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Fluctuations of eigenvalues for random Toeplitz and related matrices

Dang-Zheng Liu∗  Xin Sun†  Zheng-Dong Wang‡

Abstract

Consider random symmetric Toeplitz matrices $T_n = (a_{i-j})_{i,j=1}^n$ with matrix entries $a_j, j = 0, 1, 2, \cdots$, being independent real random variables such that

$$E[a_j] = 0, \quad E[|a_j|^4] = 1 \text{ for } j = 0, 1, 2, \cdots,$$

(homogeneity of 4-th moments)

$$\kappa = E[|a_j|^4],$$

and further (uniform boundedness)

$$\sup_{j \geq 0} E[|a_j|^k] = C_k < \infty \text{ for } k \geq 3.$$

Under the assumption of $a_0 \equiv 0$, we prove a central limit theorem for linear statistics of eigenvalues for a fixed polynomial with degree at least 2. Without this assumption, the CLT can be easily modified to a possibly non-normal limit law. In a special case where $a_j$'s are Gaussian, the result has been obtained by Chatterjee for some test functions. Our derivation is based on a simple trace formula for Toeplitz matrices and fine combinatorial analysis. Our method can apply to other related random matrix models, including Hermitian Toeplitz and symmetric Hankel matrices. Since Toeplitz matrices are quite different from Wigner and Wishart matrices, our results enrich this topic.

Keywords: Toeplitz (band) matrix; Hankel matrix; Random matrices; Linear statistics of eigenvalues; Central limit theorem.

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1 Introduction and main results

Toeplitz matrices appear very often in mathematics and physics and also in plenty of applications, see Grenander and Szegő’s book [14] for a detailed introduction to deterministic Toeplitz matrices. The study of random Toeplitz matrices with independent entries is proposed by Bai in his review paper [2]. Since then, the literature around the asymptotic distribution of eigenvalues for random Toeplitz and related matrices is very large, including the papers of Basak and Bose [4], Bose et al. [6, 7], Bryc et al. [8], Hammond and Miller [15], Kargin [19], Liu and Wang [21], Massey et al. [22], Sen and Virág [24, 25]. We refer to [4, 24, 25] for recent progress. However, the study of fluctuations of eigenvalues for random Toeplitz matrices is quite little, to the best of our knowledge, the only known result comes from Chatterjee [9] in the special case where the matrix entries are Gaussian distributions. In this paper we will derive a central limit theorem (CLT for short) for linear statistics of eigenvalues of random Toeplitz and related matrices.

In the literature fluctuations of eigenvalues for random matrices have been extensively studied. The investigation of central limit theorems for linear statistics of eigenvalues of random matrices dates back to the work of Jonsson [18] on Gaussian Wishart matrices. Similar work for the Wigner matrices was obtained by Sinai and Soshnikov [26]. For further discussion on Wigner (band) matrices and Wishart matrices and their generalized models, we refer to Bai and Silverstein’s book [3], recent papers [1, 9] and the references therein. For another class of invariant random matrix ensembles, Johansson [17] proved a general result which implies CLT for linear statistics of eigenvalues. Recently, Dumitriu and Edelman [13] and Popescu [23] proved that CLT holds for tridiagonal random matrix models.

Another important contribution is the work of Diaconis et al. [12, 11], who proved similar results for random unitary matrices. These results are closely connected to Szegő’s limit theorem (see [16]) for the determinant of Toeplitz matrices with $(j, k)$ entry \( \hat{g}(j-k) \), where \( \hat{g}(k) = \frac{1}{2\pi} \int_0^{2\pi} g(e^{i\theta})e^{-ik\theta} d\theta \), see [10, 5] and references therein for connections between random matrices and Toeplitz determinants. Here we emphasize that Szegő’s limit theorem implies a CLT (see [16]).

Now we turn to our model. The matrix of the form \( T_n = (a_{i-j})_{i,j=1}^n \) is called a Toeplitz matrix. If we introduce the Toeplitz or Jordan matrices \( B = (\delta_{i+1,j})_{i,j=1}^n \) and \( F = (\delta_{i,j+1})_{i,j=1}^n \), respectively called the “backward shift” and “forward shift” because of their effect on the elements of the standard basis \( \{e_1, \cdots, e_n\} \) of \( \mathbb{R}^n \), then an \( n \times n \) matrix \( T \) can be written in the form

\[
T = \sum_{j=0}^{n-1} a_{-j} B^j + \sum_{j=1}^{n-1} a_j F^j
\]

if and only if \( T \) is a Toeplitz matrix where \( a_{-n+1}, \cdots, a_0, \cdots, a_{n-1} \) are complex numbers [20]. It is worth emphasizing that this representation of a Toeplitz matrix is of vital importance as the starting point of our method. The “shift” matrices \( B \) and \( F \) exactly present the information of the traces.

Consider a Toeplitz band matrix as follows. Given a band width \( b_n < n \), let

\[
\eta_{ij} = \begin{cases} 
1, & |i-j| \leq b_n; \\
0, & \text{otherwise}.
\end{cases}
\]

Then a Toeplitz band matrix is

\[
T_n = (\eta_{ij} a_{i-j})_{i,j=1}^n.
\]
Moreover, the Toeplitz band matrix $T_n$ can also be rewritten in the form

$$T_n = \sum_{j=0}^{b_n} a_{-j}B^j + \sum_{j=1}^{b_n} a_jF^j = a_0I_n + \sum_{j=1}^{b_n} (a_{-j}B^j + a_jF^j),$$

(1.4)

where $I_n$ is the identity matrix. Obviously, a Toeplitz matrix can be considered as a band matrix with the bandwidth $b_n = n - 1$. In this paper, the basic model under consideration consists of $n \times n$ random symmetric Toeplitz band matrices

$$T_n = (\eta_{ij} a_{i-j})_{i,j=1}^n$$

in Eq. (1.3). We assume that $a_j = a_{-j}$ for $j = 1, 2, \cdots$, and $\{a_j\}_{j=1}^\infty$ is a sequence of independent real random variables such that

$$E[a_j] = 0, \quad E[|a_j|^2] = 1 \text{ for } j = 1, 2, \cdots, \quad \quad \text{(1.5)}$$

(homogeneity of 4-th moments)

$$\kappa = E[|a_j|^4], \quad \quad \text{(1.6)}$$

and further (uniform boundedness)

$$\sup_{j \geq 1} E[|a_j|^k] = C_k < \infty \text{ for } k \geq 3. \quad \quad \text{(1.7)}$$

In addition, we also assume $a_0 \equiv 0$ (we will explain in Remarks 1.3 and 5.5 below!) and the bandwidth $b_n \to \infty$ but $b_n/n \to b \in [0, 1]$ as $n \to \infty$.

Set $A_n = \frac{T_n}{\sqrt{b_n}}$, a linear statistic of eigenvalues $\lambda_1, \cdots, \lambda_n$ of $A_n$ is a function of the form

$$\frac{1}{n} \sum_{j=1}^{n} f(\lambda_j), \quad \quad \text{(1.8)}$$

where $f$ is some fixed function. In particular, when $f(x) = x^p$ we write

$$\omega_p = \frac{\sqrt{b_n}}{n} \sum_{j=1}^{n} (\lambda_p^j - E[\lambda_p^j]). \quad \quad \text{(1.9)}$$

These $\omega_p, \ p = 2, 3, \ldots$, are our main objects. Note that $\frac{1}{n} \sum_{j=1}^{n} (\lambda_p^j - E[\lambda_p^j])$ converges weakly to zero as $n \to \infty$, moreover under the condition

$$\sum_{j=1}^{\infty} \frac{1}{b_n^2} < \infty$$

we have a strong convergence, see [4, 19, 21]. We remark that the fluctuations

$$\frac{1}{n} \sum_{j=1}^{n} \lambda_p^j - \frac{1}{n} \sum_{j=1}^{n} E[\lambda_p^j]. \quad \quad \text{(1.10)}$$

are of order $\frac{1}{\sqrt{b_n}}$ while those for Wigner matrices are of order $\frac{1}{\sqrt{n}}$. Thus Toeplitz case is more like the case of classical central limit theorem, especially when $b_n = n$. This shows that the correlations between eigenvalues for Toeplitz matrices are much weaker than those for Wigner matrices (another different phenomenon is that the limiting distribution for random Toeplitz matrices has unbounded support, see [8, 15]). The potential reasons for this phenomenon may come from the fact that the order of the number of independent variables is $O(n)$ for Toeplitz matrices while it is $O(n^2)$ for Wigner matrices. On the other hand, the eigenvalues of random Toeplitz matrices are obviously not independent, so it is different from the case of CLT for independent variables.
In the special case that the matrix entries $a_j$ are Gaussian distributions, by using his notion of “second order Poincaré inequalities” Chatterjee in [9] proved the following theorem:

**Theorem** ([9], Theorem 4.5) Consider the Gaussian Toeplitz matrices $T_n = (a_{i-j})_{i,j=1}^n$, i.e. $a_j = a_{-j}$ for $j = 1, 2, \ldots$, and $\{a_j\}_{j=0}^\infty$ is a sequence of independent standard Gaussian random variables. Let $p_n$ be a sequence of positive integers such that $p_n = o(\log n/\log \log n)$. Let $A_n = T_n/\sqrt{n}$, then, as $n \to \infty$,

$$\frac{\text{tr}(A_n^{p_n}) - E[\text{tr}(A_n^{p_n})]}{\sqrt{\text{Var}(\text{tr}(A_n^{p_n}))}} \text{ converges in total variation to } N(0, 1).$$

The CLT also holds for $\text{tr}(f(A_n))$, when $f$ is a fixed nonzero polynomial with nonnegative coefficients.

