# THE VOLUME OF THE GIANT COMPONENT OF A RANDOM GRAPH WITH GIVEN EXPECTED DEGREES* 

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#### Abstract

We consider the random graph model $G(\mathbf{w})$ for a given expected degree sequence $\mathbf{w}=\left(w_{1}, w_{2}, \ldots, w_{n}\right)$. If the expected average degree is strictly greater than 1 , then almost surely the giant component in $G$ of $G(\mathbf{w})$ has volume (i.e., sum of weights of vertices in the giant component) equal to $\lambda_{0} \operatorname{Vol}(G)+O\left(\sqrt{n} \log ^{3.5} n\right)$, where $\lambda_{0}$ is the unique nonzero root of the equation


$$
\sum_{i=1}^{n} w_{i} e^{-w_{i} \lambda}=(1-\lambda) \sum_{i=1}^{n} w_{i},
$$

and where $\operatorname{Vol}(G)=\sum_{i} w_{i}$.
Key words. random graphs, expected degree sequences, giant connected component

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1. Introduction. Among the many celebrated results of Erdős and Rényi on random graphs, one of the most well-known theorems is a sharp estimate for the size of the giant component. For the random graph $G(n, p)$, as introduced by Erdős and Rényi in 1959 [17], every pair of a set of $n$ vertices is chosen to be an edge with probability $p$ independently. Erdős and Rényi [17] showed that the size (i.e., the number of vertices) of the giant component of $G(n, p)$ satisfies the following.

Theorem A. If $d=n p>1$, a graph $G$ of $G(n, p)$ almost surely contains a giant component with $(f(d)+o(1)) n$ vertices, where $f(d)$ is given by

$$
\begin{equation*}
f(d)=1-\frac{1}{d} \sum_{k=1}^{\infty} \frac{k^{k-1}}{k!}\left(d e^{-d}\right)^{k} \tag{1}
\end{equation*}
$$

In $G(n, p)$, every vertex has the same expected degree $n p$. Although such a random graph model is useful in some applications, most real-world networks have degree distributions far from regular $[1,4,5,6,20,21,22,25,26]$. It is therefore not surprising that the random graph model $G(n, p)$ does not capture many behaviors of numerous networks $[1,2,3,9,10,11,12,13,14,15,23]$.

Here we consider the random graph model $G(\mathbf{w})$ for a given expected degree sequence $\mathbf{w}=\left(w_{1}, w_{2}, \ldots, w_{n}\right)$, as introduced in $[10,11,12,13]$. The edges are chosen independently and randomly as follows. The probability $p_{i j}$ that there is an edge between $v_{i}$ and $v_{j}$ is proportional to the product $w_{i} w_{j}$ (as well as the loop at $v_{i}$ with probability proportional to $w_{i}^{2}$ ). Namely,

$$
\begin{equation*}
p_{i j}=\frac{w_{i} w_{j}}{\sum_{k} w_{k}}=\frac{w_{i} w_{j}}{\operatorname{Vol}(G)} \tag{2}
\end{equation*}
$$

[^0]Here the expected volume for a subset $S$ of vertices, $\operatorname{Vol}(S)$, is defined as

$$
\operatorname{Vol}(S)=\sum_{v_{i} \in S} w_{i}
$$

and $\operatorname{Vol}(G)=\operatorname{Vol}(V(G))$. The (actual) volume of $S$ in a graph $G$ is the sum of all degrees of vertices in $S$ and is denoted by $\operatorname{vol}(S)$ :

$$
\operatorname{vol}(S)=\sum_{v_{i} \in S} d_{i}
$$

where $d_{i}$ denotes the degree of vertex $v_{i}$. In order to avoid confusion when we deal with the graph $G$ in a nonprobabilistic context, we can view $w_{i}$ as a weight assigned to vertex $v_{i}$.

In [10], the following theorem was given concerning the giant components for graphs in the random graph model $G(\mathbf{w})$.

Theorem B. Suppose that $G$ is a random graph in $G(\mathbf{w})$ with expected degree sequence $\mathbf{w}$. If the expected average degree $d$ is strictly greater than 1 , then the following hold:
(1) Almost surely $G$ has a unique giant component. Furthermore, the volume of the giant component is at least $\left(1-\frac{2}{\sqrt{d e}}+o(1)\right) \operatorname{Vol}(G)$ if $d \geq \frac{4}{e}=1.4715 \ldots$, and is at least $\left(1-\frac{1+\log d}{d}+o(1)\right) \operatorname{Vol}(G)$ if $d<2$.
(2) The second-largest component almost surely has size at most $(1+o(1)) \mu(d) \log n$, where

$$
\mu(d)= \begin{cases}\frac{1}{1+\log d-\log 4} & \text { if } d>\frac{4}{e} \\ \frac{1}{d-1-\log d} & \text { if } 1<d<2\end{cases}
$$

Moreover, with probability at least $1-n^{-k}$, the second-largest component has size at most $(k+1+o(1)) \mu(d) \log n$ for any $k \geq 1 .{ }^{1}$

In this paper, we will state a sharp asymptotic estimate for the volume of the giant component for a random graph in $G(\mathbf{w})$.

ThEOREM 1. If the expected average degree is strictly greater than 1, then almost surely the giant component in a graph $G$ in $G(\mathbf{w})$ has volume $\lambda_{0} \operatorname{Vol}(G)+$ $O\left(\sqrt{n} \log ^{3.5} n\right)$, where $\lambda_{0}$ is the unique nonzero root of the following equation:

$$
\begin{equation*}
\sum_{i=1}^{n} w_{i} e^{-w_{i} \lambda}=(1-\lambda) \sum_{i=1}^{n} w_{i} \tag{3}
\end{equation*}
$$

We remark that $\operatorname{Vol}(G)$ in the statement of Theorem 1 can be replaced by $\operatorname{vol}(G)$, since it was proved in [10] that with probability at least $1-e^{-c}$,

$$
|\operatorname{Vol}(G)-\operatorname{Vol}(G)| \leq \sqrt{c \operatorname{Vol}(G)}
$$

Since the average degree is $\operatorname{vol}(G) / n$, the average degree can also be approximated by the average expected degree $\operatorname{Vol}(G) / n$.

The paper is organized as follows. Section 2 contains several facts concerning (3). In section 3, we show that the asymptotic formula for the volume of the giant components of a random graph in $G(\mathbf{w})$ is a generalization of Theorem A by Erdős

[^1]

FIG. 1. When $\tilde{d}>1, f(x)$ has a unique positive root.


