Asymptotic Distribution of Degree–Based Topological Indices

Mingao Yuan

^aDepartment of Statistics, North Dakota State University, USA mingao.yuan@ndsu.edu

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Abstract

Topological indices play a significant role in mathematical chemistry. Given a graph \mathcal{G} with vertex set $\mathcal{V} = \{1, 2, \ldots, n\}$ and edge set \mathcal{E} , let d_i be the degree of node i. The degree-based topological index is defined as $\mathcal{I}_n = \sum_{\{i,j\} \in \mathcal{E}} f(d_i, d_j)$, where f(x, y) is a symmetric function. In this paper, we investigate the asymptotic distribution of the degree-based topological indices of a heterogeneous Erdős-Rényi random graph. We show that after suitably centered and scaled, the topological indices converges in distribution to the standard normal distribution. Interestingly, we find that the general Randić index with $f(x, y) = (xy)^{\tau}$ for a constant τ exhibits a phase change at $\tau = -\frac{1}{2}$.

1 Introduction

Topological index is a numerical parameter of a graph. It is graph invariant and characterizes the topology of a graph. Topological indices are used to model many physicochemical properties in QSAR [1,4,12]. One of the most important types of topological indices is the degree-based topological indices, which are defined as a function of the degrees of nodes in a graph [12].

The first degree-based topological index is the Randić index [20]. It measures the branching extent of a graph. The Randić index is the

most popular and most studied index among all topological indices. It plays a central role in understanding quantitative structure-property and structure-activity relations in chemistry and pharmocology [21,22]. Moreover, the Randić index possesses a wealth of non-trivial and interesting mathematical properties [2,3,5,6,15]. In addition, the Randić index finds countless applications in network (graph) data analysis. For instance, it was used to quantify the similarity between different networks or subgraphs of the same network [11], it serves as a quantitative characterization of network heterogeneity [10], and [8,9] used the Randić index to measure graph robustness.

Motivated by the Randić index, various degree-based topological indices have been introduced and attracted great interest in the past years [16]. For example, [2] proposed the general Randić index, which includes the Randić index and the second Zagreb index as special cases. [26, 27] defined the general sum-connectivity index, which includes the harmonic index as a special case [28]. [23,24] introduced the inverse sum indeg index to predict the total surface area of octane isomers. The reader is referred to [12] for more references.

One of the interesting research topics in the study of topological indices is to investigate their properties of random graphs. Recently, [17,18] performed numeric and analytic analyses of the Randić index and the harmonic index in the Erdős-Rényi random graph. Their simulation studies show that the indices are approximately equal to one half of the number of nodes, and the distributions of the indices are symmetric around their expectations. [7,8,14] calculated the expectations of the Randić index, the generalized Zagreb indices and two modified Zagreb indices of the Erdős-Rényi random graph, respectively. [25] derived the asymptotic limits of the Randić index and the harmonic index of a heterogeneous random graph.

In this paper, we are interested in the asymptotic distribution of the degree-based topological index of a heterogeneous Erdős-Rényi random graph. We show that after suitably centered and scaled, the degree-based topological index converges in distribution to the standard normal distribution. We apply our results to several well-known topological indices and observe that the general Randić index exhibits a phase change phe-

nomenon.

This paper is organized as follows. In Section 2, we present the asymptotic distribution of the topological index of a heterogeneous random graph. In Section 3, we provide several examples. The proof is deferred to Section 4.

Notation: We adopt the Bachmann–Landau notation throughout this paper. Let a_n and b_n be two positive sequences. Denote $a_n = \Theta(b_n)$ if $c_1b_n \leq a_n \leq c_2b_n$ for some positive constants c_1, c_2 . Denote $a_n =$ $\omega(b_n)$ if $\lim_{n\to\infty} \frac{a_n}{b_n} = \infty$. Denote $a_n = O(b_n)$ if $a_n \leq cb_n$ for some positive constants c. Denote $a_n = o(b_n)$ if $\lim_{n\to\infty} \frac{a_n}{b_n} = 0$. Let $\mathcal{N}(0,1)$ be the standard normal distribution and X_n be a sequence of random variables. Then $X_n \Rightarrow \mathcal{N}(0,1)$ means X_n converges in distribution to the standard normal distribution as n goes to infinity. Denote $X_n = O_P(a_n)$ if $\frac{X_n}{a_n}$ is bounded in probability. Denote $X_n = o_P(a_n)$ if $\frac{X_n}{a_n}$ converges to zero in probability as n goes to infinity. Let $\mathbb{E}[X_n]$ and $Var(X_n)$ denote the expectation and variance of a random variable X_n respectively. $\mathbb{P}[E]$ denote the probability of an event E. Let f = f(x, y) be a function. For non-negative integers $s, t, f^{(s,t)} = f^{(s,t)}(x, y)$ denote the partial derivative $\frac{\partial^{s+t} f(x,y)}{\partial x^s \partial y^t}$. For convenience, sometimes we write $f_x = f^{(1,0)}, f_y = f^{(0,1)},$ $f_{xx} = f^{(2,0)}, f_{yy} = f^{(0,2)}$ and $f_{xy} = f^{(1,1)}$. exp(x) denote the exponential function e^x . For positive integer n, denote $[n] = \{1, 2, ..., n\}$. Given a finite set E, |E| represents the number of elements in E.

2 Main results

A graph consists of a set of nodes (vertices) and a set of edges. Given a positive integer n, an *undirected* graph on $\mathcal{V} = [n]$ is a pair $\mathcal{G} = (\mathcal{V}, \mathcal{E})$, where \mathcal{E} is a collection of subsets of \mathcal{V} such that |e| = 2 for every $e \in \mathcal{E}$. Elements in \mathcal{E} are called edges. A graph can be conveniently represented as an adjacency matrix A, where $A_{ij} = 1$ if $\{i, j\}$ is an edge, $A_{ij} = 0$ otherwise and $A_{ii} = 0$. Since \mathcal{G} is undirected, the adjacency matrix A is symmetric. The degree d_i of node i is the number of edges connecting it, that is, $d_i = \sum_{j \neq i} A_{ij}$. A graph is said to be random if $A_{ij} (1 \leq i < j \leq n)$ are random.

The degree-based topological index of a graph is defined as follows [1, 12].

Definition 1. The degree-based topological index of a graph $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ is defined as

$$\mathcal{I}_n = \sum_{\{i,j\}\in\mathcal{E}} f(d_i, d_j),\tag{1}$$

where f(x, y) is a real function satisfying f(x, y) = f(y, x).

Many well-known topological indices can be expressed as (1). For example, the Randić index corresponds to $f(x, y) = (xy)^{-\frac{1}{2}}$ and the hyper-Zagreb index corresponds to $f(x, y) = (x + y)^2$.

Definition 2. Let β be a constant between zero and one, n be an positive integer, and

$$W = \{ w_{ij} \in [\beta, 1] | 1 \le i < j \le n, w_{ji} = w_{ij}, w_{ii} = 0 \}.$$

Define a heterogeneous random graph $\mathcal{G}_n(\beta, W)$ as

$$\mathbb{P}(A_{ij}=1) = p_n w_{ij},$$

where A_{ij} $(1 \le i < j \le n)$ are independent, $A_{ij} = A_{ji}$ and $p_n \in (0, 1)$.

The expected degree of node i in $\mathcal{G}_n(\beta, W)$ is $\mathbb{E}[d_i] = \sum_{k \neq i} p_n w_{ik}$. In general, $\mathbb{E}[d_i] \neq \mathbb{E}[d_j]$ if $i \neq j$, that is, the expected degrees of nodes are not the same. Hence $\mathcal{G}_n(\beta, W)$ is a heterogeneous random graph. When $w_{ij} = c$ $(1 \leq i < j \leq n)$ for a constant $c \in (0, 1)$, $\mathcal{G}_n(\beta, W)$ is the Erdős-Rényi random graph. It is homogeneous in the sense that nodes in it share the same expected degree.

Several recent works have studied the expectations of some special topological indices of the Erdős-Rényi random graph [7,8,14,17,18]. In this paper, we derive the asymptotic distribution of the topological index \mathcal{I}_n of the heterogeneous random graph $\mathcal{G}_n(\beta, W)$. Our results can be applied to all the topological indices studied in [7,8,14,17,18].

Before presenting our results, we introduce several notations and as-

sumptions. Let $w_{i(k)} = 1 + \sum_{l \notin \{i,k\}} p_n w_{il}$ and

$$\sigma_n^2 = \sum_{i < j} (a_{ij} + a_{ji})^2 p_n w_{ij} (1 - p_n w_{ij}), \qquad (2)$$

where

$$a_{ij} = \frac{1}{2}f(w_{i(j)}, w_{j(i)}) + \frac{1}{2}\sum_{l \notin \{i, j\}} p_n w_{il} \big[f_x(w_{i(l)}, w_{l(i)}) + f_y(w_{i(l)}, w_{l(i)}) \big].$$

Assumption 1. Let $k_0(k_0 \ge 3)$, s,t be non-negative integers. Suppose $np_n = \omega(\log(n))$ and the following conditions hold. (C1)

$$\sum_{i < j} (a_{ij} + a_{ji})^4 p_n = o(\sigma_n^4).$$

(C2). For all non-negative integers s, t satisfying $s + t \le k_0$, there is some positive constant C such that

$$|f^{(s,t)}(x,y)| \le (xy)^C.$$

(C3). Given s,t satisfying $s + t = k_0$, $|f^{(s,t)}(x,y)|$ is monotone in x and y.

(C4). For a large positive constant M and positive sequences $a_n, b_n \in [(\log(np_n))^{-2}, M]$, the following holds. For $s + t = k_0$,

$$n(np_n)^{\frac{\kappa_0}{2}+1}|f^{(s,t)}(a_nnp_n,b_nnp_n)| = o(\sigma_n).$$

(C5). For $1 \le s + t \le k_0 - 1$,

$$n(np_n)^{2(s+t)-1}|f^{(s,t)}(np_n, np_n)|^2 = o(\sigma_n^2).$$

Assumption 1 is not restrictive and many common degree-based topological indices satisfy this assumption as shown later. Under Assumption 1, we derive the asymptotic distribution of the topological index \mathcal{I}_n of $\mathcal{G}_n(\beta, W)$ as follows. **Theorem 1.** Let \mathcal{I}_n be the topological index defined in (1) of the random graph $\mathcal{G}_n(\beta, W)$ and σ_n^2 be defined in (2). Suppose Assumption 1 holds. Then

$$\frac{\mathcal{I}_n - \mathbb{E}[\mathcal{I}_n]}{\sigma_n} \Rightarrow \mathcal{N}(0, 1), \tag{3}$$

as n goes to infinity. In addition, the expectation $\mathbb{E}[\mathcal{I}_n]$ has the following asymptotic expression

$$\mathbb{E}[\mathcal{I}_n] = \left(1 + O\left(\frac{1}{np_n}\right)\right) \sum_{1 \le i < j \le n} p_n w_{ij} f(w_{i(j)}, w_{j(i)}), \tag{4}$$

where the error rate $\frac{1}{np_n}$ cannot be improved.

Based on Theorem 1, the degree-based topological index \mathcal{I}_n (suitably centered and scaled) of the heterogeneous random graph $\mathcal{G}_n(\beta, W)$ converges in distribution to the standard normal distribution. As far as we know, this is the first theoretical result on limiting distribution of topological indices. Moreover, Theorem 1 provides the best approximation of the expectation of \mathcal{I}_n , in the sense that the error rate $\frac{1}{np_n}$ cannot be improved. For some special topological indices of the Erdős-Rényi random graph, it is possible to get an exact and compact expression of $\mathbb{E}[\mathcal{I}_n]$. For instance, [14] and [7] obtained the exact expressions of the expectation of the hyper-Zagreb index and the forgotten topological index of the Erdős-Rényi random graph respectively. However, for most topological indices, it seems impossible to get exact and closed-form expressions of the expectations [7]. Our result (4) provides an approximation of the expectations.

The proof of Theorem 1 proceeds by decomposing \mathcal{I}_n as a sum of leading term and remainder term, followed by finding the limiting distribution of the leading term and showing the remainder term is negligible. The condition (C1) of Assumption 1 is used to prove the leading term converges in distribution to the standard normal distribution. The conditions (C2)–(C5) are needed to bound the remainder term. The condition $np_n = \omega(\log(n))$ requires the random graph to be relatively dense. This condition is common in theoretical network analysis. Assumption 1 is weak. We shall provide several examples of degree–based topological indices that satisfy this assumption in the subsequent section.

3 Application to several topological indices

In this section, we apply Theorem 1 to several well-known topological indices of a special heterogeneous random graph.

Let $p_n = n^{-\alpha}$ for a constant $\alpha \in (0,1)$ and $w_{ij} = e^{-\frac{\kappa i}{n}} e^{-\frac{\kappa j}{n}}, (i \neq j)$ with non-negative constant κ . Then $e^{-2\kappa} \leq w_{ij} \leq 1$. Denote the corresponding random graph as $\mathcal{G}_n(\alpha, \kappa)$. When $\kappa = 0$, $\mathcal{G}_n(\alpha, 0)$ is the Erdős-Rényi random graph. In this case, we denote it as $\mathcal{G}_n(\alpha)$ for convenience. For $\kappa > 0$, $\mathcal{G}_n(\alpha, \kappa)$ is heterogeneous.

Denote $c(\kappa) = \frac{1-e^{-\kappa}}{\kappa}$ for $\kappa > 0$ and c(0) = 1. Note that

$$\sum_{i=1}^{n} e^{-\frac{\kappa i}{n}} = \frac{e^{-\frac{\kappa}{n}}(1-e^{-\kappa})}{1-e^{-\frac{\kappa}{n}}} = nc(\kappa) + O(1), \qquad \kappa > 0.$$

Then

$$\begin{split} w_{i(k)} &= 1 + \sum_{l \notin \{i,k\}} p_n w_{il} \\ &= 1 + p_n e^{-\frac{\kappa i}{n}} \left(nc(\kappa) + O(1) - e^{-\frac{\kappa i}{n}} - e^{-\frac{\kappa k}{n}} \right) \\ &= 1 + np_n c(\kappa) e^{-\frac{\kappa i}{n}} + O(p_n) \\ &= np_n c(\kappa) e^{-\frac{\kappa i}{n}} + O(1), \quad \kappa > 0. \end{split}$$

When $\kappa = 0$,

$$w_{i(k)} = 1 + \sum_{l \notin \{i,k\}} p_n = 1 + (n-2)p_n.$$

3.1 The general Randić index

The general Randić index is a generalization of the well-known Randić index and has been widely studied in literature [8,17,18]. Let $f(x,y) = (xy)^{\tau}$ for a non-zero constant τ . The general Randić index \mathcal{I}_n is defined as

$$\mathcal{I}_n = \sum_{\{i,j\}\in\mathcal{E}} (d_i d_j)^{\tau}.$$
(5)

When $\tau = -\frac{1}{2}$, \mathcal{I}_n is the Randić index.

Given non-negative integers s, t, straightforward computation yields

$$f^{(s,t)}(x,y) = \left(\prod_{k=0}^{s-1} (\tau-k)\right) \left(\prod_{k=0}^{t-1} (\tau-k)\right) x^{\tau-s} y^{\tau-t}.$$

Then

$$\begin{aligned} a_{ij} \\ &= \frac{1}{2} \left[(np_n)^2 c(\kappa)^2 e^{-\frac{\kappa i}{n}} e^{-\frac{\kappa j}{n}} + O\left(np_n\right) \right]^{\tau} \\ &+ \frac{1}{2} \sum_{l \notin \{i,j\}} p_n e^{-\frac{\kappa i}{n}} e^{-\frac{\kappa l}{n}} \tau \left(np_n c(\kappa) e^{-\frac{\kappa i}{n}} + O\left(1\right) \right)^{\tau-1} \\ &\times \left(np_n c(\kappa) e^{-\frac{\kappa l}{n}} + O\left(1\right) \right)^{\tau} \\ &+ \frac{1}{2} \sum_{l \notin \{i,j\}} p_n e^{-\frac{\kappa i}{n}} e^{-\frac{\kappa l}{n}} \tau \left(np_n c(\kappa) e^{-\frac{\kappa i}{n}} + O\left(1\right) \right)^{\tau} \\ &\times \left(np_n c(\kappa) e^{-\frac{\kappa l}{n}} + O\left(1\right) \right)^{\tau-1}. \end{aligned}$$

Note that $e^{-\kappa} \leq e^{-\frac{\kappa i}{n}} \leq 1$. If $\tau > 0$, then $a_{ij} = \Theta((np_n)^{2\tau})$. In this case,

$$\sigma_n^2 = \sum_{i < j} (a_{ij} + a_{ji})^2 p_n = \Theta \left(n(np_n)^{4\tau + 1} \right).$$

Hence

$$\frac{\sum_{i < j} (a_{ij} + a_{ji})^4 p_n}{\sigma_n^4} = O\left(\frac{n(np_n)^{8\tau+1}}{n^2(np_n)^{8\tau+2}}\right) = o(1).$$

For $1 \leq s+t$,

$$n(np_n)^{2(s+t)}|f^{(s,t)}(np_n, np_n)|^2 = \Theta\left(n(np_n)^{4\tau}\right) = o\left(\sigma_n^2\right)$$

Let $k_0 = \max\left\{\lfloor 1 + \frac{1}{1-\alpha} \rfloor + 1, 3\right\}$. Then $k_0 > 1 + \frac{1}{1-\alpha}$. For $s + t = k_0$, it is easy to verify that

$$n(np_n)^{\frac{k_0}{2}+1}|f^{(s,t)}(np_n,np_n)| = \Theta\left(n(np_n)^{1+2\tau-\frac{k_0}{2}}\right) = o(\sigma_n).$$

Then Assumption 1 holds and Theorem 1 applies.

Corollary 1. Let \mathcal{I}_n be the general Randić index defined in (5) of the random graph $\mathcal{G}_n(\alpha, \kappa)$ and σ_n^2 be defined in (2). If $\tau > 0$, then

$$\frac{\mathcal{I}_n - \mathbb{E}[\mathcal{I}_n]}{\sigma_n} \Rightarrow \mathcal{N}(0, 1), \tag{6}$$

as n goes to infinity. In addition, the expectation $\mathbb{E}[\mathcal{I}_n]$ has the following asymptotic expression

$$\mathbb{E}[\mathcal{I}_n] = \left(1 + O\left(\frac{1}{np_n}\right)\right) \sum_{1 \le i < j \le n} p_n w_{ij} (w_{i(j)} w_{j(i)})^{\tau}, \tag{7}$$

where the error rate $\frac{1}{np_n}$ cannot be improved.

When $\tau < 0$, Assumption 1 may not hold. To see this, let $\kappa = 0$. Then

$$a_{ij} = \frac{(1+(n-2)p_n)^{2\tau}}{2} + \tau (1+(n-2)p_n)^{2\tau-1}(n-2)p_n$$
$$= (1+(n-2)p_n)^{2\tau} \left(\frac{1+2\tau}{2} - \frac{\tau}{1+(n-2)p_n}\right).$$

If $\tau \neq -\frac{1}{2}$, then $a_{ij} = \Theta((np_n)^{2\tau})$. Similar to the case $\tau > 0$, Assumption 1 holds and Theorem 1 applies.

If
$$\tau = -\frac{1}{2}$$
, then $a_{ij} = \Theta\left(\frac{1}{(np_n)^2}\right)$. In this case, $\sigma_n^2 = \Theta\left(\frac{n}{(np_n)^3}\right)$ and
$$n(np_n)^{2(s+t)-1}|f^{(s,t)}(np_n, np_n)|^2 = \Theta\left(\frac{n}{(np_n)^3}\right).$$

Clearly, $\frac{n}{(np_n)^3} \neq o(\sigma_n^2)$. Hence Assumption 1 does not hold. We have to study this case separately.

When $\tau = -\frac{1}{2}$, \mathcal{I}_n is the well-known Randić index. The Randić index is perhaps the first degree-based topological index introduced in [20]. Recently, [17] performed simulation studies of the expectation and distribution of the Randić index in Erdős-Rényi random graph. Its asymptotic limit was given in [25]. Here we derive the asymptotic distribution of the Randić index of the Erdős-Rényi random graph as follows. **Theorem 2.** Let \mathcal{I}_n be the general Randić index of the Erdős-Rényi random graph $\mathcal{G}_n(\alpha)$ with constant $\alpha \in (0, 1)$. Then

$$\frac{\mathcal{I}_n - \mathbb{E}[\mathcal{I}_n]}{\sigma_n} \Rightarrow \mathcal{N}(0, 1), \tag{8}$$

where

$$\mathbb{E}[\mathcal{I}_n] = \frac{n(np_n)^{2\tau+1}}{2} \left(1 + O\left(\frac{1}{np_n}\right) \right),$$

$$\sigma_n^2 = \frac{n(n-1)(n-2)p_n^2(1-p_n)^2}{32(1+(n-2)p_n))^4} = \Theta\left(\frac{n}{(np_n)^2}\right),$$

for $\tau = -\frac{1}{2}$ and

$$\sigma_n^2 = \frac{(1+2\tau)^2}{2}n(n-1)p_n(1+(n-2)p_n)^{4\tau}(1+o(1)) = \Theta\left(n(np_n)^{4\tau+1}\right),$$

for $\tau \neq -\frac{1}{2}$.