The author remarked that the theorem above is only for Gaussian Toeplitz matrices based on the obvious fact: considering the function $f(x) = x$, CLT may not hold for linear statistics of non-Gaussian Toeplitz matrices. The author also remarked that the theorem above says nothing about the limiting formula of the variance $\text{Var}(\text{tr}(A_n^{p_n}))$. However, we assert that CLT holds for a test function $f(x) = x^{2p}$ even for non-Gaussian Toeplitz matrices. When $f(x) = x^{2p+1}$ the fluctuation is Gaussian if and only if the diagonal random variable $a_0$ is Gaussian. Moreover, if we suppose $a_0 \equiv 0$, we can obtain CLT for any fixed polynomial test functions. On the other hand, we can calculate the variance in terms of integrals associated with pair partitions. Unfortunately, our method fails to deal with the test function $f(x) = x^{2p}$, where $p_n$ depends on $n$.

Our study is inspired by the work of Sinai and Soshnikov [26], but new ideas are needed since the structure of Toeplitz matrices is quite different from that of Wigner matrices. More specifically, we first expand $\text{tr}(T_n^p)$ into $\sum_{i=1} \sum_{l_1,j_1} a_j a_{j_1} I_{l_1,j_1}$ where the “balanced” vector $J = (j_1, \ldots, j_p) \in \{-b_n, \ldots, b_n\}^p$ with $\sum_{j=1}^p j_j = 0$, $a_j = \prod_{l=1}^i a_l$ and $I_{l_1,j_1} = \prod_{k=1}^i \chi_{l_1,n}(i + \sum_{k=1}^p j_1)$. So we can reduce estimates about higher moments of $\text{tr}(T_n^p)$ to the combinatorial analysis of correlated “balanced” vectors. In addition, our method can apply to other related random matrix models, including Hermitian Toeplitz matrices and Hankel matrices.

Now we state the main theorem as follows.

**Theorem 1.1.** Let $T_n$ be a real symmetric ((1.5)-(1.7)) random Toeplitz band matrix with bandwidth $b_n$, where $b_n/n \to b \in [0, 1]$ and $b_n \to \infty$ as $n \to \infty$. Set $A_n = T_n/\sqrt{b_n}$ and

$$\omega_p = \frac{\sqrt{b_n}}{n} (\text{tr}(A_n^{p_n}) - E[\text{tr}(A_n^{p_n})]),$$

(1.11)

For every $p \geq 2$, we have

$$\omega_p \longrightarrow N(0, \sigma_p^2)$$

(1.12)

in distribution as $n \to \infty$. Moreover, for a given polynomial

$$Q(x) = \sum_{j=2}^p q_j x^j$$

(1.13)

with degree $p \geq 2$, set

$$\omega_Q = \frac{\sqrt{b_n}}{n} (\text{tr}(Q(A_n)) - E[\text{tr}(Q(A_n))]),$$

(1.14)

we also have

$$\omega_Q \longrightarrow N(0, \sigma_Q^2)$$

(1.15)

in distribution as $n \to \infty$. Here the variances $\sigma_p^2$ and $\sigma_Q^2$ will be given in section 4.
From the proof of our main theorem, we can easily derive an interesting result concerning product of independent variables whose subscripts satisfy certain “balance” condition. When $p = 2$, it is a direct result from the classical central limit theorem. Here we state it but omit its proof, see Remark 5.6 for detailed explanation.

**Corollary 1.2.** Suppose that $a_j = a_{-j}$ for $j = 1, 2, \ldots$, and $\{a_j\}_{j=1}^\infty$ is a sequence of independent random variables satisfying the assumptions (1.5)-(1.7).

For every $p \geq 2$,

$$\frac{1}{k^{1/2}} \sum_{0 \neq j_1, \ldots, j_p = -k}^k \left( \prod_{l=1}^p a_{j_l} - E \prod_{l=1}^p a_{j_l} \right) \delta_{0, \sum_{l=1}^p j_l}$$  \hspace{1cm} (1.16)

converges in distribution to a Gaussian distribution $N(0, \sigma_N^2)$ as $k$ tends to infinity. Here the variance $\sigma^2_p$ corresponds to the case $b = 0$ as in section 4.

**Remark 1.3** (on the diagonal entry $a_0$). Rewrite $A_n(a_0) = A_n(0) + \frac{\omega_n}{\sqrt{n}} I_n$, where $A_n(0)$ denotes the matrix with $a_0 = 0$. It is easy to see that

$$\omega_p(a_0) = \omega_p(0) + \frac{\text{tr}(A_{n-1}^p(0))}{n} (a_0 - E[a_0]) + O(b_n^{-1/2}),$$  \hspace{1cm} (1.17)

which converges in distribution to the distribution of $\sigma_n^p + pM_{p-1}(a_0 - E[a_0])$. Here $n$ is the standard normal distribution, independent of $a_0$, and $M_{p-1}$ is the $(p-1)$-th moment in Theorem 3.1. Since $M_p = 0 \iff p$ is odd, if $a_0$ is neither a constant a.s. nor Gaussian then the fluctuation is not Gaussian for odd $p$.

The remaining part of the paper is arranged as follows. Some integrals associated with pair partitions are defined in section 2. Sections 3, 4 and 5 are devoted to the proof of Theorem 1.1. We will extend our main result to other models closely related to Toeplitz matrices in section 6.

## 2 Integrals associated with pair partitions

In order to calculate the moments of the limiting distribution and the limiting covariance matrix of random variables $\omega_p$, we first review some basic combinatorical concepts, and then define some integrals associated with pair partitions.

**Definition 2.1.** Let the set $[n] = \{1, 2, \ldots, n\}$.

(1) We call $\pi = \{V_1, \ldots, V_r\}$ a partition of $[n]$ if the blocks $V_j$ $(1 \leq j \leq r)$ are pairwise disjoint, non-empty subsets of $[n]$ such that $[n] = V_1 \cup \ldots \cup V_r$. The number of blocks of $\pi$ is denoted by $|\pi|$, and the number of elements of $V_j$ is denoted by $|V_j|$.

(2) Without loss of generality, we assume that $V_1, \ldots, V_r$ have been arranged such that $s_1 < s_2 < \cdots < s_r$, where $s_j$ is the smallest number of $V_j$. Therefore we can define the projection $\pi(i) = j$ if $i$ belongs to the block $V_j$; furthermore for two elements $p, q$ of $[n]$ we write $p \sim_{\pi} q$ if $\pi(p) = \pi(q)$.

(3) The set of all partitions of $[n]$ is denoted by $\mathcal{P}(n)$, and the subset consisting of all pair partitions, i.e. all $|V_j| = 2$, $1 \leq j \leq r$, is denoted by $\mathcal{P}_2(n)$. Note that $\mathcal{P}_2(n)$ is an empty set if $n$ is odd.

(4) Suppose $p, q$ are positive integers and $p + q$ is even, we denote a subset of $\mathcal{P}_2(p+q)$ by $\mathcal{P}_2(p, q)$, which consists of such pair partitions $\pi$: there exists $1 \leq i \leq p < j \leq p + q$ such that $i \sim_{\pi} j$ (we say that there is one crossing match in $\pi$).

(5) When $p$ and $q$ are both even, we denote a subset of $\mathcal{P}(p + q)$ by $\mathcal{P}_2,4(p, q)$, which consists of such partitions $\pi = \{V_1, \ldots, V_r\}$ satisfying

(i) $|V_j| = 2$, $1 \leq j \leq r, j \neq i$ and $|V_i| = 4$ for some $i$. 

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(ii) $V_j \subseteq \{1,2,\ldots,p\}$ or $\{p+1,p+2,\ldots,p+q\}$ for $1 \leq j \leq r, j \neq i$.

(iii) two elements of $V_i$ come from $\{1,2,\ldots,p\}$ and the other two come from $\{p+1,p+2,\ldots,p+q\}$.

For other cases of $p$ and $q$, we assume $P_{2,4}(p,q)$ is an empty set.

Now we define several types of definite integrals associated with $\pi \in P_{2,4}(p,q)$ or $\pi \in P_{2,4}(p,q)$. For reader’s convenience, we suggest omitting them for the moment and refer to them when needed in sections 3 and 4. Let the parameter $b \in [0,1]$.

First, for $\pi \in P_{2,4}(p,q)$ we set

$$\epsilon_\pi(i) = \begin{cases} 1, & i \text{ is the smallest number of } \pi^{-1}(i); \\ -1, & \text{otherwise.} \end{cases}$$

(2.1)

For every pair partition $\pi \in P_{2,4}(p,q)$, we construct a projective relation between two groups of unknowns $y_1,\ldots,y_{p+q}$ and $x_1,\ldots,x_{p+q}$ as follows:

$$\epsilon_\pi(i) y_i = \epsilon_\pi(j) y_j = x_{\pi(i)}$$

(2.2)

whenever $i \sim_\pi j$. Thus, we have an identical equation

$$\sum_{j=1}^{p+q} y_j = 0.$$  

(2.3)

For $x_0, y_0 \in [0,1]$ and $x_1,\ldots,x_{p+q} \in [-1,1]$, we define two kinds of integrals with Type I by

$$f_I^+ (\pi) = \int_{[0,1]^2 \times [-1,1]} \prod_{j=1}^{p} \chi_{[0,1]}(x_0+b \sum_{i=1}^{j} y_i) \prod_{j'=p+1}^{p+q} \chi_{[0,1]}(y_0+b \sum_{i=p+1}^{j'} y_i) d y_0 \prod_{l=0}^{p+q} d x_l$$

(2.4)

and

$$f_I^- (\pi) = \int_{[0,1]^2 \times [-1,1]} \prod_{j=1}^{p} \chi_{[0,1]}(x_0+b \sum_{i=1}^{j} y_i) \prod_{j'=p+1}^{p+q} \chi_{[0,1]}(y_0-b \sum_{i=p+1}^{j'} y_i) d y_0 \prod_{l=0}^{p+q} d x_l.$$  

(2.5)

Here $\delta$ is the Dirac function and $\chi$ is the indicator function. Note that $\sum_{j=1}^{p} y_j \neq 0$ by the definition of $P_{2,4}(p,q)$, therefore the above integrals are multiple integrals in $(1+q+p+1)$ variables.

