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LOCAL CIRCULAR LAW FOR RANDOM MATRICES

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Abstract

The circular law asserts that the spectral measure of eigenvalues of rescaled random matrices without
symmetry assumption converges to the uniform measure on the unit disk. We prove a local version of
this law at any point \( z \) away from the unit circle. More precisely, if \( ||z| - 1| \geq \tau \) for arbitrarily small
\( \tau > 0 \), the circular law is valid around \( z \) up to scale \( N^{-1/2+\varepsilon} \) for any \( \varepsilon > 0 \) under the assumption that
the distributions of the matrix entries satisfy a uniform subexponential decay condition.

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A considerable literature about random matrices focuses on Hermitian or symmetric matrices with independent entries. These models are paradigms for local eigenvalues statistics of many random Hamiltonians, as envisioned by Wigner. The study of non-Hermitian random matrices goes back to Ginibre, then in Princeton and motivated by Wigner. Ginibre’s viewpoint on the problem was described as follows [12]:

Apart from the intrinsic interest of the problem, one may hope that the methods and results will provide further insight in the cases of physical interest or suggest as yet lacking applications.

In fact the eigenvalues statistics found by Ginibre, in the case of Gaussian complex or real entries, correspond to bidimensional gases, with distinct temperatures and symmetry conditions; this is therefore a model for many interacting particle systems in dimension 2 (see e.g. [10] chap. 15). The spectral statistics found in [12] in the complex case are the following: given a \( N \times N \) matrix with independent entries \( \frac{1}{\sqrt{N}} z_{ij} \), the \( z_{ij} \)’s being identically distributed according to the standard complex Gaussian measure \( \mu = \frac{1}{\pi} e^{-|z|^2} dA(z) \) (where \( dA \) denotes the Lebesgue measure on \( \mathbb{C} \)), its eigenvalues \( \mu_1, \ldots, \mu_N \) have a probability density proportional to

\[
\prod_{i<j} |\mu_i - \mu_j|^2 e^{-N \sum_k |\mu_k|^2},
\]

with respect to the Lebesgue measure on \( \mathbb{C}^N \). This law is a determinantal point process (because of the Vandermonde determinant) with an explicit kernel given by (see [12, 16] for a proof)

\[
K_N(z_1, z_2) = \frac{N}{\pi} e^{-\frac{N}{2}(|z_1|^2 + |z_2|^2)} \sum_{\ell=0}^{N-1} \frac{(Nz_1 z_2)^\ell}{\ell!},
\]

with respect to the Lebesgue measure on \( \mathbb{C} \). This integrability property allowed Ginibre to derive the circular law for the eigenvalues, i.e., the empirical spectral distribution converges to the uniform measure on the unit circle,

\[
\frac{1}{\pi} \mathbf{1}_{|z|<1} dA(z).
\]

This phenomenon is the non-Hermitian counterpart of the semicircular law for Wigner random Hermitian matrices, and the quarter circular limit for Marchenko-Pastur random covariance matrices.

In the case of real Gaussian entries, the joint distribution of the eigenvalues is more complicated but still integrable, allowing Edelman [7] to prove the limiting circular law as well; for more precise asymptotic properties of the real Ginibre ensemble, see [4, 11, 21]. We note also that the (right) eigenvalues of the quaternionic Ginibre ensemble were recently shown to converge to a (non-uniform) measure on the unit ball of the quaternions field [3].

For non-Gaussian entries, there is no explicit formula for the eigenvalues. Furthermore, the spectral measure, as a measure on \( \mathbb{C} \), cannot be characterized by computing \( \text{Tr}(M^a M^b) \). Thus the moment method, which is the popular way to prove the semicircle law, cannot be applied to solve this problem. Nevertheless, Girko [13] partially proved that the spectral measure of a non-Hermitian matrix \( M \) with independent entries converges to the circular law (1.1). The key insight of this work was the introduction of the Hermitization technique. This allows him to translate the convergence of complex empirical measures into the convergence of logarithmic transforms for a family of Hermitian matrices. More precisely, if we denote the original non-Hermitian matrix by \( X \) and the eigenvalues of \( X \) by \( \mu_j \), then for any \( \mathcal{C}^2 \) function \( F \) we have the identity

\[
\frac{1}{N} \sum_{j=1}^{N} F(\mu_j) = \frac{1}{4\pi N} \int \Delta F(z) \text{Tr} \log(X^* - z^*)(X - z) dA(z).
\]
From this formula, it is clear that the small eigenvalues of the Hermitian matrix \((X^* - z^*)(X - z)\) play a special role due to the logarithmic singularity at 0. The key question is to estimate the smallest eigenvalues of \((X^* - z^*)(X - z)\), or in other words, the smallest singular values of \((X - z)\). This problem was not treated in [13], but the gap was remedied in a series of papers. First Bai [1] was able to treat the logarithmic singularity assuming bounded density and bounded high moments for the entries of the matrix (see also [2]). Lower bounds on the smallest singular values were given in Rudelson, Vershynin [19, 20], and subsequently Tao, Vu [22], Pan, Zhou [17] and Götze, Tikhomirov [14] weakened the moments and smoothness assumptions for the circular law, till the optimal \(L^2\) assumption, under which the circular law was proved in [23].

The purpose of this paper is to prove a local version of the circular law, up to the optimal scale \(N^{-1/2+\varepsilon}\) (see Section 2 for a precise statement). Below this scale, detailed local statistics will be important and that is beyond the scope of the current paper. The main tool of this paper is a detailed analysis of the self-consistent equations of the Green functions

\[
G_{ij}(w) = [(X^* - z^*)(X - z) - w]^{-1}_{ij}.
\]

Our method is related to the proof of a local semicircular law in [9] or to a local Marchenko-Pastur law in [13]. We are able to control \(G_{ij}(E + i\eta)\) for the energy parameter \(E\) in any compact set and sufficient small \(\eta\). This provides sufficient information to use the formula (1.2) for functions \(F\) at the scales \(N^{-1/2+\varepsilon}\). We also notice that a local Marchenko-Pastur law for \(X^*X\) was proved in [5], simultaneously with the present article.

Finally, we remark that the local circular law demonstrates that the eigenvalue distribution in the unit disk is extremely “uniform”. If the eigenvalues are distributed in the unit disk by a uniform statistics or any other statistics with summable decay of correlations, then there will be big holes or some clusterings of eigenvalues in the disk. While the usual circular law does not rule out these phenomena, the local law established in this paper does. This implies that the eigenvalue statistics cannot be any probability laws with summable decay of correlations

2 The local circular law

We first introduce some notations. Let \(X\) be an \(N \times N\) matrix with independent centered entries of variance \(N^{-1}\). The matrix elements can be either real or complex, but for the sake of simplicity we will consider real entries in this paper. Denote the eigenvalues of \(X\) by \(\mu_j, j = 1, \ldots, N\). We will use the following notion of stochastic domination which simplifies the presentation of the results and their proofs.

**Definition 2.1 (Stochastic domination).** Let \(W = (W_N)_{N \geq 1}\) be family a random variables and \(\Psi = (\Psi_N)_{N \geq 1}\) be deterministic parameters. We say that \(W\) is stochastically dominated by \(\Psi\) if for any \(\sigma > 0\) and \(D > 0\) we have

\[
\Pr\left(|W_N| > N^\sigma \Psi_N\right) \leq N^{-D}
\]

for sufficiently large \(N\). We denote this stochastic domination property by

\[
W \prec \Psi, \quad \text{or} \quad W = O_\prec(\Psi).
\]

In this paper, we will assume that the probability distributions for the matrix elements have the uniform subexponential decay property, i.e.,

\[
\sup_{(i,j) \in [1,N]^2} \Pr\left(|\sqrt{N}X_{i,j}| > \lambda\right) \leq \vartheta^{-1} e^{-\lambda^\vartheta}
\]  

(2.1)
for some constant $\vartheta > 0$ independent of $N$. This condition can of course be weakened to an hypothesis of boundedness on sufficiently high moments, but the error estimates in the following Theorem would be weakened as well. We now state our local circular law, which holds up to the optimal scale $N^{-1/2+\varepsilon}$.

**Theorem 2.2.** Let $X$ be an $N \times N$ matrix with independent centered entries of variance $N^{-1}$. Suppose that the probability distributions of the matrix elements satisfy the uniformly subexponentially decay condition \[\text{[2.1]}\]. We assume that for some fixed $\tau > 0$, for any $N$ we have $\tau \leq |z_0| - 1 \leq \tau^{-1}$ ($z_0$ can depend on $N$). Let $f$ be a smooth non-negative function which may depend on $N$, such that $\|f\|_\infty \leq C$, $\|f'\|_\infty \leq NC$ and $f(z) = 0$ for $|z| \geq C$, for some constant $C$ independent of $N$. Let $f_{z_0}(z) = N^{2a} f(N^a(z - z_0))$ be the approximate delta function obtained from rescaling $f$ to the size order $N^{-a}$ around $z_0$. We denote by $D$ the unit disk. Then for any $a \in (0, 1/2)$,

\[
\left( N^{-1} \sum_j f_{z_0}(\mu_j) - \frac{1}{\pi} \int_D f_{z_0}(z) \, dA(z) \right) \prec N^{-1/2+2a} \|\Delta f\|_{L^1}.
\tag{2.2}
\]

### 3 Hermitization and local Green function estimate

In the following, we will use the notation

\[Y_z = X - zI\]

where $I$ is the identity operator. Let $\lambda_j(z)$ be the $j$-th eigenvalue (in the increasing ordering) of $Y_z^* Y_z$. We will generally omit the $z$-dependence in these notations. Thanks to the Hermitization technique of Girko \[\text{[13]}\], the first step in proving the local circular law is to understand the local statistics of eigenvalues of $Y_z^* Y_z$, for $z$ strictly inside the unit circle. In this section, we first recall some well-known facts about the Stieltjes transform of the empirical measure of eigenvalues of $Y_z^* Y_z$. We then present the key estimate concerning the Green function of $Y_z^* Y_z$ in almost optimal spectral windows. This result will be used later on to prove a local version of the circular law.

#### 3.1 Properties of the limiting density of the Hermitization matrix

Define the Green function of $Y_z^* Y_z$ and its trace by

\[G(w) := G(w, z) = (Y_z^* Y_z - w)^{-1}, \quad m(w) := m(w, z) = \frac{1}{N} \text{Tr} G(w, z) = \frac{1}{N} \sum_{j=1}^N \left. \frac{1}{\lambda_j(z) - w} \right|_{w = E + i\eta}.
\]

We will also need the following version of the Green function later on:

\[G(w) := G(w, z) = (Y_z Y_z^* - w)^{-1}.
\]

As we will see, with high probability $m(w, z)$ converges to $m_c(w, z)$ pointwise, as $N \to \infty$ where $m_c(w, z)$ is the unique solution of

\[m_c^{-1} = -w(1 + m_c) + |z|^2(1 + m_c)^{-1}
\]

with positive imaginary part (see Section 3 in \[\text{[14]}\] for the existence and uniqueness of such a solution). The limit $m_c(w, z)$ is the Stieltjes transform of a density $\rho_c(x, z)$ and we have

\[m_c(w, z) = \int_{\mathbb{R}} \frac{\rho_c(x, z)}{x - w} \, dx.
\]
whenever $\eta > 0$. The function $\rho_c(x, z)$ is the limiting eigenvalue density of the matrix $Y_z^*Y_z$ (cf. Lemmas 4.2 and 4.3 in [1]). Let
\[
\lambda_{\pm} := \lambda_{\pm}(z) := \frac{(\alpha \pm 3)^3}{8(\alpha \pm 1)}, \quad \alpha := \sqrt{1 + 8|z|^2}.
\] (3.2)

Note that $\lambda_-$ has the same sign as $|z| - 1$. The following two propositions summarize the properties of $\rho_c$ and $m_c$ that we will need to understand the main results in this section. They will be proved in Appendix A.

Proposition 3.1. The limiting density $\rho_c$ is compactly supported and the following properties regarding $\rho_c$ hold.

(i) The support of $\rho_c(x, z)$ is $[\max\{0, \lambda_-\}, \lambda_+]$.

(ii) As $x \to \lambda_+$ from below, the behavior of $\rho_c(x, z)$ is given by $\rho_c(x, z) \sim \sqrt{\lambda_+-x}$.

(iii) For any $\varepsilon > 0$, if $\max\{0, \lambda_-\} + \varepsilon \leq x \leq \lambda_+ - \varepsilon$, then $\rho_c(x, z) \sim 1$.

(iv) Near $\max\max\{0, \lambda_-\}$, the behavior of $\rho_c(x, z)$ can be classified as follows.

- If $|z| \geq 1 + \tau$ for some fixed $\tau > 0$, then $\lambda_- > \varepsilon(\tau) > 0$ and $\rho_c(x, z) \sim 1_{x > \lambda_-}(x - \lambda_-)^{1/2}$.
- If $|z| \leq 1 - \tau$ for some fixed $\tau > 0$, then $\lambda_- < -\varepsilon(\tau) < 0$ and $\rho_c(x, z) \sim 1/\sqrt{x}$.

All of the estimates in this proposition are uniform in $|z| < 1 - \tau$, or $\tau^{-1} \geq |z| \geq 1 + \tau$ for fixed $\tau > 0$.

Proposition 3.2. The preceding Proposition implies that, uniformly in $w$ in any compact set,
\[
|m_c(w, z)| = O(|w|^{-1/2})
\]
Moreover, the following estimates on $m_c(w, z)$ hold.

- If $|z| \geq 1 + \tau$ for some fixed $\tau > 0$, then $m_c \sim 1$ for $w$ in any compact set.
- If $|z| \leq 1 - \tau$ for some fixed $\tau > 0$, then $m_c \sim |w|^{-1/2}$ for $w$ in any compact set.

3.2 Concentration estimate of the Green function up to the optimal scale. We now state precisely the estimate regarding the convergence of $m$ to $m_c$. Since the matrix $Y_z^*Y_z$ is symmetric, we will follow the approach of [1]. We will use extensively the following definition of high probability events.

Definition 3.3 (High probability events). Define
\[
\varphi := (\log N)^{\log \log N}.
\] (3.3)

Let $\zeta > 0$. We say that an $N$-dependent event $\Omega$ holds with $\zeta$-high probability if there is some constant $C$ such that
\[
P(\Omega^\zeta) \leq N^C \exp(-\varphi^\zeta)
\]
for large enough $N$.

For $\alpha \geq 0$, define the $z$-dependent set
\[
S(\alpha) := \{w \in \mathbb{C} : \max(\lambda_-/5, 0) \leq E \leq 5\lambda_+, \varphi^\alpha N^{-1}|m_c|^{-1} \leq \eta \leq 10\},
\] (3.4)
where $\varphi$ is defined in (3.3). Here we have suppressed the explicit $z$-dependence. Notice that for $|z| < 1 - \varepsilon$, as $|m_c| \sim |w|^{-1/2}$ we allow $\eta \sim |w| \sim N^{-2}\varphi^{2m}$ in the set $S(\alpha)$. This is a key feature of our approach which shows that the Green function estimates hold until a scale much smaller than the typical $N^{-1}$ value of $\eta$.  

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Theorem 3.4 (Strong local Green function estimates). Suppose \( \tau \leq |z| - 1 \leq \tau^{-1} \) for some \( \tau > 0 \) independent of \( N \). Then for any \( \zeta > 0 \), there exists \( C_\zeta > 0 \) such that the following event holds with \( \zeta \)-high probability:

\[
\bigcap_{w \in S(C_\zeta)} \left\{ |m(w) - m_c(w)| \leq \phi C_\zeta \frac{1}{N\eta} \right\}.
\]

Moreover, the individual matrix elements of the Green function satisfy, with \( \zeta \)-high probability,

\[
\bigcap_{w \in S(C_\zeta)} \left\{ \max_{ij} |G_{ij} - m_c \delta_{ij}| \leq \phi C_\zeta \left( \sqrt{\frac{\text{Im} m_c}{N\eta}} + \frac{1}{N\eta} \right) \right\}.
\]

4 Properties of \( \rho_c \) and \( m_c \)

We have introduced some basic properties of \( \rho_c \) and \( m_c \) in Proposition 3.1 and 3.2. In this section, we collect some more useful properties used in this paper, proved in Appendix A. Recall that \( w = E + i\eta \), \( \alpha = \sqrt{1 + 8|z|^2} \) from (3.2), and define \( \kappa := \kappa(w, z) \) as the distance from \( E \) to \( \{\lambda_+ \pm \} \):

\[
\kappa = \min_{\lambda \in \{\lambda_+ \pm \}} \{|E - \lambda_+ \pm |, |E - \lambda_+ |\}.
\]  

(4.1)

For \( |z| < 1 \), we have \( \lambda_- < 0 \) (see Proposition 3.1), so in this case we define \( \kappa := |E - \lambda_+| \).

Lemma 4.1. There exists \( \tau_0 > 0 \) such that for any \( \tau \leq \tau_0 \) if \( |z| \leq 1 - \tau \) and \( |w| \leq \tau^{-1} \) then the following properties concerning \( m_c \) hold. All constants in the following estimates depend on \( \tau \).

Case 1: \( E \geq \lambda_+ \) and \( |w - \lambda_+| \geq \tau \). We have

\[
|\text{Re} m_c| \sim 1, \quad \frac{1}{2} \leq \text{Re} m_c < 0, \quad \text{Im} m_c \sim \eta.
\]

(4.2)

Case 2: \( |w - \lambda_+| \leq \tau \) (Notice that there is no restriction on whether \( E \leq \lambda_+ \) or not). We have

\[
m_c(w, z) = -\frac{2}{3 + \alpha} + \sqrt{\frac{8(1 + \alpha)^3}{\alpha(3 + \alpha)^5}} (w - \lambda_+)^{1/2} + O(\lambda_+ - w),
\]

(4.3)

and

\[
\text{Im} m_c \sim \begin{cases} \frac{\eta}{\kappa} & \text{if } \kappa \geq \eta \text{ and } E \geq \lambda_+, \\ \eta & \text{if } \kappa \leq \eta \text{ or } E \leq \lambda_+. \end{cases}
\]

(4.4)

Case 3: \( |w| \leq \tau \). We have

\[
m_c(w, z) = \frac{(1 - |z|^2)}{\sqrt{w}} + \frac{1 - 2|z|^2}{2|z|^2 - 2} + O(\sqrt{w})
\]

(4.5)

as \( w \to 0 \), and

\[
\text{Im} m_c(w, z) \sim |w|^{-1/2}.
\]

(4.6)
Case 4: \(|w| \geq \tau, |w - \lambda_+| \geq \tau\) and \(E \leq \lambda_+\). We have

\[
|m_c| \sim 1, \quad \text{Im } m_c \sim 1. \tag{4.7}
\]

Here Case 1 covers the regime where \(E \geq \lambda_+\) and \(w\) is far away from \(\lambda_+\). Case 2 concerns the regime that \(w\) is near \(\lambda_+\), while Case 3 is for \(w\) is near the origin. Finally Case 4 is for \(w\) not covered by the first three cases.

