On fluctuations of eigenvalues of random band matrices

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Abstract
We consider the fluctuation of linear eigenvalue statistics of random band \( n \times n \) matrices whose entries have the form \( M_{ij} = b^{-1/2}u^{1/2}(|i-j|)^{\varepsilon} \tilde{w}_{ij} \) with i.i.d. \( w_{ij} \) possessing the \((4+\varepsilon)\)th moment, where the function \( u \) has a finite support \([-C^*,C^*]\), so that \( M \) has only \( 2C^*b+1 \) nonzero diagonals. The parameter \( b \) (called the bandwidth) is assumed to grow with \( n \) in a way that \( b/n \to 0 \). Without any additional assumptions on the growth of \( b \) we prove CLT for linear eigenvalue statistics for a rather wide class of test functions. Thus we improve and generalize the results of the previous papers [8] and [11], where CLT was proven under the assumption \( n >> b >> n^{1/2} \). Moreover, we develop a method which allows to prove automatically the CLT for linear eigenvalue statistics of the smooth test functions for almost all classical models of random matrix theory: deformed Wigner and sample covariance matrices, sparse matrices, diluted random matrices, matrices with heavy tales etc.

1 Introduction and main results
Consider an ensemble of random symmetric \( n \times n \) matrices with entries of the form
\[
\mathcal{M}_{ij} = (u_{ij}/b)^{1/2}\tilde{w}_{ij}, \quad u_{ij} = u(|i-j|/b)
\]
where \( \{\tilde{w}_{ij}\}_{ij} \) are i.i.d. (up to the symmetry \( \tilde{w}_{ij} = \tilde{w}_{ji} \)) random variables, satisfying the moment conditions
\[
E[\tilde{w}_{ij}] = 0, \quad E[|\tilde{w}_{ij}|^2] = 1, \quad E[|\tilde{w}_{ij}|^4] = 3 + \kappa_4, \quad E[|\tilde{w}_{ij}|^{4+\varepsilon}] \leq C < \infty,
\]
and \( u(x) \) is a piece-wise continuous (with a finite number of jumps) continuous at \( x = 0 \) function with a compact support, satisfying the conditions
\[
u(x) = u(-x), \quad 0 \leq u(x) \leq C, \quad \int u(x)dx = 1, \quad \text{supp } u \subset [-C^*,C^*].
\]

It is easy to see that the entries of \( \mathcal{M} \) are nonzero only inside the band \(|i-j| \leq C^*b \). Hence for fixed \( b \) we have a matrix with a finite numbers of diagonals, while if \( b \sim n \), we obtain some kind of the Wigner matrix, with all of the entries having the variances of the same order (see [20]). The model is now widely discussed in mathematical literatures, since by non rigorous conjecture of [6] it is expected that the behavior of local eigenvalue statistics demonstrates a kind of phase transition: for \( b << n^{1/2} \) the statistics is of Poisson type and for \( b >> n^{1/2} \) it is of the same type as for Wigner matrices. Till now this result is not proven rigorously, but the problem is one of the most challenging in the random matrix theory (see, e.g. [19], [3], [4], [18] and references therein).

It was proved many years ago (see [10]) that in the limit
\[
b \to \infty, \quad b/n \to 0, \quad \text{as } n \to \infty,
\]
the normalized eigenvalue counting measure converges weakly to the Wigner semicircle low, which has the density
\[
\rho_{sc}(\lambda) = \frac{1}{2\pi} \sqrt{4 - \lambda^2} 2_{[-2,2]}.
\]
This means that if we denote \( \{\lambda_i\}_{i=1}^n \) the eigenvalues of \( M \), choose any bounded integrable test function \( \varphi \), and consider the linear eigenvalue statistics of the form

\[
N_n[\varphi] = \sum_{j=1}^n \varphi(\lambda_j), \quad N_n^\circ[\varphi] = N_n[\varphi] - E[N_n[\varphi]],
\]

then in the limit (1.4) we have

\[
E\{n^{-1}N_n[\varphi]\} \to \int \varphi(\lambda)\rho_{ac}(\lambda) d\lambda, \quad \text{Var}\{N_n[\varphi]\} \to 0.
\]

In particular, for \( \varphi(\lambda) = (\lambda - z)^{-1} \)

\[
n^{-1}N_n[\varphi] = n^{-1}\text{Tr}(M - z)^{-1} \to g(z),
\]

\[
g(z) = \frac{1}{2}(-z + \sqrt{z^2 - 4}).
\]

The next natural question is the behavior of the fluctuations \( N_n^\circ[\varphi] \) in the same limit, in particular, the behavior of its variance. This question was solved partially in the paper [9], where the main term of the covariance of the traces of two resolvents was found in the case of Gaussian \( \omega \), under the additional restriction \( b = n^\theta, \, 1/3 < \theta < 1 \). The next step was done in the papers [11] and [8], where the Central Limit Theorem (CLT) for the random variable \( \sqrt{b/n}N_n^\circ[\varphi] \) was proved for sufficiently smooth test functions, but again under the technical condition \( n >> b >> n^{1/2} \).

The main result of the present paper is the proof of CLT for the linear eigenvalue statistics (1.6) of the band matrices under the limiting transition (1.4) without any additional restriction on the growth of \( b \).

We consider the test functions from the space \( \mathcal{H}_s \), possessing the norm

\[
\|\varphi\|_s^2 = \int (1 + 2|k|)^s |\hat{\varphi}(k)|^2 dk, \quad s > 2, \quad \hat{\varphi}(k) = \int e^{ikx}\varphi(x)dx.
\]

**Theorem 1** Consider the model of band matrices (1.1)-(1.3) and any test function possessing the norm (1.8) with \( s > 2 \). Then the sequence of random variables \( \sqrt{b/n}N_n^\circ[\varphi] \) with \( N_n^\circ[\varphi] \) of (1.6) converges in distribution in the limit (1.4) to the normal random variable with zero mean and the variance

\[
V[\varphi] = \frac{1}{\pi^2} \int_0^\pi \int_0^\pi dx dy \varphi(2\cos x)\varphi(2\cos y) \int \frac{\partial^2}{\partial x\partial y} \log \frac{1 - \hat{u}(k)e^{ix+y}}{1 - \hat{u}(k)e^{ix-y}} dk \\
+ \frac{(u,u)\kappa_k}{\pi^2} \left( \int_0^\pi \varphi(2\cos x)\cos 2xdx \right)^2 + \frac{u(0)}{2\pi^2} \left( \int_0^\pi \varphi(2\cos x)\cos xdx \right)^2,
\]

where \( (u,u) = \int u^2(x)dx \) and \( \hat{u}(k) \) is the Fourier transform of the function \( u \) defined as in (1.8).

To prove CLT for the band matrices, we use the CLT for martingales (see [2] Theorem 35.12)).

