

ON THE SPECTRUM OF A RANDOM GRAPH

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The spectrum of a graph is the spectrum of its adjacency matrix.

In the present paper we deal with the asymptotic behaviour of the spectrum of a random graph. We will show that the first eigenvalue is of order n , and then there is a sudden drop: the second largest eigenvalue is $O(n^{\frac{1}{2}+\epsilon})$.

Proposition 1. *Let A be an $n \times n$ symmetric $(0, 1)$ matrix, in which the density of ones is d (i.e. the number of ones is n^2d). Then the largest eigenvalue λ_1 of A satisfies the inequality*

$$nd \leq \lambda_1 \leq n\sqrt{d}.$$

Proof. Let $e = (1, \dots, 1)$ be the n -dimensional all-one vector. Then $\lambda_1 \geq \frac{(e, Ae)}{(e, e)} = \frac{n^2d}{n}$. If λ_i are the eigenvalues of the matrix, then $\sum_{i=1}^n \lambda_i^2 = n^2d$. Hence $\lambda_1 \leq n\sqrt{d}$.

Q.E.D.

Proposition 2. Let $A_n = (a_{ij})$ be an $n \times n$ symmetric matrix where $a_{ii} \equiv 0$, a_{ij} for $i > j$ are independent random variables. Suppose that $P(a_{ij} = 1) = p$, $P(a_{ij} = 0) = q = 1 - p$. If $\lambda_1 = \lambda_1(n)$ is the largest eigenvalue of the matrix A_n , then $\lim_{n \rightarrow \infty} \frac{\lambda_1}{n} = p$ is probability.

Proof. We know that

$$\min_{1 < i < n} \sum_{j=1}^n a_{ij} \leq \lambda_1 \leq \max_{1 < i < n} \sum_{j=1}^n a_{ij}.$$

(Perron – Frobenius theorem). Since for a given value of i the probability $P\left(\left|\frac{1}{n} \sum_{j=1}^n a_{ij} - p\right| > \delta\right)$ is exponentially small,

$$\lim_{n \rightarrow \infty} P\left(\max_{1 < i < n} \left|\frac{1}{n} \sum_{j=1}^n a_{ij} - p\right| > \delta\right) = 0$$

Q.E.D.

Now we are interested in the behaviour of the other eigenvalues. The Wigner theorem – generalized by Ludwig Arnold – describes the overall behaviour of the eigenvalues.

Theorem (Wigner). Let $A_n = (a_{ij})$ be an $n \times n$ symmetric matrix, where a_{ij} for $i \geq j$ are independent random variables. Suppose that a_{ii} are identically distributed with the distribution function G and a_{ij} for $i > j$ are identically distributed with the distribution function H . Assume that $\int x^2 dH(x) = \sigma^2 < \infty$. Denote the empirical distribution function of the eigenvalues of $\frac{1}{\sqrt{n}} A_n$ by $F_n(x) = \frac{1}{n} \sum_{\lambda_i < x \sqrt{n}} 1$ where

λ_i are the eigenvalues of A_n . Then, for arbitrary x we have

$$\lim_{n \rightarrow \infty} F_n(x) = \int_{-\infty}^x f(x) dx \quad \text{in probability}$$

where

$$f(x) = \begin{cases} \frac{1}{2\pi\sigma^2} \sqrt{4\sigma^2 - x^2} & |x| < 2\sigma \\ 0 & \text{otherwise.} \end{cases}$$

For the proof and generalizations see [2], [3] and [4].

This theorem describes only the behaviour of the majority of the eigenvalues, and allows many of them (in fact $o(n)$) to be as large as n . We show now that this is not the case and only the first eigenvalue of a graph is of order n , the rest is near \sqrt{n} . (Actually this happens with any symmetric random matrix if the expectation of the entries is non-zero.)

Proposition 3. *Let A_n be as in Proposition 2. Then for the second largest eigenvalue $\lambda_2 = \lambda_2(n)$ of A_n we have $\lambda_2 = O(n^{\frac{1}{2} + \epsilon})$ in probability.*

Proof. Let us consider the matrix $B_n = P_n A_n P_n$ where P_n is the orthogonal projection parallel to the all-one vector $e_n = (1, \dots, 1)$. By the Courant - Fischer theorem $\lambda_2 \leq \|B_n\|$ where $\|\cdot\|$ denotes the maximal eigenvalue of the matrix. It is easy to see that $B_n = (b_{ij})$, $b_{ij} = a_{ij} - d_i - d_j + d$ where d_i and d are the density of ones in the i -th row and in the whole matrix, respectively. We can write $B_n = C_n + R_n$ where $C_n = (c_{ij})$, $c_{ij} = a_{ij} - p$ and $R_n = (r_{ij})$, $r_{ij} = d_i + d_j - d - p$. So $\|B_n\| \leq \|C_n\| + \|R_n\| \leq \|C_n\| + \|R_n\|_2$, where $\|R_n\|_2^2 = \sum_{i=1}^n \sum_{j=1}^n r_{ij}^2$. We have

$$D^2(d_i + d_j - d) = \frac{1}{n^4} D^2\left(\sum_{k=1}^n \sum_{l=k+1}^n h_{kl} a_{kl}\right) \leq \frac{3pq}{n}$$

where $D^2(\cdot)$ stands for variance. Hence the expectation $E\|R_n\|_2^2 \leq n^2 \frac{3pq}{n}$. The Markov inequality implies that the probability that the maximal eigenvalue of the matrix R_n is larger than $K\sqrt{n}$ is $O\left(\frac{1}{K}\right)$.

For the estimation of $\|C_n\|$ we use the following statement.

Proposition 4. *Let A_n be as in Wigner's theorem, and assume that the expectation of the entries is zero: $\int x dH(x) = 0$. If also all moments of H are finite, then for any $\epsilon > 0$ we have for the maximal eigenvalue $\lambda = \lambda(n)$ of A_n the relation $\lim P(|\lambda| > n^{\frac{1}{2} + \epsilon}) = 0$.*

Proof. It can be seen from [5] and [1] that $E\lambda^{2l} \leq E \sum_{i=1}^n \lambda_i^{2l} = O(n^{l+1})$. Whence

$$P(|\lambda| > n^{\frac{1}{2} + \epsilon}) < n^{-(l+2l\epsilon)} E\lambda^{2l} = O(n^{-(2l\epsilon-1)}) = o(1)$$

if l was chosen large enough.

Q.E.D.

We remark that if we do not assume the existence of higher moments in Proposition 4, then the conclusion $\lambda = o(n)$ in probability still holds.

REFERENCES

- [1] L. Arnold, On the asymptotic distribution of the eigenvalues of random matrices, *J. Math. Analysis Appl.*, 20 (1967), 262-268.
- [2] L. Arnold, On Wigner's semicircle law for the eigenvalues of random matrices, *Z. Wahrscheinlichkeitstheorie verw. Geb.*, 19 (1971), 191-198.
- [3] L. Arnold, Deterministic version of Wigner's semicircle law for the distribution of matrix eigenvalues, *Linear Algebra and its Appl.*, 13 (1976), 185-199.
- [4] V.L. Girko, *Random Matrices*, Kiev 1975 (in Russian).
- [5] E.P. Wigner, Characteristic vectors of bordered matrices with infinite dimensions, *Ann. of Math.*, 62 (1955), 548-564.

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