Fig. 2. When $\tilde{d}<1, f(x)>0$ for all $x>0$.
and Rényi. Section 4 includes some improved lower bounds for the volume of the giant component of $G \in G(\mathbf{w})$ as a function of the expected average degree. In section 5 , we give the complete proof of Theorem 1. In section 6 , we derive a sharp estimate for the number of vertices in the giant components.
2. Preliminaries. Before we proceed, we examine some basic properties of the solutions to the equation in (3). The proof is quite straightforward and will be omitted here.

Let $\tilde{d}$ denote the expected second order average degree:

$$
\tilde{d}=\frac{\sum_{i} w_{i}^{2}}{\sum_{i} w_{i}}
$$

Lemma 1. Suppose that the expected second order average degree satisfies $\tilde{d}>1$. Define

$$
f(\lambda)=\sum_{i=1}^{n} w_{i} e^{-w_{i} \lambda}-(1-\lambda) \sum_{i=1}^{n} w_{i}
$$

We have $f(0)=0, f^{\prime}(0)<0$, and $f^{\prime \prime}(\lambda)>0$. Hence $f(\lambda)=0$ has a unique positive solution $\lambda_{0}$ (see Figure 1). In particular,

1. if $f\left(x_{1}\right) \leq 0$ for some positive $x_{1}$, then $\lambda_{0} \geq x_{1}$;
2. if $f\left(x_{2}\right) \geq 0$ for some positive $x_{2}$, then $\lambda_{0} \leq x_{2}$;
3. $\lambda_{0}<1$ since $f(1)>0$.

When $\tilde{d}<1$, we have $f(0)=0, f^{\prime}(0)>0$, and $f^{\prime \prime}(\lambda)>0$. Zero is the only nonnegative root for $f(x)$ (see Figure 2). This corresponds to the case in which there is no giant component.

The following fact is useful in the proof of the main theorem.
Lemma 2. Suppose that the expected average degree d satisfies

$$
d=\frac{1}{n} \sum_{i=1}^{n} w_{i} \geq 1+\delta>1
$$

for some positive constant $\delta$. Define $f(\lambda)=\sum_{i=1}^{n} w_{i} e^{-w_{i} \lambda}-(1-\lambda) \sum_{i=1}^{n} w_{i}$, and let $\lambda_{0}$ denote the unique nonzero root of $f(\lambda)=0$. Then there is a positive constant $c=c(\delta)$ such that

$$
f^{\prime}\left(\lambda_{0}\right) \geq c \sum_{i=1}^{n} w_{i}
$$

Proof. Since $\tilde{d} \geq d>1$, the unique root $\lambda_{0}$ of $f$ exists. We have

$$
f^{\prime}\left(\lambda_{0}\right)=\sum_{i=1}^{n} w_{i}-\sum_{i=1}^{n} w_{i}^{2} e^{-w_{i} \lambda_{0}}
$$

Case 1. $\lambda_{0} \geq \frac{1}{2}$. Since $x e^{-x \lambda_{0}}$ attains its maximum at $x=1 / \lambda_{0}$, we have

$$
\begin{aligned}
f^{\prime}\left(\lambda_{0}\right) & =\sum_{i=1}^{n} w_{i}-\sum_{i=1}^{n} w_{i}^{2} e^{-w_{i} \lambda_{0}} \\
& \geq \sum_{i=1}^{n} w_{i}-\sum_{i=1}^{n} w_{i} \frac{1}{e \lambda_{0}} \\
& =\left(1-\frac{1}{e \lambda_{0}}\right) \sum_{i=1}^{n} w_{i} \\
& \geq\left(1-\frac{2}{e}\right) \sum_{i=1}^{n} w_{i}
\end{aligned}
$$

The statement holds for this case.
Case 2. $\lambda_{0}<\frac{1}{2}$. We will utilize some convexity inequalities. First we will prove the following claim.

Now we consider the function $h(x)=\left(x^{2}+\frac{x}{\lambda_{0}}\right) e^{-\lambda_{0} x}$. We have

$$
\begin{align*}
h^{\prime}(x) & =\left(\frac{1}{\lambda_{0}}+x-\lambda_{0} x^{2}\right) e^{-\lambda_{0} x} \\
h^{\prime \prime}(x) & =-\lambda_{0} x\left(3-\lambda_{0} x\right) e^{-\lambda_{0} x} \tag{4}
\end{align*}
$$

We need the following facts, whose proofs will be given at the end of this section.
Claim A.
(i) $h(x)$ is concave downward over $x$ in $\left(0, \frac{3}{\lambda_{0}}\right)$. The maximum value of $h(x)$ for $x$ in $[0, \infty)$ is reached at $x_{0}=\frac{\sqrt{5}+1}{2 \lambda_{0}}$.
(ii) $d<\frac{2}{e \lambda_{0}}<x_{0}$.
(iii) $\lambda_{0}>1-\frac{1}{d}$.

Now, we consider the following function:

$$
H(x)= \begin{cases}h(x), & 0 \leq x \leq x_{0} \\ h\left(x_{0}\right), & x \geq x_{0}\end{cases}
$$

Using Claim A(i), $H(x)$ is concave downward and $H(x) \geq h(x)$ for all $x \geq 0$. We have

$$
\begin{aligned}
f^{\prime}\left(\lambda_{0}\right) & =\sum_{i=1}^{n} w_{i}-\sum_{i=1}^{n} w_{i}^{2} e^{-w_{i} \lambda_{0}} \\
& =\sum_{i=1}^{n} w_{i}+\frac{1}{\lambda_{0}} \sum_{i=1}^{n} w_{i} e^{-w_{i} \lambda_{0}}-\sum_{i=1}^{n} h\left(w_{i}\right) \\
& =\sum_{i=1}^{n} w_{i}+\frac{1}{\lambda_{0}}\left(1-\lambda_{0}\right) \sum_{i=1}^{n} w_{i}-\sum_{i=1}^{n} h\left(w_{i}\right) \\
& =\frac{1}{\lambda_{0}} \sum_{i=1}^{n} w_{i}-\sum_{i=1}^{n} h\left(w_{i}\right) \\
& \geq \frac{1}{\lambda_{0}} \sum_{i=1}^{n} w_{i}-\sum_{i=1}^{n} H\left(w_{i}\right) \\
& \geq \frac{1}{\lambda_{0}} \sum_{i=1}^{n} w_{i}-n H\left(\frac{1}{n} \sum_{i=1}^{n} w_{i}\right) \\
& =\frac{1}{\lambda_{0}} n d-n H(d) .
\end{aligned}
$$

