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# Maximum eigenvalue of symmetric random matrices with dependent heavy tailed entries

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# MAXIMUM EIGENVALUE OF SYMMETRIC RANDOM MATRICES WITH DEPENDENT HEAVY TAILED ENTRIES

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ABSTRACT. This paper deals with symmetric random matrices whose upper diagonal entries are obtained from a linear random field with heavy tailed noise. It is shown that the maximum eigenvalue and the spectral radius of such a random matrix with dependent entries converge to the Frechét distribution after appropriate scaling. This extends a seminal result of [Soshnikov \(2004\)](#) when the tail index is strictly less than one.

## 1. INTRODUCTION

In this article, we study the asymptotic behaviour of the maximum eigenvalue of an  $n \times n$  symmetric random matrix  $A_n$ , whose upper diagonal entries are given by the linear random field

$$(1.1) \quad Y_{k,l} := \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} c_{ij} X_{i+k,j+l}, \quad 1 \leq k \leq l \leq n,$$

where  $\{c_{i,j}\}_{i,j \geq 0}$  is a sequence of real numbers satisfying

$$(1.2) \quad \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} |c_{i,j}|^{\delta} < \infty$$

for some  $\delta \in (0, \alpha)$ , and  $\{X_{i,j}, i, j \in \mathbb{N}\}$  is a family of i.i.d. positive random variables with distribution function  $F$  satisfying

$$(1.3) \quad 1 - F(x) = L(x)x^{-\alpha}, \quad x > 0$$

for some slowly varying function  $L$  and for some  $0 < \alpha < 1$ . It is easy to check, following the arguments of [Cline \(1983\)](#) (see also [Davis and Resnick \(1985\)](#)), that (1.2) ensures the almost sure convergence of the series in (1.1).

Random matrices with heavy tailed entries have generated considerable interest in the recent years; see [Soshnikov \(2004\)](#), [Ben Arous and Guionnet \(2008\)](#), [Belinschi et al. \(2009\)](#), [Auffinger et al. \(2009\)](#), [Davis et al. \(2011\)](#), [Bordenave et al. \(2011\)](#). [Soshnikov \(2004\)](#) investigated the edge behavior of Wigner matrices with

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i.i.d. heavy tailed upper diagonal entries whose distribution function  $F$  satisfies (1.3) for some  $0 < \alpha < 2$ . It was established that in this case, the largest eigenvalue converges to Frechét distribution after scaling by

$$(1.4) \quad b_n := \inf \left\{ x : 1 - F(x) \leq \frac{2}{n(n+1)} \right\}.$$

This result was later extended by Auffinger et al. (2009) to the case  $2 \leq \alpha < 4$  with centered entries. One of the important features of the proof of the above results is that eigenvalues behave similar to the largest entries of the matrix in absolute value, and as a consequence, the point process of the normalized positive eigenvalues converges to a Poisson point process.

This edge behavior is drastically different from the case of  $\alpha > 4$  which is supposed to be governed by the Tracy-Widom law; see, for example, Lee and Yin (2013). For a relaxation of identically distributed condition and further results on edge universality see Bourgade et al. (2013). Few similar results are also known for sample covariance matrices; see, for example, Yin et al. (1988) and Auffinger et al. (2009). Davis et al. (2011) studied the edge behavior for sample covariance matrix  $XX^T$ , where the rows of  $X$  are independent copies of a linear process with heavy tailed noise. Dependence across both rows and columns have also been investigated in the context of bulk asymptotics of sample covariance matrices whose entries have lighter tails; see Hachem et al. (2005) and Pfaffel and Schlemm (2012). For a review of the existing literature on random matrices, we refer the readers to the articles Ben Arous and Guionnet (2011), Erdős and Yau (2012).

