# New scaling of Itzykson-Zuber integrals 

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#### Abstract

We study asymptotics of the Itzykson-Zuber integrals in the scaling when one of the matrices has a small rank compared to the full rank. We show that the result is basically the same as in the case when one of the matrices has a fixed rank. In this way we extend the recent results of Guionnet and Maïda who showed that for the fixed rank scaling, the Itzykson-Zuber integral is given in terms of the Voiculescu's $R$-transform of the full rank matrix. © 2006 Elsevier Masson SAS. All rights reserved.

\section*{Résumé}

Nous étudions les asymptotiques de l'intégrale de Itzykson-Zuber dans le cas où l'une des matrices est de rang petit. Nous prouvons que l'asymptotique obtenue est la même que lorsque la matrice est de rang fixe. De cette manière, nous étendons les récents résultats de Guionnet et Maïda, qui prouvent que en rang fixe, l'intégrale de Itzykson-Zuber s'exprime à l'aide de la $R$-transformée de Voiculescu de la matrice en rang plein. © 2006 Elsevier Masson SAS. All rights reserved.


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## 1. Introduction and the main result

For diagonal matrices $A_{N}, B_{N} \in M_{N}(\mathbb{C})$ we consider the Itzykson-Zuber integral

$$
I_{N}^{(\beta)}\left(A_{N}, B_{N}\right)=\int \mathrm{e}^{N \operatorname{Tr} U A_{N} U^{\star} B_{N}} \mathrm{~d} m_{N}^{\beta}(U)
$$

where $m_{N}^{\beta}$ denotes the Haar measure on the orthogonal group $O_{N}$ when $\beta=1$, on the unitary group $U_{N}$ when $\beta=2$, and on the symplectic group $\operatorname{Sp}(N / 2)$ when $\beta=4$ (in the latter case, $N$ is even). Usually one is interested in the study

[^0]of the asymptotics of the Itzykson-Zuber integral as the size $N$ of the matrices tends to infinity and for this reason one would like to have an insight into the limit
\[

$$
\begin{equation*}
\lim _{N \rightarrow \infty} \frac{1}{N^{2}} \log I_{N}^{(\beta)}\left(A_{N}, B_{N}\right)=: \tilde{I}^{(\beta)}\left(\mu_{A}, \mu_{B}\right) \tag{1}
\end{equation*}
$$

\]

when the spectral measures $\mu_{A_{N}}, \mu_{B_{N}}$ converge weakly towards some probability measures on $\mathbb{R}$. Existence of the limit of (1) was proved by Guionnet and Zeitouni [4].

In [1], the study of another scaling of Itzykson-Zuber integrals was initiated, namely when the rank $M$ of the matrix $A_{N}$ is small compared to the full rank $N$. In order to obtain a non-trivial limit one should consider rather the limit

$$
\begin{equation*}
\lim _{N \rightarrow \infty} \frac{1}{M N} \log I_{N}^{(\beta)}\left(A_{N}, B_{N}\right) . \tag{2}
\end{equation*}
$$

In this paper we shall say that a family of matrices $B_{N} \in M_{N}(\mathbb{C})$ converges in distribution if and only if for all integer $k, N^{-1} \operatorname{Tr}\left(B_{N}^{k}\right)$ tends towards a finite limit as $N \rightarrow \infty$.

The first author proved that if $A_{N}=\operatorname{diag}(t, 0,0, \ldots)$ has rank one and spectral measures of $B_{N}$ converge to some probability measure $\mu_{B}$ then for any integer $k$,

$$
\left.\lim _{N \rightarrow \infty} \frac{1}{N} \frac{\partial^{k}}{\partial t^{k}} \log I_{N}^{(2)}\left(A_{N}, B_{N}\right)\right|_{t=0}=\left.\frac{\mathrm{d}^{k}}{\mathrm{~d} t^{k}} \int_{0}^{t} R_{\mu_{B}}(s) \mathrm{d} s\right|_{t=0}
$$

The function $R$ is called Voiculescu's $R$-transform and is of central importance in free probability theory. We recall its definition in Section 2.

