

УДК 519.179.4

LIMIT DISTRIBUTIONS OF VERTEX DEGREES IN A CONDITIONAL CONFIGURATION GRAPH

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The configuration graph where vertex degrees are independent identically distributed random variables is often used for modeling of complex networks such as the Internet. We consider a random graph consisting of N vertices. The random variables η_1, \dots, η_N are equal to the degrees of vertices with the numbers $1, \dots, N$. The probability $\mathbf{P}\{\eta_i = k\}$, $i = 1, \dots, N$, $k = 1, 2, \dots$, is equivalent to $h(k)/k^\tau$ as $k \rightarrow \infty$, where $h(x)$ is a slowly varying function integrable in any finite interval, $\tau > 1$. We obtain the limit distributions of the maximum vertex degree and the number of vertices with a given degree under the condition that the sum of degrees is equal to n and $N, n \rightarrow \infty$.

Key words: configuration graph; the limit distribution; vertex degree.

И. А. Чеплюкова, Ю. Л. Павлов. ПРЕДЕЛЬНЫЕ РАСПРЕДЕЛЕНИЯ СТЕПЕНЕЙ ВЕРШИН В УСЛОВНОМ КОНФИГУРАЦИОННОМ ГРАФЕ

Конфигурационный граф, степени вершин которого являются независимыми одинаково распределенными случайными величинами, часто используют для моделирования сложных сетей, таких как Интернет. Мы рассматриваем случайный граф с N вершинами. Случайные величины η_1, \dots, η_N равны степеням вершин с номерами $1, \dots, N$. Вероятность $\mathbf{P}\{\eta_i = k\}$, $i = 1, \dots, N$, $k = 1, 2, \dots$, пропорциональна величине $h(k)/k^\tau$ при $k \rightarrow \infty$, где $h(x)$ – интегрируемая на любом конечном интервале медленно меняющаяся функция и $\tau > 1$. Найдены предельные распределения максимальной степени вершин и числа вершин заданной степени при условии, что сумма степеней равна n при $N, n \rightarrow \infty$.

Ключевые слова: конфигурационный граф; предельное распределение; степень вершины.

INTRODUCTION

The study of random graphs has been causing growing interest in connection with the wide use of these models for the description of complex networks (see, e. g. [3, 6, 11]). Such models can be used to adequately describe the topology of transport, electricity, social, telecommunication networks and global Internet. Observations on

real networks showed that their topology can be described by random graphs with vertex degrees being independent identically distributed random variables with power-law distribution. In [3] it was suggested that for large k the number of vertices with the degree k is proportional to $k^{-\tau}$, where $\tau > 1$. That is why in [11] it was suggested that the distribution of the vertex degree η is

$$\mathbf{P}\{\eta \geq k\} = h(k)k^{-\tau+1}, \quad k = 1, 2, \dots, \quad (1)$$

where $h(k)$ is a slowly varying function.

We consider a random graph consisting of $N+1$ vertices. Let random variables η_1, \dots, η_N be equal to the degrees of vertices with the numbers $1, \dots, N$. Each vertex is assigned a certain degree in accordance with the distribution (1). The vertex degree is the number of stubs (or semiedges) that are numbered in an arbitrary order. Stubs are vertex edges for which adjacent nodes are not yet determined. The vertex 0 is auxiliary and has degree 0 if the sum of all other vertices is even, else the degree is 1. It is clear that we need to use the auxiliary vertex 0 for the sum of degrees to be even. The graph is constructed by joining all the stubs pairwise equiprobably to form links.

There are many papers where the results describing the limit behaviour of different random graph characteristics were obtained. The authors of [11] were sure (without proof) that the function $h(k)$ in (1) does not influence the limit results, and that to study the configuration graph one can replace $h(k)$ with the constant 1. In our work we will show that the role of the slowly varying function $h(k)$ is more complicated.