Next, for $\pi = \{V_1,\ldots,V_{p+q-1}\} \in P_{2,4}(p,q)$ (denoting the block with four elements by $V_j$), we set for $\pi(k) \neq i$

$$\tau_\pi(k) = \begin{cases} 1, & k \text{ is the smallest number of } \pi^{-1}(\pi(k)); \\ -1, & \text{otherwise.} \end{cases}$$

(2.6)

while for $\pi(k) = i$

$$\tau_\pi(k) = \begin{cases} 1, & k \text{ is the smallest or largest number of } \pi^{-1}(\pi(k)); \\ -1, & \text{otherwise.} \end{cases}$$

(2.7)
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To every partition $\pi \in \mathcal{P}_{2,4}(p,q)$, we construct a projective relation between two groups of unknowns $y_1, \ldots, y_{p+q}$ and $x_1, \ldots, x_{\frac{p+q}{2} - 1}$ as follows:

$$\tau_\pi(i) y_i = \tau_\pi(j) y_j = x_{\pi(i)} \quad (2.8)$$

whenever $i \sim \pi j$. Then two kinds of integrals with Type II are defined respectively by

$$f^-_{II}(\pi) = \int_{[0,1]^2 \times [-1,1]} \frac{1}{2} \sum_{j=1}^p \chi_{[0,1]}(x_0 + b \sum_{i=1}^j y_i) \prod_{j'=p+1}^{p+q} \chi_{[0,1]}(y_0 + b \sum_{i=p+1}^{j'} y_i) \, dy_0 \prod_{l=0}^{\frac{p+q}{2} - 1} \, dx_l$$

and

$$f^+_{II}(\pi) = \int_{[0,1]^2 \times [-1,1]} \frac{1}{2} \sum_{j=1}^p \chi_{[0,1]}(x_0 + b \sum_{i=1}^j y_i) \prod_{j'=p+1}^{p+q} \chi_{[0,1]}(y_0 - b \sum_{i=p+1}^{j'} y_i) \, dy_0 \prod_{l=0}^{\frac{p+q}{2} - 1} \, dx_l. \quad (2.9)$$

3 Mathematical expectation

In this section, we will review some results about the moments of the limiting distribution of eigenvalues in [21], for the convenience of the readers and further discussion.

**Theorem 3.1.** $E[\frac{1}{n} \text{tr}(A_n^{2k})] = M_{2k} + o(1)$ and $E[\frac{1}{n} \text{tr}(A_n^{2k+1})] = o(1)$ as $n \to \infty$ where

$$M_{2k} = \sum_{\pi \in \mathcal{P}_{2,4}(2k)} \int_{[0,1]^2 \times [-1,1]^k} 2 \chi_{[0,1]}(x_0 + b \sum_{i=1}^j y_i) \prod_{j'=p+1}^{p+q} \chi_{[0,1]}(y_0 - b \sum_{i=p+1}^{j'} y_i) \prod_{l=0}^{\frac{p+q}{2} - 1} \, dx_l. \quad (3.1)$$

Let us first give a lemma about traces of Toeplitz band matrices. Although its proof is simple, it is very useful in treating random matrix models closely related to Toeplitz matrices.

**Lemma 3.2.** For Toeplitz band matrices $T_{i,n} = (\eta_{j,i-j})_{i,j=1}^n$ with the bandwidth $b_n$ where $a_{i,-n+1}, \ldots, a_{i,-1}$ are complex numbers and $l = 1, \ldots, p$, we have the trace formula

$$\text{tr}(T_{i,n} \cdots T_{p,n}) = \sum_{j} a_j I_j \delta_{0, \sum_{l=1}^p j_l}, \quad p \in \mathbb{N}. \quad (3.2)$$

Here $J = (j_1, \ldots, j_p) \in \{-b_n, \ldots, b_n\}^p$, $a_J = \prod_{l=1}^p a_{l,j_l}$, $I_J = \prod_{l=1}^p \chi_{[1,n]}(i + \sum_{l=1}^k j_l)$ and the summation $\sum_J$ runs over all possibilities that $J \in \{-b_n, \ldots, b_n\}^p$.

**Proof.** For the standard basis $\{e_1, \ldots, e_n\}$ of the Euclidean space $\mathbb{R}^n$, we have

$$T_{p,n} e_i = \sum_{j=0}^{b_n} a_{p,j} B^j e_i + \sum_{j=1}^{b_n} a_{p,j} F^j e_i = \sum_{j=-b_n}^{b_n} a_{p,j} \chi_{[1,n]}(i + j) e_{i+j}. \quad \text{Repeating } T_{i,n}^t's \text{ effect on the basis, we have}$$

$$T_{1,n} \cdots T_{p,n} e_i = \sum_{j_1, \ldots, j_p = -b_n}^{b_n} \prod_{l=1}^p a_{l,j_l} \prod_{k=1}^p \chi_{[1,n]}(i + \sum_{l=1}^k j_l) e_{i + \sum_{l=1}^k j_l}.$$
We will mainly use the above trace formula in the case where \( T_{1,n} = \cdots = T_{p,n} = T_n \). Since \( a_0 \equiv 0 \), from the Kronecker delta symbol in the trace formula of (3.2), it suffices to consider these \( J = (j_1, \ldots, j_p) \in \{ \pm 1, \ldots, \pm b_n \}^p \) with the addition of \( \sum_{k=1}^{b_n} j_k = 0 \). We remark that \( a_0 \equiv 0 \) is not necessary to Theorem 3.1. In fact it is sufficient to ensure Theorem 3.1 if all finite moments of random variable \( a_0 \) exist and its expectation is zero.

**Definition 3.3.** Let \( J = (j_1, \ldots, j_p) \in \{ \pm 1, \ldots, \pm b_n \}^p \), we say \( J \) is balanced if \( \sum_{k=1}^{b_n} j_k = 0 \).

The component \( j_u \) of \( J \) is said to be coincident with \( j_v \) if \( |j_u| = |j_v| \) for \( 1 \leq u \neq v \leq p \).

For \( J \in \{ \pm 1, \ldots, \pm b_n \}^p \), we construct a set of numbers with multiplicities

\[
S_J = \{|j_1|, \ldots, |j_p|\}. \tag{3.3}
\]

We call \( S_J \) the projection of \( J \).

The balanced \( J \)'s can be classified into three categories.

Category 1 (denoted by \( \Gamma_1(p) \)): \( J \) is said to belong to category 1 if each of its components is coincident with exactly one other component of the opposite sign. It is obvious that \( \Gamma_1(p) \) is an empty set when \( p \) is odd.

Category 2 (\( \Gamma_2(p) \)) consists of all those vectors such that \( S_J \) has at least one number with multiplicity 1.

Category 3 (\( \Gamma_3(p) \)) consists of all other balanced vectors in \( \{ \pm 1, \ldots, \pm b_n \}^p \). For \( J \in \Gamma_3(p) \), either \( S_J \) has one number with multiplicity at least 3, or each of \( S_J \) has multiplicity 2 but at least two of the components are the same, which are denoted respectively by \( \Gamma_3(1) \) and \( \Gamma_3(2) \).

We are now ready to prove Theorem 3.1.

**Proof of Theorem 3.1.** By Lemma 3.2, we have

\[
\frac{1}{n} \text{tr}(A_n^2) = \frac{1}{nb_n} \sum_{i=1}^{n} \sum_{J} \text{E}[a_J] I_J \delta_{0, \sum_{l=1}^{2k} j_l} = \sum_1 + \sum_2 + \sum_3, \tag{3.4}
\]

where

\[
\sum_1 = \frac{1}{nb_n} \sum_{i=1}^{n} \sum_{J \in \Gamma_1(2k)} \text{E}[a_J] I_J, \quad l = 1, 2, 3. \tag{3.5}
\]

By the definition of the categories and the assumptions on the entries of the random matrices, we obtain

\[
\sum_2 = 0.
\]

Next, we divide \( \sum_3 \) into two parts

\[
\sum_3 = \sum_3[1] + \sum_3[2],
\]

where

\[
\sum_3[1] = \frac{1}{nb_n} \sum_{i=1}^{n} \sum_{J \in \Gamma_3(2k)} \text{E}[a_J] I_J, \quad l = 1, 2. \tag{3.6}
\]

For \( J \in \Gamma_3(2k) \), we denote the number of distinct elements of \( S_J \) by \( t \). By the definition of the category, we have \( t \leq k \). Note that the random variables whose subscripts have different absolute values are independent. Once we have specified the distinct numbers of \( S_J \), the subscripts \( j_1, \cdots, j_{2k} \) are determined in at most \( 2^{2k} k^{2k} \) ways. If \( J \in \Gamma_3(2k) \),
then \( t \leq \frac{2k-1}{2} \). Again by independence and the assumptions on the matrix elements (1.7), we find

\[
| \sum_{31} | \leq \frac{1}{n b_n^k} \sum_{i=1}^{n} O(b_n^{2k-1}) = o(1).
\]

When \( J \in \Gamma_{32}(2k) \), there exist \( p_0, q_0 \in [2k] \) such that

\[
j_{p_0} = j_{q_0} = \frac{1}{2} (n b_n + j_q - \sum_{q=1}^{2k})\]

We can choose the other at most \( k - 1 \) distinct numbers, which determine \( j_{p_0} = j_{q_0} \). This shows that there is a loss of at least one degree of freedom, thus the contribution of such terms is \( O(b_n^{k-1}) \), i.e.

\[
| \sum_{32} | = o(1).
\]

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\[
| \sum_{32} | = o(1).
\]

Since the main contribution comes from the category 1, each term \( E[\alpha_J] = 1 \) for \( J \in \Gamma_{1}(2k) \). So we can rewrite

\[
E[\frac{1}{n} \text{tr}(A_n^{2k})] = o(1) + \frac{1}{n b_n^k} \sum_{\pi \in \mathcal{P}_2(2k)} \sum_{i=1}^{n} b_n \sum_{j_1, \ldots, j_k=1 \atop \Pi_j}^{2k} \chi_{[1,n]}(i + \sum_{q=1}^{l} \epsilon_{\pi}(q) j_{\pi(q)}).
\]

(3.7)

For fixed \( \pi \in \mathcal{P}_2(2k) \),

\[
\frac{1}{n b_n^k} \sum_{i=1}^{n} b_n \sum_{j_1, \ldots, j_k=1 \atop \Pi_j}^{2k} \chi_{[1,n]}(i + \sum_{q=1}^{l} \epsilon_{\pi}(q) j_{\pi(q)}),
\]

i.e.

\[
\frac{1}{n b_n^k} \sum_{i=1}^{n} b_n \sum_{j_1, \ldots, j_k=1 \atop \Pi_j}^{2k} \chi_{[1,n]}(i + \sum_{q=1}^{l} \epsilon_{\pi}(q) j_{\pi(q)})
\]

can be considered as a Riemann sum of the definite integral

\[
\int_{[0.1] \times [-1,1]}^{2k} \prod_{l=1}^{k} \chi_{[0,1]}(x_0 + b \sum_{q=1}^{l} \epsilon_{\pi}(q) x_{\pi(q)}) \prod_{l=0}^{k} d x_l.
\]

As in the above arguments, by Lemma 3.2 and the assumptions on the matrix elements (1.7), we have

\[
\left| E[\frac{1}{n} \text{tr}(A_n^{2k+1})] \right| \leq \frac{1}{n b_n^{k+1}} \sum_{i=1}^{n} O(b_n^k) = o(1)
\]

(3.8)

since \( \mathcal{P}_2(2k - 1) = \emptyset \).