**Theorem 2** Let \( X_{n,k} = E_{<k}[Y - E_{<k}Y] \) be a martingale differences array with respect to independent random vectors \( V_1, \ldots, V_n, S_n = \sum_{k=1}^n X_k, \sigma_n = \sum_{k=1}^n E[X_k^2] = O(1) \). Assume that

\[
(1) \quad \sum E\{X_{k}^4\} \leq \varepsilon_n, \quad (2) \quad \text{Var}\left\{ \sum_{k=1}^n X_k^2 \right\} \leq \tilde{\varepsilon}_n
\]

Then

\[
|E\{e^{itS_n}\} - e^{-t^2\sigma_n/2}| \leq C'(t)(\varepsilon_n^{1/2} + \tilde{\varepsilon}_n^{1/2}).
\]

**Remark:** Here we have replaced a more general condition \( \sum E\{X_k^4|X_k^2 > \delta\} \to 0 \) used in [2] by condition (1) which is more easy to check for the random matrix models.
The idea to use Theorem 2 for the proof of CLT in the random matrix theory is not new. Since the paper of [1] it was used many times (see, e.g., [2, 15] and [13]), but the method of the proof of CLT used in the present paper allows to prove CLT by the same way for all classical models of random matrix theory: deformed Wigner and sample covariance matrices, sparse and diluted random matrices etc. It becomes even simpler than that for band matrices, since the proof of condition (2) becomes simpler.

The paper is organized as follows. In Section 2.1 we give the sketch of the proof of CLT, introduce truncated band matrix and explain how one can extend CLT from some special class of the test functions to all functions of \( H_s \). In Section 2.2 we check conditions (1.10) and in Section 2.3 prove Lemma 1 (given in Section 2.1) about the difference of linear eigenvalue statistics of initial and truncated matrices. In Section 3 we compute the variance (1.9). And in Section 4 the proofs of some auxiliary results (partially known before) are given in order to make the proof of Theorem 1 more self consistent.

## 2 Proof of CLT

### 2.1 Strategy of the proof

We start from the proof of CLT for the truncated and "periodically continued" model:

\[
M_{ij} = (u_{ij}/b)^{1/2}w_{ij}, \quad u_{ij} = u(|i-j|_n/b)
\]

\[
w_{ij} = \begin{cases} \tilde{w}_{ij}1_{|\tilde{w}_{ij}| \leq b^{1/2}} - E\{\tilde{w}_{ij}1_{|\tilde{w}_{ij}| \leq b^{1/2}}\}, & |i-j| \leq C^*b \\ \omega_{ij}, & ||i-j|-n| \leq C^*b \end{cases}
\]

Here and below

\[
|i-j|_n := \max\{|i-j|, ||i-j|-n|\},
\]

and \( \{\omega_{ij}\}_{|i-j|-n| \leq C^*b} \) are independent (up to the symmetry conditions) and independent from \( M \) copies of \( w_{12} \). Thus we not only truncated the entries of \( M \), but also add entries in upper right and lower left parts of it, in order to obtain the periodic distribution, i.e., invariant with respect to the shift \( i \to |i + 1|_n \).

Then the standard argument gives us that for \(|i-j|_n \leq C^*b\)

\[
E\{|w_{ij}|\} = 0, \quad E\{|w_{ij}|^2\} = 1 + O(b^{-1-\varepsilon/2}), \\
E\{|w_{ij}|^4\} = 3 + \kappa_4 + O(b^{-\varepsilon/2}), \quad E\{|\omega_{ij}|^8\} \leq Cb^{4-\varepsilon/2}.
\]

Moreover, it is easy to see that

\[
n^{-1}E\{\text{Tr}(M-M)^2\} \leq Cb^{-\varepsilon/2}.
\]

Then, using Theorem 2, we prove CLT for \( \nu_{1n} := (b/n)^{1/2}\mathcal{N}_n[\varphi, M] \) with the test functions of the form

\[
\varphi\eta = \varphi \ast P_{\eta},
\]

where \( \ast \) means a convolution, \( P_{\eta} \) is a Poisson kernel

\[
P_{\eta}(\lambda) = \frac{\pi^{-1}\eta}{\lambda^2 + \eta^2},
\]

and \( \varphi \in H_s \cap L_1(\mathbb{R}) \). It is easy to see that then

\[
\mathcal{N}_n[\varphi\eta, M] = \pi^{-1} \int \varphi(\lambda)3\gamma_n(\lambda + i\eta)d\lambda.
\]

Then we shall prove the lemma
Lemma 1. Set \( G(z) = (M - z)^{-1} \), \( \tilde{\gamma}_n(z) := \text{Tr} G(z) \). Then for any \( z : \Re z > \eta \)
\[
\frac{b}{n} \text{Var}\{\gamma_n(z) - \tilde{\gamma}_n(z)\} \leq Cb^{-\varepsilon/2}.
\] (2.8)

The lemma implies that for any \( \varphi \in H_s \cap L_1(\Re) \) if we set \( \nu_{2n} := (b/n)^{1/2} N^n_n(\varphi, M) \), then
\[
\text{Var}\{\nu_{2n} - \nu_{1n}\} = \frac{b}{n} \text{Var}\{N_n[\varphi, M] - N_n[\varphi, M]\}
\]
\[
= \frac{b}{\pi^{2n}} \int \int \text{d}\lambda_1 \text{d}\lambda_2 \varphi(\lambda_1) \varphi(\lambda_2)
\]
\[
\times \text{Cov}\{\Im \gamma_n(\lambda_1 + i\eta) - \Im \tilde{\gamma}_n(\lambda_1 + i\eta), \Im \gamma_n(\lambda_2 + i\eta) - \Im \tilde{\gamma}_n(\lambda_2 + i\eta)\}
\]
\[
\leq Cb^{-\varepsilon} \int \int \text{d}\lambda_1 \text{d}\lambda_2 |\varphi(\lambda_1) \varphi(\lambda_2)| \leq Cb^{-\varepsilon/2}.
\]

Hence, for any fixed \( x \in \Re \)
\[
|E\{e^{ix\nu_{1n}}\} - E\{e^{ix\nu_{2n}}\}| \leq x \text{Var}^{1/2}\{\nu_{1n} - \nu_{2n}\} \leq b \text{C}b^{-\varepsilon/4}.
\]

Thus, CLT for \( \nu_{1n} \) and Lemma \( \text{1} \) imply CLT for \( \nu_{2n} \) if the test function has the form \( (2.5) \).

To extend CLT to the test functions from \( H_s \), we use a proposition (see [14, Proposition 3.2.9]).

Proposition 1 Let \( \{\xi^{(n)}_l\}_{l=1}^n \) be a triangular array of random variables, \( N_n[\varphi] = \sum_{l=1}^n \varphi(\xi^{(n)}_l) \) be its linear statistics, corresponding to a test function \( \varphi : \Re \rightarrow \Re \), and \( \{d_n\} \) is some sequence of positive numbers. Assume that

(a) there exists a vector space \( L \) endowed with a norm \(|\cdot|\) and such that \( V_n \) is defined on \( L \) and admits the bound
\[
d_n \text{Var}\{N_n[\varphi]\} \leq C|\varphi|^2, \ \forall \varphi \in L;
\] (2.9)

(b) there exists a dense linear manifold \( L_1 \subset L \) such that CLT is valid for \( N_n[\varphi], \varphi \in L_1 \), i.e., there exists a continuous quadratic functional \( V : L_1 \rightarrow \Re_1 \) such that we have uniformly in \( x \), varying on any compact interval
\[
\lim_{n \rightarrow \infty} Z_n[x]\varphi = e^{-x^2V[\varphi]/2}, \ \forall \varphi \in L_1, \ \text{where} \ Z_n[x]\varphi := E\{e^{ixd_{1/2}N^n_n[\varphi]}\}.
\] (2.10)

Then \( V \) admits a continuous extension to \( L \) and CLT is valid for all \( N_n[\varphi], \varphi \in L \).

