$.

Combining (4.9), (4.30), (4.31), (4.11), (4.12), (4.37) and (4.48), we see that $M_n(z)$ converges to a Gaussian process $M(z)$ with the mean function and covariance function as follows

$$\begin{aligned}
 \text{EM}(z) = & \frac{\rho \underline{s}^3(z)(1 + \underline{s}(z))^{-3}}{(1 - \rho \underline{s}^2(z)(1 + \underline{s}(z))^{-2})^2} + \frac{\beta_x \cdot \rho \underline{s}^3(z)(1 + \underline{s}(z))^{-3}}{1 - \rho \underline{s}^2(z)(1 + \underline{s}(z))^{-2}} \\
 & + \left(-\frac{\underline{s}'(z)}{(1 + \underline{s}(z))^2} + \frac{2\underline{s}'(z)}{(1 + \underline{s}(z))^3} \right) \frac{[8 - 2a_{\mathbf{R}} + \beta_x(4a_g - h_{\mathbf{R}})]\rho}{4} \\
 & - \left(\frac{\underline{s}'(z)}{(1 + \underline{s}(z))^2} + \frac{2\underline{s}'(z)}{(1 + \underline{s}(z))^3} \right) \cdot \frac{(\beta_x c_g + 2)\rho}{4}
 \end{aligned} \tag{4.49}$$

and

$$\begin{aligned}
 & \text{Cov}(M(z_1), M(z_2)) \\
 = & 2 \left(\frac{\underline{s}'(z_1)\underline{s}'(z_2)}{(\underline{s}(z_2) - \underline{s}(z_1))^2} - \frac{1}{(z_2 - z_1)^2} \right) + \rho \frac{\underline{s}'(z_1)\underline{s}'(z_2)}{(1 + \underline{s}(z_1))^2(1 + \underline{s}(z_2))^2} \\
 & \cdot (\beta_x + 2d_{\mathbf{R}} - \beta_x - 4). \tag{4.50}
 \end{aligned}$$

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The remaining is to prove the tightness, which should have been proved in Step III. We need to obtain the tightness of $\text{tr}\mathbf{A}^{-1}(z) - \mathbb{E}(\text{tr}\mathbf{A}^{-1}(z))$ and $\sum_{j=1}^n Q_{j1}(z)$. Notice that the tightness of $\text{tr}\mathbf{A}^{-1}(z) - \mathbb{E}(\text{tr}\mathbf{A}^{-1}(z))$ has been proved by Bai and Silverstein (2004), now we will devote to prove the tightness of $\sum_{j=1}^n Q_{j1}(z)$.

We want to prove, as a sufficient condition,

$$\sup_{n, z_1, z_2 \in \mathcal{C}^+} \frac{\mathbb{E} \left| \sum_{j=1}^n (Q_{j1}(z_1) - Q_{j1}(z_2)) \right|^2}{|z_1 - z_2|^2} \leq K.$$

To this end, we only need to prove

$$\sup_{n, z_1, z_2 \in \mathcal{C}_n} \frac{\mathbb{E} \left| \sum_{j=1}^n (Q_{j1}(z_1) - Q_{j1}(z_2)) \right|^2}{|z_1 - z_2|^2} \leq K.$$

Apparently, we have

$$\begin{aligned} & \frac{\sum_{j=1}^n (Q_{j1}(z_1) - Q_{j1}(z_2))}{z_1 - z_2} \\ &= \frac{1}{n} \sum_{j=1}^n \mathbb{E}_j \left[\text{tr} \mathbf{A}_j^{-1}(z_1) \mathbf{A}_j^{-1}(z_2) \mathbf{R}^{-1/2} (\text{diag}(\mathbf{S}_j) - \mathbf{I}_p) \mathbf{R}^{1/2} \right] \\ &= \sum_{j=1}^n \sum_{k=1}^p \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbb{E}_j \left[\mathbf{A}_j^{-1}(z_1) \mathbf{A}_j^{-1}(z_2) \right] \mathbf{R}^{-1/2} \mathbf{e}_k \cdot (\mathbf{r}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{r}_j - \frac{1}{n}). \end{aligned}$$

Thus we obtain

$$\begin{aligned} & \frac{\mathbb{E} \left| \sum_{j=1}^n (Q_{j1}(z_1) - Q_{j1}(z_2)) \right|^2}{|z_1 - z_2|^2} \\ &= \mathbb{E} \left| \sum_{j=1}^n \sum_{k=1}^p \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbb{E}_j \left[\mathbf{A}_j^{-1}(z_1) \mathbf{A}_j^{-1}(z_2) \right] \mathbf{R}^{-1/2} \mathbf{e}_k \cdot (\mathbf{r}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{r}_j - \frac{1}{n}) \right|^2 \\ &= \sum_{j=1}^n \mathbb{E} \left| \sum_{k=1}^p \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbb{E}_j \left[\mathbf{A}_j^{-1}(z_1) \mathbf{A}_j^{-1}(z_2) \right] \mathbf{R}^{-1/2} \mathbf{e}_k \cdot (\mathbf{r}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{r}_j - \frac{1}{n}) \right|^2 \\ &= \sum_{j=1}^n \sum_{k, \ell=1}^p \mathbb{E} \left\{ \mathbf{e}_k^T \mathbf{R}^{1/2} \mathbb{E}_j \left[\mathbf{A}_j^{-1}(z_1) \mathbf{A}_j^{-1}(z_2) \right] \mathbf{R}^{-1/2} \mathbf{e}_k \mathbf{e}_\ell^T \mathbf{R}^{1/2} \right. \\ & \quad \left. \cdot \mathbb{E}_j \left[\mathbf{A}_j^{-1}(z_1) \mathbf{A}_j^{-1}(z_2) \right] \mathbf{R}^{-1/2} \mathbf{e}_\ell \right\} \cdot \mathbb{E} \left[(\mathbf{r}_j^T \mathbf{g}_k \mathbf{g}_k^T \mathbf{r}_j - \frac{1}{n}) (\mathbf{r}_j^T \mathbf{g}_\ell \mathbf{g}_\ell^T \mathbf{r}_j - \frac{1}{n}) \right] \end{aligned}$$

$$\leq \frac{1}{n^2} \sum_{j=1}^n \sum_{k,\ell=1}^p (2r_{k\ell}^2 + \beta_x \sum_{h=1}^p g_{\ell h}^2 g_{kh}^2) \leq K,$$

where K is a constant and $E|\mathbf{A}_j^{-1}(z)|^p \leq K$. So the tightness is proved.