**Lemma 4.2.** There exists \(\tau_0 > 0\) such that for any \(\tau \leq \tau_0\), if \(|z| \geq 1 + \tau\) and \(|w| \leq \tau^{-1}\) then the following properties concerning \(m_c\) hold. All constants in the following estimates depend on \(\tau\). Recall from \(3.2\) that \(\lambda_+ = \frac{(\alpha - 3 \beta)}{8(\alpha - 1)} > 0\).

Case 1: \(E \geq \lambda_+\) and \(|w - \lambda_+| \geq \tau\). We have

\[
|\text{Re } m_c| \sim 1, \quad -\frac{1}{2} \leq \text{Re } m_c < 0, \quad \text{Im } m_c \sim \eta.
\]

Case 2: \(E \leq \lambda_-\) and \(|w - \lambda_-| \geq \tau\). We have

\[
|\text{Re } m_c| \sim 1, \quad 0 \leq \text{Re } m_c, \quad \text{Im } m_c \sim \eta.
\]

Case 3: \(|\kappa + \eta| \leq \tau\). We have

\[
m_c(w, z) = \frac{2}{-3 + \alpha} + \sqrt{\frac{8(\pm \alpha)^3}{\pm \alpha(\pm \alpha)^5}} (w - \lambda_{\pm})^{1/2} + O(\lambda_{\pm} - w),
\]

\[
\text{Im } m_c \sim \begin{cases} 
\sqrt{\kappa} & \text{if } \kappa \geq \eta \text{ and } E \notin [\lambda_-, \lambda_+], \\
\sqrt{\eta} & \text{if } \kappa \leq \eta \text{ or } E \in [\lambda_-, \lambda_+].
\end{cases}
\tag{4.8}
\]

Case 4: \(|w| \geq \tau, |w - \lambda_+| \geq \tau\) and \(\lambda_- \leq E \leq \lambda_+\). We have

\[
|m_c| \sim 1, \quad \text{Im } m_c \sim 1.
\]

Here Case 1 covers the regime \(E \geq \lambda_+\) and \(w\) is far away from \(\lambda_+\). Case 2 concerns the regime \(E \leq \lambda_-\) and \(w\) is far away from \(\lambda_-\). Case 3 is for \(w\) near \(\lambda_\pm\). Finally Case 4 is for \(w\) not covered by the first three cases.

The following lemma concerns the two cases covered in Lemmas \(4.1\) and \(4.2\), i.e., \(z\) is either strictly inside or outside of the unit disk.

**Lemma 4.3.** There exists \(\tau_0 > 0\) such that for any \(\tau \leq \tau_0\) if either the conditions \(|z| \leq 1 - \tau\) and \(|w| \leq \tau^{-1}\) hold or the conditions \(|z| \geq 1 + \tau, |w| \leq \tau^{-1}\), \(\text{Re } \omega \geq \lambda_- / 5\) hold, then we have the following three bounds concerning \(m_c\) (all constants in the following estimates depend on \(\tau\)):

\[
|m_c + 1| \sim |m_c| \sim |w|^{-1/2},
\]

\[
\left|\text{Im } \frac{1}{w(1 + m_c)}\right| \leq C \text{Im } m_c, \tag{4.10}
\]

\[
\left|(-1 + |z|^2) \left( m_c - \frac{-2}{3 + \alpha} \right) \left( m_c - \frac{-2}{3 - \alpha} \right) \right| \geq C \frac{\sqrt{\kappa + \eta}}{|w|}. \tag{4.11}
\]

\[7\]
5 Proof of Theorem 2.2, local circular law in the bulk

Our main tool in this section will be Theorem 3.4 which critically uses the hypothesis \(|z| - 1 \geq \tau|
when \(z\) is on the unit circle the self-consistent equation (which is a fixed point equation for the function
\(g(m) = (1 + w m (1 + m^2) / \|z\|^2 - 1)\) see [6,21] later in this paper) becomes unstable.

We follow Girko’s idea [13] of Hermitization, which can be reformulated as the following identity (see e.g. [15]):
for \(F\) smooth
\[
\frac{1}{N} \sum_{j=1}^{N} F(\mu_j) = \frac{1}{4\pi N} \int \Delta F(z) \left( \sum_{j} \log(z - \mu_j)(z - \bar{\mu}_j) \right) dA(z) = \frac{1}{4\pi N} \int \Delta F(z) \text{Tr} \log Y_z^* Y_z dA(z) \tag{5.1}
\]

We will use the notation \(z = z(\xi) = z_0 + N^{-a} \xi\). Choosing \(F = f_{z_0}\) defined in Theorem 2.2, and changing the variable to \(\xi\), we can rewrite the identity (5.1) as
\[
N^{-1} \sum_{j} f_{z_0}(\mu_j) = \frac{1}{4\pi} N^{-1 + 2a} \int (\Delta f)(\xi) \text{Tr} \log Y_z^* Y_z dA(\xi) = \frac{1}{4\pi} N^{-1 + 2a} \int (\Delta f)(\xi) \sum_{j} \log \lambda_j(z) dA(\xi).
\]

Recall that \(\lambda_j(z)\)'s are the ordered eigenvalues of \(Y_z^* Y_z\), and define \(\gamma_j(z)\) as the classical location of \(\lambda_j(z)\), i.e.
\[
\int_{0}^{\gamma_j(z)} \rho_{c}(x, z) dx = j/N.
\tag{5.2}
\]

Suppose we have
\[
\left| \int \Delta f(\xi) \left( \sum_{j} \log \lambda_j(z(\xi)) - \sum_{j} \log \gamma_j(z(\xi)) \right) dA(\xi) \right| < \|\Delta f\|_{L_1}. \tag{5.3}
\]

Thanks to Proposition 3.1 one can check that uniformly in \(|z| < 1 - \tau\), and also in the domain \(1 + \tau \leq |z| \leq \tau^{-1}\) \((\tau > 0)\), for any \(\delta > 0\) we have
\[
\left| \sum_{j} \log \gamma_j(z) - N \left( \int_{0}^{\infty} (\log x) \rho_{c}(x, z) dx \right) \right| \leq N^{\delta}
\]
for large enough \(N\). We therefore have
\[
N^{-1} \sum_{j} f_{z_0}(\mu_j) = \frac{1}{4\pi} \int \Delta f(\xi) \left( \int_{0}^{\infty} (\log x) \Delta_z \rho_{c}(x, z) dx \right) dA(\xi) + O_{\infty} \|\Delta f\|_{L_1} \tag{5.4}
\]
where we have used that
\[
\frac{1}{4\pi} N^{2a} \int \Delta f(\xi) \int_{0}^{\infty} (\log x) \rho_{c}(x, z) dx dA(\xi) = \frac{1}{4\pi} \int \Delta f(\xi) \left( \int_{0}^{\infty} (\log x) \Delta_z \rho_{c}(x, z) dx \right) dA(\xi).
\]

It is known, by Lemma 4.4 of [11], that
\[
\int_{0}^{\infty} (\log x) \Delta_z \rho_{c}(x, z) dx = 4 \chi_D(z). \tag{5.5}
\]

Combining (5.4) and (5.5), we have proved 2.2 provided that we can prove 3.3. To prove 3.3, we need the following rigidity estimate which is a consequence of Theorem 3.4.
Lemma 5.1. Suppose $\tau \leq |z| - 1 \leq \tau^{-1}$ for some $\tau > 0$ independent of $N$. Then for any $\zeta > 0$, there exists $C_{\zeta} > 0$ such that the following event holds with $\zeta$-high probability: for any $\varphi^{C_{\zeta}} < j < N - \varphi^{C_{\zeta}}$ we have

$$
\gamma_{j-\varphi^{C_{\zeta}}} \leq \gamma_j \leq \gamma_{j+\varphi^{C_{\zeta}}}.
$$

and in the case $|z| \leq 1 - \tau$,

$$
\frac{|\lambda_j - \gamma_j|}{\gamma_j} \leq \frac{C\varphi^{C_{\zeta}}}{j(1 - \frac{1}{N})^{1/3}},
$$

in the case $|z| \geq 1 + \tau$,

$$
\frac{|\lambda_j - \gamma_j|}{\gamma_j} \leq \frac{C\varphi^{C_{\zeta}}}{(\min\{\frac{1}{N}, 1 - \frac{1}{N}\})^{1/3}}.
$$

Proof. First, with \(8.4\) and the definition \(8.3\), for any $\zeta$ holds with $\zeta$-high probability. It also implies that for $\eta = \varphi^{C_{\zeta}} N^{-1} |m_c|^{-1}$,

$$
\eta \Im m(E + i\eta) \leq C\varphi^{2C_{\zeta}} N^{-1}.
$$

Then using the fact that $\eta \Im m(E + i\eta)$ and $\eta \Im m_c(E + i\eta)$ are increasing with $\eta$, we obtain that holds for any $0 \leq \eta \leq O(\varphi^{C_{\zeta}} N^{-1} |m_c|^{-1})$ with $\zeta$-high probability. Notice that $\Im m$ and $\Im m_c$ are positive number. Define the interval

$$
I_E = [E_1, E_2] = [\gamma_j, 4\lambda_+]
$$

and define $\eta_j \geq 0$ as the smallest positive solution of

$$
\eta_j = 2\varphi^{C_{\zeta}} |m_c(E_j + i\eta_j)|^{-1} N^{-1}, \quad j = 1, 2.
$$

Since

$$
\#\{j : E - \eta \leq \lambda_j \leq E + \eta\} \leq C N \eta \Im m(E + i\eta),
$$

we have by \(8.4\) that

$$
\#\{j : E_1 - \eta_1 \leq \lambda_j \leq E_1 + \eta_1\} + \#\{j : E_2 - \eta_2 \leq \lambda_j \leq E_2 + \eta_2\} \leq C\varphi^{2C_{\zeta}}.
$$

Using the Helffer-Sjöstrand functional calculus (see e.g. \(8.4\)), letting $\chi(\eta)$ be a smooth cutoff function with support in $[-1, 1]$, with $\chi(\eta) = 1$ for $|\eta| \leq 1/2$ and with bounded derivatives, we have for any $q : \mathbb{R} \to \mathbb{R}$,

$$
q(\lambda) = \frac{1}{2\pi} \int_{\mathbb{R}^2} \frac{iy q''(x) \chi(y) + i(q(x) + iyq'(x))\chi'(y)}{\lambda - x - iy} \, dx \, dy.
$$

To prove \(8.6\), we choose $q$ to be supported in $[E_1, E_2]$ such that $q(x) = 1$ if $x \in [E_1 + \eta_1, E_2 - \eta_2]$ and $|q'| \leq C(\eta_1)^{-1}$, $|q''| \leq C(\eta_1)^{-2}$ if $|x - E_1| \leq \eta_1$. We now claim that

$$
\left| \int q(\lambda) \Delta \rho(\lambda) \, d\lambda \right| \leq C\varphi^{2C_{\zeta}} N^{-1}, \text{ where } \Delta \rho = \rho - \rho_c, \quad \rho = \frac{1}{N} \sum_j \delta_{\lambda_j(z)}.
$$
Combining (5.12) and (5.14), we have for any $1 \leq j \leq N$,
\[
\#\{k : \lambda_k \geq \gamma_j\} - (N - j) = O(\varphi^{2\zeta})
\]
which implies (5.6) with $C\zeta$ in (5.6) replaced by $2C\zeta$.

It remains to prove (5.12). Since $q$ and $\chi$ are real, with $\Delta m = m - m_c$

\[
\left| \int q(\lambda)\Delta p(\lambda)d\lambda \right| \leq C \int_{\mathbb{R}^2} (|q(E)| + |\eta||q'(E)|)|\chi'(\eta)||\Delta m(E + i\eta)|dEd\eta
\]

\[
+ C \sum_{i} \left| \int_{|\eta| \leq \eta_i} \int_{E - E_i \leq \eta} \eta q''(E)\chi(\eta) \text{Im} \Delta m(E + i\eta)dEd\eta \right|
\]

\[
+ C \sum_{i} \left| \int_{|\eta| \geq \eta_i} \int_{E - E_i \leq \eta} \eta q''(E)\chi(\eta) \text{Im} \Delta m(E + i\eta)dEd\eta \right|
\]

(5.13)
The first term is estimated by

\[
\int_{\mathbb{R}^2} (|q(E)| + |\eta||q'(E)|)|\chi'(\eta)||\Delta m(E + i\eta)|dEd\eta \leq CN^{-1}\varphi^{C\zeta},
\]

using (5.14) and that on the support of $\chi'$ is in $1 \geq |\eta| \geq 1/2$.

For the second term in the r.h.s. of (5.13), with $|q''| \leq C\eta_i^{-2}$, (5.9) and (5.10), we obtain

second term in r.h.s. of (5.13) $\leq CN^{-1}\varphi^{C\zeta}$.

(5.15)

We now integrate the third term in (5.13) by parts first in $E$, then in $\eta$ (and use the Cauchy-Riemann equation $\frac{\partial}{\partial E} \text{Im}(\Delta m) = -\frac{\partial}{\partial \eta} \text{Re}(\Delta m)$) so that

\[
\int \eta q''(E)\chi(\eta) \text{Im} \Delta m(E + i\eta)dEd\eta = -\int_{E - E_i \leq \eta_i} \eta_i \chi(\eta)q'(E) \text{Re} \Delta m(E + i\eta)dE
\]

\[
- \int (\eta\chi'(\eta) + \chi(\eta))q'(E) \text{Re} \Delta m(E + i\eta)dEd\eta
\]

We therefore can bound the third term in (5.13) with absolute value by

\[
C \sum_{i} \int_{|E - E_i| \leq \eta_i} \eta_i |q'(E)||\text{Re} \Delta m(E + i\eta_i)|dE
\]

\[
+ C \sum_{i} \eta_i^{-1} \int_{\eta_i \leq \eta \leq 1} \int_{E - E_i \leq \eta} |\text{Re} \Delta m(E + i\eta)|dEd\eta + \int_{\mathbb{R}^2} |\eta||q'(E)||\chi'(\eta)||\Delta m(E + i\eta)|dEd\eta
\]

(5.16)

where the last term can be bounded as the first term in r.h.s. of (5.13). By using (5.9) we have

\[
(5.10) \leq CN^{-1}\varphi^{C\zeta} + CN^{-1}\varphi^{C\zeta} \sum_{i} \eta_i^{-1} \int_{|E - E_i| \leq \eta_i} dE \int_{\eta_i \leq \eta \leq 1} \frac{1}{\eta N} d\eta \leq CN^{-1}\varphi^{C\zeta + 1}
\]

where we used $\eta_i \geq N^{-C}$. Together with (5.14) and (5.15), we obtain (5.12) and complete the proof of (5.6).
Now we prove (5.7). Using (5.2) and Proposition 3.1, we have

$$\gamma_j = O(j^2 N^{-2})$$, \( j \leq N/2 \); \hspace{1cm} \gamma_j = \lambda_+ - O \left( \frac{N-j}{N} \right)^{2/3}, \hspace{1cm} j \geq N/2. \tag{5.17}$$

One can check easily that

$$\gamma_j - \gamma_{j-1} = O \left( j N^{-1/3} \right) \tag{5.18}$$

and for \( j \geq 2 \)

$$\frac{|\gamma_j - \gamma_{j-1}|}{\gamma_j} \leq C j^{-1/3} (N - j)^{-1/3} \leq \frac{C \varphi^{C \zeta}}{j (1 - \frac{j}{N})^{1/3}}. \tag{5.18}$$

Combining (5.18) with (5.6), we obtain (5.7).

For (5.8), the proof is similar to the above reasoning, but simpler: in this case \( \gamma_j \sim 1 \) for \( j \leq N/2 \). For \( j \geq N/2 \), \( \gamma_j \) is bounded as (5.17), and one can check if

$$1 + \tau \leq |z| \leq \tau - 1,$$

then

$$\gamma_j - \gamma_{j-1} = O \left( \min \left\{ \frac{j}{N}, 1 - \frac{j}{N} \right\} \right)^{-1/3} N^{-1},$$

which implies (5.8).

We return to the proof of the local circular law, Theorem 2.2. We now only need to prove (5.8) from Lemma 5.1. From (5.7) and (5.8), we have

$$|\log \lambda_j(z) - \log \gamma_j(z)| \leq C \frac{\lambda_j - \gamma_j}{\gamma_j} \leq \frac{C \varphi^{C \zeta}}{j (1 - \frac{j}{N})^{1/3}} \leq \frac{C \varphi^{C \zeta}}{j (1 - \frac{j}{N})^{1/3}} \leq C \varphi^{C \zeta} \left( \min \left\{ \frac{j}{N}, 1 - \frac{j}{N} \right\} \right)^{-1/3} N^{-1},$$

Notice that, for large enough \( C \), there is a constant \( c > 0 \) such that for any \( j \) we have

$$\lambda_j \leq N^C$$

with probability larger than \( 1 - \exp(-N^C) \) (for this elementary fact, one can for example see that the entries of \( X \) are smaller than 1 with probability greater than \( 1 - \vartheta^{-1} e^{-N^2} \) by the subexponential decay assumption (2.1) and then use \( \sum \lambda_j = \text{Tr} Y^* Y \)), so together with the above bounds on \( |\log \lambda_j(z) - \log \gamma_j(z)| \) this proves that for any \( \zeta > 0 \), there exists \( C_\zeta > 0 \) such that

$$\sum_{j > e^{C \zeta}} (|\log \lambda_j(z) - \log \gamma_j(z)|) \leq \varphi^{2C \zeta} \tag{5.19}$$

with \( \zeta \)-high probability. Furthermore, one can see that the estimates hold uniformly for \( z \)'s in this region.