Based on Theorem 2, we have an interesting finding. For $\tau \neq -\frac{1}{2}$, the order of σ_n^2 is $n(np_n)^{4\tau+1}$, while for $\tau = -\frac{1}{2}$, the order of σ_n^2 is $\frac{n}{(np_n)^2}$. Note that

$$n(np_n)^{4(-\frac{1}{2})+1} = \frac{n}{np_n} = \omega\left(\frac{n}{(np_n)^2}\right).$$

Therefore, as a function of τ , the order of σ_n^2 is continuous at $\tau \neq -\frac{1}{2}$, but discontinuous at $\tau = -\frac{1}{2}$. In this sense, the general Randić index exhibits a phase change at $\tau = -\frac{1}{2}$.

3.2 Hyper–Zagreb index

The Zagreb indices and its variants are frequently used to measure physical-chemical properties of compounds. Recently, [14] studied the expectation of the hyper-Zagreb index of the Erdős-Rényi random graph. Let $f(x, y) = (x + y)^2$. The hyper-Zagreb index \mathcal{I}_n is defined as

$$\mathcal{I}_n = \sum_{\{i,j\}\in\mathcal{E}} (d_i + d_j)^2.$$
(9)

Clearly, $f^{(s,t)}(x,y) = 0$ for $s + t \ge 3$ and

$$f_x(x,y) = f_y(x,y) = 2(x+y), \quad f_{xx}(x,y) = f_{yy}(x,y) = f_{xy}(x,y) = 2.$$

Straightforward calculation yields

$$\sum_{\substack{l \notin \{i,j\}}} w_{il} \left[f_x(w_{i(l)}, w_{l(i)}) + f_y(w_{i(l)}, w_{l(i)}) \right]$$

$$= 4e^{-\frac{\kappa i}{n}} \sum_{\substack{l \notin \{i,j\}}} e^{-\frac{\kappa l}{n}} \left(np_n c(\kappa) e^{-\frac{\kappa i}{n}} + np_n c(\kappa) e^{-\frac{\kappa l}{n}} + O\left(1\right) \right)$$

$$= 4e^{-\frac{\kappa i}{n}} \left[n^2 p_n c(\kappa)^2 e^{-\frac{\kappa i}{n}} + n^2 p_n c(\kappa) c(2\kappa) + O(n) \right]$$

$$= 4n^2 p_n c(\kappa)^2 e^{-\frac{2\kappa i}{n}} + 4n^2 p_n c(\kappa) c(2\kappa) e^{-\frac{\kappa i}{n}} + O(n).$$

Since $e^{-\kappa} \le e^{-\frac{\kappa i}{n}} \le 1$ and $e^{-2\kappa} \le e^{-\frac{2\kappa i}{n}} \le 1$, then

$$\begin{aligned} a_{ij} &= \frac{(np_n)^2 c(\kappa)^2}{2} \left(e^{-\frac{\kappa i}{n}} + e^{-\frac{\kappa j}{n}} \right)^2 \\ &+ 2(np_n)^2 c(\kappa)^2 e^{-\frac{2\kappa i}{n}} + 2(np_n)^2 c(\kappa) c(2\kappa) e^{-\frac{\kappa i}{n}} + O(np_n) \\ &= (np_n)^2 \left[\frac{5}{2} c(\kappa)^2 e^{-\frac{2\kappa i}{n}} + \frac{1}{2} c(\kappa)^2 e^{-\frac{2\kappa j}{n}} \\ &+ c(\kappa)^2 e^{-\frac{\kappa i}{n}} e^{-\frac{\kappa j}{n}} + 2c(\kappa) c(2\kappa) e^{-\frac{\kappa i}{n}} \right] + O(np_n) \\ &= \Theta\left((np_n)^2 \right). \end{aligned}$$

Consequently, we have $\sigma_n^2 = \Theta\left(n(np_n)^5\right)$ and

$$\frac{\sum_{i < j} (a_{ij} + a_{ji})^4 p_n}{\sigma_n^4} = O\left(\frac{n(np_n)^9}{n^2(np_n)^{10}}\right) = o(1).$$

For $s + t \ge 3$, $f^{(s,t)}(x,y) = 0$. For s + t = 2, we have

$$n(np_n)^4 = o\left(\sigma_n^2\right).$$

Then Assumption 1 holds and Theorem 1 applies.

Corollary 2. Let \mathcal{I}_n be the hyper-Zagreb index defined in (9) of the ran-

dom graph $\mathcal{G}_n(\alpha,\kappa)$ and σ_n^2 be defined in (2). Then

$$\frac{\mathcal{I}_n - \mathbb{E}[\mathcal{I}_n]}{\sigma_n} \Rightarrow \mathcal{N}(0, 1), \tag{10}$$

as n goes to infinity. In addition, the expectation $\mathbb{E}[\mathcal{I}_n]$ has the following asymptotic expression

$$\mathbb{E}[\mathcal{I}_n] = \left(1 + O\left(\frac{1}{np_n}\right)\right) \sum_{1 \le i < j \le n} p_n w_{ij} (w_{i(j)} + w_{j(i)})^2, \qquad (11)$$

where the error rate $\frac{1}{np_n}$ cannot be improved.

By Theorem 1 in [14], the expectation of the hyper-Zagreb index of the Erdős-Rényi random graph $\mathcal{G}_n(\alpha)$ is equal to

$$\mathbb{E}[\mathcal{I}_n] = n(n-1)(n-2)(2n-5)p_n^3 + 5n(n-1)(n-2)p_n^2 + 2n(n-1)p_n.$$
(12)

By (11), we have

$$\mathbb{E}[\mathcal{I}_n] = 2n(n-1)(n-2)^2 p_n^3 \left(1 + O\left(\frac{1}{np_n}\right)\right).$$
(13)

Then our result (13) is consistent with (12). In addition, by (12), the error rate $\frac{1}{np_n}$ cannot be improved, as stated in Corollary 2.

3.3 Forgotten topological index

The forgotten topological index is another chemical index. [7] studied the expectation of the forgotten topological index of the Erdős-Rényi random graph. Let $f(x, y) = x^2 + y^2$. The forgotten topological index \mathcal{I}_n is defined as

$$\mathcal{I}_n = \sum_{\{i,j\}\in\mathcal{E}} (d_i^2 + d_j^2).$$
(14)

Similar to the hyper-Zagreb index, it is easy to verify that Assumption 1 holds. Then Theorem 1 applies.

Corollary 3. Let \mathcal{I}_n be the forgotten topological index defined in (14) of

the random graph $\mathcal{G}_n(\alpha,\kappa)$ and σ_n^2 be defined in (2). Then

$$\frac{\mathcal{I}_n - \mathbb{E}[\mathcal{I}_n]}{\sigma_n} \Rightarrow \mathcal{N}(0, 1), \tag{15}$$

as n goes to infinity. In addition, the expectation $\mathbb{E}[\mathcal{I}_n]$ has the following asymptotic expression

$$\mathbb{E}[\mathcal{I}_n] = \left(1 + O\left(\frac{1}{np_n}\right)\right) \sum_{1 \le i < j \le n} p_n w_{ij} (w_{i(j)}^2 + w_{j(i)}^2), \qquad (16)$$

where the error rate $\frac{1}{np_n}$ cannot be improved.

For the Erdős-Rényi random graph $\mathcal{G}_n(\alpha)$, the expectation of the forgotten topological index [7] is equal to

$$\mathbb{E}[\mathcal{I}_n] = n(n-1)(n-2)(n-3)p_n^3 + 3n(n-1)(n-2)p_n^2 + n(n-1)p_n.$$
(17)

By (16), we have

$$\mathbb{E}[\mathcal{I}_n] \\ = \left(1 + O\left(\frac{1}{np_n}\right)\right) \frac{n(n-1)p_n}{2} \left[(1 + (n-2)p_n)^2 + (1 + (n-2)p_n)^2\right] \\ = n(n-1)(n-2)^2 p_n^3 \left(1 + O\left(\frac{1}{np_n}\right)\right).$$

Hence our approximation (16) is consistent with (17). Moreover, by (17), the error rate $\frac{1}{np_n}$ cannot be improved as in Corollary 3.

3.4 The inverse sum indeg index

The inverse sum indeg index is a significant predictor of total surface area of octane isomers [19, 23, 24]. Let $f(x, y) = \frac{xy}{x+y}$. The inverse sum indeg index \mathcal{I}_n is defined as

$$\mathcal{I}_n = \sum_{\{i,j\}\in\mathcal{E}} \frac{d_i d_j}{d_i + d_j}.$$
(18)

As far as we know, the inverse sum indeg index of random graph has not been studied in literature. Here we provide its asymptotic distribution and an approximation of its expectation.

Let g(x, y) = xy and $h(x, y) = \frac{1}{x+y}$. For simplicity, we denote f(x, y) as f. Then f = gh. Given positive integers $s, t, h^{(s,t)} = \frac{c_{s,t}}{(x+y)^{1+s+t}}$, where $c_{s,t}$ is a constant dependent on s, t. Straightforward calculation yields

$$\begin{split} f^{(s,0)} &= \sum_{r=0}^{s} \binom{s}{r} g^{(r,0)} h^{(s-r,0)} = g h^{(s,0)} + s g^{(1,0)} h^{(s-1,0)} \\ &= \frac{c_{s,0} x y}{(x+y)^{1+s}} + \frac{s c_{s-1,0} y}{(x+y)^{s}}, \\ f^{(0,t)} &= \sum_{r=0}^{t} \binom{t}{r} g^{(0,r)} h^{(0,t-r)} = g h^{(0,t)} + t g^{(0,1)} h^{(0,t-1)} \\ &= \frac{c_{0,t} x y}{(x+y)^{1+t}} + \frac{t c_{0,t-1} x}{(x+y)^{t}}. \end{split}$$

Further, for $s \ge 1$ or $t \ge 1$, we have

$$\begin{split} f^{(s,t)} &= \sum_{r=0}^{t} \binom{t}{r} g^{(0,r)} h^{(s,t-r)} + s \sum_{r=0}^{t} \binom{t}{r} g^{(1,r)} h^{(s-1,t-r)} \\ &= g h^{(s,t)} + t g^{(0,1)} h^{(s,t-1)} + s g^{(1,0)} h^{(s-1,t)} + s t g^{(1,1)} h^{(s-1,t-1)} \\ &= \frac{c_{s,t} x y}{(x+y)^{1+s+t}} + \frac{t c_{s,t-1} x}{(x+y)^{s+t}} + \frac{s c_{s-1,t} y}{(x+y)^{s+t}} + \frac{s t c_{s-1,t-1}}{(x+y)^{s+t-1}}. \end{split}$$

Hence, for $s + t \ge 1$, $|f^{(s,t)}(np_n, np_n)|$ can be bounded as follows

$$|f^{(s,t)}(np_n, np_n)| = O\left(\frac{1}{(np_n)^{s+t-1}}\right)$$

Note that

$$f_x(np_n, np_n) = f_y(np_n, np_n) = \frac{(np_n)^2}{(np_n + np_n)^2} = \frac{1}{4}$$

Then $a_{ij} = \Theta(np_n), \ \sigma_n^2 = \Theta(n(np_n)^3)$ and

$$\frac{\sum_{i < j} (a_{ij} + a_{ji})^4 p_n}{\sigma_n^4} = O\left(\frac{n(np_n)^5}{n^2(np_n)^6}\right) = o(1).$$

When $1 \leq s + t$, we have

$$n(np_n)^{2(s+t)}|f^{(s,t)}(np_n,np_n)|^2 = O\left(n(np_n)^2\right) = o\left(\sigma_n^2\right).$$

Let $k_0 = \max\{\lfloor 1 + \frac{1}{1-\alpha} \rfloor + 1, 3\}$. Then $k_0 > 1 + \frac{1}{1-\alpha}$ and

$$n(np_n)^{\frac{k_0}{2}+1}|f^{(s,t)}(np_n,np_n)| = O\left(n(np_n)^{2-\frac{k_0}{2}}\right) = o(\sigma_n).$$

Assumption 1 holds. Then Theorem 1 applies.

Corollary 4. Let \mathcal{I}_n be the inverse sum indeg index defined in (18) of the random graph $\mathcal{G}_n(\alpha, \kappa)$ and σ_n^2 be defined in (2). Then

$$\frac{\mathcal{I}_n - \mathbb{E}[\mathcal{I}_n]}{\sigma_n} \Rightarrow \mathcal{N}(0, 1), \tag{19}$$

as n goes to infinity. In addition, the expectation $\mathbb{E}[\mathcal{I}_n]$ has the following asymptotic expression

$$\mathbb{E}[\mathcal{I}_n] = \left(1 + O\left(\frac{1}{np_n}\right)\right) \sum_{1 \le i < j \le n} p_n w_{ij} \frac{w_{i(j)} w_{j(i)}}{w_{i(j)} + w_{j(i)}}, \qquad (20)$$

where the error rate $\frac{1}{np_n}$ cannot be improved.

For the Erdős-Rényi random graph $\mathcal{G}_n(\alpha)$, the expectation of the inverse sum indeg index can be expressed as

$$\mathbb{E}[\mathcal{I}_n] = \left(1 + O\left(\frac{1}{np_n}\right)\right) \frac{n(n-1)(n-2)p_n^2}{4}.$$

4 Proof of main results

In this section, we provide detailed proofs of Theorem 1 and Theorem 2. It is not easy to work on \mathcal{I}_n defined in (1) directly. In stead, we provide an alternative expression of \mathcal{I}_n as follows

$$\mathcal{I}_n = \sum_{1 \le i < j \le n} A_{ij} f(d_{i(j)}, d_{j(i)}) = \frac{1}{2} \sum_{i \ne j} A_{ij} f(d_{i(j)}, d_{j(i)}), \qquad (21)$$

where $d_{i(j)} = 1 + \sum_{l \notin \{i,j\}} A_{il}$ and $d_{j(i)} = 1 + \sum_{l \notin \{i,j\}} A_{jl}$. Note that A_{ij} , $d_{i(j)}$ and $d_{j(i)}$ are independent, $\mathbb{E}[d_{i(j)}] = w_{i(j)}$ and $\mathbb{E}[d_{j(i)}] = w_{j(i)}$. We will use these facts frequently in the proof.

4.1 Lemmas

Before proving Theorem 1 and Theorem 2, we present two lemmas.

Lemma 1. Let $\mathcal{G}_n(\beta, W)$ be defined in Definition 2, $\delta_n = (\log(np_n))^{-2}$ and M be a constant greater than $\frac{e^2}{1-p_n\beta}$. For any $i \in [n]$, we have

 $\mathbb{P}(d_{i(j)} - 1 = k) \le \exp(-np_n\beta(1 + o(1))), \quad k \le \delta_n np_n,$ $\mathbb{P}(d_{i(j)} - 1 = k) \le \exp(-np_n\beta(1 + o(1))), \quad k \ge Mnp_n.$

Proof of Lemma 1: Given distinct indices i, j, let $\theta_{ij} = \{p_n w_{il} | l \in [n] \setminus \{i, j\}\}$. Then $d_{i(j)} - 1$ follows the Poisson-Binomial distribution $PB(\theta_{ij})$. Recall that $\beta \leq w_{ij} \leq 1$. Then

$$\mathbb{P}(d_{i(j)} - 1 = k) = \sum_{S \subset [n] \setminus \{i,j\}, |S| = k} \prod_{l \in S} p_n w_{il} \prod_{l \in S^C \setminus \{i,j\}} (1 - p_n w_{il}) \\
\leq \sum_{S \subset [n] \setminus \{i,j\}, |S| = k} \prod_{l \in S} p_n \prod_{l \in S^C \setminus \{i,j\}} (1 - p_n \beta) \\
= \binom{n-2}{k} p_n^k (1 - p_n \beta)^{n-2-k}.$$
(22)

Note that $\binom{n-2}{k} \leq e^{k \log n - k \log k + k}$ and $(1 - p_n \beta)^{n-2-k} = e^{(n-2-k) \log(1-p_n \beta)}$. Then by (22) we get

$$\mathbb{P}(d_{i(j)} - 1 = k) \le \exp(k \log(np_n) - k \log k + k + (n - 2 - k) \log(1 - p_n \beta)).$$
(23)

Let $g(k) = k \log(np_n) - k \log k + k + (n - 2 - k) \log(1 - p_n\beta)$. Considering k as continuous variable, the derivative of g(k) with respect to k is equal to

$$g'(k) = \log\left(\frac{np_n}{1-p_n\beta}\right) - \log k.$$

Clearly, g'(k) > 0 for $k < \frac{np_n}{1-p_n\beta}$ and g'(k) < 0 for $k > \frac{np_n}{1-p_n\beta}$. Then g(k) achieves its maximum at $k = \frac{np_n}{1-p_n\beta}$. For $k \le \delta_n np_n$, $g(k) \le g(\delta_n np_n)$. Hence

$$\mathbb{P}(d_{i(j)} - 1 = k)$$

$$\leq \exp\left(\delta_n n p_n \log \frac{1}{\delta_n (1 - p_n \beta)} + \delta_n n p_n + n \log(1 - p_n \beta)\right)$$

$$\leq \exp\left(-n p_n \beta (1 + o(1))\right).$$

Since $M > \frac{e^2}{1-p_n\beta}$, then $Mnp_n \ge \frac{np_n}{1-p_n\beta}$. For $k \ge Mnp_n$, $g(k) \le g(Mnp_n)$. Hence

$$\mathbb{P}(d_{i(j)} - 1 = k)$$

$$\leq \exp\left(-M\log(M)np_n + Mnp_n + (n - 2 - Mnp_n)\log(1 - p_n\beta)\right)$$

$$= \exp\left(-M(\log M - 1)np_n + |n - 2 - Mnp_n|p_n\beta\right)$$

$$\leq \exp\left(-np_n\beta(1 + o(1))\right).$$

Lemma 2. Suppose (C1) of Assumption 1 holds. For the random graph $\mathcal{G}_n(\beta, W)$, we have

$$\frac{\sum_{i \neq j} a_{ij}(A_{ij} - p_n w_{ij})}{\sqrt{\sum_{i < j} (a_{ij} + a_{ji})^2 p_n w_{ij}(1 - p_n w_{ij})}} \Rightarrow \mathcal{N}(0, 1),$$

where

$$a_{ij} = \frac{1}{2}f(w_{i(j)}, w_{j(i)}) + \frac{1}{2}\sum_{l \notin \{i, j\}} p_n w_{il} \big[f_x(w_{i(l)}, w_{l(i)}) + f_y(w_{i(l)}, w_{l(i)}) \big].$$

Proof of Lemma 2: Let

$$\mathcal{Z}_{n} = \frac{\sum_{i \neq j} a_{ij} (A_{ij} - p_{n} w_{ij})}{\sqrt{\sum_{i < j} (a_{ij} + a_{ji})^{2} p_{n} w_{ij} (1 - p_{n} w_{ij})}}.$$

Then $\mathbb{E}[\overline{\mathcal{Z}_n}] = 0$. Note that

$$\sum_{i \neq j} a_{ij} (A_{ij} - p_n w_{ij}) = \sum_{i < j} (a_{ij} + a_{ji}) (A_{ij} - p_n w_{ij}).$$

Then

$$\mathcal{Z}_n = \frac{\sum_{i < j} (a_{ij} + a_{ji}) (A_{ij} - p_n w_{ij})}{\sqrt{\sum_{i < j} (a_{ij} + a_{ji})^2 p_n w_{ij} (1 - p_n w_{ij})}}.$$

Recall that $A_{ij} (1 \le i < j \le n)$ are independent. It is easy to verify that

$$\mathbb{E}\left[\left(\sum_{i < j} (a_{ij} + a_{ji})(A_{ij} - p_n w_{ij})\right)^2\right]$$

= $\sum_{i < j} (a_{ij} + a_{ji})^2 \mathbb{E}[(A_{ij} - p_n w_{ij})^2]$
= $\sum_{i < j} (a_{ij} + a_{ji})^2 p_n w_{ij}(1 - p_n w_{ij}).$

Hence $Var[\mathcal{Z}_n] = 1$. Straightforward calculation yields

$$\mathbb{E}[(A_{ij} - p_n w_{ij})^4] = p_n w_{ij} [(1 - p_n w_{ij})^4 + p_n^3 w_{ij}^3 (1 - p_n w_{ij})] \le 2p_n w_{ij}.$$

By (C1) of Assumption 1, we have

$$\frac{\sum_{i < j} (a_{ij} + a_{ji})^4 \mathbb{E}[(A_{ij} - p_n w_{ij})^4]}{\left(\sum_{i < j} (a_{ij} + a_{ji})^2 p_n w_{ij} (1 - p_n w_{ij})\right)^2} \\
\leq \frac{2\sum_{i < j} (a_{ij} + a_{ji})^4 p_n w_{ij}}{\left(\sum_{i < j} (a_{ij} + a_{ji})^2 p_n w_{ij} (1 - p_n w_{ij})\right)^2} \\
= o(1).$$

According to the Lyapunov Central Limit Theorem, \mathcal{Z}_n converges in distribution to the standard normal distribution.