Following the representation (4.1), the LSS of sample correlation matrix $\mathbf{R}^{-1}\widehat{\mathbf{R}}$ is as follows

$$\sum_{k=1}^p g_\ell(\lambda_k) - p \int g_\ell(x) dF_{\rho_n}(x) = -\frac{1}{2\pi\mathbf{i}} \int g_\ell(z) \left(\mathbf{R}^{-1}\widehat{\mathbf{R}} - z\mathbf{I} \right)^{-1} - p s_{\rho_n}(z) dz$$

for $\ell = 1, \dots, m$, where $F_\rho(\cdot)$ is the LSD of $\mathbf{R}^{-1}\widehat{\mathbf{R}}$ and F_{ρ_n} is obtained by replacing ρ by ρ_n .

Thus the random vector

$$\left(\sum_{k=1}^p g_1(\lambda_k) - p \int g_1(x) dF_{\rho_n}(x), \dots, \sum_{k=1}^p g_m(\lambda_k) - p \int g_m(x) dF_{\rho_n}(x) \right) \quad (4.51)$$

converges to an m -dimensional Gaussian random vector X_{g_1}, \dots, X_{g_m} with mean functions and

covariance functions as follows

$$EX_{g_\ell} = -\frac{1}{2\pi\mathbf{i}} \oint g_\ell(z) \cdot EM(z) dz$$

and

$$\text{Cov}(X_{g_\ell}, X_{g_j}) = -\frac{1}{4\pi^2} \oint \oint g_\ell(z_1) g_j(z_2) \cdot \text{Cov}(M(z_1), M(z_2)) dz_1 dz_2.$$

We finally complete the proof of Theorem 3.1.

5. Proof of Corollary 3.1 and Example 3.2

5.1 Proof of Corollary 3.1

Let $\underline{s}(z) = -\frac{1}{1+\sqrt{\rho}\xi}$, then

$$dz = h(1 - \xi^{-2})d\xi \quad \text{and} \quad z = 1 + h\xi + h\xi^{-1} + h^2 = |1 + h\xi|^2$$

where $|\xi| = 1$ and $h = \sqrt{\rho}$, because $z = -\underline{s}^{-1}(z) + \rho(1 + \underline{s}(z))^{-1}$ (e.g. Wang and Yao (2013)).

When ξ runs counterclockwisely the unit circle, z will run counterclockwisely a contour that

5. PROOF OF COROLLARY 3.1 AND EXAMPLE 3.2

encloses the support interval $[a, b] = [(1-h)^2, (1+h)^2]$. So now we regard z as a function of ξ .

Moreover, we have

$$\underline{s}'(z) = \frac{\underline{s}^2(z)[1 + \underline{s}(z)]^2}{[1 + \underline{s}(z)]^2 - \rho \underline{s}^2(z)},$$

since

$$z = -\frac{1}{\underline{s}(z)} + \frac{\rho}{1 + \underline{s}(z)} = -\frac{[1 + (1-\rho)\underline{s}(z)]}{\underline{s}(z)[1 + \underline{s}(z)]} = -(1-\rho) \frac{[\underline{s}(z) + (1-\rho)^{-1}]}{\underline{s}(z)[1 + \underline{s}(z)]}.$$

Substitute the above expression into Theorem 3.1, we obtain Corollary 3.1.

5.2 Proof of Example 3.2

Let $g_1(x) = x$, $g_2(x) = x^2$, $g_3(x) = x^3$, $g_4(x) = x^4$, and $g_5(x) = \log(x)$. It is well known that the moments of the standard M-P distribution with index ρ take values

$$m_k(\rho) = \sum_{r=0}^{k-1} \frac{1}{r+1} \binom{r}{k} \binom{k-1}{r} \rho^k.$$

From this, it is easy to calculate the centering terms

$$\begin{aligned} \int g_1(x) f_{\rho_{n-1}}(x) dx &= m_1(\rho_{n-1}) = 1, & \int g_2(x) f_{\rho_{n-1}}(x) dx &= m_2(\rho_{n-1}) = 1 + \rho_{n-1}, \\ \int g_3(x) f_{\rho_{n-1}}(x) dx &= m_3(\rho_{n-1}) = 1 + 3\rho_{n-1} + \rho_{n-1}^2, \\ \int g_4(x) f_{\rho_{n-1}}(x) dx &= m_4(\rho_{n-1}) = 1 + 6\rho_{n-1} + 6\rho_{n-1}^2 + \rho_{n-1}^3. \end{aligned}$$

While the centering term $\int g_5(x) f_{\rho_{n-1}}(x) dx = \frac{\rho_{n-1}-1}{\rho_{n-1}} \log(1 - y_{n-1}) - 1$, $\rho_{n-1} < 1$, has been proved in Section 9.12.3 in Bai and Silverstein (2010).

Using Theorem 3.1, now we focus on the expressions of the expectation parameters $E X_{g_i}$ of LSS for $j = 1, 2, 3, 4, 5$ and (co)variance parameters $\text{Cov}(X_{g_j}, X_{g_k})$ for $j, k = 1, 2, 3, 4, 5$.