We consider the subset of random graphs under the condition $\eta_1 + \dots + \eta_N = n$. Such conditional graphs can be useful for modeling of networks for which we can estimate the number of communications. They are useful also for studying networks without conditions on the number of links by averaging the results of conditional graphs with respect to the distribution of the sum of degrees. Conditional random graphs were first analyzed in [9], where $h(k) \equiv 1$. Obviously, the limit behaviour of a random graph depends on the degree structure. In [9] the limit distributions were obtained for the maximum vertex degree and the number of vertices of a given degree as N and n tend to infinity in such a way that $1 < n/N < \zeta(\tau)$, where $\zeta(\tau)$ is the value of the Riemann's zeta-function at the point τ . For other zones of parameters analogous results were obtained in papers [7, 8, 10].

Here we extend the results on the maximum vertex degree and the number of vertices of a given degree to the configuration graphs with degree distribution (1), where $h(k)$ is not constant. In the following section the main results are formulated, then auxiliary statements are

proved. And the last section contains proofs of the main results.

MAIN RESULTS

In the paper we assume that the distributions of node degrees are

$$p_k = \mathbf{P}\{\eta_i = k\} = \frac{h(k)}{k^\tau \Sigma(1, \tau)}, \quad (2)$$

where $i = 1, \dots, N$, $k = 1, 2, \dots$, $\tau > 1$, $h(k)$ is a slowly varying function integrable in any finite interval and

$$\Sigma(x, y) = \sum_{k=1}^{\infty} x^k h(k) k^{-y}. \quad (3)$$

We denote also by ξ_1, \dots, ξ_N the auxiliary independent identically distributed random variables such that

$$p_r(\lambda) = \mathbf{P}\{\xi_i = k\} = \lambda^k p_k \Sigma(1, \tau) / \Sigma(\lambda, \tau), \quad (4)$$

where $i = 1, \dots, N$, $k = 1, 2, \dots$ and the parameter $\lambda = \lambda(n, N)$ belongs to the interval $(0, 1)$. From (2)–(4) we obtain

$$m = \mathbf{E}\xi_1 = \Sigma(\lambda, \tau - 1) / \Sigma(\lambda, \tau), \quad (5)$$

$$\sigma^2 = \mathbf{D}\xi_1 = \Sigma(\lambda, \tau - 2) / \Sigma(\lambda, \tau) - m^2.$$

Let the parameter $\lambda = \lambda(n, N)$ of the distribution (4) be determined by the relation

$$m = \Sigma(\lambda, \tau - 1) / \Sigma(\lambda, \tau) = n/N. \quad (6)$$

We denote by $\eta_{(N)}$ and μ_r the maximum vertex degree and the number of vertices with the degree r , respectively. We get the following results.

Theorem 1. *Let $n, N \rightarrow \infty, n/N \rightarrow 1, (n - N)^3/N^2 \rightarrow \infty$ and let r be such that*

$$\frac{N\lambda^{r-1}h(r)}{r^\tau} \rightarrow \infty, \quad \frac{N\lambda^{r+1}h(r+1)}{\Sigma(\lambda, \tau)(r+1)^\tau} \rightarrow \gamma,$$

where γ is a nonnegative constant. Then

$$\mathbf{P}\{\eta_{(N)} = r\} \rightarrow e^{-\gamma},$$

$$\mathbf{P}\{\eta_{(N)} = r + 1\} \rightarrow 1 - e^{-\gamma}.$$

We introduce the conditions:

- (A1) $\tau > 4$;
(A2) $3 < \tau \leq 4$, $(1 - \lambda)^{\tau-4-\varepsilon}/\sqrt{N} \rightarrow 0$;
(A3) $5/2 < \tau \leq 3$, $N(1 - \lambda)^{11-3\tau+\varepsilon} \geq C_3 > 0$;
(A4) $\tau = 5/2$, $N(-\ln(1 - \lambda))^2(1 - \lambda)^{7/2+\varepsilon} \geq C_4 > 0$;
(A5) $1 < \tau < 5/2$, $N(1 - \lambda)^{6-\tau+\varepsilon} \geq C_5 > 0$,

where ε is some sufficiently small positive constant.