4.2 Proof of Theorem 1

Let k_0 be the integer in Assumption 1. The proof strategy is as follows: firstly we use the Taylor expansion to expand the function $f(d_{i(j)}, d_{j(i)})$ at $(w_{i(j)}, w_{j(i)})$ to k_0 -th order; then we write \mathcal{I}_n as the sum of the leading term and remainder terms; finally we show the leading term (after suitable scaling) converges in distribution to the standard normal distribution and the remainder terms are negligible.

By the Taylor expansion, $f(d_{i(j)}, d_{j(i)})$ can be decomposed as

$$f(d_{i(j)}, d_{j(i)}) = M_{ij} + S_{ij} + T_{ij} + R_{ij}, \qquad (24)$$

where

$$\begin{split} M_{ij} &= f(w_{i(j)}, w_{j(i)}) + f_x(w_{i(j)}, w_{j(i)})(d_{i(j)} - w_{i(j)}) \\ &+ f_y(w_{i(j)}, w_{j(i)})(d_{j(i)} - w_{j(i)}), \\ S_{ij} &= \frac{1}{2} f_{xx}(w_{i(j)}, w_{j(i)})(d_{i(j)} - w_{i(j)})^2 \\ &+ \frac{1}{2} f_{yy}(w_{i(j)}, w_{j(i)})(d_{j(i)} - w_{j(i)})^2 \\ &+ f_{xy}(w_{i(j)}, w_{j(i)})(d_{i(j)} - w_{i(j)})(d_{j(i)} - w_{j(i)}), \\ T_{ij} &= \sum_{k=3}^{k_0 - 1} \sum_{s+t=k} \frac{f^{(s,t)}(w_{i(j)}, w_{j(i)})}{s!t!} (d_{i(j)} - w_{i(j)})^s (d_{j(i)} - w_{j(i)})^t, \\ R_{ij} &= \sum_{s+t=k_0} \frac{f^{(s,t)}(X_{i(j)}, X_{j(i)})}{s!t!} (d_{i(j)} - w_{i(j)})^s (d_{j(i)} - w_{j(i)})^t, \end{split}$$

 $X_{i(j)}$ is between $d_{i(j)}$ and $w_{i(j)}$, and $X_{j(i)}$ is between $d_{j(i)}$ and $w_{j(i)}$. By (21), the topological index \mathcal{I}_n is equal to

$$\mathcal{I}_n = \frac{1}{2} \sum_{i \neq j} M_{ij} A_{ij} + \frac{1}{2} \sum_{i \neq j} S_{ij} A_{ij} + \frac{1}{2} \sum_{i \neq j} T_{ij} A_{ij} + \frac{1}{2} \sum_{i \neq j} R_{ij} A_{ij}.$$
 (25)

Then

$$\frac{\mathcal{I}_n - \mathbb{E}[\mathcal{I}_n]}{\sigma_n}$$

$$= \frac{\frac{1}{2}\sum_{i\neq j}(M_{ij}A_{ij} - \mathbb{E}[M_{ij}A_{ij}])}{\sigma_n} + \frac{\frac{1}{2}\sum_{i\neq j}(S_{ij}A_{ij} - \mathbb{E}[S_{ij}A_{ij}])}{\sigma_n} + \frac{\frac{1}{2}\sum_{i\neq j}(T_{ij}A_{ij} - \mathbb{E}[T_{ij}A_{ij}])}{\sigma_n} + \frac{\frac{1}{2}\sum_{i\neq j}(R_{ij}A_{ij} - \mathbb{E}[R_{ij}A_{ij}])}{\sigma_n}, (26)$$

where σ_n^2 is defined in (2).

Next we show the first term in (26) is leading term and the last three terms are negligible.

4.2.1 Asymptotic normality of the first term in (26)

To begin with, we study the first term of (26). Note that

$$\sum_{i \neq j} M_{ij} A_{ij}$$

$$= \sum_{i \neq j} f(w_{i(j)}, w_{j(i)}) A_{ij} + \sum_{i \neq j} f_x(w_{i(j)}, w_{j(i)}) (d_{i(j)} - w_{i(j)}) A_{ij}$$

$$+ \sum_{i \neq j} f_y(w_{i(j)}, w_{j(i)}) (d_{j(i)} - w_{j(i)}) A_{ij}.$$
(27)

Since $d_{i(j)} - w_{i(j)} = \sum_{l \notin \{i,j\}} (A_{il} - p_n w_{il})$, then $(d_{i(j)} - w_{i(j)})$ does not contain A_{ij} . Hence, $(d_{i(j)} - w_{i(j)})$ and A_{ij} are independent. Similarly, $d_{j(i)} - w_{j(i)}$ and A_{ij} are independent. In addition, $\mathbb{E}[d_{i(j)}] = w_{i(j)}$ and $\mathbb{E}[d_{j(i)}] = w_{j(i)}$. Then the expectation of $\sum_{i \neq j} M_{ij} A_{ij}$ is equal to

$$\mathbb{E}\left[\sum_{i\neq j} M_{ij}A_{ij}\right] = \sum_{i\neq j} f(w_{i(j)}, w_{j(i)})p_n w_{ij}.$$
(28)

The first term of (27) can be written as

$$\sum_{i \neq j} f(w_{i(j)}, w_{j(i)}) A_{ij} = \sum_{i \neq j} f(w_{i(j)}, w_{j(i)}) (A_{ij} - p_n w_{ij}) + \sum_{i \neq j} f(w_{i(j)}, w_{j(i)}) p_n w_{ij}.$$
(29)

Similarly, the second term of (27) is written as

$$\sum_{i \neq j} f_x(w_{i(j)}, w_{j(i)})(d_{i(j)} - w_{i(j)})A_{ij}$$

$$= \sum_{i \neq j} f_x(w_{i(j)}, w_{j(i)})(d_{i(j)} - w_{i(j)})(A_{ij} - p_n w_{ij})$$

$$+ \sum_{i \neq j} p_n w_{ij} f_x(w_{i(j)}, w_{j(i)})(d_{i(j)} - w_{i(j)})$$

$$= \sum_{i \neq j \neq l} f_x(w_{i(j)}, w_{j(i)})(A_{il} - p_n w_{il})(A_{ij} - p_n w_{ij})$$

$$+ \sum_{i \neq j} \left(\sum_{l \notin \{i, j\}} p_n w_{il} f_x(w_{i(l)}, w_{l(i)})\right) (A_{ij} - p_n w_{ij}). \quad (30)$$

We will show the first term of (30) is of smaller order than the second term. To this end, we find the second moment of it. Recall that if $\{i, j\} \neq \{s, t\}$, A_{ij} and A_{st} are independent. Let i, j, l be three arbitrary distinct indices and i_1, j_1, l_1 be another three arbitrary distinct indices. If $\{i, j, l\} \neq \{i_1, j_1, l_1\}$, then

$$\mathbb{E}[(A_{ij} - p_n w_{ij})(A_{il} - p_n w_{il})(A_{i_1j_1} - p_n w_{i_1j_1})(A_{i_1l_1} - p_n w_{i_1l_1})] = 0.$$

When $\{i, j, l\} = \{i_1, j_1, l_1\}$, it is easy to get

$$\mathbb{E}[(A_{ij} - p_n w_{ij})(A_{il} - p_n w_{il})(A_{i_1 j_1} - p_n w_{i_1 j_1})(A_{i_1 l_1} - p_n w_{i_1 l_1})]$$

= $\mathbb{E}[(A_{ij} - p_n w_{ij})^2 (A_{il} - p_n w_{il})^2]$
= $p_n w_{ij} (1 - p_n w_{ij}) p_n w_{il} (1 - p_n w_{il}).$

Then the second moment of the first term of (30) can be calculated as follows.

$$\mathbb{E}\left[\left(\sum_{\substack{i\neq j\neq l\\i_1\neq j_1\neq l}} f_x(w_{i(j)}, w_{j(i)})(A_{il} - p_n w_{il})(A_{ij} - p_n w_{ij})\right)^2\right]$$

=
$$\sum_{\substack{i\neq j\neq l\\i_1\neq j_1\neq l_1}} f_x(w_{i(j)}, w_{j(i)})f_x(w_{i_1(j_1)}, w_{j_1(i_1)})$$

$$\times \mathbb{E}[(A_{il} - p_n w_{il})(A_{ij} - p_n w_{ij})(A_{i_1 l_1} - p_n w_{i_1 l_1})(A_{i_1 j_1} - p_n w_{i_1 j_1})]$$

$$= \sum_{i \neq j \neq l} f_x(w_{i(j)}, w_{j(i)})^2 \mathbb{E}[(A_{il} - p_n w_{il})^2(A_{ij} - p_n w_{ij})^2]$$

$$= \sum_{i \neq j \neq l} f_x(w_{i(j)}, w_{j(i)})^2 p_n w_{il}(1 - p_n w_{il}) p_n w_{ij}(1 - p_n w_{ij})$$

$$= \Theta\left(n^3 p_n^2 f_x(n p_n, n p_n)^2\right).$$

By Markov's inequality, it follows that

$$\sum_{i \neq j \neq l} f_x(w_{i(j)}, w_{j(i)})(A_{il} - p_n w_{il})(A_{ij} - p_n w_{ij})$$

= $O_P\left(\sqrt{n^3 p_n^2 f_x(np_n, np_n)^2}\right).$ (31)

Similarly, one has

$$\sum_{i \neq j \neq l} f_y(w_{i(j)}, w_{j(i)}) (A_{il} - p_n w_{il}) (A_{ij} - p_n w_{ij})$$

= $O_P\left(\sqrt{n^3 p_n^2 f_y(np_n, np_n)^2}\right).$ (32)

Denote

$$a_{ij} = \frac{1}{2}f(w_{i(j)}, w_{j(i)}) + \frac{1}{2}\sum_{l \notin \{i, j\}} p_n w_{il} \big[f_x(w_{i(l)}, w_{l(i)}) + f_y(w_{i(l)}, w_{l(i)}) \big].$$

Then combining (27)- (32) yields

$$= \frac{\frac{\frac{1}{2}\sum_{i\neq j}(M_{ij}A_{ij} - \mathbb{E}[M_{ij}A_{ij}])}{\sigma_n}}{\sum_{i\neq j}a_{ij}(A_{ij} - p_nw_{ij})} + O_P\left(\sqrt{\frac{n^3p_n^2f_x(np_n, np_n)^2}{\sigma_n^2}}\right).$$

By Lemma 2 and (C5) of Assumption 1 (let s + t = 1), we conclude that

$$\frac{\frac{1}{2}\sum_{i\neq j}(M_{ij}A_{ij} - \mathbb{E}[M_{ij}A_{ij}])}{\sigma_n} \Rightarrow \mathcal{N}(0, 1).$$

Then the proof is complete if the second, third and last term of (26)

converge to zero in probability.

4.2.2 Bound the second term of (26)

We prove the second term of (26) is equal to $o_P(1)$. By the definition of S_{ij} , we have

$$\sum_{i \neq j} S_{ij} A_{ij}$$

$$= \frac{1}{2} \sum_{i \neq j} f_{xx}(w_{i(j)}, w_{j(i)}) (d_{i(j)} - w_{i(j)})^2 A_{ij}$$

$$+ \frac{1}{2} \sum_{i \neq j} f_{yy}(w_{i(j)}, w_{j(i)}) (d_{j(i)} - w_{j(i)})^2 A_{ij}$$

$$+ \sum_{i \neq j} f_{xy}(w_{i(j)}, w_{j(i)}) (d_{i(j)} - w_{i(j)}) (d_{j(i)} - w_{j(i)}) A_{ij}.$$
(33)

Then

$$\mathbb{E}\left[\sum_{i\neq j} S_{ij}A_{ij}\right] \\
= \frac{1}{2}\sum_{i\neq j} f_{xx}(w_{i(j)}, w_{j(i)})\mathbb{E}\left[(d_{i(j)} - w_{i(j)})^{2}\right] p_{n}w_{ij} \\
+ \frac{1}{2}\sum_{i\neq j} f_{yy}(w_{i(j)}, w_{j(i)})\mathbb{E}\left[(d_{j(i)} - w_{j(i)})^{2}\right] p_{n}w_{ij} \\
= \frac{1}{2}\sum_{i\neq j\neq l} f_{xx}(w_{i(j)}, w_{j(i)})p_{n}w_{il}(1 - p_{n}w_{il})p_{n}w_{ij} \\
+ \frac{1}{2}\sum_{i\neq j\neq l} f_{yy}(w_{i(j)}, w_{j(i)})p_{n}w_{jl}(1 - p_{n}w_{jl})p_{n}w_{ij}.$$
(34)

The first term of (33) can be expressed as

$$\sum_{i \neq j} f_{xx}(w_{i(j)}, w_{j(i)}) (d_{i(j)} - w_{i(j)})^2 A_{ij}$$

=
$$\sum_{i \neq j} f_{xx}(w_{i(j)}, w_{j(i)}) (d_{i(j)} - w_{i(j)})^2 (A_{ij} - p_n w_{ij})$$

$$+\sum_{i\neq j} p_n w_{ij} f_{xx}(w_{i(j)}, w_{j(i)}) (d_{i(j)} - w_{i(j)})^2.$$
(35)

We will find an upper bound of (35). Note that

$$\sum_{i \neq j} f_{xx}(w_{i(j)}, w_{j(i)}) (d_{i(j)} - w_{i(j)})^2 (A_{ij} - p_n w_{ij})$$

$$= \sum_{i \neq j} f_{xx}(w_{i(j)}, w_{j(i)}) (A_{ij} - p_n w_{ij}) \sum_{\substack{s \neq t \\ s, t \notin \{i, j\}}} (A_{is} - p_n w_{is}) (A_{it} - p_n w_{it})$$

$$+ \sum_{i \neq j} f_{xx}(w_{i(j)}, w_{j(i)}) (A_{ij} - p_n w_{ij}) \sum_{s \notin \{i, j\}} (A_{is} - p_n w_{is})^2.$$
(36)

The second moment of the first term of (36) is equal to

$$\mathbb{E}\left[\sum_{i\neq j} f_{xx}(w_{i(j)}, w_{j(i)})(A_{ij} - p_n w_{ij}) \times \sum_{\substack{s\neq t \\ s,t\notin\{i,j\}}} (A_{is} - p_n w_{is})(A_{it} - p_n w_{it})\right]^2 \times (A_{is} - p_n w_{ij})^2 \mathbb{E}\left[(A_{ij} - p_n w_{ij})^2 \times (A_{is} - p_n w_{is})^2(A_{it} - p_n w_{it})^2\right] \times (A_{is} - p_n w_{is})^2(A_{it} - p_n w_{it})^2 + \sum_{\substack{i\neq j, s\neq t \\ s,t\notin\{i,j\}}} f_{xx}(w_{i(j)}, w_{j(i)})^2 p_n w_{ij}(1 - p_n w_{ij}) \times p_n w_{is}(1 - p_n w_{is}) p_n w_{it}(1 - p_n w_{it}) = \Theta\left(n^4 p_n^3 f_{xx}(n p_n, n p_n)^2\right).$$
(37)

The second moment of the second term of (36) is equal to

$$\mathbb{E}\left[\sum_{i\neq j} f_{xx}(w_{i(j)}, w_{j(i)})(A_{ij} - p_n w_{ij}) \sum_{s\notin\{i,j\}} (A_{is} - p_n w_{is})^2\right]^2$$

=
$$\sum_{i\neq j\neq s} f_{xx}(w_{i(j)}, w_{j(i)})^2 \mathbb{E}\left[(A_{ij} - p_n w_{ij})^2 (A_{is} - p_n w_{is})^4\right]$$

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$$+\sum_{\substack{i\neq j\neq s\neq t}} f_{xx}(w_{i(j)}, w_{j(i)})^{2} \mathbb{E} [(A_{ij} - p_{n}w_{ij})^{2}(A_{is} - p_{n}w_{is})^{2} \\ \times (A_{it} - p_{n}w_{it})^{2}] \\ +\sum_{\substack{i\neq j\neq s}} f_{xx}(w_{i(j)}, w_{j(i)})f_{xx}(w_{i(s)}, w_{s(i)}) \\ \times \mathbb{E} [(A_{ij} - p_{n}w_{ij})^{3}(A_{is} - p_{n}w_{is})^{3}] \\ \leq \sum_{\substack{i\neq j\neq s}} f_{xx}(w_{i(j)}, w_{j(i)})^{2}p_{n}^{2} + \sum_{\substack{i\neq j\neq s\neq t}} f_{xx}(w_{i(j)}, w_{j(i)})^{2}p_{n}^{3} \\ + \sum_{\substack{i\neq j\neq s}} f_{xx}(w_{i(j)}, w_{j(i)})f_{xx}(w_{i(s)}, w_{s(i)})p_{n}^{2} \\ = \Theta \left(n^{4}p_{n}^{3}f_{xx}(np_{n}, np_{n})^{2}\right).$$
(38)

Now we consider the second term of (35). Note that

$$\sum_{i \neq j} p_n w_{ij} f_{xx}(w_{i(j)}, w_{j(i)}) (d_{i(j)} - w_{i(j)})^2$$

$$= \sum_{i \neq j} p_n w_{ij} f_{xx}(w_{i(j)}, w_{j(i)}) \sum_{\substack{s \neq t \\ s, t \notin \{i, j\}}} (A_{is} - p_n w_{is}) (A_{it} - p_n w_{it})$$

$$+ \sum_{i \neq j} p_n w_{ij} f_{xx}(w_{i(j)}, w_{j(i)})$$

$$\times \sum_{s \notin \{i, j\}} \left[(A_{is} - p_n w_{is})^2 - \mathbb{E}[(A_{is} - p_n w_{is})^2] \right]$$

$$+ \sum_{i \neq j} p_n w_{ij} f_{xx}(w_{i(j)}, w_{j(i)}) \sum_{s \notin \{i, j\}} \mathbb{E}\left[(A_{is} - p_n w_{is})^2 \right]. \quad (39)$$

The second moment of the first term of (39) is equal to

$$\mathbb{E}\left[\sum_{\substack{i\neq j \\ i\neq j}} p_n w_{ij} f_{xx}(w_{i(j)}, w_{j(i)}) \sum_{\substack{s\neq t \\ s,t\notin\{i,j\}}} (A_{is} - p_n w_{is})(A_{it} - p_n w_{it})\right]^2$$