Firstly, we compute $E X_{g_j}$ for $j = 1, 2, 3, 4$. Denoting

$$\mu_1(g) = -\frac{1}{2\pi i} \oint_{\mathcal{C}} g(z) \left(-\frac{1}{(1 + \underline{s}(z))^2} + \frac{2}{(1 + \underline{s}(z))^3} \right) \underline{s}'(z) dz,$$

$$\begin{aligned}\mu_2(g) &= -\frac{1}{2\pi i} \oint_C g(z) \left(\frac{1}{(1+\underline{s}(z))^2} + \frac{2}{(1+\underline{s}(z))^3} \right) \underline{s}'(z) dz, \\ \mu_3(g) &= -\frac{1}{2\pi i} \oint g(z) \frac{\rho \underline{s}^3(z)(1+\underline{s}(z))^{-3}}{(1-\rho \underline{s}^2(z)(1+\underline{s}(z))^{-2})^2} dz, \\ \mu_4(g) &= -\frac{1}{2\pi i} \oint g(z) \frac{\beta_x \cdot \rho \underline{s}^3(z)(1+\underline{s}(z))^{-3}}{1-\rho \underline{s}^2(z)(1+\underline{s}(z))^{-2}} dz,\end{aligned}$$

it follows that

$$EX_g = \mu_3(g) + \beta_x \mu_4(g) + \frac{[8 - 2a_{\mathbf{R}} + \beta_x(4a_g - h_{\mathbf{R}})]\rho}{4} \mu_1(g) - \frac{(\beta_x c_g + 2)\rho}{4} \mu_2(g).$$

By Wang and Yao (2013), one gets

$$\begin{aligned}\mu_3(g) &= \lim_{r \downarrow 1} \frac{1}{2\pi i} \oint_{|\xi|=1} g(|1+h\xi|^2) \left[\frac{\xi}{\xi^2 - r^{-2}} - \frac{1}{\xi} \right] d\xi, \\ \mu_4(g) &= \frac{1}{2\pi i} \oint_{|\xi|=1} g(|1+h\xi|^2) \frac{1}{\xi^3} d\xi.\end{aligned}$$

Similarly, we can show that

$$\begin{aligned}\mu_1(g) &= -\frac{1}{2\pi i} \oint_{|\xi|=1} g(|1+h\xi|^2) \left(-\left(1 + \frac{1}{h\xi}\right)^2 + 2\left(1 + \frac{1}{h\xi}\right)^3 \right) \frac{h}{(1+h\xi)^2} d\xi \\ &= -\frac{1}{2\pi i} \oint_{|\xi|=1} g(|1+h\xi|^2) \left(\frac{1}{h\xi^2} + \frac{2}{h^2\xi^3} \right) d\xi, \\ \mu_2(g) &= -\frac{1}{2\pi i} \oint_{|\xi|=1} g(|1+h\xi|^2) \left(\frac{3}{h\xi^2} + \frac{2}{h^2\xi^3} \right) d\xi.\end{aligned}$$

Note that

$$\begin{aligned}|1+h\xi|^8 &= (1+h^2)^4 + 12(1+h^2)^2 h^2 + 6h^4 + h^4(\xi^{-4} + \xi^4) \\ &\quad + 4(1+h^2)h^3(\xi^{-3} + \xi^3) + (4h^4 + 6(1+h^2)^2 h^2)(\xi^{-2} + \xi^2) \\ &\quad + (12(1+h^2)h^3 + 4(1+h^2)^3 h)(\xi^{-1} + \xi), \\ |1+h\xi|^6 &= (1+h^2)^3 + 6h^2(1+h^2) + (3h(1+h^2)^2 + 3h^3)(\xi + \xi^{-1}) \\ &\quad + 3h^2(1+h^2)(\xi^2 + \xi^{-2}) + h^3(\xi^3 + \xi^{-3}), \\ |1+h\xi|^4 &= (1+h^2)^2 + 2h^2 + 2h(1+h^2)(\xi + \xi^{-1}) + h^2(\xi^2 + \xi^{-2}).\end{aligned}$$

5. PROOF OF COROLLARY 3.1 AND EXAMPLE 3.2

Then we have

$$\begin{aligned}
\mu_1(g_4) &= -\frac{1}{2\pi i} \oint_{|\xi|=1} |1+h\xi|^8 \left(\frac{1}{h\xi^2} + \frac{2}{h^2\xi^3} \right) d\xi, \\
&= -\frac{1}{2\pi i} \oint_{|\xi|=1} \frac{12(1+h^2)h^2 + 4(1+h^2)^3}{\xi} d\xi - \frac{1}{\pi i} \oint_{|\xi|=1} \frac{4h^2 + 6(1+h^2)^2}{\xi} d\xi, \\
&= -4(3(1+h^2)h^2 + (1+h^2)^3) - 4(2h^2 + 3(1+h^2)^2) \\
&= -4(3(1+\rho)\rho + (1+\rho)^3 + 2\rho + 3(1+\rho)^2), \\
\mu_2(g_4) &= -4(9(1+\rho)\rho + 3(1+\rho)^3 + 2\rho + 3(1+\rho)^2).
\end{aligned}$$

$$\begin{aligned}
\mu_3(g_4) &= \lim_{r \downarrow 1} \frac{1}{2\pi i} \oint_{|\xi|=1} |1+h\xi|^8 \left[\frac{\xi}{\xi^2 - r^{-2}} - \frac{1}{\xi} \right] d\xi \\
&= \lim_{r \downarrow 1} \frac{1}{2\pi i} \oint_{|\xi|=1} \frac{[(1+h^2)\xi + h\xi^2 + h]^4}{\xi^3(\xi^2 - r^{-2})} d\xi \\
&\quad - \frac{1}{2\pi i} \oint_{|\xi|=1} \frac{(1+h^2)^4 + 12(1+h^2)^2h^2 + 6h^4}{\xi} d\xi \\
&= \lim_{r \downarrow 1} \left. \frac{[(1+h^2)\xi + h\xi^2 + h]^4}{\xi^3(\xi + r^{-1})} \right|_{\xi=r^{-1}} + \lim_{r \downarrow 1} \left. \frac{[(1+h^2)\xi + h\xi^2 + h]^4}{\xi^3(\xi - r^{-1})} \right|_{\xi=-r^{-1}} \\
&\quad + \lim_{r \downarrow 1} \left. \frac{4h^4 + 6h^2(1+h^2)^2}{(\xi^2 - r^{-2})} \right|_{\xi=0} - (1+h^2)^4 - 12(1+h^2)^2h^2 - 6h^4 \\
&= \frac{(1+h)^8}{2} + \frac{(1-h)^8}{2} - 4h^4 - 6h^2(1+h^2)^2 - (1+h^2)^4 - 12(1+h^2)^2h^2 - 6h^4 \\
&= 6\rho^2 + 6(1+\rho)^2\rho,
\end{aligned}$$

$$\mu_4(g_4) = \frac{1}{2\pi i} \oint_{|\xi|=1} |1+h\xi|^8 \frac{1}{\xi^3} dz = 4\rho^2 + 6(1+\rho)^2\rho.$$