In this paper, the following hypothesis and notation about the sequence of matrices $A_{N} \in M_{N}(\mathbb{C})$ will be used frequently.

Hypothesis 1. $A_{N}$ is diagonal and it has rank $M(N)=\mathrm{o}(N)$. Its non-zero eigenvalues are denoted by $a_{1, N} \geqslant \cdots \geqslant$ $a_{M, N}$.

Using the results presented in our previous article [2] it is possible to prove that the following quite general statement holds true:

Theorem 2. Assume that $B_{N}$ has a limiting distribution $\mu_{B}$ and that $A_{N}$ is uniformly bounded and satisfies Hypothesis 1 . Then for $\beta=1$ and 2 ,

$$
\left.\left|\frac{1}{N M(N)} \frac{\partial^{k}}{\partial t^{k}} \log I_{N}^{(\beta)}\left(t A_{N}, B_{N}\right)\right|_{t=0}-\left.\frac{\beta}{2 M(N)} \sum_{i=1}^{M(N)} \frac{\mathrm{d}^{k}}{\mathrm{~d} t^{k}} \int_{0}^{2 t / \beta} R_{\mu}\left(t a_{i, N}\right) \mathrm{d} s\right|_{t=0} \right\rvert\,=\mathrm{o}(1) .
$$

If $\beta=4$ then

$$
\left.\left|\frac{1}{N M(N)} \frac{\partial^{k}}{\partial t^{k}} \log I_{N}^{(\beta)}\left(t A_{N}, B_{N}\right)\right|_{t=0}+\left.\frac{1}{M(N)} \sum_{i=1}^{M(N)} \frac{\mathrm{d}^{k}}{\mathrm{~d} t^{k}} \int_{0}^{t} R_{\mu}\left(-t a_{i, N}\right) \mathrm{d} s\right|_{t=0} \right\rvert\,=\mathrm{o}(1) .
$$

Observe that it is convenient to separate the case $\beta=4$ from the cases $\beta=1,2$.
We shall not prove this theorem since it is an immediate consequence of Theorem 5.5 of [2]. On the other hand, it has been proved recently in Theorem 7 of [3], that if the $L^{\infty}$-norm of $A_{N}$ is bounded by a constant depending on $B_{N}$ (see Theorem 4 in this paper), under Hypothesis 3 for $B_{N}$, and provided that $M(N)=\mathrm{O}\left(N^{1 / 2-\varepsilon}\right)$ for some $\varepsilon>0$, one has for $\beta=1,2$ :

$$
\begin{equation*}
\left|\frac{1}{N M(N)} \log I_{N}^{(\beta)}\left(A_{N}, B_{N}\right)-\frac{\beta}{2 M(N)} \sum_{i=1}^{M(N)} \int_{0}^{2 t / \beta} R_{\mu}\left(s a_{i, N}\right) \mathrm{d} s\right|=\mathrm{o}(1) . \tag{3}
\end{equation*}
$$

The main result of this paper is Theorem 6. It shows that Eq. (3) still holds true if one replaces the restriction $M(N)=\mathrm{O}\left(N^{1 / 2-\varepsilon}\right)$ by Hypothesis 1 , and that under additional assumption on $B_{N}$, it is enough to assume that the $L^{\infty}$-norm of $A_{N}$ is uniformly bounded by any constant, independently on $B_{N}$.

The techniques used in this paper are elementary, and substantially simplify the proofs of [3]. Better, it becomes possible to express the $R$-transform as a limit of Guionnet-Zeitouni's limits obtained in [4].

Let $b_{1, N} \geqslant \cdots \geqslant b_{N, N}$ be the real eigenvalues of the diagonal matrix $B_{N}$ and let $\mu_{B_{N}}$ be the probability measure $N^{-1} \sum_{i=1}^{N} \delta_{b_{i, N}}$. We make the following hypothesis on the family of matrices $\left(B_{N}\right)_{N \in \mathbb{N}}$ with $B_{N} \in M_{N}(\mathbb{C})$ :

Hypothesis 3. The probability measure $\mu_{B_{N}}$ converges weakly towards a compactly supported probability measure $\mu_{B}$ and that $b_{1, N}$ (resp. $b_{N, N}$ ) converge towards the upper bound $\lambda_{\max }$ (resp. the lower bound $\lambda_{\min }$ ) of the support of $\mu_{B}$.