On the other hand, the following important Lemma 5.2 holds, concerning the smallest eigenvalue. It implies that

$$\sum_{j \leq e^{C \zeta}} |\log \lambda_j(z)| \sim 1$$
holds uniformly for $z$ in any fixed compact set. It is easy to check that for any $\delta > 0$, for large enough $N$,
\[
\sum_{j \in \varphi_{C\zeta}} |\log \gamma_j(z)| \leq N^\delta.
\]
Hence we can extend the summation in (5.19) to all $j \geq 1$, which gives (5.3) and completes the proof of Theorem 2.2.

Lemma 5.2 (Lower bound on the smallest eigenvalue). Under the same assumptions of Theorem 2.2,
\[
|\log \lambda_1(z)| \prec 1
\]
holds uniformly for $z$ in any fixed compact set.

Proof. This lemma follows\(^1\) from [20] or Theorem 2.1 of [22], which gives the required estimate uniformly in $z$. Note that the typical size of $\lambda_1$ is $N^{-2}$ [20], and we need a much weaker bound of type $\mathbb{P}(\lambda_1(z) \leq e^{-N^{-\varepsilon}}) \leq N^{-C}$ for any $\varepsilon, C > 0$. This estimate is very simple to prove if, for example, the entries of $X$ have a density bounded by $N C$. Then, from the variational characterization $\lambda_1(z) = \min_{|u|=1} \|X(z)u\|^2$, one easily gets
\[
\lambda_1(z)^{1/2} \geq N^{-1/2} \min_{k \in [1,N]} \text{dist}(X(z)e_k, \text{span}\{X(z)e_\ell, \ell \neq k\}) = N^{-1/2} \min_{k \in [1,N]} \|\langle X(z)e_k, u_k(z) \rangle\|,
\]
where $u_k(z)$ is a unit vector independent of $X(z)e_k$. By conditioning on $u_k(z)$, the result of this lemma is straightforward since the matrix entries have a density. \(\square\)

6 Weak local Green function estimate

In this section, we make a first step towards Theorem 3.4, with a weaker version of it, stated hereafter.

Theorem 6.1 (Weak local Green function estimates). Under the assumption of Theorem 3.4, the following event hold with $\zeta$-high probability (see (3.4) for the definition of $S$):
\[
\bigcap_{w \in S(b)} \left\{ \max_{ij} |G_{ij}(w) - m_c(w)\delta_{ij}| \leq \varphi_{C\zeta} \frac{1}{|w|^{1/2}} \left( \frac{|w|^{1/2}}{N\eta} \right)^{1/4}, \quad b > 5C\zeta \right\}, \tag{6.1}
\]

This theorem will be proved in the subsequent subsections.

6.1 Identities for Green functions and their minors. There are many different ways to form minors for the matrices $Y^*Y$ and $YY^*$. We will use the following definition (where we use the notation $[a,b] = [a,b] \cap \mathbb{Z}$).

Definition 6.2. Let $T, U \subset [1,N]$. Then we define $Y^{(T,U)}$ as the $(N - |U|) \times (N - |T|)$ matrix obtained by removing all columns of $Y$ indexed by $i \in T$ and all rows of $Y$ indexed by $i \in U$. Notice that we keep the labels of indices of $Y$ when defining $Y^{(T,U)}$.\(^2\)

\(^1\)Strictly speaking, this bound was proved for identically distributed entries, but the proof extends to the case of distinct distributions, provided that, for example, a uniform subexponential decay holds.

\(^2\)Stated in a way that is different from the one used in the text.
Let $y_i$ be the $i$-th column of $Y$ and $y_i^{(S)}$ be the vector obtained by removing $y_i(j)$ for all $j \in \mathbb{S}$. Similarly we define $y_i$ be the $i$-th row of $Y$. Define

$G_{(T,U)}^{(T,U)} = \left( (Y^{(T,U)})^* Y^{(T,U)} - w \right)^{-1}$, \hspace{1em} $m_G^{(T,U)} = \frac{1}{N} \text{Tr} G_{(T,U)}^{(T,U)}$,

$G_{(T,U)}^\ast = \left( Y^{(T,U)} Y^{(T,U)} \ast - w \right)^{-1}$, \hspace{1em} $m_G^\ast = \frac{1}{N} \text{Tr} G_{(T,U)}^\ast$.

By definition, $m^{(0,0)} = m$. Since the eigenvalues of $Y^* Y$ and $YY^*$ are the same except the zero eigenvalue, it is easy to check that

$$m_{G}^{(T,U)}(w) = m_{G}^\ast + \frac{|U| - |T|}{Nw} \tag{6.2}$$

For $|U| = |T|$, we define

$$m_{G}^{(T,U)} := m_{G}^{(T,U)} = m_{G}^\ast \tag{6.3}$$

By definition, $G_{(T,U)}$ is a $(N - |T|) \times (N - |T|)$ matrix and $G_{(T,U)}^\ast$ is a $(N - |U|) \times (N - |U|)$ matrix. For $i$ or $j \in \mathbb{T}$, $G_{ij}$ has no meaning from the previous definition. But we define $G_{ij} = 0$ whenever either $i$ or $j \in U$. Similar convention applies to $G_{ij}^\ast$, which is zero if $i$ or $j \in U$.

Notice that we can view $G_{ij} = (W_z^*)^* W_z$ where $W_z = Y_z^*$, so all properties of $G_{(T,U)}$ have parallel versions for $G_{(U,T)}^\ast$. We shall call this property row-column reflection symmetry, i.e., we interchange $G_{(U,T)}, Y, z, y, i$ by $G_{(T,U)}^\ast, Y^*, z^*, y^i$. Here $y_i$ is a $N \times 1$ column vector and $y_i$ a $1 \times N$ row vector. The following lemma provides the formulas relating Green functions and their minors.

**Lemma 6.3** (Relation between $G$, $G_{(T,U)}$ and $G_{(U,T)}^\ast$). For $i, j \neq k$ ($i = j$ is allowed) we have

$$G_{ij}^{(k,0)} = G_{ij} - \frac{G_{ik} G_{kj}}{G_{kk}}, \hspace{1em} G_{ij}^{(0,k)} = G_{ij} - \frac{G_{ik} G_{kj}}{G_{kk}}, \tag{6.4}$$

$$G_{ij}^{(0,i)} = G + \frac{(G_{ij}^*) (y_i G)}{1 - y_i G_{ii}}, \hspace{1em} G_{ij}^{(0,i)} = G + \frac{(G_{ij}^*) (y_i G)}{1 + y_i G_{ii}} \tag{6.5}$$

and

$$G_{ij}^{(i,0)} = G + \frac{(G_{ij}^*) (y_i^* G)}{1 - y_i^* G_{ii}}, \hspace{1em} G_{ij}^{(i,0)} = G + \frac{(G_{ij}^*) (y_i^* G)}{1 + y_i^* G_{ii}} \tag{6.6}$$

Furthermore, the following crude bound on the difference between $m$ and $m_{G}^{(U,T)}$ holds: for $U, T \subset [1, N]$ we have

$$|m - m_{G}^{(U,T)}| + |m - m_{G}^{(U,T)}| \leq \frac{|U| + |T|}{N \eta}. \tag{6.7}$$

**Proof.** By the row-column reflection symmetry, we only need to prove these formulas involving $G$. We first prove (6.4). In [8–9], was proved a lemma concerning Green functions of matrices and their minors. This lemma is stated as Lemma [14.2] in Appendix B. Let $H := Y^* Y$.

For $T \subset [1, N]$, denote $H_T^{(T)}$ as the $N - |T|$ by $N - |T|$ minor of $H$ after removing the $i$-th rows and columns index by $i \in T$. Following the convention in Definition [14.1], we define

$$G_T^{(T)} = (H_T^{(T)} - wI)^{-1}. \tag{6.8}$$
By definition, we have
\[ G^T = G^{(T,0)}. \] (6.9)

Then we can apply (B.4) to \( G^{(T,0)} \) and obtain (6.4).

We now prove (6.5). Recall the rank one perturbation formula
\[ (A + \mathbf{v}^*\mathbf{v})^{-1} = A^{-1} - \frac{(A^{-1}\mathbf{v}^*)(\mathbf{v}A^{-1})}{1 + \mathbf{v}A^{-1}\mathbf{v}^*} \]
where \( \mathbf{v} \) is a row vector and \( \mathbf{v}^* \) is its Hermitian conjugate. Together with
\[ G^{-1} = Y^*Y - wI = \sum_j y_j^*y_j - wI = (G^{(0,i)})^{-1} + y_i^*y_i \]
we obtain (6.5).

We now prove (6.6). With (6.4), we have
\[ m^{(i,\emptyset)}_G - m = -\frac{1}{N} \sum_j G_{ji}G_{ij} \]
Moreover, by diagonalization in an orthonormal basis and the obvious identity \( |(\lambda - \omega)^{-2}| = \eta^{-1} \text{Im}[(\lambda - \omega)^{-1}] \) \((\lambda \in \mathbb{R})\), we have
\[ \left| \sum_j G_{ji}G_{ij} \right| = |(G^2)_{ii}| = \frac{\text{Im}G_{ii}}{\eta}, \]
so we have proved that
\[ |m - m^{(i,\emptyset)}_G| \leq \frac{1}{N\eta}. \] (6.10)

By (6.3), (6.10) holds for \( m^{(i,\emptyset)}_G \) as well. Similar arguments can be used to prove (6.6) for \( m^{(i,i)}_G \), \( m^{(i,i)}_G \) and the general cases. This completes the proof of Lemma 6.3.

The next step is to derive equations between the matrix and its minors. The main results are stated as the following Lemma 6.5. We first need the following definition.

**Definition 6.4.** In the following, \( \mathbb{E}_X \) means the integration with respect to the random variable \( X \). For any \( \mathcal{T} \subset [1,N] \), we introduce the notations
\[ Z_i^{(T)} := (1 - \mathbb{E}_{y_i})y_i^{(T)}G^{(T,i)}y_i^{(T)*} \]
and
\[ Z_i^{(T)} := (1 - \mathbb{E}_{y_i})y_i^{(T)*}G^{(i,T)}y_i^{(T)}. \]
Recall by our convention that \( y_i \) is a \( N \times 1 \) column vector and \( y_i \) is a \( 1 \times N \) row vector. For simplicity we will write
\[ Z_i = Z_i^{(\emptyset)}, \quad \bar{Z}_i = Z_i^{(\emptyset)}. \]
Lemma 6.5 (Identities for $G, \mathcal{G}, Z$ and $\mathcal{Z}$). For any $T \subset [1, N]$, we have

$$G^{(0,T)}_{ii} = -w^{-1} \left[ 1 + m^{(i,T)}_G + |z|^2 G^{(i,T)}_i + Z^{(T)}_i \right]^{-1}, \quad (6.11)$$

$$G^{(0,T)}_{ij} = -w G^{(0,T)}_{ii} G^{(i,T)}_{jj} \left( y^{(T)}_i G^{(i,j,T)} y^{(T)}_j \right), \quad i \neq j, \quad (6.12)$$

where, by definition, $G^{(i,T)}_{ii} = 0$ if $i \in T$. Similar results hold for $\mathcal{G}$:

$$\left[G^{(T,0)}_{ii}\right]^{-1} = -w \left[ 1 + m^{(T,i)}_G + |z|^2 G^{(T,i)}_i + Z^{(T)}_i \right], \quad (6.13)$$

$$G^{(T,0)}_{ij} = -w G^{(T,0)}_{ii} G^{(T,i)}_{jj} \left( y^{(T)}_i G^{(T,i,j)} y^{(T)}_j \right), \quad i \neq j. \quad (6.14)$$

Proof. By the row-column reflection symmetry, we only need to prove the $G$ part of this lemma. Furthermore, for simplicity, we prove the case $T = \emptyset$, the general case can be proved in the same way.

We first prove (6.11). Let $H = Y^* Y$. Similarly to (6.7) and (6.8), we define $G^{[i]}$ and $H^{[i]}$. Then using (B.2) and (6.9), we have

$$[G_{ii}]^{-1} = h_{ii} - w - \sum_{k,l \neq i} h_{ik} G^{(i,0)}_{kl} h_{li}. \quad (6.15)$$

From the definition of $H$, we have $h_{ik} = y^*_i y_k$. Then

$$[G_{ii}]^{-1} = y_i^* y_i - w - y_i^* Y^{(i,0)} G^{(i,0)} (Y^{(i,0)})^* y_i. \quad (6.16)$$

For any matrix $A$, we have the identity

$$A (A^* A - w)^{-1} A^* = 1 + w (AA^* - w)^{-1}, \quad (6.17)$$

and as a consequence

$$Y^{(i,0)} G^{(i,0)} (Y^{(i,0)})^* = 1 + w G^{(i,0)}. \quad (6.18)$$

Combining (6.15) and (6.17), we have

$$[G_{ii}]^{-1} = -w - w y_i^* G^{(i,0)} y_i. \quad (6.19)$$

We now write

$$y_i^* G^{(i,0)} y_i = E_{y_i} y_i^* G^{(i,0)} y_i + Z_i$$

By definition

$$E_{y_i} y_i^* G^{(i,0)} y_i = \frac{1}{N} \text{Tr} G^{(i,0)} + |z|^2 G^{(i,0)}_i = m^{(i,0)}_G + |z|^2 G^{(i,0)}_i$$

which complete the proof of (6.11).

We now prove (6.12). As above, using now (B.3), we have

$$G^{(0,T)}_{ij} = G^{(0,T)}_{ii} G^{(i,T)}_{jj} \left( h_{ij} - \sum_{kl \neq ij} h_{ik} G^{(i,j,0)}_{kl} h_{lj} \right)$$

where

$$h_{ij} - \sum_{kl \neq ij} h_{ik} G^{(i,j,0)}_{kl} h_{lj} = y_i^* y_j - y_i^* Y^{(i,0)} G^{(i,0)} (Y^{(i,0)})^* y_j. \quad (6.20)$$

Then using (6.16) again, we obtain (6.12). \qed
6.2 The self-consistent equation and its stability. We now derive the self-consistent equation for \( m(w) \) and its stability estimates. Following [3, we introduce the following control parameter:

**Definition 6.6. Define the control parameter**

\[
\Psi = \left( \frac{\text{Im} m_c + \Lambda}{N\eta} + \frac{1}{N\eta} \right), \quad \Lambda = |m - m_c|
\]

Notice that all quantities depend on \( w \) and \( z \). Furthermore, if \( \Lambda \leq C|m_c| \) then for \( w \in S(\beta) \) (see (5.4)),

\[
|m_c|^{-1}\Psi \leq \frac{1}{\sqrt{N\eta|m_c|}} + \frac{1}{N\eta|m_c|} \leq C\varphi^{-b/2}.
\]  

(6.19)

The quantity \( |m_c|^{-1}\Psi \) will be our controlling small parameter in this paper.

Before we start to prove Theorem 8.3, we make the following observation. The parameter \( z \) can be either inside the unit ball or outside of it. Recall the properties of \( m_c \) in section 3. By Lemma 3.1, the limiting density \( \rho_c \) is supported on \( [\lambda_-, \lambda_+] \), where \( \lambda_- < 0 \) and \( \lambda_+ \sim 1 \) when \( |z| \leq 1 - \tau \). Since \( \lambda_- < 0 \) in this case, we will never approach \( \lambda_- \). On the other hand, we will have to consider the behavior when \( w \sim 0 \). When \( 1 + \tau \leq |z| \leq \tau^{-1} \), we have \( \lambda_+ > 0 \) and \( w \) stays away from the origin by definition of \( S(C) \), i.e., the condition \( E \geq \lambda_-/5 \). Our approach to the local Green function estimates will use the self-consistent equation of \( m(w) \). This approach depends crucially on the stability properties of this equation which can be divided roughly into three cases: \( w \) near the edges \( \lambda_\pm \), \( w \sim 0 \) or \( w \) in the bulk (defined here as the rest of possible \( w \in S(C) \)). From Lemma 4.1 and Lemma 4.2, the behavior of \( m_c \) near the edges \( \lambda_\pm \) when \( |z| \geq 1 + \tau \) are identical to its behavior near the edge \( \lambda_+ \) when \( |z| \leq 1 - \tau \). In the bulk, the behavior for both cases are the same. Thus we will only consider the case \( |z| \leq 1 - \tau \) since it covers all three different behaviors. Hence from now on, we will assume that \( |z| \leq 1 - \tau \). We emphasize that \( \text{Im} m_c \ll |m_c| \) when \( |\lambda_+ - w| \ll 1 \). All stability results concerning the self-consistent equation will be under the following assumption (6.20).

**Lemma 6.7** (Self consistent equation). Suppose \( |z| \leq 1 - \tau \) for some \( \tau > 0 \). Then there exists a small constant \( \alpha > 0 \) independent of \( N \) such that if the estimate

\[
\Lambda \leq \alpha|m_c|
\]  

(6.20)

holds for some \( |w| \leq C \) on a set \( A \) in the probability space of matrix elements for \( X \), then in the set \( A \) we have with \( \zeta \)-high probability

\[
w m(1 + m)^2 - m|z|^2 + 1 + m = Y, \quad Y = O(\varphi^{Q_\zeta} \Psi),
\]  

(6.21)

provided that \( w \in S(\beta) \) for some \( b > 5Q_\zeta \) with \( Q_\zeta \) defined in Lemma 6.7.