Therefore, we get

$$\begin{aligned}
EX_{g_4} &= 6\rho^2 + 6(1+\rho)^2\rho + 2\beta_x\rho [2\rho + 3(1+\rho)^2] \\
&\quad - \rho[8 - 2a_{\mathbf{R}} + \beta_x(4a_g - h_{\mathbf{R}})] [3(1+\rho)\rho + (1+\rho)^3 + 2\rho + 3(1+\rho)^2] \\
&\quad + \rho(\beta_x c_g + 2) [9(1+\rho)\rho + 3(1+\rho)^3 + 2\rho + 3(1+\rho)^2].
\end{aligned}$$

Moreover, we have

$$\mu_1(g_3) = -\frac{1}{2\pi i} \oint_{|\xi|=1} |1+h\xi|^6 \left(\frac{1}{h\xi^2} + \frac{2}{h^2\xi^3} \right) d\xi$$

$$= -(3(1+h^2)^2 + 3h^2) - 6(1+h^2) = -3((1+\rho)^2 + 3\rho + 2),$$

$$\mu_2(g_3) = -3(3(1+\rho)^2 + 5\rho + 2),$$

$$\begin{aligned} \mu_3(g_3) &= \lim_{r \downarrow 1} \frac{1}{2\pi i} \oint_{|\xi|=1} |1+h\xi|^6 \left[\frac{\xi}{\xi^2 - r^{-2}} - \frac{1}{\xi} \right] d\xi \\ &= \frac{(1+h)^6}{2} + \frac{(1-h)^6}{2} - 3h^2(1+h^2) - (1+h^2)^3 - 6h^2(1+h^2) = 3\rho(1+\rho), \end{aligned}$$

$$\mu_4(g_3) = \frac{1}{2\pi i} \oint_{|\xi|=1} |1+h\xi|^6 \frac{1}{\xi^3} d\xi = 3\rho(1+\rho).$$

Hence, we obtain

$$\begin{aligned} EX_{g_3} &= 3\rho(1+\rho) + 3\beta_x \rho(1+\rho) + \frac{3\rho(\beta_x c_g + 2)}{4} (3(1+\rho)^2 + 5\rho + 2) \\ &\quad - \frac{3\rho[8 - 2a_{\mathbf{R}} + \beta_x(4a_g - h_{\mathbf{R}})]}{4} ((1+\rho)^2 + 3\rho + 2). \end{aligned}$$

By Wang and Yao (2013), it yields that

$$\begin{aligned} \mu_3(g_1) + \beta_x \mu_4(g_1) &= 0, \\ \mu_3(g_2) + \beta_x \mu_4(g_2) &= \rho + \beta_x \rho, \\ \mu_3(g_5) + \beta_x \mu_4(g_5) &= \frac{\log(1-\rho)}{2} - \frac{\beta_x \rho}{2}. \end{aligned} \tag{5.1}$$

Furthermore, we have

$$\begin{aligned} \mu_1(g_1) &= -\frac{1}{2\pi i} \oint_{|\xi|=1} |1+h\xi|^2 \left(\frac{1}{h\xi^2} + \frac{2}{h^2\xi^3} \right) d\xi = -1, \quad \mu_2(g_1) = -3, \\ \mu_1(g_2) &= -\frac{1}{2\pi i} \oint_{|\xi|=1} |1+h\xi|^4 \left(\frac{1}{h\xi^2} + \frac{2}{h^2\xi^3} \right) d\xi = -2(1+h^2) - 2 = -2(2+\rho), \\ \mu_2(g_2) &= -2(4+3\rho). \end{aligned}$$

Therefore, we have

$$\begin{aligned} EX_{g_1} &= -\frac{[8 - 2a_{\mathbf{R}} + \beta_x(4a_g - h_{\mathbf{R}})]\rho}{4} + \frac{3(\beta_x c_g + 2)\rho}{4} \\ EX_{g_2} &= \rho + \beta_x \rho - \frac{[8 - 2a_{\mathbf{R}} + \beta_x(4a_g - h_{\mathbf{R}})]\rho}{2} (2+\rho) + \frac{(\beta_x c_g + 2)\rho}{2} (4+3\rho). \end{aligned}$$

5. PROOF OF COROLLARY 3.1 AND EXAMPLE 3.2

Note that

$$\begin{aligned}
\oint_{|\xi|=1} \log(1+h\xi) \frac{1}{\xi} d\xi &= 0, & \oint_{|\xi|=1} \log(1+h\xi^{-1}) \frac{1}{\xi} d\xi &= 0, \\
\oint_{|\xi|=1} \log(1+h\xi) \frac{1}{\xi^2} d\xi &= 2\pi i \cdot h, & \oint_{|\xi|=1} \log(1+h\xi^{-1}) \frac{1}{\xi^2} d\xi &= 0, \\
\oint_{|\xi|=1} \log(1+h\xi) \frac{1}{\xi^3} d\xi &= -2\pi i \cdot \frac{1}{2}h^2, & \oint_{|\xi|=1} \log(1+h\xi^{-1}) \frac{1}{\xi^3} d\xi &= 0, \\
\oint_{|\xi|=1} \log(1+h\xi) \frac{1}{\xi^4} d\xi &= 2\pi i \cdot \frac{1}{3}h^3, & \oint_{|\xi|=1} \log(1+h\xi^{-1}) \frac{1}{\xi^4} d\xi &= 0, \\
\oint_{|\xi|=1} \log(1+h\xi) \frac{1}{\xi^5} d\xi &= -2\pi i \cdot \frac{1}{4}h^4, & \oint_{|\xi|=1} \log(1+h\xi^{-1}) \frac{1}{\xi^5} d\xi &= 0.
\end{aligned}$$