Consider the Hilbert space  $l^2 := \{(a_n : n \in \mathbb{Z}) \subset \mathbb{R} : \sum_{n \in \mathbb{Z}} a_n^2 < \infty\}$ . We shall define an operator  $T$  on  $l^2$  as follows. For  $i, j \in \mathbb{Z}$ , let

$$(1.5) \quad T(i, j) := \begin{cases} c_{i+1, j-1}, & i \leq -1, j \geq 1, \\ c_{j+1, i-1}, & j \leq -1, i \geq 1, \\ 0, & \text{otherwise.} \end{cases}$$

$T$  acts on  $l^2$  in the natural way by  $(Ta)_i := \sum_{j=-\infty}^{\infty} T(i, j)a_j$ ,  $i \in \mathbb{Z}$ . By the Cauchy-Schwarz inequality, the operator norm  $\|T\|$  of  $T$  can be bounded above by  $(2 \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} c_{i, j}^2)^{1/2}$ , which is finite because of (1.2). Let  $\rho(A_n)$  and  $\lambda_{\max}(A_n)$  denote the spectral radius (same as the spectral norm in this case) and the maximum eigenvalue of  $A_n$ , respectively. With these notations, we can now state the main result of this paper.

**Theorem 1.1.** *Let  $A_n$  be as above and  $b_n$  be as in (1.4), then for all  $x > 0$ ,*

$$\lim_{n \rightarrow \infty} P(\lambda_{\max}(A_n) \leq \|T\| b_n x) = \lim_{n \rightarrow \infty} P(\rho(A_n) \leq \|T\| b_n x) = e^{-x^{-\alpha}}.$$

Note that we partially recover Theorem 1.1 of Soshnikov (2004) as a consequence of the above result by choosing  $c_{ij} = 1$  when  $i = j$  and  $c_{ij} = 0$  when  $i \neq j$ . However our methods are limited to  $0 < \alpha < 1$  and the case  $1 \leq \alpha < 4$  is still open. We would also like to point that in general, it is very difficult to calculate  $\|T\|$ . In the following special case, it can be computed using Lemma 3.2 below.

**Corollary 1.2.** *Let  $A_n, b_n$  be as in Theorem 1.1 with  $c_{ij} = \alpha_i \beta_j$ , where  $\{\alpha_i\}_{i \geq 0}$  and  $\{\beta_j\}_{j \geq 0}$  satisfy the summability conditions  $\sum_{i=0}^{\infty} |\alpha_i|^\delta < \infty$  and  $\sum_{j=0}^{\infty} |\beta_j|^\delta < \infty$ ,*

respectively, for some  $\delta \in (0, \alpha)$ . Then for all  $x > 0$ ,

$$\lim_{n \rightarrow \infty} P(\lambda_{\max}(A_n) \leq \|\alpha\|_2 \|\beta\|_2 b_n x) = \lim_{n \rightarrow \infty} P(\rho(A_n) \leq \|\alpha\|_2 \|\beta\|_2 b_n x) = e^{-x^{-\alpha}},$$

where  $\|\cdot\|_2$  denotes the  $\ell^2$  norm.

The setup of Davis et al. (2011) can be thought of as a special case of the above example (except that their random matrix  $X$  is not necessarily symmetric) with only nonzero  $\alpha_i$  being  $\alpha_0 = 1$ . They also obtained a similar constant in the limit.

In the rest of the paper, we prove Theorem 1.1. In Section 2, we deal with the finite linear random fields and then using a truncation argument pass on to the infinite case in Section 3. The novel technique used in this paper is the well-known fact from linear algebra that for any matrix norm  $\|\cdot\|$ ,

$$(1.6) \quad \rho(A) = \lim_{r \rightarrow \infty} \left\| \|A^{2r}\| \right\|^{\frac{1}{2r}}.$$

In this paper, for any  $n \times n$  matrix  $M$ , the  $(i, j)^{\text{th}}$  entry is denoted by  $M(i, j)$  and the matrix norms  $\|\cdot\|_\infty$  and  $\max(\cdot)$  are defined by  $\|M\|_\infty = \max_{1 \leq i \leq n} \sum_{j=1}^n |M(i, j)|$  and  $\max(M) = \max_{1 \leq i, j \leq n} |M(i, j)|$ , respectively.

## 2. FINITE LINEAR RANDOM FIELD

In this section, we first show an upper bound of the spectral radius of  $A_n$  using (1.6) and by splitting the matrix  $A_n^{2r}$  into two parts, one having entries only with the  $2r^{\text{th}}$  powers and the other with lower order terms. The matrix with lower order terms has negligible contribution and the main contribution comes from the  $2r^{\text{th}}$  powers. A careful analysis of the coefficient matrices leads to the upper bound. Then we prove a lower bound of maximum eigenvalue using the Rayleigh's characterisation.