The Harish-Chandra formula in the case when $\beta=2$ and when the eigenvalues have no multiplicity tells that

$$
I_{N}^{(2)}\left(A_{N}, B_{N}\right)=\frac{\operatorname{det}\left(\mathrm{e}^{\left.N \cdot a_{i, N} b_{j, N}\right)_{i, j=1}^{N}}\right.}{\Delta\left(a_{i, N}\right) \Delta\left(b_{i, N}\right)},
$$

where $\Delta$ denotes Vandermonde determinant. In particular it implies that the value of $I_{N}^{(2)}$ depends only on the eigenvalues of $A_{N}, B_{N}$ therefore the assumption that $A_{N}, B_{N}$ are diagonal can be removed in the case $\beta=2$. However a priori it cannot be removed in the other cases (although it is an open question to the authors' knowledge).

We recall some definitions that can be found in [3]. For a compactly supported probability measure $\mu$, call $H_{\mu}$ its Hilbert transform

$$
H_{\mu}(z)=\int \frac{1}{z-\lambda} \mathrm{d} \mu(\lambda) .
$$

It is invertible under composition in a complex neighborhood of infinity; and let us call $z^{-1}+R_{\mu}$ its inverse. $R_{\mu}$ is called the $R$-transform of Voiculescu.

Let $H_{\text {min }}=\lim _{z \uparrow \lambda_{\text {min }}} H_{\mu_{B}}(z)$ and $H_{\text {max }}=\lim _{z \downarrow \lambda_{\text {max }}} H_{\mu_{B}}(z)$.
Definition. For $\beta \in\{1,2,4\}$, a compactly supported probability measure $\mu$ and a real parameter $t$ we define $f_{\mu}^{(\beta)}(t)$ as follows.

If $\beta=1,2$, the function is

$$
f_{\mu}^{(\beta)}(t)=t v(t)-\frac{\beta}{2} \int \log \left(1+\frac{2}{\beta} t v(t)-\frac{2}{\beta} t \lambda\right) \mathrm{d} \mu(\lambda)
$$

with

$$
v(t)= \begin{cases}R_{\mu}\left(\frac{2}{\beta} t\right) & \text { if } H_{\min } \leqslant \frac{2 t}{\beta} \leqslant H_{\max }, \\ \lambda_{\max }-\frac{\beta}{2 t} & \text { if } \frac{2 t}{\beta}>H_{\max }, \\ \lambda_{\min }-\frac{\beta}{2 t} & \text { if } \frac{2 t}{\beta}<H_{\min }\end{cases}
$$

Observe that in the particular case $t \in\left[H_{\min }, H_{\max }\right]$,

$$
f_{\mu}^{(\beta)}(t)=\int_{0}^{2 t / \beta} R_{\mu}(s t) \mathrm{d} s
$$

In the case $\beta=4$ the function is defined by

$$
f_{\mu}^{(4)}(t)=-f_{\mu}^{(2)}(-t) .
$$

One can prove that this function is continuous with respect to $t$ and also with respect to $\mu_{B}$ for the metric on the measures given by

$$
\begin{align*}
d\left(\mu, \mu^{\prime}\right)= & \left|\lambda_{\max }-\lambda_{\max }^{\prime}\right|+\left|\lambda_{\min }-\lambda_{\min }^{\prime}\right| \\
& +\sup \left\{\int f\left(\mathrm{~d} \mu-\mathrm{d} \mu^{\prime}\right),\|f\|_{\infty}=1 \text { and } f \text { of Lipschitz constant } \leqslant 1\right\} . \tag{4}
\end{align*}
$$

The following result of [3] will be of fundamental importance for us.
Theorem 4. ([3], Theorems 1.2 and 1.6) Assume that the family of matrices $B_{N}$ satisfies Hypothesis 3. For $\beta=1,2$, for any real number $t$,

$$
\lim _{N \rightarrow \infty} N^{-1} \log I_{N}\left(B_{N}, \operatorname{diag}(t, 0, \ldots, 0)\right)=f_{\mu_{B}}^{(\beta)}(t) .
$$

Note that although Guionnet and Maïda in [3] do not handle the case $\beta=4$, it is easy to extend their results to this case and we will take these results for granted.