This yields that

$$\mu_1(g_5) = -\frac{1}{2\pi i} \oint_{|\xi|=1} \log(|1+h\xi|^2) \left(\frac{1}{h\xi^2} + \frac{2}{h^2\xi^3} \right) d\xi = 0, \mu_2(g_5) = -2.$$

Together with (5.1), we have

$$EX_{g_5} = \frac{\log(1-\rho)}{2} - \frac{\beta_x \rho}{2} + \frac{(\beta_x c_g + 2)\rho}{2}.$$

Next, we will show the (co)variances of X_{g_j} and X_{g_k} , $j, k = 1, 2, 3, 4$. By Wang and Yao

(2013), it implies that

$$\begin{aligned}
\text{Cov}(X_{g_j}, X_{g_k}) &= -\frac{1}{2\pi^2} \lim_{r \downarrow 1} \oint_{|\xi_1|=1} \oint_{|\xi_2|=1} \frac{g_j(|1+h\xi_1|^2) g_k(|1+h\xi_2|^2)}{(\xi_1 - r\xi_2)^2} d\xi_1 d\xi_2 \\
&\quad - \frac{1}{4\pi^2} \oint_{|\xi_1|=1} \frac{g_j(|1+h\xi_1|^2)}{\xi_1^2} d\xi_1 \oint_{|\xi_2|=1} \frac{g_k(|1+h\xi_2|^2)}{\xi_2^2} d\xi_2 \\
&\quad \cdot (2\beta_x + 2d_{\mathbf{R}} - 2\beta_x q_{\mathbf{G}} - 4).
\end{aligned}$$

In the following, we will compute for $j, k = 1, 2, 3, 4$

$$\begin{aligned}
a_1(g_k, \xi_1) &= \frac{1}{2\pi i} \lim_{r \downarrow 1} \oint_{|\xi_2|=1} \frac{g_k(|1+h\xi_2|^2)}{(\xi_1 - r\xi_2)^2} d\xi_2, \\
a_2(g_j, g_k) &= \frac{1}{2\pi i} \oint_{|\xi_1|=1} a_1(g_k, \xi_1) g_j(|1+h\xi_1|^2) d\xi_1,
\end{aligned}$$

$$a_3(g_j) = \frac{1}{2\pi i} \oint_{|\xi|=1} \frac{g_j(|1+h\xi|^2)}{\xi^2} d\xi.$$

To begin with, we will compute some integrals. By the residue theorem, it follows that

$$\begin{aligned} a_3(g_4) &= \frac{1}{2\pi i} \oint_{|\xi|=1} \frac{12(1+h^2)h^3 + 4(1+h^2)^3h}{\xi} d\xi = 12(1+h^2)h^3 + 4(1+h^2)^3h, \\ a_3(g_3) &= \frac{1}{2\pi i} \oint_{|\xi|=1} \frac{3h(1+h^2)^2 + 3h^3}{\xi} d\xi = 3h(1+h^2)^2 + 3h^3, \\ a_3(g_2) &= \frac{1}{2\pi i} \oint_{|\xi|=1} \frac{2h(1+h^2)}{\xi} d\xi = 2h(1+h^2), \\ a_3(g_1) &= \frac{1}{2\pi i} \oint_{|\xi|=1} \frac{h}{\xi} d\xi = h, \\ a_3(g_5) &= \frac{1}{2\pi i} \oint_{|\xi|=1} \frac{\log(1+h\xi)}{\xi^2} d\xi = h. \end{aligned} \tag{5.2}$$

Furthermore, we obtain

$$\begin{aligned} a_1(g_4, \xi_1) &= \lim_{r \downarrow 1} \frac{1}{2\pi i} \oint_{|\xi_2|=1} \frac{h^4}{\xi_2^4(\xi_1 - r\xi_2)^2} d\xi_2 + \lim_{r \downarrow 1} \frac{1}{2\pi i} \oint_{|\xi_2|=1} \frac{4(1+h^2)h^3}{\xi_2^3(\xi_1 - r\xi_2)^2} d\xi_2 \\ &\quad + \lim_{r \downarrow 1} \frac{1}{2\pi i} \oint_{|\xi_2|=1} \frac{4h^4 + 6(1+h^2)^2h^2}{\xi_2^2(\xi_1 - r\xi_2)^2} d\xi_2 \\ &\quad + \lim_{r \downarrow 1} \frac{1}{2\pi i} \oint_{|\xi_2|=1} \frac{12(1+h^2)h^3 + 4(1+h^2)^3h}{\xi_2(\xi_1 - r\xi_2)^2} d\xi_2 \\ &= \frac{4h^4}{\xi_1^5} + \frac{12(1+h^2)h^3}{\xi_1^4} + \frac{8h^4 + 12(1+h^2)^2h^2}{\xi_1^3} \\ &\quad + \frac{12(1+h^2)h^3 + 4(1+h^2)^3h}{\xi_1^2} \end{aligned} \tag{5.3}$$

and

$$\begin{aligned} a_1(g_3, \xi_1) &= \lim_{r \downarrow 1} \frac{1}{2\pi i} \oint_{|\xi_2|=1} \frac{h^3}{\xi_2^3(\xi_1 - r\xi_2)^2} d\xi_2 + \lim_{r \downarrow 1} \frac{1}{2\pi i} \oint_{|\xi_2|=1} \frac{3h^2(1+h^2)}{\xi_2^2(\xi_1 - r\xi_2)^2} d\xi_2 \\ &\quad + \lim_{r \downarrow 1} \frac{1}{2\pi i} \oint_{|\xi_2|=1} \frac{3h(1+h^2)^2 + 3h^3}{\xi_2(\xi_1 - r\xi_2)^2} d\xi_2 \\ &= \frac{3h^3}{\xi_1^4} + \frac{6h^2(1+h^2)}{\xi_1^3} + \frac{3h(1+h^2)^2 + 3h^3}{\xi_1^2}, \end{aligned} \tag{5.4}$$

