

*Pacific
Journal of
Mathematics*

A PROPERTY OF FREE ENTROPY

S.T. BELINSCHI AND H. BERCOVICI

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We show that the restriction on the uniform norms of approximating matricial microstates can be removed when defining free entropy.

1. Introduction.

Denote by \mathfrak{M}_k the algebra of complex $k \times k$ matrices, and by τ_k the normalized trace on \mathfrak{M}_k , i.e., $\tau_k(A) = \frac{1}{k} \text{Tr}(A)$ for $A \in \mathfrak{M}_k$. Consider for each k a standard Gaussian Hermitian random matrix X_k . Thus, if E denotes expected value, $E\tau_k(X_k) = 0$ and $E\tau_k(X_k^2) = 1$. It was shown by E. Wigner [9] that, as $k \rightarrow \infty$, X_k tends in distribution to a semicircular law, i.e., the limits

$$\mu_p = \lim_{k \rightarrow \infty} E\tau_k(X_k^p)$$

exist, and they can be calculated as

$$\mu_p = \frac{1}{2\pi} \int_{-2}^2 t^p \sqrt{4 - t^2} dt$$

for $p = 1, 2, \dots$. If we have several independent standard Gaussian Hermitian random matrices $(X_k(i))_{i=1}^n$, D. Voiculescu [4] proved that, as $k \rightarrow \infty$, these sets of variables converge in distribution to a free semicircular family. Briefly, this means that given indices $i_j \in \{1, 2, \dots, n\}$ such that $i_j \neq i_{j+1}$ for $j = 1, 2, \dots, m - 1$, and given positive integers p_1, p_2, \dots, p_m , the limit

$$\lim_{k \rightarrow \infty} E\tau_k(X_k(i_1)^{p_1} X_k(i_2)^{p_2} \dots X_k(i_m)^{p_m})$$

exists, and

$$\lim_{k \rightarrow \infty} E\tau_k[(X_k(i_1)^{p_1} - \mu_{p_1})(X_k(i_2)^{p_2} - \mu_{p_2}) \dots (X_k(i_m)^{p_m} - \mu_{p_m})] = 0.$$

It is natural to look for large deviation principles associated with these limit laws. For this purpose (and also with motivation from information theory and statistical physics) Voiculescu introduced in [6] (cf. also [5]) the notion of free entropy. The original definition of free entropy, which will be reviewed below, involves a bound $R > 0$ on the operator norm of approximating matricial microstates, and this may perhaps obscure its significance for large deviations. It is our purpose here to show that this bound can be removed — roughly speaking, one can set $R = \infty$ in the

definition of free entropy. This result applies to other notions of free entropy which appeared subsequently (see for instance [7] for free entropy in the presence of additional variables, [8] for free entropy using an ultrafilter, and [3] for free entropy of a nonselfadjoint variable). We will only provide the proof for the original quantity χ defined in [6], but it should be obvious how the argument applies in the other situations.

It should be noted that a large deviation theorem for Wigner's result has been proved by G. Ben Arous and A. Guionnet [1], where the natural topology of weak convergence of probability measures on the real line is used. The rate function is closely related with free entropy. For several variables, a thorough study of large deviations was undertaken by T. Cabanal Duvillard and A. Guionnet [2]. The rate function they determine is related with another version of free entropy (microstate free).

2. The main result.

For the remainder of this note we fix a positive integer n . We will denote by I the collection of all multiindices $\alpha = (\alpha_1, \alpha_2, \dots, \alpha_m)$ with $m \geq 1$ and $\alpha_j \in \{1, 2, \dots, n\}$ for all $j = 1, 2, \dots, m$. In other words, $I = \bigcup_{m=1}^{\infty} \{1, 2, \dots, n\}^m$. A multiindex of the form $(\alpha, \alpha, \dots, \alpha)$ will also be denoted α^m . We consider the space \mathbb{S} consisting of all families $(\mu(\alpha))_{\alpha \in I}$ of complex numbers indexed by I . The space \mathbb{S} will be endowed with the topology of componentwise convergence.