This completes the proof.

\( \square \)

**Remark 3.4.** When \( b = 0 \), we can easily get \( M_{2k} = 2^k(2k - 1)!! \). This is just the 2k-moment of the normal distribution with variance 2, which is also obtained independently by Basak and Bose [4] and Kargin [19]. However, for \( b > 0 \) it is quite difficult to calculate \( M_{2k} \) because the integrals in the sum of (3.1) are not all the same for different partitions \( \pi \)'s, some of which are too hard to evaluate.
4 covariance

In this section we evaluate the covariance of $\omega_p$ and $\omega_q$. Recall

$$\omega_p = \frac{\sqrt{b_n}}{n} (\text{tr}(A_p^n) - E[\text{tr}(A_p^n)]) = \frac{1}{n} b_n^{-p+1} \sum_{i=1}^{n} \sum_{J} I_J (a_J - E[a_J]) \delta_{0, \sum_{j=1}^{p} j_i},$$

thus

$$E[\omega_p \omega_q] = \frac{1}{n^2} b_n^{-p+q+1} \sum_{i,i',J,J'} I_J I_{J'} E[(a_J - E[a_J]) (a_{J'} - E[a_{J'}])]$$

$$= \frac{1}{n^2} b_n^{-p+q+1} \sum_{i,i',J,J'} I_J I_{J'} (E[a_J a_{J'}] - E[a_J]E[a_{J'}]),$$

where the summation $\sum_{J,J'}$ runs over all balanced vectors $J \in \{\pm 1, \ldots, \pm b_n\}^p$ and $J' \in \{\pm 1, \ldots, \pm b_n\}^q$.

The main result of this section can be stated as follows:

**Theorem 4.1.** Using the notations in section 2, for fixed $p, q \geq 2$, as $n \to \infty$ we have

$$E[\omega_p \omega_q] \to \sigma_{p,q} = \sum_{\pi \in \mathcal{P}_2(p,q)} (f_I^\pi (\pi) + f_{I'}^\pi (\pi)) + (\kappa - 1) \sum_{\pi \in \mathcal{P}_2(u,p,q)} (f_{I_1^u}^\pi (\pi) + f_{I_1^u}^\pi (\pi))$$

(4.3)

when $p + q$ is even and

$$E[\omega_p \omega_q] = o(1)$$

(4.4)

when $p + q$ is odd.

When $p = q$ we denote $\sigma_{p,q}$ by $\sigma_p^2$, where $\sigma_p$ denotes the standard deviation. From the theorem above, we obtain the variance of $\omega_Q$ in Theorem 1.1

$$\sigma_Q^2 = \sum_{i=2}^{p} \sum_{j=2}^{p} q_i q_j \sigma_{i,j}.$$  

(4.5)

By the independence of matrix entries, the only non-zero terms in the sum of (4.2) come from pairs of balanced vectors $J = (j_1, \ldots, j_p)$ and $J' = (j'_1, \ldots, j'_q)$ such that

(i) The projections $S_J$ and $S_{J'}$ of $J$ and $J'$ have at least one element in common;

(ii) Each number in the union of $S_J$ and $S_{J'}$ occurs at least two times.

**Definition 4.2.** Any ordered pair of balanced vectors $J = (j_1, \ldots, j_p)$ and $J' = (j'_1, \ldots, j'_q)$ satisfying (i) above is called correlated. If $j_u \in J$ ($j_u$ is a component of $J$) and $|j_u| \in S_J \bigcap S_{J'}$, then $j_u$ is called a joint point of the ordered correlated pair.

To observe which correlated pairs lead to the main contribution to the covariance, we next construct a balanced vector of dimension $(p + q - 2)$ from each correlated pair $J$ of dimension $p$ and $J'$ of dimension $q$. Although the corresponding map of correlated pairs to such balanced vectors is not one to one, the number of pre-images for a balanced vector of dimension $(p + q - 2)$ is finite (only depending on $p$ and $q$). We will study the resulting balanced vectors in a similar way as in section 3.

**Proof of Theorem 4.1.** Let us first construct a map from the ordered correlated pair $J$ and $J'$ as follows. Let $j_u \in J$ be the first joint point (whose subscript is the smallest)
of the ordered correlated pair $J$ and $J'$, and let $j'_u$ be the first element in $J'$ such that $|J_u| = |j'_u|$. If $j_u = - j'_u$, we construct a vector $L = (l_1, \ldots, l_{p+q-2})$ such that

$$l_1 = j_1, \ldots, l_{u-1} = j_{u-1}, l_u = j'_1, \ldots, l_{u+v-2} = j'_v-1,$$

$$l_{u+v-1} = j'_v+1, \ldots, l_{u+q-2} = j'_q , l_{u+q-1} = j_{u+1}, \ldots, l_{p+q-2} = j_p.$$ 

It is obvious that

$$\sum_{k=1}^{p+q-2} l_k = 0,$$

so $L$ is balanced. If $j_u = j'_u$, then from $J$ and $-J' = (-j'_1, \ldots, -j'_q)$ we proceed as in the same way above. We call this process of constructing $L$ from $J$ and $J'$ a reduction step and denote it by $L = J \cup |j_u, J'|$.

**Remark 4.3.** From the construction above, for any joint point of $J$ and $J'$ a reduction step can be done in the same way. Given $\theta \in S_J \cap S_{J'}$, when saying $J \cap \theta J'$, we mean that there exists some joint point of $J$ and $J'$ $j_u$ satisfying $|J_u| = \theta$ and $J \cap \theta J'$ is the vector after this reduction step. In this section, $j_u$ is always the first joint point. While in section 5, $j_u$ may denote other joint points which is clear in the context.

Notice that the reduction might cause the appearance of one number with multiplicity 1 in $S_L$, although each number in the union of $S_J$ and $S_{J'}$ occurs at least two times. If so, the resulting number with multiplicity 1 in $S_L$ must be coincident with the joint point $j_u$. In addition, to estimate which terms lead to main contribution to higher moments of $tr(A_n^p)$, we will use the reduction steps and mark the appearance of the numbers with multiplicity 1 in section 5.

Next, assume we have a balanced vector $L$ of dimension $(p+q-2)$, we shall estimate in how many different ways it can be obtained from correlated pairs of dimensions $p$ and $q$. First, we have to choose some component $l_u$ in the first half of the vector, $1 \leq u \leq p$ such that

$$\sum_{i=1}^{u+q-2} l_i \neq |l_j|, \quad j = 1, \ldots, u - 1. \quad (4.6)$$

Set $J = (j_1, \ldots, j_p)$ with

$$j_1 = l_1, \ldots, j_{u-1} = l_{u-1}, j_u = \sum_{i=u}^{u+q-2} l_i, j_{u+1} = l_{u+q-1}, \ldots, j_p = l_{p+q-2}. \quad (4.7)$$

We also have to choose some component $l_{u+v-1}$, $1 \leq v \leq q - 1$ such that

$$\sum_{i=u}^{u+q-2} l_i \neq |l_j|, \quad j = u, \ldots, u + v - 2 \quad (4.8)$$

whenever $v \geq 2$. Set $J' = (j'_1, \ldots, j'_q)$ with

$$j'_1 = l_u, \ldots, j'_{u-1} = l_{u+v-2}, j'_v = - \sum_{i=u}^{u+q-2} l_i, j'_{v+1} = l_{u+v-1}, \ldots, j'_q = l_{u+q-2}. \quad (4.9)$$

If $j_u$ is the joint point of the constructed correlated pair $J$ and $J'$ and $j'_u$ is the corresponding element in $J'$, then the pair $\{J, J'\}$ or $\{J, -J'\}$ is the pre-image of $L$. Note that since when $u = v = 1$ the conditions (4.6) and (4.8) are satisfied, the pre-image of $L$ always exists. A simple estimation shows that the number of pre-images of $L$ is at most $2pq$, not depending on $n$ (we will see this fact plays an important role in the estimation of higher moments in section 5).
There is at most one element with multiplicity 1 in $S_L$. If so, this number will be determined by others because of the balance of $L$. Consequently, the degree of freedom for such terms is at most $p + q - 2 - 1$. Therefore, the sum of these terms will be $O(b_n^{-1/2})$, which can be omitted. Now we suppose each number in $S_L$ occurs at least two times. Recall the procedure in section 3, and we know that the main contribution to the covariance (4.2) comes from the $L \in \Gamma_1(p + q - 2)$, which implies $E[\omega_p \omega_q] = o(1)$ when $p + q$ is even. When $p + q$ is even, for $L \in \Gamma_1(p + q - 2)$ the weight

$$E[a_J a_{J'}] - E[a_J]E[a_{J'}] = E[\prod_{s=1}^{p} a_{j_s} \prod_{t=1}^{q} a_{j'_t}] - E[\prod_{s=1}^{p} a_{j_s}]E[\prod_{t=1}^{q} a_{j'_t}]$$

equals to 1 if $j_s$ is not coincident with any component of $L$; otherwise the weight is either $E[|a_{j_s}|^4] = \kappa$ or $E[|a_{j_s}|^4] - (E[|a_{j_s}|^2])^2 = \kappa - 1$.

