

## Normal Approximation of Number of Isolated Vertices in a Random Graph

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**Abstract :** In this paper, we give bounds in normal approximation of number of isolated vertices in a random graph on  $n$  vertices. The technique we used here is the Stein's method.

**Keywords :** Random graph, isolated vertices, normal approximation, Stein's method.

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### 1 Introduction

A random graph is a graph generated by some random procedure. The study of random graphs has a long history. A systematic study began with the influential work of Erdős and Rényi in 1959-1961 ([2], [3], [4]) and has developed into one of the mainstays of modern discrete mathematics.

Let  $G(n, p)$  be a random graph on  $n$  labeled vertices  $\{1, 2, \dots, n\}$  where each possible edge  $\{i, j\}$  is present randomly and independently with the probability  $p$ ,  $0 < p < 1$ . Let

$$X_i = \begin{cases} 1, & \text{if the } i\text{-th vertex is isolated;} \\ 0, & \text{otherwise,} \end{cases}$$

and

$$W_n = X_1 + X_2 + \dots + X_n.$$

Note that  $W_n$  is the number of isolated vertices in  $G(n, p)$  and the expectation of  $W_n$  is

$$\begin{aligned} \lambda_n &:= EW_n = nq^{n-1}, \\ \sigma_n^2 &:= \text{Var } W_n = nq^{n-1} \left( 1 + nq^{n-2} \left( p - \frac{1}{n} \right) \right) \end{aligned}$$

where  $q = 1 - p$  (see [8], p. 137).

In [7], Teerapabolarn, Neammanee and Chongcharoen showed that  $X_1, X_2, \dots, X_n$  are not independent and the distribution of  $W_n$  can be approximated by Poisson with parameter  $\lambda_n$ . Here is their result.

**Theorem 1.1**

$$|P(W_n = 0) - e^{-\lambda_n}| \leq (\lambda_n + e^{-\lambda_n} - 1) \left( \frac{(n-2)p+1}{n(1-p)} \right).$$

In 1987, Kordecki solved the problem by using normal approximation. His bound is as follows :

**Theorem 1.2** *If  $p = w_n/n, \log n - w_n \rightarrow \infty$  and  $nw_n \rightarrow \infty$ , then there exists a constant  $C \equiv C(z)$  such that*

$$\sup_{z \in \mathbb{R}} \left| P \left( \frac{W_n - \lambda_n}{\sigma_n} \leq z \right) - \Phi(z) \right| \leq \frac{C}{\sigma_n} \quad (1.1)$$

where

$$\Phi(z) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^z e^{-\frac{t^2}{2}} dt$$

is the standard normal distribution function.

Notice that the constant  $C$  in the above theorem is not good enough because it depends on  $z$  which goes to infinity as  $z \rightarrow \infty$  (see [8], Lemma 2.1(3)). In this work, we improve the upper bound in (1.1) to the case of non-uniform and uniform bounds. The followings are our main results.

**Theorem 1.3** *(non-uniform) If  $p = w_n/n, \log n - w_n \rightarrow \infty$  and  $nw_n \rightarrow \infty$ , then there exists a constant  $C$ , independent of  $z$ , such that*

$$\left| P \left( \frac{W_n - \lambda_n}{\sigma_n} \leq z \right) - \Phi(z) \right| \leq \frac{C}{(1+|z|)\sigma_n}. \quad (1.2)$$

**Corollary 1.4** *(uniform) If  $p = w_n/n, \log n - w_n \rightarrow \infty$  and  $nw_n \rightarrow \infty$ , then there exists a constant  $C$ , independent of  $z$ , such that*

$$\sup_{z \in \mathbb{R}} \left| P \left( \frac{W_n - \lambda_n}{\sigma_n} \leq z \right) - \Phi(z) \right| \leq \frac{C}{\sigma_n}.$$

Note that if  $np = c$  where  $c$  is a constant, then the Poisson convergence is impossible since

$$\lim_{n \rightarrow \infty} (\lambda_n + e^{-\lambda_n} - 1) \left( \frac{(n-2)p+1}{n(1-p)} \right) = (1+c)e^{-c}$$

and in this case the rate of convergence to the normal distribution has the order  $\frac{1}{\sqrt{n}}$ .