Consider now a tracial W^* -probability space (\mathfrak{A}, τ) . That is, \mathfrak{A} is a von Neumann algebra, and τ is a normal trace state on \mathfrak{A} . We will write \mathfrak{A}^{sa} for the space of selfadjoint elements of \mathfrak{A} . Given an n -tuple $X = (X_1, X_2, \dots, X_n) \in (\mathfrak{A}^{\text{sa}})^n$, its distribution $\mu_X \in \mathbb{S}$ is defined by

$$\mu_X(\alpha) = \tau(X_\alpha),$$

where $X_\alpha = X_{\alpha_1} X_{\alpha_2} \dots X_{\alpha_m}$ for $\alpha = (\alpha_1, \alpha_2, \dots, \alpha_m) \in I$. This notation applies in particular to n -tuples of selfadjoint matrices in \mathfrak{M}_k . Voiculescu's entropy measures the extent to which the distribution of X can be approximated by distributions of the form μ_A with $A \in (\mathfrak{M}_k^{\text{sa}})^n$. Note first that $\mathfrak{M}_k^{\text{sa}}$ is a real Hilbert space with the Hilbert-Schmidt norm $\|A\|_2 = \text{Tr}(A^2)$, and λ_k will denote the corresponding Lebesgue measure (i.e., a cube whose sides form an orthonormal basis has measure equal to one). On the space $(\mathfrak{M}_k^{\text{sa}})^n$ we have the product measure $\lambda_k^{\otimes n}$.

Given $X \in (\mathfrak{A}^{\text{sa}})^n$, and a neighborhood U of μ_X in \mathbb{S} , we set

$$\Gamma(X; k, U) = \{A \in (\mathfrak{M}_k^{\text{sa}})^n : \mu_A \in U\}.$$

Given in addition a positive number R ,

$$\Gamma_R(X; k, U) = \{A \in \Gamma(X; k, U) : \|A_j\| < R \text{ for all } j\}.$$

We can then define the quantities

$$\chi_R(X; U) = \liminf_{k \rightarrow \infty} \left[\frac{1}{k^2} \log \lambda_k^{\otimes n}(\Gamma_R(X; k, U)) + \frac{n}{2} \log k \right],$$

and

$$\chi_R(X) = \inf_U \chi_R(X; U),$$

where U runs over a neighborhood base of μ_X in \mathbb{S} . Finally, the free entropy is defined as

$$\chi(X) = \sup_{R > 0} \chi_R(X).$$

We also set

$$\chi_\infty(X; U) = \liminf_{k \rightarrow \infty} \left[\frac{1}{k^2} \log \lambda_k^{\otimes n}(\Gamma(X; k, U)) + \frac{n}{2} \log k \right],$$

and $\chi_\infty(X) = \inf_U \chi_\infty(X; U)$. This quantity was introduced in the concluding remarks of [6], where other possible definitions of free entropy are discussed briefly. The inequalities

$$\chi_R(X) \leq \chi(X) \leq \chi_\infty(X)$$

are obvious for $R > 0$, and Proposition 2.4 of [6] states that $\chi_R(X) = \chi(X)$ if R is sufficiently large; $R > \max_j \|X_j\|$ will suffice. Our main result is as follows:

Proposition 2.1. *For every $X \in (\mathfrak{A}^{\text{sa}})^n$ we have $\chi(X) = \chi_\infty(X)$.*

The proof of this result is a refinement of the proof of Proposition 2.4 in [6]. We begin by considering the diffeomorphism f of the real line onto $(-2, 2)$ defined by $f(t) = t$ for $t \in [-1, 1]$, $f(t) = 2 - \frac{1}{t}$ for $t > 1$, and $f(t) = -2 - \frac{1}{t}$ for $t < -1$. Observe that f' does not have any local minimum, and therefore

$$\frac{f(s) - f(t)}{s - t} \geq \min\{f'(s), f'(t)\} \geq f'(s)f'(t)$$

for all s and t . The function $F_n : (\mathfrak{M}_k^{\text{sa}})^n \rightarrow (\mathfrak{M}_k^{\text{sa}})^n$ defined by