The proposition allows to extend CLT from any dense subset of \( H_s \) for which we are able to prove CLT on the whole \( H_s \), if we can check \( (2.9) \). This can be done by using the another proposition (proven in [16] and also [17] and Lemma 2).

Proposition 2 For any \( s > 0 \) and any \( M \)
\[
\text{Var}\{N_n[\varphi, M]\} \leq C_s|\varphi|^2 \int_0^\infty \text{d}y y^{2s-1} \int_{-\infty}^\infty \text{Var}\{\text{Tr} G(x + iy)\} \text{d}x.
\] (2.11)

Lemma 2 If the conditions \( (1.1) \) and \( (1.3) \) are satisfied, then for any \( 0 < y < 1 \)
\[
\frac{b}{n} \int \text{d}x \text{Var}\{\text{Tr} G(x + iy)\} \leq Cy^{-4} \log y^{-1}
\] (2.12)

The proof of the lemma is given in Section 4.

Combining the proposition with \( (2.12) \), we prove \( (2.10) \).
2.2 Checking of conditions (1.10)

To apply Theorem 2, we denote $E_p$ the averaging with respect to the variable $\{w_{p,j}\}_{j \geq p}$, $E_{<p} = E_1 \ldots E_{p-1}$ and consider

$$X_p[\varphi_\eta] = \pi^{-1}(b/n)^{1/2} \int \varphi(\lambda)\Delta X_p[\lambda + i\eta]d\lambda,$$

$$\tilde{X}_p[z] = E_{<p}\{\gamma_n(z) - E_p\gamma_n(z)\}.$$  

Then, according to Theorem 2, we have to check condition (1)-(2) of (1.10) for $\{X_p[\varphi_\eta]\}$. It is evident, that conditions (1) follow from the bounds

$$E_p\{|X_p[z]|^2\} \leq Cb^{-1}, \quad |\tilde{X}_p[z]| \leq C,$$  

(2.14)

valid uniformly in $|\Im z| \geq \eta$. And since

$$\text{Var}\left\{\sum X_p^2 - \sum E_p\{X_p^2\}\right\} = \sum_p E\{|X_p^2 - E_p\{X_p^2\}|^2\} \leq \sum_p E\{X_p^4\} \leq \varepsilon_n,$$

condition (2) of (1.10) follows from the uniform in $|\Im z_1|, |\Im z_2| \geq \eta$ bound

$$\text{Var}\{\Sigma(z_1, z_2)\} \leq \varepsilon_n,$$

$$\Sigma(z_1, z_2) := \frac{b}{n} \sum_p E_p\{\tilde{X}_p[z_1]\tilde{X}_p[z_2]\}.$$  

Let us prove (2.14) and (2.15).

Denote $M^{(p)}$ the $(n-1) \times (n-1)$ matrix which is obtained from $M$ by removing the $p$th line and column. Set also

$$G^{(p)} = (M^{(p)} - z)^{-1}, \quad v^{(p)} := (v_{p1}, \ldots, v_{pm}) \in \mathbb{R}^{n-1}, \quad v_{ij} := u_{ij}^{1/2} w_{ij},$$

(2.16)

Use the identities

$$G_{pp} = -A_p^{-1}, \quad G_{ij} = G^{(p)}_{ij} - Q^{(p)}_{ij}, \quad \text{Tr } G - \text{Tr } G^{(p)} = -\frac{\partial}{\partial z} \log A_p(z),$$

(2.17)

where

$$A_p := z + b^{-1/2}v_{pp} + b^{-1}(G^{(p)}v^{(p)}, v^{(p)}),$$

$$Q^{(p)}_{ij} = b^{-1}A_p^{-1}(G^{(p)}v^{(p)}, v^{(p)})_{ij}.$$  

(2.18)

Since for the resolvent $G(z) = (M - z)^{-1}$ of any symmetric or hermitian matrix $M$ and any vector $m$

$$\Im(G(z)m, m) = \Im z(G(z)m, G(z)m),$$

(2.19)

we have for $|\Im z| \geq \eta$

$$|A_p(z)| \geq |\Im A_p(z)| = |\Im z|(1 + b^{-1}(G^{(p)}v^{(p)}, G^{(p)}v^{(p)})) \geq \eta,$$  

$$|\tilde{A}_p| \geq |\Im \tilde{A}_p| \geq \eta, \quad \text{where } \tilde{A}_p := E_p\{A_p\},$$

$$||Q^{(p)}|| \leq |A_p|^{-1}|b^{-1}(G^{(p)}v^{(p)}, G^{(p)}v^{(p)})| \leq \eta$$  

(2.20)

and

$$\left|\frac{A_{p}'(z)}{A_p}\right| \leq \frac{|1 + b^{-1}((G^{(p)})^2v^{(p)}, v^{(p)})|}{\Im A_p} \leq \eta^{-1} \Rightarrow |\tilde{X}_p| \leq 2\eta^{-1},$$

which implies the second inequality of (2.14).
The last relation of \((2.17)\) yields

\[
E_p\{X_p(z_1)X_p(z_2)\} = \frac{\partial^2}{\partial z_1 \partial z_2} D_p(z_1, z_2)
\]

\[
D_p(z_1, z_2) := E_p\left\{ E_{<p}\{\log A_p(z_1)\} E_{<p}\{\log A_p(z_2)\} \right\}.
\]

Here and below for any random variable \(\xi\) we denote \(\xi_p = \xi - E_p\{\xi\}\).

Since \(D_p(z_1, z_2)\) is an analytic function on \(z_1, z_2 : |\Re z_1|, |\Re z_2| \geq \eta/2\), in order to prove the first bound of \((2.14)\) it suffices to prove that uniformly in \(|\Re z| \geq \eta/2\)

\[
E_p\{|E_{<p}\{\log A_p(z)\}p\}|^2 \leq \eta^{-2} E_p\{|E_{<p}\{A_p(z)\}|^2\} \leq Cb^{-1}.
\]

Evidently

\[
E_{<p}\{A_p(z)\} = b^{-1/2}v_{pp} + b^{-1} \sum_{i,j \neq p} G_{ij}(z)v_{pi}v_{pj} + \sum_{i>p} G_{ii}(z)(v_{pi} - u_{pi}).
\]

Hence, averaging with respect to \(E_p\) and using \((2.23)\), we obtain the first bound of \((2.14)\). Similarly one can get the relation which we need below

\[
E_p\{|E_{<p}\{A_p(z)\}p\}|^4 \leq Cb^{-1-\varepsilon/2}.
\]  \((2.22)\)