=
$$\sum_{\substack{i\neq j, i\neq j_1, s\neq t \\ s,t\notin\{i,j,j_1\}}} p_n^2 w_{ij} w_{ij_1} f_{xx}(w_{i(j)}, w_{j(i)}) f_{xx}(w_{i(j_1)}, w_{j_1(i)})$$

$$\times \mathbb{E}\left[(A_{is} - p_n w_{is})^2 (A_{it} - p_n w_{it})^2\right]$$

$$+ \sum_{\substack{i \neq j, s \neq t \\ s,t \notin \{i,j\}}} p_n^2 w_{ij}^2 f_{xx}(w_{i(j)}, w_{j(i)})^2 \mathbb{E} \left[(A_{is} - p_n w_{is})^2 (A_{it} - p_n w_{it})^2 \right]$$

$$= \sum_{\substack{i \neq j, i \neq j, s \neq t \\ s,t \notin \{i,j,j_1\}}} p_n^2 w_{ij} w_{ij_1} f_{xx}(w_{i(j)}, w_{j(i)}) f_{xx}(w_{i(j_1)}, w_{j_1(i)})$$

$$\times p_n w_{is}(1 - p_n w_{is}) p_n w_{it}(1 - p_n w_{it})$$

$$+ \sum_{\substack{i \neq j, s \neq t \\ s,t \notin \{i,j\}}} p_n^2 w_{ij}^2 f_{xx}(w_{i(j)}, w_{j(i)})^2 p_n w_{is}(1 - p_n w_{is}) p_n w_{it}(1 - p_n w_{it})$$

$$= \Theta \left(n^5 p_n^4 f_{xx}(np_n, np_n)^2 \right)$$

$$(40)$$

The second moment of the second term of (39) is equal to

$$\mathbb{E}\left[\sum_{i \neq j} p_{n} w_{ij} f_{xx}(w_{i(j)}, w_{j(i)}) \times \sum_{s \notin \{i,j\}} \left[(A_{is} - p_{n} w_{is})^{2} - \mathbb{E} \left[(A_{is} - p_{n} w_{is})^{2} \right] \right]^{2} \\
= \sum_{\substack{i \neq j \neq s \\ i \neq j_{1} \neq s}} p_{n} w_{ij} f_{xx}(w_{i(j)}, w_{j(i)}) p_{n} w_{ij_{1}} f_{xx}(w_{i(j_{1})}, w_{j_{1}(i)}) \\
\times \mathbb{E} \left[\left((A_{is} - p_{n} w_{is})^{2} - \mathbb{E} \left[(A_{is} - p_{n} w_{is})^{2} \right] \right)^{2} \right] \\
+ \sum_{i \neq j \neq s} p_{n}^{2} w_{ij}^{2} f_{xx}(w_{i(j)}, w_{j(i)})^{2} \\
\times \mathbb{E} \left[\left((A_{is} - p_{n} w_{is})^{2} - \mathbb{E} \left[(A_{is} - p_{n} w_{is})^{2} \right] \right)^{2} \right] \\
= \Theta \left(n^{4} p_{n}^{3} f_{xx}(np_{n}, np_{n})^{2} \right).$$
(41)

By Markov's inequality and equations (35)-(41), we have

$$\sum_{i \neq j} f_{xx}(w_{i(j)}, w_{j(i)}) \left[(d_{i(j)} - w_{i(j)})^2 A_{ij} - \mathbb{E} \left[(d_{i(j)} - w_{i(j)})^2 \right] A_{ij} \right]$$

= $O_P \left(\sqrt{n^5 p_n^4 f_{xx}(np_n, np_n)^2} \right).$ (42)

Similarly, one has

$$\sum_{i \neq j} f_{yy}(w_{i(j)}, w_{j(i)}) \left[(d_{i(j)} - w_{i(j)})^2 A_{ij} - \mathbb{E} \left[(d_{i(j)} - w_{i(j)})^2 \right] A_{ij} \right]$$

= $O_P \left(\sqrt{n^5 p_n^4 f_{yy} (np_n, np_n)^2} \right).$ (43)

Now we bound the third term of (33). Note that

$$\sum_{i \neq j} f_{xy}(w_{i(j)}, w_{j(i)})(d_{i(j)} - w_{i(j)})(d_{j(i)} - w_{j(i)})A_{ij}$$

$$= \sum_{i \neq j} f_{xy}(w_{i(j)}, w_{j(i)})(d_{i(j)} - w_{i(j)})(d_{j(i)} - w_{j(i)})(A_{ij} - p_n w_{ij})$$

$$+ \sum_{i \neq j} f_{xy}(w_{i(j)}, w_{j(i)})(d_{i(j)} - w_{i(j)})(d_{j(i)} - w_{j(i)})p_n w_{ij}$$

$$= \sum_{\substack{i \neq j, s \neq j \\ t \neq i}} f_{xy}(w_{i(j)}, w_{j(i)})(A_{ij} - p_n w_{ij})(A_{is} - p_n w_{is})(A_{jt} - p_n w_{jt})$$

$$+ \sum_{\substack{i \neq j, s \neq j \\ t \neq i}} f_{xy}(w_{i(j)}, w_{j(i)})p_n w_{ij}(A_{is} - p_n w_{is})(A_{jt} - p_n w_{jt}). \quad (44)$$

The second moment of the first term of (44) is equal to

$$\mathbb{E}\left[\sum_{\substack{i \neq j, s \neq j, t \neq i}} f_{xy}(w_{i(j)}, w_{j(i)})(A_{ij} - p_n w_{ij}) \\ \times (A_{is} - p_n w_{is})(A_{jt} - p_n w_{jt})\right]^2 \\ = \sum_{\substack{i \neq j, s \neq j, \\ t \neq i, s \neq t}} f_{xy}(w_{i(j)}, w_{j(i)})^2 \mathbb{E}[(A_{ij} - p_n w_{ij})^2 \\ \times (A_{is} - p_n w_{is})^2 (A_{jt} - p_n w_{jt})^2] \\ + \sum_{\substack{i \neq j, s \neq j, \\ t \neq i, s \neq t}} f_{xy}(w_{i(j)}, w_{j(i)})^2 \mathbb{E}[(A_{ij} - p_n w_{ij})^2 \\ \times (A_{is} - p_n w_{is})^2 (A_{js} - p_n w_{js})^2] \\ = \sum_{\substack{i \neq j, s \neq j, \\ t \neq i, s \neq t}} f_{xy}(w_{i(j)}, w_{j(i)})^2 p_n w_{ij}(1 - p_n w_{ij})$$

$$\times p_{n}w_{is}(1-p_{n}w_{is})p_{n}w_{jt}(1-p_{n}w_{jt}) + \sum_{\substack{i \neq j, \\ s \neq j}} f_{xy}(w_{i(j)}, w_{j(i)})^{2}p_{n}w_{ij}(1-p_{n}w_{ij}) \times p_{n}w_{is}(1-p_{n}w_{is})p_{n}w_{js}(1-p_{n}w_{js}) = \Theta \left(n^{4}p_{n}^{3}f_{xy}(np_{n}, np_{n})^{2}\right).$$

$$(45)$$

The second moment of the second term of (44) is equal to

$$\mathbb{E}\left[\sum_{\substack{i\neq j, s\neq j, t\neq i}} f_{xy}(w_{i(j)}, w_{j(i)})p_{n}w_{ij}(A_{is} - p_{n}w_{is})(A_{jt} - p_{n}w_{jt})\right]^{2} \\
= \sum_{\substack{i\neq j, s\neq j\\ t\neq i, s\neq t}} f_{xy}(w_{i(j)}, w_{j(i)})^{2}p_{n}^{2}w_{ij}^{2}\mathbb{E}\left[(A_{is} - p_{n}w_{is})^{2}(A_{jt} - p_{n}w_{jt})^{2}\right] \\
+ \sum_{\substack{i\neq j, s\neq j\\ t\neq i, s\neq t}} f_{xy}(w_{i(j)}, w_{j(i)})^{2}p_{n}^{2}w_{ij}^{2}\mathbb{E}\left[(A_{is} - p_{n}w_{is})^{2}(A_{js} - p_{n}w_{js})^{2}\right] \\
= \sum_{\substack{i\neq j, s\neq j\\ t\neq i, s\neq t}} f_{xy}(w_{i(j)}, w_{j(i)})^{2}p_{n}^{2}w_{ij}^{2}p_{n}w_{is}(1 - p_{n}w_{is})p_{n}w_{jt}(1 - p_{n}w_{jt}) \\
+ \sum_{\substack{i\neq j, s}} f_{xy}(w_{i(j)}, w_{j(i)})^{2}p_{n}^{2}w_{ij}^{2}p_{n}w_{is}(1 - p_{n}w_{is})p_{n}w_{js}(1 - p_{n}w_{js}) \\
= \Theta\left(n^{4}p_{n}^{4}f_{xy}(np_{n}, np_{n})^{2}\right).$$
(46)

Combining (44), (45) and (46) yields

$$\sum_{i \neq j} f_{xy}(w_{i(j)}, w_{j(i)})(d_{i(j)} - w_{i(j)})(d_{j(i)} - w_{j(i)})A_{ij}$$

= $O_P\left(\sqrt{n^4 p_n^3 f_{xy}(np_n, np_n)^2}\right).$ (47)

By (42), (43), (47) and (C5) of Assumption 1 (let s + t = 2), we get

$$\sum_{i \neq j} (S_{ij}A_{ij} - \mathbb{E}[S_{ij}A_{ij}]) = O_P\left(\sqrt{n^5 p_n^4 [f_{xx}(np_n, np_n)^2 + f_{yy}(np_n, np_n)^2] + n^4 p_n^3 f_{xy}(np_n, np_n)^2}\right) = o_P(\sigma_n).$$
(48)

Hence the second term of (26) is equal to $o_P(1)$.

4.2.3 Bound the third term of (26)

Now we prove the third term of (26) converges in probability to zero. This is the most complex part of the proof. Note that

$$\sum_{i \neq j} T_{ij} A_{ij}$$

$$= \sum_{k=3}^{k_0-1} \sum_{s+t=k} \sum_{i \neq j} \frac{f^{(s,t)}(w_{i(j)}, w_{j(i)})}{s!t!} (d_{i(j)} - w_{i(j)})^s (d_{j(i)} - w_{j(i)})^t A_{ij}$$

$$= \sum_{k=3}^{k_0-1} \sum_{s+t=k} \sum_{i \neq j} \frac{f^{(s,t)}(w_{i(j)}, w_{j(i)})}{s!t!} (d_{i(j)} - w_{i(j)})^s (d_{j(i)} - w_{j(i)})^t$$

$$\times (A_{ij} - p_n w_{ij})$$

$$+ \sum_{k=3}^{k_0-1} \sum_{s+t=k} \sum_{i \neq j} \frac{f^{(s,t)}(w_{i(j)}, w_{j(i)})}{s!t!} (d_{i(j)} - w_{i(j)})^s$$

$$\times (d_{j(i)} - w_{j(i)})^t p_n w_{ij}.$$
(49)

Next we bound the second moment of the first term of (49) and the variance of the second term. Since k_0 is a fixed finite integer, the quantities s!, t!in (49) are finite. We will ignore them in the subsequent analysis for simplicity. Given finite integer $k_0 \ge 4$, there are finitely many non-negative integers s, t such that s + t = k for any $k = 3, 4, \ldots, k_0 - 1$. Hence we only need to bound the second moment of

$$\sum_{i \neq j} f^{(s,t)}(w_{i(j)}, w_{j(i)}) (d_{i(j)} - w_{i(j)})^s (d_{j(i)} - w_{j(i)})^t (A_{ij} - p_n w_{ij}), \quad (50)$$

and the variance of

$$\sum_{i \neq j} f^{(s,t)}(w_{i(j)}, w_{j(i)}) (d_{i(j)} - w_{i(j)})^s (d_{j(i)} - w_{j(i)})^t p_n w_{ij},$$
(51)

where s, t are given non-negative integers with s + t = k for $k = 3, 4, ..., k_0 - 1$.

We consider the variance of (51) first. Fix integer $k \in \{3, 4, \dots, k_0 - 1\}$

and integers $s, t \in \{0, 1, 2, ..., k\}$ satisfying s + t = k. For positive integers $r \leq s$ and $v \leq t$, let $\lambda_{r;1}, \lambda_{r;2}, ..., \lambda_{r;r}, \gamma_{v;1}, \gamma_{v;2}, ..., \gamma_{v;v}$ be positive integers such that $\lambda_{r;1} + \lambda_{r;2} + \cdots + \lambda_{r;r} = s$ and $\gamma_{v;1} + \gamma_{v;2} + \cdots + \gamma_{v;v} = t$. Given indices i, j, we have

$$(d_{i(j)} - w_{i(j)})^{s} = \sum_{\substack{j_{1}, j_{2}, \dots, j_{s} \notin \{i, j\} \ l = 1}} \prod_{l=1}^{s} (A_{ij_{l}} - p_{n}w_{ij_{l}})$$

$$= \sum_{r=1}^{s} \sum_{\substack{j_{1}, j_{2}, \dots, j_{r} \notin \{i, j\} \ l = 1}} \prod_{l=1}^{r} (A_{ij_{l}} - p_{n}w_{ij_{l}})^{\lambda_{r;l}},$$

$$(d_{j(i)} - w_{j(i)})^{t} = \sum_{\substack{i_{1}, i_{2}, \dots, i_{t} \notin \{i, j\} \ l = 1}} \prod_{l=1}^{t} (A_{ji_{l}} - p_{n}w_{ji_{l}})$$

$$= \sum_{v=1}^{t} \sum_{\substack{i_{1}, i_{2}, \dots, i_{v} \notin \{i, j\} \ l = 1}} \prod_{l=1}^{v} (A_{ji_{l}} - p_{n}w_{ji_{l}})^{\gamma_{v;l}}.$$
(52)

Then (51) can be written as

$$\sum_{i \neq j} p_n w_{ij} f^{(s,t)}(w_{i(j)}, w_{j(i)}) (d_{i(j)} - w_{i(j)})^s (d_{j(i)} - w_{j(i)})^t = \sum_{r=1}^s \sum_{\nu=1}^t V_{r\nu},$$

where

$$V_{rv} = \sum_{\substack{i \neq j \\ j_1, \dots, j_r \notin \{i, j\} \\ j_1 \neq j_2 \neq \dots \neq j_r \\ i_1, \dots, i_v \notin \{i, j\} \\ i_1 \neq i_2 \neq \dots \neq i_v}} p_n w_{ij} f^{(s,t)}(w_{i(j)}, w_{j(i)}) \prod_{l=1}^r (A_{ij_l} - p_n w_{ij_l})^{\lambda_{r;l}} \\ \times \prod_{m=1}^v (A_{ji_m} - p_n w_{ji_m})^{\gamma_{v;m}}.$$
(53)

Note that

$$Var\left(\sum_{r=1}^{s}\sum_{v=1}^{t}V_{rv}\right) = \sum_{r=1}^{s}\sum_{v=1}^{t}\sum_{r_{1}=1}^{s}\sum_{v_{1}=1}^{t}Cov(V_{rv}, V_{r_{1}v_{1}})$$

$$\leq \sum_{r=1}^{s} \sum_{v=1}^{t} \sum_{r_1=1}^{s} \sum_{v_1=1}^{t} \Big(Var(V_{rv}) + Var(V_{r_1v_1}) \Big),$$

and s, t are finite non-negative integers, we only need to bound $Var(V_{rv})$ for each given r, v. Fix $r \in \{1, 2, ..., s\}$ and $v \in \{1, 2, ..., t\}$. There are two cases: (I) there exists $l_0 \in \{1, 2, ..., r\}$ or $m_0 \in \{1, 2, ..., v\}$ such that $\lambda_{r;l_0} = 1$ or $\gamma_{v;m_0} = 1$; (II) $\lambda_{r;l} \geq 2$ for all $l \in \{1, 2, ..., r\}$ and $\gamma_{v;m} \geq 2$ for all $m \in \{1, 2, ..., v\}$.

We study case (I) first. Suppose there are some $\lambda_{r;l}$ or $\gamma_{v;m}$ which are equal to one. Without loss of generality, let $\lambda_{r;1} = \lambda_{r;2} = \cdots = \lambda_{r;r_0} = 1$ and $\lambda_{r;l} \geq 2$ for $l \in \{r_0 + 1, \ldots, r\}$, $\gamma_{v;1} = \gamma_{v;2} = \cdots = \gamma_{v;v_0} = 1$ and $\gamma_{v;l} \geq 2$ for $l \in \{v_0 + 1, \ldots, v\}$. Here, either $r_0 \geq 1$ or $v_0 \geq 1$. Without loss of generality, let $r_0 \geq 1$. In this case,

$$\prod_{l=1}^{r} (A_{ij_l} - p_n w_{ij_l})^{\lambda_{r;l}} = \left(\prod_{l=1}^{r_0} (A_{ij_l} - p_n w_{ij_l}) \right) \left(\prod_{l=r_0+1}^{r} (A_{ij_l} - p_n w_{ij_l})^{\lambda_{r;l}} \right), \quad (54)$$

$$\prod_{m=1}^{v} (A_{ji_m} - p_n w_{ji_m})^{\gamma_{v;m}} = \left(\prod_{m=1}^{v_0} (A_{ji_m} - p_n w_{ji_m})\right) \left(\prod_{m=v_0+1}^{v} (A_{ji_m} - p_n w_{ji_m})^{\gamma_{v;m}}\right), (55)$$

and $\mathbb{E}[V_{rv}] = 0$. Then $Var(V_{rv}) = \mathbb{E}[V_{rv}^2]$. For convenience, denote $\bar{A}_{ij} = A_{ij} - p_n w_{ij}$. By (53), (54) and (55), we have

$$\mathbb{E}[V_{rv}^{2}] = \sum_{\substack{i \neq j \\ j_{1}, \dots, j_{r} \notin \{i, j\} \\ j_{1} \neq j \neq \dots, \neq j_{r} \\ j_{1} \neq j_{2} \neq \dots, \neq j_{r} \\ i_{1} \neq j_{2} \neq \dots, \neq j_{r} \\ i_{1} \neq i_{2} \neq \dots, \neq i_{v} \\ k p_{n} w_{i'j'} f^{(s,t)} (w_{i'(j')}, w_{j'(i')})$$

$$\times \mathbb{E}\left[\left(\prod_{l=1}^{r_{0}} \bar{A}_{ij_{l}}\right)\left(\prod_{l=r_{0}+1}^{r} \bar{A}_{ij_{l}}^{\lambda_{r;l}}\right)\left(\prod_{l=1}^{r_{0}} \bar{A}_{i'j_{l}'}\right)\left(\prod_{l=r_{0}+1}^{r} \bar{A}_{i'j_{l}'}^{\lambda_{r;l}}\right)\right) \times \left(\prod_{m=1}^{v_{0}} \bar{A}_{ji_{m}}\right)\left(\prod_{m=v_{0}+1}^{v} \bar{A}_{j'i_{m}'}^{\gamma_{v;m}}\right)\left(\prod_{m=1}^{v_{0}} \bar{A}_{j'i_{m}'}\right)\left(\prod_{m=v_{0}+1}^{v} \bar{A}_{j'i_{m}'}^{\gamma_{v;m}}\right)\right].$$

$$(56)$$

Next we find an upper bound of (56). Recall that $i \neq j$ and $i' \neq j'$. We shall decompose the summation in (56) into six cases: $i \neq i'$ and j = j'; i = i' and $j \neq j'$; $i \neq j'$ and j = i'; i = j' and $j \neq i'$; $\{i, j\} = \{i', j'\}$; $\{i, j\} \cap \{i', j'\} = \emptyset$. For convenience, denote the expectation in (56) as E.

Firstly, we consider the case $\{i, j\} = \{i', j'\}$. There are two scenarios: (i) i = i' and j = j', (ii) i = j' and j = i'.