For each  $N \geq 1$ , let  $A_n^{(N)}$  be the  $n \times n$  symmetric matrix whose upper diagonal elements are given by the finite linear random field

$$(2.1) \quad Y_{k,l}^{(N)} := \sum_{i=0}^N \sum_{j=0}^N c_{ij} X_{i+k, j+l}, \quad 1 \leq k \leq l \leq n,$$

where  $\{X_{i,j}, i, j \in \mathbb{N}\}$  and  $\{c_{ij}\}_{i,j \geq 0}$  are as in Section 1 and  $C_N$  be the matrix

$$(2.2) \quad C_N := \begin{bmatrix} \mathbf{0} & \widehat{C}_N \\ \widehat{C}_N^T & \mathbf{0} \end{bmatrix},$$

where  $\widehat{C}_N$  is the  $(N+1) \times (N+1)$  matrix whose  $(i, j)^{\text{th}}$  entry is  $c_{N+1-i, N+1-j}$ .

**2.1. Upper bound for spectral radius.** Our first lemma gives an upper bound for the spectral radius of  $A_n^{(N)}$  and is the most crucial step towards proving Theorem 1.1.

**Lemma 2.1.**

$$(2.3) \quad \lim_{n \rightarrow \infty} P \left( \frac{\rho(A_n^{(N)})}{\max_{1 \leq k, l \leq n} |X_{k,l}|} > \rho(C_N) + \varepsilon \right) = 0,$$

for all  $\varepsilon > 0$ .

*Proof.* Fix an integer  $r \geq 1$ . Write  $\left(A_n^{(N)}\right)^{2r} = U_n + V_n$ , where  $U_n$  contains the terms  $X_{k,l}^{2r}$  with their coefficients and  $V_n$  consists of the cross terms. For all  $n > 2N$ , we shall decompose  $U_n$  further into the sum of two matrices  $W_n$  and  $Z_n$ , show that  $W_n$  gives the correct upper bound and  $\|Z_n\|_\infty, \|V_n\|_\infty$  have negligible contributions. These steps are elaborated below.

**Step 1: Decomposition of  $U_n$ .**

Fix  $n > 2N$ . Let  $k, l$  be integers such that  $N < k < l - N \leq n - N$  and  $\tilde{T}$  be the set of all pairs  $(k, l)$  satisfying these inequalities. Define  $\tilde{C}_U$  to be the  $n \times n$  matrix whose  $(N + 1) \times (N + 1)$  submatrix formed by the  $(l - N)^{\text{th}}, (l - N + 1)^{\text{th}}, \dots, l^{\text{th}}$  rows and  $(k - N)^{\text{th}}, (k - N + 1)^{\text{th}}, \dots, k^{\text{th}}$  columns is  $\tilde{C}_N$  defined as above and the other entries are all zero. The condition  $k < l - N$  ensures that  $\tilde{C}_U$  is upper triangular with diagonal entries zero. Set  $\tilde{C}_{k,l} := \tilde{C}_U + \tilde{C}_U^T$ . Thus,  $\tilde{C}_{k,l}$  is a symmetric matrix, and the coefficient of  $X_{k,l}^{2r}$  in the  $(i, j)^{\text{th}}$  entry of  $U_n$  equals  $\tilde{C}_{k,l}^{2r}(i, j)$ . It is easy to see that  $\tilde{C}_{k,l}^{2r}(i, j)$  is zero unless  $(i, j)$  belongs to either  $[k - N, k] \times [l - n, l]$  or  $[l - N, l] \times [k - n, k]$ . Write  $U_n = W_n + Z_n$ , where  $W_n$  contains the entry  $X_{k,l}^{2r}$  (with its coefficient) if and only if  $(k, l) \in \tilde{T}$  and the remaining entries are in  $Z_n$ .