We are left to check \((2.15)\). Writing \(A_p = \tilde{A}_p + A_p^0\), expanding \(\log A_p\) around \(\tilde{A}_p\), and using \((2.22)\), we obtain

\[
\Sigma(z_1, z_2) = \frac{b}{n} \sum_p E_p\{X_p(z_1)X_p(z_2)\} = \frac{\partial^2}{\partial z_1 \partial z_2} \hat{\Sigma}(z_1, z_2)
\]  \((2.23)\)

\[
\hat{\Sigma}(z_1, z_2) := \frac{b}{n} \sum_p D_p(z_1, z_2) = \frac{b}{n} \sum_p (\tilde{A}_p(z_1)\tilde{A}_p(z_2))^{-1} T_p(z_1, z_2)
\]

\[
+ \frac{b}{n} \sum_p (O(E_p\{|A_p^0(z_1)|^3\}) + O(E_p\{|A_p^0(z_2)|^3\}))
\]

\[
= \frac{b}{n} \sum_p (\tilde{A}_p(z_1)\tilde{A}_p(z_2))^{-1} T_p(z_1, z_2) + O(b^{-\varepsilon/4}),
\]

where

\[
T_p(z_1, z_2) := E_p\left\{ E_{<p}\{A_p^0(z_1)\} E_{<p}\{A_p^0(z_2)\} \right\}
\]  \((2.24)\)

\[
= 2b^{-2} \sum_{i,j > p} u_{pi}u_{pj} E_p\{ E_{<p}\{G_{ij}(z_1)\} E_{<p}\{G_{ij}(z_2)\} \} + \kappa_4 b^{-2} \sum_{i,j > p} u_{pi}^2 E_p\{ E_{<p}\{G_{ij}(z_1)\} E_{<p}\{G_{ij}(z_2)\} \} + b^{-1} u_{pp}.
\]

**Lemma 3** Given \(\eta > 0\) there exists \(\delta(\eta) > 0\) such that uniformly in \(z : |\Re z| > \eta\)

\[
\Var\{G_{ij}(z)\} \leq b^{-\delta}, \quad E\{|G_{ij}(z) - E\{G_{ij}(z)\}| \} \leq C_0 b^{-1},
\]

\[
E\{|G_{ij}(z) - g(z)| \} \leq C_0 b^{-\delta},
\]  \((2.25)\)

where \(g(z)\) is defined by \((1.7)\).

The proof of the lemma is given in Section 4.

**Remark 1** Below we will often use a simple observation. If for some random variables \(|R_k| \leq C_k\), \(\sum_k C_k \leq C\), and \(f_k : E\{|f_k - f_k^0| \} \leq C_1 b^{-\delta}\), where \(f_k^0\) are some constants, then we have with the same \(C\) and \(C_0\) of \((2.25)\)

\[
\sum_k R_k f_k = \sum_k R_k f_k^0 + r, \quad E\{|r| \} \leq CC_1 b^{-\delta}.
\]  \((2.26)\)
In particular, since in view of (2.24) \(|T_p(z_1, z_2)| \leq Cb^{-1}\), we have

\[
\tilde{\Sigma}(z_1, z_2) = \frac{b}{n} \sum \tilde{A}_p(z_1) \tilde{A}_p(z_2) \left( T_p(z_1, z_2) + o(1) \right)
\]

(2.27)

where \(T_p(z_1, z_2)\) is the first sum in the r.h.s. of (2.24). The constant term here does not contribute into the variance of \(\Sigma(z_1, z_2)\), so it is not important in the proof of (2.15).

Let us denote \(\tilde{M}^{<p}\) the matrix \(M\) whose entries \(w_{ij}\) with \(\min\{i, j\} < p\) are replaced by \(\tilde{w}_{ij}\) which are independent from all \(\{w_{kl}\}_{k,l=1}^n\) and have the same distribution as \(w_{ij}\). Let also \(\tilde{M}^{(p,q)}\) be the matrix \(\tilde{M}^{(<p,q)}\) without \(q\)th line and column. We denote also \(\tilde{E}_{<p}\) the averaging with respect to all \(w_{ij}\) and \(\tilde{E}_{ij}\) with \(\min\{i, j\} < p\). Set

\[
\tilde{G}^{(p,q)} = \left( \tilde{M}^{(p,q)} \right)^{-1}, \quad \tilde{G}^{(p)} = \left( \tilde{M}^{(p)} \right)^{-1}.
\]

(2.28)

Then evidently

\[
T_p(z_1, z_2) = \sum_{jk} E_p\{E_{<p}\{\tilde{G}^{(p)}>_{jk}(z_1)G^{(p)}(z_2)\}\}\{u_{jp}u_{kp}\}
\]

(2.29)

Moreover, if we replace \(G^{(p)}\) in (2.24) by \(G\) and set

\[
T_p^{o}(z_1, z_2) = \sum_{i,j>p} u_{pi}u_{pj}E_p\{E_{<p}\{G_{ij}(z_1)\}\}{\}\{G_{ij}(z_1)\}\}
\]

(2.30)

then in view of (2.17) and (2.24)

\[
\left| T_p^{o}(z_1, z_2) - T_p(z_1, z_2) \right| \leq \left| E_p\{E_{<p}\{\tilde{G}^{(p)}>_{ij}(z_1)G(z_2)\}\}\{G^{(p)}(z_2)\}\}\{G^{(p)}(z_2)\}\right| \leq C,
\]

(2.31)

where we have used that since \(Q^{(p)}\) is a rank one matrix with a bounded norm, we have for any bounded matrix \(B\)

\[
\text{Tr} Q^{(p)}B \leq \|B\|\|Q\|.
\]

Thus we need to study the variance of

\[
\Sigma_1 = \frac{1}{bn} \sum_p T_p^{o}(z_1, z_2).
\]

(2.32)

To prove (2.15), it suffices to show that

\[
\text{Var}\{\Sigma_1\} = \sum_r E\{|E_{<r}\{(\Sigma_1)^2_r\}|^2\} \to 0.
\]

The last relation is a corollary of the bounds, which we are going to prove

\[
n^2 E\{|(\Sigma_1)^2_r\} \leq C, \quad r = 1, \ldots, n.
\]

(2.33)

By (2.32),

\[
n(\Sigma_1)^2_r = \frac{1}{b} \sum_{p \leq r} (T_p^{o}(z_1, z_2))^{2}_r
\]