## 2 Proof of Main Results

In this section, we give the proof of main results by using Stein's method for normal approximation and the idea from Kordecki [8].

Stein's method was given by Stein [5] in 1972. His technique was relied on the elementary differential equation

$$f'(w) - wf(w) = I_{\{w \leq z\}} - \Phi(z) \quad (2.1)$$

where  $I_A$  is defined by

$$I_A(w) = \begin{cases} 1, & \text{if } w \in A; \\ 0, & \text{if } w \notin A. \end{cases}$$

It is well-known that the unique solution  $\varphi_z$  of Stein's equation (2.1) is of the form

$$\varphi_z(x) = \begin{cases} \sqrt{2\pi}e^{\frac{x^2}{2}} \Phi(x)[1 - \Phi(z)], & \text{if } x \leq z; \\ \sqrt{2\pi}e^{\frac{x^2}{2}} \Phi(z)[1 - \Phi(x)], & \text{if } x \geq z. \end{cases} \quad (2.2)$$

**Lemma 2.1** *For  $z > 0$  and  $x \neq z$ , we have*

- (i)  $|\varphi_z''(x)| \leq 4.32$  for  $0 < z < 1$ ,
- (ii)  $|\varphi_z''(x)| \leq \frac{C}{1+z}$  for  $0 < x < \frac{z}{2}$  and some constant  $C$  and
- (iii)  $|\varphi_z''(x)| \leq \frac{e^{-z^2/2}}{\sqrt{2\pi}} + \Phi(z)\left(\frac{1}{z} + z\right)$  for  $0 < x < z$  and  $z \geq 1$ .

**Proof.** From (2.2), we note that

$$\varphi_z'(x) = \begin{cases} \sqrt{2\pi}e^{\frac{x^2}{2}}(1 - \Phi(x))\Phi(z), & \text{if } x \geq z; \\ \sqrt{2\pi}e^{\frac{x^2}{2}}(1 - \Phi(z))\Phi(x), & \text{if } x < z, \end{cases}$$

and

$$\varphi_z''(x) = \begin{cases} -\Phi(z)[x - \sqrt{2\pi}(1 - \Phi(x))(1 + x^2)e^{x^2/2}], & \text{if } x > z; \\ (1 - \Phi(z))[x + \sqrt{2\pi}\Phi(x)(1 + x^2)e^{x^2/2}], & \text{if } x < z. \end{cases}$$

(i) Let  $0 < z < 1$ .

**Case 1**  $0 < x < z$ .

Since

$$\varphi_z'''(x) = \begin{cases} -\Phi(z)[2 + x^2 - \sqrt{2\pi}xe^{x^2/2}(1 - \Phi(x))(3 + x^2)], & \text{if } x > z; \\ (1 - \Phi(z))[2 + x^2 + \sqrt{2\pi}xe^{x^2/2}\Phi(x)(3 + x^2)], & \text{if } x < z, \end{cases}$$

$\varphi_z'''(x) > 0$  for  $0 < x < z$ , so that  $\varphi_z''$  is an increasing function on  $(0, z)$ . Hence

$$0 \leq \sqrt{\frac{\pi}{2}}(1 - \Phi(z)) = \varphi_z''(0) \leq \varphi_z''(x) \leq \frac{1}{2}[1 + 2\sqrt{2\pi}\Phi(1)e^{1/2}] \leq 3.98. \quad (2.3)$$

**Case 2**  $x < -1$ .

Since

$$\frac{1}{x^3} < x + \sqrt{2\pi}\Phi(x)(1 + x^2)e^{x^2/2} < -\frac{1}{x} \quad \text{if } x \leq 0, \quad (2.4)$$

([8], p. 133), we have

$$\left| x + \sqrt{2\pi}\Phi(x)(1 + x^2)e^{x^2/2} \right| < \left| \frac{1}{x} \right| < 1.$$

Therefore,

$$|\varphi_z''(x)| \leq 1. \quad (2.5)$$

**Case 3**  $-1 < x < 0$ .

$$\begin{aligned} |\varphi_z''(x)| &\leq \frac{1}{2}[|x| + \sqrt{2\pi}\Phi(x)(1 + x^2)e^{1/2}] \\ &\leq \frac{1}{2}(1 + 2\sqrt{2\pi}\Phi(0)e^{1/2}) \\ &\leq 2.57. \end{aligned} \quad (2.6)$$

**Case 4**  $x > z$ .

We note that

$$-\frac{1}{x} < x - \sqrt{2\pi}(1 - \Phi(x))(1 + x^2)e^{x^2/2} < \frac{1}{x^3} \quad \text{if } x > 0 \quad (2.7)$$

([8], p. 133).