**Step 2: Bound for  $\|W_n\|_\infty$ .**

Define an  $n \times n$  symmetric matrix  $B_n$  whose  $(u, v)^{\text{th}}$  upper diagonal entry is  $X_{u,v}^{2r}$ . Fix  $1 \leq i, j \leq n$ . Clearly, if  $(k, l) \notin S_{ij}$ , the union of  $[i, i + N] \times [j, j + N]$  and  $[j, j + N] \times [i, i + N]$ , then the coefficient of  $X_{k,l}^{2r}$  in  $W_n(i, j)$  is zero. Thus,

$$(2.4) \quad |W_n(i, j)| \leq \sum_{(k,l) \in S_{ij} \cap \tilde{T}} \tilde{C}_{k,l}^{2r}(i, j) X_{k,l}^{2r} \leq \max(C_N^{2r}) \sum_{(k,l) \in S_{ij} \cap \tilde{T}} X_{k,l}^{2r}.$$

Using the above inequality, we have that for each  $i$ ,

$$\sum_{j=i}^n |W_n(i, j)| \leq \max(C_N^{2r}) \sum_{(k,l) \in \tilde{T}} \#\{j \in [i, n] : (k, l) \in S_{ij}\} B_n(k, l).$$

Since  $(k, l) \in S_{ij} \cap \tilde{T}$  and  $i \leq j$  yield  $k - N \leq i \leq k < l - N \leq j \leq l \leq n$ , the above upper bound can be further bounded by

$$\leq \max(C_N^{2r}) \sum_{k=i}^{i+N} \sum_{l=1}^n \#\{j \in [i, n] : (k, l) \in S_{ij}\} B_n(k, l) \leq (N + 1)^2 \max(C_N^{2r}) \|B_n\|_\infty.$$

Similarly, using the symmetry of  $B_n$ ,  $\sum_{j=1}^{i-1} |W_n(i, j)|$  can also be bounded by the same quantity. Thus, we have  $\|W_n\|_\infty \leq 2(N + 1)^2 \max(C_N^{2r}) \|B_n\|_\infty$ , from which using equation (34) in [Soshnikov \(2004\)](#) it follows that

$$\lim_{n \rightarrow \infty} P \left( \frac{\|W_n\|_\infty}{\max_{1 \leq k \leq l \leq n} X_{k,l}^{2r}} > 2(N + 1)^2 \max(C^{2r}) + \varepsilon \right) = 0$$

for all  $\varepsilon > 0$ .

Step 3: Negligibility of  $\|Z_n\|_\infty$ .

Let us now try to upper bound  $\|Z_n\|_\infty$ . For all  $0 \leq k \leq N$ , let  $\bar{C}_k$  be a symmetric  $(k+N+1) \times (k+N+1)$  matrix whose  $(i, j)^{\text{th}}$  upper triangular entry is  $c_{N+1-i, N+k+1-j}$  whenever  $1 \leq i \leq N+1$  and  $k+1 \leq j \leq N+k+1$  (with  $i \leq j$ ) and other upper triangular entries are all zero. By a reasoning similar to the proof of (2.4), it follows that

$$|Z_n(i, j)| \leq \max_{0 \leq k \leq N} \max(\bar{C}_k) \sum_{(u, v) \in \tilde{T}^c} X_{u, v}^{2r},$$

for all  $1 \leq i, j \leq n$ , where  $\tilde{T}^c := \{(u, v) : 1 \leq u \leq v \leq n+N\} \setminus \tilde{T}$ . It is easy to see that the cardinality of  $\tilde{T}^c$  and the number of non-zero entries of  $Z_n$  are both  $O(n)$ . Therefore, there exists a constant  $K$  independent of  $n$  such that

$$\|Z_n\|_\infty \leq \sum_{i=1}^n \sum_{j=1}^n |Z_n(i, j)| \leq Kn \sum_{(u, v) \in \tilde{T}^c} X_{u, v}^{2r},$$

which implies  $\|Z_n\|_\infty = o_p(n^{4r/\alpha})$ .

Step 4: Negligibility of  $\|V_n\|_\infty$ .