(2.34)
Notice also that \((T_p''(z_1, z_2))_r^o = 0\) for \(p \geq r + 1\), hence the sum in (2.33) is over \(p \leq r\). Then (2.17) yields
\[
(T_p''(z_1, z_2))_r^o = \left(\hat{E}_{p}\{\text{Tr} \hat{G}^{<p}(z_1) I(p) G(z_2) I(p)\} - \hat{E}_{p}\{\text{Tr} \hat{G}^{<p,r}(z_1) I(p) G(r) (z_2) v(r)\}\right)_r^o
\]
\[
+ \text{sim} + \left(\hat{E}_{p}\{(A, b)^{-2}(G(r)(z_2) I(p) \hat{G}^{(r)}(z_1) \hat{G}^{(r)}(z_1)) v(r), v(r)^2\}\right)_r^o
\]
\[
= : (\hat{F}_1^r(z_1, z_2))_r^o + (\hat{F}_2^r(z_1, z_2))_r^o + (\hat{F}_2^r(z_1, z_2))_r^o,
\]
where "+sim" means the adding of the term which can be obtained from the previous one by replacing \(z_2\) and \(z_1\). Since \(E\{\|\xi^o\|^2\} \leq E\{\|\xi^o\|^2\}\) for any random variable \(\xi\), (2.34) yields
\[
n^2 E\{\|\Sigma_1\|^2\} \leq CE\left\{b^{-1} \sum_{p \leq r} (\hat{F}_1^p(z_1, z_2) + \hat{F}_2^p(z_1, z_2))\right\}^2\]
(2.35)
\[
\leq CE\left\{b^{-2} \sum_{p \leq r} E_{\leq p}\{(I(p) G(r)(z_1) v(r), G(r)(z_1) v(r)) (1 + b^{-1}(v(r), v(r)))\}\right\}^2 + \text{sim}
\]
\[
= : CE\left\{b^{-2} \sum_{p \leq r} (F_1^p(z_1) + F_2^p(z_2))\right\}^2.
\]
To sum in the r.h.s of (2.35) with respect to \(p\) we would like to use the property
\[
\sum_{p=1}^n I(p) \leq CbI,
\]
(2.36)
but since \(p\) appears not only in \(I(p)\), we need to remove \(p\) from the other places first. Write
\[
E\{u^2|\Sigma_1^o|\} \leq Cb^{-4} \sum_{p \leq q \leq r} E\{F_1^p F_1^q\}
\]
\[
\leq Cb^{-4} \sum_{q=1}^r \sum_{p=1}^q E\{E_{\leq q}\{(I(p) G(r)(v(r), G(r)(v(r)) (1 + b^{-1}(v(r), v(r)))\}\} F_1^q\}
\]
\[
\leq Cb^{-4} \sum_{q=1}^r E\{E_{\leq q}\{b^{-1}(v(r), v(r)) (1 + b^{-1}(v(r), v(r)))\}\} F_1^q\}
\]
\[
\leq Cb^{-4} \sum_{q=1}^r E\{E_{\leq q}\{b^{-1}(v(r), v(r)) (1 + b^{-1}(v(r), v(r)))\}^3\}
\]
\[
+ CE\{(1 + b^{-1}(v(r), v(r)))^4\} \leq C'E\{(1 + b^{-1}(v(r), v(r)))^4\}.
\]
Here in the first line we use (2.35), in the second line we use first that for \(p \leq q\) the averaging \(E_{\leq p}\) can be replaced by \(E_{\leq q}\), and then use (2.36) for summation over \(p \leq r\). The third line follows from the second one in view of the bound \(\|G(r)\| \leq C\). Next we split the sum over \(q\) into two parts: one over \(q < r - C^*b\) and another over \(r - C^*b \leq q \leq r\), and observed that for the \(q\) in the first part \((v(r), v(r))\) is a constant with respect to the averaging \(E_{\leq q}\), hence
\[
E\{E_{\leq q}\{b^{-1}(v(r), v(r)) (1 + b^{-1}(v(r), v(r)))\}\} F_1^q\}
\]
\[
= E\{\{(G(r) v(r), G(r) v(r)) b^{-1}(v(r), v(r)) (1 + b^{-1}(v(r), v(r)))\}^2\},
\]
Then we can take the sum over \(q < r - C^*b\), using again the bound (2.30), and finish to estimate the sum using the bound \(\|G(r)\| \leq C\). As for the terms with \(r - C^*b \leq q \leq r\), they are estimated just using the boundedness of \(\|G(r)\|\) and \(\|I(p)\|\). Thus we have proved (2.33).
\[\square\]
2.3 Proof of Lemma \[1\]

Set 

\[ G^{(p)} := (\mathcal{M}^{(p)} - z)^{-1}, \quad A_p := z + b^{-1}(G^{(p)}\tilde{G}^{(p)}, v^{(p)}), \quad \Delta A_p := A_p - A_p. \]

The same argument as in the previous section implies that it suffices to check that

\[ \frac{b}{n} \sum_p E\{|\Delta A_p - E_p\{\Delta A_p\}|^2\} \rightarrow 0. \tag{2.37} \]

Since we know that (see (2.24))

\[ \frac{b}{n} \sum_{|p|,n \leq C^*b} E\{|\Delta A_p - E_p\{\Delta A_p\}|^2\} \leq \frac{b}{n} \sum_{|p|,n \leq C^*b} 2\left(E\{|A_p|\}^2 + E\{|A_p|^2\}\right) \leq Cb, \]

we conclude that it suffices to prove that

\[ \frac{b}{n} \sum_{|p|,n > C^*b} E\{|\Delta A_p - E_p\{\Delta A_p\}|^2\} \rightarrow 0. \tag{2.38} \]

Let us write

\[ \Delta A_p = b^{-1/2}\Delta v_{pp} + b^{-1}(G^{(p)}\Delta v^{(p)}, \Delta v^{(p)}) + 2b^{-1}(G^{(p)}\Delta v^{(p)}, v^{(p)}) \tag{2.39} \]

\[ + b^{-1}((G^{(p)} - G^{(p)})v^{(p)}, v^{(p)}) =: J_{0p} + J_{1p} + 2J_{2p} + J_{3p}. \]

Averaging with respect to \(v^{(p)}\) and \(\tilde{v}^{(p)}\) we get similarly to (2.24) for \(|p|,n \geq cb\)

\[ E\{|J_{1p} - E_p\{J_{1p}\}|^2\} = b^{-2} \sum_{i \neq j} E\{|G^{(p)}_{ij}|^2(v_{pi} - \tilde{v}_{pi})^2(v_{pj} - \tilde{v}_{pj})^2\} \tag{2.40} \]

\[ + b^{-2} \sum_{i} E\{|G^{(p)}_{ii}|^2(v_{pi} - \tilde{v}_{pi})^4\} \leq b^{-4-\varepsilon} \sum_{i \neq j} E\{|G^{(p)}_{ij}|^2|I_{ij}^{(p)}|I_{jj}^{(p)} + b^{-2-\varepsilon/2} \sum_{i} E\{|G^{(p)}_{ii}|^2|I_{ii}^{(p)}| \leq Cb^{-1-\varepsilon/2}. \]

Similarly

\[ E\{|J_{2p} - E_p\{J_{2p}\}|^2\} \leq Cb^{-2-\varepsilon/2}, \quad E\{|J_{0p} - E_p\{J_{0p}\}|^2\} \leq Cb^{-2-\varepsilon/2}. \tag{2.41} \]

In addition, again similarly to (2.24) we have

\[ E\{|J_{3p} - E_p\{J_{3p}\}|^2\} \leq Cb^{-2}E\{|\text{Tr} I^{(p)}(G^{(p)} - G^{(p)})I^{(p)}(G^{(p)} - G^{(p)})\}. \tag{2.42} \]