Theorem 2. Let $N, n \rightarrow \infty, n/N \nearrow \Sigma(1, \tau - 1)/\Sigma(1, \tau)$, one of the following conditions (A1) – (A5) is fulfilled, and $r = r(N, n)$ take values in such a way that

$$\frac{N\lambda^{r+1}h(r+1)}{(r+1)^\tau \Sigma(\lambda, \tau)(1-\lambda)} \rightarrow \gamma,$$

where γ is a positive constant. Then

$$\mathbf{P}\{\eta_{(N)} \leq r\} = e^{-\gamma}(1 + o(1)).$$

Theorem 3. Let $n, N \rightarrow \infty$ and one of the following conditions is fulfilled

1. $n/N \rightarrow 1$, $r = 1, 2$, $(n - N)^2/N \rightarrow \infty$;
2. $n/N \rightarrow 1$, $r \geq 3$, $N\lambda^{r-1} \rightarrow \infty$;
3. $n/N \nearrow \Sigma(1, \tau - 1)/\Sigma(1, \tau)$, parameters τ, N, n are determined by one of the conditions (A1) – (A5).

Then for a nonnegative integer k uniformly with respect to $u = (k - Np_r(\lambda))/(\sigma_{rr}\sqrt{N})$ lies in any fixed finite interval

$$\mathbf{P}\{\mu_r = k\} = \frac{1}{\sigma_{rr}\sqrt{2\pi N}} e^{-u^2/2}(1 + o(1)),$$

where

$$\sigma_{rr}^2 = p_r(\lambda) \left(1 - p_r(\lambda) - \frac{(n/N - r)^2}{\sigma^2} p_r(\lambda) \right).$$

Theorem 4. Let $n, N \rightarrow \infty, n/N \rightarrow 1, n - N \rightarrow \infty, r \geq 2$. Then

$$\mathbf{P}\{\mu_r = k\} = \frac{1 + o(1)}{k!} (Np_r(\lambda))^k \exp\{-Np_r(\lambda)\}$$

uniformly with respect to $(k - Np_r(\lambda))/\sqrt{Np_r(\lambda)}$ lies in any fixed finite interval. This assertion remains true for $r \rightarrow \infty$ if $1 < n/N < \Sigma(1, \tau - 1)/\Sigma(1, \tau)$ under one of the following conditions:

1. $n/N \rightarrow 1$, $n - N \rightarrow \infty$;

2. $n/N \nearrow \Sigma(1, \tau - 1)/\Sigma(1, \tau)$, parameters τ, N, n are determined by one of the conditions (A1) – (A5).

Remark. In [2], a case of these theorems under the condition $1 < C_1 \leq n/N \leq C_2 < \Sigma(1, \tau - 1)/\Sigma(1, \tau)$ was proved.

AUXILIARY RESULTS

We prove some auxiliary statements (Lemmas 1–6), and use them to prove Theorems 1–5. The technique of obtaining these theorems is based on the generalized allocation scheme suggested by V. F. Kolchin [5]. It is readily seen that for our subset of random graphs

$$\begin{aligned} & \mathbf{P}\{\eta_1 = k_1, \dots, \eta_N = k_N\} = \\ & = \mathbf{P}\{\xi_1 = k_1, \dots, \xi_N = k_N | \xi_1 + \dots + \xi_N = n\}. \end{aligned}$$

Therefore, the conditions of the generalized allocation scheme are valid (see [5]). Let $\xi_1^{(r)}, \dots, \xi_N^{(r)}$ and $\tilde{\xi}_1^{(r)}, \dots, \tilde{\xi}_N^{(r)}$ be two sets of independent identically distributed random variables such that

$$\mathbf{P}\{\xi_1^{(r)} = k\} = \mathbf{P}\{\xi_1 = k | \xi_1 \leq r\}, \quad (7)$$

$$\mathbf{P}\{\tilde{\xi}_1^{(r)} = k\} = \mathbf{P}\{\xi_1 = k | \xi_1 \neq r\}, \quad k = 1, 2, \dots$$