By Claim A(ii), we have $d<\frac{2}{e \lambda_{0}}<x_{0}$. Hence, $H(d)=h(d)$.

$$
\begin{aligned}
f^{\prime}\left(\lambda_{0}\right) & \geq \frac{1}{\lambda_{0}} n d-n h(d) \\
& =\frac{1}{\lambda_{0}} n d-n\left(d^{2}+\frac{d}{\lambda_{0}}\right) e^{-\lambda_{0} d} \\
& =n d \frac{1}{\lambda_{0}}\left(1-\left(1+d \lambda_{0}\right) e^{-\lambda_{0} d}\right) \\
& \geq n d\left(1-\left(1+d \lambda_{0}\right) e^{-\lambda_{0} d}\right)
\end{aligned}
$$

The function $\psi(x)=1-(1+x) e^{-x}$ is increasing for $x$ in $[0, \infty)$. For any $x>0$, $\psi(x)>\psi(0)=0$. Hence we have

$$
\begin{aligned}
f^{\prime}\left(\lambda_{0}\right) & \geq n d \psi\left(\lambda_{0} d\right) \\
& \geq n d \psi(d-1) \\
& \geq c n d
\end{aligned}
$$

by choosing $c=c(\delta)=\min \{\psi(\delta), 1-2 / e\}$.
It remains to prove Claim A.
Proof of Claim A. (i) follows from (4).
To prove (ii), we use the facts that $\lambda_{0}$ is a root of $f$ and $x e^{-\lambda_{0} x}$ has its maximum value $\frac{1}{e \lambda_{0}}$ at $x=1 / \lambda_{0}$. Then

$$
\begin{aligned}
\left(1-\lambda_{0}\right) n d & =\left(1-\lambda_{0}\right) \sum_{i=1}^{n} w_{i} \\
& =\sum_{i=1}^{n} w_{i} e^{-\lambda_{0} w_{i}} \\
& \leq \sum_{i=1}^{n} \frac{1}{e \lambda_{0}} \\
& =\frac{n}{e \lambda_{0}}
\end{aligned}
$$

Thus,

$$
\lambda_{0}\left(1-\lambda_{0}\right) \leq \frac{1}{d e}
$$

We have

$$
\lambda_{0} \leq \frac{1}{2}\left(1-\sqrt{1-\frac{4}{d e}}\right) \quad \text { or } \quad \lambda_{0} \geq \frac{1}{2}\left(1+\sqrt{1-\frac{4}{d e}}\right)
$$

Then $\lambda_{0}<\frac{1}{2}$ implies

$$
\begin{aligned}
\lambda_{0} & \leq \frac{1}{2}\left(1-\sqrt{1-\frac{4}{d e}}\right) \\
& =\frac{2}{d e} \frac{1}{1+\sqrt{1-\frac{4}{d e}}} \\
& <\frac{2}{d e} .
\end{aligned}
$$

Hence, we have $d<\frac{2}{e \lambda_{0}}<x_{0}$, as desired.
To prove (iii), we consider the function

$$
g(x)= \begin{cases}x e^{-\lambda_{0} x}, & 0 \leq x \leq \frac{1}{\lambda_{0}} \\ \frac{1}{e \lambda_{0}}, & x>\frac{1}{\lambda_{0}},\end{cases}
$$

We observe that $g(x)$ is concave downward and $g(x) \geq x e^{-\lambda_{0} x}$ for all $x \geq 0$.
By the definition of $\lambda_{0}$, we have

$$
\begin{aligned}
\left(1-\lambda_{0}\right) n d & =\left(1-\lambda_{0}\right) \sum_{i=1}^{n} w_{i} \\
& =\sum_{i=1}^{n} w_{i} e^{-\lambda_{0} w_{i}} \\
& \leq \sum_{i=1}^{n} g\left(w_{i}\right) \\
& \leq n g(d)
\end{aligned}
$$

By Claim A(ii), $d<\frac{2}{e \lambda_{0}}$. Thus, $g(d)=d e^{-\lambda_{0} d}$. We have

$$
1-\lambda_{0} \leq e^{-\lambda_{0} d}
$$

Note that $\phi(\lambda)=(1-\lambda)-e^{-\lambda d}$ is concave downward over $[0, \infty)$. Since $\phi(0)=0$ and $\phi^{\prime}(0)=d-1>0, \phi(x)$ has a unique positive root, which we denote by $s$. We have $\phi(x)>0$ for any $0<x<s$. Since $\phi\left(\lambda_{0}\right) \leq 0$ and $\lambda_{0} \neq 0$, we have $\lambda_{0} \geq s$.

Define $t=(1-s) d$; then we have

$$
\frac{t}{d}=1-s=e^{-s d}=e^{-d+t}
$$

Thus $t$ satisfies the following equation:

$$
\begin{equation*}
t e^{-t}=d e^{-d} \tag{5}
\end{equation*}
$$

The function $x e^{-x}$ increases in $[0,1]$ and decreases in $[1, \infty]$. There is a unique $t<1$ satisfying (5).

We have

$$
\lambda_{0} \geq s=1-\frac{t}{d}>1-\frac{1}{d}
$$

The proof of Claim A is now finished, and therefore the proof of Lemma 2 is complete.
3. Theorem $1 \Rightarrow$ Theorem A. In this section we want to show that the formula for the size of the giant component for a random graph in $G(n, p)$ as derived by Erdős and Rényi in Theorem A is a special case of Theorem 1. In other words, if we restrict the expected degree sequence to the case when all degrees are equal, then we recover the theorem of Erdős and Rényi.