**Proof.** By (4.3), (4.10) and (6.20), for \( |z| \leq 1 - t \) the following inequalities hold on the set \( A \):

\[
|w|^{-1}\frac{1}{1 + m^2} \leq |w|^{-1}\frac{1}{|1 + m_c + O(\Lambda)|^2} \leq C,
\]  

(6.22)

\[
|\text{Im} \frac{1}{w(1 + m)}| \leq |\text{Im} \frac{1}{w(1 + m_c)}| + \frac{1}{w(1 + m_c)(m - m_c)} \frac{1}{(1 + m)} \leq \text{Im} m_c + CA.
\]  

(6.23)
Furthermore, using (6.22), (4.9), (6.10), (6.20) and (6.1), we have in the set $A$

$$1 + m - \frac{|z|^2}{w(1 + m)} = 1 + m_c - \frac{|z|^2}{w(1 + m_c)} + O(\Lambda) = \frac{1}{wm_c} + O(\Lambda).$$

(6.24)

The origin of the self-consistent equation (6.21) relies on the choice $T = \{i\}$ in (6.13):

$$\left[ G_{ii}(i,\emptyset) \right]^{-1} = -w \left[ 1 + m_G(i,i) + Z^{(i)}_i \right].$$

(6.25)

By definition of $\Psi$ and (6.3),

$$|m_G(i,i) - m| \leq C \frac{\zeta}{N\eta} \leq C\Psi.$$

(6.26)

Moreover, we have from (6.19) that with $\zeta$-high probability in $A$

$$|Z^{(i)}_i| \leq \varphi Q_{\zeta}/2 \sqrt{\text{Im} m_G(i,i) + |z|^2 \text{Im} G_{ii}(i,i)} \leq \varphi Q_{\zeta}/2 \Psi$$

(6.27)

where we have used (6.20), (6.20) and, by definition, $G_{ii}(i,\emptyset) = 0$. We would like to estimate $(G_{ii}(i,\emptyset))^{-1}$ in (6.25) by treating $(1 + m)$ as the main term and the rest as error terms. From the equations (6.20) and (6.19), the ratio between the error terms and the main term for $w \in S(b)$ with $b > 5Q_{\zeta}$ is bounded by

$$|m|^{-1} |Z^{(i)}_i| + |m|^{-1} |m_G(i,i) - m| \leq \varphi^{-Q_{\zeta}}.$$

(6.28)

Therefore for any $w \in S(b)$ with $b > 5Q_{\zeta}$ we have with $\zeta$-high probability

$$G_{ii}(i,\emptyset) = -\frac{1}{w(1 + m)} + \mathcal{E}_1$$

(6.29)

where

$$\mathcal{E}_1 = w^{-1} \frac{1}{(1 + m)^2} \left[ m_G(i,i) - m + Z^{(i)}_i \right] + O \left( \frac{|Z^{(i)}_i|^2}{|w||1 + m|^3} \right) = O(\varphi Q_{\zeta}/2 \Psi).$$

(6.30)

where we have used (6.22) and $|m_c| \sim |w|^{-1/2}$. Together with (6.23), we thus have with $\zeta$-high probability

$$|\text{Im} G_{ii}(i,\emptyset)| \leq \left| \text{Im} \frac{1}{w(1 + m)} \right| + O(\varphi Q_{\zeta}/2 \Psi) \leq |\text{Im} m_c + CA + O(\varphi Q_{\zeta}/2 \Psi).$$

(6.31)

Using this estimate, (6.6) and (6.29), we can estimate $Z_i := Z^{(i)}_i$ by

$$|Z_i| \leq \varphi Q_{\zeta}/2 \sqrt{\text{Im} m_G(i,\emptyset) + |z|^2 \text{Im} G_{ii}(i,\emptyset)} \leq \varphi Q_{\zeta}/2 \sqrt{\frac{\text{Im} m + \text{Im} m_c + A + \varphi Q_{\zeta}/2 \Psi}{N\eta} + \frac{\varphi Q_{\zeta}}{N\eta}} \leq \varphi Q_{\zeta} \Psi$$

(6.32)

We can now use (6.32), (6.28) and (6.6) to estimate the right hand side of (6.11) such that

$$G_{ii} = -w^{-1} \left[ 1 + m_G(i,\emptyset) + |z|^2 G_{ii}(i,\emptyset) + Z_i \right]^{-1}
$$

$$= -w^{-1} \left[ 1 + m - \frac{|z|^2}{w(1 + m)} + (m_G(i,\emptyset) - m) + \mathcal{E}_1 + Z_i \right]^{-1}$$

(6.33)

$$= -w^{-1} \left[ 1 + m - \frac{|z|^2}{w(1 + m)} \right]^{-1} - \mathcal{E}_2$$

(6.34)
where $E_1$ and $Z_i$ are bounded in (6.30) and (6.32) and $E_2$ is bounded by

$$E_2 = O \left( w^{-1} \left[ 1 + m - \frac{|z|^2}{w(1 + m)} \right]^{-2} \varphi^{Q}\Psi \right) \leq O(\varphi^{Q}\Psi).$$

In the last inequality, we have used (6.21) to bound $1 + m - \frac{|z|^2}{w(1 + m)}$ and (6.19) for $m\xi$.

Summing over the index $i$ in (6.34), we have

$$0 = w m + \left[ 1 + m - \frac{|z|^2}{w(1 + m)} \right]^{-1} + O(|w|\varphi^{Q}\Psi)$$

(6.35)

Hence we have proved

$$0 = w m(1 + m)^2 - m|z|^2 + 1 + m = O(\varphi^{Q}\Psi).$$

Together with the assumption (6.20) on $\Lambda$ and (4.9) on the order of $m\xi$, this proves (6.21).

**Corollary 6.8.** Under the assumptions of Lemma 6.7, the following properties hold. Let $T, U \in \mathbb{1}, N] $ such that $i / U$ and $|T| + |N| \leq C$. For any $\zeta > 0$ and $w \in S(b)$ for some $b > 5Q\xi$ with $Q\xi$ defined in Lemma C.1, we have with $\zeta$-high probability for any $i \in U$ that

$$G(T, U)_{ii}^T - G(\emptyset, i)_{ii} = O(\varphi^{Q}\Psi)$$

(6.36)

If $i \notin U$, then

$$G(T, U)_{ii}^T - G_{ii} = O(\varphi^{Q}\Psi).$$

(6.37)

**Proof.** We first prove the case $i \notin U$. We claim that the parallel version of (6.34) holds as well, i.e.,

$$G(T, U)_{ii} = -w^{-1} \left[ 1 + m - \frac{|z|^2}{w(1 + m)} \right]^{-1} + O(\varphi^{Q}\Psi)$$

(6.38)

Comparing (6.38) with (6.34), we have proved (6.37).

We now prove the case $i \in U$. By row-column symmetry, we have

$$G(T, U) = \left[ (Y(T, U))^* Y(T, U) - w \right]^{-1} = \left[ A(U, T) (A(U, T))^* - w \right]^{-1} := G(A)_{ii} = A = Y^*.$$

Hence we have to prove, for $i \in U$ and $i \notin T$, that

$$G(A)_{ii} - G(A)_{i, \emptyset} = O(\varphi^{Q}\Psi).$$

We will omit $A$ in the following argument.

One can extend (6.25) to $g_{ii}^T$ and obtain

$$g_{ii}^T = -\frac{1}{w(1 + m)} + E_1^{(T, U)} = O(\varphi^{Q}\Psi)$$

(6.39)

as in (6.24). Comparing (6.39) with the equation for $g_{ii}^{(i, \emptyset)}$ (6.24), we obtain (6.36) in the case $i \in U$. 

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We define for any sequence $A_i$ ($1 \leq i \leq N$) the quantity
\[ [A] := N^{-1} \sum_i A_i. \]

In application, we often use $A = Z$ or $A = \bar{Z}$. Define
\[ D(m) = m^{-1} + w + wm - \frac{|z|^2}{1 + m}. \]

The following lemma is our stability estimate for the equation $D(m) = 0$. Notice that it is a deterministic result. It assumes that $|D(m)|$ has a crude upper bound and then derives a more precise estimate on $\Lambda = |m - m_c|$.

**Lemma 6.9** (Stability of the self-consistent Equation). Suppose that $1 - |z|^2 > t > 0$. Let $\delta : \mathbb{C} \to \mathbb{R}_+$ be a continuous function satisfying the bound
\[ |\delta(w)| \leq (\log N)^{-8}|w|^{1/2}. \] (6.40)

Suppose that, for a fixed $E$ with $0 \leq E \leq C$ for some constant $C$ independent of $N$, (6.20) and the estimate
\[ |\Upsilon(m)(w, z)| = |D(m)m(1 + m)(w, z)| \leq \delta(w)|m_c|^2 \] (6.41)
hold for $10 \geq \eta \geq \tilde{\eta}$ which may depend on $N$. Denote $\varepsilon^2 := \kappa + \eta$ where $\kappa = |E - \lambda_+| \approx$ in our case that $1 - |z|^2 > t > 0$. Then there is an $M_0$ large enough independent of $N$ such that for any fixed $M > M_0$ and $N$ large enough (depending on $M$) the following estimates for $\Lambda = |m - m_c|$ hold for $10 \geq \eta \geq \tilde{\eta}$:

**Case 1**: $\Lambda \leq \frac{M^3/2 \delta}{|w|}$ or $\Lambda \geq \frac{1}{M^2|w|^{1/2}}$ if $\varepsilon^2 \geq 1/M^2$ (6.42)

**Case 2a**: $\Lambda \leq \frac{M \delta}{\varepsilon}$ or $\Lambda \geq \frac{2M \delta}{\varepsilon}$ if $\varepsilon^2 \leq 1/M^2$ and $\delta \leq \frac{\varepsilon^2}{M^{3/2}}$ (6.43)

**Case 2b**: $\Lambda \leq M \sqrt{\delta}$, or $\Lambda \geq 2M \sqrt{\delta}$ if $\varepsilon^2 \leq 1/M^2$ and $\delta \geq \frac{\varepsilon^2}{M^{3/2}}$ (6.44)

The three upper bounds (i.e., the first inequalities in (6.42)-(6.44)) can be summarized as
\[ \Lambda \leq C \frac{\delta(w)|w|^{-1}}{\sqrt{\kappa + \eta + \delta}} \] (6.45)

**Proof.** Define the polynomial
\[ P_{w,z}(x) = wx(1 + x)^2 + x(1 - |z|^2) + 1. \]

By definition of $\Upsilon$ (6.21), we have
\[ P_{w,z}(m) = wm(1 + m)^2 + m(1 - |z|^2) + 1 = \Upsilon = D(m)m(1 + m). \]

Since $P_{w,z}(m_c) = 0$, we have
\[ wu^3 + B(w, z)u^2 + A(w, z)u = \Upsilon, \quad u = m - m_c, \]
\[ B = w(3m_c + 2), \]
\[ A(w, z) = w(3m_c + 1)(m_c + 1) + 1 - |z|^2 = 2wm_c(1 + m_c) - \frac{1}{m_c}. \]

By definition of \( P_{w,z} \), we can express \( A \) and \( B \) by
\[ P'_{w,z}(m_c(w, z)) = A(w, z), \quad P''_{w,z}(m_c(w, z)) = 2B(w, z). \]

**Case 1:** In this case, we claim that the following estimates concerning \( A \) and \( B \) hold:
\[ |A| \geq C/M, \quad B = O(|w|^{1/2}). \] (6.46)

Since \( A \) and \( B \) are explicit functions of \( m_c \), equation (6.46) is just properties of the solution \( m_c \) of the third order polynomial \( P_{w,z}(m) \). We now give a sketch of the proof. Consider first the case \(|w| \ll 1\). Then (6.46) follows from (4.9), (4.10), (4.6) and the definitions of \( A \) and \( B \).

We now assume that \( w \sim 1 \). Clearly, \(|B| \leq O(1) \sim |w|^{1/2}|, which gives (6.46) for \( B \). To prove \(|A| \geq C/M\), by definition of \( m_c \), we have \( w = \frac{1 - m_c + m_z |z|^2}{m_c(1 + m_c)} \). Thus we can rewrite \( A \) as
\[ A = \frac{-1 - 3m_c + 2m_c(-1 + |z|^2)}{m_c(1 + m_c)} = \frac{2(-1 + |z|^2)}{m_c(1 + m_c)}(m_c - a_+)(m_c - a_-), \]
where \( a_\pm := \frac{3 \pm \sqrt{1 + 8|z|^2}}{4(-1 + |z|^2)} = \frac{-2}{3 \pm \sqrt{1 + 8|z|^2}} \).

By (4.9) and (6.11) (where \( \alpha = \sqrt{1 + 8|z|^2} \)), we obtain (6.41).

We now prove (6.42) by contradiction. If (6.42) is violated then with \( u = m - m_c \) we have
\[ |\Upsilon| = |u||A(w, z) + B(w, z)u + wu^2| \geq \frac{M^{3/2} \delta}{|w|} \left[ \frac{C_1}{M} - \frac{C_2}{M^2} - \frac{C_3}{M^3} \right] \geq \frac{C \sqrt{M} \delta}{|w|}, \]
where \( M \) is a large constant in the last inequality. By (6.14) and (4.3), \(|\Upsilon| \leq C\delta/|w|\). Thus we have
\[ \frac{C \sqrt{M} \delta}{|w|} \leq |\Upsilon| \leq \frac{C\delta}{|w|} \]
which is a contradiction provided that \( M \) is large enough.

**Case 2:** \( \varepsilon^2 := \kappa + \eta \leq 1/M^2 \). Note in this case \( w \sim 1 \). Then by (4.3) we have
\[ B = w, \quad \kappa(z) = 0 \] (6.47)
where the last equation can be checked by direct computation and we used \(|z|^2 < 1 - t < 1 \). There is a more intrinsic reason why the last equation for \( A \) holds. Notice that \( \lambda_+ \) is a point that the polynomial \( P_{w,z}(m)|_{m=\lambda_+} \) has a double root. Therefore, we have \( 0 = P'_{w,z}(m_c(\lambda_+, z)) = A(\lambda_+, z) \).

Notice that in the case \( \kappa + \eta \) is small enough, we can approximate \( A(w, z) \) by linearizing \( w \) r.t. \( w = \lambda_+ \). Thus by the defining equation \( P''_{w,z}(m_c(\lambda_+, z)) = A(\lambda_+, z) \), we have
\[ A(w, z) \sim P''_{w,z}(m_c(\lambda_+, z))(m_c(w, z) - m_c(\lambda_+, z)) + \frac{\partial P_{w,z}}{\partial w}(m_c(\lambda_+, z))(w - \lambda_+) \sim \sqrt{\kappa + \eta} = \varepsilon \] (6.48)
where we have used that \( P_{m}(\lambda, z) = B(\lambda, z) \sim 1 \), \( \frac{\partial P_{m}}{\partial w}(m_{c}(\lambda, z)) \sim 1 \) and, by (1.9), that \( (m_{c}(w, z) - m_{c}(\lambda, z)) \sim \frac{C}{\sqrt{\kappa + \eta}}. \) While we can also check the conclusion of (6.43) by direction computation, the current derivation provides a more intrinsic reason why it is correct.

Case 2a: Suppose (6.43) is violated. We first choose \( M \) large enough so that \( |m_{c}(1 + m_{c})| \leq M^{1/4} \) in this regime. Then by (6.47) and (6.48), with \( w \) large, we have

\[
C \delta M^{1/4} \geq |\partial P| = |u| |A(w, z) + B(w, z)u + wu^{2}| \geq \frac{\delta M}{\varepsilon} \left[ C_{1} \varepsilon - \frac{C_{2} M \delta}{\varepsilon} - \frac{C_{3} M^{2} \delta^{2}}{\varepsilon^{2}} \right] \geq C_{1} \delta M / 2,
\]

which is a contradiction provided that \( M \) is large enough. Here we have used that, by the restriction of \( \varepsilon \) and \( \delta \) in (6.33) that \( \varepsilon \geq M^{3/4} \sqrt{\delta}, \) \( M \) is large enough constant and \( \delta \ll 1. \)

Case 2b: Suppose (6.44) is violated. Similarly we have

\[
C \delta M^{1/4} \geq |\partial P| = |u| |B(w, z)u + A(w, z) + wu^{2}| \geq |u| \left[ C_{1} M \sqrt{\delta} - C_{2} \varepsilon - C_{3} M^{2} \delta \right] \geq C_{1} |u| \left[ M \sqrt{\delta}/2 - C_{2} \varepsilon \right] \geq C_{1} M^{2} \delta / 4
\]

which is a contradiction. Here we have used, by the restriction of \( \varepsilon \) and \( \delta \) in (6.44) and \( M \) is large enough constant, that \( C_{2} \varepsilon \leq C_{2} M^{3/4} \sqrt{\delta} \leq M \sqrt{\delta}/20. \)

With a slight stronger condition on \( \delta \) and an initial estimate \( \Lambda \ll 1 \) when \( \eta \sim 1, \) the first inequalities in (6.32)-(6.44), i.e., (6.43), always hold. We state this as the following Corollary, which is a deterministic statement.

**Corollary 6.10 (Deterministic continuity argument).** Suppose that the assumptions of Lemma 6.9 hold. If we have

\[
\Lambda(E + 10i) \ll 1
\]

and that \( \delta \) is decreasing in \( \eta \) for \( \varepsilon = \sqrt{\kappa + \eta} \) small enough, then (6.43) holds all \( \eta \in [\tilde{\eta}, 10]. \)

**Proof.** By assumption \( \Lambda(E + 10i) \ll 1 \) and the left inequality of (6.42) holds for \( \eta = 10. \) By continuity of \( \Lambda, \) the same inequality,

\[
\Lambda \leq \frac{M^{3/2} \delta}{|w|},
\]

holds for \( w = E + i \eta \) as long as \( \eta \in [\tilde{\eta}, 10] \) and \( \varepsilon \geq 1/M. \)

Suppose that as \( \eta \) decreases, we get to Case 2a. Notice that when we decrease \( \eta, \) by the conditions on \( \varepsilon \) we will not go back to Case 1 from either Case 2a or Case 2b. For any \( \varepsilon \ll 1/M \) with \( M \) large, we have

\[
\frac{M^{3/2} \delta}{|w|} \leq \frac{M \delta}{2 \varepsilon}.
\]

Hence at the transition point from Case 1 to Case 2a, the inequality \( \Lambda(E + i \eta) \leq \frac{M \delta}{\varepsilon} \) holds. Thus by continuity of \( \Lambda, \) the bound \( \Lambda(E + i \eta) \leq \frac{M \delta}{\varepsilon} \) in (6.44) holds until we leave Case 2a.