We shall show that

$$(2.5) \quad P[\|V_n\|_\infty > b_n^{2r} \varepsilon] \rightarrow 0 \text{ as } n \rightarrow \infty.$$

To this end, note that a typical entry of  $V_n$  has the following form:

$$V(k, l) = \sum'_{i_1, i_2, \dots, i_{2r-1}=1}^n \sum'_{\substack{m_1, \dots, m_{2r}=0 \\ n_1, \dots, n_{2r}=0}}^N \prod_{i=1}^{2r} c_{m_i, n_i} X_{m_1+k, n_1+i_1} \cdots X_{m_{2r}+i_{2r-1}, n_{2r}+l},$$

where  $\sum'$  denotes that the sum is over all those indices which do not give  $2r$ -th power. To simplify the notations we denote by  $\mathcal{I}$  the set of indices  $1 \leq l \leq n$ ,  $\{i_k, 1 \leq k \leq 2r\}$  and  $\{m_i, n_i\}_{1 \leq i \leq 2r}$  such that  $i_k \in \{1, \dots, n\}$  and  $m_i, n_i \in \{0, \dots, N\}$  with the constraint that  $(m_1+k, n_1+i_1) \neq \dots \neq (m_{2r}+i_{2r-1}, n_{2r}+l)$ . Also for any index  $\mathbf{j} \in \mathcal{I}$  we denote the product of random variables  $X_{m_1+k, n_1+i_1} \cdots X_{m_{2r}+i_{2r-1}, n_{2r}+l}$  by  $Z_{\mathbf{j}}$  and product of coefficients  $\prod_{i=1}^{2r} c_{m_i, n_i}$  by  $c_{\mathbf{j}}$ .

Now note that for any  $\mathbf{u} \in \mathcal{I}$ , if  $\{Y_i, 1 \leq i \leq 2r\}$  are i.i.d. random variables with regularly varying tail of index  $-\alpha$  then,  $Z_{\mathbf{u}} \stackrel{d}{=} Y_1^{l_1} \cdots Y_{2r}^{l_{2r}}$  for some  $0 \leq l_i \leq 2r-1$  satisfying  $l_1 + l_2 + \dots + l_{2r} = 2r$ . It is easy to see that  $Z_{\mathbf{u}}$  is regularly varying with index  $-\alpha/k$  for some  $1 \leq k \leq 2r-1$ . Now partition  $\mathcal{I}$  into equivalence classes  $\mathbf{C}_k$  such that for any  $\mathbf{u} \in \mathbf{C}_k$ ,  $Z_{\mathbf{u}}$  is regularly varying  $-\alpha/k$ . Also note that there exists constants  $L_k$  such that,  $|\mathbf{C}_k| \leq L_k N^{4r} n^{2r}$ . As  $\sum_{l=1}^n |V(k, l)| \leq \sum_{\mathbf{u} \in \mathcal{I}} |c_{\mathbf{u}}| Z_{\mathbf{u}}$ , it is enough to show that for some  $\delta > 0$ , we have  $P[\sum_{\mathbf{u} \in \mathcal{I}} |c_{\mathbf{u}}| Z_{\mathbf{u}} > b_n^{2r} \eta] \leq C n^{-(1+\delta)}$ , where  $C$  does not depend on  $k$ .

For  $k \geq 2$  and  $r \geq 2$  denote

$$\beta_{\alpha,r}(k) := \frac{k(4r - (1 + 2r)\alpha)}{\alpha(k - \alpha)} > 0.$$

Let  $\delta_1 = 4r/\alpha$  if  $r = 1$  and also for  $k = 1$ . For  $k \geq 2$  choose  $\delta_k \in (0, \beta_{\alpha,r}(k))$ .

$$\begin{aligned} P \left[ \sum_{\mathbf{u} \in \mathbf{C}_k} |c_{\mathbf{u}}| Z_{\mathbf{u}} > b_n^{2r} \eta \right] &\leq P \left[ \sum_{\mathbf{u} \in \mathbf{C}_k} |c_{\mathbf{u}}| |Z_{\mathbf{u}}| \mathbb{I}(|Z_{\mathbf{u}}| \leq n^{\delta_k}) > b_n^{2r} \epsilon \right] \\ &\quad + P \left[ \max_{\mathbf{u} \in \mathbf{C}_k} Z_{\mathbf{u}} > n^{\delta_k} \right] =: P_1 + P_2. \end{aligned}$$

First we bound  $P_2$ . Choose  $\kappa$  such that  $\kappa > (\frac{2r+1}{\delta_k} \wedge \frac{\alpha}{k})$ .

$$\begin{aligned} P \left[ \max_{\mathbf{u} \in \mathbf{C}_k} |Z_{\mathbf{u}}| > n^{\delta_k} \right] &\leq \sum_{\mathbf{u} \in \mathbf{C}_k} P[|Z_{\mathbf{u}}| > n^{\delta_k}] \leq n^{-\delta_k \kappa} \sum_{\mathbf{u} \in \mathbf{C}_k} E[|Z_{\mathbf{u}}|^{\kappa}] \\ &\leq l_k n^{2r - \delta_k \kappa} = l_k n^{-(1+\epsilon_1)}. \end{aligned}$$

Where we have used that  $|\mathbf{C}_k| \leq L_k N^{4r} n^{2r}$  and  $l_k = L_k N^{4r}$ . Also note that  $\epsilon_1 = \delta_k \kappa - 2r - 1 > 0$ .