Consider (i) first. In this case, A_{ij_l} , $A_{ij'_l}$ are independent of A_{ji_m} , $A_{ji'_m}$. Then

$$E = \mathbb{E}\left[\left(\prod_{l=1}^{r_0} \bar{A}_{ij_l} \bar{A}_{ij'_l}\right) \left(\prod_{l=r_0+1}^{r} \bar{A}_{ij_l}^{\lambda_{r;l}} \bar{A}_{ij'_l}^{\lambda_{r;l}}\right)\right] \times \mathbb{E}\left[\left(\prod_{m=1}^{v_0} \bar{A}_{ji_m} \bar{A}_{ji'_m}\right) \left(\prod_{m=v_0+1}^{v} \bar{A}_{ji_m}^{\gamma_{v;m}} \bar{A}_{ji'_m}^{\gamma_{v;m}}\right)\right].$$
 (57)

Recall that j_1, j_2, \ldots, j_r are mutually distinct and j'_1, j'_2, \ldots, j'_r are mutually distinct. Moreover, $\mathbb{E}[\bar{A}_{ij_l}] = 0$ for all $l = 1, 2, \ldots, r$. If there exists an index j_{l_1} with $1 \leq l_1 \leq r_0$ such that $j_{l_1} \notin \{j'_1, j'_2, \ldots, j'_r\}$, then

$$\mathbb{E}\left[\left(\prod_{l=1}^{r_0} \bar{A}_{ij_l} \bar{A}_{ij'_l}\right) \left(\prod_{l=r_0+1}^{r} \bar{A}_{ij_l}^{\lambda_{r;l}} \bar{A}_{ij'_l}^{\lambda_{r;l}}\right)\right] \\
= \mathbb{E}[\bar{A}_{ij_{l_1}}]\mathbb{E}\left[\left(\prod_{l=1,j_l \neq j_{l_1}}^{r_0} \bar{A}_{ij_l}\right) \left(\prod_{l=1}^{r_0} \bar{A}_{ij'_l}\right) \left(\prod_{l=r_0+1}^{r} \bar{A}_{ij_l}^{\lambda_{r;l}} \bar{A}_{ij'_l}^{\lambda_{r;l}}\right)\right] \\
= 0.$$
(58)

Hence E = 0 by (57). Similarly, if there exists an index l_1 with $1 \le l_1 \le r_0$ such that $j'_{l_1} \notin \{j_1, j_2, \ldots, j_r\}$, then E = 0. In addition, if there is an index m_1 with $1 \le m_1 \le v_0$ such that $i_{m_1} \notin \{i'_1, i'_2, \ldots, i'_v\}$ or $i'_{m_1} \notin$ $\{i_1, i_2, \ldots, i_v\}$, then E = 0. Consequently, $E \neq 0$ implies the following

$$\{j_1, j_2, \dots, j_{r_0}\} \subset \{j'_1, j'_2, \dots, j'_r\}, \quad \{j'_1, j'_2, \dots, j'_{r_0}\} \subset \{j_1, j_2, \dots, j_r\},$$
$$\{i_1, i_2, \dots, i_{v_0}\} \subset \{i'_1, i'_2, \dots, i'_v\}, \quad \{i'_1, i'_2, \dots, i'_{v_0}\} \subset \{i_1, i_2, \dots, i_v\}.$$

Without loss of generality, suppose

$$\{j_1, j_2, \dots, j_{r_1}\} = \{j'_1, j'_2, \dots, j'_{r_1}\}, \qquad \{j_{r_1+1}, \dots, j_r\} \cap \{j'_{r_1+1}, \dots, j'_r\} = \emptyset,$$

$$\{i_1, i_2, \dots, i_{v_1}\} = \{i'_1, i'_2, \dots, i'_{v_1}\}, \qquad \{i_{v_1+1}, \dots, i_v\} \cap \{i'_{v_1+1}, \dots, i'_v\} = \emptyset,$$

$$(60)$$

for some r_1 $(r_0 \leq r_1 \leq r)$ and v_1 $(v_0 \leq v_1 \leq v)$. There are at most $n^{2+2r-r_1+2v-v_1}$ possible choices for the indices $i, j, i_1, \ldots, i_v, j_1, \ldots, j_r, i', j', i'_1, \ldots, i'_v, j'_1, \ldots, j'_r$ satisfying i = i', j = j', (59) and (60). Let σ_1 be a permutation of $\{1, 2, \ldots, r_1\}$ such that $j_l = j'_{\sigma_1(l)}$ and σ_2 be a permutation of $\{1, 2, \ldots, v_1\}$ such that $i_m = i'_{\sigma_2(m)}$. The numbers of the permutations σ_1 and σ_2 are $r_1!$ and $v_1!$ respectively. Then

$$\mathbb{E}\left[\left(\prod_{l=1}^{r_0} \bar{A}_{ij_l} \bar{A}_{ij'_l}\right) \left(\prod_{l=r_0+1}^{r} \bar{A}_{ij_l}^{\lambda_{r;l}} \bar{A}_{ij'_l}^{\lambda_{r;l}}\right)\right] \\
= \mathbb{E}\left[\left(\prod_{l=1}^{r_1} \bar{A}_{ij_l}^{\lambda_{r;l}+\lambda_{r;\sigma_1}(l)}\right)\right] \left(\prod_{l=r_1+1}^{r} \mathbb{E}[\bar{A}_{ij_l}^{\lambda_{r;l}}]\mathbb{E}[\bar{A}_{ij'_l}^{\lambda_{r;l}}]\right) \\
= O\left(p_n^{2r-r_1}\right).$$
(61)

Similarly, we have

$$\mathbb{E}\left[\left(\prod_{m=1}^{v_0} \bar{A}_{ji_m} \bar{A}_{ji'_m}\right) \left(\prod_{m=v_0+1}^{v} \bar{A}_{ji_m}^{\gamma_{v;m}} \bar{A}_{ji'_m}^{\gamma_{v;m}}\right)\right] \\
= \mathbb{E}\left[\left(\prod_{m=1}^{v_1} \bar{A}_{ji_l}^{\gamma_{v;m}+\gamma_{v;\sigma_2(m)}}\right)\right] \left(\prod_{m=v_1+1}^{v} \mathbb{E}[\bar{A}_{ji_m}^{\gamma_{v;m}}]\mathbb{E}[\bar{A}_{ji'_m}^{\gamma_{v;m}}]\right) \\
= O\left(p_n^{2v-v_1}\right).$$
(62)

Note that $2(r+v) - (r_1 + v_1) \le 2(s+t) - 1$. By (56), (57), (61), (62) and (C5) of Assumption 1, the sum in (56) over the indices i = i', j = j', (59)

and (60) is bounded by

$$(np_n)^{2+2r-r_1+2v-v_1} f^{(s,t)}(np_n, np_n)^2 \leq (np_n)(np_n)^{2(s+t)} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(63)

Consider case (ii) i = j' and j = i'. If there is an index l_1 with $1 \le l_1 \le r_0$ such that $j_{l_1} \notin \{i'_1, \ldots, i'_v\}$, then E = 0 (the same argument as in (58)). If there is an index m_1 with $1 \le m_1 \le v_0$ such that $i'_{m_1} \notin \{j_1, \ldots, j_r\}$, then E = 0 (the same argument as in (58)). Then $E \ne 0$ implies the following

$$\{j_1, j_2, \dots, j_{r_0}\} \subset \{i'_1, i'_2, \dots, i'_v\}, \quad \{i'_1, i'_2, \dots, i'_{v_0}\} \subset \{j_1, j_2, \dots, j_r\},$$
$$\{i_1, i_2, \dots, i_{v_0}\} \subset \{j'_1, j'_2, \dots, j'_v\}, \quad \{j'_1, j'_2, \dots, j'_{r_0}\} \subset \{i_1, i_2, \dots, i_v\}.$$

Without loss of generality, suppose

$$\{j_1, j_2, \dots, j_{r_1}\} = \{i'_1, i'_2, \dots, i'_{r_1}\}, \qquad \{j_{r_1+1}, \dots, j_r\} \cap \{i'_{r_1+1}, \dots, i'_v\} = \emptyset,$$

$$\{i_1, i_2, \dots, i_{v_1}\} = \{j'_1, j'_2, \dots, j'_{v_1}\}, \qquad \{i_{v_1+1}, \dots, i_v\} \cap \{j'_{v_1+1}, \dots, j'_v\} = \emptyset,$$

$$(65)$$

for some r_1 with $\max\{r_0, v_0\} \leq r_1 \leq \min\{r, v\}$ and v_1 with $\max\{r_0, v_0\} \leq v_1 \leq \min\{r, v\}$. There are at most $n^{2+2r-r_1+2v-v_1}$ possible choices for indices $i, j, i_1, \ldots, i_v, j_1, \ldots, j_r, i', j', i'_1, \ldots, i'_v, j'_1, \ldots, j'_r$ satisfying i = j', j = i', (64) and (65). Let σ_1 be a permutation of $\{1, 2, \ldots, r_1\}$ such that $j_l = i'_{\sigma_1(l)}$ and σ_2 be a permutation of $\{1, 2, \ldots, v_1\}$ such that $i_m = j'_{\sigma_2(m)}$. Then

$$E = \mathbb{E}\left[\left(\prod_{l=1}^{r_1} \bar{A}_{ij_l}^{\lambda_{r;l}+\gamma_{v;\sigma_1(l)}}\right) \left(\prod_{l=r_1+1}^{r} \bar{A}_{ij_l}^{\lambda_{r;l}}\right) \left(\prod_{m=r_1+1}^{r} \bar{A}_{ii'_m}^{\gamma_{v;m}}\right) \\ \times \left(\prod_{m=1}^{v_1} \bar{A}_{ji_m}^{\lambda_{r;m}+\gamma_{v;\sigma_2(m)}}\right) \left(\prod_{m=v_1+1}^{v} \bar{A}_{ji_m}^{\gamma_{v;m}}\right) \left(\prod_{m=v_1+1}^{r} \bar{A}_{jj'_m}^{\lambda_{v;m}}\right)\right] \\ = \Theta\left(p_n^{2(r+v)-r_1-v_1}\right).$$

Then the sum in (56) over the indices i = j', j = i', (64) and (65) is

bounded by

$$(np_n)^{2+2r-r_1+2v-v_1} f^{(s,t)}(np_n, np_n)^2 \le (np_n)(np_n)^{2(s+t)} f^{(s,t)}(np_n, np_n)^2 = o\left(\sigma_n^2\right).$$
(66)

Consider the case $i \neq i'$ and j = j'. For any $l_0 \in \{1, 2, ..., r\}$, $\{i, j_{l_0}\} \neq \{j, i_l\}$ and $\{i, j_{l_0}\} \neq \{j, i'_l\}$ for any $1 \leq l \leq v$. If $r_0 \geq 2$, it is not possible that $\{i, j_1\} = \{i', j'_{l_1}\}$ and $\{i, j_2\} = \{i', j'_{l_2}\}$ for distinct l_1 and l_2 . Then E = 0 (the same argument as in (58)). Let $r_0 = 1$. In this case, $i = j'_1$ and $i' = j_1$. Otherwise E = 0. Suppose $\{i_1, \ldots, i_{v_1}\} = \{i'_1, \ldots, i'_{v_1}\}$ and $\{i_{v_1+1}, \ldots, i_v\} \cap \{i'_{v_1+1}, \ldots, i'_v\} = \emptyset$ for $v_0 \leq v_1 \leq v$. Suppose $\{j_2, \ldots, j_{r_1}\} = \{j'_2, \ldots, j'_{r_1}\}$ and $\{j_{r_1+1}, \ldots, j_r\} \cap \{j'_{r_1+1}, \ldots, j'_r\} = \emptyset$ for $1 \leq r_1 \leq r$. There are at most $n^{2+2(r+v)-v_1-r_1}$ choices for the indices $i, j, i_1, \ldots, i_v, j_1, \ldots, j_r, i', j', i'_1, \ldots, i'_v, j'_1, \ldots, j'_r$ satisfying these conditions. Then

$$\begin{split} E &= \mathbb{E}\bigg[\bar{A}_{ij_1}^2 \left(\prod_{l=2}^r \bar{A}_{ij_l}^{\lambda_{r;l}}\right) \left(\prod_{l=2}^{r_1} \bar{A}_{j_1j_l}^{\lambda_{r;l}}\right) \left(\prod_{l=r_1+1}^r \bar{A}_{j_1j_l'}^{\lambda_{r;l}}\right)\bigg] \\ &\times \mathbb{E}\bigg[\left(\prod_{m=1}^{v_1} \bar{A}_{ji_m}^{2\gamma_{v;m}}\right) \left(\prod_{m=v_1+1}^v \bar{A}_{ji_m}^{\gamma_{v;m}} \bar{A}_{ji_m'}^{\gamma_{v;m}}\right)\bigg] \\ &= O\left(p_n^{1+2(r-1)+v_1+2(v-v_1)}\right), \end{split}$$

and hence the sum over $i \neq i'$ and j = j' in (56) is bounded by

$$(np_n)^{2(r+v)-v_1+1}f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(67)

Consider the case i = i' and $j \neq j'$. If $v_0 \geq 1$, the summation is similarly bounded by (67). Suppose $v_0 = 0$. Suppose $\{j_1, \ldots, j_{r_1}\} =$ $\{j'_1, \ldots, j'_{r_1}\}$ and $\{j_{r_1+1}, \ldots, j_r\} \cap \{j'_{r_1+1}, \ldots, j'_r\} = \emptyset$ for $r_0 \leq r_1 \leq r$. There are at most n^{3+2r-r_1+2v} choices for the indices $i, j, i_1, \ldots, i_v, j_1, \ldots, j_r, i', j', i'_1, \ldots, i'_v, j'_1, \ldots, j'_r$ satisfying these conditions. Then

$$E = \mathbb{E}\left[\left(\prod_{l=1}^{r_1} \bar{A}_{ij_l}^{2\lambda_{r;l}}\right) \left(\prod_{l=r_1+1}^r \bar{A}_{ij_l}^{\lambda_{r;l}}\right) \left(\prod_{l=r_1+1}^r \bar{A}_{ij_l'}^{\lambda_{r;l}}\right)\right]$$

$$\times \left(\prod_{m=1}^{v} \bar{A}_{ji_m}^{\gamma_{v;m}}\right) \left(\prod_{m=1}^{v} \bar{A}_{j'i_m'}^{\gamma_{v;m}}\right) \\ = O\left(p_n^{2r-r_1+2v}\right).$$

Then the sum over i = i' and $j \neq j'$ in (56) is bounded by

$$n(np_n)^{2+2r-r_1+2\nu}f^{(s,t)}(np_n,np_n)^2 = o(\sigma_n^2).$$
(68)

Suppose i = j' and $j \neq i'$. If $r_0 \geq 2$ or $v_0 \geq 2$, then E = 0 (similar to the argument in the case $i \neq i'$ and j = j'). Let $r_0 = 1$. If $v_0 = 1$, then $j = j'_1$ and $i_1 = i'$. If $v_0 = 0$, then $j_1 = i'_{m_1}$ for some $1 \leq m_1 \leq v$, $j = j'_{l_1}$ for some $1 \leq l_1 \leq r$ and $i_1 = i'$. Without loss of generality, let $m_1 = l_1 = 1$. Suppose

$$\{j_1, \dots, j_{r_1}\} = \{i'_1, \dots, i'_{r_1}\}, \quad \{j_{r_1+1}, \dots, j_r\} \cap \{i'_{r_1+1}, \dots, i'_v\} = \emptyset.$$
$$\{i_2, \dots, i_{v_1}\} = \{j'_2, \dots, j'_{v_1}\}, \quad \{i_{v_1+1}, \dots, i_v\} \cap \{j'_{v_1+1}, \dots, j'_r\} = \emptyset,$$

where $1 \leq r_1, v_1 \leq \min\{r, v\}$. There are at most $n^{2(r+v)-r_1-v_1+2}$ choices for the indices $i, j, i_1, \ldots, i_v, j_1, \ldots, j_r, i', j', i'_1, \ldots, i'_v, j'_1, \ldots, j'_r$ satisfying these conditions. In this case,

$$E = \mathbb{E}\left[\bar{A}_{ji_1}^2 \left(\prod_{l=1}^{r_1} \bar{A}_{ij_l}^{\lambda_{r;l}+\gamma_{v;l}}\right) \left(\prod_{l=r_1+1}^r \bar{A}_{ij_l}^{\lambda_{r;l}}\right) \left(\prod_{l=r_1+1}^v \bar{A}_{ii'_l}^{\gamma_{v;l}}\right)\right] \\ \times \mathbb{E}\left[\left(\prod_{m=2}^v \bar{A}_{ji_m}^{\gamma_{v;m}}\right) \left(\prod_{l=2}^v \bar{A}_{i_1i_l}^{\lambda_{r;l}}\right) \left(\prod_{l=v_1+1}^r \bar{A}_{i_1j'_l}^{\lambda_{r;l}}\right)\right] \\ = O\left(p_n^{1+r+(r-v_1)+v-r_1+v-1+v_1-1}\right).$$

Then the sum over i = j' and $j \neq i'$ in (56) is bounded by

$$(np_n)^{2(r+v)-r_1-1}f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(69)

The case $i \neq j'$ and j = i' can be similarly bounded as in (69).

Now we consider $\{i, j\} \cap \{i', j'\} = \emptyset$. If $r_0 \ge 3$, then at least one of $\{i, j_1\}$, $\{i, j_2\}$ and $\{i, j_3\}$ is not in $\{\{j', i'_{m_1}\}, \{i', j'_l\}\}$ for any m, m_1, l .

Hence E = 0. Similarly, if $v_0 \ge 3$, E = 0.

Suppose $r_0 = v_0 = 2$. Then $\{i, j_1\} = \{i', j'_l\}$ for some $1 \le l \le r$ or $\{i, j_1\} = \{j', i'_m\}$ for some $1 \le m \le v$. Otherwise E = 0. Without loss of generality, suppose $\{i, j_1\} = \{i', j'_l\}$. In this case, $i = j'_l$ and $j_1 = i'$. If $l \geq 3$, then $\{i', j_1'\} = \{j, i_{m_1}\}$ and $\{i', j_2'\} = \{j', i_{m_2}'\}$ or $\{i', j_1'\} = \{j', i_{m_2}'\}$ and $\{i', j_2'\} = \{j, i_{m_1}\}$ (otherwise E = 0). Either case is impossible, due to the fact that $j' \neq i'$, $j' \neq j'_{l_1}$ for any l_1 . Hence l = 1 or l = 2. Without loss of generality, let l = 1. In this case, $\{i', j'_2\} = \{j, i_{m_3}\}$ and $i' = i_{m_3}$ and $j'_2 = j$. If $m_3 \ge 3$, then $\{j, i_1\} = \{j', i'_{m_4}\}$ and $\{j, i_2\} = \{j', i'_{m_5}\}$ (otherwise E = 0), which is not possible. Hence $m_3 = 1$ or $m_3 = 2$. Let $m_3 = 1$ (the argument for $m_3 = 2$ is the same). Then $\{j, i_2\} = \{j', i'_{m_6}\}$. If $m_6 \ge 3$, then $\{j', i'_1\} = \{i, j_{l_2}\}$ and $\{j', i'_2\} = \{i, j_{l_3}\}$ (otherwise E = 0), which is not possible. Hence, $m_6 = 1$ or $m_6 = 2$. Let $m_6 = 1$ (the argument for $m_6 = 2$ is the same). Then $\{j', i_2'\} = \{i, j_2\},\$ $i = i'_2$ and $j_2 = j'$. There are at most $n^{2(r+v)-4}$ possible choices of the indices $i, j, i_1, \ldots, i_v, j_1, \ldots, j_r, i', j', i'_1, \ldots, i'_v, j'_1, \ldots, j'_r$ satisfying these conditions. Then the sum in (56) over these indices is bounded by

$$(np_n)^{2(r+v)-4} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(70)

Suppose $r_0 = 2$ and $v_0 = 1$. Then $\{i, j_1\} = \{i', j'_l\}$ and $\{i, j_2\} = \{j', i'_m\}$ or $\{i, j_1\} = \{j', i'_m\}$ and $\{i, j_2\} = \{i', j'_l\}$ for some l, m. Otherwise E = 0. Without loss of generality, let $\{i, j_1\} = \{i', j'_l\}$ and $\{i, j_2\} = \{j', i'_m\}$. By a similar argument as in the previous paragraph, l = 1 or l = 2. Let l = 1. Then $\{i', j'_2\} = \{j, i_{m_3}\}$. If $m_3 = 1$ and m = 1, the sum in (56) over these indices is bounded by

$$p_n(np_n)^{2(r+v)-2} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2),$$
(71)

If $m_3 \ge 2$ or $m \ge 2$, the sum in (56) over these indices is bounded by

$$p_n^2 (np_n)^{2(r+\nu)-4} f^{(s,t)} (np_n, np_n)^2 = o(\sigma_n^2).$$
(72)

Suppose $r_0 = 2$ and $v_0 = 0$. Then $\{i, j_1\} = \{i', j'_l\}$ and $\{i, j_2\} = \{j', i'_m\}$ or $\{i, j_1\} = \{j', i'_m\}$ and $\{i, j_2\} = \{i', j'_l\}$ for some l, m. Otherwise

E = 0. Without loss of generality, let $\{i, j_1\} = \{i', j'_l\}$ and $\{i, j_2\} = \{j', i'_m\}$. Then l = 1 or l = 2. Without loss of generality, let l = 1. Then $\{i', j'_2\} = \{j, i_{m_1}\}$. The sum in (56) over these indices is bounded by

$$p_n(np_n)^{2(r+\nu)-2} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(73)

Suppose $r_0 = 1$ and $v_0 = 1$. Then $\{i, j_1\} = \{i', j'_l\}$ or $\{i, j_1\} = \{j', i'_m\}$ for some l, m. Without loss of generality, let $\{i, j_1\} = \{i', j'_l\}$. Suppose l = 1. Then $\{j, i_1\} = \{i', j'_{l_1}\}$ or $\{j, i_1\} = \{j', i'_{m_1}\}$. If $\{j, i_1\} = \{j', i'_{m_1}\}$ with $m_1 = 1$, the sum in (56) over these indices is bounded by

$$(np_n)^{2(r+v)} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(74)

If $m_1 \ge 2$, then $\{j', i'_1\} = \{i, j_{m_2}\}$ for $m_2 \ge 2$. Then the sum in (56) over these indices is bounded by

$$p_n(np_n)^{2(r+v)-2} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(75)

The case $\{j, i_1\} = \{i', j'_{l_1}\}$ can be similarly bounded as in (75).