$$F_n(A_1, A_2, \dots, A_n) = (f(A_1), f(A_2), \dots, f(A_n))$$

is also differentiable, and we need to estimate the Jacobian determinant $(JF_n)(A)$. Since

$$(JF_n)(A) = (JF_1)(A_1)(JF_1)(A_2) \dots (JF_1)(A_n),$$

it suffices to do this in one variable. As pointed out in [6], if A is a $k \times k$ matrix with eigenvalues $\mu_1, \mu_2, \dots, \mu_k$, we have

$$(JF_1)(A) = \left(\prod_{i \neq j} \frac{f(\mu_i) - f(\mu_j)}{\mu_i - \mu_j} \right) \cdot \prod_{i=1}^k f'(\mu_i).$$

By the estimate for difference quotients shown above,

$$\begin{aligned} (JF_1)(A) &\geq \left(\prod_{i \neq j} f'(\mu_i) f'(\mu_j) \right) \cdot \prod_{i=1}^k f'(\mu_i) \\ &= \prod_{i=1}^k f'(\mu_i)^{2k-1} = \prod_{|\mu_i| > 1} \mu_i^{-2(2k-1)}. \end{aligned}$$

Denoting $\log^+(t) = \max\{\log t, 0\}$, we obtain

$$\begin{aligned} \log(JF_1)(A) &\geq -2(2k-1) \sum_{i=1}^k \log^+ \mu_i \\ &= -2k(2k-1) \frac{1}{k} \sum_{i=1}^k \log^+ \mu_i \\ &= -2k(2k-1) \tau_k(\log^+ |A|). \end{aligned}$$

We have therefore proved the following estimate:

Lemma 2.2. *Given $A = (A_1, A_2, \dots, A_n) \in (\mathfrak{M}_k^{\text{sa}})^n$, we have*

$$(JF_n)(A) \geq \exp \left[-2k(2k-1) \sum_{j=1}^n \tau_k(\log^+ |A_j|) \right].$$

Note for further use that, for a selfadjoint $k \times k$ matrix A , $\tau_k(\log^+ |A|)$ can be estimated in terms of the moments $\tau_k(A^{2p})$, $p \geq 1$. In fact, $\log^+ t = \frac{1}{2p} \log^+ t^{2p} \leq \frac{1}{2p} t^{2p}$, and therefore

$$\tau_k(\log^+ |A|) \leq \frac{1}{2p} \tau_k(A^{2p}).$$

We need one more ingredient.

Lemma 2.3. *Let $X \in (\mathfrak{A}^{\text{sa}})^n$ satisfy $\max_j \|X_j\| < 1$, and let U be a neighborhood of μ_X in \mathbb{S} . There exists a neighborhood V of μ_X in \mathbb{S} such that*

$$F_n(\Gamma(X; k, V)) \subset \Gamma_2(X; k, U) \text{ for all } k.$$

Proof. Clearly it suffices to prove the lemma for neighborhoods of the form

$$U = \{\mu \in \mathbb{S} : |\mu(\alpha) - \tau(X_\alpha)| < \varepsilon\},$$

where $\alpha = (\alpha_1, \alpha_2, \dots, \alpha_m) \in I$ and $\varepsilon > 0$ are fixed. Using the Hölder inequality

$$\|A_\alpha\|_1 \leq \|A_{\alpha_1}\|_m \|A_{\alpha_2}\|_m \dots \|A_{\alpha_m}\|_m,$$

we see that it is sufficient to choose V so that, for all $A \in \Gamma(X; k, V)$, we have $|\tau_k(A_\alpha) - \tau(X_\alpha)| < \varepsilon/2$, $\|A_j\|_m \leq 1$, and $\|A_j - f(A_j)\|_m < \varepsilon/2m$ for