It is possible that we cross from Case 2a to Case 2b. At the transition point, we have \( \delta = \frac{M \delta}{\varepsilon} \) and thus

\[
\frac{M \delta}{\varepsilon} \leq \frac{1}{2} M \sqrt{\delta}
\]
for $M$ large. Hence the first inequality of Case 2b, i.e., $\Lambda \leq M\sqrt{\delta}$ holds. By continuity, this bound continues to hold unless we leave Case 2b. Since $\delta$ is decreasing in $\eta$ when $\varepsilon$ is small, once we get to Case 2b, we will not go back to Case 2a (or Case 1 as explained before).

It is possible that the Case 2a is omitted and we get to Case 2b directly from Case 1. Notice that $\varepsilon = 1/M$ at such a transition point and we have $|w| \sim 1$. Furthermore, by (6.40), we get $\delta \leq 1/\log N$ at the transition point. Putting these together, we have for $M$ large,

$$
\frac{M^{3/2}\delta}{|w|} \leq \frac{1}{2} M\sqrt{\delta}.
$$

Hence the bound $\Lambda(E + i\eta) \leq M\sqrt{\delta}$ in (6.44) holds.

### 6.3 The large $\eta$ case.

Our method to estimate the Green functions and the Stieltjes transform is to fix the energy $E$ and apply a continuity argument in $\eta$ by first showing that the crude bound in Lemma 6.9 holds for large $\eta$. In order to start this scheme, we need to establish estimates on the Green functions when $\eta = O(1)$. This is the main focus of this subsection. We start with the following lemma which provide a crude bound on the Green functions.

**Lemma 6.11.** For any $w \in S(0)$ and $\eta > c > 0$ for fixed $c$, we have the bound

$$
\max_{i,j \notin U} |G_{ij}^{(U,T)}(w)| \leq C.
$$

(6.49)

for some $C > 0$. Notice that this bound is deterministic and is independent of the randomness.

**Proof.** By definition, we have

$$
|G_{ij}| = \left| \sum_{\alpha} \frac{u_{\alpha}(i)\Pi_{\alpha}(j)}{\lambda_{\alpha} - w} \right| \leq \frac{1}{\eta} \sum_{\alpha} u_{\alpha}(i)\Pi_{\alpha}(j) \leq \frac{1}{\eta} \leq C
$$

where we have used $|\lambda_{\alpha} - w| \geq \Im w = \eta$. Furthermore, $G_{ij}^{(U,T)}$ can be bounded similarly. \hfill \Box

The main result of this subsection is the following bound on $\Lambda$.

**Lemma 6.12.** For any $\zeta > 0$ and $\varepsilon > 0$, we have

$$
\max_{w \in S(0), \eta = 10} \Lambda(w) \leq N^{-1/2 + \varepsilon}
$$

(6.50)

with $\zeta$-high probability.

**Proof.** From (6.25)-(6.27), for $\eta = O(1)$ we have

$$
\left[ G_{ii}^{(i,\emptyset)} \right]^{-1} = -w \left[ 1 + m_{G}^{(i,i)} + Z_{i}^{(i)} \right], \quad |m_{G}^{(i,i)} - m| \leq \frac{C}{N}.
$$

From (6.49), we have $|G_{ij}| + |G_{ji}| \leq \eta^{-1} \leq O(1)$ and $|m_{G}^{(i,i)}| \leq O(1)$. Hence the large deviation estimate (6.27) becomes, with $\zeta$-high probability,

$$
|Z_{i}^{(i)}| \leq \varphi^{C_{\zeta}} \sqrt{\frac{\Im m_{G}^{(i,i)}}{N}} \leq \varphi^{C_{\zeta}} N^{-1/2}.
$$

(6.51)

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Thus for any $\varepsilon > 0$ we have

$$ g^{(i, \emptyset)}_{ii} := -\frac{1}{w(1 + m + O(N^{-1/2+\varepsilon}))} $$

Together with (6.11), we obtain

$$ G_{ii}^{-1} = -w - w m^{(i, \emptyset)} + \frac{|z|^2}{1 + m + O(N^{-1/2+\varepsilon})} - w Z_i. $$

By an argument similar to the one used in (6.51), we can estimate $Z_i$ by

$$ |Z_i| \leq N^{-1/2+\varepsilon} $$

for any $\varepsilon > 0$ with $\zeta$-high probability. This implies that, with $\zeta$-high probability,

$$ G_{ii}^{-1} = -w - w m + \frac{|z|^2}{1 + m + O(N^{-1/2+\varepsilon})} + O(w N^{-1/2+\varepsilon}). \quad (6.52) $$

For any $\eta$ fixed, we claim that the following inequality between the real and imaginary parts of $m$ holds:

$$ |\text{Re} m| \leq 2 \sqrt{\frac{\text{Im} m}{\eta}}. \quad (6.53) $$

To prove this, we note that for any $\ell \geq 1$

$$ N^{-1} \sum_{|\lambda_j - E| \geq \ell \eta} \frac{E - \lambda_j}{(E - \lambda_j)^2 + \eta^2} \leq \frac{1}{\ell \eta}, $$

$$ N^{-1} \sum_{|\lambda_j - E| \leq \ell \eta} \frac{|E - \lambda_j|}{(E - \lambda_j)^2 + \eta^2} \leq N^{-1} \sum_{|\lambda_j - E| \leq \ell \eta} \frac{\ell \eta}{(E - \lambda_j)^2 + \eta^2} \leq \ell \text{Im} m. $$

Summing up these two inequalities and optimizing $\ell$, we have proved (6.53).

Assume that $\text{Im} m \leq c(\log N)^{-1}$. From (6.53), we have $|m| \leq c(\log N)^{-1/2}$. Together with $\text{Im} w = \eta \sim 1$,

$$ |m| = N^{-1} \left| \sum_i G_{ii} \right| = N^{-1} \left| \sum_i \left( -w - w m + \frac{|z|^2}{1 + m} \right)^{-1} \right| + O(N^{-1/2+\varepsilon}) \geq (-w + |z|^2 + o(1))^{-1} \geq C $$

for some constant $C$. This contradicts $|m| \leq c(\log N)^{-1/2}$ and we can thus assume that $\text{Im} m \geq c(\log N)^{-1}$ when $\eta \sim 1$ and $w = O(1)$. In this case, we also have

$$ |1 + m| \geq C(\log N)^{-1}. $$

Then (6.52) implies for any $\varepsilon > 0$ that with $\zeta$-high probability

$$ G_{ii} = \left( -w - w m + \frac{|z|^2}{1 + m} \right)^{-1} + O(N^{-1/2+\varepsilon}) $$

Summing up all $i$, we have the following equation for $m$ with $\zeta$-high probability:

$$ m = \frac{-1 - m}{w(1 + m)^2 - |z|^2} + O(N^{-1/2+\varepsilon}). $$
We can rewrite this equation into the following form:

$$P_{w,z}(m) = w(1+m)^2m - |z|^2m + m + 1 = O(N^{-1/2+\varepsilon}).$$  \hfill (6.54)

It can be checked (with computer calculation or rather complicated but elementary algebraic calculation) that for $0 \leq E \leq 5\lambda_+$ and $\eta = O(1)$, the third order polynomial $P_{w,z}(m)$ has no double root and there is only one root with positive real part. We denote this root by $m_1$ and the other two roots by $m_2$ and $m_3$. For $0 \leq E \leq 5\lambda_+$ and $t \leq \eta \leq t^{-1}$ for any $t$ fixed, the three roots are separate by order one due to compactness. Since there is no double root, we have $|P_{w,z}(m_1)| \geq c > 0$ whenever $0 \leq E \leq 5\lambda_+$ and $t \leq \eta \leq t^{-1}$. Thus the stability of (6.54) is trivial and we have proved that in this range of parameters

$$|m(w,z) - m_1(w,z)| = O(N^{-1/2+\varepsilon})$$

for any $\varepsilon > 0$ with $\zeta$-high probability.

### 6.4 Proof of the weak local Green function estimates.

In this subsection, we finish the proof of Theorem 6.1. We fix an energy $E$ and we will decrease the imaginary part $\eta$ of $w = E + i\eta$. Recall all stability results are based on assumption (6.20), i.e., $\Lambda \leq \alpha|m| \sim |w|^{-1/2}$ for some small constant $\alpha$, which so far was established only for large $\eta$ in (6.54). We would like to know that this condition continue to hold for smaller $\eta$. More precisely, suppose that (6.20) holds in a set $A$ for all $w = E + \eta$ with $\eta \in [\tilde{\eta}, 10]$ where $\tilde{\eta}$ satisfies

$$\tilde{\eta} \geq \varphi^5 N^{-1/2}, \quad b > 5Q\zeta.$$  \hfill (6.55)

We can choose $\tilde{\eta} = \eta_1 < \eta_2 \ldots < \eta_n = 10$ such that $|\eta_{j+1} - \eta_j| \leq N^{-20}$ and $n = O(N^{20})$. By (6.21) and (6.50) we have with $\zeta$-high probability in $A$,

$$\Upsilon(w) \leq O(\varphi^Q \Psi)(w) \leq \varphi^Q \sqrt{\frac{|w|^{-1/2}}{N\eta}}$$  \hfill (6.56)

for all $w = E + i\eta_j$ for all $1 \leq j \leq n$. Since $\Lambda(E + i\eta)$ is continuous in $\eta$ at a scale, say, $N^{-10}$, (6.56) holds for all $\eta \in [\tilde{\eta}, 10]$ with $\zeta$-high probability in $A$. Hence for $\eta$ satisfying (6.55) the estimate (6.41) holds with

$$\delta = C\varphi^Q |w| \left( \frac{|w|^{-1/2}}{N\eta} \right)^{1/2}$$

With this choice, we can check that the assumption on $\delta$, (6.40), holds as well. Furthermore $\delta$ is decreasing in $\eta$ when $\varepsilon = \sqrt{\kappa + \tilde{\eta}}$ is small enough. By Corollary 6.10, (6.45) holds all $\eta \in [\tilde{\eta}, 10]$.

For $|z| < 1 - t$ for some $t > 0$, if $\kappa \ll 1$ then $|w| \sim 1$ and (6.45) implies

$$\Lambda \leq C\sqrt{\delta(w)} \leq \varphi^Q |w| \left( \frac{1}{N\eta} \right)^{1/4}. $$

If $\kappa \gg c > 0$ for some $c > 0$ then

$$\Lambda \leq C\delta(w)|w|^{-1} \leq C\varphi^Q \varphi \left( \frac{|w|^{-1/2}}{N\eta} \right)^{1/2} \leq C\varphi^Q \varphi \frac{1}{|w|^{1/2}} \left( \frac{|w|^{1/2}}{N\eta} \right)^{1/4}. $$  \hfill (6.57)

Combining both cases, for any $w \in S(b), b > 5Q\zeta$, we have with $\zeta$-high probability in $A$ that

$$\Lambda \leq \varphi^Q \varphi \frac{1}{|w|^{1/2}} \left( \frac{|w|^{1/2}}{N\eta} \right)^{1/4} \leq C\varphi^{-Q\zeta/5}|w|^{-1/2} \sim C\varphi^{-Q\zeta/5}|m_e|.$$  \hfill (6.58)
Suppose that \( \hat{\eta} := \tilde{\eta} - N^{-20} \in S(b) \) for some \( b > 5Q \). Then for any \( \eta \in [\tilde{\eta} - N^{-20}, \tilde{\eta}] \), by (6.58) and the continuity of \( \Lambda \), we have

\[
\Lambda(E + i\eta) \leq \Lambda(E + i\tilde{\eta}) + N^{-10} \leq C_0 \varphi^{-Q} \cdot \frac{1}{N^{1/2}} + N^{-10} \leq \alpha \|m_c(E + i\tilde{\eta})\|/2
\]

Thus the condition (6.20) in Lemma 6.7 is satisfied with \( \zeta \)-high probability in \( A \). Since we can start this procedure with \( \tilde{\eta} = 10 \) and there are only \( N^C \) steps to get to \( \tilde{\eta} = \varphi^{5Q} \cdot N^{-1} |w|^{1/2} \), we have proved that (6.58) holds for all \( w \in S(b) \) with \( b > 5Q \). Notice that from now on the assumption (6.20) holds with \( \zeta \)-high probability.

We can now prove the estimate (6.1) on the diagonal term. Comparing (6.35) with (6.38) \( (T = U = \emptyset) \), for any \( w \in S(b) \), \( b > 5Q \), we have with \( \zeta \)-high probability

\[
|G_{ii} - m_c| \leq O(\varphi^Q \cdot \Psi) \tag{6.59}
\]

By definition of \( \Psi \), (6.58) and \( m_c \sim |w|^{-1/2} \), we have

\[
\Psi = \left( \sqrt{\frac{\text{Im} m_c + \Lambda}{N \eta}} + \frac{1}{N \eta} \right) \leq \left( \sqrt{\frac{|w|^{-1/2}}{N \eta}} + \frac{1}{N \eta} \right).
\]

Using the restriction on \( \eta \) so that \( N \eta \geq |w|^{1/2} \varphi^{5Q} \), we have

\[
\Psi \leq C \left( \sqrt{\frac{|w|^{-1/2}}{N \eta}} \right) \leq C |w|^{-1/2} \left( \frac{\sqrt{w}}{N \eta} \right)^{1/4} \tag{6.60}
\]

With (6.57) and (6.59), we have thus proved that

\[
\max_i |G_{ii} - m_c| \leq \varphi^Q |w|^{-1/2} \left( \frac{\sqrt{w}}{N \eta} \right)^{1/4}
\]

for any \( w \in S(b) \), \( b > 5Q \). Hence the estimate (6.1) on the diagonal element \( G_{ii} \) holds.

To conclude Theorem 6.1, it remains to prove the estimate on the off-diagonal elements. Recall the identity (6.12) for \( G_{ij} \) and the equations (C.3) and (C.4). We can estimate the off-diagonal Green function by

\[
|G_{ij}| = \left| \text{Im} G_{ij}^{(i,0)} |z|^2 (G_{ij}^{(i,j,0)})^* \right| + O \left( \varphi^Q \sqrt{\frac{\text{Im} m_G^{(i,j,0)} + \text{Im} G_{ij}^{(i,j,0)} + \text{Im} G_{ij}^{(i,j,0)}}{N \eta}} \right), \quad i \neq j,
\]

\[
|G_{ij}| = \left| |z|^2 G_{ij}^{(i,j,0)} \right| + O \left( \varphi^Q \Psi \right), \quad i \neq j. \tag{6.61}
\]

Here we have used \( |G_{ij} G_{ij}^{(i,0)}| = O(|w|^{-1}) \), which follows from (6.36), \( \Lambda \ll m_c \) and \( |m_c| \sim |w|^{-1/2} \).

Recall the identity (6.14) that

\[
G_{ij}^{(i,j,0)} = -w G_{ii}^{(i,0)} G_{ij}^{(i,j)} \left( y_i^{(i,j)} G_{ij}^{(i,j,0)} y_j^{(i,j,0)} \right), \quad i \neq j.
\]

By (C.2), we have

\[
\left| \left( y_i^{(i,j)} G_{ij}^{(i,j,0)}, y_j^{(i,j,0)} \right) \right| \leq \varphi^Q \sqrt{\frac{\text{Im} m_G^{(i,j)}}{N \eta}}.
\]
where we have used (C.4) and that, by definition, \( \text{Im} G_{ij}^{(ij,ij)} = 0 = \text{Im} G_{jj}^{(ij,ij)} \). Therefore, we have with \( \zeta \)-high probability,
\[
|G_{ij}^{(ij,ij)}| \leq \varphi^{Q_\zeta} \sqrt{\frac{\text{Im} m_c + \Lambda + (N\eta)^{-1}}{N\eta}} \leq \varphi^{Q_\zeta} \Psi, \quad i \neq j,
\]
(6.62)
where we also used \( |G_{ij}^{(ij,ij)}| G_{jj}^{(ij,ij)}| \leq C|m_c|^2 \leq C|w|^{-1} \). Together with (6.61) and (6.36), we have proved that with \( \zeta \)-high probability
\[
|G_{ij}| \leq \varphi^{Q_\zeta} \Psi, \quad i \neq j.
\]
(6.63)
With (6.60), it proves Theorem 6.1 for the off-diagonal elements provided that \( w \in S(b) \) with \( b > 5Q_\zeta \). Finally, we rename \( b \) as the \( C_\zeta \) and this concludes the proof of Theorem 6.1.

7 Proof of the strong local Green function estimates

Lemma 6.7 provides an error estimate to the self-consistent equation of \( m \) linearly in \( \Psi \). The following Lemma improves this estimate to quadratic in \( \Psi \). This is the key improvement leading to a proof of the strong local Green function estimates, i.e., Theorem 6.1.

Lemma 7.1. For any \( \zeta > 1 \), there exists \( R_\zeta > 0 \) such that the following statement holds. Suppose for some deterministic number \( \Lambda(w, z) \) (which can depend on \( \zeta \)) we have
\[
\Lambda(w, z) \leq \bar{\Lambda}(w, z) \ll m_c(w, z)
\]
for \( w \in S(b) \), \( b > 5R_\zeta \), in a set \( \Xi \) with \( P(\Xi^c) \leq e^{-p_N (\log N)^2} \) and \( p_N \) satisfies that
\[
\varphi \leq p_N \leq \varphi^{2\zeta}.
\]
(7.1)
Then there exists a set \( \Xi' \) such that \( P(\Xi'^c) \leq e^{-p_N} \) and
\[
\mathcal{D}(m(w, z)) \leq \frac{1}{2} \varphi^{R_\zeta |m_c|^{-3}} \bar{\Psi}^2, \quad \bar{\Psi} := \sqrt{\frac{\text{Im} m_c + \bar{\Lambda}}{N\eta}} + \frac{1}{N\eta}, \quad \text{in } \Xi'.
\]
(7.2)
Notice that the probability deteriorates in the exponent by a \( (\log N)^{-2} \) factor.