If  $x \geq 1$ , then, by (2.7),

$$\left| x - \sqrt{2\pi}(1 - \Phi(x))(1 + x^2)e^{x^2/2} \right| < \left| \frac{1}{x} \right| \leq 1$$

which implies  $|\varphi_z''(x)| \leq 1$ .

If  $x < 1$ , then

$$|\varphi_z''(x)| \leq \Phi(1)[|x| + 2\sqrt{2\pi}(1 - \Phi(0))e^{1/2}] \leq \Phi(1)[1 + \sqrt{2\pi}e^{1/2}] \leq 4.32. \quad (2.8)$$

By (2.3), (2.5), (2.6) and (2.8), we completed the proof.

(ii) For  $0 < x \leq \frac{z}{2}$ , we have

$$\begin{aligned}
 |\varphi_z''(x)| &\leq |1 - \Phi(z)| |x + \Phi(x)\sqrt{2\pi}(1+x^2)e^{x^2/2}| \\
 &\leq \frac{e^{-z^2/2}}{z\sqrt{2\pi}} \left( \frac{z}{2} + \sqrt{2\pi}(1 + \frac{z^2}{4})e^{z^2/8} \right) \\
 &\leq 0.2e^{-z^2/2} + \frac{1 + \frac{z^2}{4}}{z} e^{-3z^2/8} \\
 &\leq 0.2e^{-z^2/2} + Ce^{-z/2} \\
 &\leq Ce^{-z/2} \\
 &\leq \frac{C}{1+z}.
 \end{aligned}$$

(iii) If  $0 < x < z$  and  $z \geq 1$ , then

$$\begin{aligned}
 0 \leq \varphi_z''(x) &= (1 - \Phi(z)) [x + \sqrt{2\pi}\Phi(x)(1+x^2)e^{\frac{x^2}{2}}] \\
 &\leq \frac{e^{-\frac{z^2}{2}}}{\sqrt{2\pi}z} [z + \sqrt{2\pi}\Phi(z)(1+z^2)e^{\frac{z^2}{2}}] \\
 &= \frac{e^{-\frac{z^2}{2}}}{\sqrt{2\pi}} + \Phi(z) \left( \frac{1}{z} + z \right)
 \end{aligned}$$

where we have used the fact that  $1 - \Phi(z) \leq \frac{e^{-\frac{z^2}{2}}}{\sqrt{2\pi}z}$  for  $z > 0$  in the first inequality.  $\square$

**Lemma 2.2** *If  $p = \frac{w_n}{n}$ ,  $\log n - w_n \rightarrow \infty$  and  $nw_n \rightarrow \infty$ , then*

- (i)  $E(W_{n-1} - W_n + 1)^2 = O(1)$  and
- (ii)  $E(W_{n-1} - W_n + 1)^4 = O(1)$ .

**Proof.** (i) See [8].

(ii) For  $i = 1, 2, \dots, n-1$ , let

$$Y_i = \begin{cases} 1, & \text{if the } i\text{-th vertex has degree 1 and it is jointed with the } n\text{-th vertex;} \\ 0, & \text{otherwise.} \end{cases}$$

Hence

$$\begin{aligned}
 E(W_{n-1} - W_n + 1)^4 &= E(Y_1 + Y_2 + \dots + Y_{n-1} - X_n + 1)^4 \\
 &\leq C \left\{ E(Y_1 + Y_2 + \dots + Y_{n-1})^4 + EX_n^4 + 1 \right\}. \quad (2.9)
 \end{aligned}$$