Now by the same way as in (2.30)-(2.31) we can replace here \(G^{(p)}\) by \(G\) and \(G^{(p)}\) by \(G\) with an error \(O(b^{-2})\):

\[ E\{|J_{3p} - E_p\{J_{3p}\}|^2\} \leq 2b^{-2}E\{|\text{Tr} I^{(p)}(G - G)I^{(p)}(G - G)|\} \tag{2.43} \]

The resolvent identity implies

\[ G - G = G(M^{(p)} - \mathcal{M})G = -G\Delta MG. \]

Hence, the last term in the r.h.s. of (2.42) can be estimated as

\[ b^{-2}E\{|\text{Tr} I^{(p)}(G - G)I^{(p)}(G - G)|\} = b^{-2}E\{|\text{Tr} I^{(p)}G\Delta MG I^{(p)}G^*\Delta MG^*|\} \leq Cb^{-2}E\{|\text{Tr} I^{(p)}G(\Delta M)^2 G^*|\}. \]

Hence, using (2.30) and (2.4), we obtain

\[ \frac{b}{n} \sum_{C^*b < p < n - C^*b} E\{|J_{3p} - E_p\{J_{3p}\}|^2\} \leq Cn^{-1}b^{-1}E\{\text{Tr}(\Delta M)^2 G^*\} \tag{2.44} \]

\[ \leq Cn^{-1}b^{-1}E\{\text{Tr}(\Delta M)^2\} \leq Cb^{-1-\varepsilon/2}. \]

Combining (2.44) with (2.39)-(2.42), we get (2.38).

\[ \square \]
3 Variance

In view of (2.32) to find $S_1$, it suffices to find the main order of $b^{-1}E\{T_p'(z_1, z_2)\}$ defined in (2.31). For this aim it suffices to compute for any $i$ the main order of

$$ t_i = \sum_{j > p} u_{pj} \tilde{E}_{<p} \{ \tilde{G}_{ij}(z_1)G_{ij}(z_2) \}. $$

Consider

$$ s_i := \sum_{j > p} u_{pj} \tilde{E}_{<p} \left\{ \tilde{G}_{ij}(z_1) \sum_k b^{-1/2}v_{ik}G_{kj}(z_2) \right\} $$

$$ = \sum_{j > p} u_{pj} \tilde{E}_{<p} \left\{ \tilde{G}_{ij}(z_1) \sum_k \left( b^{-1/2}v_{ik} - z_2\delta_{ik} + z_2\delta_{ik} \right)G_{kj}(z_2) \right\} $$

$$ = \sum_j u_{pj} \delta_{ij} E\{G_{ii}(z_1)\} + z_2t_i = u_{pi}g(z_1) + z_2t_i + O(b^{-\delta/2}), $$

where we used Lemma 3 for the last equality.

The idea is to compute the l.h.s. above in a way which gives us an equation with respect to $\{t_i\}_{i > p}$. It is possible by using the formula (see e.g. [14]) valid for any random variable $G$ where

$$ \text{cumulant is } b^{-1}\sum_{j > p} u_{pj} \tilde{E}_{<p} \{ \tilde{G}_{ij}(z_1)G_{ij}(z_2) \}. $$

For this aim it suffices to compute for any $\xi$ the main order

$$ \xi = \sum_{i > p} u_{pi} \tilde{E}_{<p} \{ \tilde{G}_{ii}(z_1)G_{ii}(z_2) \}. $$

In view of (2.32) we obtain

$$ \xi = \sum_{i > p} u_{pi} \tilde{E}_{<p} \{ \tilde{G}_{ii}(z_1)G_{ii}(z_2) \} + z_2t_i = u_{pi}g(z_1) + z_2t_i + O(b^{-\delta/2}). $$

Here we used the differentiation formula for the resolvent of any symmetric matrix $M$

$$ \frac{d}{dM_{lk}} G_{st}(z) = -G_{sk}(z)G_{tl}(z) - G_{sl}(z)G_{kt}(z) $$

Two sums written in the r.h.s. of (3.3) collect the terms, corresponding to $s = 1$ in the r.h.s. of (3.4). The remainder $R_1$ collects the terms, corresponding to $s = 2$ in the r.h.s. of (3.4). The remainders $R_2$ and $R_3$ collect the terms, corresponding to $s = 3$ and $s = 4$ respectively. And the remainder $R_4$ appears because of the remainder in (3.2). Let us analyze the order of each of these terms. By (3.4)

$$ R_1 = -b^{-1} \sum_{j > p,k} u_{jp}u_{ik} \tilde{E}_{<p} \{ \tilde{G}_{ij}(z_1)G_{ik}(z_2)G_{kj}(z_2) \} $$

$$ = -b^{-1} \tilde{E}_{<p} \{ (\tilde{G}^{(i,p)} G^{(i,p)} G)_{ij} \} - b^{-1} \tilde{E}_{<p} \{ (\tilde{G}^{(i,p)} G^{(i,p)} G)_{ii} \} = O(b^{-1}). $$

where $I_{ik} = \delta_{ik}u_{ik}1_{k > p}$.

To estimate $R_2$, observe that by (3.4) after two differentiation we obtain the sum of terms of the type $\tilde{G}_{l_1l_2} \tilde{G}_{l_3l_4} \tilde{G}_{l_5l_6} \tilde{G}_{l_7l_8}$, where $\tilde{G}$ can be $G$ or $\tilde{G}$ and the set of indexes $l_1, l_2 \ldots l_7, l_8$ contains $3$ times $i$, $3$ times $k$, and $2$ times $j$, but $\tilde{G}_{jj}$ can not appear. Thus, each term contains either $\tilde{G}_{jk} \tilde{G}_{ji}$ or $\tilde{G}_{jk} \tilde{G}_{jj}$, or $\tilde{G}_{ij} \tilde{G}_{ji}$. Any of this combinations after summation with respect to $j$ gives us $O(1)$. Hence, after summation with respect to $k$ we obtain $O(b)$. But the factor which appears because of the third cumulant is $b^{-3/2}$, hence $R_2 = O(b^{-1/2})$. By the same argument $R_3 = O(b^{-1})$. 

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Finally, to estimate $R_4$, observe that we have two summations with respect to $p < j < p + C_\varepsilon b$ and $i - C_\varepsilon b < k < i + C_\varepsilon b$, and the factor which appears because of $b^{-3}E\{\varepsilon_{ik}\}$ is bounded by $b^{-2-\varepsilon/2}$. At the last step of transformations of (3.3) we write

$$G_{kk}(z_2) = g(z_2) + (G_{kk}(z_2) - g(z_2)), \quad G_{ii}(z_1) = g(z_1) + (G_{ii}(z_1) - g(z_1))$$

and use the bound (3.2). Then we obtain

$$s_i = -g(z_2)t_i - g(z_1)\sum_k U^{(p)}_{ik}t_k + r_i, \quad r_i \leq Cb^{-\varepsilon/2},$$

where

$$U_{ik} = b^{-1}u_{ik}, \quad U^{(p)}_{ik} = b^{-1}u_{ik}1_{i \geq p}1_{k > p}. \quad (3.5)$$

Combining (3.1) and (3.3) with above estimates for the reminders and using that by (1.7) we have $(z_2 + g(z_2)) = -g^{-1}(z_2)$, we obtain the system of equations

$$((\zeta - U^{(p)})t)_i = u_i^{(p)} + r_i, \quad r_i \leq C(b^{-\varepsilon/2} + b^{-\delta/2}),$$

with $\zeta = (g(z_1)g(z_2))^{-1}, \quad u_i^{(p)} = 1_{i > p}u_{pi}$.