We remark that, by Lemma 4.1, \( \text{Im} m_c \ll |m_c| \) when \( \eta + \kappa \ll 1 \). Hence we have to track the dependence of \( \text{Im} m_c \) carefully in the previous Lemma. This is one major difference between the weak and strong local Green function estimates. Similar phenomena occur for the Stieltjes transforms of the eigenvalue distributions of Wigner matrices. Lemma 7.1 will be proved later in this section; we now use it to prove Theorem 3.4. We first give a heuristic argument.

Suppose that we have the estimate (7.2) with \( \bar{\Psi} \) replaced by \( \Psi \). We assume \( \Lambda \geq (N\eta)^{-1} \) for convenience so that \( \Psi^2 \sim (\text{Im} m_c + \Lambda)/(N\eta) \) (If this assumption is violated then (3.5) holds automatically and we have nothing to prove). Then we can apply Corollary 6.10 by choosing
\[
\delta = \varphi^{R_\zeta |w|^{3/2}} \left[ \frac{\text{Im} m_c + \Lambda}{N\eta} \right]
\]
(7.3)
which implies (6.45). Consider first the case $\kappa + \eta \sim O(1)$. Using (6.45) with the choice of $\delta$ in (7.3) and $\kappa + \eta + \delta \geq O(1)$, we have
\[
\Lambda \leq \varphi_R \frac{1}{N \eta}.
\]
When $\eta$ satisfies the condition (6.55), the coefficient of $\Lambda$ on the right side of the last equation is smaller than $1/2$. Hence, using $\text{Im} m_c \leq |m_c| \leq C|w|^{-1/2}$ (see Proposition 3.2), we have
\[
\Lambda \leq C\varphi_R \frac{1}{N \eta}.
\]
We now consider the case $\kappa + \eta \ll 1$ and thus $|w| \sim O(1)$. From the first inequality of (6.45), we have
\[
\Lambda \leq C \frac{\delta(w)|w|^{-1}}{\sqrt{\kappa + \eta + \delta}|w|} \leq C \sqrt{\delta(w)}.
\]
(7.4)
Also, in the regime $\kappa + \eta \ll 1$, (4.4) asserts that
\[
\text{Im} m_c \leq C\sqrt{\kappa + \eta}, \quad \frac{\text{Im} m_c}{N \eta \sqrt{\kappa + \eta + \delta}} \leq C.
\]
Using the choice of $\delta$ in (7.3), we have
\[
\Lambda \leq C \varphi_R \frac{1}{N \eta \sqrt{\kappa + \eta}} \leq C \varphi_R \frac{1}{N \eta} + C \varphi_R \frac{\Lambda}{N \eta \sqrt{\kappa + \eta + \delta}} \leq C' \varphi_R \frac{1}{N \eta}
\]
where we have used (7.3) to absorb the last term involving $\Lambda$ in the last inequality with a change of constant $C$. This completes the heuristic proof of Theorem 3.4. We now give a formal proof of this theorem assuming Lemma 7.1.

**Proof of Theorem 3.4.** We first prove (3.6) assuming (3.5). By (3.6) and the definition of $\Psi$, we have for $i \neq j$,
\[
|G_{ij}| \leq \varphi_R \left( \sqrt{\frac{\text{Im} m_c + \Lambda}{N \eta}} + \frac{1}{N \eta} \right) \leq \varphi_R \left( \sqrt{\frac{\text{Im} m_c}{N \eta}} + \frac{1}{N \eta} \right)
\]
where we have used (3.5) in the last step. This proves (3.6).

The main task in proving Theorem 3.4 is to prove (3.5). We first consider the case that $|z| \leq 1 - \delta$. We assume that $\zeta$ is large enough, e.g., $\zeta \geq 10$. By Theorem 6.1 and $m_c \sim |w|^{-1/2}$ (1.9) for $|z| < 1 - \delta$, there exists a constant $C_{\zeta+5}$ such that for any $w \in S(b)$, $b > 5C_{\zeta+5}$ and $\alpha \ll 1$, we have
\[
\Lambda(w) \leq \Lambda_1 := \alpha |m_c| \sim O(\alpha |w|^{-1/2}),
\]
holds with the probability larger than $1 - \exp(-\varphi^{C_{\zeta+5}})$ (here we have replaced $\zeta$ in Theorem 6.1 by $\zeta + 5$ for the convenience of the following argument). Since $S(b)$ is decreasing in $b$, we can choose $D_\zeta = 5 \max(C_{\zeta+5}, R_{\zeta})$ so that we can apply Lemma 6.1 with $p_N = \varphi^{C_{\zeta+5}}$ (which guarantees (7.1)). Together with $\Lambda_1 \leq |m_c|$, we have, for any $w \in S(D_\zeta)$ fixed,
\[
\mathcal{D}(m) \leq \frac{1}{2} \varphi_{\zeta+5} |m_c|^{-3} \Psi_1^2, \quad \Psi_1 := \sqrt{\frac{\text{Im} m_c + |m_c|}{N \eta}} + \frac{1}{N \eta},
\]
(7.6)
holds with the probability larger than $1 - \exp(-\varphi^6 (\log N)^{-2})$. Notice that the application of Lemma 7.4 causes the probability in the exponent to deteriorate by a $(\log N)^{-2}$ factor.

Using (7.6), we can apply Corollary 6.10 with

$$\delta = \delta_1 := \varphi R_c |m_c|^{-3} \Psi_1^2.$$  \hspace{1cm} (7.7)

Here the assumption of $\Lambda(E + 10i)$ is guaranteed by (6.30). By definition of $\Psi_1$, $\varphi$ and $|m_c| \sim |w|^{-1/2}$, we have

$$\delta \leq \varphi R_c |w| \ll (\log N)^{-8}|w|^{1/2}.$$  \hspace{1cm}

Furthermore, it is easy to prove that $\delta$ is decreasing in $\eta$ when $\kappa + \eta$ is small. We have thus verified the assumptions on $\delta$ in Corollary 6.10 with the choice $\delta = \delta_1$ given in (7.7). From (6.45), we obtain for $w \in S(D_\zeta)$, with $C_0$ being the $C$ in (6.45),

$$\Lambda \leq C_0 \frac{\delta_1 |w|^{-1}}{\sqrt{\kappa + \eta + \delta_1}} \leq C_0 \frac{\varphi R_c}{N \eta \sqrt{\kappa + \eta + \delta_1}}$$

holds with the probability larger than $1 - \exp(-\varphi^6 (\log N)^{-2})$. We have thus proved (3.3) provided that $\kappa + \eta \geq (\log N)^{-1}$.

We now prove (3.3) when $\kappa + \eta \leq (\log N)^{-1}$. We have in this case $|w| \sim 1$. We apply Lemma 7.4 with $\tilde{\Lambda} = \Lambda_1 = |m_c| \sim 1$ given by (7.5). Thus (7.6) holds and we apply Corollary 6.10 with $\delta = \delta_1$ (7.7). Since $\Lambda_1 \geq (N \eta)^{-1}$ and $\Im m_c \sim \sqrt{\kappa + \eta}$ (4.4), the conclusion of Corollary 6.10 implies that for $w \in S(D_\zeta)$,

$$\Lambda \leq C_0 \varphi R_c |w|^{1/2} \frac{\Im m_c + \Lambda_1}{N \eta \sqrt{\kappa + \eta + \delta_1}} \leq C_1 \varphi R_c \frac{1}{N \eta} + C_1 \varphi R_c \frac{\Lambda_1}{N \eta \sqrt{\delta_1}}$$

holds with probability larger than $1 - \exp(-\varphi^6 (\log N)^{-2})$. Here $C_1$ depends only on $C_0$. From the definition of $\delta_1$ and $\Psi_1$, we have

$$\frac{\varphi R_c \Lambda_1}{N \eta \sqrt{\delta_1}} \leq \varphi R_c |m_c|^{3/2} \frac{\Lambda_1}{N \eta} \leq \frac{\Lambda_1}{\sqrt{\delta_1}} \leq \frac{\Lambda_1}{N \eta},$$

where for the last inequality we used

$$\Psi_1 \geq \sqrt{\Lambda_1/(N \eta)}.$$  

Since $\Lambda_1 \geq (N \eta)^{-1}$, combining the last two inequalities, for $w \in S(D_\zeta)$, we have

$$N \eta |\Lambda| \leq C_3 \varphi R_c + C_3 \varphi R_c^2 (N \eta \Lambda_1)^{1/2} \leq \varphi R_c (N \eta \Lambda_1)^{1/2}$$  \hspace{1cm} (7.8)

holds with the probability larger than $1 - \exp(-\varphi^6 (\log N)^{-2})$ for some $C_3$. Notice that we have used $N \eta \geq \varphi^{5R_c}$ in the last step in (7.8).

Repeating this process with the choices

$$N \eta \Lambda_2 := \varphi R_c (N \eta \Lambda_1)^{1/2}, \quad \Psi_2 := \sqrt{\frac{\Im m_c + \Lambda_2}{N \eta}}, \quad \delta_2 := \varphi R_c |m_c|^{-3} \Psi_2^2,$$

for $w \in S(D_\zeta)$, we obtain that

$$N \eta |\Lambda| \leq C_3 \varphi R_c + C_3 \varphi R_c^2 (N \eta \Lambda_2)^{1/2} \leq \varphi R_c (N \eta \Lambda_2)^{1/2}$$

for some $C_3$. Notice that we have used $N \eta \geq \varphi^{5R_c}$ in the last step in (7.8).
holds with the probability larger than $1 - \exp(-\varphi^{5}(\log N)^{-4})$. Notice that the last constant $C_3$ is the same as the one appears in (7.8) and it does not change in the iteration procedure. We now iterate this process $K$ times to have
\[
N\eta|\Lambda| \leq \varphi^{R_\zeta} (N\eta \Lambda_1)^{1/2} \leq \varphi^{2R_\zeta} (N\eta\Lambda_1)^{1/2}
\]
holds with the probability larger than $1 - \exp(-\varphi^{5}(\log N)^{-2K})$. We need $K$ so large that
\[
(A_1 N\eta)^{1/(2K)} \leq (CN)^{1/(2K)} \leq \varphi,
\]
i.e.,
\[
K \geq \frac{(\log \log (CN) - \log \log \varphi)}{\log 2} = \frac{(\log \log (CN) - 2 \log \log \log N)}{\log 2}
\]
on the other hand, we need $K$ small enough so that
\[
1 - \exp(-\varphi^{5}(\log N)^{-2K}) \geq 1 - \exp(-\varphi^{5}), \quad \text{i.e.,} \quad \varphi^{5}(\log N)^{-2K} \geq 1. \quad (7.9)
\]
We note that it also guarantees (7.10), since
\[
N\eta|\Lambda| \leq \varphi^{2R_\zeta+1}
\]
with the probability larger than $1 - \exp(-\varphi^5)$ which implies (7.10) when $\kappa + \eta \leq (\log N)^{-1}$. This completes the proof of Theorem 3.4.

7.1 Proof of Lemma 7.1

The first step in proving Lemma 7.1 is to derive a second order self-consistent equation which identifies the first order dependence of the correction in the self-consistent equation derived in Lemma 6.7. The second order terms will be bounded by $\Psi^2$; the first order terms are of the forms of averages of $Z_{i}^{(i)}$ and $Z_{i}$. In Lemma 7.3, the averages of $Z_{i}^{(i)}$ and $Z_{i}$ will be estimated by $\Psi^2$. This improvement from the naive order $\Psi$ to $\Psi^2$ is the key ingredient to obtain the strong local law. We remark that $\text{Im} \ m_c \ll |m_c|$ when $\eta + \kappa \ll 1$. Hence the dependence of $\text{Im} \ m_c$ versus $m_c$ has to be tracked carefully. We now state the second order self-consistent equation: the following lemma.

Lemma 7.2 (second order self-consistent equation). For any constant $\zeta > 0$, there exists $C_\zeta > 0$ such that
\[
\mathcal{D}(m) \leq O \left( \varphi^{C_\zeta} \frac{1}{m_c^2} \Psi^2 + w[Z] + m_c^{-2}[Z^*_i] \right) \quad (7.11)
\]
where
\[
[Z^*_i] = N^{-1} \sum_i Z_{i}^{(i)}, \quad [Z] = N^{-1} \sum_i Z_{i}.
\]

Proof. We have proved the weak local Green function estimate, i.e., Theorem 6.1 in Section 6. This in particular implies that (6.20) holds with $\zeta$-high probability in $S(b)$ for large enough $b$ with $\zeta$-high probability. With this remark in mind, we now prove Lemma 7.2.
We first take the inverse of both sides of (6.33) and sum up \(i\) to get, with \(\zeta\)-high probability,

\[
N^{-1} \sum_i G_{ii}^{-1} = -w - wm + \frac{|z|^2}{1 + m} + w[Z] - \frac{|z|^2}{(1 + m)^2} [Z^*_i]
\]

(7.12)

\[
+ N^{-1} \sum_i O \left( \frac{(Z_i^{(i)})^2 + (1/2m)^2}{(1 + m)^2} \right) + |w| O \left( \frac{1}{N} \sum_i m_{\tilde{G}}^{(i,\emptyset)} - m \right) + |m_c|^{-2} O \left( \frac{1}{N} \sum_i (m_{\tilde{G}}^{(i,\emptyset)} - m) \right).
\]

where we have used (6.30) and the bound (6.22). Recall the estimates of \(Z_i\) and \(Z_i^{(i)}\) by \(\Psi\) in (6.27) and (6.24). Hence we have

\[
N^{-1} \sum_i G_{ii}^{-1} = -w - wm + \frac{|z|^2}{1 + m} + \varphi^c O(m_c^{-3}\Psi^2)
\]

(7.13)

\[
+ O(w[Z]) + O(m_c^{-2}[Z^*_i]) + |w| O \left( \frac{1}{N} \sum_i m_{\tilde{G}}^{(i,\emptyset)} - m \right) + |m_c|^{-2} O \left( \frac{1}{N} \sum_i (m_{\tilde{G}}^{(i,\emptyset)} - m) \right).
\]

By (6.30) - (6.60), we have

\[
|G_{ii} - m| \leq O(\varphi^c \Psi) \ll |m_c|,
\]

(7.14)

where \(b \geq 5Q\xi\) and \(Q\xi\) is defined in Lemma C.1. We now perform the expansion \(G_{ii}^{-1} = [(G_{ii} - m) + m]^{-1}\) to have

\[
G_{ii}^{-1} = m^{-1} - \frac{G_{ii} - m}{m^2} + O(\varphi^2 Q|m_c|^{-3}\Psi^2).
\]

Using this approximation in (7.13), we have

\[
m^{-1} + w + wm - \frac{|z|^2}{1 + m} = \varphi^2 Q\xi O(m_c^{-3}\Psi^2) + O(w[Z]) + O(m_c^{-2}[Z])
\]

(7.15)

\[
+ |w| O \left( \frac{1}{N} \sum_i m_{\tilde{G}}^{(i,\emptyset)} - m \right) + |m_c|^{-2} O \left( \frac{1}{N} \sum_i (m_{\tilde{G}}^{(i,\emptyset)} - m) \right).
\]

Using (6.2), we have

\[
\frac{1}{N} \sum_i m_{\tilde{G}}^{(i,\emptyset)} - m = \frac{1}{N} \sum_i m_{\tilde{G}}^{(i,\emptyset)} - m + \frac{C}{Nw}.
\]

Furthermore, with (6.4) we have

\[
m_{\tilde{G}}^{(i,\emptyset)} - m = \frac{1}{N} \left( G_{ii} + \sum_{j \neq i} G_{ji} G_{ij} \right) = \frac{1}{N} \sum_j G_{ji} G_{ij} = O \left( \frac{\text{Im} G_{ii}}{N|\eta|G_{ii}} \right).
\]

(7.17)

The diagonal element \(G_{ii}\) can be estimated by (7.14) so that

\[
\left| \frac{\text{Im} G_{ii}}{N|\eta|G_{ii}} \right| \leq \varphi Q\xi \frac{\text{Im} m_c + \Lambda + \Psi}{N|\eta|m_c} \leq \varphi Q\xi \frac{\Psi^2}{m_c^2}.
\]

Therefore, we have

\[
O \left( \frac{1}{N} \sum_i m_{\tilde{G}}^{(i,\emptyset)} - m \right) \leq O \left( \frac{1}{N} \sum_i m_{\tilde{G}}^{(i,\emptyset)} - m \right) + \frac{C}{N|w|} \leq \varphi Q\xi |m_c|^{-1}\Psi^2 + \frac{C}{N|w|}.
\]

(7.18)
Notice that only the imaginary part of \( m_c \) appears through \( \Psi \) instead of \( m_c \) which can be much bigger near the spectral edge.

We now estimate the last term in (7.10). Notice that \( G^{(i,\emptyset)} \) is the Green function of the matrix \( A^+A \) where \( A = (Y^{(i,\emptyset)})^* \). Then \( m^{(i,i)} \) is the Green function of \( A^{(i,i)} + A^{(i,i)} \) where we have used \( A^{(i,i)} = Y^{(i,i)} \). Thus we can apply (7.17) (which holds for matrices of the form \( A^+A \) with \( A \) not necessarily a square matrix) to get

\[
|m^{(i,\emptyset)}_G - m^{(i,i)}| \lesssim O\left( \frac{\text{Im} G^{(i,\emptyset)}_{ii}}{N\eta} \right).
\]

By (6.31), we have

\[
\text{Im} G^{(i,\emptyset)}_{ii} \lesssim C \left( \text{Im} m_c + \Lambda + \varphi \zeta \Psi \right).
\]

By (6.30) and (6.29),

\[
|G^{(i,\emptyset)}_{ii}| \sim |w|^{-1/2} \sim |m_c|.
\]

These estimates imply that

\[
\frac{1}{N} \sum_i m^{(i,i)} - m \leq \frac{1}{N} \sum_i m^{(i,\emptyset)}_G - m + \frac{1}{N} \sum_i |m^{(i,i)} - m^{(i,\emptyset)}_G| \lesssim \varphi \zeta |m_c|^{-1} \Psi^2.
\]  

(7.19)

Inserting (7.18) and (7.19) into (7.14), we obtain

\[ D(m) \leq O \left( \varphi \zeta \left( \frac{1}{m_c^3} \Psi^2 + N^{-1} \right) + w[\mathcal{Z}] + m_c^{-2}[\mathcal{Z}^*] \right). \]

To conclude Lemma 7.2, we choose \( C_\zeta = 2Q_\zeta \) and it remains to prove \( \frac{1}{m_c} \Psi^2 \geq O(N^{-1}) \). By definition of \( \Psi \) and the fact that \( |m_c| \sim |w|^{-1/2} \) (4.9), this inequality follows from the following property of \( \text{Im} m_c \):

\[
\left| \frac{\text{Im} m_c}{N\eta} \right| \geq O(N^{-1}).
\]

This estimate on \( \text{Im} m_c \) is a direct consequence of (4.2), (4.4), (4.6) and (4.7). This completes the proof of Lemma 7.2 (with \( C_\zeta \) increasing by 1).