Now to bound  $P_1$  choose  $\gamma > (1 \wedge \frac{\alpha}{k})$ . Using Hölder's inequality we get,

$$\begin{aligned} P \left[ \sum_{\mathbf{u} \in \mathbf{C}_k} |c_{\mathbf{u}}| |Z_{\mathbf{u}}| \mathbb{I}(|Z_{\mathbf{u}}| \leq n^{\delta_k}) > b_n^{2r} \epsilon \right] \\ \leq P \left[ \sum_{\mathbf{u} \in \mathbf{C}_k} |c_{\mathbf{u}}|^{\gamma} |Z_{\mathbf{u}}|^{\gamma} \mathbb{I}(|Z_{\mathbf{u}}| \leq n^{\delta_k}) > K_1 b_n^{2\gamma r} n^{2r(\gamma-1)} \right] \\ \leq K_2 b_n^{-2\gamma r} n^{-2r\gamma} E[|Z_{\mathbf{u}}|^{\gamma} \mathbb{I}(|Z_{\mathbf{u}}| \leq n^{\delta_k})]. \end{aligned}$$

By Karamata's theorem (see Theorem 2.1 in [Resnick \(2007\)](#)), we get,

$$E[|Z_{\mathbf{u}}|^{\gamma} \mathbb{I}(|Z_{\mathbf{u}}| \leq n^{\delta_k})] \sim K_3 n^{\gamma \delta_k} n^{-\frac{\delta_k \alpha}{k}}.$$

As  $b_n^{2\gamma r} \sim n^{\frac{4\gamma r}{\alpha}} L_1(n)$  for some slowly varying function  $L_1$  we have

$$b_n^{-2\gamma r} n^{-2r\gamma} E[|Z_i|^{\gamma} \mathbb{I}(|Z_i| \leq n^{\delta_k})] \sim n^{-\frac{4\gamma r}{\alpha} - 2r\gamma + \gamma \delta_k - \frac{\delta_k \alpha}{k}} L_1(n).$$

Now choosing appropriately  $\gamma$  and using the fact that  $\alpha < 1$  it easily follows that,  $P_1 \leq l'_j n^{-(1+\epsilon_2)}$  for some  $\epsilon_2 > 0$ . Note that the constants only depend on  $\epsilon, N, r$ . So there exists  $C$  independent of  $k$  such that,  $P \left[ \sum_{\mathbf{u} \in \mathcal{I}} |c_{\mathbf{u}}| Z_{\mathbf{u}} > b_n^{2r} \eta \right] \leq C n^{-(1+\delta)}$  This establishes (2.5).

#### Step 5: **Combine Steps 1 - 4.**

Combining the above steps, we get that there is  $K > 0$  such that for all  $r \geq 1$  and  $\epsilon > 0$ ,

$$\lim_{n \rightarrow \infty} P \left( \frac{\rho(A_n)}{\max_{1 \leq k \leq l \leq n} |X_{k,l}|} > K^{1/r} \max(C^{2r})^{1/2r} + \epsilon \right) = 0,$$



from which (2.3) follows using (1.6).  $\square$

**2.2. Lower bound for maximum eigenvalue.** Let us now proceed to lower bound  $\lambda_{\max}(A_n^{(N)})$ .

**Lemma 2.2.**

$$\lim_{n \rightarrow \infty} P \left( \frac{\lambda_{\max}(A_n^{(N)})}{|X_{i^*, j^*}|} \leq \lambda_{\max}(C_N) - \varepsilon \right) = 0.$$