So far we have found the terms leading to the main contribution, now we calculate the variance. Based on whether or not the fourth moment appears, we evaluate the variance. If the fourth moment doesn’t appear, then $j_1, \ldots, j_p, j'_1, \ldots, j'_q$ match in pairs. By their subscripts they can be treated as pair partitions of $\{1, 2, \ldots, p, p + 1, \ldots, p + q\}$ but with at least one crossing match (i.e., $\mathcal{P}_2(p, q)$ as in section 2). Thus, for every $\pi \in \mathcal{P}_2(p, q)$, the summation can be a Riemann sum and its limit becomes $f^+_{II}(\pi)$ (it is $f^+_{II}(\pi)$ when the first coincident components in $J$ and $J'$ have the same sign). On the other hand, if the fourth moment does appear, then $j_1, \ldots, j_p, j'_1, \ldots, j'_q$ match in pairs except that there exists a block with four elements. Therefore, from the balance of $\sum_{k=1}^{p} j_k = 0$ and $\sum_{k=1}^{q} j'_k = 0$, we know that the main contribution must come from such partitions: $j_1, \ldots, j_p$ and $j'_1, \ldots, j'_q$ both form pair partitions; the block with four elements take respectively from a pair of $j_1, \ldots, j_p$ and $j'_1, \ldots, j'_q$. Otherwise, the degree of freedom decreases by at least one. Similarly, for every $\pi \in \mathcal{P}_{2,4}(p, q)$, the corresponding summation can be a Riemann sum and its limit becomes $f^+_{II}(\pi)$ (it is $f^+_{II}(\pi)$ when the first coincident components in $J$ and $J'$ have the same sign).

In a similar way as in section 3, noting that the coincident components in $J$ and $J'$ may have the same or opposite sign, we conclude with the notations in section 2 that

$$E[\omega_p \omega_q] \to \sum_{\pi \in \mathcal{P}_2(p, q)} (f^+_{II}(\pi) + f^+_{II}(\pi)) + (\kappa - 1) \sum_{\pi \in \mathcal{P}_{2,4}(p, q)} (f^+_{II}(\pi) + f^+_{II}(\pi))$$

as $n \to \infty$.

This completes the proof. \hfill \Box

5 Higher Moments of $\text{tr}(A_n^p)$

Let $\mathfrak{B}_{n, p}$ denote the set of all balanced vectors $J = (j_1, \ldots, j_p) \in \{\pm 1, \ldots, \pm b_n\}^p$. Let $\mathfrak{B}_{n, p, i}$ ($1 \leq i \leq n$) be a subset of $\mathfrak{B}_{n, p}$ such that $J \in \mathfrak{B}_{n, p, i}$ if and only if

$$\forall i \in \{1, \ldots, p\}, \quad 1 \leq i + \sum_{q=1}^{i} j_q \leq n.$$

With these notations, Lemma 3.2 can be rewritten as

$$\text{tr}(T_n^p) = \sum_{i=1}^{n} \sum_{J \in \mathfrak{B}_{n, p, i}} a_J.$$  \hfill (5.1)
To finish the proof of Theorem 1.1, it is sufficient to show that given \( p_1, p_2, \cdots, p_l \geq 2 \) and \( l \geq 1 \), as \( n \to \infty \) we have

\[
E[\omega_{p_1} \omega_{p_2} \cdots \omega_{p_l}] = E[g_{p_1} g_{p_2} \cdots g_{p_l}],
\]

(5.2)

where \( \{g_p\}_{p \geq 2} \) is a centered Gaussian family with covariances \( \sigma_{p,q} = E[g_p g_q] \).

Then a CLT for

\[
\omega_Q = \sqrt{\frac{b_n}{n}} (\text{tr}Q(A_n) - E[\text{tr}Q(A_n)])
\]

(5.3)

follows, with the variance

\[
\sigma_Q^2 = \sum_{i=2}^{p} \sum_{j=2}^{p} q_i q_j \sigma_{i,j}.
\]

(5.4)

The main idea is rather straightforward: in an analogous way to the one used in Eq. (4.2), we will deal with

\[
E[\omega_{p_1} \cdots \omega_{p_l}] =
\]

\[
n^{-l} b_n^{\frac{p_1 + \cdots + p_l}{2}} \sum_{i_1, \ldots, i_l = 1}^{n} \sum_{J_1 \in \mathcal{B}_{n,p_1}, \ldots, J_l \in \mathcal{B}_{n,p_l}}^* \ E\left[ \prod_{i=1}^{l} (a_{iJ_i} - E[a_{iJ_i}]) \right].
\]

(5.5)

Remember that two balanced vectors \( J = (j_1, \ldots, j_p) \) and \( J' = (j'_1, \ldots, j'_q) \) are called correlated if the corresponding projections \( S_J \) and \( S_{J'} \) of \( J \) and \( J' \) have at least one element in common.

**Definition 5.1.** Given a set of balanced vectors \( \{J_1, J_2, \ldots, J_l\} \), a subset of balanced vectors \( J_{m_1}, J_{m_2}, \ldots, J_{m_l} \) is called a cluster if

1) for any pair \( J_{m_i}, J_{m_j} \) from the subset one can find a chain of vectors \( J_{m_k} \), also belonging to the subset, which starts with \( J_{m_i} \) and ends with \( J_{m_j} \), such that any two neighboring vectors are correlated;

2) the subset \( \{J_{m_1}, J_{m_2}, \ldots, J_{m_l}\} \) cannot be enlarged with the preservation of 1).

It is clear that the vectors corresponding to different clusters are disjoint. By this reason the mathematical expectation in (5.5) decomposes into the product of mathematical expectations corresponding to different clusters. We will show that the leading contribution to (5.5) comes from products where all clusters consist exactly of two vectors, as is stated in Lemma 5.2 below.

**Lemma 5.2.** Provided \( l \geq 3 \), we have

\[
n^{-l} b_n^{\frac{p_1 + \cdots + p_l}{2}} \sum_{i_1, \ldots, i_l = 1}^{n} \sum_{J_1 \in \mathcal{B}_{n,p_1}, \ldots, J_l \in \mathcal{B}_{n,p_l}}^* \ E\left[ \prod_{i=1}^{l} (a_{iJ_i} - E[a_{iJ_i}]) \right] = o(1)
\]

(5.6)

where the sum \( \sum^* \) in (5.6) is taken over \( l \) vectors which exactly form a cluster.

For fixed \( p \), all the involved moments no higher than \( p \) are \( O(1) \) because of uniform boundedness of matrix entries. On the other hand, \( 0 \leq \prod_{i=1}^{l} I_{J_i} \leq 1 \). So to prove Lemma 5.2, we just need to count the number of terms in (5.6). As before, to complete the estimation it suffices to replace \( \mathcal{B}_{n,p_i} \) by \( \mathcal{B}_{n,p} \). That is, it suffices to prove

\[
b_n^{\frac{p_1 + \cdots + p_l}{2}} \sum_{J_1 \in \mathcal{B}_{n,p_1}, \ldots, J_l \in \mathcal{B}_{n,p_l}}^* 1 = o(1)
\]

(5.7)

where the summation \( \sum^* \) is taken over \( l \) vectors which exactly form a cluster.

Instead of Lemma 5.2, we will prove
Lemma 5.3. Provided \( l \geq 3, p = (p_1, \ldots, p_l) \) with positive integers \( p_1, \ldots, p_l \geq 2 \). Let \( \mathcal{B}_p \) be a subset of the Cartesian product \( \mathcal{B}_{n,p_1} \times \cdots \times \mathcal{B}_{n,p_l} \) such that \( (J_1, J_2, \ldots, J_l) \in \mathcal{B}_p \) if and only if

(i) any element in \( \bigcup_{i=1}^{l} S_{J_i} \) has at least multiplicity two in the union;
(ii) \( J_1, J_2, \ldots, J_l \) make a cluster;
(iii) \( 0 \notin \bigcup_{i=1}^{l} S_{J_i} \).

Then we claim that

\[
\text{card}(\mathcal{B}_p) = o\left(b_n^{p_1 + p_2 + \cdots + p_l - \frac{1}{2}}\right). \quad (5.8)
\]

Notice that we list the condition (iii) which looks redundant from \( J \in \{ \pm 1, \ldots, \pm b_n \}^p \) to emphasize the importance of \( 0 \) (i.e., the diagonal matrix entry \( a_0 \)). In fact, if \( p_1, p_2, \ldots, p_l \) are all even, even for these \( J \in \{ 0, \pm 1, \ldots, \pm b_n \}^p \) under condition (i) and (ii), the estimation above is true.

Proof of Lemma 5.3. The intuitive idea of the proof is as follows: from condition (i) in Lemma 5.3, regardless of the correlating condition, the cardinality of \( \mathcal{B}_p \) (denoted by \( \text{card}(\mathcal{B}_p) \) for short) is \( O(b_n^{p_1 + p_2 + \cdots + p_l}) \). We can say that the freedom degree is \( \frac{p_1 + p_2 + \cdots + p_l}{2} \). But each correlation means two vectors share a common element so that it will decrease the freedom degree by one. To form a cluster we need \( l - 1 \) correlations. So

\[
\text{card}(\mathcal{B}_p) = O\left(b_n^{p_1 + p_2 + \cdots + p_l - (l-1)}\right). \quad \text{When } l \geq 3, l - 1 > \frac{l}{2} \text{.}
\]

Thus we obtain the desired estimation in Lemma 5.3. However, adding one correlation does not necessarily lead the freedom degree to decrease by one. Some may be redundant. So we have to make use of the correlations more efficiently.