where the second equality holds because $\frac{1}{r}\xi_1$ is not a pole as $r > 1$. By Taylor formula

$$\log(1+x) = \sum_{k=1}^{\infty} (-1)^{k-1} \frac{x^k}{k},$$

it follows that

$$\begin{aligned}
 a_1(g_5, \xi_1) &= \lim_{r \downarrow 1} \frac{1}{2\pi i} \oint_{|\xi_2|=1} \frac{\log(1+h\xi_2)}{(\xi_1-r\xi_2)^2} d\xi_2 + \lim_{r \downarrow 1} \frac{1}{2\pi i} \oint_{|\xi_2|=1} \frac{\log(1+h\xi_2^{-1})}{(\xi_1-r\xi_2)^2} d\xi_2 \quad (5.5) \\
 &= \lim_{r \downarrow 1} \sum_{k=1}^{\infty} \frac{(-1)^{k-1} h^k}{k} \frac{1}{2\pi i} \oint_{|\xi_2|=1} \frac{1}{\xi_2^k (\xi_1-r\xi_2)^2} d\xi_2 \\
 &= \lim_{r \downarrow 1} \sum_{k=1}^{\infty} \frac{(-1)^{k-1} h^k}{k} \frac{1}{(k-1)!} \frac{d}{d\xi_2^{k-1}} (\xi_1-r\xi_2)^{-2} \Big|_{\xi_2=0} \\
 &= \sum_{k=1}^{\infty} \frac{(-1)^{k-1} h^k}{\xi_1^{k+1}} = \frac{h}{\xi_1(\xi_1+h)}.
 \end{aligned}$$

Next, we will show the covariances of X_{g_j} and X_{g_5} , $j = 1, 2, 3, 4, 5$. By Wang and Yao (2013), one has

$$\text{Cov}(X_{g_5}, X_{g_5}) = -2\log(1-\rho) + \rho(2d_{\mathbf{R}} - 4)$$

$$\text{Cov}(X_{g_1}, X_{g_5}) = 2\rho + \rho(2d_{\mathbf{R}} - 4).$$

Using (5.3), we compute that

$$\begin{aligned}
 a_2(g_4, g_5) &= \frac{1}{2\pi i} \oint_{|\xi_1|=1} a_1(g_4, \xi_1) \log(|1+h\xi_1|^2) d\xi_1 \\
 &= \frac{4h^4}{2\pi i} \oint_{|\xi_1|=1} \frac{\log(1+h\xi_1)}{\xi_1^5} d\xi_1 + \frac{12(1+h^2)h^3}{2\pi i} \oint_{|\xi_1|=1} \frac{\log(1+h\xi_1)}{\xi_1^4} d\xi_1 \\
 &\quad + \frac{8h^4 + 12(1+h^2)^2 h^2}{2\pi i} \lim_{r \downarrow 1} \oint_{|\xi_1|=1} \frac{\log(1+h\xi_1)}{\xi_1^3} d\xi_1 \\
 &\quad + \frac{12(1+h^2)h^3 + 4(1+h^2)^3 h}{2\pi i} \oint_{|\xi_1|=1} \frac{\log(1+h\xi_1)}{\xi_1^2} d\xi_1 \\
 &= -h^8 + 4(1+h^2)^2 h^6 - 2h^4(2h^2 + 3(1+h^2)^2) \\
 &\quad + 4h^2(1+h^2)(3h^2 + (1+h^2)^2).
 \end{aligned}$$

Moreover, it follows from (5.4) that

$$\begin{aligned}
 a_2(g_3, g_5) &= \frac{1}{2\pi i} \oint_{|\xi_1|=1} a_1(g_3, \xi_1) \log(|1+h\xi_1|^2) d\xi_1 \\
 &= \frac{3h^3}{2\pi i} \oint_{|\xi_1|=1} \frac{\log(1+h\xi_1)}{\xi_1^4} d\xi_1 + \frac{6h^2(1+h^2)}{2\pi i} \oint_{|\xi_1|=1} \frac{\log(1+h\xi_1)}{\xi_1^3} d\xi_1
 \end{aligned}$$

$$\begin{aligned}
 & + \frac{3h(1+h^2)^2 + 3h^3}{2\pi i} \oint_{|\xi_1|=1} \frac{\log(1+h\xi_1)}{\xi_1^2} d\xi_1 \\
 & = h^6 - 3h^4(1+h^2) + 3h^2((1+h^2)^2 + h^2).
 \end{aligned}$$

Applying (5.5), we see that

$$\begin{aligned}
 a_2(g_2, g_5) & = \frac{1}{2\pi i} \oint_{|\xi_1|=1} a_1(g_5, \xi_1) |1+h\xi_1|^4 d\xi_1 \\
 & = \frac{(1+h^2)^2 h + 2h^3}{2\pi i} \oint_{|\xi_1|=1} \frac{1}{\xi_1(\xi_1+h)} d\xi_1 + \frac{2h^2(1+h^2)}{2\pi i} \oint_{|\xi_1|=1} \frac{1}{\xi_1+h} d\xi_1 \\
 & \quad + \frac{2h^2(1+h^2)}{2\pi i} \oint_{|\xi_1|=1} \frac{1}{\xi_1^2(\xi_1+h)} d\xi_1 + \frac{h^3}{2\pi i} \oint_{|\xi_1|=1} \frac{\xi_1}{\xi_1+h} d\xi_1 \\
 & \quad + \frac{h^3}{2\pi i} \oint_{|\xi_1|=1} \frac{1}{\xi_1^3(\xi_1+h)} d\xi_1 = 2h^2(1+h^2) - h^4.
 \end{aligned}$$