At some points, we will need to make the following assumption on the matrix $B_{N}$ :
Hypothesis 5. There exists a constant $c>0$ such that

$$
b_{i+1, N}+\frac{1}{c N} \geqslant b_{i, N}
$$

This assumption implies that the distribution of $\mu_{B}$ has an absolutely continuous part with respect to the Lebesgue measure whose density is $\geqslant c$ almost everywhere on its support, and the support should be an interval.

With these preliminaries, we are ready to state our main result:
Theorem 6. Let $\beta \in\{1,2,4\}$.
(1) Let $A_{N}$ be a sequence of matrices satisfying Hypothesis 1, and

$$
H_{\min }<\liminf _{N} a_{N, N} \leqslant \limsup _{N} a_{1, N}<H_{\max } .
$$

Assume that $B_{N}$ satisfies Hypothesis 3. Then

$$
\left|\frac{1}{N M(N)} \log I_{N}^{(\beta)}\left(A_{N}, B_{N}\right)-\frac{\beta}{2 M(N)} \sum_{i=1}^{M(N)} f_{\mu_{B}}^{(\beta)}\left(a_{i, N}\right)\right|=\mathrm{o}(1) .
$$

(2) Let $A_{N}$ be a sequence of matrices satisfying Hypothesis 1. Assume that $B_{N}$ satisfies Hypotheses 3 and 5 . Then

$$
\left|\frac{1}{N M(N)} \log I_{N}^{(\beta)}\left(A_{N}, B_{N}\right)-\frac{\beta}{2 M(N)} \sum_{i=1}^{M(N)} f_{\mu_{B}}^{(\beta)}\left(a_{i, N}\right)\right|=\mathrm{o}(1) .
$$

(3) Let $\mu, \nu$ be compactly supported real probability measures. Let $v$ be supported in a compact subset of $\left(H_{\min }, H_{\max }\right)$. For $a \in[0,1]$, let $v_{a}=a v+(1-a) \delta_{0}$. Then

$$
\lim _{a \rightarrow 0} a^{-1} \tilde{I}^{(\beta)}\left(v_{a}, \mu\right)=\int_{t \in \mathbb{R}} f_{\mu}^{(\beta)}(t) \mathrm{d} \nu(t)
$$

(4) Let $\mu, \nu$ be compactly supported real probability measures. Assume that $\mu$ has a connected support $\left[\lambda_{\min }, \lambda_{\max }\right]$, and that there exists a constant $c>0$ such that

$$
c \cdot \mathrm{~d} x 1_{\left[\lambda_{\min }, \lambda_{\max }\right]} \leqslant \mu .
$$

Then

$$
\lim _{a \rightarrow 0} a^{-1} \tilde{I}^{(\beta)}\left(v_{a}, \mu\right)=\int_{t \in \mathbb{R}} f_{\mu}^{(\beta)}(t) \mathrm{d} \nu(t)
$$

Remark. In Theorem 6, we do not handle the limit $\lim _{a \rightarrow 0} a^{-1} \tilde{I}^{(\beta)}\left(\nu_{a}, \mu\right)$ under the most general assumption that $\mu, \nu$ are compactly supported real probability measures. It is tempting to conjecture that Hypothesis 5 is irrelevant but we could not settle this question.

The next section is devoted to proving Theorem 6.