*Proof.* Let  $(i^*, j^*) := \arg \max_{1 \leq i \leq j \leq n} |X_{i,j}|$  and  $E_n := \{(i^*, j^*) \in \tilde{T}\}$ , where  $\tilde{T}$  is as defined in Step 1 of the proof of Lemma 2.1. It's easy to see that

$$(2.6) \quad \lim_{n \rightarrow \infty} P(E_n) = 1.$$

Let  $v \in \mathbb{R}^n$  be the unit vector whose  $(i^* - N)^{\text{th}}, (i^* - N + 1)^{\text{th}}, \dots, i^{*\text{th}}$  and  $(j^* - N)^{\text{th}}, (j^* - N + 1)^{\text{th}}, \dots, j^{*\text{th}}$  coordinates are  $1/\sqrt{2(N+1)}$  and the other coordinates are zero. Now it is easy to see that on  $E_n$  we have,

$$|v^T A_n^{(N)} v - |X_{i^*, j^*}| v^T C_N v| \leq \sum_{(k,l) \in ([i^*-N, i^*] \times [j^*-N, j^*]) \setminus \{(i^*, j^*)\}} |u_{kl}| |X_{k,l}|,$$

where  $u_{k,l}$  are fixed numbers independent of  $n$ . Thus by Rayleigh's characterization, we have that for fixed  $\varepsilon > 0$ ,

$$\begin{aligned} & P \left( \frac{\lambda_{\max}(A_n^{(N)})}{|X_{i^*, j^*}|} \leq \lambda_{\max}(C_N) - \varepsilon \right) \\ & \leq P(E_n^c) + P \left( \sum_{(k,l) \in ([i^*-N, i^*] \times [j^*-N, j^*]) \setminus \{(i^*, j^*)\}} |u_{kl}| |X_{k,l}| > \varepsilon |X_{i^*, j^*}| \right), \end{aligned}$$

from which Lemma 2.2 follows using (2.6) since for each  $(k, l)$  in  $[i^* - N, i^*] \times [j^* - N, j^*]$ , the ratio  $X_{k,l}/X_{i^*, j^*}$  converges to zero in probability.  $\square$

### 3. INFINITE LINEAR RANDOM FIELD

Let  $\{X_{i,j}\}$  be as before, that is, regularly varying  $-\alpha$  with  $\alpha \in (0, 1)$  and it is easy to see that  $b_n$  satisfies

$$n^2 P(X_{i,j} > b_n x) \rightarrow x^{-\alpha}.$$

Note that  $b_n = n^{\frac{2}{\alpha}} \tilde{L}(n)$  for slowly varying function  $\tilde{L}(n)$ . The following lemma asserts that if the truncation level of the linear random field goes to infinity then one can approximate the original matrix  $A_n$  with the matrix with truncated entries.

**Lemma 3.1.** *Let  $A_n^{(N)}$  and  $A_n$  are  $n \times n$  symmetric matrices with entries given by (2.1) and (1.1), respectively. Then the following holds,*

$$\lim_{N \rightarrow \infty} \limsup_{n \rightarrow \infty} P(|\rho(A_n) - \rho(A_n^{(N)})| > b_n \eta) = 0,$$

and,

$$\lim_{N \rightarrow \infty} \limsup_{n \rightarrow \infty} P(|\lambda_{\max}(A_n) - \lambda_{\max}(A_n^{(N)})| > b_n \eta) = 0.$$

*Proof.* It is well known that

$$|\lambda_{\max}(A_n) - \lambda_{\max}(A_n^{(N)})| \leq |\rho(A_n) - \rho(A_n^{(N)})| \leq 2 \max_{1 \leq k \leq n} \sum_{l=k}^n |Y_{k,l} - Y_{k,l}^{(N)}|.$$

Therefore, it is enough to show the following two limits

$$(3.1) \quad \lim_{N \rightarrow \infty} \limsup_{n \rightarrow \infty} \sum_{k=1}^n P \left( \sum_{l=k}^n \sum_{i=0}^N \sum_{j=N+1}^{\infty} |c_{i,j}| X_{i+k,j+l} > b_n \frac{\eta}{4} \right) = 0,$$

and

$$(3.2) \quad \lim_{N \rightarrow \infty} \limsup_{n \rightarrow \infty} \sum_{k=1}^n P \left( \sum_{l=k}^n \sum_{i=N+1}^{\infty} \sum_{j=0}^{\infty} |c_{i,j}| X_{i+k,j+l} > b_n \frac{\eta}{4} \right) = 0.$$