We now estimate the averages \( |\mathcal{Z}| \) and \( |\mathcal{Z}^*| \). Our goal is to catch cancellation effects due to the average over the indices \( i \). This is the content of the next lemma, to be proved in next subsection. Clearly this lemma completes the proof of Lemma 7.1.

**Lemma 7.3.** For any \( \zeta > 1 \), there exists \( R_\zeta > 0 \) such that the following statement holds. Suppose for some deterministic number \( \Lambda(w, z) \) (which can depend on \( \zeta \)) we have

\[
\Lambda(w, z) \leq \Lambda(w, z) \ll m_c(w, z)
\]

for \( w \in S(b), b > 5R_\zeta \), in a set \( \Xi \) with \( \mathbb{P}(\Xi^c) \leq e^{-p_N(\log N)^2} \) and \( p_N \) satisfies that

\[
\varphi \leq p_N \leq \varphi^{2\zeta}.
\]

(7.20)

Then there exists a set \( \Xi' \) such that \( \mathbb{P}(\Xi'^c) \leq e^{-p_N} \) and

\[
|\mathcal{Z}| + |\mathcal{Z}^*| \lesssim \varphi \zeta |w|^{1/2} \tilde{\Psi}^2, \quad \text{in} \quad \Xi',
\]

(7.21)

where \( \tilde{\Psi} \) is defined in (7.2).
7.2 Strong bounds on $|Z|$. In this subsection, we prove Lemma 7.3. The main tool is the abstract cancellation Lemma [C.3.1]

We first perform a cutoff for all random variables $X_{ij}$ in $X$ so that $|X_{ij}| \leq N^{10}$. Due to the subexponential decay assumption, the probability of the complement of this event is $e^{-N^n}$, which is negligible.

Define $P_i$ and $P_i^c$ as the operator for the expectation value w.r.t. the $i$-th row and $i$-th column. Let

$$Q_i = 1 - P_i, \quad Q_i^c = 1 - P_i^c$$

With this convention and Lemma [6.3] we can rewrite $Z_i$ and $Z_i^{(i)}$, from Definition [6.4], as

$$Z_i = Q_i (wG_{ii})^{-1}, \quad Z_i^{(i)} = Q_i (wG_{ii}^{(i, \emptyset)})^{-1}.$$

By definition, for any $i, j, U, T$, we know $|G_{ij}^{(i, \emptyset)}| \leq \eta^{-1}$. From the identities of $G_{ii}$ and $G_{ii}^{(i, \emptyset)}$ in Lemma [6.3] and $|X_{ij}| \leq N^C$, we have, for any $1 \leq i \leq N$,

$$|G_{ii}|^{-1} + |G_{ii}^{(i, \emptyset)}|^{-1} \leq N^C. \quad (7.22)$$

Let $D_\zeta = \max\{C_6\zeta + 10, Q_6\zeta + 1\}$ with $C_6$ defined in Lemma [6.1] and $Q_6$ in Lemma [C.3.1]. Then for any fixed $T, U: |T|, |U| \leq p$ there exists a set $\Xi_{T, U}$ with

$$P(\Xi_{T, U}) \geq 1 - e^{-\varphi^{6\zeta + 10}}$$

such that for any $w \in S(b), b > 5D_\zeta$ the following properties hold.

(i) for $w \in S(b)$

$$\Lambda \leq \varphi^{-D_\zeta/4}|w|^{-1/2}, \quad \Psi \leq \varphi^{-2D_\zeta}|w|^{-1/2} \quad (7.23)$$

(ii) for $w \in S(b)$

$$\max_{i,j} |G_{ij}(z) - m_{\epsilon}(z)\delta_{ij}| \leq \varphi^{D_\zeta} \frac{1}{|w|^{1/2}} \left( \frac{|w|^{1/2}}{N\eta} \right)^{1/4}, \quad b > 5D_\zeta. \quad (7.24)$$

(iii) for any $i \neq j$,

$$|(1 - E_{Y_i})y_i^* G^{(i, \emptyset)}(y_i^*)| + |y_i^* G^{(i, \emptyset)}(y_i)| \leq \varphi^{D_\zeta} \Psi \quad (7.25)$$

$$|(1 - E_{Y_j})y_j^{(i)} G^{(i, U)}(y_j^{(i)})^* + |y_j^{(i)} G^{(i, U)}(y_j^{(i)})^*| \leq \varphi^{D_\zeta} \Psi \quad (7.26)$$

(iv) for any $i$ and $T, U: |T| + |U| \leq p$,

$$|G_{ii}^{(i, \emptyset)} - \frac{-1}{w(1 + m^{(i, \emptyset)})}| \leq \varphi^{D_\zeta} \Psi \quad (7.27)$$

Here (i) and (ii) follow from Lemma [6.3]. (iv) follows from [6.32] and the case (iii) with $T = \emptyset = U$ follows from Lemma [C.3.1] and [6.62]. The general case, i.e., $T, U \neq \emptyset$ can be proved similarly using [5.6]. Furthermore, since $|T|, |U| \leq p$ and $p \leq \varphi^{K_6}$, there exists a set $\Xi_0$ with

$$P(\Xi_0) \geq 1 - e^{-\varphi^{2\zeta + \delta}}.$$
such that for any \( w \in \mathbb{S}(b), b > 5D_\zeta \) the above properties \((7.23)-(7.27)\) hold for all \(|T|, |U| \leq p\). The reason is the number of the \( T, U \) satisfying \(|T|, |U| \leq p\) is bounded by \( N^{2p} \leq \varphi^{4c+1}, \) where we have used \((7.20)\).

Since \( \Psi \) is a monotonic in \( \Lambda \), we can replace \( \Psi \) in \((7.23)-(7.27)\) by \( \tilde{\Psi} \) in the set \( \Xi \cap \Xi_0 \). By \((7.20)\), we have \( P[\Xi_0] \leq e^{-pN(\log N)^2} \). For notation simplicity we will use \( \Xi \) for the set \( \Xi \cap \Xi_0 \) from now on. We claim that, for any \( i \in A \subset [1, N], |A| \leq p \), there exist decompositions

\[
Q_A(wG_{ii})^{-1} = Z_{i,A} + Q_A 1(\Xi)\tilde{Z}_{i,A} \tag{7.28}
\]

\[
Q_A \left( wG_{ii}^{(i,\theta)} \right)^{-1} = Z_{i,A} + Q_A 1(\Xi)\tilde{Z}_{i,A} \tag{7.29}
\]

so that \((D.2)\) holds with \( \mathcal{Y} = \left| w \right|^{-1/2} \) and \( \mathcal{X} = \varphi^{D_\zeta+2\zeta} |w|^{1/2} |\tilde{\Psi}|. \) Notice that the condition \( \mathcal{X} < 1 \) follows from \( \tilde{\Lambda} \ll |m_c| \) and \( N\eta \gg \varphi^{5D_\zeta} |m_c| \) if \( w \in \mathbb{S}(b), b > 5D_\zeta \) is large enough. Thus we obtain that

\[
E \| Z \|_p + E \| Z^*_\|_p \leq |w|^{1/2} |p(Cp)^4p(\varphi^{2D_\zeta+4c} |\tilde{\Psi}|^2)^p \tag{7.30}
\]

Choosing \( C_\zeta = 2D_\zeta + 20\zeta \) one can see that \((7.31)\) follows from \((7.20), (7.30)\) and the Markov inequality.

It remains to prove \((7.28)\) and \((7.29)\). We prove \((7.28)\) first. For simplicity, we assume that \( A = \{1, \ldots, |A|\} \). Denote the first \(|A|\) column of \( Y \) by \( a \) so that \( a \) is a \( N \times |A| \) matrix. Similarly, denote by \( B \) the matrix obtained after removing the first \( K \)-columns of \( Y \). Then we have the identity

\[
Y^*Y - w = \left( \begin{array}{cc} a^*a - w & a^*B \\ B^*a & B^*B - w \end{array} \right).
\]

Recall the identity \((6.16)\): for any matrix \( M \),

\[
M(M^*M - w)^{-1}M^* = 1 + w(MM^* - w)^{-1}.
\]

Then we have for \( i, j \in A \)

\[
G_{ij} = \left( \frac{1}{a^*a - w - a^*B(B^*B - w)^{-1}B^*a} \right)_{ij} + \left( \frac{1}{a^*a - w - a^*a^*(1 + w(BB^* - w)^{-1})a} \right)_{ij}
\]

\[
= \left( \frac{1}{-w - w a^*G^{(A,\theta)}a} \right)_{ij}, \quad G^{(A,\theta)} = (BB^* - w)^{-1}. \tag{7.31}
\]

Rewrite

\[
I + a^*G^{(A,\theta)}a = \alpha (I + R), \quad R := \alpha^{-1} \left( a^*G^{(A,\theta)}a + I - \alpha I \right)
\]

where

\[
\alpha := \left( N^{-1} \sum_{j=1}^{N} G_{jj}^{(A,\theta)} + \left| z \right|^2 \frac{-1}{w(1 + m_G^{(A,\theta)})} + 1 \right) = m_G^{(A,\theta)} - \frac{|z|^2}{w(1 + m_G^{(A,\theta)})} + 1
\]

We will prove \( |R| \ll 1 \) with high probability. Using \((3.1), \tilde{\Lambda} \ll m_c \leq \zeta \) and \((6.6)\), we have

\[
\alpha \sim w^{-1/2}, \quad \text{in } \Xi
\]

By \((7.25), (7.27)\) and \((6.6)\), we have

\[
\alpha R_{ii} = (1 - \mathbb{E}_\mathcal{Y})y_i^*G^{(A,\theta)}y_i + \left| z \right|^2 \left( G_{ii}^{(A,\theta)} - \frac{-1}{w(1 + m_G^{(A,\theta)})} \right) = O(\varphi^{D_\zeta} |\tilde{\Psi}|), \quad \text{in } \Xi,
\]

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\[ \alpha R_{ij} = y_j^* G^{(A,b)} y_j \leq O(\varphi^{D_c} \tilde{\Psi}), \quad \text{in } \Xi. \]

Therefore, we have the bound
\[ \|1(\Xi) R\| = O(\varphi^{D_c} \tilde{\Psi} \alpha^{-1}) = O(\varphi^{D_c} |w|^{1/2} \tilde{\Psi}) \leq 1, \quad \|1(\Xi) R^k\| = O(\varphi^{D_c} \tilde{\Psi} \alpha^{-1}|A|^{k-1}), \quad k = 1, 2, \ldots \]
(7.32)

With (7.31) and the definition of \( R \), we have \(-w \alpha G_{ij} = [(I + R)^{-1}]_{ij} \) for \( i, j \in A \). Therefore,
\[ -w G_{ii} \alpha = [(I + R)^{-1}]_{ii} = 1 + \sum_{j=1}^{|A|-1} ((-R)^j)_{ii} + \alpha w \sum_{j \in A} ((-R)^{|A|})_{ij} G_{ji} \]

Then, together with (7.31), (7.32) and \( m_c \sim |w|^{-1/2} \sim \alpha \), we have thus proved that, in \( \Xi \),
\[ -w G_{ii} \alpha = 1 + \sum_{j=1}^{|A|-1} (R^j)_{ii} + O \left( |A| \varphi^{D_c} |w|^{1/2} \tilde{\Psi} \right)^{|A|}, \quad \text{in } \Xi \]

Thus
\[ \frac{-1}{w G_{ii}} = \alpha U_A + O(|w|^{-1/2} |A| \varphi^{D_c} |w|^{1/2} \tilde{\Psi}^{|A|}) \quad \text{(7.33)} \]
\[ = \alpha U_A + O(|w|^{-1/2} |A| \varphi^{D_c+2\zeta} |w|^{1/2} \tilde{\Psi}^{|A|})), \quad \text{in } \Xi \]

where we used \( |A| \leq p \leq \varphi^{2\zeta} \) and \( U_A \) is a linear combination of the following products of \( (R^j)_{ii} \)'s
\[ \prod_k (R^j)_{ii}, \quad 0 \leq \sum_k j_k \leq |A| - 1. \]

Notice we have
\[ Q_A \left( \prod_k \alpha (R^j)_{ii} \right) = 0, \quad \text{(7.34)} \]
provided that \( 0 \leq \sum_k j_k \leq |A| - 1 \). This is because that \( \alpha \) is independent of \( \{y_k : k \in A\} \) and \( R_{ab} \) is independent of \( \{y_k : k \in A, k \neq a, b\} \). Hence there exists \( \ell \in A \) such that \( y_\ell \) does not appear in \( \prod_k \alpha (R^j)_{ii} \) and this proves (7.34). Therefore, we have proved that
\[ Q_A \alpha U_A = 0. \quad \text{(7.35)} \]

Define \( \Omega_A \) as the probability space for the columns \( \{y_k : k \in A\} \) and \( \Omega_{A^c} \) the one for the columns \( \{y_k : k \in A^c\} \). Then the full probability space \( \Omega \) equals to \( \Omega = \Omega_A \times \Omega_{A^c} \). Define \( \pi_{A^c} \) to be the projection onto \( \Omega_{A^c} \) and \( \Xi^* = (\pi_{A^c}^{-1} \cdot \pi_{A^c} \cdot \Xi) \). Then \( 1(\Xi^*) \) is independent of \( \{y_k : k \in A\} \). Hence we can extend (7.35) to
\[ Q_A 1(\Xi^*) \alpha U_A = 0. \]

Let
\[ \tilde{z}_{i,A} = (w G_{ii})^{-1} + 1(\Xi^* \setminus \Xi) \alpha U_A, \quad \Delta_{i,A} = Q_A 1(\Xi) \left[ (w G_{ii})^{-1} + \alpha U_A \right] \]
so that (D.1) is satisfied, i.e.,

\[
Z_{i,A} + Q_A(Ξ) \tilde{Z}_{i,A} = Q_A(Ξ) (wG_{ii})^{-1} + 1(Ξ^*) + Q_A(Ξ) (wG_{ii})^{-1} + 1(Ξ) aU_A
\]

\[
= (Q_A wG_{ii})^{-1} + Q_A [1(Ξ) aU_A + 1(Ξ^*)] aU_A
\]

\[
= (Q_A wG_{ii})^{-1} + Q_A [1(Ξ) aU_A + 1(Ξ^*)] aU_A = (Q_A wG_{ii})^{-1}.
\]

By (7.33), \(|Z_{i,A}| \leq O(|w|^{-1/2}(|A| \varphi_{0c}^{1/2} |w|^{1/2} \Psi)|A|)\) in \(Ξ\). We now prove that

\[
\tilde{Z}_{i,A} = (wG_{ii})^{-1} + 1(Ξ^*) aU_A \leq N^{C|A|}.
\]

By (7.22), we have \((wG_{ii})^{-1} = O(N^C)\). Notice that \(α\) is independent of \(\{y_k : k \in A\}\). Since \(α \sim |w|^{-1/2}\) in \(Ξ\), the same asymptotic holds in \(Ξ^* \setminus Ξ\). By definitions of \(U_A (7.33)\) and \(R\), and the assumption \(X_{ij} = O(N^C)\), we obtain (7.36) and this completes the proof of (7.28). Similarly, we can prove (7.29) and this completes the proof of Lemma 7.3.

A Proof of the properties of \(m_c\) and \(ρ_c\)

In this appendix we are going to prove the lemma 4.1 4.2 and 4.3. We can solve \(m_c\) explicitly by the following formula.

**Lemma A.1** (Explicit expression of \(m_c\)). For any \(E \in \mathbb{R}\), let

\[
A_± := A_±(E, z) := 2E^{3/2} - 9E^{1/2}(1 + 2|z|^2) ± 6\sqrt{3}|z|\sqrt{(λ_+ - E)(E - λ_-)} ± .
\]

Then we have

\[
\lim_{n → 0^+} m_c(E + iη, z) = -\frac{2}{3} - \frac{1}{2^{1/3}\sqrt{E}} \left( \frac{1 - \sqrt{3}}{2} A_+^{1/3}(E, z) + \frac{1 + \sqrt{3}}{2} A_-^{1/3}(E, z) \right),
\]

where we note \(x^{1/3} = \text{sgn}(x)x^{1/3}\). Moreover, for general \(w ∈ \mathbb{C}\), \(m_c(w, z)\) is the analytic extension of \(\lim_{n → 0^+} m_c(E + iη, z)\).

**Proof of Lemma A.1** By definition, \(m_c\) is an analytic function, so we only need to prove (A.1). By definition, \(m_c\) is one of the three solutions of (3.1), and needs to have positive imaginary part. Solving explicitly this degree three polynomial equation proves that there is just one such solution, with the limit (A.1) close to the critical axis. 

Since \(ρ_c(E) = \frac{1}{π} \text{Im} m_c(E + i0^+)\), by (A.1) and \(A_+ ≥ A_-\), we have: for \(0 ≤ E ≤ λ_+\),

\[
ρ_c(E, z) = \frac{1}{2^{1/3}\sqrt{3^{1/2}π}E} \left( A_+^{1/3} - A_-^{1/3} \right) ≥ 0.
\]

With Lemma A.1 and A.2, one can easily prove Proposition 3.1.
Proof of Lemma 4.1. By definition,
\[
\text{Re } m_c(w, z) = \int_R \rho_c(x, z)(x - E) \frac{dx}{(x - E)^2 + \eta^2}
\] (A.3)
so for the first case this implies
\[
0 > \text{Re } m_c(w, z) \geq \int \frac{\rho_c(x, z)}{x - E} dx.
\]
Moreover, recall that \( \alpha = \sqrt{1 + 8|z|^2} \), so (still in the first case)
\[
0 \geq \int \frac{\rho_c(x, z)}{x - E} dx \geq \int m_c(\lambda_+, z) \frac{-2}{\alpha + 3} \geq -\frac{1}{2}.
\]
We also have easily \( |m_c| \sim 1 \) easily from (A.3), we therefore obtained the l.h.s. of (4.2). Similarly, one can prove \( \text{Im } m_c \sim \eta \) thanks to
\[
\text{Im } m_c(w, z) = \eta \int \frac{\rho_c(x, z)}{x - E} dx
\]
and complete the proof for the first case.