Theorem 2. Theorem 1 implies Theorem A of Erdős and Rényi for $G(n, p)$.
Proof. In $G(n, p)$, we have $w_{1}=w_{2}=\cdots=w_{n}=n p=d$. Equation (3) becomes

$$
e^{-d \lambda}=1-\lambda
$$

Let $\lambda=1-\frac{1}{d} z$. We have

$$
e^{-d+z}=\frac{z}{d},
$$

or equivalently,

$$
z=d e^{-d} e^{z}
$$

Here we use the following version of the well-known Lagrange inversion formula.
LAGRANGE INVERSION FORMULA. Suppose that $z$ is a function of $x$ and $y$ in terms of another analytic function $\phi$ as follows:

$$
z=x+y \phi(z) .
$$

Then $z$ can be written as a power series in $y$ as follows:

$$
z=x+\sum_{k=1}^{\infty} \frac{y^{k}}{k!} D^{(k-1)} \phi^{k}(x),
$$

where $D^{(t)}$ denotes the th derivative.
We apply the above formula with $x=0, y=d e^{-d}$, and $\phi(z)=e^{z}$. Then we have

$$
\begin{aligned}
z & =\left.\sum_{k=1}^{\infty} \frac{y^{k}}{k!} D^{(k-1)} e^{k x}\right|_{x=0} \\
& =\sum_{k=1}^{\infty} \frac{k^{k-1}}{k!} y^{k} \\
& =\sum_{k=1}^{\infty} \frac{k^{k-1}}{k!}\left(d e^{-d}\right)^{k}
\end{aligned}
$$

This is exactly (1) in Theorem A of Erdős and Rényi.
4. Lower bounds. Theorem 1 gives an implicit formula for the volume of the giant component for a random graph with a given expected degree sequence. It is often useful to deduce some bounds which depend only on the expected average degree d. Of particular interest is the following question.

Among all random graphs $G(\mathbf{w})$ with the same expected average degree $d$, which degree distributions minimize or maximize the volume of the giant component?

One obvious example comes to mind. Almost surely $G(m, p)$ with $m p=\Omega(\log m)$ is connected. By adding $n-m$ vertices to $G(m, p)$ with weights zero, we get a random graph $G(\mathbf{w})$ with the expected average degree $d=\frac{m p}{n}$, which almost surely has a giant component with volume $\operatorname{Vol}(G)$.

One might be inclined to conjecture that the random graph with equal expected degrees generates the smallest giant component among all possible degree distributions with the same volume. The answer is "yes" for $1<d \leq \frac{e}{e-1}$, and a surprising "no" if $d$ is sufficiently large.

We will prove the following theorem.
Theorem 3. When $d \geq \frac{4}{e}$, almost surely the giant component of $G \in G(\mathbf{w})$ has volume at least

$$
\left(\frac{1}{2}\left(1+\sqrt{1-\frac{4}{d e}}\right)+o(1)\right) \operatorname{Vol}(G)
$$

We remark that $\frac{1}{2}\left(1+\sqrt{1-\frac{4}{d e}}\right)=1-\frac{1}{d e}+O\left(\frac{1}{d^{2}}\right)$ improves the bound in Theorem B. In fact, this bound is best possible as $d$ approaches infinity, as shown by the following example.

Example. Let $m=\left\lfloor n^{3 / 4}\right\rfloor$ and $y=1+\frac{n}{m}(d-1) \approx(d-1) n^{1 / 4}$. We choose the expected degrees

$$
w_{1}=w_{2}=\cdots=w_{m}=y, \quad w_{m+1}=\cdots=w_{n}=1
$$

The expected average degree of this random graph $G(\mathbf{w})$ is

$$
\frac{m y+(n-m)}{n}=d
$$

Let $x_{0}=1-\frac{1}{d e}$. To show that the giant component of $G$ has volume at most $\left(x_{0}+o(1)\right) \operatorname{Vol}(G)$, it is sufficient to verify $f\left(x_{0}\right) \geq 0$. Here

$$
f(\lambda)=\sum_{i=1}^{n} w_{i} e^{-w_{i} \lambda}-(1-\lambda) \sum_{i=1}^{n} w_{i}
$$

We have

$$
\begin{aligned}
f\left(x_{0}\right) & =\sum_{i=1}^{n} w_{i} e^{-w_{i} x_{0}}-\left(1-x_{0}\right) \sum_{i=1}^{n} w_{i} \\
& =m y e^{-y x_{0}}+(n-m) e^{-x_{0}}-\left(1-x_{0}\right) n d \\
& \geq \frac{n}{e}\left(e^{\frac{1}{d e}}-1-O\left(n^{-1 / 4}\right)\right) \\
& \geq 0
\end{aligned}
$$

as desired.

We are now ready to prove Theorem 3.
Proof of Theorem 3. We note that the function $g(z)=z e^{-z \lambda}$ reaches its maximum value at $z=\frac{1}{\lambda}$. We have

$$
\begin{aligned}
f(\lambda) & =\sum_{i=1}^{n} w_{i} e^{-w_{i} \lambda}-(1-\lambda) \sum_{i=1}^{n} w_{i} \\
& \leq \sum_{i=1}^{n} \frac{1}{\lambda} e^{-1}-(1-\lambda) \sum_{i=1}^{n} w_{i} \\
& =\frac{n}{e \lambda}(1-\lambda(1-\lambda) d e) .
\end{aligned}
$$

Since $\lambda_{0}$ is a solution of $f(\lambda)=0$, we have

$$
\lambda_{0}\left(1-\lambda_{0}\right) \leq \frac{1}{d e},
$$

which implies either $\lambda_{0} \leq \frac{1}{2}\left(1-\sqrt{1-\frac{4}{d e}}\right)$ or $\lambda_{0} \geq \frac{1}{2}\left(1+\sqrt{1-\frac{4}{d e}}\right)$.
We will show that $\lambda_{0} \leq \frac{1}{2}\left(1-\sqrt{1-\frac{4}{d e}}\right)$ is not true by proving $f\left(\frac{1}{2}\right) \leq 0$.
We note that

$$
\begin{aligned}
f\left(\frac{1}{2}\right) & =\sum_{i=1}^{n} w_{i} e^{-w_{i} / 2}-\frac{1}{2} \sum_{i=1}^{n} w_{i} \\
& \leq 2 n e^{-1}-\frac{1}{2} n d \\
& =\frac{n}{2}\left(\frac{4}{e}-d\right) \\
& \leq 0 .
\end{aligned}
$$

Thus we conclude that $\lambda_{0} \geq \frac{1}{2}\left(1+\sqrt{1-\frac{4}{d e}}\right)$.
When $d$ is small and not in the range covered by Theorem 3, we can still derive the following lower bound.

Theorem 4. When $1<d \leq \frac{e}{e-1}$, then almost surely $G(\mathbf{w})$ has a giant component of size at least $\left(\lambda_{1}+o(1)\right) \operatorname{Vol}(G)$, where $\lambda_{1}$ is the nonzero root of the following equation:

$$
\begin{equation*}
e^{-\lambda d}=1-\lambda . \tag{6}
\end{equation*}
$$

In other words, among all random graphs $G(\mathbf{w})$ with fixed expected average degree $d$, the Erdős-Rényi random graph $G\left(n, \frac{d}{n}\right)$ has the smallest giant component (measured in volume).