Suppose $l \geq 2$. Then $\{i', j_1'\} = \{j, i_{m_1}\}$. If $m_1 = 1$, then $\{j', i_1'\} = \{i, j_{l_1}\}$ for $l_1 \geq 2$ or $\{j', i_1'\} = \{j, i_{m_2}\}$ for $m_2 \geq 2$. In this case, the sum in (56) over these indices is bounded by

$$p_n(np_n)^{2(r+v)-2} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(76)

If $m_1 \ge 2$, then $\{j', i'_1\} = \{j, i_1\}$. The sum in (56) over these indices is bounded by

$$p_n(np_n)^{2(r+v)-2} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(77)

Suppose $r_0 = 1$ and $v_0 = 0$. Then $\{i, j_1\} = \{i', j'_{l_1}\}$ or $\{i, j_1\} = \{j', i'_{m_1}\}$. Suppose $\{i, j_1\} = \{j', i'_{m_1}\}$. Then $\{i', j'_1\} = \{j, i_{m_2}\}$. In this case, the sum in (56) over these indices is bounded by

$$(np_n)^{2(r+v)} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(78)

Suppose $\{i, j_1\} = \{i', j'_{l_1}\}$. If $l_1 = 1$, then the sum in (56) over these indices is bounded by

$$n(np_n)^{2(r+\nu)+1} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(79)

If $l_1 \ge 2$, then $\{i', j'_1\} = \{j, i_{m_3}\}$. In this case, the sum in (56) over these indices is bounded by

$$(np_n)^{2(r+v)}f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(80)

Now we consider the case (II): $\lambda_{r;l} \geq 2$ for all $l = 1, 2, \ldots, r$ and $\gamma_{v;m} \geq 2$ for all $m = 1, 2, \ldots, v$. In this case, $r \leq \frac{s}{2}$ and $v \leq \frac{t}{2}$. The expectation of V_{rv} is equal to

$$\mathbb{E}[V_{rv}] \\
= \sum_{\substack{i \neq j \\ j_1, j_2, \dots, j_r \notin \{i, j\} \\ j_1 \neq j_2 \neq \dots \neq j_r \\ i_1, i_2, \dots, i_v \notin \{i, j\} \\ i_1 \neq i_2 \neq \dots \neq i_v}} p_n w_{ij} f^{(s,t)}(w_{i(j)}, w_{j(i)}) \prod_{l=1}^r \mathbb{E}[(A_{ij_l} - p_n w_{ij_l})^{\lambda_{r;l}}] \\
\times \prod_{m=1}^v \mathbb{E}[(A_{ji_m} - p_n w_{ji_m})^{\gamma_{v;m}}].$$
(81)

Next we bound the variance of V_{rv} , that is, $Var(V_{rv}) = \mathbb{E}\left[(V_{rv} - \mathbb{E}[V_{rv}])^2\right]$. Let $\eta_l \in \{0,1\}$ for l = 1, 2, ..., r and $\xi_l \in \{0,1\}$ for l = 1, 2, ..., v. Then we have

$$\prod_{l=1}^{r} (A_{ij_{l}} - p_{n}w_{ij_{l}})^{\lambda_{r;l}} \prod_{m=1}^{v} (A_{ji_{m}} - p_{n}w_{ji_{m}})^{\gamma_{v;m}}$$

$$= \prod_{l=1}^{r} \left[(A_{ij_{l}} - p_{n}w_{ij_{l}})^{\lambda_{r;l}} - \mathbb{E} \left[(A_{ij_{l}} - p_{n}w_{ij_{l}})^{\lambda_{r;l}} \right] \right]$$

$$+ \mathbb{E} \left[(A_{ij_{l}} - p_{n}w_{ij_{l}})^{\gamma_{v;l}} - \mathbb{E} \left[(A_{ji_{l}} - p_{n}w_{ji_{l}})^{\gamma_{v;l}} \right] \right]$$

$$+ \mathbb{E} \left[(A_{ji_{l}} - p_{n}w_{ji_{l}})^{\gamma_{v;l}} - \mathbb{E} \left[(A_{ji_{l}} - p_{n}w_{ji_{l}})^{\gamma_{v;l}} \right] \right]$$

$$= \sum_{\substack{\eta_{1},...,\eta_{r} \in \{0,1\} \\ \xi_{1},...,\xi_{v} \in \{0,1\} \\}} \prod_{l=1}^{r} \left[(A_{ij_{l}} - p_{n}w_{ij_{l}})^{\lambda_{r;l}} - \mathbb{E} \left[(A_{ij_{l}} - p_{n}w_{ij_{l}})^{\lambda_{r;l}} \right] \right]^{\eta_{l}} \\\times \left[\mathbb{E} \left[(A_{ij_{l}} - p_{n}w_{ij_{l}})^{\lambda_{r;l}} \right] \right]^{1-\eta_{l}} \\\times \prod_{l=1}^{r} \left[(A_{ji_{l}} - p_{n}w_{ji_{l}})^{\gamma_{v;l}} - \mathbb{E} \left[(A_{ji_{l}} - p_{n}w_{ji_{l}})^{\gamma_{v;l}} \right] \right]^{\xi_{l}} \\\times \left[\mathbb{E} \left[(A_{ji_{l}} - p_{n}w_{ji_{l}})^{\gamma_{v;l}} \right] \right]^{1-\xi_{l}}.$$
(82)

For convenience, denote

$$X_{ij}(x) = (A_{ij} - p_n w_{ij})^x - \mathbb{E}[(A_{ij} - p_n w_{ij})^x]$$

and $Y_{ij}(x) = \mathbb{E}[(A_{ij} - p_n w_{ij})^x].$

Since $i \neq j$, $j_l \notin \{i, j\}$ for $l = 1, 2, \ldots, r$ and $i_l \notin \{i, j\}$ for $l = 1, 2, \ldots, v$, then $\{i, j_l\} \neq \{j, i_m\}$ for any $l = 1, 2, \ldots, r$ and $m = 1, 2, \ldots, v$. Consequently, A_{ij_l} $(l = 1, 2, \ldots, r)$ and A_{ji_m} $(m = 1, 2, \ldots, v)$ are independent.

By (53), (81) and (82), $V_{rv} - \mathbb{E}[V_{rv}]$ does not contain the term corresponding to $\eta_l = 0$ for all $l \in \{1, 2, \ldots, r\}$ and $\xi_m = 0$ for all $m \in \{1, 2, \ldots, v\}$. Hence we assume $\eta_1 + \cdots + \eta_r \ge 1$ or $\xi_1 + \cdots + \xi_v \ge 1$. Without loss of generality, let $\eta_1 = \eta_2 = \cdots = \eta_{l_0} = 1$, $\eta_{l_0+1} = \cdots = \eta_r = 0$, $\xi_1 = \xi_2 = \cdots = \xi_{m_0} = 1$ and $\xi_{m_0+1} = \cdots = \xi_v = 0$, where $1 \le l_0 + m_0 \le r + v$. In this case,

$$\prod_{l=1}^{r} \left[X_{ij_{l}}(\lambda_{r;l}) \right]^{\eta_{l}} \left[Y_{ij_{l}}(\lambda_{r;l}) \right]^{1-\eta_{l}} = \left(\prod_{l=1}^{l_{0}} X_{ij_{l}}(\lambda_{r;l}) \right) \left(\prod_{l=l_{0}+1}^{r} Y_{ij_{l}}(\lambda_{r;l}) \right),$$
$$\prod_{l=1}^{v} \left[X_{ji_{l}}(\gamma_{v;l}) \right]^{\xi_{l}} \left[Y_{ji_{l}}(\gamma_{v;l}) \right]^{1-\xi_{l}} = \left(\prod_{l=1}^{m_{0}} X_{ji_{l}}(\gamma_{v;l}) \right) \left(\prod_{l=m_{0}+1}^{v} Y_{ji_{l}}(\gamma_{v;l}) \right).$$

Denote

$$V_{rv}(l_0, m_0)$$

$$= \sum_{\substack{i \neq j \\ j_{1}, j_{2}, \dots, j_{r} \notin \{i, j\} \\ j_{1} \neq j_{2} \neq \dots \neq j_{r} \\ i_{1}, i_{2}, \dots, i_{v} \notin \{i, j\} \\ i_{1} \neq i_{2} \neq \dots \neq i_{v}}} p_{n} w_{ij} f^{(s,t)}(w_{i(j)}, w_{j(i)}) \left(\prod_{l=1}^{l_{0}} X_{ij_{l}}(\lambda_{r;l})\right) \\ \times \left(\prod_{l=l_{0}+1}^{r} Y_{ij_{l}}(\lambda_{r;l})\right) \left(\prod_{l=1}^{m_{0}} X_{ji_{l}}(\gamma_{v;l})\right) \left(\prod_{l=m_{0}+1}^{v} Y_{ji_{l}}(\gamma_{v;l})\right). (83)$$

We only need to consider the variance of $V_{rv}(l_0, m_0)$. Since $\lambda_{r;l} \ge 2$ and $w_{ij} \in [\beta, 1]$, then

$$\mathbb{E}[(A_{ij_l} - p_n w_{ij_l})^{\lambda_{r;l}}]$$

= $(1 - p_n w_{ij_l})^{\lambda_{r;l}} p_n w_{ij_l} + (-p_n w_{ij_l})^{\lambda_{r;l}} (1 - p_n w_{ij_l})$
= $\Theta(p_n).$

Hence one has

$$\prod_{l=l_0+1}^{r} Y_{ij_l}(\lambda_{r;l}) = \Theta(p_n^{r-l_0}), \qquad \prod_{l=m_0+1}^{v} Y_{ji_l}(\gamma_{v;l}) = \Theta(p_n^{v-m_0}).$$

For convenience, denote

$$a_{ij_1,\dots,j_{l_0}} = \sum_{j_{l_0+1},\dots,j_r \notin \{j_1,\dots,j_{l_0}\}} \prod_{l=l_0+1}^r Y_{ij_l}(\lambda_{r;l}),$$

$$b_{ji_1,\dots,i_{m_0}} = \sum_{i_{m_0+1},\dots,i_v \notin \{i_1,\dots,i_{m_0}\}} \prod_{l=m_0+1}^v Y_{ji_l}(\gamma_{v;l}).$$

Then

$$a_{ij_1,\dots,j_{l_0}} = \Theta((np_n)^{r-l_0}), \qquad b_{ji_1,\dots,i_{m_0}} = \Theta((np_n)^{v-m_0}).$$
(84)

In this case, $V_{rv}(l_0, m_0)$ is written as

$$V_{rv}(l_{0}, m_{0}) = \sum_{\substack{i \neq j \\ j_{1}, j_{2}, \dots, j_{r} \notin \{i, j\} \\ j_{1} \neq j_{2} \neq \dots \neq j_{r} \\ i_{1}, i_{2}, \dots, i_{v} \notin \{i, j\} \\ i_{1} \neq i_{2} \neq \dots \neq i_{v}}} p_{n} w_{ij} f^{(s,t)}(w_{i(j)}, w_{j(i)}) a_{ij_{1}, \dots, j_{l_{0}}} b_{ji_{1}, \dots, i_{m_{0}}} \\ \times \left(\prod_{i=1}^{l_{0}} X_{ij_{i}}(\lambda_{r; l})\right) \left(\prod_{l=1}^{m_{0}} X_{ji_{l}}(\gamma_{v; l})\right).$$
(85)

Next we bound the variance of $V_{rv}(l_0, m_0)$. Since $i \neq j$, $j_l \notin \{i, j\}$ and $i_m \notin \{i, j\}$ for all l, m, $X_{ij_l}(\lambda_{r;l})$ and $X_{ji_l}(\gamma_{v;l})$ are independent. By definition, $\mathbb{E}[X_{ij_l}(\lambda_{r;l})] = \mathbb{E}[X_{ji_l}(\gamma_{v;l})] = 0$. Then $\mathbb{E}[V_{rv}(l_0, m_0)] = 0$ and $Var(V_{rv}(l_0, m_0)) = \mathbb{E}[V_{rv}(l_0, m_0)^2]$, that is,

$$Var\left(V_{rv}(l_{0}, m_{0})\right) = \mathbb{E}\left[\sum_{i,j,j_{1},...,j_{l_{0}},i_{1},...,i_{m_{0}}} p_{n}w_{ij}f^{(s,t)}(w_{i(j)}, w_{j(i)})a_{ij_{1},...,j_{l_{0}}}b_{ji_{1},...,i_{m_{0}}}\right] \times \left(\prod_{l=1}^{l} X_{ij_{l}}(\lambda_{r;l})\right) \left(\prod_{l=1}^{m_{0}} X_{ji_{l}}(\gamma_{v;l})\right)\right]^{2} = \sum_{\substack{i,j,j_{1},...,j_{l_{0}},i_{1},...,i_{m_{0}}\\i',j',j'_{1},...,j_{l_{0}},i'_{1},...,i'_{m_{0}}}} p_{n}w_{ij}f^{(s,t)}(w_{i(j)}, w_{j(i)})p_{n}w_{i'j'}f^{(s,t)}(w_{i'(j')}, w_{j'(i')}) \times a_{ij_{1},...,i'_{m_{0}}} a_{i'j'_{1},...,j'_{l_{0}}}b_{j'i'_{1},...,i'_{m_{0}}}} \times \mathbb{E}\left(\prod_{l=1}^{l} X_{ij_{l}}(\lambda_{r;l})X_{i'j'_{l}}(\lambda_{r;l})\prod_{l=1}^{m_{0}} X_{ji_{l}}(\gamma_{v;l})X_{j'i'_{l}}(\gamma_{v;l})\right).$$
(86)

For each j_{l_1} with $1 \leq l_1 \leq l_0$, if $\{i, j_{l_1}\} \neq \{i', j'_l\}$ or $\{j', i'_m\}$ for some $l \in \{1, 2, \ldots, l_0\}$ or $m \in \{1, 2, \ldots, m_0\}$, then $X_{ij_{l_1}}(\lambda_{r;l_1})$ is independent of $X_{ji_l}(\lambda_{r;l}), X_{i'j'_l}(\lambda_{r;l})$ and $X_{j'i'_m}(\lambda_{r;m})$. In this case,

$$\mathbb{E}\left[\prod_{l=1}^{l_{0}} X_{ij_{l}}(\lambda_{r;l}) X_{i'j_{l}'}(\lambda_{r;l}) \prod_{l=1}^{m_{0}} X_{ji_{l}}(\gamma_{v;l}) X_{j'i_{l}'}(\gamma_{v;l})\right]$$

$$= \mathbb{E}\left[X_{ij_{l_{1}}}\right] \mathbb{E}\left[\prod_{l=1, l \neq l_{1}}^{l_{0}} X_{ij_{l}}(\lambda_{r;l}) X_{i'j_{l}'}(\lambda_{r;l}) \prod_{l=1}^{m_{0}} X_{ji_{l}}(\gamma_{v;l}) X_{j'i_{l}'}(\gamma_{v;l})\right]$$

$$= 0.$$
(87)

Hence, for each j_{l_1} with $1 \leq l_1 \leq l_0$, there exists l or m such that $\{i, j_{l_1}\} = \{i', j'_l\}$ or $\{i, j_{l_1}\} = \{j', i'_m\}$. Otherwise, (87) holds. Moreover, if $\{i, j_{l_1}\} = \{i', j'_l\}$ and $\{i, j_{l_1}\} = \{j', i'_m\}$, then $\{i', j'_l\} = \{j', i'_m\}$. In this case, $i' = i'_m$ and $j' = j'_l$ (since $i' \neq j'$). This is not possible, due to the fact that $i'_m \notin \{i', j'\}$ and $j'_l \notin \{i', j'\}$ for all m, l. Therefore, $\{i, j_{l_1}\}$ can only be equal to one of $\{i', j'_l\}$ and $\{j', i'_m\}$, but not both. Similarly, $\{j, i_m\}$ can only be equal to one of $\{i', j'_l\}$ and $\{j', i'_m\}$.

Let $m_0 = 0$. If $l_0 \ge 2$, then i = i', $\{j_1, j_2, \dots, j_{l_0}\} = \{j'_1, j'_2, \dots, j'_{l_0}\}$. Then

$$\mathbb{E}\left[\prod_{l=1}^{l_0} X_{ij_l}(\lambda_{r;l}) X_{i'j_l'}(\lambda_{r;l}) \prod_{l=1}^{m_0} X_{ji_l}(\gamma_{v;l}) X_{j'i_l'}(\gamma_{v;l})\right] \\ = \mathbb{E}\left[\prod_{l=1}^{l_0} X_{ij_l}(\lambda_{r;l})^2\right] = O(p_n^{l_0}).$$
(88)

Then the sum over these indices in (86) is bounded by

$$O\left(n(np_n)^{2(r+v)-l_0+2}f^{(s,t)}(np_n,np_n)^2\right) = o(\sigma_n^2).$$
(89)

If $l_0 = 1$, then $\{i, j_1\} = \{i', j'_1\}$. In this case, (89) still holds.

Let $m_0 = 1$. If $l_0 \ge 3$, then i = i', $\{j_1, j_2, \dots, j_{l_0}\} = \{j'_1, j'_2, \dots, j'_{l_0}\}$ and $\{j, i_1\} = \{j', i'_1\}$. Then the sum over these indices in (86) is bounded by

$$O\left((np_n)^{2(r+\nu)-l_0+1}f^{(s,t)}(np_n,np_n)^2\right) = o(\sigma_n^2).$$
(90)

If $l_0 = 2$, there are two situations: (i) i = i', $\{j_1, j_2\} = \{j'_1, j'_2\}$ and $\{j, i_1\} = \{j', i'_1\}$; (ii) $\{i, j_1\} = \{j'_{l_1}, i'\}$, $\{i', j'_{l_2}\} = \{j, i_1\}$, $\{i, j_2\} = \{j', i'_1\}$ with $l_1 \neq l_2$. For the case (i), the sum over these indices in (86) is bounded by (90) with $l_0 = 2$. For case (ii), the sum over these indices in (86) is bounded by

$$O\left((np_n)^{2(r+v)-2}f^{(s,t)}(np_n, np_n)^2\right) = o(\sigma_n^2).$$
(91)

If $l_0 = 1$, then $\{i, j_1\} = \{i', j'_1\}$ and $\{j, i_1\} = \{j', i'_1\}$ or $\{i, j_1\} = \{j', i'_1\}$ and $\{j, i_1\} = \{i', j'_1\}$. Either case, the sum over these indices in (86) is bounded by

$$O\left((np_n)^{2(r+v)}f^{(s,t)}(np_n,np_n)^2\right) = o(\sigma_n^2).$$
(92)

If $l_0 = 0$, then (89) holds with l_0 replaced by $m_0 = 1$.

Let $m_0 = 2$. If $l_0 \ge 3$, then i = i', $\{j_1, j_2, \dots, j_{l_0}\} = \{j'_1, j'_2, \dots, j'_{l_0}\}, j = j'$ and $\{i_1, i_2\} = \{i'_1, i'_2\}$. In this case, the sum over these indices in (86) is bounded by

$$O\left((np_n)^{2(r+v)-l_0}f^{(s,t)}(np_n,np_n)^2\right) = o(\sigma_n^2).$$
(93)

Let $l_0 = 2$. There are two cases (i) i = i', $\{j_1, j_2\} = \{j'_1, j'_2\}$, j = j' and $\{i_1, i_2\} = \{i'_1, i'_2\}$; (ii) $\{i, j_1\} = \{j'_{l_1}, i'\}$, $\{j'_{l_2}, i'\} = \{j, i_{m_1}\}$, $\{j, i_{m_2}\} = \{j', i'_{m_3}\}$, $\{j', i'_{m_4}\} = \{i, j_2\}$. For case (i), (93) holds. For case (ii), the sum over these indices in (86) is bounded by

$$O\left((np_n)^{2(r+v)-2}f^{(s,t)}(np_n, np_n)^2\right) = o(\sigma_n^2).$$
(94)

The case $l_0 = 0, 1$ are similar to the case $m_0 = 0, 1$ and $l_0 = 2$ discussed earlier.

The case $m_0 \ge 3$ is similar to the case $l_0 \ge 3$.