Therefore, we obtain

$$\begin{aligned}
 \text{Cov}(X_{g_4}, X_{g_5}) & = 2a_2(g_4, g_5) + a_3(g_4)a_3(g_5)(2d_{\mathbf{R}} - 4) \\
 & = -2\rho^4 + 8(1+\rho)^2\rho^3 - 4\rho^2[2\rho + 3(1+\rho)^2] \\
 & \quad + 4\rho(1+\rho)[3\rho + (1+\rho)^2](2d_{\mathbf{R}} - 2), \\
 \text{Cov}(X_{g_3}, X_{g_5}) & = 2a_2(g_3, g_5) + a_3(g_3)a_3(g_5)(2d_{\mathbf{R}} - \beta_x - 4) \\
 & = 2\rho^3 - 6\rho^2(1+\rho) + 3\rho[(1+\rho)^2 + \rho](2d_{\mathbf{R}} - 2), \\
 \text{Cov}(X_{g_2}, X_{g_5}) & = 2a_2(g_2, g_5) + a_3(g_2)a_3(g_5)(2d_{\mathbf{R}} - \beta_x - 4) \\
 & = -2\rho^2 + 2\rho(1+\rho)(2d_{\mathbf{R}} - 2).
 \end{aligned}$$

Subsequently, we deduce the covariance of X_{g_j} , $j = 1, 2, 3, 4$. From (5.3), we see that

$$\begin{aligned}
 a_2(g_4, g_4) & = \frac{1}{2\pi i} \oint_{|\xi_1|=1} a_1(g_4, \xi_1) |1+h\xi_1|^8 d\xi_1 \\
 & = \frac{4h^4}{2\pi i} \oint_{|\xi_1|=1} \frac{|1+h\xi_1|^8}{\xi_1^5} d\xi_1 + \frac{12(1+h^2)h^3}{2\pi i} \oint_{|\xi_1|=1} \frac{|1+h\xi_1|^8}{\xi_1^4} d\xi_1 \\
 & \quad + \frac{8h^4 + 12(1+h^2)^2 h^2}{2\pi i} \lim_{r \downarrow 1} \oint_{|\xi_1|=1} \frac{|1+h\xi_1|^8}{\xi_1^3} d\xi_1
 \end{aligned}$$

5. PROOF OF COROLLARY 3.1 AND EXAMPLE 3.2

$$\begin{aligned}
& + \frac{12(1+h^2)h^3 + 4(1+h^2)^3h}{2\pi i} \oint_{|\xi_1|=1} \frac{|1+h\xi_1|^8}{\xi_1^2} d\xi_1 \\
& = 4h^8 + 48(1+h^2)^2h^6 + 8h^4[2h^2 + 3(1+h^2)^2]^2 \\
& \quad + 16h^2(1+h^2)^2[3h^2 + (1+h^2)^2]^2, \\
a_2(g_3, g_4) & = \frac{12(1+h^2)h^3}{2\pi i} \oint_{|\xi_1|=1} \frac{|1+h\xi_1|^6}{\xi_1^4} d\xi_1 \\
& \quad + \frac{8h^4 + 12(1+h^2)^2h^2}{2\pi i} \lim_{r \downarrow 1} \oint_{|\xi_1|=1} \frac{|1+h\xi_1|^6}{\xi_1^3} d\xi_1 \\
& \quad + \frac{12(1+h^2)h^3 + 4(1+h^2)^3h}{2\pi i} \oint_{|\xi_1|=1} \frac{|1+h\xi_1|^6}{\xi_1^2} d\xi_1 \\
& = 12(1+h^2)h^6 + 12h^4(1+h^2)[2h^2 + 3(1+h^2)^2] \\
& \quad + 12h^2[3(1+h^2)h^2 + (1+h^2)^3][(1+h^2)^2 + h^2], \\
a_2(g_2, g_4) & = \frac{8h^4 + 12(1+h^2)^2h^2}{2\pi i} \lim_{r \downarrow 1} \oint_{|\xi_1|=1} \frac{|1+h\xi_1|^4}{\xi_1^3} d\xi_1 \\
& \quad + \frac{12(1+h^2)h^3 + 4(1+h^2)^3h}{2\pi i} \oint_{|\xi_1|=1} \frac{|1+h\xi_1|^4}{\xi_1^2} d\xi_1 \\
& = 4h^4[2h^2 + 3(1+h^2)^2] + 8h^2(1+h^2)[3(1+h^2)h^2 + (1+h^2)^3], \\
a_2(g_1, g_4) & = \frac{12(1+h^2)h^3 + 4(1+h^2)^3h}{2\pi i} \oint_{|\xi_1|=1} \frac{|1+h\xi_1|^2}{\xi_1^2} d\xi_1 \\
& = 4h^2[3(1+h^2)h^2 + (1+h^2)^3].
\end{aligned}$$

Hence, we get

$$\begin{aligned}
\text{Cov}(X_{g_4}, X_{g_4}) & = 2a_2(g_4, g_4) + a_3^2(g_4) (2d_{\mathbf{R}} - 4) \\
& = 8\rho^4 + 96\rho^3(1+\rho)^2 + 16\rho^2(2\rho + 3(1+\rho)^2)^2 \\
& \quad + 16\rho(1+\rho)^2(3\rho + (1+\rho)^2)^2(2d_{\mathbf{R}} - 2),
\end{aligned}$$

$$\begin{aligned}
\text{Cov}(X_{g_3}, X_{g_4}) & = 2a_2(g_3, g_4) + a_3(g_3)a_3(g_4)(\beta_x + 2d_{\mathbf{R}} - \beta_x - 4) \\
& = 24(1+\rho)\rho^3 + 24\rho^2(1+\rho)(2\rho + 3(1+\rho)^2) + 12\rho(3(1+\rho)\rho + (1+\rho)^3) \\
& \quad \cdot ((1+\rho)^2 + \rho)(2d_{\mathbf{R}} - 2),
\end{aligned}$$

$$\begin{aligned}\text{Cov}(X_{g_2}, X_{g_4}) &= 2a_2(g_2, g_4) + a_3(g_2)a_3(g_4) (\beta_x + 2d_{\mathbf{R}} - \beta_x - 4) \\ &= 8\rho^2 (2\rho + 3(1 + \rho)^2) + 8\rho(1 + \rho) (3(1 + \rho)\rho + (1 + \rho)^3) (2d_{\mathbf{R}} - 2),\end{aligned}$$

$$\begin{aligned}\text{Cov}(X_{g_1}, X_{g_4}) &= 2a_2(g_1, g_4) + a_3(g_1)a_3(g_4) (2d_{\mathbf{R}} - 4) \\ &= 4\rho (3(1 + \rho)\rho + (1 + \rho)^3) (2d_{\mathbf{R}} - 2).\end{aligned}$$