Proof. Consider the function

$$
g(x)= \begin{cases}x e^{-\lambda_{1} x}, & 0 \leq x \leq \frac{1}{\lambda_{1}}, \\ \frac{1}{e \lambda_{1}}, & x>\frac{1}{\lambda_{1}} .\end{cases}
$$

We observe that $g(x)$ is concave downward and $g(x) \geq x e^{-\lambda_{1} x}$ for all $x \geq 0$. We have

$$
\begin{aligned}
f\left(\lambda_{1}\right) & =\sum_{i=1}^{n} w_{i} e^{-\lambda_{1} w_{i}}-\left(1-\lambda_{1}\right) n d \\
& \leq \sum_{i=1}^{n} g\left(w_{i}\right)-\left(1-\lambda_{1}\right) n d \\
& \leq n g\left(\frac{1}{n} \sum_{i=1}^{n} w_{i}\right)-\left(1-\lambda_{1}\right) n d \\
& \leq n\left(g(d)-\left(1-\lambda_{1}\right) d\right)
\end{aligned}
$$

Since $\lambda_{1}$ is an increasing function of $d, d \lambda_{1}$ is also an increasing function of $d$. When $d=\frac{e}{e-1}$, it is easy to verify that $\lambda=1-\frac{1}{e}$ is the other root of (6). Therefore, $d \lambda_{1} \leq 1$ when $d \leq \frac{e}{e-1}$. In particular, we have

$$
g(d)=d e^{-\lambda_{1} d}
$$

Hence

$$
\begin{aligned}
f\left(\lambda_{1}\right) & \leq n\left(g(d)-\left(1-\lambda_{1}\right) d\right) \\
& =n d\left(e^{-\lambda_{1} d}-\left(1-\lambda_{1}\right)\right) \\
& =0
\end{aligned}
$$

By Remark 1, we have $\lambda_{0} \geq \lambda_{1}$, as desired.
5. The proof of the main theorem. A central tool that we use in the proof of the main theorem is a relaxed version of the Azuma inequality (as seen in Theorem 1 of [12]), which can be described as follows.

Suppose that $\Omega$ is a probability space, and that $\mathcal{F}$ denote a $\sigma$-field on $\Omega$ (i.e., a collection of subsets of $\Omega$, which contains $\emptyset$ and $\Omega$ and is closed under unions, intersections, and complementation). A filter $\mathbf{F}$ is an increasing chain of $\sigma$-subfields

$$
\{0, \Omega\}=\mathcal{F}_{0} \subset \mathcal{F}_{1} \subset \cdots \subset \mathcal{F}_{n}=\mathcal{F}
$$

A martingale (obtained from) $X$ is associated with a filter $\mathbf{F}$ and a sequence of random variables $X_{0}, X_{1}, \ldots, X_{n}$ satisfying $X_{i}=E\left(X \mid \mathcal{F}_{i}\right)$ and, in particular, $X_{0}=E(X)$ and $X_{n}=X$. For undefined terminology on martingales, the reader is referred to [19].

For $\mathbf{c}=\left(c_{1}, c_{2}, \ldots, c_{n}\right)$ a vector with positive entries, a martingale $X$ is said to be $c$-Lipschitz if

$$
\begin{equation*}
\left|X_{i}-X_{i-1}\right| \leq c_{i} \tag{7}
\end{equation*}
$$

for $i=1,2, \ldots, n$.
If the $c$-Lipschitz condition is not satisfied, we can still consider the following relaxed version.

A martingale $X$ is said to be near- $c$-Lipschitz with an exceptional probability $\eta$ if

$$
\begin{equation*}
\sum_{i} \operatorname{Pr}\left(\left|X_{i}-X_{i-1}\right| \geq c_{i}\right) \leq \eta \tag{8}
\end{equation*}
$$

Theorem C (Theorem 1 as in [12]). For nonnegative values, $c_{1}, c_{2}, \ldots, c_{n}, a$ martingale $X$ is near-c-Lipschitz with an exceptional probability $\eta$. Then $X$ satisfies

$$
\operatorname{Pr}(|X-E(X)|<a) \leq 2 e^{-\frac{a^{2}}{2 \sum_{i=1}^{n} c_{i}^{2}}}+\eta
$$

The idea for the proof of Theorem 1 is to first prove that the volume of giant component concentrates on its expected value $\mathrm{E}(\operatorname{Vol}(G C C))$ and then show that $\mathrm{E}(\operatorname{Vol}(G C C)) / \operatorname{Vol}(G)$ can be approximated by the nonzero root of (3). To do so, we need to establish several useful facts.

Lemma 3. With probability at least $1-2 n^{-k}$, a vertex with weight greater than $\max \{8 k, 2(k+1+o(1)) \mu(d)\} \log n$ is in the giant component of $G(\mathbf{w})$.

Proof. Consider a vertex $v_{i}$ with weight $w_{i} \geq \max \{8 k, 2(k+1+o(1)) \mu(d)\} \log n$. For a random graph $G$ in $G(\mathbf{w})$, let $d_{i}$ denote the degree of $v_{i}$ in $G$. Then, $d_{i}$ is the sum of independent 0-1 random variables with $\mathrm{E}\left(d_{i}\right)=w_{i}$. For any nonnegative value $\lambda$, we have

$$
\operatorname{Pr}\left(d_{i}-\mathrm{E}\left(d_{i}\right)<-\lambda\right) \leq e^{-\frac{\lambda^{2}}{2 \mathrm{E}\left(d_{i}\right)}} .
$$

By choosing $\lambda=w_{i} / 2$, we have

$$
\operatorname{Pr}\left(d_{i}<w_{i} / 2\right) \leq e^{-w_{i} / 8} \leq n^{-k}
$$

With probability at least $1-n^{-k}, v_{i}$ is in a connected component of size at least $w_{i} / 2$. If this connected component is not the giant component, then the second largest component must have size at least $w_{i} / 2$. However, from Theorem B , this can happen with probability only at most $n^{-k}$ because of the assumption that

$$
w_{i} / 2 \geq(k+1+o(1)) \mu(d) \log n .
$$

Hence, with probability at least $1-2 n^{-k}$, a vertex with weight greater than $\max \{8 k$, $2(k+1+o(1)) \mu(d)\} \log n$ is in the giant component.