Now we study the second moment of (50). Note that

$$\sum_{i \neq j} f^{(s,t)}(w_{i(j)}, w_{j(i)}) (d_{i(j)} - w_{i(j)})^s (d_{j(i)} - w_{j(i)})^t (A_{ij} - p_n w_{ij})$$
$$= \sum_{r=1}^s \sum_{v=1}^t U_{rv},$$

where

$$U_{rv} = \sum_{\substack{i \neq j \ j_{1}, j_{2}, \dots, j_{r} \notin \{i, j\} \\ j_{1} \neq j_{2} \neq \dots \neq j_{r} \\ i_{1}, i_{2}, \dots, i_{v} \notin \{i, j\} \\ i_{1} \neq i_{2} \neq \dots \neq i_{v}}} f^{(s,t)}(w_{i(j)}, w_{j(i)})(A_{ij} - p_{n}w_{ij}) \\ \times \prod_{l=1}^{r} (A_{ij_{l}} - p_{n}w_{ij_{l}})^{\lambda_{r;l}} \prod_{m=1}^{v} (A_{ji_{m}} - p_{n}w_{ji_{m}})^{\gamma_{v;m}}.$$

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Since s, t are fixed finite integers less than k_0 , we only need to bound the variance of U_{rv} . Obviously, $\mathbb{E}[U_{rv}] = 0$. Then $Var(U_{rv})$ is equal to

$$\mathbb{E}[U_{rv}^{2}] = \sum_{\substack{i \neq j \\ i' \neq j' \\ j_{1}, j_{2}, \dots, j_{r} \notin \{i, j\} \\ j_{1} \neq j_{2} \neq \dots \neq j_{r} \\ i_{1}, i_{2}, \dots, i_{v} \notin \{i, j\} \\ j_{1}' \neq i_{2} \neq \dots \neq i_{v} \\ j_{1}' \neq i_{2} \neq \dots \neq i_{v} \\ j_{1}' \neq j_{2}' \neq \dots \neq j_{v}' \\ j_{1}' \neq j_{2}' \neq \dots \neq j_{v}'} f^{(s,t)}(w_{i'(j')}, w_{j'(i')}) \mathbb{E}\left[(A_{ij} - p_{n}w_{ij})(A_{i'j'} - p_{n}w_{i'j'}) \left(\prod_{l=1}^{r} \bar{A}_{ij_{l}}^{\lambda_{r;l}}\right) \right. \\ \times \left(\prod_{l=1}^{r} \bar{A}_{i'j_{l}'}^{\lambda_{r;l}}\right) \left(\prod_{m=1}^{v} \bar{A}_{ji_{m}}^{\gamma_{v;m}}\right) \left(\prod_{m=1}^{v} \bar{A}_{j'i'_{m}}^{\gamma_{v;m}}\right) \right]. \tag{95}$$

Denote the expectation in (95) as E_1 . Fix $r \in \{1, 2, ..., s\}$ and $v \in \{1, 2, ..., t\}$. There are two cases: (a) there are some indices $\lambda_{r;l}$ or $\gamma_{v;m}$ are equal to one; (b) $\lambda_{r;l} \geq 2$ and $\gamma_{v;m} \geq 2$ for all $l \in \{1, 2, ..., r\}$ and $m \in \{1, 2, ..., v\}$.

Consider case (a) first. Suppose there are some indices $\lambda_{r;l}$ or $\gamma_{v;m}$ which are equal to one. Without loss of generality, let $\lambda_{r;1} = \lambda_{r;2} = \cdots = \lambda_{r;r_0} = 1$ and $\lambda_{r;l} \geq 2$ for $l \in \{r_0 + 1, \ldots, r\}$. Let $\gamma_{v;1} = \gamma_{v;2} = \cdots = \gamma_{v;v_0} = 1$ and $\gamma_{v;l} \geq 2$ for $l \in \{v_0 + 1, \ldots, v\}$. Here $r_0 + v_0 \geq 1$. Then

$$E_{1} = \mathbb{E}\left[(A_{ij} - p_{n}w_{ij})(A_{i'j'} - p_{n}w_{i'j'}) \left(\prod_{l=1}^{r_{0}} \bar{A}_{ij_{l}}\right) \left(\prod_{l=r_{0}+1}^{r} \bar{A}_{ij_{l}}^{\lambda_{r;l}}\right) \times \left(\prod_{l=r_{0}+1}^{r} \bar{A}_{i'j_{l}'}^{\lambda_{r;l}}\right) \left(\prod_{l=r_{0}+1}^{r} \bar{A}_{j'i_{l}'}^{\lambda_{r;l}}\right) \left(\prod_{m=1}^{v_{0}} \bar{A}_{ji_{m}}\right) \left(\prod_{m=v_{0}+1}^{v} \bar{A}_{ji_{m}}^{\gamma_{v;m}}\right) \times \left(\prod_{m=1}^{v_{0}} \bar{A}_{j'i_{m}'}\right) \left(\prod_{m=v_{0}+1}^{v} \bar{A}_{j'i_{m}'}^{\gamma_{v;m}}\right) \right].$$

We split the sum of (95) into two cases: $\{i, j\} \neq \{i', j'\}$ and $\{i, j\} = \{i', j'\}$.

Suppose $\{i, j\} \neq \{i', j'\}$. Let $v_0 \geq 2$ and $r_0 \geq 2$. For $1 \leq m_1 \leq v_0$, if $\{j, i_{m_1}\} \neq \{i', j'_l\}$ for any $l, \{j, i_{m_1}\} \neq \{j', i'_m\}$ for any m and $\{j, i_{m_1}\} \neq \{j', i'_m\}$ for any m and $\{j, i_{m_1}\} \neq \{j', j'_m\}$

 $\{i',j'\},$ then $A_{ji_{m_1}}$ is independent of $A_{i'j_l'}$ for all l, $A_{j'i_m'}$ for all m and $A_{i'j'}$. Then

$$E_{1} = \mathbb{E}\left[(A_{ij} - p_{n}w_{ij})(A_{i'j'} - p_{n}w_{i'j'}) \left(\prod_{l=1}^{r_{0}} \bar{A}_{ij_{l}}\right) \left(\prod_{l=r_{0}+1}^{r} \bar{A}_{ij_{l}}^{\lambda_{r;l}}\right) \times \left(\prod_{l=r_{0}+1}^{r_{0}} \bar{A}_{i'j_{l}'}^{\lambda_{r;l}}\right) \left(\prod_{l=r_{0}+1}^{v_{0}} \bar{A}_{i'j_{l}'}^{\lambda_{r;l}}\right) \left(\prod_{m=1,m\neq m_{1}}^{v_{0}} \bar{A}_{ji_{m}}\right) \left(\prod_{m=v_{0}+1}^{v} \bar{A}_{ji_{m}}^{\gamma_{v;m}}\right) \times \left(\prod_{m=1}^{v_{0}} \bar{A}_{j'i_{m}'}^{\lambda_{r;l}}\right) \left(\prod_{m=v_{0}+1}^{v} \bar{A}_{j'i_{m}'}^{\gamma_{v;m}}\right) \right] \mathbb{E}\left[\bar{A}_{ji_{m}}\right] = 0.$$
(96)

Hence, $\{j, i_{m_1}\} = \{i', j_l'\}$ for some l or $\{j, i_{m_1}\} = \{j', i_m'\}$ for some m or $\{j, i_{m_1}\} = \{i', j'\}$. Similar results hold for $\{i, j\}, \{i, j_l\}$ with $1 \le l \le r_0$, $\{j', i_m'\}$ with $1 \le m \le v_0, \{i', j'\}$, and $\{i', j_l'\}$ with $1 \le l \le r_0$. (i) Suppose $\{j, i_1\} = \{i', j_{l_1}'\}$. Then $\{j, i_2\} = \{i', j_{l_2}'\}$ or $\{j, i_2\} = \{i', j'\}$. If $\{j, i_2\} = \{i', j_{l_2}'\}$, then $\{i, j\} = \{i', j_{l_3}'\}$. Since $j_l' \ne j'$ for all l, then either $\{j', i_1'\} \ne \{j_{l_3}', j_l\}$ for all l or $\{j', i_2'\} \ne \{j_{l_3}', j_l\}$ for all l. By a similar argument as in (96), $E_1 = 0$. The same result hods if $\{j, i_2\} = \{i', j'\}$. (ii) Suppose $\{j, i_1\} = \{j', i_{m_1}'\}$ or $\{j, i_1\} = \{i', j'\}$. By a similar argument as in the case $\{j, i_1\} = \{i', j_{l_1}'\}$, it is easy to get $E_1 = 0$.

Let $v_0 \ge 2, r_0 \le 1$. Given $1 \le m \le v_0$, if $\{j, i_m\} = \{i', j'_l\}$ for some l, then $E_1 = 0$. Hence $\{j, i_m\} = \{j', i'_l\}$ for some l. Then j = j' and

$$\{\{j,i\},\{j,i_1\},\ldots,\{j,i_{v_0}\}\} \subset \{\{j',i'\},\{j',i_1'\},\ldots,\{j',i_v'\}\}.$$
(97)

Similarly,

 $\{\{j',i'\},\{j',i_1'\},\ldots,\{j',i_{v_0}'\}\} \subset \{\{j,i\},\{j,i_1\},\ldots,\{j,i_v\}\}.$ (98)

Without loss of generality, let $i = i'_1$, $i_1 = i'$, $i_2 = i'_2$, ..., $i'_{v_1} = i'_{v_1}$ for $v_0 \leq v_1 \leq v$. There are at most $n^{2(r+v)+2-v_1}$ choices of these indices. If $r_0 = 0$, E_1 is bounded by $O\left(p_n^{2(r+v)+1-v_1}\right)$. Then the sum of (95) over these indices is bounded by

$$n(np_n)^{2(r+v)+1-v_1} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(99)

If $r_0 = 1$, then $\{i, j_1\} = \{i', j'_1\}$. In this case, the sum of (95) over these indices is bounded by

$$(np_n)^{2(r+v)-v_1} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(100)

Let $v_0 = r_0 = 1$. In this case, $\{j, i_1\} = \{i', j'_l\}$ for some l or $\{j, i_1\} = \{j', i'_m\}$ for some m or $\{j, i_1\} = \{i', j'\}$. Let $\{j, i_1\} = \{i', j'_{l_1}\}$. If j = i', then $\{i, j\} = \{i', j'_{l_2}\}$. In this case, $\{i', j'\} = \{j, i_{m_1}\}$ for some $m_1 \ge 2$, $\{i, j_1\} = \{j', i'_1\}$ and $l_1 = 1$. Otherwise $E_1 = 0$. There are at most $n^{2(r+v)-2}$ choices of these indices. In this case, the sum of (95) over these indices is bounded by

$$(np_n)^{2(r+v)-2} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(101)

If $j = j'_1$, then $\{i, j\} = \{j', i'_1\}$, $\{i, j_1\} = \{i', j'\}$ and $l_1 = 1$. Otherwise $E_1 = 0$. There are at most $n^{2(r+v)-1}$ choices of these indices. In this case, the sum of (95) over these indices is bounded by

$$(np_n)^{2(r+\nu)-1} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(102)

The cases $\{j, i_1\} = \{j', i'_m\}$ for some m and $\{j, i_1\} = \{i', j'\}$ can be similarly bounded as in (100) and (102).

Let $v_0 = 1$ and $r_0 = 0$. In this case, $\{j, i_1\} = \{i', j'_l\}$ for some l or $\{j, i_1\} = \{j', i'_m\}$ for some m or $\{j, i_1\} = \{i', j'\}$. Let $\{j, i_1\} = \{j', i'_{m_1}\}$. If $m_1 = 1$, then $\{i, j\} = \{j', i'_{m_2}\}$ and $\{i', j'\} = \{j, i_{m_3}\}$. There are at most $n^{2(r+v)-1}$ choices of these indices. Then the sum of (95) over these indices is bounded by (102). If $m_1 \ge 2$ and $m_3 = 1$, then the sum of (95) over these indices is bounded by (102). If $m_1 \ge 2$ and $m_3 \ge 2$, then $\{j', i'_1\} = \{j, i_{m_3}\}$. There are at most $n^{2(r+v)-1}$ choices of these indices. Then the sum of (95) over these indices is bounded by (102). If $m_1 \ge 2$ and $m_3 \ge 2$, then $\{j', i'_1\} = \{j, i_{m_3}\}$. There are at most $n^{2(r+v)-1}$ choices of these indices. Then the sum of (95) over these indices is bounded by $n^{2(r+v)-1}$ choices of these indices.

$$n(np_n)^{2(r+v)-2} f^{(s,t)}(np_n, np_n)^2 = o(\sigma_n^2).$$
(103)

The case $\{j, i_1\} = \{i', j'_l\}$ for some l and $\{j, i_1\} = \{i', j'\}$ can be similarly studied.

Suppose $\{i, j\} = \{i', j'\}$. Consider i = i' and j = j' first. For any m with $1 \leq m \leq v_0$, if $i_m \neq i'_{m_1}$ for all $1 \leq m_1 \leq v$, then the expectation $E_1 = 0$. Hence, $\{i_1, \ldots, i_{v_0}\} \subset \{i'_1, \ldots, i'_v\}$. Similarly, $\{i'_1, \ldots, i'_{v_0}\} \subset \{i_1, \ldots, j_r\}, \{j_1, \ldots, j_r\} \subset \{j'_1, \ldots, j'_r\}$. Without loss of generality, let

$$\{i'_1, \dots, i'_{v_1}\} = \{i_1, \dots, i_{v_1}\}, \qquad \{i'_{v_1+1}, \dots, i'_{v_n}\} \cap \{i_{v_1+1}, \dots, i_{v_n}\} = \emptyset.$$

$$\{j'_1, \dots, j'_{r_1}\} = \{j_1, \dots, j_{r_1}\}, \quad \{j'_{r_1+1}, \dots, j'_r\} \cap \{j_{r_1+1}, \dots, j_r\} = \emptyset,$$

where $v_0 \leq v_1 \leq v$, $r_0 \leq r_1 \leq r$. There are at most $n^{2+r_1+2(r-r_1)+v_1+2(v-v_1)}$ indices. Let σ_1 be a one-to-one map from $\{i_1, \ldots, i_{v_1}\}$ to $\{i'_1, \ldots, i'_{v_1}\}$ and σ_2 be a one-to-one map from $\{j_1, \ldots, j_{r_1}\}$ to $\{j'_1, \ldots, j'_{r_1}\}$. Then

$$E_{1} = \mathbb{E}\left[\bar{A}_{ij}^{2}\left(\prod_{l=1}^{r_{1}}\bar{A}_{ijl}^{\lambda_{r;l}+\lambda_{r;\sigma_{2}(l)}}\right)\left(\prod_{l=r_{1}+1}^{r}\bar{A}_{ijl}^{\lambda_{r;l}}\bar{A}_{ijl'}^{\lambda_{r;l}}\right) \times \left(\prod_{m=1}^{v_{1}}\bar{A}_{ji_{m}}^{\gamma_{v;m}+\gamma_{v;\sigma_{1}(m)}}\right)\left(\prod_{m=v_{1}+1}^{v}\bar{A}_{ji_{m}}^{\gamma_{v;m}}\bar{A}_{ji'_{m}}^{\gamma_{v;m}}\right)\right] \\ = \Theta\left(p_{n}^{1+r_{1}+2(r-r_{1})+v_{1}+2(v-v_{1})}\right).$$
(104)

Then the sum over indices $\{i, j\} = \{i', j'\}$ in (95) is bounded by

$$n(np_n)^{2(r+\nu)-(r_1+\nu_1)+1}f^{(s,t)}(np_n,np_n)^2 = o(\sigma_n^2).$$
 (105)

Similarly, (105) holds for the case i = j' and j = i' with $\max\{r_0, v_0\} \le r_1, v_1 \le \min\{r, v\}.$

Now we consider case (b). Suppose $\lambda_{r;l} \geq 2$ for $l \in \{1, \ldots, r\}$ and $\gamma_{v;m} \geq 2$ for $m \in \{1, \ldots, v\}$. In this case $r \leq \frac{s}{2}$ and $v \leq \frac{t}{2}$. If $\{i, j\} \neq \{i', j'\}, \{i, j\} \neq \{i', j'_l\}$ for all $l, \{i, j\} \neq \{j', i'_m\}$ for all m, then $E_1 = 0$.

Suppose $\{i, j\} = \{i', j'\}$. There are two cases: i = i' and j = j'or i = j' and j = i'. Let i = i' and j = j'. Suppose $|\{j_1, \ldots, j_r\} \cap$ $\{j'_1, \ldots, j'_r\}| = r_1$ and $|\{i_1, \ldots, i_v\} \cap \{i'_1, \ldots, i'_v\}| = v_1$. There are at most $n^{2+r_1+2(r-r_1)+v_1+2(v-v_1)}$ possible indices. Without loss of generality, assume $j_l = j'_l$ for $1 \le l \le r_1$ and $i_l = i'_l$ for $1 \le l \le v_1$. Then

$$E_{1} = \mathbb{E}\left[(A_{ij} - p_{n}w_{ij})^{2} \left(\prod_{l=1}^{r_{1}} \bar{A}_{ij_{l}}^{2\lambda_{r;l}} \right) \left(\prod_{l=r_{1}+1}^{r} \bar{A}_{ij_{l}}^{\lambda_{r;l}} \right) \left(\prod_{l=r_{1}+1}^{r} \bar{A}_{ij_{l}'}^{\lambda_{r;l}} \right) \\ \times \left(\prod_{m=1}^{v_{1}} \bar{A}_{ji_{m}}^{2\gamma_{v;m}} \right) \left(\prod_{m=v_{1}+1}^{v} \bar{A}_{ji_{m}}^{\gamma_{v;m}} \right) \left(\prod_{m=v_{1}+1}^{v} \bar{A}_{ji_{m}'}^{\gamma_{v;m}} \right) \right] \\ = O\left(p_{n}^{1+r_{1}+2(r-r_{1})+v_{1}+2(v-v_{1})} \right).$$
(106)

Then the sum over indices $\{i, j\} = \{i', j'\}$ with i = i' and j = j' in (95) is bounded by

$$n(np_n)^{2(r+v)-r_1-v_1+1}f^{(s,t)}(np_n,np_n)^2 = o(\sigma_n^2).$$

Let i = j' and j = i'. Suppose $|\{j_1, \ldots, j_r\} \cap \{i'_1, \ldots, i'_v\}| = r_1$ and $|\{i_1, \ldots, i_v\} \cap \{j'_1, \ldots, j'_r\}| = v_1$, where $0 \le r_1 \le \min\{r, v\}$ and $0 \le v_1 \le \min\{r, v\}$. There are at most $n^{2+r_1+(r-r_1)+(r-v_1)+v_1+(v-v_1)+(v-r_1)}$ such indices. Without loss of generality, let $j_l = i'_l$ for $l \le r_1$ and $i_l = j'_l$ for $l \le v_1$. Then

$$E_{1} = \mathbb{E}\left[(A_{ij} - p_{n}w_{ij})^{2} \left(\prod_{l=1}^{r_{1}} \bar{A}_{ijl}^{\lambda_{r;l} + \gamma_{v;l}} \right) \left(\prod_{l=r_{1}+1}^{r} \bar{A}_{ijl}^{\lambda_{r;l}} \right) \left(\prod_{l=v_{1}+1}^{r} \bar{A}_{jjl}^{\gamma_{v;m}} \right) \right. \\ \times \left(\prod_{m=1}^{v_{1}} \bar{A}_{ji_{m}}^{\gamma_{v;m} + \lambda_{r;m}} \right) \left(\prod_{m=v_{1}+1}^{v} \bar{A}_{ji_{m}}^{\gamma_{v;m}} \right) \left(\prod_{m=r_{1}+1}^{v} \bar{A}_{ii'_{m}}^{\gamma_{v;m}} \right) \right] \\ = \Theta\left(p_{n}^{1+r_{1}+(r-r_{1})+(r-v_{1})+v_{1}+(v-v_{1})+(v-r_{1})} \right).$$
(107)

Then the sum over indices $\{i, j\} = \{i', j'\}$ with i = j' and j = i' in (95) is bounded by

$$O\left(n(np_n)^{1+2(r+v)-(r_1+v_1)}f^{(s,t)}(np_n,np_n)^2\right) = o(\sigma_n^2).$$

Suppose $\{i, j\} = \{i', j'_{l_1}\}$ for some l_1 . If $i = j'_{l_1}$ and j = i', then $j' = i_{m_1}$ for some m_1 , otherwise $E_1 = 0$. There are at most $n^{3+r+(r-1)+v+(v-1)}$

possible nodes. In this case,

$$E_{1} = \mathbb{E}\left[(A_{jj_{l_{1}}} - p_{n}w_{jj_{l_{1}}})^{1+\lambda_{r;l_{1}}} \left(\prod_{l=1}^{r} \bar{A}_{ij_{l}}^{\lambda_{r;l}} \right) \left(\prod_{l=1, l \neq l_{1}}^{r} \bar{A}_{jj_{l}'}^{\lambda_{r;l}} \right) \times (A_{ji_{m_{1}}} - p_{n}w_{ji_{m_{1}}})^{1+\gamma_{v;m_{1}}} \left(\prod_{m=1, m \neq m_{1}}^{v} \bar{A}_{ji_{m}}^{\gamma_{v;m}} \right) \left(\prod_{m=1}^{v} \bar{A}_{i_{m_{1}}i_{m}'}^{\gamma_{v;m}} \right) \right]$$
$$= \Theta\left(p_{n}^{2+r+(r-1)+v+(v-1)} \right).$$
(108)

Then the sum over indices $\{i, j\} = \{i', j'_{l_1}\}$ in (95) is bounded by

$$O\left(n(np_n)^{2(r+v)}f^{(s,t)}(np_n,np_n)^2\right) = o(\sigma_n^2).$$
 (109)

Similarly, (109) holds for i = i' and $j = j'_{l_1}$ or $\{i, j\} = \{j', i'_{m_1}\}$ for some m_1 .