By calculating and (5.4), one has

$$\begin{aligned}a_2(g_3, g_3) &= \frac{1}{2\pi i} \oint_{|\xi_1|=1} a_1(g_3, \xi_1) |1 + h\xi_1|^6 d\xi_1 \\ &= \frac{3h^3}{2\pi i} \oint_{|\xi_1|=1} \frac{|1 + h\xi_1|^6}{\xi_1^4} d\xi_1 + \frac{6h^2(1 + h^2)}{2\pi i} \oint_{|\xi_1|=1} \frac{|1 + h\xi_1|^6}{\xi_1^3} d\xi_1 \\ &\quad + \frac{3h(1 + h^2)^2 + 3h^3}{2\pi i} \oint_{|\xi_1|=1} \frac{|1 + h\xi_1|^6}{\xi_1^2} d\xi_1 \\ &= 3h^6 + 18h^4(1 + h^2)^2 + 9h^2((1 + h^2)^2 + h^2)^2,\end{aligned}$$

$$\begin{aligned}a_2(g_2, g_3) &= \frac{6h^2(1 + h^2)}{2\pi i} \oint_{|\xi_1|=1} \frac{|1 + h\xi_1|^4}{\xi_1^3} d\xi_1 \\ &\quad + \frac{3h(1 + h^2)^2 + 3h^3}{2\pi i} \oint_{|\xi_1|=1} \frac{|1 + h\xi_1|^4}{\xi_1^2} d\xi_1 \\ &= 6h^4(1 + h^2) + 6h^2(1 + h^2)((1 + h^2)^2 + h^2),\end{aligned}$$

$$a_2(g_1, g_3) = \frac{3h(1 + h^2)^2 + 3h^3}{2\pi i} \oint_{|\xi_1|=1} \frac{|1 + h\xi_1|^2}{\xi_1^2} d\xi_1 = 3h^2((1 + h^2)^2 + h^2).$$

Consequently, we acquire

$$\begin{aligned}\text{Cov}(X_{g_3}, X_{g_3}) &= 2a_2(g_3, g_3) + a_3^2(f_3) (\beta_x + 2d_{\mathbf{R}} - \beta_x - 4) \\ &= 6\rho^3 + 36\rho^2(1 + \rho)^2 + 9\rho((1 + \rho)^2 + \rho)^2 \cdot (2d_{\mathbf{R}} - 2),\end{aligned}$$

$$\begin{aligned}\text{Cov}(X_{g_2}, X_{g_3}) &= 2a_2(g_2, g_3) + a_3(g_2)a_3(g_3) (\beta_x + 2d_{\mathbf{R}} - \beta_x - 4) \\ &= 12\rho^2(1 + \rho) + 6\rho(1 + \rho)((1 + \rho)^2 + \rho) \cdot (2d_{\mathbf{R}} - 2),\end{aligned}$$

$$\text{Cov}(X_{g_1}, X_{g_3}) = 2a_2(g_1, g_3) + a_3(g_1)a_3(g_3) (2d_{\mathbf{R}} - 4) = 3\rho((1 + \rho)^2 + \rho) (2d_{\mathbf{R}} - 2).$$

Finally, from Wang and Yao (2013) it can be proved that

$$\begin{aligned} \text{Cov}(X_{g_1}, X_{g_1}) &= \rho(2d_{\mathbf{R}} - 2), & \text{Cov}(X_{g_1}, X_{g_2}) &= 2\rho(1 + \rho)(2d_{\mathbf{R}} - 2), \\ \text{Cov}(X_{g_2}, X_{g_2}) &= 4\rho^2 + 4\rho(1 + \rho)^2(2d_{\mathbf{R}} - 2). \end{aligned}$$

Thus the proof of Example 3.2 is complete.

6. Some Auxiliary Lemmas

Lemma 6.1 (Lemma B.26 in Bai and Silverstein (2010)). Let $\mathbf{A} = (a_{jk})$ be an $n \times n$ nonrandom matrix and $\mathbf{X} = (x_1, \dots, x_n)'$ be a random vector of independent entries. Assume that $\mathbb{E}x_j = 0$, $\mathbb{E}|x_j|^2 = 1$ and $\mathbb{E}|x_j|^l \leq \nu_l$. Then for $p \geq 1$,

$$\mathbb{E}|\mathbf{X}^* \mathbf{A} \mathbf{X} - \text{tr} \mathbf{A}|^p \leq C_p \left[(\nu_4 \text{tr} \mathbf{A} \mathbf{A}^*)^{p/2} + \nu_{2p} \text{tr} (\mathbf{A} \mathbf{A}^*)^{p/2} \right]$$

where C_p is a constant depending on p only.

Lemma 6.2 (Lemma 4 in Karoui (2009)). Let us focus on $\mathbf{S} = \frac{1}{n} \mathbf{G} \mathbf{X} \mathbf{X}^T \mathbf{G}^T$. When $p/n \rightarrow c$, we have $\max_{j=1, \dots, n} \left| \sqrt{\mathbf{e}_j^T \mathbf{S} \mathbf{e}_j} - 1 \right| \xrightarrow{a.s.} 0$.

Lemma 6.3. For rectangular matrix \mathbf{A} , complex vectors \mathbf{a} and \mathbf{b} , we have $|\mathbf{a}^* \mathbf{A} \mathbf{b}| \leq \|\mathbf{A}\| \sqrt{\mathbf{a}^* \mathbf{a}} \sqrt{\mathbf{b}^* \mathbf{b}}$.

Lemma 6.4. For rectangular matrices $\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D}$, $|\text{tr}(\mathbf{A} \mathbf{B} \mathbf{C} \mathbf{D})| \leq \|\mathbf{A}\| \|\mathbf{C}\| [\text{tr}(\mathbf{B} \mathbf{B}^*)]^{1/2} [\text{tr}(\mathbf{D} \mathbf{D}^*)]^{1/2}$.

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