Lemma 4. For any $k>2$, with probability at least $1-6 n^{-k+2}$, we have

$$
|\operatorname{Vol}(G C C)-\mathrm{E}(\operatorname{Vol}(G C C))| \leq 2 C_{1}(k+1)^{2} \sqrt{k-2} \sqrt{n} \log ^{2.5} n
$$

for some positive constant $C_{1}$.
Proof. Let $L=L(k)$ be the set of vertices with weight greater than $\max \{8 k, 2(k+$ $1+o(1)) \mu(d)\} \log n$. If $L \neq \emptyset$, we form a new graph $G^{*}$ by adding a new vertex $v_{*}$ to $G(\mathbf{w})$ and add edges from $v_{*}$ to each vertex in $L . G(\mathbf{w})$ almost surely has a giant component, and so does $G^{*}$. Let $X$ denote the volume of the giant component in $G^{*}$. (While computing the values for Vol of the giant component in $G^{*}$, we use the convention that the weight of $v_{*}$ is zero.) If $L=\emptyset$, we simply let $X=\operatorname{Vol}(G C C)$.

We wish to show the concentration of the random variable $X$. It is sufficient to prove the following claim.

Claim B.

$$
\operatorname{Pr}(|X-\mathrm{E}(X)|<\lambda) \leq 4 n^{-k+2}
$$

where $\lambda=2 C_{1}(k+1)^{2} \sqrt{k-2} \sqrt{n} \log ^{2.5} n$.
We observe that $X$ does not depend on whether $\{u, v\}$ is an edge if both $u$ and $v$ are in $L$. We list all pairs of vertices with at least one vertex not in $L$ by $\left\{f_{1}, f_{2}, \ldots, f_{m}\right\}$, where $m=\binom{n}{2}-\binom{|L|}{2}$. (The order of edges in the list is arbitrarily chosen.) For $i=0,1,2, \ldots, m$, let $\mathcal{F}_{i}$ denote the $\sigma$-field generated by exposing pairs $f_{1}, f_{2}, \ldots, f_{i}$. We apply Theorem C on the edge-exposing martingale $X$ with $X_{i}=$ $\mathrm{E}\left(X \mid \mathcal{F}_{i}\right)$ and $X_{m}=X$. We wish to find a good Lipschitz or near-Lipschitz bound for $\left|X_{i}-X_{i-1}\right|$. By definition, $X_{i-1}$ is the conditional expectation of $X_{i}$. Choosing the
pair $f_{i}$ as an edge can change $X$ by at most the volume of a small component. Let $v_{i}$ be a vertex of the pair $f_{i}$ not in $L$. (If there is a tie, break arbitrarily.) Let $G_{v_{i}}$ be the random graph obtained by deleting $v_{i}$ from $G(\mathbf{w})$. The possible small component containing $v$ before $f_{i}$ is exposed can be broken into at most $d_{i}$ largest connected components excluding the giant component in $G_{v_{i}}$.

First, we apply Theorem B to the random graph $G_{v_{i}}$. Note that the average degree of $G_{v_{i}}$ is $(1+o(1)) d$. Thus, with probability at least $1-n^{-k}$, all small components of $G_{v_{i}}$ have size at most $(k+1+o(1)) \mu(d) \log n$. Similarly, with probability at least $1-n^{-k}$, all small components of $G_{v_{i}}$ have volumes at most $C(k+1) \log n$ for some positive constant depending only on $d$. Also, for any positive $\lambda^{\prime}$, the degree $d_{i}$ of $v_{i}$ can be upper bounded by

$$
\operatorname{Pr}\left(d_{i}>w_{i}+\lambda^{\prime}\right)<e^{-\frac{\lambda^{2}}{2\left(w_{i}+\lambda^{\prime} / 3\right)}} .
$$

By choosing $\lambda^{\prime}=w_{i}+2 k \log n$, we have

$$
\begin{aligned}
\operatorname{Pr}\left(d_{i}>2 w_{i}+2 k \log n\right) & <e^{-\frac{\lambda^{2}}{2\left(w_{i}+\lambda / 3\right)}} \\
& =e^{-\frac{\left(w_{i}+2 k \log n\right)^{2}}{2\left(w_{i}+\left(w_{i}+2 k \log n\right) / 3\right)}} \\
& <n^{-k} .
\end{aligned}
$$

Thus, with probability at least $1-2 n^{-k}$, we have

$$
\begin{aligned}
\left|X_{i}-X_{i-1}\right| & \leq d_{i} \times(k+1) C \log n \\
& <\left(2 w_{i}+2 k \log n\right)(k+1) C \log n \\
& <(10 k \log n+2(k+1+o(1)) \mu(d) \log n)(k+1) C \log n \\
& <C_{1}(k+1)^{2} \log ^{2} n
\end{aligned}
$$

where $C_{1}=C(10+2 \mu(d))$ is a bounded positive number.
Now we apply Theorem C on martingale $X$ with $c_{i}=C_{1}(k+1)^{2} \log ^{2} n$ and $\eta \leq\binom{ n}{2} 2 n^{-k}$. For any positive $\lambda$, we have

$$
\begin{aligned}
\operatorname{Pr}(|X-\mathrm{E}(X)|>\lambda) & \leq 2 e^{-\frac{\lambda^{2}}{2 \sum_{i=1}^{\lambda_{i}^{2}}}+\eta} \\
& \leq 2 e^{-\frac{\lambda^{2}}{2 C_{1}^{2}(k+1)^{4} n \log ^{4} n}}+2 n^{-k+2}
\end{aligned}
$$

For $\lambda=2 C_{1}(k+1)^{2} \sqrt{k-2} \sqrt{n} \log ^{2.5} n$, we have

$$
\operatorname{Pr}(|X-\mathrm{E}(X)|>\lambda) \leq 4 n^{-k+2}
$$

as desired. $\quad$
Proof of Theorem 1. For any vertex $v$ with weight $w_{v}$, the probability that $v$ is not in the giant component of $G(\mathbf{w})$ can be estimated as follows. To simplify the notation, we write $C_{k}=\max \{8 k, 2(k+1+o(1)) \mu(d)\}$.