As in section 4, we do the reduction steps as long as the structure of the cluster is preserved. To say precisely, we start from checking \( J_1 \) and \( J_2 \). If \( J_a \) is the first joint point of \( J_1 \) and \( J_2 \) satisfying the condition that \( J_1 \bigcup J_a \bigcup J_2 \) can still form a cluster with the other vectors, then we do this reduction step. If this kind of \( J_a \) does not exist, we turn to check \( J_1 \) and \( J_2 \) in the same way. After each reduction step, we have a new cluster of vectors \( \tilde{J}_1, \tilde{J}_2, \ldots, \tilde{J}_l \). We continue to check \( \tilde{J}_1 \) and \( \tilde{J}_2 \) as before. If we cannot do any reduction step, we stop.

Suppose that we did \( m \) reduction steps in total. Then we have a new cluster of vectors \( \tilde{J}_1', \tilde{J}_2', \ldots, \tilde{J}_l' \) and the dimension of \( \tilde{J}_i' \) is \( p_i' \) for \( 1 \leq i \leq l' \). From the reduction process, for any \( \theta \in S_{\tilde{J}_a}, J_{a1} \bigcap S_{J_a}, J_{a2} \bigcap S_{J_a} \) cannot form a cluster with the other vectors. The resulting cluster still satisfies condition (ii) in Lemma 5.3. However, the condition (i) may fail because the joint point of a pair of correlated vectors can be a tripartite one, thus after a reduction step its multiplicity becomes one.

Note that after a reduction step the number of pre-images of the correlating vector only depends on the dimensions of the involving vectors, not depending on \( n \). Thus we only need to estimate the degree of freedom of the reduced vectors \( J_1', J_2', \ldots, J_l' \).

Since after one reduction step the total dimension of vectors will decrease by two and the number of vectors will decrease by one, we have

\[
\sum_{i=1}^{l} p_i = \sum_{i=1}^{l'} p_i' + 2m \quad (5.9)
\]

and

\[
l = l' + m. \quad (5.10)
\]

Denote by \( l_0 \) the number of single multiplicity elements in \( \bigcup_{i=1}^{l'} S_{J_i} \). Since one reduction step will add at most one element with single multiplicity, therefore

\[
l_0 \leq m. \quad (5.11)
\]

Below, we will proceed according to two cases: \( l' > 1 \) and \( l' = 1 \).
In the case \( l' > 1 \)

To complete the proof of this case, we need some definitions and notations.

Let \( \mathcal{U} \) be the set consisting of all elements which belong to at least two of \( S_{J_1}, \ldots, S_{J_t} \), i.e., \( \mathcal{U} = \{ \theta | \exists i 
eq j \text{ s.t. } \theta \in S_{J_i} \cap S_{J_j} \} \). Since \( l' > 1 \) and \( J_1', \ldots, J_t' \) form a cluster, \( \mathcal{U} \neq \emptyset \).

For any \( \theta \in \bigcup_{i=1}^{l'} S_{J_i} \), set \( H_\theta = \{ J_i' | \theta - \theta \in S_{J_i}', 1 \leq i \leq l' \} \) and denote the number of vectors in \( H_\theta \) by \( h_\theta = \text{card}(H_\theta) \). Obviously, \( \{ J_i'|i \leq l' \} = \bigcup_{\theta \in \mathcal{U}} H_\theta \).

We notice the following three facts.

Fact 1: For any \( \theta \in \mathcal{U} \), \( h_\theta \geq 2 \). If \( h_\theta = 2 \), the two vectors in \( H_\theta \) can still be reduced to one vector and the reduction doesn’t affect their connection with the other vectors, which is a contradiction with our assumption that \( J_1', J_2', \ldots, J_t' \) cannot be reduced.

Fact 2: For any \( \theta \in \mathcal{U} \) and \( J_i' \in H_\theta \), the multiplicity of \( \theta \) in \( S_{J_i} \) is one. Otherwise, there are two vectors belonging to \( H_\theta \), for example, \( J_i', J_j' \in H_\theta \) but \( S_{J_i} \) has two \( \theta \)'s, then \( J_i' \cup J_j' \) can be a reduction step and \( J_i' \cup J_j' \) still form a cluster.

Fact 3: For any different elements \( \theta \) and \( \gamma \) in \( \mathcal{U} \), \( \text{card}(H_\theta \cap H_\gamma) \leq 1 \). Otherwise, suppose \( J_i' \) and \( J_j' \) belong to \( H_\theta \cap H_\gamma \) and \( J_k' \) is an element of \( H_\gamma \) other than \( J_i' \) and \( J_j' \). From Fact 1, \( J_k' \) must exist. Now \( J_i' \cup J_k' \) can form a reduction step since \( J_i' \cup J_k' \) can be correlated with \( J_j' \) by \( \gamma \) and other correlations won’t be broken.

**Definition 5.4.** \( \mathcal{V} \subset \mathcal{U} \) is called a dominating set of \( \{ J_1', \ldots, J_t' \} \) if \( \{ J_i'|1 \leq i \leq l' \} = \bigcup_{\theta \in \mathcal{V}} H_\theta \).

Choose a minimal dominating set denoted by \( \mathcal{U}_0 \), which means that any proper subset of \( \mathcal{U}_0 \) is not a dominating set. Since \( \mathcal{U} \) is a finite set, \( \mathcal{U}_0 \) must exist. Let \( \mathcal{U}_0 = \{ \theta_i | 1 \leq i \leq t \} \). For any \( 1 \leq j \leq t \), since \( \mathcal{U}_0 \setminus \{ \theta_j \} \) is not a dominating set, there exists \( J_{k_j}' \in H_{\theta_j} \setminus \bigcup_{j \neq \theta_j} H_{\theta_j} \). Once we have already known the elements of \( \bigcup_{i=1}^{l'} S_{J_i}' \setminus \mathcal{U}_0 \), we know all the elements in \( S_{J_{k_j}}' \) other than \( \theta_j \), thus \( \theta_j \) will be determined by the balance of some \( \gamma \) in \( \mathcal{U} \).

Set

\[
h = \sum_{i=1}^{t} h_{\theta_i},
\]

then the different way of choice of \( \bigcup_{i=1}^{l'} S_{J_i}' \setminus \mathcal{U}_0 \) is \( O\left(b_n \frac{\sum_{i=1}^{t} p_i - l - h}{2} \right) \). From Eqs. (5.9) and (5.10), we have

\[
\frac{\sum_{i=1}^{l'} p_i - l - h}{2} + l_0 = \left( \frac{\sum_{i=1}^{t} p_i - l}{2} \right) - \left( m - l_0 \right) - (h - l') .
\]

(5.12)

Note that \( m - l_0 \geq 0 \) from Eq. (5.11). If \( h' \), we have

\[
O\left(b_n \frac{\sum_{i=1}^{t} p_i - l}{2} \right) = o\left(b_n \frac{\sum_{i=1}^{t} p_i - l}{2} \right).
\]

(5.13)

If \( h = l' \) and \( t = 1 \), the analysis is easy but a little complex. We will deal with it later.

Now we focus on the case that \( h = l' \) and \( t > 1 \). In this case, since \( \{ J_i'|i \leq l' \} = \bigcup_{\theta \in \mathcal{U}_0} H_\theta \), \( l' \leq \sum_{i=1}^{l'} \text{card}(H_\theta) = h \) and the equity is true iff \( H_{\theta_1} \cap H_{\theta_2} = \emptyset \) for any \( 1 \leq i \neq j \leq t \). Because \( J_1', J_2', \ldots, J_t' \) form a cluster, without lost of generality, we can assume that \( \bigcup_{\theta \in H_{\theta_1} \cap H_{\theta_2}} S_{J_1}' \cap \bigcup_{\theta \in H_{\theta_1}} S_{J_i}' \neq \emptyset \). This way there exist \( J_{s_1}' \in H_{\theta_1}, J_{s_2}' \in H_{\theta_2} \) and \( \gamma \in S_{J_{s_1}}' \cap S_{J_{s_2}}' \). We know that \( \gamma \in \mathcal{U} \setminus H_\theta \) since \( H_{\theta_1} \cap H_{\theta_2} = \emptyset \) for any \( 1 \leq i \neq j \leq t \). From Fact 3, \( \text{card}(H_{\theta_1} \cap H_\gamma) = 0 \) or 1. Given the elements of \( \bigcup_{i=1}^{l'} S_{J_i}' \setminus (\mathcal{U}_0 \cup \{ \gamma \}) \), \( \theta_i \) can be decided by the balance of some \( J_i' \in H_{\theta_i} \setminus H_\gamma \). Then \( \gamma \) can be decided by any vector in
$H_\gamma$. Thus we have $O(b_n^{\frac{\sum_{i=1}^{l'} p_i' - h_0 - h_n}{2} + l_0})$ ways to decide $J'_1, J'_2, \ldots, J'_l$. From Eq. (5.12), one gets

$$O(b_n^{\frac{\sum_{i=1}^{l'} p_i' - h_0 - h_n}{2} + l_0}) = \frac{O\left(b_n^{\left(\frac{\sum_{i=1}^{l'} p_i - (m - l_0) - (h - l') - h_n}{2}\right)}\right)}{O\left(b_n^{\frac{\sum_{i=1}^{l} p_i - l}{2}}\right)} = o\left(b_n^{\frac{\sum_{i=1}^{l} p_i - l}{2}}\right).$$

(5.14)

If $h = l'$ and $t = 1$, which means that $\theta_1$ appears exactly one time in each $S_{j_i} (1 \leq i \leq l')$, we will divide the situation into three subcases.

