Accepted Manuscript

Random matrices: Overcrowding estimates for the spectrum

Hoi H. Nguyen



PII:S0022-1236(18)30234-9DOI:https://doi.org/10.1016/j.jfa.2018.06.010Reference:YJFAN 8041To appear in:Journal of Functional Analysis

Received date:21 September 2017Accepted date:20 June 2018

Please cite this article in press as: H.H. Nguyen, Random matrices: Overcrowding estimates for the spectrum, *J. Funct. Anal.* (2018), https://doi.org/10.1016/j.jfa.2018.06.010

This is a PDF file of an unedited manuscript that has been accepted for publication. As a service to our customers we are providing this early version of the manuscript. The manuscript will undergo copyediting, typesetting, and review of the resulting proof before it is published in its final form. Please note that during the production process errors may be discovered which could affect the content, and all legal disclaimers that apply to the journal pertain.

ACCEPTED MANUSCRIPT

RANDOM MATRICES: OVERCROWDING ESTIMATES FOR THE SPECTRUM

HOI H. NGUYEN

ABSTRACT. We address overcrowding estimates for the singular values of random iid matrices, as well as for the eigenvalues of random Wigner matrices. We show evidence of long range separation under arbitrary perturbation even in matrices of discrete entry distributions. In many cases our method yields nearly optimal bounds.

1. INTRODUCTION

1.1. Random iid matrices with subgaussian tails. Consider a random matrix $M = (m_{ij})_{1 \le i,j \le n}$, where m_{ij} are iid copies of a random variable ξ of mean zero and variance one. Let $\sigma_n \le \cdots \le \sigma_1$ be the singular values of M.

An important problem with practical applications is to bound the condition number of M. As the asymptotic behavior of the largest singular value σ_1 is well understood under natural assumption on ξ , the main problem is to study the lower bound of the least singular value σ_n . This problem was first raised by Goldstine and von Neumann [10] well back in the 1940s, with connection to their investigation of the complexity of inverting a matrix.

To answer Goldstine and von Neuman's question, Edelman [7] computed the distribution of the least singular value of Ginibre matrix (where ξ is standard gaussian). He showed that for all fixed $\varepsilon > 0$

$$\mathbf{P}(\sigma_n \le \varepsilon n^{-1/2}) = \int_0^{\varepsilon^2} \frac{1 + \sqrt{x}}{2\sqrt{x}} e^{-(x/2 + \sqrt{x})} \, dx + o(1) = \varepsilon - \frac{1}{3}\varepsilon^3 + O(\varepsilon^4) + o(1).$$

Note that the same asymptotic continues to hold for any $\varepsilon > 0$ which can go to zero with n (see also [22])

$$\mathbf{P}(\sigma_n \le \varepsilon n^{-1/2}) \le \varepsilon. \tag{1}$$

For other singular values of Ginibre matrices, an elegant result by Szarek [24] shows that the σ_{n-k+1} are separated away from zero with an extremely fast rate.

Theorem 1.2. Assume that ξ is standard gaussian, then there exist absolute constants C_1, C_2 such that for all $\varepsilon > 0$, and all $1 \le k \le n$

$$(\frac{C_1}{k}\varepsilon)^{k^2} \le \mathbf{P}(\sigma_{n-k+1} \le \frac{\varepsilon}{\sqrt{n}}) \le (\frac{C_2}{k}\varepsilon)^{k^2}.$$

In what follows ε is always bounded by O(1). Motivated by the universality phenomenon in random matrix theory, we expect similar repulsion bounds for general random matrix ensembles. More specifically, we will

The author is supported by research grant DMS-1600782.

Download English Version:

https://daneshyari.com/en/article/8896573

Download Persian Version:

https://daneshyari.com/article/8896573

Daneshyari.com