Since $|g(z_1)g(z_2)| < 1$ and

$$\|U^{(p)}\| \leq \max_k \sum_i |U_{ki}| \leq 1 + o(1),$$

the operator $(\zeta - U^{(p)})^{-1}$ can be defined by the Neumann series

$$(\zeta - U^{(p)})^{-1} = \sum_{m=0}^{\infty} \zeta^{-m-1}(U^{(p)})^m,$$

and it possesses the properties

$$\sum_k \|(U^{(p)} - \zeta)_{ik}^{-1}\| \leq C, \quad i > p, \quad \|(U^{(p)})_{ii}\| \leq Cb^{-1}, \quad 1 \leq i \leq n. \quad (3.7)$$

Application of $(\zeta - U^{(p)})^{-1}$ to both parts of (3.6) and (3.7) imply

$$t_i = ((\zeta - U^{(p)})^{-1}u_i^{(p)})_i + \tilde{r}_i, \quad |\tilde{r}_i| \leq C(b^{-\varepsilon/2} + b^{-\delta/2}),$$

$$\Rightarrow b^{-1}T''_{p}(z_1, z_2) = b^{-1}((\zeta - U^{(p)})^{-1}u^{(p)}), \quad u^{(p)} + o(1),$$

$$\Rightarrow E\{\Sigma_1\} = \frac{2}{n\zeta} \sum_{p} b^{-1}((\zeta - U^{(p)})^{-1}u^{(p)}), \quad u^{(p)} + o(1),$$

where $\Sigma_1$ was defined in (3.2).

**Proposition 3** Let the matrices $U$ and $U^{(p)}$ be defined by (3.4), where $\{u_{i,j}\}$ satisfy conditions (1.3), the vectors $u^{(p)}$ be defined by (3.4), and $|\zeta| > 1$. Then

$$\frac{1}{\zeta n} \sum_{p=1}^{n} b^{-1}((\zeta - U^{(p)})^{-1}u^{(p)}, u^{(p)}) = -\frac{b}{n}(\text{Tr}\log(1 - \zeta^{-1}U) + \zeta^{-1}\text{Tr}U) + O(b^{-1}). \quad (3.9)$$

**Proof.** Denoting by $S_1(z)$ the l.h.s. of (3.9) and by $S_2(z)$ the main term in the r.h.s. of (3.9), we have

$$S_2(z) = \frac{b}{n} \sum_{m=2}^{\infty} m^{-1}\zeta^{-m} \sum U_{1i_2} \cdots U_{im_1}$$

$$= \frac{b}{n} \sum_{p=1}^{n} \sum_{m=2}^{\infty} m^{-1}\zeta^{-m} \sum_{\min\{i_1, \ldots, i_m\} = p} U_{1i_2} \cdots U_{im_1}$$

$$= \frac{1}{bn} \sum_{p=1}^{n} \sum_{m=2}^{\infty} \zeta^{-m} \sum_{i_2, \ldots, i_m > p} U_{pi_2} \cdots U_{im_p} + O(b^{-1})$$

$$= \frac{1}{bn} \sum_{p=1}^{n} \sum_{m=2}^{\infty} \zeta^{-m}((U^{(p)})^{m-2}u^{(p)}, u^{(p)}) + O(b^{-1}) = S_1(z) + O(b^{-1}).$$
The term $O(b^{-1})$ appears in the third line above as a sum of the terms, which have at least two $p$ among $\{i_1, \ldots, i_m\}$. But the contribution of these terms for fixed $m$ in view of (3.7) can be estimated as
\[
|z|^{-m-1} \sum_{k=1}^{m-1} (U_k)_{pp}(U^{m-k})_{pp} \leq m^2|z|^{-m-1}b^{-2}.
\]
After summation with respect to $m$ and multiplication by $b$ we obtain $O(b^{-1})$.

Now observe that the r.h.s. of (3.9) has a limit, as $n,b \to \infty$ like in (1.4).

\[
S_2(z) = \sum_{m=2}^{\infty} \frac{1}{m\zeta^m} \int u(x_1 - x_2) \cdots u(x_{m-1} - x_m)(u(x_m) - u(x_1))d\bar{c} + r_{n,b}
\]
\[
= -\int \log (1 - \zeta^{-1}u(k)) dk - \zeta^{-1}u(0) + o(1),
\]
where $\hat{u}$ is the Fourier transform of the function $u$ defined as in (1.8). Hence, the proposition and the last line of (3.8) yield
\[
E\{\Sigma_1\} = -2 \left( \int \log (1 - \zeta^{-1}\hat{u}(k)) dk + \zeta^{-1}u(0) \right) + o(1).
\]
Thus by (2.23) and (2.24) we obtain
\[
\frac{b}{n} \text{Var}\{N_n[\varphi]\} \to \int d\lambda_1d\lambda_2\varphi(\lambda_1)\varphi(\lambda_2)C_\eta(\lambda_1, \lambda_2),
\]
where
\[
C_\eta(\lambda_1, \lambda_2) = \frac{1}{4\pi^2} \left( C(\lambda_1 + i\eta, \lambda_2 - i\eta) + C(\lambda_1 - i\eta, \lambda_2 + i\eta) 
- C(\lambda_1 + i\eta, \lambda_2 + i\eta) - C(\lambda_1 - i\eta, \lambda_2 - i\eta) \right).
\]
Now by Proposition 11 for any $\varphi$ possessing the norm (1.8) we have
\[
\frac{b}{n} \text{Var}\{N_n[\varphi]\} \to \lim_{\eta \to 0} \int d\lambda_1d\lambda_2\varphi(\lambda_1)\varphi(\lambda_2)C_\eta(\lambda_1, \lambda_2).
\]
Let us make the change of variables $\lambda_1 = 2\cos x_1$, $\lambda_2 = 2\cos x_2$. Then, using that (see (1.7))
\[
\lim_{\eta \to 0} g(\lambda_\alpha \pm i\eta) = -e^{\mp i\alpha\pi}, \quad \alpha = 1, 2,
\]
we obtain (1.9) by a simple calculus.