**Case I:** $l_0 > 0$.

Without loss of generality, we suppose $\alpha \in J'_1$ is an element with single multiplicity in $\bigcup_{i=1}^{l'} J'_i$. If the elements except for $\theta_1$ and $\alpha$ are known, $\theta_1$ can be determined from the balance of $J'_2$ and then $\alpha$ can be determined from the balance of $J'_1$. So we have

$$O(b_n^{\frac{\sum_{i=1}^{l'} p_i' - h_0 - h_n}{2} + l_0 - 1})$$

ways of choice in sum. From $h_{i_0} = h = l'$ and Eqs. (5.9) and (5.10), we get

$$O(b_n^{\frac{\sum_{i=1}^{l'} p_i' - h_0 - h_n}{2} + l_0 - 1}) = O(b_n^{\frac{\sum_{i=1}^{l} p_i - l}{2} - 1}) = o\left(b_n^{\frac{\sum_{i=1}^{l} p_i - l}{2}}\right).$$

**Case II:** $l_0 = 0$ and $m > 0$.

As in case I above, $\theta_1$ is determined by other elements. To determine all the elements other than $\theta_1$, we have $O(b_n^{\sum_{i=1}^{l'} p_i' - l})$ ways. From Eqs. (5.9) and (5.10), we have

$$O(b_n^{\frac{\sum_{i=1}^{l'} p_i' - l}{2}}) = O(b_n^{\frac{\sum_{i=1}^{l} p_i - l}{2} - m}) = o\left(b_n^{\frac{\sum_{i=1}^{l} p_i - l}{2}}\right).$$

**Case III:** $l_0 = 0$ and $m = 0$.

In this case, we cannot do any reduction step.

We claim that there exists $\gamma \in S_{J_1}$ other than $\theta_1$ such that the number of $+\gamma$ and $-\gamma$ in $J_1$ are not equal (when $p_1$ is even we can always find some $\gamma$ other than $\theta_1$ such that $\gamma$ occurs odd times in $S_{J_1}$). Otherwise, $\theta_1 = 0$ because of the balance of $J_1$, which contradicts condition (iii) in Lemma 5.3.

To determine all the elements except for $\theta_1$ and $\gamma$, we have

$$O\left(b_n^{\frac{(p_1 + \ldots + p_l) - (b_{b_1} + 1)}{2}}\right)$$

ways. Since $H_{\theta_1} = \{J_1, J_2, \ldots, J_l\}$, we have $H_\gamma = \{J_1\}$ (otherwise a reduction can be done). So we can determine $\theta_1$ from the balance of $J_2$. Then $\gamma$ will be determined by the balance of $J_1$. Since $h_{b_1} = l$, we have

$$O\left(b_n^{\frac{(p_1 + \ldots + p_l) - (b_{b_1} + 1)}{2}}\right) = o\left(b_n^{\frac{(p_1 + \ldots + p_l) - l}{2}}\right).$$

Now we complete the proof in the case of $l' > 1$.

**In the case $l' = 1$**
We also divide this case into two subcases.

**Case I**: \( l_0 > 0 \).

Suppose \( \alpha \in J'_1 \) is an element with single multiplicity. If the elements other than \( \alpha \) are known, \( \alpha \) can be determined from the balance of \( J'_1 \). So we have totally \( O(b_n^{\sum_{i=1}^p l'_i - l_0 + 1} \delta_{l_0 - 1}) \) ways. From \( l' = 1 \) and Eqs. (5.9), (5.10) and (5.11), we have

\[
O\left(b_n^{\sum_{i=1}^p l'_i - l_0 + 1} \delta_{l_0 - 1}\right) = O\left(b_n^{\sum_{i=1}^p p_i - 1} \delta_{l_0 - 1}\right) = o\left(b_n^{\sum_{i=1}^p p_i - 1} \delta_{l_0 - 1}\right).
\]

**Case II**: \( l_0 = 0 \).

In this case, every element in \( S_{J'_1} \) appears at least twice. Thus we have totally \( O(b_n^{\sum_{i=1}^p l'_i}) \) ways. From \( l' = 1 \) and Eqs. (5.9) and (5.10), it follows that

\[
O\left(b_n^{\sum_{i=1}^p l'_i} \delta_{l_0 - 1}\right) = O\left(b_n^{\sum_{i=1}^p p_i - 1} \delta_{l_0 - 1}\right) = o\left(b_n^{\sum_{i=1}^p p_i - 1} \delta_{l_0 - 1}\right)
\]

since \( \delta \geq 3 \).

Now we have completed the proof in the case of \( l' = 1 \).

The proof of Lemma 5.3 is then complete.

We now turn to the proof of the main Theorem 1.1.

**Proof of Theorem 1.1.** By Theorem 4.1, Lemma 5.3 and Wick formulas for centered Gaussian distributions, it’s easy to complete the proof.

**Remark 5.5.** From the analysis of case III above, we can also understand why the technical but necessary condition of \( a_0 \equiv 0 \) is assumed in the introduction. But when \( p \) is even, the assumption of \( a_0 \equiv 0 \) is not necessary.

**Remark 5.6.** From the calculation of mathematical expectation, variance, and higher moments in sections 3, 4, 5, the factor \( I_J \) in the trace formula of Lemma 3.2 can be replaced by a more general non-negative integrable bounded function. For example, if \( I_J \equiv 1 \) for all \( J \), then the results of asymptotic distribution and fluctuation hold the same as in case where \( b = 0 \). More precisely, as \( n \to \infty \)

\[
\frac{1}{b_n^{\sum_{i=1}^p j_i}} \sum_{j_1, \ldots, j_p = -b_n}^{b_n} \mathbb{E}\left[ \prod_{l=1}^p a_{j_l} \right] \delta_{\sum_{i=1}^p j_i + 1} \to \begin{cases} 2\pi^2 (p - 1)!! & \text{if } p \text{ is even;} \\ 0 & \text{if } p \text{ is odd} \end{cases}
\]

and the fluctuation can be stated as in Corollary 1.2 (For conciseness we replace \( b_n \) by \( k \) there). From this point of view, our method gives a mechanism of CLT.

### 6 Extensions to other models

In this section, we will make use of our method to deal with some random matrix models closely related to Toeplitz matrices, including Hermitian Toeplitz band matrices and Hankel band matrices. Before we start our extensions, we first review and generalize the key procedures or arguments in calculating mathematical expectation (the distribution of eigenvalues), covariance and higher moments in sections 3, 4, 5 respectively as follows:

1) "Good" trace formulae. See Lemma 3.2. The simple formula both represents the form of the matrix and translates our object of matrix entries to its subscripts \( j_1, \ldots, j_p \).
which are integers satisfying some homogeneous equation. In the present paper, we mainly encounter such homogeneous equations as

\[ \sum_{l=1}^{p} \tau_l j_l = 0, \]  

(6.1)

where \( \tau_l \) can take \( \pm 1 \). We write \( J = (j_1, \ldots, j_p) \).

2) “Balanced” vectors via some homogeneous equations. We can generalize the concept of balance: a vector \( J = (j_1, \ldots, j_p) \) is said to be balanced if its components satisfy one of finite fixed equations with the form of (6.1). From the balance of a vector, one can determine an element by the others and solve one of the equations.

3) Reduction via a joint point. Eliminate the joint point from two correlated vectors in some definite way, and one gets a new balanced vector.

4) Choose a minimal dominating set by which the freedom degree can be reduced case-by-case. We should particularly be careful about 0.

The CLT is essentially a consequence of the fact that we can omit the terms which have a cluster consisting of more than two vectors. Thus if the argument in section 5 is valid, by Wick formulas the CLT is true. In the following models, we will establish a good trace formula in each case, and then the balanced vector and reduction step can be defined in a natural way. With these equipment the analysis in section 5 is still valid after a small adaption.

The mathematical expectation and covariance vary from case to case. But the way to calculate them is similar. We will only state main results without details, since the proofs would have been overloaded with unnecessary notations and minor differences. When necessary, we will point out the differences.

6.1 Hermitian Toeplitz band matrices

The case is very similar to real symmetric Toeplitz band matrices, except that we now consider \( n \)-dimensional complex Hermitian matrices \( T_n = (\eta_{ij} a_{i-j})_{i,j=1}^{n} \). We assume that \( \text{Re} a_j = \text{Re} a_{-j} \) and \( \text{Im} a_j = -\text{Im} a_{-j} \) for \( j = 1, 2, \ldots \), and \( \{\text{Re} a_j, \text{Im} a_j\}_{j \in \mathbb{N}} \) is a sequence of independent real random variables such that

\[ E[|a_j|^2] = 1 \quad \text{and} \quad E[a_j^2] = 0 \quad \text{for} \quad j \in \mathbb{N}, \]  

(6.2)

(homogeneity of 4-th moments)

\[ \kappa = E[|a_j|^4], \]  

(6.3)

and further (uniform boundedness)

\[ \sup_{j \in \mathbb{Z}} E[|a_j|^k] = C_k < \infty \quad \text{for} \quad k \in \mathbb{N}. \]  

(6.4)

In addition, we also assume \( a_0 \equiv 0 \) and the bandwidth \( b_n \to \infty \) but \( b_n/n \to b \in [0,1] \).

**Theorem 6.1.** With above assumptions and notations, Theorem 1.1 also holds for random Hermitian Toeplitz band matrices.

**Remark 6.2.** The distribution of eigenvalues for this case has been proved to be the same as the real case in [21]. The covariance like in (4.2) is slightly different from the real case because \( E[a_j^2] = 0 \), which is given by

\[ \sum_{\pi \in \mathcal{P}_2(p,q)} f_1^{-}(\pi) + (\kappa - 1) \sum_{\pi \in \mathcal{P}_{2,4}(p,q)} (f_1^{-}(\pi) + f_1^{+}(\pi)). \]  

(6.5)
6.2 Hankel band matrices

A Hankel matrix $H_n = (h_{i+j-1})_{i,j=1}^n$ is closely related to a Toeplitz matrix. Explicitly, let $P_n = (\delta_{i-j,n-j})_{i,j=1}^n$ be the “backward identity” permutation, then for a Toeplitz matrix of the form $T_n = (a_{i-j})_{i,j=1}^n$ and a Hankel matrix of the form $H_n = (h_{i+j-1})_{i,j=1}^n$, $P_n T_n$ is a Hankel matrix and $P_n H_n$ is a Toeplitz matrix. In this paper we always write a Hankel band matrix $H_n = P_n T_n$ where $T_n = (\eta_j a_{i-j})_{i,j=1}^n$ is a Toeplitz band matrix with bandwidth $b_n$ and the matrix entries $a_{-n+1}, \ldots, a_0, \ldots, a_{n-1}$ are real-valued, thus $H_n$ is a real symmetric matrix.