To establish (3.1), note that the expression inside the limits can be bounded by

$$\begin{aligned} & \sum_{k=1}^n \sum_{l=k}^n \sum_{i=0}^N \sum_{j=N+1}^{\infty} P \left( |c_{i,j}| X_{i+k,j+l} > b_n \frac{\eta}{4} \right) \\ & + \sum_{k=1}^n \sum_{l=k}^n \sum_{i=0}^N \sum_{j=N+1}^{\infty} \frac{4|c_{i,j}|}{b_n \eta} E \left( X_{i+k,j+l} \mathbb{I}(|c_{i,j}| X_{i+k,j+l} \leq b_n \frac{\eta}{4}) \right). \end{aligned}$$

Since  $\alpha \in (0, 1)$ , by Karamata's theorem each  $E(X_{i+k,j+l} \mathbb{I}(X_{i+k,j+l} \leq t)) \sim \frac{\alpha}{1-\alpha} t(1-F(t))$  as  $t \rightarrow \infty$  and hence by applying Potter bound (see Proposition 2.6 in Resnick (2007)) on both the terms above, we can bound their sum by

$$\leq \sum_{k=1}^n \sum_{l=k}^n \sum_{i=0}^N \sum_{j=N+1}^{\infty} O(n^{-2}) |c_{ij}|^\delta,$$

from which (3.1) follows using (1.2) as the  $O(n^{-2})$  term above does not depend on  $i, j, k$  and  $l$ . Similarly, we can also establish (3.2). This completes the proof.  $\square$

The next lemma evaluates the limit of spectral radius and maximum eigenvalue of the coefficient matrix  $C_N$  defined in (2.2). This result was applied to establish Corollary 1.2 and will be used in the proof of Theorem 1.1 as well.

**Lemma 3.2.** *Let  $C_N$  be as in (2.2) and the coefficients satisfy (1.2) with  $0 < \alpha < 1$ . Then we have,*

- (a) *As  $N \rightarrow \infty$ ,  $\rho(C_N) \rightarrow \|T\|$  where  $T$  is defined in (1.5) and  $\|\cdot\|$  denotes the operator norm.*
- (b)  *$\lambda_{\max}(C_N) = \rho(C_N)$  and hence the convergence of maximum eigenvalue follows from (a).*

*Proof.* Part (b) is obvious from the structure of the matrix  $C_N$  and hence we just briefly sketch a proof of Part (a). To this end, note that by pre and post-multiplying  $C_N$  by a permutation matrix, it is easy to see that for all  $N \geq 1$ ,  $\rho(C_N) = \rho(T_N)$ , where  $T_N(i, j) := T(i, j) \mathbf{1}(|i| \vee |j| \leq N)$  with  $T$  is given by (1.5). Furthermore,  $T_N$

is a self adjoint operator, and hence by Spectral Theorem, its spectral radius and operator norm are equal. As  $N \rightarrow \infty$  we have,

$$\|T_N - T\|^2 \leq \sum_{i,j} (T_N(i,j) - T(i,j))^2 = \sum_{(i,j) \in \mathbb{Z}^2: |i| \vee |j| > N} T(i,j)^2 \rightarrow 0,$$

from which Part (a) follows. □

**3.1. Proof of Theorem 1.1.** We denote by  $Z_\alpha$  a Frechét random variable with parameter  $\alpha$ . Using Proposition 1.11 of Resnick (1987) and Lemmas 2.1, 2.2 and 3.2 above, we have that for each fixed truncation level  $N$  when  $n \rightarrow \infty$ , both  $\lambda_{\max}(A_n^{(N)})/b_n$  and  $\rho(A_n^{(N)})/b_n$  converge weakly to  $\rho(C_N)Z_\alpha$ , which in turn converges to  $\|T\|Z_\alpha$  when  $N \rightarrow \infty$ . Therefore by Lemma 3.1, Theorem 1.1 follows.

**Remark 3.1.** The proofs of Lemma 3.1 and Step 4 of Lemma 2.1 rely heavily on the fact that  $0 < \alpha < 1$ . If these can be established (with additional assumptions if necessary) for higher values of  $\alpha$ , then our result can be extended.

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