## 2. Proof of the main result

$I_{N}^{(\beta)}$ satisfies the following obvious translation invariance property:

$$
\begin{equation*}
I_{N}^{(\beta)}(A, B)=\mathrm{e}^{-N x \operatorname{Tr} A} I_{N}^{(\beta)}(A, B+x \mathrm{Id}) \tag{5}
\end{equation*}
$$

therefore there is no loss of generality in assuming that $B_{N}$ is positive.
We shall need the following technical results:
Lemma 7. Let $B \in M_{N}(\mathbb{C})$ be Hermitian, with eigenvalues $b_{1} \geqslant \cdots \geqslant b_{d} \geqslant 0$. Let $\Pi$ be a projector of rank $N-1$ and $B^{\prime}=\Pi В П$ and let $b_{1}^{\prime} \geqslant \cdots \geqslant b_{d}^{\prime}=0$ its eigenvalues. Then one has

$$
b_{1} \geqslant b_{1}^{\prime} \geqslant b_{2} \geqslant \cdots \geqslant b_{d} \geqslant b_{d}^{\prime}=0 .
$$

Proof. Let $V$ be the set of Hermitian projections of rank $n+1-i$ and $V^{\prime}$ be the subset of projections of $V$ dominating $1-\Pi$. According to the "minimax" theorem, $b_{i}=\min _{\pi \in V}\|\pi B \pi\|$, and $b_{i-1}^{\prime}=\min _{\pi \in V^{\prime}}\|\pi B \pi\|$. Since $V^{\prime} \subset V$ this implies that $b_{i+1}^{\prime} \geqslant b_{i}$. Replacing $B$ by $\|B\| \mathrm{Id}-B$ shows that $b_{i} \geqslant b_{i}^{\prime}$, which concludes the proof.

The following lemma is the keystone towards the proof of Theorem 6:
Lemma 8. One has

$$
\begin{align*}
& \prod_{i=1}^{M(N)} \inf _{C_{i, N}} I_{N+1-i}\left(\frac{N}{N+1-i} \operatorname{diag}\left(a_{i, N}, 0, \ldots, 0\right), C_{i, N}\right) \\
& \quad \leqslant I_{N}\left(A_{N}, B_{N}\right) \leqslant \prod_{i=1}^{M(N)} \sup _{C_{i, N}} I_{N+1-i}\left(\frac{N}{N+1-i} \operatorname{diag}\left(a_{i, N}, 0, \ldots, 0\right), C_{i, N}\right), \tag{6}
\end{align*}
$$

where the inf and sup are taken over all diagonal matrices

$$
\begin{equation*}
C_{i, N}=\operatorname{diag}\left(c_{1}^{(i)}, \ldots, c_{N+1-i}^{(i)}\right) \quad \text { such that } b_{k} \leqslant c_{k}^{(i)} \leqslant b_{k+i-1} . \tag{7}
\end{equation*}
$$

Proof. We make this proof by induction on $M(N)$. For $M(N)=1$ there is nothing to prove.
Suppose that the result holds at the level $M(N)-1$. In the case $\beta=2$, we identify $U_{N-1}$ with a subgroup of $U_{N}$ via the morphism

$$
U_{N-1} \rightarrow\left(\begin{array}{cc}
1 & 0 \\
0 & U_{N-1}
\end{array}\right) \subset U_{N}
$$

In the case $\beta=1$ we consider instead the embedding of $O_{N-1}$ into $O_{N}$

$$
O_{N-1} \rightarrow\left(\begin{array}{cc}
1 & 0 \\
0 & O_{N-1}
\end{array}\right) \subset O_{N},
$$

and in the case $\beta=4$, the embedding $\mathrm{SP}_{N-1}$ into $\mathrm{SP}_{N}$

$$
\mathrm{SP}_{N-1} \rightarrow\left(\begin{array}{cc}
\mathrm{Id}_{2} & 0 \\
0 & \mathrm{SP}_{N-1}
\end{array}\right) \subset \mathrm{SP}_{N}
$$

We shall only deal with the case $\beta=2$, the cases $\beta=1,4$ being similar.