For Hankel band matrices, as in Toeplitz case we also have a trace formula and its derivation is similar; see [21].

**Lemma 6.3.** \( \text{tr}(H_n^p) = \)

\[
\begin{cases}
\sum_{i=1}^{n} \frac{b_n}{p} \prod_{l=1}^{p} a_{j_l} \prod_{l=1}^{p} X(1,n)(i - \sum_{q=1}^{l} (-1)^q j_q) \delta_{0, \sum_{q=1}^{p} (-1)^q j_q}, & \text{p even;} \\
\sum_{i=1}^{n} \sum_{j_p=b_n} b_n \prod_{l=1}^{p} a_{j_l} \prod_{l=1}^{p} X(1,n)(i - \sum_{q=1}^{l} (-1)^q j_q) \delta_{2p-1-n, \sum_{q=1}^{p} (-1)^q j_q}, & \text{p odd.}
\end{cases}
\]

From this trace formula, our method can apply to the case that $p$ is even. We consider random Hankel matrices satisfying the following assumptions: \( \{a_j : j \in \mathbb{Z}\} \) is a sequence of independent real random variables such that

\[
E[a_j] = 0, \ E[|a_j|^2] = 1 \quad \text{for} \quad j \in \mathbb{Z},
\]

(homogeneity of 4-th moments)

\[
\kappa = E[|a_j|^4], \quad j \in \mathbb{Z}
\]

and further (uniform boundedness)

\[
\sup_{j \in \mathbb{Z}} E[|a_j|^k] = C_k < \infty \quad \text{for} \quad k \geq 3.
\]

In addition, we also assume the bandwidth $b_n \to \infty$ but $b_n/n \to b \in [0, 1]$ as $n \to \infty$.

**Theorem 6.4.** Let $H_n$ be a real symmetric ((6.6)-(6.8)) random Hankel band matrix with the bandwidth $b_n$, where $b_n/n \to b \in [0, 1]$ but $b_n \to \infty$ as $n \to \infty$. Set $A_n = H_n/\sqrt{b_n}$ and

\[
\zeta_p = \frac{\sqrt{b_n}}{n} \left( \text{tr}(A_n^{2p}) - E[\text{tr}(A_n^{2p})] \right).
\]

Then

\[
\zeta_p \longrightarrow N(0, \sigma_p^2)
\]

in distribution as $n \to \infty$. Moreover, for a given polynomial

\[
Q(x) = \sum_{j=0}^{p} q_j x^j
\]

with degree $p \geq 1$, set

\[
\zeta_Q = \frac{\sqrt{T_n}}{n} \left( \text{tr}Q(A_n^2) - E[\text{tr}Q(A_n^2)] \right).
\]

we also have

\[
\zeta_Q \longrightarrow N(0, \sigma_Q^2)
\]

in distribution as $n \to \infty$. Here the variances $\sigma_p^2$ and $\sigma_Q^2$ will be given below.
We remark that the limit of \( \frac{1}{n} \text{tr}(H_n/\sqrt{b_n})^p \) can be calculated in the same way as in Toeplitz case, see also [21]. Next, we derive briefly the variances \( \sigma^2 \) and \( \tilde{\sigma}^2 \).

By Lemma 6.3, rewrite

\[
\zeta_p = \frac{1}{nb_n} \sum_{i=1}^{2p} \sum_{J} I_J \left( a_J - E[a_J] \right) \delta_{0, \sum_{i=1}^{2p} (-1)^{j_i}}. \tag{6.14}
\]

Here \( J = (j_1, \ldots, j_{2p}) \in \{-b_n, \ldots, b_n\}^{2p}, a_J = \prod_{i=1}^{2p} a_{j_i}, I_J = \prod_{k=1}^{2} \chi_{[1,n]}(i - \sum_{l=1}^{k} (-1)^{j_l}) \) and the summation \( \sum_J \) runs over all possible \( J \in \{-b_n, \ldots, b_n\}^{2p} \). We call a vector \( J = (j_1, \ldots, j_{2p}) \in \{-b_n, \ldots, b_n\}^{2p} \) is balanced if

\[
\sum_{i=1}^{2p} (-1)^{j_i} = 0. \tag{6.15}
\]

Since \( a_j \) and \( a_{-j} \) are independent when \( j \neq 0 \), \( S_J \), the projection of \( J \), should be \( J \) itself (forget the order of its components).

The only difference is the definition of balance and projection. Using the same technique in section 4, we get

\[
E[\zeta_p \zeta_q] \longrightarrow \tilde{\sigma}_{p,q} = \sum_{\pi \in \mathcal{P}_2(2p,2q)} g_l(\pi) + (\kappa - 1) \sum_{\pi \in \mathcal{P}_{2,4}(2p,2q)} g_{1f}(\pi) \tag{6.16}
\]

as \( n \rightarrow \infty \).

Here for a fixed pair partition \( \pi \), we construct a projective relation between two groups of unknowns \( y_1, \ldots, y_{2p+2q} \) and \( x_1, \ldots, x_{p+q} \) as follows:

\[
y_i = y_j = x_{\pi(i)} \tag{6.17}
\]

whenever \( i \sim_{\pi} j \).

If \( \pi \in \mathcal{P}_2(2p,2q) \), let

\[
g_l(\pi) = \int_{[0,1]^2 \times [-1,1]^{2p+q}} \delta \left( \sum_{i=1}^{2p} (-1)^{j_i} y_i \right) \left\{ \begin{array}{c}
\sum_{i=1}^{2p} (-1)^{j_i} y_i = 0 \\
\sum_{i=2p+1}^{2p+q} (-1)^{j_i} = 0
\end{array} \right\} \prod_{j=1}^{2p} \chi_{[0,1]}(x_0 - b \sum_{i=1}^{j} (-1)^{j_i}) \prod_{j=2p+1}^{2p+q} \chi_{[0,1]}(y_0 - b \sum_{i=2p+1}^{j} (-1)^{j_i}) \prod_{l=0}^{p+q} dx_l. \tag{6.18}
\]

If \( \pi \in \mathcal{P}_{2,4}(2p,2q) \), let

\[
g_{1f}(\pi) = \int_{[0,1]^2 \times [-1,1]^{2p+q+1}} \chi \left\{ \sum_{i=1}^{2p} (-1)^{j_i} y_i = 0 \right\} \chi \left\{ \sum_{i=2p+1}^{2p+q} (-1)^{j_i} = 0 \right\} \prod_{j=1}^{2p} \chi_{[0,1]}(x_0 - b \sum_{i=1}^{j} (-1)^{j_i}) \prod_{j=2p+1}^{2p+q} \chi_{[0,1]}(y_0 - b \sum_{i=2p+1}^{j} (-1)^{j_i}) \prod_{l=0}^{p+q-1} dx_l. \tag{6.19}
\]

Because of the existence of the characteristic function in the integrals of type I and II, we see that \( g_l(\pi) \neq 0 \) if and only if \( \sum_{i=2p+1}^{2p+q} (-1)^{j_i} = 0 \) when \( \sum_{i=1}^{2p} (-1)^{j_i} = 0 \). Denote the subset of \( \mathcal{P}_2(2p,2q) \) consisting of this kind of \( \pi \) by \( \mathcal{P}_2^I(2p,2q) \).
Remark 6.6. The derivation of higher moments can be calculated in the same way the variance \( \tilde{\sigma}^2 \) since \( 2p \) (the existence of \( \gamma \) as in section 5. Careful examination shows that the argument in case III in section 5 is still valid so that one needn’t care much about the definition of balance. Moreover, \( j+k \) is odd provided \( j, k \notin V_i \) and \( i \sim \pi j \).

From the discussion above, we get

\[
E[\zeta_p \zeta_q] \rightarrow \tilde{\sigma}_{p,q} = \sum_{\pi \in \mathcal{P}_2(2p, 2q)} g_{11}(\pi) + (\kappa - 1) \sum_{\pi \in \mathcal{P}_2^1(2p, 2q)} g_{11}(\pi),
\]

(6.20)

and

\[
\tilde{\sigma}_{p,p}^2 = \tilde{\sigma}_{p,p}
\]

(6.21)

and

\[
\tilde{\sigma}_{Q_i}^2 = \sum_{i=1}^{p} \sum_{j=1}^{p} q_{ij} \tilde{\sigma}_{i,j}.
\]

(6.22)

Similar to Corollary 1.2, we get another central limit theorem for product of independent random variables.

**Corollary 6.5.** Suppose that \( \{a_j : j \in \mathbb{Z}\} \) is a sequence of independent random variables satisfying the assumptions (6.6)–(6.8). For every \( p \geq 1, \)

\[
\frac{1}{k^{2p-1}} \sum_{j_1, \ldots, j_{2p} = -k}^{k} \left( \prod_{i=1}^{2p} a_{j_i} - E[\prod_{i=1}^{2p} a_{j_i}] \right) \delta_{0, \sum_{i=1}^{2p} (-1)^{j_i}}
\]

(6.23)

converges in distribution to a Gaussian distribution \( N(0, \tilde{\sigma}_p^2) \) as \( k \) tends to infinity. Here the variance \( \tilde{\sigma}_p^2 \) corresponds to the case \( b = 0 \) above.

**Remark 6.6.** The derivation of higher moments can be calculated in the same way as in section 5. Careful examination shows that the argument in case III in section 5 (the existence of \( \gamma \)) is the only part depending on the definition of balance. However, since \( 2p \) and \( 2q \) are even, the dimensions of all vectors involved are even. Remark 5.5 is still valid so that one needn’t care much about the definition of balance. Moreover, \( a_0 \) needn’t equal to 0.

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