Case a. $w_{v} \geq C_{k} \log n$. By Lemma 3, we have

$$
\operatorname{Pr}(v \notin G C C) \leq \frac{2}{n^{k}}
$$

Case b. $w_{v} \leq C_{k} \log n$. Let $G_{v}$ be the random graph by removing $v$ from $G$. Expose every pair of vertices in $G_{v}$. Let $H$ be the giant component of $G_{v}$. Applying Lemma 4 to $G_{v}$, with probability at least $1-\frac{6}{(n-1)^{k-2}}$, we have

$$
|\operatorname{Vol}(H)-\mathrm{E}(\operatorname{Vol}(H))| \leq 2 C_{1}(k+1)^{2} \sqrt{k-2} \sqrt{n} \log ^{2.5} n
$$

Now we expose the pairs of vertices containing $v$. We have

$$
\begin{aligned}
\operatorname{Pr}(v \notin G C C \mid H) & =\prod_{v_{j} \in V(H)}\left(1-w_{v} w_{j} \rho\right) \\
& =e^{-\sum_{v_{j} \in V(H)} w_{v} w_{j} \rho+\sum_{v_{j} \in V(H)} w_{v}^{2} w_{j}^{2} \rho^{2}} \\
& =e^{-w_{v} \operatorname{Vol}(H) \rho\left(1+O\left(w_{v} \tilde{d} \rho\right)\right)}
\end{aligned}
$$

The probability that $v$ is not in the giant component can be estimated as follows:

$$
\begin{align*}
\operatorname{Pr}(v \notin G C C) & =\mathrm{E}(\operatorname{Pr}(v \notin G C C \mid H))+O\left(n^{-k+2}\right) \\
& =\mathrm{E}\left(e^{-w_{v} \operatorname{Vol}(H) \rho}\right)+O\left(n^{-k+2}\right) \\
& =e^{-w_{v} \mathrm{E}(\operatorname{Vol}(H)) \rho+O\left(k^{2} w_{i} \rho \sqrt{n} \log ^{2.5} n\right)}+O\left(n^{-k+2}\right) . \tag{9}
\end{align*}
$$

Note that $G C C$ can be formed from $H$ by joining at most $d_{v}$ 's small components. Thus, we have

$$
\begin{aligned}
|\mathrm{E}(G C C)-\mathrm{E}(H)| & \leq \mathrm{E}\left(d_{v}\right)(k+1) C \log n+2 n^{-k} \\
& =w_{v}(k+1) C \log n+2 n^{-k} \\
& =O\left(w_{v} k \log n\right)
\end{aligned}
$$

By substituting $\mathrm{E}(H)$ by $\mathrm{E}(\operatorname{Vol}(G C C))+O\left(w_{v} k \log n\right)$ in (9), we have

$$
\begin{aligned}
\operatorname{Pr}(v \notin G C C) & =e^{-w_{v} \mathrm{E}(\operatorname{Vol}(G C C)) \rho+O\left(w_{v}^{2} k \rho \log n\right)+O\left(k^{2} w_{v} \rho \sqrt{n} \log ^{2.5} n\right)}+O\left(n^{-k+2}\right) \\
& =\left(1+O\left(k^{3} \rho \sqrt{n} \log ^{3.5} n\right)\right) e^{-w_{v} \mathrm{E}(\operatorname{Vol}(G C C)) \rho}+O\left(n^{-k+2}\right) .
\end{aligned}
$$

Putting these together, we have

$$
\begin{aligned}
\operatorname{Vol}(G)- & \mathrm{E}(\operatorname{vol}(G C C)) \\
= & \sum_{v} w_{v} \operatorname{Pr}(v \notin G C C) \\
= & \sum_{w_{v}<C_{k} \log n} w_{v} \operatorname{Pr}(v \notin G C C)+\sum_{w_{v} \geq C_{k} \log n} w_{v} \operatorname{Pr}(v \notin G C C) \\
= & \sum_{w_{v}<C_{k} \log n} w_{v}\left[\left(1+O\left(k^{3} \rho \sqrt{n} \log ^{3.5} n\right)\right) e^{-w_{v} \mathrm{E}(\operatorname{Vol}(G C C)) \rho}+O\left(n^{-k+2}\right)\right] \\
& +\sum_{w_{v} \geq C_{k} \log n} w_{v} O\left(2 n^{-k}\right) \\
= & \sum_{w_{v}<C_{k} \log n} w_{v} e^{-w_{v} \mathrm{E}(\operatorname{Vol}(G C C)) \rho}+O\left(k^{3} \sqrt{n} \log ^{3.5} n\right) .
\end{aligned}
$$

We choose $k$ to be a constant large enough satisfying

$$
C_{k} \geq \begin{cases}\frac{2}{\left(1-\frac{2}{\sqrt{d e}}\right)} & \text { if } d>\frac{4}{e} \\ \frac{2}{\left(1-\frac{1+\log d}{d}\right)} & \text { if } 1<d<2\end{cases}
$$

By Theorem A, we have $C_{k} \mathrm{E}(\operatorname{Vol}(G C C)) \rho \geq 2$. In particular, for any vertex $v$ with $w_{v} \geq C_{k} \log n$, we have

$$
e^{-w_{v} \mathrm{E}(\operatorname{Vol}(G C C)) \rho} \leq n^{-2}
$$

Thus,

$$
\sum_{w_{v} \geq C_{k} \log n} w_{v} e^{-w_{v} \mathrm{E}(\operatorname{Vol}(G C C)) \rho}=O\left(n^{-1}\right)
$$

Therefore we have

$$
\operatorname{Vol}(G)-\mathrm{E}(\operatorname{vol}(G C C))=\sum_{v} w_{v} e^{-w_{v} \mathrm{E}(\operatorname{Vol}(G C C)) \rho}+O\left(\sqrt{n} \log ^{3.5} n\right)
$$

Letting $x_{0}=\frac{\operatorname{Vol}(G C C)}{\operatorname{Vol}(G)}$ and $f(x)=\sum_{i=1}^{n} w_{i} e^{-w_{i} x}-(1-x) \sum_{i=1}^{n} w_{i}$, we have

$$
\begin{equation*}
f\left(x_{0}\right)=O\left(\sqrt{n} \log ^{3.5} n\right) \tag{10}
\end{equation*}
$$

The equation $f(x)=0$ has only two roots, $x=0$ and $x=\lambda_{0}$. Note that $f(x)$ is concave upward with $\left|f^{\prime}(0)\right|=n\left(d^{2}-d\right)$, and $\left|f^{\prime}\left(\lambda_{0}\right)\right|>c n d$. Consider a small interval $I$ around 0 with diameter $O\left(\sqrt{n} \log ^{3.5} n\right)$. The preimage $f^{-1}(I)$ has diameter at most $O\left(n^{-1 / 2} \log ^{3.5} n\right)$. Since $x_{0}$ is bounded away from 0 by a small constant, we have $\left|x_{0}-\lambda_{0}\right|=O\left(n^{-1 / 2} \log ^{3.5} n\right)$. Therefore, almost surely the giant component has volume

$$
\lambda_{0} \operatorname{Vol}(G)+O\left(\sqrt{n} \log ^{3.5} n\right)
$$

Theorem 1 is proved.
6. The complement of the giant component and its size. As we know, the giant component almost surely exists if the expected average degree $d>1$. We consider the remaining graph $G^{\prime}$ after removing the giant component.