Through the above identification, and using the invariance of Haar measure by convolution, we have

$$
\begin{equation*}
I_{N}^{(\beta)}\left(A_{N}, B_{N}\right)=\int_{U \in U_{N}} \int_{V \in U_{N-1}} \exp \left(N \operatorname{Tr} A_{N} V U B_{N} U^{*} V^{*}\right) \mathrm{d} m_{N}^{\beta}(U) \mathrm{d} m_{N-1}^{\beta}(V) . \tag{8}
\end{equation*}
$$

In the remainder of the proof, $N$ is fixed, so in order to lighten the notation we omit the subscript $N$ for the matrices $A_{N}$ and $B_{N}$, and replace $a_{i, N}$ by $a_{i}$ (resp. $b_{i, N}$ by $b_{i}$ ).

Since a unitary conjugation of $A$ by a unitary leaves $I_{N}^{(\beta)}(A, B)$ invariant, one can assume without loss of generality that

$$
A=\operatorname{diag}\left(a_{1}, \ldots, a_{N}\right)=\left(\begin{array}{cc}
a_{1} & 0 \\
0 & A_{-}
\end{array}\right)
$$

with $a_{i} \geqslant a_{j}$ if $i<j$. We shall also adopt the following notation:

$$
\tilde{B}=U B U^{*}=\left(\begin{array}{cc}
\tilde{b}_{11} & \tilde{B}_{1-} \\
\tilde{B}_{-1} & \tilde{B}_{--}
\end{array}\right) .
$$

Observe that $\tilde{B}_{--} \in M_{N-1}(\mathbb{C})$ and $\tilde{b}_{11} \in \mathbb{C}$ are random variables depending on the matrix random variable $U$. With these notations we have

$$
\begin{align*}
I_{N}^{(\beta)}(A, B) & =\int_{U \in U_{N}} \int_{V \in U_{N-1}} \exp \left(N \operatorname{Tr} A V \tilde{B} V^{*}\right) \mathrm{d} m_{N}^{\beta}(U) \mathrm{d} m_{N-1}^{\beta}(V) \\
& =\int_{U \in U_{N}} \int_{V \in U_{N-1}} \exp \left(N a_{1} \tilde{b}_{11}+N \operatorname{Tr} A_{-} V \tilde{B}_{--} V^{*}\right) \mathrm{d} m_{N}^{\beta}(U) \mathrm{d} m_{N-1}^{\beta}(V) \\
& =\int_{U \in U_{N}} \exp \left(N a_{1} \tilde{b}_{11}\right) \int_{V \in U_{N-1}} \exp \left(N \operatorname{Tr} A_{-} V \tilde{B}_{--} V^{*}\right) \mathrm{d} m_{N}^{\beta}(U) \mathrm{d} m_{N-1}^{\beta}(V) \\
& =\int_{U \in U_{N}} \exp \left(N a_{1} \tilde{b}_{11}\right) I_{N-1}^{(\beta)}\left(\frac{N}{N-1} A_{-}, \tilde{B}_{--}\right) \mathrm{d} m_{N}^{\beta}(U) . \tag{9}
\end{align*}
$$

We are in position to apply the recursion hypothesis to

$$
I_{N-1}^{(\beta)}\left(\frac{N}{N-1} A_{-}, \tilde{B}_{--}\right) .
$$

Lemma 7 implies that

$$
\begin{align*}
& \prod_{i=2}^{M(N)} \inf _{C_{i, N}} I_{N+1-i}\left(\frac{N}{N+1-i} \operatorname{diag}\left(a_{i}, 0, \ldots, 0\right), C_{i, N}\right) \\
& \quad \leqslant I_{N-1}^{(\beta)}\left(\frac{N}{N-1} A_{-}, \tilde{B}_{--}\right) \leqslant \prod_{i=2}^{M(N)} \sup _{C_{i, N}} I_{N+1-i}\left(\frac{N}{N+1-i} \operatorname{diag}\left(a_{i}, 0, \ldots, 0\right), C_{i, N}\right), \tag{10}
\end{align*}
$$

where the inf and sup are taken over the same set of matrices $C_{i, N}$ as in (7).
In addition, it is obvious that

$$
\int_{U \in U_{N}} \exp \left(N a_{1} \tilde{b}_{11}\right) \mathrm{d} m_{N}^{\beta}(U)=I_{N}^{(\beta)}\left(\operatorname{diag}\left(a_{1}, 0, \ldots, 0\right), B\right)
$$

The above remark and the inequality (10) plugged into Eq. (9) validate the recursion hypothesis at rank $M(N)$ and complete the proof of the lemma.