For the second case, it is easy to prove (4.3) when \( w = \lambda_+ \), as we did from an explicit calculation. Then one obtains (4.3) by expanding \( m_c \) around \( m_c(\lambda_+, z) \), using (3.1). The estimate (4.4) directly follows from (4.3).

Similarly, for the third case, first \( m_c = \infty \), i.e., \( m_c^{-1} = 0 \) when \( w = 0 \), then one can easily obtain (4.5) in case 3 by solving (3.1) with expanding \( m_c^{-1} \) around \( (m_c(0, z))^{-1} \). The estimate (4.6) directly follows from (4.5). The fourth case follows from
\[
m_c(w, z) = \int \frac{\rho_c(x, z)}{x - w} dx
\] (A.4)
and the properties of \( \rho \) stated in proposition 3.1.

Proof of Lemma 4.2. This is similar to the proof of Lemma 4.1.

Proof of Lemma 4.3. We are going to prove this lemma in the case \( |z| \leq 1 - \tau \), the other cases can be proved similarly. Note first that (4.9) is a consequence of all possible cases in Lemma 4.1.

We now prove (4.10) in the four different cases, which have been classified in Lemma 4.1. In the first case, if additionally \( \eta \sim 1 \), as \( 0 > \text{Re}(m_c) > -1/2 \), the l.h.s. in (4.10) is bounded by \( \text{O}(1) \), which implies (4.10). For the first case if \( \eta \) is small enough, since \( |\text{Re } w| \sim (1 + m_c) \sim 1 \) and \( |\text{Im } (m_c)| \sim \eta \), so
\[
\left| \frac{1}{w(1 + m_c)} \right| \leq C |\text{Im}(w(m_c + 1))| \leq C \text{Im } m_c
\] (A.5)
which gives (4.10) in the first case. In the same way we get (4.10) in the second case, where \( \text{Im } m_c \geq c_0 \). For the third case, using (4.3), one can easily prove (4.10). Finally, the fourth case is simple since the l.h.s. in (4.10) is clearly \( \text{O}(1) \).

We now prove (4.11). Using (4.6) and (4.7), \( (\alpha = \sqrt{1 + 8|z|^2} \) is a real number) we have that, in the cases three and four,
\[
\left| (-1 + |z|^2) \left( m_c - \frac{-2}{3 + \alpha} \right) \left( m_c - \frac{-2}{3 - \alpha} \right) \right| \geq C |\text{Im } m_c|^2 \geq C |w|^{-1}
\] (A.6)

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For case two, using (4.3),

\[
\left| (-1 + |z^2|) \left( m_c - \frac{-2}{3 + \alpha} \right) \left( m_c - \frac{-2}{3 - \alpha} \right) \right| \geq C \left| m_c - \frac{-2}{3 + \alpha} \right| \geq C \left| \frac{\sqrt{\kappa + \eta}}{w} \right|
\]  (A.7)

Note \( m_c(\lambda_+) = -2/(3 + \alpha) \). For case one, with (A.4), it is easy to prove that either \( \text{Im} m_c \sim 1 \) or \( \text{Re} m_c - m_c(\lambda_+) = \text{Re} m_c + 2/(3 + \alpha) \sim 1 \). It implies that \( \left| m_c - \frac{-2}{3 + \alpha} \right| \sim 1 \). This completes the proof. \( \square \)

**B Perturbation theorem**

In this section, we introduce the theorem on the relations between the Green function \( G \) of the matrix \( H \) and the Green function of the minor of the matrix. This theorem was proved in [8]. We first introduce some notations (here we use [] instead of () in [8], since upper index () has been used in the main part of the paper).

**Definition B.1.** Let \( H \) be \( N \times N \) matrix, \( T \subset [1,N] \) and \( H^{[T]} \) be the \( N - |T| \) by \( N - |T| \) minor of \( H \) after removing the \( i \)-th rows and columns index by \( i \in T \). For \( T = \emptyset \), we define \( H^{(\emptyset)} = H \). For any \( T \subset [1,N] \) we introduce the following notations:

\[
G_{ij}^{[T]} := (H^{[T]} - w)^{-1}(i,j), \quad i, j \notin T
\]

\[
Z_{ij}^{[T]} := \sum_{k, \ell \notin T} h_{ik} G_{k\ell}^{[T]} h_{\ell j}
\]

\[
K_{ij}^{[T]} := h_{ij} - w \delta_{ij} - Z_{ij}^{[T]}. \quad (B.1)
\]

The following formulas were proved in Lemma 4.2 from [8].

**Lemma B.2** (Self-consistent perturbation formulas). Let \( T \subset [1,N] \). For simplicity, we use the notation \([i \ T]\) for \([\{i\} \cup T]\) and \([ij \ T]\) for \([\{i,j\} \cup T]\). Then we have the following identities:

(i) For any \( i \notin T \)

\[
G_{ii}^{[T]} = (K_{ii}^{[i \ T]})^{-1}. \quad (B.2)
\]

(ii) For \( i \neq j \) and \( i, j \notin T \)

\[
G_{ij}^{[T]} = -G_{jj}^{[T]} G_{ii}^{[T]} K_{ij}^{[T]} = -G_{ji}^{[T]} G_{jj}^{[T]} K_{ij}^{[T]}. \quad (B.3)
\]

(iii) For any indices \( i, j, k \notin T \) with \( k \notin \{i,j\} \) (but \( i = j \) is allowed)

\[
G_{ij}^{[T]} - G_{ij}^{[k \ T]} = G_{ik}^{[T]} G_{kj}^{[T]} (G_{kk}^{[T]})^{-1}. \quad (B.4)
\]

**C Large deviation estimates.**

In order to obtain the self-consistent equations for the Green functions, we needed the following large deviation estimate.
Lemma C.1 (Large deviation estimate). For any $\zeta > 0$, there exists $Q_\zeta > 0$ such that for $T \subset [1, N]$, $|T| \leq N/2$ the following estimates hold with $\zeta$-high probability:

$$
|Z_i^{(T)}| = \left| (1 - E_{y_i}) \left( y_i^{(T)} G^{(T,i)} y_i^{(T)*} \right) \right| \leq \frac{\varphi Q_\zeta/2}{N} \sqrt{\text{Im} m_{G}^{(T,i)} + |z|^2 \text{Im} G_{ii}^{(T,i)}},
$$  \hspace{1cm} (C.1)

$$
|Z_i^{(T)}| = \left| (1 - E_{y_i}) \left( y_i^{(T)*} G^{(i,T)} y_i^{(T)} \right) \right| \leq \frac{\varphi Q_\zeta/2}{N} \sqrt{\text{Im} m_{G}^{(i,T)} + |z|^2 \text{Im} G_{ii}^{(i,T)}},
$$  \hspace{1cm} (C.2)

Furthermore, for $i \neq j$, we have

$$
\left| (1 - E_{y_i,y_j}) \left( y_i^{(T)} G^{(T,i)} y_j^{(T)*} \right) \right| \leq \frac{\varphi Q_\zeta/2}{N} \sqrt{\text{Im} m_{G}^{(T,i,j)} + |z|^2 \text{Im} G_{ii}^{(T,i,j)} + |z|^2 \text{Im} G_{jj}^{(T,i,j)}},
$$  \hspace{1cm} (C.3)

where

$$
E_{y_i,y_j} \left( y_i^{(T)} G^{(T,i,j)} y_j^{(T)*} \right) = |z|^2 G_{ij}^{(T,i,j)}, \hspace{0.5cm} E_{y_i,y_j} \left( y_i^{(T)*} G^{(i,j,T)} y_j^{(T)} \right) = |z|^2 G_{ij}^{(i,j,T)}. \hspace{1cm} (C.4)
$$

We first recall the following large deviation estimates concerning independent random variables, which were proved in Appendix B of [8].

Lemma C.2. Let $a_i$ ($1 \leq i \leq N$) be independent complex random variables with mean zero, variance $\sigma^2$ and having a uniform subexponential decay

$$
P(|a_i| \geq x\sigma) \leq \theta^{-1} \exp \left( -x^\theta \right), \hspace{1cm} \forall x \geq 1,
$$

with some $\theta > 0$. Let $A_i, B_{ij} \in \mathbb{C}$ ($1 \leq i, j \leq N$). Then there exists a constant $0 < \phi < 1$, depending on $\theta$, such that for any $\xi > 1$ we have

$$
P \left\{ \sum_{i=1}^{N} |a_i A_i| \geq (\log N)^\xi \sigma \left( \sum_{i=1}^{N} |A_i|^2 \right)^{1/2} \right\} \leq \exp \left[ - (\log N)^{\phi\xi} \right], \hspace{1cm} (C.5)
$$

$$
P \left\{ \sum_{i=1}^{N} \sigma^2 B_{ii} \geq (\log N)^\xi \sigma^2 \left( \sum_{i=1}^{N} |B_{ii}|^2 \right)^{1/2} \right\} \leq \exp \left[ - (\log N)^{\phi\xi} \right], \hspace{1cm} (C.6)
$$

$$
P \left\{ \sum_{i\neq j} |a_i B_{ij} a_j| \geq (\log N)^\xi \sigma^2 \left( \sum_{i\neq j} |B_{ij}|^2 \right)^{1/2} \right\} \leq \exp \left[ - (\log N)^{\phi\xi} \right], \hspace{1cm} (C.7)
$$

for any sufficiently large $N \geq N_0$, where $N_0 = N_0(\theta)$ depends on $\theta$.

Proof of Lemma C.2. We will only prove the assertion of this lemma concerning the Green function $G$. Similar statement for $G$ can be proved with the row-column symmetry. From now on, we will only prove all statements concerning $G$ if identical proofs are valid for $G$ and we will not repeat this comment.

We first prove (C.1) by writing

$$
(1 - E_{y_i}) \left( y_i^{(T)} G^{(T,i)} y_i^{(T)*} \right) = (1 - E_{y_i}) |z|^2 G_{ii}^{(T,i)} - (1 - E_{y_i}) \sum_k \left[ z G_{ik}^{(T,i)} X_k + z^* X_k G_{ki}^{(T,i)} \right] + (1 - E_{y_i}) \sum_{jk} X_j G_{jk}^{(T,i)} X_k^* \hspace{1cm} (C.8)
$$

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with $Y = X - zI$. Since $G_{ik}^{(T,i)}$ is independent of $y_i$, the first term on the right hand side vanishes. For any $\zeta > 0$, we apply (C.6) and (C.7) in Lemma C.2 with $\phi \xi = \zeta \log \log N$. Denote $\xi = Q_{\zeta}/2$ and the last term in (C.8) is bounded by
\[
\varphi Q_{\zeta/2} \sqrt{N^{-2} \sum_{jk} |G_{jk}^{(T,i)}|^2} \leq \varphi Q_{\zeta/2} \sqrt{\frac{\text{Im} G_{ij}^{(T,i)}}{N \eta}}
\]
with $\zeta$-high probability. Similarly, with (C.8), the second term on the right hand side is bounded by
\[
\varphi Q_{\zeta/2} |z| \sqrt{N^{-1} \sum_{k} \left| \left( G_{ik}^{(T,i)} \right|^2 + |G_{ki}^{(T,i)}|^2 \right)} \leq \varphi Q_{\zeta/2} \sqrt{\frac{|z|^2 \text{Im} G_{ij}^{(T,i)}}{N \eta}}
\]
The proofs for the other bounds follow from similar arguments.

### D Abstract decoupling lemma

We recall an abstract cancellation Lemma proved in [18].

**Lemma D.1.** Let $\mathcal{I}$ be a finite set which may depend on $N$ and $I_i \subset \mathcal{I}$, $1 \leq i \leq N$.

Let $S_1, \ldots, S_N$ be random variables which depend on the independent random variables $\{x_\alpha, \alpha \in \mathcal{I}\}$. In application, we often take $\mathcal{I} = [1, N]$ and $I_i = \{i\}$.

Recall $\mathbb{E}_i$ denote the conditional expectation with respect to the complement of $\{x_\alpha, \alpha \in I_i\}$, i.e., we integrate out the variables $\{x_\alpha, \alpha \in I_i\}$. Define the commuting projection operators
\[
Q_i = 1 - P_i, \quad P_i = \mathbb{E}_i, \quad P_i^2 = P_i, \quad Q_i^2 = Q_i, \quad [Q_i, P_j] = [P_i, P_j] = [Q_i, Q_j] = 0.
\]

For $A \subset [1, N]$
\[
Q_A := \prod_{i \in A} Q_i, \quad P_A := \prod_{i \in A} P_i
\]
We use the notation
\[
[Z] = \frac{1}{N} \sum_{i=1}^{N} Z_i, \quad Z_i := Q_i S_i.
\]

Let $p$ be an even integer Suppose for some constants $C_0, c_0 > 0$ there is a set $\Xi$ (the "good configurations") so that the following assumptions hold:

(i) (Bound on $Q_A S_i$ in $\Xi$). There exist deterministic positive numbers $X < 1$ and $Y$ such that for any set $A \subset [1, N]$ with $i \in A$ and $|A| \leq p$, $Q_A S_i$ in $\Xi$ can be written as the sum of two random variables
\[
(Q_A S_i) = Z_{i,A} + Q_A 1(\Xi^c)Z_{i,A}, \quad \text{in} \quad \Xi
\]
and
\[
|Z_{i,A}| \leq Y(C_0 X |A|)^{|A|}, \quad |Z_{i,A}| \leq Y N^{c_0 |A|}
\]
(ii) (Crude bound on $S_i$).

$$\max_i |S_i| \leq YN^{C_0}.$$ 

(iii) ($\Xi$ has high probability).

$$\mathbb{P}[\Xi^c] \leq e^{-c_0(\log N)^{3/2}}.$$ 

Then, under the assumptions (i) – (iii), we have

$$\mathbb{E}[Z]^p \leq (Cp)^4 [X^2 + N^{-1}]^{3p} Y^p$$ 

for some $C > 0$ and any sufficiently large $N$.

Roughly speaking, this lemma increase the estimate of $Z_i$ from $X$ to $X^2$ after averaging over $i$.

REFERENCES

[1] Z. D. Bai, Circular law, Ann. Probab. 25 (1997), no. 1, 494–529.
[2] Z. D. Bai and J. Silverstein, Spectral Analysis of Large Dimensional Random Matrices, Mathematics Monograph Series, vol. 2, Science Press, Beijing, 2006.
[3] F. Benaych-Georges and F. Chapon, Random right eigenvalues of Gaussian quaternionic matrices, Random Matrices: Theory and Applications 2 (2012).
[4] A. Borodin and C. D. Sinclair, The Ginibre ensemble of real random matrices and its scaling limits, Comm. Math. Phys. 291 (2009), no. 1, 177–224.
[5] C. Cacciapuoti, A. Maltsev, and B. Schlein, Local Marchenko-Pastur law at the hard edge of sample covariance matrices, to appear in Journal of Mathematical Physics (2012).
[6] E. B. Davies, The functional calculus, J. London Math. Soc. (2) 52 (1995), no. 1, 166–176.
[7] A. Edelman, The probability that a random real Gaussian matrix has $k$ real eigenvalues, related distributions, and the circular law, J. Multivariate Anal. 60 (1997), no. 2, 203–232.
[8] L. Erdős, H.-T. Yau, and J. Yin, Bulk universality for generalized Wigner matrices, Probability Theory and Related Fields 154 (2012), no. 1-2, 341–407.
[9] ____, Rigidity of Eigenvalues of Generalized Wigner Matrices, Adv. Mat. 229 (2012), no. 3, 1435–1515.
[10] P. J. Forrester, Log-gases and random matrices, London Mathematical Society Monographs Series, vol. 34, Princeton University Press, Princeton, NJ, 2010.
[11] P. J. Forrester and T. Nagao, Eigenvalue Statistics of the Real Ginibre Ensemble, Phys. Rev. Lett. 99 (2007).
[12] J. Ginibre, Statistical ensembles of complex, quaternion, and real matrices, J. Mathematical Phys. 6 (1965), 440–449.
[13] V. L. Girko, The circular law, Teor. Veroyatnost. i Primenen. 29 (1984), no. 4, 669–679 (Russian).
[14] F. Götze and A. Tikhomirov, The circular law for random matrices, Ann. Probab. 38 (2010), no. 4, 1444–1491.
[15] A. Guionnet, M. Krishnapur, and O. Zeitouni, The single ring theorem, Ann. of Math. 174 (2011), no. 2, 1189–1217.
[16] M. Mehta, Random matrices, 3rd ed., Pure and Applied Mathematics (Amsterdam), vol. 142, Elsevier/Academic Press, Amsterdam, 2004.
[17] G. Pan and W. Zhou, Circular law, extreme singular values and potential theory, J. Multivariate Anal. 101 (2010), no. 3, 645–656.
[18] N. Pillai and J. Yin, Universality of Covariance matrices, preprint [arXiv:1110.2501] (2011).
[19] M. Rudelson, Invertibility of random matrices: Norm of the inverse, Ann. of Math. 168 (2008), no. 2, 575–600.
[20] M. Rudelson and R. Vershynin, The Littlewood-Offord problem and invertibility of random matrices, Adv. Math. 218 (2008), no. 2, 600–633.
[21] C. D. Sinclair, Averages over Ginibre’s ensemble of random real matrices, Int. Math. Res. Not. IMRN 5 (2007).
[22] T. Tao and V. Vu, Random matrices: the circular law, Commun. Contemp. Math. 10 (2008), no. 2, 261–307.
[23] ______, Random matrices: universality of ESDs and the circular law, Ann. Probab. 38 (2010), no. 5, 2023–2065. With an appendix by Manjunath Krishnapur.