Note that  $EX_n^4 = q^{n-1} \leq 1$  and

$$EY_i^4 = P(Y_i = 1) = pq^{n-2}.$$

If  $i \neq j$ , then

$$EY_i^3 Y_j = P(Y_i = 1, Y_j = 1) = p^2 q^{n-2} q^{n-3} = p^2 q^{2n-5} = EY_i^2 Y_j^2.$$

If  $i \neq j \neq k$ , then

$$EY_i^2 Y_j Y_k = P(Y_i = 1, Y_j = 1, Y_k = 1) = p^3 q^{n-2} q^{n-3} q^{n-4} = p^3 q^{3n-9}.$$

If  $i \neq j \neq k \neq l$ , then

$$EY_i Y_j Y_k Y_l = P(Y_i = 1, Y_j = 1, Y_k = 1, Y_l = 1) = p^4 q^{n-2} q^{n-3} q^{n-4} q^{n-5} = p^4 q^{4n-14}.$$

Therefore,

$$\begin{aligned} E(Y_1 + Y_2 + \cdots + Y_{n-1})^4 &= \sum_{i=1}^{n-1} EY_i^4 + \sum_{i=1}^{n-1} \sum_{\substack{j=1 \\ j \neq i}}^{n-1} \{EY_i^3 Y_j + EY_i^2 Y_j^2\} \\ &\quad + \sum_{i=1}^{n-1} \sum_{\substack{j=1 \\ j \neq i}}^{n-1} \sum_{\substack{k=1 \\ k \neq i, j}}^{n-1} EY_i^2 Y_j Y_k + \sum_{i=1}^{n-1} \sum_{\substack{j=1 \\ i \neq j}}^{n-1} \sum_{\substack{k=1 \\ k \neq l}}^{n-1} \sum_{l=1}^{n-1} EY_i Y_j Y_k Y_l \\ &= (n-1)pq^{n-2} + (n-1)(n-2)(p^2 q^{2n-5} + p^2 q^{2n-5}) \\ &\quad + (n-1)(n-2)(n-3)p^3 q^{3n-9} \\ &\quad + (n-1)(n-2)(n-3)(n-4)p^4 q^{4n-14} \\ &= (n-1)pq^{n-2} \left\{ 1 + 2(n-2)pq^{n-3} + (n-2)(n-3)p^2 q^{2n-7} \right. \\ &\quad \left. + (n-2)(n-3)(n-4)p^3 q^{3n-10} \right\}. \end{aligned}$$

Let  $p = \frac{w_n}{n}$ . Then  $q^n = (1 - \frac{w_n}{n})^n \approx e^{-w_n} = e^{-np}$ . Therefore,

$$(n-1)pq^{n-2} = \frac{(n-1)p}{q^2} e^{-np} \leq \frac{1}{q^2} = \frac{1}{(1-p)^2}$$

and

$$(n-2)pq^{n-3} = \frac{(n-2)p}{q^3} e^{-np} \leq \frac{1}{q^3} = \frac{1}{(1-p)^3}.$$

Similarly,

$$(n-2)(n-3)p^2 q^{2n-7} \leq \frac{1}{(1-p)^7} \quad \text{and} \quad (n-2)pq^{n-3} \leq \frac{1}{(1-p)^{10}}.$$

Hence

$$E(Y_1 + Y_2 + \cdots + Y_{n-1})^4 \leq \frac{1}{(1-p)^{12}}. \quad (2.10)$$

Therefore, by (2.9) and (2.10),

$$E(W_{n-1} - W_n + 1)^4 = O(1).$$

□

### Proof of Theorem 1.3.

To prove Theorem 1.3, we let

$$U_n = \frac{W_n - \lambda_n}{\sigma_n}.$$

It suffices to consider  $z \geq 0$  as we can apply the result to  $-U_n$  when  $z < 0$ . Let  $z > 0$ . By Stein equation (2.1),

$$E(\varphi'(U_n) - U_n \varphi(U_n)) = P(U_n \leq z) - \Phi(z).$$

In [8], Kordecki showed that

$$\begin{aligned} |E(\varphi'_z(U_n) - U_n \varphi_z(U_n))| &\leq \left| E \left( \varphi'_z(U_n) \left( 1 - \frac{\lambda_n}{\sigma_n^2} (W_{n-1} - W_n + 1) \right) \right) \right| \\ &\quad + \frac{\lambda_n}{2\sigma_n^3} E \left( \left| \varphi''_z \left( U_n + \nu \frac{W_{n-1} - W_n + 1}{\sigma_n} \right) \right| (W_{n-1} - W_n + 1)^2 \right) \\ &:= A_1 + A_2 \end{aligned}$$

where  $0 < \nu < 1$  and

$$E(\varphi'_z(U_n)(W_{n-1} - W_n + 1)) = E(\varphi'_z(U_n)E(W_{n-1} - W_n + 1|W_n)).$$