For a random graph $G$ in the Erdős-Rényi model $G(n, p)$, where $p=d / n$, if $d>1$, there is a unique $c<1$ satisfying

$$
c e^{-c}=d e^{-d}
$$

We write $\lambda_{0}=1-\frac{c}{d}$. For any vertex $v$, the probability that $v \in S$ is known [19] to be

$$
e^{-\lambda_{0} d}=e^{-d+c}=\frac{c}{d}
$$

Hence $S$ has $\left(\frac{c}{d}+o(1)\right) n$ vertices. After removing the giant component from $G(n, p)$, the remaining graph can be viewed as a random graph in $G\left(n^{\prime}, p\right)$, where $n^{\prime} \approx \frac{c}{d} n$.

The above fact can be generalized to the random graph model $G(\mathbf{w})$. The following theorem is based on the proof of Theorem 1, and we omit the proof here.

Theorem 5. Suppose that the expected average degree d is strictly greater than 1. Let $G^{\prime}$ denote the remaining graph of a random graph $G$ in $G(\mathbf{w})$ by removing the giant component. Then almost surely $G^{\prime}$ is an induced subgraph on a random subset $S$ satisfying the following:

1. Any vertex $v_{i}$ is contained in $S$ with probability $e^{-\lambda_{0} w_{i}}$, where $\lambda_{0}$ is as defined in (3).
2. For any $v_{i}, v_{j} \in S$, the probability that $v_{i} v_{j}$ is an edge of $G_{S}$ is $w_{i} w_{j} / \operatorname{Vol}(G)$. The induced subgraph $G_{S}$ is a random graph with given expected degrees

$$
\left\{\left(1-\lambda_{0}\right) w_{i}\right\}_{v_{i} \in S} .
$$

3. $G^{\prime} \backslash G_{S}$ consists of at most $O(\log n)$ components each with size $O(\log n)$.

We further analyze the size of the giant component. The proof is similar and will be omitted.

THEOREM 6. If the expected average degree is strictly greater than 1, then almost surely the giant component in a random graph of given expected degrees $w_{i}$, $i=1, \ldots, n$, has $n-\sum_{i=1}^{n} e^{-w_{i} \lambda_{0}}+O\left(\sqrt{n} \log ^{4.5} n\right)$ vertices and $\left(\lambda_{0}-\frac{1}{2} \lambda_{0}^{2}\right) \operatorname{Vol}(G)+$ $O\left(\sqrt{\operatorname{Vol}(G)} \log ^{3.5} n\right)$ edges, where $\lambda_{0}$ is as defined in (3).
7. Comparing theoretical results with the data from the collaboration graph. To illustrate the effectiveness of our results, we use an example of the collaboration graph of the second kind. Based on the data of Mathematics Review [18], there are about 401,000 authors as vertices. Two vertices are joined by an edge if there is a paper by exactly two authors. There are about 284,000 edges. The giant component has 176,000 vertices and 248,000 edges. Suppose that we model this collaboration graph as a random graph with some given expected degrees $w_{i}$. Although we do not know the exact values of the $w_{i}$ 's, we can make the following deductions using the theorems in the previous section.

By Theorem 6, we have

$$
\lambda_{0}\left(2-\lambda_{0}\right) \approx \frac{\operatorname{Vol}(G C C)}{\operatorname{Vol}(G)} \approx \frac{248000}{284000}
$$

Solving the above equation, we have $\lambda_{0} \approx 0.644$.
For a fixed vertex $v_{i}$, the degree of $v_{i}$ follows the Poisson distribution with expected value $w_{i}$. Namely, for a fixed $k$, the probability that $v_{i}$ has degree $k$ is $\frac{w_{i}^{k}}{k!} e^{-w_{i}}$. Let $n_{k}$ denote the number of vertices of degree $k$. Then by the linearity of the expectation, we have

$$
\mathrm{E}\left(n_{k}\right) \approx \sum_{i=0} \frac{w_{i}^{k}}{k!} e^{-w_{i}}
$$

Theorem 6 implies that the size of the giant component satisfies

$$
\begin{align*}
|G C C| & \approx n-\sum_{i=1}^{n} e^{-\lambda_{0} w_{i}} \\
& =n-\sum_{i=1}^{n} e^{\left(1-\lambda_{0}\right) w_{i}} e^{-w_{i}} \\
& =\sum_{k \geq 0} n_{k}-\sum_{i=1}^{n} \sum_{k=0}^{\infty} \frac{\left(1-\lambda_{0}\right)^{k}}{k!} w_{i}^{k} e^{-w_{i}} \\
& \approx \sum_{k \geq 0} n_{k}\left(1-\left(1-\lambda_{0}\right)^{k}\right) \\
& =\sum_{k \geq 1} n_{k}\left(1-\left(1-\lambda_{0}\right)^{k}\right) . \tag{11}
\end{align*}
$$

Here we estimate $n_{k}$ by

$$
n_{k} \approx \mathrm{E}\left(n_{k}\right) \approx \sum_{k=1}^{n} \frac{w_{i}^{k}}{k!} e^{-w_{i}}
$$

Grossman, Ion, and De Castro [18] have computed the $n_{k}$ 's as shown in Table 1.

Table 1
The degree sequence of the collaboration graph of the second kind.

| $n_{0}$ | $n_{1}$ | $n_{2}$ | $n_{3}$ | $n_{4}$ | $n_{5}$ | $n_{6}$ | $n_{7}$ | $n_{8}$ | $n_{9}$ | $\cdots$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 166381 | 145872 | 34227 | 16426 | 9913 | 6670 | 4643 | 3529 | 2611 | 2032 | $\cdots$ |



Fig. 3. Degree distribution of the collaboration graph of the second kind.


Fig. 4. Size distribution of connected components of the collaboration graph of the second kind.

By substituting the above $n_{k}$ 's into (11), the size of the giant component is estimated to be about 177,400 . This is rather close to the actual value 176,000 , within an error bound of less than $1 \%$.

In Figures 3 and 4, we have plotted the degree distribution and the distribution of the sizes of connected components of the collaboration graph of the second kind.

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[^1]:    ${ }^{1}$ The quantitative estimate of this probability is in the proof of Theorem 1 in [10].