$j = 1, 2, \dots, n$. Choose a number $r < 1$ so that $r > \|X_j\|$ for all j , and choose an even integer $q > m$ such that $r^{q/m} < \varepsilon/2m$. Define next

$$V = \left\{ \mu \in \mathbb{S} : |\mu(\alpha) - \tau(X_\alpha)| < \frac{\varepsilon}{2} \text{ and } |\mu(j^q)| < r^q \text{ for } j = 1, 2, \dots, n \right\};$$

recall that j^q denotes the q -multiindex with all entries equal to j . Consider now $A \in \Gamma(X; k, V)$, and note that the inequalities $|\tau_k(A_\alpha) - \tau(X_\alpha)| < \varepsilon/2$ are obviously satisfied. Also,

$$\|A_j\|_m \leq \|A_j\|_q = \tau(A_j^q)^{1/q} < r < 1.$$

Finally, if $\mu_1, \mu_2, \dots, \mu_k$ are the eigenvalues of A_j ,

$$\begin{aligned} \|A_j - f(A_j)\|_m &\leq \left(\frac{1}{k} \sum_{|\mu_j| \geq 1} |\mu_j|^m \right)^{1/m} \leq \left(\frac{1}{k} \sum_{|\mu_j| \geq 1} |\mu_j|^q \right)^{1/m} \\ &\leq (\tau_k(A_j^q))^{1/m} \leq r^{q/m}, \end{aligned}$$

and this quantity is less than $\varepsilon/2m$. □

Proof of Proposition 2.1. It suffices to prove the proposition in case $\|X_j\| < 1$ for all j . From the results of [6] we know that $\chi_2(X) = \chi(X)$, and clearly $\chi_2(X) \leq \chi_\infty(X)$. To prove the opposite inequality $\chi_2(X) \geq \chi_\infty(X)$, let U be a neighborhood of μ_X in \mathbb{S} , and let V be the neighborhood of μ_X furnished by Lemma 2.3, i.e., $F_n(\Gamma(X; k, V)) \subset \Gamma_2(X; k, U)$ for all $k \geq 1$. Given a positive integer p , we may also assume that $\tau_k(A_j^{2p}) \leq 1$ whenever $A = (A_1, A_2, \dots, A_n) \in \Gamma(X; k, V)$. It follows then from Lemma 2.2 (and the remark following its statement) that

$$(JF_n)(A) \geq \exp \left[-2k(2k-1) \frac{n}{2p} \right]$$

for all $A \in \Gamma(X; k, V)$. Since the function F_n is one-to-one, we deduce that

$$\begin{aligned} \lambda_k^{\otimes n}(\Gamma_2(X; k, U)) &\geq \lambda_k^{\otimes n}(F_n(\Gamma(X; k, V))) \\ &\geq \exp \left[-2k(2k-1) \frac{n}{2p} \right] \lambda_k^{\otimes n}(\Gamma(X; k, V)). \end{aligned}$$

Therefore

$$\begin{aligned} &\frac{1}{k^2} \log \lambda_k^{\otimes n}(\Gamma_2(X; k, U)) + \frac{n}{2} \log k \\ &\geq \frac{1}{k^2} \log \lambda_k^{\otimes n}(\Gamma(X; k, V)) + \frac{n}{2} \log k - \left(2 - \frac{1}{k} \right) \frac{n}{p}, \end{aligned}$$

and as $k \rightarrow \infty$ this yields

$$\chi_2(X; U) \geq \chi_\infty(X; V) - \frac{2n}{p}.$$

Since p is arbitrary, we deduce that $\chi_2(X; U) \geq \chi_\infty(X; V) \geq \chi_\infty(X)$, and the proof is concluded by taking the infimum over U . \square

We remark that a suitable modification of the above proof yields directly that $\chi_\infty(X) = \chi_R(X)$ if $\|X_j\| < R$. One needs an appropriate version of the function f , and that is easily constructed.

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Received May 6, 2002 and revised June 20, 2002. The second author was supported in part by a grant from the National Science Foundation.

INSTITUTE OF MATHEMATICS
 ROMANIAN ACADEMY
 P. O. BOX 1-764
 BUCHAREST RO-70700
 ROMANIA
E-mail address: sbelinsc@indiana.edu

MATHEMATICS DEPARTMENT
 INDIANA UNIVERSITY
 BLOOMINGTON, IN 47405
E-mail address: bercovic@indiana.edu