4.2.4 Bound the last term of (26)

Now we prove the last term of (26) converges in probability to zero. To this end, we will show that

$$\mathbb{E}\left[\left|\sum_{i\neq j} R_{ij} A_{ij}\right|\right] = o\left(\sigma_n\right).$$
(110)

Let s, t satisfy $s + t = k_0$. By (C3) of Assumption 1, $|f^{(s,t)}(x,y)|$ is monotone in x and y. There are four cases: $|f^{(s,t)}(x,y)|$ is decreasing in xand y, $|f^{(s,t)}(x,y)|$ is increasing in x and y, $|f^{(s,t)}(x,y)|$ is increasing in xand decreasing in y, $|f^{(s,t)}(x,y)|$ is decreasing in x and increasing in y.

Suppose $|f^{(s,t)}(x,y)|$ is decreasing in x and y. Let $\delta_n = [\log(np_n)]^{-2}$. Then

$$\mathbb{E}[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t}]$$

$$= \mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t}$$

$$\times I[X_{i(j)} \ge \delta_{n}w_{i(j)}, X_{j(i)} \ge \delta_{n}w_{j(i)}]\right]$$

$$+\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\ \times I[X_{i(j)} \ge \delta_{n}w_{i(j)}, X_{j(i)} < \delta_{n}w_{j(i)}]\right] \\ +\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\ \times I[X_{i(j)} < \delta_{n}w_{i(j)}, X_{j(i)} \ge \delta_{n}w_{j(i)}]\right] \\ +\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\ \times I[X_{i(j)} < \delta_{n}w_{i(j)}, X_{j(i)} < \delta_{n}w_{j(i)}]\right].$$
(111)

By the Cauchy–Schwarz inequality and (52), we have

$$\begin{split} \mathbb{E}\left[|d_{i(j)} - w_{i(j)}|^{s}\right] &\leq \sqrt{\mathbb{E}\left[(d_{i(j)} - w_{i(j)})^{2s}\right]} \\ &= \sqrt{\sum_{r=1}^{2s} \sum_{\substack{j_{1}, j_{2}, \dots, j_{r} \notin \{i, j\} \\ j_{1} \neq j_{2} \neq \dots \neq j_{r}}} \prod_{l=1}^{r} \mathbb{E}\left[(A_{ij_{l}} - p_{n}w_{ij_{l}})^{\lambda_{r;l}}\right] \\ &= O\left(\sqrt{\sum_{r=1}^{s} (np_{n})^{r}}\right) = O\left((np_{n})^{\frac{s}{2}}\right). \end{split}$$

Then

$$\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\ \times I[X_{i(j)} \ge \delta_{n}w_{i(j)}, X_{j(i)} \ge \delta_{n}w_{j(i)}]\right] \\ \le \mathbb{E}\left[|f^{(s,t)}(\delta_{n}w_{i(j)}, \delta_{n}w_{j(i)})|d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\ \times I[X_{i(j)} \ge \delta_{n}w_{i(j)}, X_{j(i)} \ge \delta_{n}w_{j(i)}]\right]$$

$$\leq |f^{(s,t)}(\delta_{n}w_{i(j)}, \delta_{n}w_{j(i)})|\mathbb{E}\left[|d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t}\right]$$

$$= |f^{(s,t)}(\delta_{n}w_{i(j)}, \delta_{n}w_{j(i)})|\mathbb{E}\left[|d_{i(j)} - w_{i(j)}|^{s}\right]\mathbb{E}\left[|d_{j(i)} - w_{j(i)}|^{t}\right]$$

$$\leq (np_{n})^{\frac{k_{0}}{2}}|f^{(s,t)}(\delta_{n}w_{i(j)}, \delta_{n}w_{j(i)})|.$$

$$(112)$$

On the event $\{X_{i(j)} < \delta_n w_{i(j)}\}$, if $X_{i(j)} < d_{i(j)}$, then $X_{i(j)}$ cannot be between $d_{i(j)}$ and $w_{i(j)}$. Hence $X_{i(j)} < \delta_n w_{i(j)}$ implies $d_{i(j)} \leq X_{i(j)} < \delta_n w_{i(j)}$. Similar result holds for $X_{j(i)}$. By definition, $d_{i(j)}$ and $d_{j(i)}$ are independent if $i \neq j$. By Lemma 1 and (C2) of Assumption 1, one has

$$\mathbb{E}\Big[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[X_{i(j)} < \delta_{n}w_{i(j)}, X_{j(i)} < \delta_{n}w_{j(i)}]\Big] \\
\leq \mathbb{E}\Big[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[d_{i(j)} \leq X_{i(j)} < \delta_{n}w_{i(j)}, d_{j(i)} \leq X_{j(i)} < \delta_{n}w_{j(i)}]\Big] \\
\leq \mathbb{E}\Big[|f^{(s,t)}(d_{i(j)}, d_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[d_{i(j)} < \delta_{n}w_{i(j)}, d_{j(i)} < \delta_{n}w_{j(i)}]\Big] \\
= \sum_{k=1}^{\delta_{n}w_{i(j)}} \sum_{l=1}^{\delta_{n}w_{j(i)}} |f^{(s,t)}(k,l)||k - w_{i(j)}|^{s}|l - w_{j(i)}|^{t} \mathbb{P}(d_{i(j)} = k)\mathbb{P}(d_{j(i)} = l) \\
= O\left(((\delta_{n}np_{n})^{M}\exp(-2np_{n}\beta(1+o(1)))\right) \\
= \exp(-2np_{n}\beta(1+o(1)), \qquad (113)$$

where M is a positive constant.

Similarly, the second term of (111) is bounded as follows.

$$\mathbb{E}\Big[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[X_{i(j)} \ge \delta_{n}w_{i(j)}, X_{j(i)} < \delta_{n}w_{j(i)}]\Big] \\
\le \mathbb{E}\Big[|f^{(s,t)}(\delta_{n}w_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[X_{i(j)} \ge \delta_{n}w_{i(j)}, d_{j(i)} \le X_{j(i)} < \delta_{n}w_{j(i)}]\Big] \\
\le \mathbb{E}\Big[|f^{(s,t)}(\delta_{n}w_{i(j)}, d_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[d_{j(i)} < \delta_{n}w_{j(i)}]\Big]$$

$$= \sum_{k=1}^{\delta_n w_{j(i)}} |f^{(s,t)}(\delta_n w_{i(j)}, k)| (k - w_{j(i)})^t \mathbb{P}(d_{j(i)} = k) \mathbb{E}[|d_{i(j)} - w_{i(j)}|^s]$$

$$= O\left((\delta_n n p_n)^M \exp(-n p_n \beta (1 + o(1))) \right)$$

$$= \exp(-n p_n \beta (1 + o(1)).$$
(114)

The third term of (111) can be similarly bounded.

Combining (111)-(114) yields

$$\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t}\right]$$

= $O\left((np_{n})^{\frac{k_{0}}{2}}|f^{(s,t)}(\delta_{n}w_{i(j)}, \delta_{n}w_{j(i)})| + \exp(-np_{n}\beta(1+o(1)))\right).$

By (C4) of Assumption 1, we have

$$\mathbb{E}\left[\left|\sum_{i\neq j}R_{ij}A_{ij}\right|\right] = O\left(n(np_n)^{\frac{k_0}{2}+1}|f^{(s,t)}(\delta_n np_n,\delta_n np_n)|\right) = o\left(\sigma_n\right).$$

Then (110) holds.

Suppose $|f^{(s,t)}(x,y)|$ is increasing in x and y. Let M be a large positive constant. Then

$$\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t}\right]$$

$$= \mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \times I[X_{i(j)} \ge Mw_{i(j)}, X_{j(i)} \ge Mw_{j(i)}]\right]$$

$$+\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \times I[X_{i(j)} \ge Mw_{i(j)}, X_{j(i)} < Mw_{j(i)}]\right]$$

$$+\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t}\right]$$

$$\times I[X_{i(j)} < Mw_{i(j)}, X_{j(i)} \ge Mw_{j(i)}] \\ + \mathbb{E} \bigg[|f^{(s,t)}(X_{i(j)}, X_{j(i)})| |d_{i(j)} - w_{i(j)}|^s |d_{j(i)} - w_{j(i)}|^t \\ \times I[X_{i(j)} < Mw_{i(j)}, X_{j(i)} < Mw_{j(i)}] \bigg].$$
(115)

On the event $\{X_{i(j)} \geq Mw_{i(j)}\}$, if $X_{i(j)} > d_{i(j)}$, then $X_{i(j)}$ cannot be between $d_{i(j)}$ and $w_{i(j)}$. Hence $X_{i(j)} \geq Mw_{i(j)}$ implies $Mw_{i(j)} \leq X_{i(j)} \leq d_{i(j)}$. Similar result holds for $X_{j(i)}$. By definition, $d_{i(j)}$ and $d_{j(i)}$ are independent if $i \neq j$. Suppose $np_n = \omega(\log n)$. By Lemma 1 and (C2) of Assumption 1, one has

$$\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[X_{i(j)} \ge Mw_{i(j)}, X_{j(i)} \ge Mw_{j(i)}]\right] \\
\le \mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[d_{i(j)} \ge X_{i(j)} \ge Mw_{i(j)}, d_{j(i)} \ge X_{j(i)} \ge Mw_{j(i)}]\right] \\
\le \mathbb{E}\left[|f^{(s,t)}(d_{i(j)}, d_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[d_{i(j)} - 1 \ge Mw_{i(j)} - 1, d_{j(i)} - 1 \ge Mw_{j(i)} - 1]\right] \\
= \sum_{k=Mw_{i(j)}-1}^{n-2} \sum_{l=Mw_{j(i)}-1}^{n-2} |f^{(s,t)}(k+1,l+1)||k+1 - w_{i(j)}|^{s}|l+1 - w_{j(i)}|^{t} \\
\times \mathbb{P}(d_{i(j)} - 1 = k)\mathbb{P}(d_{i(j)} - 1 = l) \\
= O\left(n^{C_{s,t,f}}e^{-np_{n}\beta(1+o(1))}\right) \\
= O\left(e^{-np_{n}\beta(1+o(1))}\right), \quad (116)$$

where $C_{s,t,f}$ is some constant dependent on s, t and f. Similarly,

$$\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[X_{i(j)} \ge Mw_{i(j)}, X_{j(i)} < Mw_{j(i)}]\right] \\
\le \mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[d_{i(j)} \ge X_{i(j)} \ge Mw_{i(j)}, X_{j(i)} < Mw_{j(i)}]\right] \\
\le \mathbb{E}\left[|f^{(s,t)}(d_{i(j)}, Mw_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[d_{i(j)} \ge Mw_{i(j)}]\right] \\
= \mathbb{E}|d_{j(i)} - w_{j(i)}|^{t} \sum_{k=Mw_{i(j)}-1}^{n-2} |f^{(s,t)}(k+1, Mw_{j(i)})||k+1 - w_{i(j)}|^{s} \\
\times \mathbb{P}(d_{j(i)} - 1 = k) \\
= O\left(e^{-np_{n}\beta(1+o(1))}\right), \qquad (117)$$

and

$$\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[X_{i(j)} < Mw_{i(j)}, X_{j(i)} < Mw_{j(i)}]\right] \\
\leq \mathbb{E}\left[|f^{(s,t)}(Mw_{i(j)}, Mw_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t}\right] \\
= \mathbb{E}\left[|f^{(s,t)}(Mw_{i(j)}, Mw_{j(i)})|\right] \mathbb{E}\left[|d_{i(j)} - w_{i(j)}|^{s}\right] \mathbb{E}\left[|d_{j(i)} - w_{j(i)}|^{t}\right] \\
= O\left((np_{n})^{\frac{k_{0}}{2}}|f^{(s,t)}(Mnp_{n}, Mnp_{n})|\right), \quad (118)$$

By (115), (116),(117) and (118), it follows that

$$\mathbb{E}\left[\left|\sum_{i\neq j}R_{ij}A_{ij}\right|\right] = O\left(n(np_n)^{\frac{k_0}{2}+1}|f^{(s,t)}(Mnp_n,Mnp_n)|\right) = o\left(\sigma_n\right).$$

Then (110) holds.

Suppose $|f^{(s,t)}(x,y)|$ is decreasing in x and increasing in y. Let $\delta_n = [\log(np_n)]^{-2}$ and M be a large positive constant. Then

$$\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t}\right] \\
= \mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[X_{i(j)} \ge \delta_{n}w_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[X_{i(j)} \ge \delta_{n}w_{i(j)}, X_{j(i)} < Mw_{j(i)}]\right] \\
+ \mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[X_{i(j)} < \delta_{n}w_{i(j)}, X_{j(i)} \ge Mw_{j(i)}]\right] \\
+ \mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[X_{i(j)} < \delta_{n}w_{i(j)}, X_{j(i)} < Mw_{j(i)}]\right] .$$
(119)

The first term of (119) can be bounded by

$$\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\ \times I[X_{i(j)} \ge \delta_{n}w_{i(j)}, X_{j(i)} \ge Mw_{j(i)}]\right] \\ \le \mathbb{E}\left[|f^{(s,t)}(\delta_{n}w_{i(j)}, d_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t}\right]$$

$$\times I[X_{i(j)} \ge \delta_n w_{i(j)}, d_{j(i)} \ge M w_{j(i)}]$$

$$= (np_n)^{\frac{s}{2}} \sum_{k=Mw_{j(i)}}^{n-2} |f^{(s,t)}(\delta_n w_{i(j)}, k)| |k - w_{j(i)}|^t \mathbb{P}(d_{j(i)} = k)$$

$$= \exp(-np_n\beta(1+o(1)), \qquad (120)$$

The second term of (119) can be bounded by

$$\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \times I[X_{i(j)} \ge \delta_{n}w_{i(j)}, X_{j(i)} < Mw_{j(i)}]\right]$$

$$\leq |f^{(s,t)}(\delta_{n}w_{i(j)}, Mw_{j(i)})|(np_{n})^{\frac{k_{0}}{2}}.$$
(121)

The third term of (119) can be bounded by

$$\mathbb{E}\left[|f^{(s,t)}(X_{i(j)}, X_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[X_{i(j)} < \delta_{n}w_{i(j)}, X_{j(i)} \ge Mw_{j(i)}]\right] \\
\leq \mathbb{E}\left[|f^{(s,t)}(d_{i(j)}, d_{j(i)})||d_{i(j)} - w_{i(j)}|^{s}|d_{j(i)} - w_{j(i)}|^{t} \\
\times I[d_{i(j)} < \delta_{n}w_{i(j)}, d_{j(i)} \ge Mw_{j(i)}]\right] \\
= \sum_{k=0}^{\delta_{n}w_{i(j)}} \sum_{l=Mw_{j(i)}}^{n-2} |f^{(s,t)}(k,l)||k - w_{i(j)}|^{s}|l - w_{j(i)}|^{t} \\
\times \mathbb{P}(d_{i(j)} = k)\mathbb{P}(d_{j(i)} = l) \\
= \exp(-np_{n}\beta(1 + o(1)). \quad (122)$$

By (119)-(122), (110) holds.

The case that $|f^{(s,t)}(x,y)|$ is increasing in x and decreasing in y can be similarly processed. We omit it. Then the proof is complete.

4.3 Proof of Theorem 2 in Subsection 3.1

To prove Theorem 2 in Subsection 3.1, we only need to derive the asymptotic distribution of the Randić index of the Erdős-Rényi random graph $\mathcal{G}_n(\alpha)$, that is, $\tau = -\frac{1}{2}$. Recall that for the Erdős-Rényi random graph $\mathcal{G}_n(\alpha)$, $w_{i(k)} = 1 + (n-2)p_n$. Then

$$\begin{aligned} f_x(w_{i(j)}, w_{j(i)}) &= f_y(w_{i(j)}, w_{j(i)}) = -\frac{1}{2(1 + (n-2)p_n))^2}, \\ f_{xx}(w_{i(j)}, w_{j(i)}) &= f_{yy}(w_{i(j)}, w_{j(i)}) = \frac{3}{4(1 + (n-2)p_n))^3}, \\ f_{xy}(w_{i(j)}, w_{j(i)}) &= \frac{1}{4(1 + (n-2)p_n))^3}, \\ &|f^{(s,t)}(np_n, np_n)| = O\left(\frac{1}{(np_n)^{1 + (s+t)}}\right). \end{aligned}$$

Let $k_0 = \max\left\{\lfloor 2 + \frac{1}{1-\alpha} \rfloor + 1, 3\right\}$. By (25) and the proof of Theorem 1, we have

$$\mathcal{I}_n - \mathbb{E}[\mathcal{I}_n] = \frac{1}{2} \sum_{i \neq j} (M_{ij} A_{ij} - \mathbb{E}[M_{ij} A_{ij}]) + \frac{1}{2} \sum_{i \neq j} (S_{ij} A_{ij} - \mathbb{E}[S_{ij} A_{ij}]) + O_P \left(\sqrt{\frac{n}{(np_n)^3}}\right).$$

By the calculations in Section 3.1, $a_{ij} = \Theta\left(\frac{1}{(np_n)^2}\right)$ for $\tau = -\frac{1}{2}$. Then

$$Var\left(\sum_{i\neq j}a_{ij}(A_{ij}-p_nw_{ij})\right)=O\left(\frac{n}{(np_n)^3}\right).$$

By equations (27)-(32), we have

$$\frac{1}{2} \sum_{i \neq j} (M_{ij} A_{ij} - \mathbb{E}[M_{ij} A_{ij}])$$

= $-\frac{\sum_{i \neq j \neq l} (A_{il} - p_n) (A_{ij} - p_n)}{2(1 + (n-2)p_n)^2} + O_P\left(\sqrt{\frac{n}{(np_n)^3}}\right)$

By equations (33)-(47), we have

$$\frac{1}{2} \sum_{i \neq j} (S_{ij}A_{ij} - \mathbb{E}[S_{ij}A_{ij}]) \\
= \frac{1}{4} \sum_{\substack{i \neq j, s \neq t \\ s, t \notin \{i,j\}}} p_n \left(f_{xx}(w_{i(j)}, w_{j(i)}) + f_{yy}(w_{i(j)}, w_{j(i)}) \right) \\
\times (A_{is} - p_n)(A_{it} - p_n) + O_P \left(\sqrt{\frac{n}{(np_n)^3}} \right) \\
= \frac{3(n-3)p_n}{8(1+(n-2)p_n))^3} \sum_{s \neq t \neq i} (A_{is} - p_n)(A_{it} - p_n) + O_P \left(\sqrt{\frac{n}{(np_n)^3}} \right).$$

Hence, we get

$$\mathcal{I}_n - \mathbb{E}[\mathcal{I}_n] = \mathcal{X}_n + O_P\left(\sqrt{\frac{n}{(np_n)^3}}\right), \qquad (123)$$

where

$$\mathcal{X}_n = -\frac{1}{8(1+(n-2)p_n))^2} \sum_{s \neq t \neq i} (A_{is} - p_n)(A_{it} - p_n).$$

Note that

$$\begin{aligned} -\mathcal{X}_n &= \sum_{i < j < k} \frac{(A_{ij} - p_n)(A_{ik} - p_n)}{4(1 + (n-2)p_n))^2} + \sum_{i < j < k} \frac{(A_{ji} - p_n)(A_{jk} - p_n)}{4(1 + (n-2)p_n))^2} \\ &+ \sum_{i < j < k} \frac{(A_{ki} - p_n)(A_{kj} - p_n)}{4(1 + (n-2)p_n))^2}, \end{aligned}$$

and the variance σ_n^2 of \mathcal{X}_n is equal to

$$\sigma_n^2 = \frac{n(n-1)(n-2)p_n^2(1-p_n)^2}{32(1+(n-2)p_n))^4} = \Theta\left(\frac{n}{(np_n)^2}\right).$$

By Theorem 6.1 in [13], we have

$$\frac{\mathcal{X}_n}{\sigma_n} \Rightarrow \mathcal{N}(0,1),$$

from which and (123) it follows that

$$\frac{\mathcal{I}_n - \mathbb{E}[\mathcal{I}_n]}{\sigma_n} \Rightarrow \mathcal{N}(0, 1).$$

Then the proof is complete.

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