4 Auxiliary results

Proof of Lemma 2
The first identity of (2.18) yields that it suffices to estimate $E\{|A_p' A_p^{-1} - E_1\{A_p' A_p^{-1}\}|^2\}$. Note that for any $a$ independent of $\{w_{11}\}$ we have
\[
E_p\{|\xi|^2\} \leq E_p\{|a|^2\}.
\]
Hence it suffices to estimate
\[
\left| A'_p - \frac{E_p[A'_p]}{E_p[A_p]} \right| = \left| A'_p - \frac{A'_p}{A_p} \right| \leq \left| \frac{A'_p}{E_p[A_p]} \right| + \left| \frac{A'_p}{yE_p[A_p]} \right|.
\]
Here and below \( z = x + iy, y > 0 \). Let us use also the relation \( 2.19 \) which yields, in particular, that \( |A'_p/A_p| \leq y^{-1} \). Using \( 2.18 \), we get
\[
E_p\left\{ \left| \frac{A'_p}{E_p[A_p]} \right|^2 \right\} \leq \frac{Cb^{-2} \text{Tr} G^{(p)} I_p G^{(p)*}}{|E_p[A_p]|^2},
\]
Similarly
\[
E_p\left\{ \left| \frac{A'_p}{E_p[A_p]} \right|^2 \right\} \leq \frac{Cb^{-2} \text{Tr} (G^{(p)})^2 I_p (G^{(p)*})^2}{|E_p[A_p]|^2} \leq \frac{Cb^{-2} \text{Tr} G^{(p)} I_p G^{(p)*}}{|E_p[A_p]|^2}.
\]
Thus
\[
\frac{b}{n} E\{ |(\gamma_n(z))^{\circ}|^2 \} \leq Cn^{-1} \sum_p \text{Tr} G^{(p)} I_p G^{(p)*} \leq \frac{b}{ny^2 |E_p[A_p]|^2}.
\]
Notice that the Hölder inequality implies for any \( \delta > 0 \)
\[
\int |b^{-1} \sum_{|j-p| \leq bC} G^{(p)}_{jj}(x + iy)|^{1+\delta} dx \leq Cb^{-1} \sum_{|j-p| \leq bC} \int |G^{(p)}_{jj}(x + iy)|^{1+\delta} dx
\]
\[
\leq b^{-1} \sum_j \left( \sum_k \frac{|(\psi_j, \epsilon_j)|^2}{|(x - \lambda_k)^2 + y^2|^{1+\delta/2}} \right) dx \leq C\delta^{-1} y^{-\delta}.
\]
Hence, denoting \( L_p = \{ x : |\sum u_{pj} G^{(p)}_{jj}(x + iy)| > 1 \} \), we obtain
\[
\int 1_{L_p} dx \leq C \min_\delta \delta^{-1} y^{-\delta} \leq C \log y^{-1}.
\]
Then, using once more that by \( 2.19 \) each summand in the r.h.s. of \( 4.2 \) is bounded by \( y^{-4} \), we get
\[
\int \frac{b}{n} E\{ |(\gamma_n(zx + iy))^{\circ}|^2 \} dx
\]
\[
\leq C n^{-1} \sum_p \left( \int_{R \setminus [-1,1] \cup L_p} (y^2 b)^{-1} \text{Tr} G(x + iy) I_p G^{(p)*}(x + iy) dx + Cy^{-4} \int_{[-1,1] \cup L_p} dx \right)
\]
\[
\leq Cy^3 + Cy^{-4} \log y^{-1} \leq C'y^{-4} \log y^{-1}.
\]
□

**Proof of Lemma**

It follows from \( 2.17 \) that
\[
E\{|G_{pp} - E\{G_{pp}\}|^2 \} \leq |3z|^{-2} E\{|A_p - E\{A_p\}|^2 \}
\]
\[
\leq 2|3z|^{-1} \{E\{|A_p - E\{A_p\}|^2 \} + \text{Var}\{A^2_p\} \}
\]
\[
\leq 2|3z|^{-2} b^{-1} \sum u_{pi} \text{Var}\{G^{(p)}_{ii}\} + C b^{-1}.
\]
But since
\[
E\{|G_{ii} - G_{ii}^{(p)}|\} \leq |3z|^{-2} b^{-1} E\{ |(G^{(p)}_{ii})^{(p)}|\} \leq C |3z|^{-2} b^{-1},
\]
we have
\[
\text{Var}\{C_{ii}^{(p)}\} = \text{Var}\{G_{ii}\} + O(b^{-1}) = \text{Var}\{G_{pp}\} + O(b^{-1}).
\]
Here the last equality is due to the invariance of the distribution of \( M \) with respect to the "shift" \( i \rightarrow (i + 1) \) mod \( (n) \). Hence for any \( z : |3z| \geq 2 \) we obtain from \( 4.3 \)
\[
\text{Var}\{G_{pp}\} \leq 2|3z|^{-2} \text{Var}\{G_{pp}\} + C b^{-1} \text{Var}\{G_{pp}\} \leq 2Cb^{-1}.
\]

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Let us fix any $z = x + i\eta$ with $0 < \eta < 2$ and consider the function

$$\phi(\zeta) = \log(c_0b^{1/2}\text{Cov}\{G_{pp}(\zeta), G_{pp}(z)\})$$

in the half-circle $\Omega = \{\Im \zeta < 2\} \cap \{|\zeta - x - 2i| \leq |2 - \eta/2|\}$. It is a harmonic function, and in view of (4.5) for $\Im \zeta = 2$ we can choose $c_0$ sufficiently small to have

$$c_0b^{1/2}\text{Cov}\{G_{pp}(\zeta), G_{pp}(z)\} \leq c_0b^{1/2}\text{Var}^{1/2}\{G_{pp}(\zeta)\}\text{Var}^{1/2}\{G_{pp}(z)\} \leq 1 \Rightarrow \phi(\zeta) \leq 0, \quad \zeta \in \gamma_1 := \partial \Omega \cap \{\Im \zeta = 2\}.$$

Moreover, in view of the trivial bound $|G_{pp}(\zeta)| \leq |\Im \zeta|^{-1}$, we have

$$\phi(\zeta) \leq \log b^{1/2} + \log c_0\eta^{-2}, \quad \zeta \in \gamma_2 := \partial \Omega \cap \{|\zeta - x - 2i| = |2 - \eta/2|\}.$$

Hence, by the theorem on two constants (see [5], p. 296), we have

$$\phi(\zeta) \leq (\log b^{1/2} + \log c_0\eta^{-2})\omega(\zeta), \quad (4.6)$$

where the harmonic function

$$\omega(\zeta) := \frac{2}{\pi}\Im \log \frac{2 - \eta/2 - (\zeta - x - 2i)}{2 - \eta/2 + (\zeta - x - 2i)},$$

satisfy the conditions

$$\omega(\zeta) = 0, \quad \zeta \in \gamma_1, \quad \omega(\zeta) = 1, \quad \zeta \in \gamma_2.$$

Since $\omega(z) = 1 - 2\delta$ with some $\delta(\eta) > 0$, (4.6) implies the first line of (2.25):

$$c_0b^{1/2}\text{Var}\{G_{pp}(z)\} \leq (c_0b^{1/2})^{1-2\delta} \Rightarrow \text{Var}\{G_{pp}(z)\} \leq Cb^{-\delta}.$$

Using (2.17), (4.3), and (4.6), we get similarly to (4.4),

$$E\{G_{pp}(z)\} = - (z + E\{G_{pp}(z)\})^{-1} + O(b^{-\delta}) \Rightarrow E\{G_{pp}(z)\} = g(z) + O(b^{-\delta}).$$

Thus, we have proved the second line of (2.25).

□

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