We rephrase a continuity statement obtained by [3] suited for our purposes

Lemma 9. Let $K$ be a compact interval contained in $\left(H_{\min }, H_{\max }\right)$. There exists a function $g: \mathbb{R}_{+} \rightarrow \mathbb{R}_{+}$continuous around 0 and such that $g(0)=0$ satisfying the following property:

For all $\varepsilon>0$, for all $t \in K$, and for all Hermitian matrices $B_{N}, B_{N}^{\prime}$ whose distributions have distance d (defined at Eq. (4)) less than $\varepsilon$ to $\mu$,

$$
\left|N^{-1} \log I_{N}\left(B_{N}, \operatorname{diag}(t, 0, \ldots, 0)\right)-N^{-1} \log I_{N}\left(B_{N}^{\prime}, \operatorname{diag}(t, 0, \ldots, 0)\right)\right| \leqslant g(\varepsilon)
$$

Now we are in position to prove the first two statements of Theorem 6.
Proof of Theorem 6, statements (1) and (3). Under Hypothesis 1, one observes that

$$
d\left(\mu_{C_{N, i}}, \mu\right) \rightarrow 0
$$

where $d$ is the distance defined at Eq. (4) and $C_{N, i}$ is any sequence of matrices fulfilling the condition (7).
Taking the logarithm in the inequality of Lemma 8, and using the continuity result quoted in Lemma 9 concludes the proof of statement (1).

A minor modification of the above reasoning also shows statement (3).
Remark. Lemma 8 and the strategy used in the above proof allow us to state and prove a version of Lemma 9 where one replaces $\operatorname{diag}(t, 0, \ldots, 0)$ ) by a matrix $A_{N}$ satisfying Hypothesis 1 . We skip these details.

The following lemma is a direct observation from the definitions, and we skip its proof.
Lemma 10. Let $C_{N, i}$ and $C_{N, i}^{\prime}$ be a pair of matrices fulfilling the condition (7). Under Hypothesis 5, one has

$$
C_{N, i}-\mathrm{Id} \cdot \frac{i}{c N} \leqslant C_{N, i}^{\prime} \leqslant C_{N, i}+\mathrm{Id} \cdot \frac{i}{c N}
$$

Now we are in position to prove statements involving Hypothesis 5.

## Proof of Theorem 6, statements (2) and (4).

We prove statement (4), statement (2) being a straightforward adaptation. Fix $a$ in the open interval ( 0,1 ) and choose sequences of random matrices $A_{N}, B_{N}$ such that $\mu_{A_{N}} \rightarrow v_{a}$ and $\mu_{B_{N}} \rightarrow \mu$.

In agreement with statement (4) of Theorem 6, it is possible to chose $B_{N}$ such that it satisfies Hypotheses 3 and 5. We chose $A_{N}$ such that it satisfies Hypothesis 3 , and such that its rank $M(N) \sim a N$.

Taking the logarithm of the inequality of Lemma 8 yields:

$$
\begin{align*}
& \frac{1}{M N} \sum_{i=1}^{M(N)} \inf _{C_{N, i}} \log I_{N+1-i}\left(\frac{N}{N+1-i} \operatorname{diag}\left(a_{i}, 0, \ldots, 0\right), C_{N, i}\right) \\
& \quad \leqslant \frac{1}{M N} \log I_{N}(A, B) \leqslant \frac{1}{M N} \sum_{i=1}^{M(N)} \sup _{C_{N, i}} \log I_{N+1-i}\left(\frac{N}{N+1-i} \operatorname{diag}\left(a_{i}, 0, \ldots, 0\right), C_{N, i}\right), \tag{11}
\end{align*}
$$