By Chen and Shao [1], we have

$$E|\varphi'_z(U_n)| \leq \frac{C}{(1+z)^2}$$

for some constant  $C$ . Therefore

$$\begin{aligned} &\left| E \left( \varphi'_z(U_n) E \left( 1 - \frac{\lambda_n}{\sigma_n^2} (W_{n-1} - W_n + 1) | W_n \right) - \varphi'_z(U_n) E \left( 1 - \frac{\lambda_n}{\sigma_n^2} (W_{n-1} - W_n + 1) \right) \right) \right| \\ &\leq E|\varphi'_z(U_n)| E \left| E \left( 1 - \frac{\lambda_n}{\sigma_n^2} (W_{n-1} - W_n + 1) | W_n \right) - E \left( 1 - \frac{\lambda_n}{\sigma_n^2} (W_{n-1} - W_n + 1) \right) \right| \\ &\leq \frac{C}{(1+z)^2} \left\{ \text{Var} \left( E \left( 1 - \frac{\lambda_n}{\sigma_n^2} (W_{n-1} - W_n + 1) | W_n \right) \right) \right\}^{1/2} \\ &= \frac{C}{(1+z)^2} \frac{\lambda_n}{\sigma_n^2} \left\{ \text{Var} \left( E((W_{n-1} - W_n + 1) | W_n) \right) \right\}^{1/2} \\ &= \frac{C}{(1+z)^2} \frac{\lambda_n}{\sigma_n^2} d'_n \end{aligned}$$

where

$$d'_n = \{\text{Var}(E((W_{n-1} - W_n + 1|W_n))\})^{1/2}.$$

Hence we obtain that

$$\begin{aligned} A_1 &\leq \left| E\varphi'_z(U_n)E\left(1 - \frac{\lambda_n}{\sigma_n^2}(W_{n-1} - W_n + 1)\right) \right| + \frac{C}{(1+z)^2} \frac{\lambda_n}{\sigma_n^2} d'_n \\ &\leq \frac{C}{(1+z)^2} \left| 1 - \frac{\lambda_n(\lambda_{n-1} - \lambda_n + 1)}{\sigma_n^2} \right| + \frac{C}{(1+z)^2} \frac{\lambda_n}{\sigma_n^2} d'_n \\ &= \frac{C}{(1+z)^2 \sigma_n} \end{aligned}$$

where we have used the fact that  $|E(1 - \frac{\lambda_n}{\sigma_n^2}(W_{n-1} - W_n + 1))| = 0$  and

$\frac{\lambda_n}{\sigma_n^2} d'_n = O(\sigma_n^{-1})$  (see [8], p. 139) in the last inequality.

Consider  $A_2$ : if  $0 < z < 1$ , then by Lemma 2.1(i) and Lemma 2.2(i),

$$A_2 \leq \frac{\lambda_n}{2\sigma_n^3} E(W_{n-1} - W_n + 1)^2 = \frac{\lambda_n}{2\sigma_n^3} \frac{C}{1+z}.$$