where inf and sup are taken over $C_{N, i}$ as in (7).
Let $C_{i, N}^{+}=\sup \left\{C, C \in C_{i, N}\right\}$ and $C_{i, N}^{-}=\inf \left\{C, C \in C_{i, N}\right\}$. The notions of sup and inf are well-defined because $C_{i, N}$ consists of diagonal matrices. Consider the function

$$
\phi_{i, N}: C \rightarrow \log I_{N+1-i}\left(\frac{N}{N+1-i} \operatorname{diag}\left(a_{i}, 0, \ldots, 0\right), C_{N, i}\right)
$$

Assume that $a_{i}>0$. It is straightforward to check that $\phi_{i, N}$ is increasing on $C_{i, N}$ for any $i, N$ in the sense that $C>C^{\prime}$ implies that $\phi\left(C^{\prime}\right) \geqslant \phi(C)$. Therefore its sup is reached on $C_{i, N}^{+}$and its inf on $C_{i, N}^{-}$. If $a_{i}<0$, equivalent statements hold if one replaces "increasing" by "decreasing" and sup by inf.

Let us introduce

$$
C_{N}^{+}=\max _{i: i \leqslant M(N+i)} C_{N+i, i}^{+} \quad \text { and } \quad C_{N}^{-}=\min _{i: i \leqslant M(N+i)} C_{N+i, i}^{-}
$$

where the maximum is taken pointwise (the matrices are diagonal and the index set is finite).
It is a simple observation that $C_{N}^{+}$and $C_{N}^{-}$have a limiting distribution and that this distribution can be expressed in simple terms as a function of $\mu$ and $a$. Let us call respectively $\mu_{a}^{+}$and $\mu_{a}^{-}$these distributions.

According to Theorem 4, the function

$$
t \rightarrow N^{-1} \log I_{N}\left(\operatorname{diag}(t, 0, \ldots, 0), C_{N}^{+}\right)
$$

is increasing and converges pointwise towards $f_{\mu_{a}^{+}}$as $N$ tends towards $\infty$. Dini's lemma shows that this convergence is uniform and Cesaro's theorem proves that one has

$$
\tilde{I}^{(\beta)}\left(v_{a}, \mu\right) \leqslant \int_{t \in \mathbb{R}_{+}} f_{\mu_{a}^{+}}^{(\beta)}(t) \mathrm{d} v_{a}(t)+\int_{t \in \mathbb{R}_{-}} f_{\mu_{a}^{-}}^{(\beta)}(t) \mathrm{d} v_{a}(t) .
$$

In the same way,

$$
\int_{t \in \mathbb{R}_{+}} f_{\mu_{a}^{-}}^{(\beta)}(t) \mathrm{d} v_{a}(t)+\int_{t \in \mathbb{R}_{-}} f_{\mu_{a}^{+}}^{(\beta)}(t) \mathrm{d} v_{a}(t) \leqslant \tilde{I}^{(\beta)}\left(v_{a}, \mu\right)
$$

In addition, Lemma 10 imply that there exists a constant $K$ depending only on $v$ and $\mu$ such that for all $t$ in the spectrum of $\nu$,

$$
\left|f_{\mu_{a}^{+}}(t)-f_{\mu_{a}^{-}}(t)\right| \leqslant c a K
$$

This concludes the proof of statement (4).

## 3. Concluding remarks

In comparison to [3], our method uses much weaker assumptions on the rank and norm of the matrices. Better, it allows us to state a continuity result about the limit of $I Z$ in a new scaling and thus validate an "inversion of limit" phenomenon.

Last, it shows that Hypotheses 3 and 5 are relevant one to perform computations outside the phase transition zone (that is, to say something about the limit outside the open interval ( $H_{\min }, H_{\text {max }}$ )). This is a substantial improvement to the paper [3] in which nothing was proved at "low temperature" for the finite scaling of rank $M>1$.

Unfortunately, our approach heavily relies on real-number inequalities and we believe that finer estimates, as well as complex valued estimates of [3] cannot be established with our methods.

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