If  $z \geq 1$ , then

$$\begin{aligned} A_2 &= \frac{\lambda_n}{2\sigma_n^3} E\left(|\varphi''_z(U_n + \nu \frac{W_{n-1} - W_n + 1}{\sigma_n})| \right. \\ &\quad \times (W_{n-1} - W_n + 1)^2 I(U_n + \nu \frac{W_{n-1} - W_n + 1}{\sigma_n} < 0)) \\ &\quad + \frac{\lambda_n}{2\sigma_n^3} E\left(|\varphi''_z(U_n + \nu \frac{W_{n-1} - W_n + 1}{\sigma_n})| \right. \\ &\quad \times (W_{n-1} - W_n + 1)^2 I(0 \leq U_n + \nu \frac{W_{n-1} - W_n + 1}{\sigma_n} \leq \frac{z}{2})) \\ &\quad + \frac{\lambda_n}{2\sigma_n^3} E\left(|\varphi''_z(U_n + \nu \frac{W_{n-1} - W_n + 1}{\sigma_n})| \right. \\ &\quad \times (W_{n-1} - W_n + 1)^2 I(U_n + \nu \frac{W_{n-1} - W_n + 1}{\sigma_n} > \frac{z}{2})) \\ &:= A_{21} + A_{22} + A_{23}. \end{aligned}$$

First we estimate  $A_{21}$ . If  $-1 < x < 0$ , then

$$|\varphi''_z(x)| \leq C(1 - \Phi(z)) \leq \frac{Ce^{-z^2/2}}{\sqrt{2\pi}z}.$$

Suppose that  $x < -1$ . By (2.4), we have

$$|\varphi''_z(x)| \leq (1 - \Phi(z)) \leq \frac{e^{-z^2/2}}{\sqrt{2\pi}z}$$



which implies that,

$$A_{21} \leq \frac{\lambda_n}{2\sigma_n^3} \frac{Ce^{-z^2/2}}{\sqrt{2\pi z}}.$$

By Lemma 2.1(ii) and Lemma 2.2(i),

$$A_{22} \leq \frac{\lambda_n}{2\sigma_n^3} \cdot \frac{C}{1+z} E(W_{n-1} - W_n + 1)^2 = \frac{\lambda_n}{2\sigma_n^3} \cdot \frac{C}{1+z}.$$

By (2.7), we have  $|\varphi_z''(x)| \leq 2$  for  $x > \frac{z}{2}$  and  $z \geq 1$ , so

$$\begin{aligned} A_{23} &\leq \frac{\lambda_n}{\sigma_n^3} |E(W_{n-1} - W_n + 1)^2 I(U_n + \nu \frac{W_{n-1} - W_n + 1}{\sigma_n} > \frac{z}{2})| \\ &\leq \frac{\lambda_n}{\sigma_n^3} \{E(W_{n-1} - W_n + 1)^4\}^{1/2} \{P(U_n + \nu \frac{W_{n-1} - W_n + 1}{\sigma_n} > \frac{z}{2})\}^{1/2} \\ &\leq \frac{\lambda_n}{\sigma_n^3} \{E(W_{n-1} - W_n + 1)^4\}^{1/2} \left\{ \frac{E(U_n + \nu \frac{W_{n-1} - W_n + 1}{\sigma_n})^2}{\frac{z^2}{4}} \right\}^{1/2} \\ &\leq \frac{2\sqrt{2}\lambda_n}{\sigma_n^3 z} \{E(W_{n-1} - W_n + 1)^4\}^{1/2} \{E(U_n^2) + E(\nu \frac{W_{n-1} - W_n + 1}{\sigma_n})^2\}^{1/2} \\ &\leq \frac{C\lambda_n}{\sigma_n^3} \frac{1}{1+z} \{E(W_{n-1} - W_n + 1)^4\}^{1/2} \left\{1 + \frac{O(1)}{\sigma_n^2}\right\}^{1/2}. \end{aligned}$$

Since  $\sigma_n^2 \rightarrow \infty$  (see [8], p.137) and by Lemma 2.2(2),

$$A_{23} \leq \frac{C\lambda_n}{2\sigma_n^3} \frac{1}{1+z} \left\{1 + \frac{O(1)}{\sigma_n^2}\right\}^{1/2} = \frac{C\lambda_n}{2\sigma_n^3} \frac{1}{1+z}.$$

Since  $\lambda_n/\sigma_n^2 = O(1)$  (see [8], p. 137),

$$A_2 \leq \frac{C\lambda_n}{2\sigma_n^3} \frac{1}{1+z} = \frac{C}{(1+z)\sigma_n}.$$

The comparison of  $A_1$  and  $A_2$  gives that

$$|P(\frac{W_n - \lambda_n}{\sigma_n} \leq z) - \Phi(z)| \leq \frac{C}{(1+|z|)\sigma_n}.$$

□

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