We also put

$$\zeta_N = \xi_1 + \dots + \xi_N, \quad \zeta_N^{(r)} = \xi_1^{(r)} + \dots + \xi_N^{(r)},$$

$$\tilde{\zeta}_N^{(r)} = \tilde{\xi}_1^{(r)} + \dots + \tilde{\xi}_N^{(r)}, \quad P_r = \mathbf{P}\{\xi_1 > r\}.$$

It is shown in [5] that

$$\mathbf{P}\{\eta_{(N)} \leq r\} = (1 - P_r)^N \frac{\mathbf{P}\{\zeta_N^{(r)} = n\}}{\mathbf{P}\{\zeta_N = n\}}, \quad (8)$$

$$\begin{aligned} \mathbf{P}\{\mu_r = k\} &= \binom{N}{k} p_r^k(\lambda) (1 - p_r(\lambda))^{N-k} \times \\ & \times \frac{\mathbf{P}\{\tilde{\zeta}_{N-k}^{(r)} = n - kr\}}{\mathbf{P}\{\zeta_N = n\}}. \end{aligned} \quad (9)$$

From (2)–(6) we can deduce the next lemma.

Lemma 1. Let $N, n \rightarrow \infty$. The next assertions are true:

1. if $n/N \rightarrow 1$ then $\lambda = ((n/N - 1)p_1/p_2)(1 + o(1))$;
2. if $n/N \nearrow \Sigma(1, \tau - 1)/\Sigma(1, \tau)$ then $\lambda \rightarrow 1$.

Let us consider the limit behaviour of ζ_N .

Lemma 2. Under the conditions of Theorems 1–4

$$\mathbf{P}\{\zeta_N = k\} = \frac{1 + o(1)}{\sigma\sqrt{2\pi N}} \exp\left\{-\frac{(k-n)^2}{2\sigma^2 N}\right\}$$

uniformly with respect to integers k such that $(k-n)/(\sigma\sqrt{N})$ lies in any fixed finite interval.

Proof. Let $\varphi(t)$ be the characteristic function of the random variable ξ_1 . Then

$$\varphi(t) = \Sigma(e^{it}\lambda, \tau)/\Sigma(\lambda, \tau). \quad (10)$$

Further we will need an explicit form of the third derivative of $\ln \varphi(t)$. From (4) it is not hard to get that

$$\begin{aligned} (\ln \varphi(t))''' &= i \left(-\frac{\Sigma(e^{it}\lambda, \tau - 3)}{\Sigma(e^{it}\lambda, \tau)} + \right. \\ &+ 3 \frac{\Sigma(e^{it}\lambda, \tau - 2)\Sigma(e^{it}\lambda, \tau - 1)}{\Sigma^2(e^{it}\lambda, \tau)} - \\ &\left. - 2 \frac{\Sigma^3(e^{it}\lambda, \tau - 1)}{\Sigma^3(e^{it}\lambda, \tau)} \right). \quad (11) \end{aligned}$$

Let $n/N \rightarrow 1$. From (2)–(4) it is easy to obtain that

$$\sigma^2 = O(\lambda), \quad |(\ln \varphi(t))'''| \leq C_3 \lambda. \quad (12)$$

Let $\varphi_N(t)$ be the characteristic function of the random variable $(\zeta_N - n)/(\sigma\sqrt{N})$. Then

$$\begin{aligned} \ln \varphi_N(t) &= -\frac{int}{\sigma\sqrt{N}} + N \ln \varphi\left(\frac{t}{\sigma\sqrt{N}}\right) = \\ &= -\frac{t^2}{2} + \frac{t^3 Q(t/(\sigma\sqrt{N}))}{6\sigma^3\sqrt{N}}. \quad (13) \end{aligned}$$

Then from Lemma 1, (12) and (13) follows relation

$$\ln \varphi_N(t) = -t^2/2 + o(1). \quad (14)$$

Let $n/N \nearrow \Sigma(1, \tau - 1)/\Sigma(1, \tau)$. It is well known (see e.g. [4]) that the slowly varying

function integrable in any finite interval has the following properties:

1. $h(x) > 1/\sqrt{x}$ for sufficiently large x ;
2. $\lim_{x \rightarrow \infty} h(x+t)/h(x) = 1, \quad t \geq 0$;
3. $\lim_{x \rightarrow \infty} h(x)/x^\varepsilon = 0, \quad \lim_{x \rightarrow \infty} h(x)x^\varepsilon = \infty$ for any $\varepsilon > 0$;
4. $h(x) = c(x) \exp\left\{\int_\alpha^x \varepsilon(t)/tdt\right\}$, where $\alpha > 0, c(x) \rightarrow c \neq 0, \varepsilon(x) \rightarrow 0$, as $x \rightarrow \infty$.

Using the properties (15) and Lemma 1 we can deduce that for $j = 0, 1, 2, 3$

$$|\Sigma(e^{it}\lambda, \tau - j)| \leq C_4 \lambda \Phi(\lambda, \tau - j, 1) + \lambda \Phi(\lambda, \tau - j - \varepsilon, 1), \quad (16)$$

$$|\Sigma(e^{it}\lambda, \tau)| \geq C_5 \quad \text{as } t \rightarrow 0, \quad (17)$$

where $\Phi(x, s, a)$ is the Lerch transcendent function:

$$\Phi(x, s, a) = \sum_{k=0}^{\infty} x^k (k+a)^{-s}. \quad (18)$$

It is well known (see e.g. [1]) that for the Lerch transcendent function the following properties are valid:

1. $\lambda \Phi(\lambda, 1, 1) = -\ln(1 - \lambda),$
 $(1 - \lambda)\Phi(\lambda, 0, 1) = 1;$
2. $(1 - \lambda)\Phi(\lambda, \tau, 1) = O((1 - \lambda)^\tau),$
 $\tau < 1, \quad \lambda \rightarrow 1.$

From (3), (5), (11), (16)–(19) it is not hard to get that

$$\sigma^2 \geq \begin{cases} C_6 > 0, & \tau > 5/2; \\ C_7(-\ln(1 - \lambda)), & \tau = 5/2; \\ C_8(1 - \lambda)^{\tau-5/2}, & 1 < \tau < 5/2, \end{cases} \quad (20)$$

$$\sigma^2 \leq \begin{cases} C_9 > 0, & \tau > 3; \\ C_{10}(1 - \lambda)^{\tau-3-\varepsilon}, & 1 < \tau \leq 3, \end{cases} \quad (21)$$

$$|\varphi'''(t)| = \begin{cases} O(1), & \tau > 4; \\ O((1 - \lambda)^{\tau-4-\varepsilon}), & 1 < \tau \leq 4. \end{cases} \quad (22)$$

The next expression is valid for a sufficiently small t :

$$\begin{aligned} \ln \varphi(t) &= t (\ln \varphi(t))'|_{t=0} + \frac{t^2}{2} (\ln \varphi(t))''|_{t=0} + \\ &+ \frac{t^3}{6} Q(t), \quad (23) \end{aligned}$$

where $|Q(t)| \leq 2 \max_{|u| \leq |t|} |(\ln \varphi(u))'''|$.

Using (13), (20)–(23) and (A1)–(A5) we get (14).

According to the inversion formula we represent the probability $\mathbf{P}\{\zeta_N = k\}$ as the integral

$$\mathbf{P}\{\zeta_N = k\} = \frac{1}{\sigma\sqrt{2\pi N}} \int_{-\pi\sigma\sqrt{N}}^{\pi\sigma\sqrt{N}} e^{-izt} \varphi_N(t) dt,$$

where $z = (k - n)/(\sigma\sqrt{N})$. Since

$$(\sqrt{2\pi})^{-1} e^{-z^2/2} = (2\pi)^{-1} \int_{-\infty}^{\infty} e^{-izt - t^2/2} dt, \quad (24)$$

the difference

$$R = 2\pi[\sigma\sqrt{N}\mathbf{P}\{\zeta_N = k\} - (2\pi)^{-1/2} e^{-z^2/2}]$$

can be rewritten as the sum of four integrals: $R = I_1 + I_2 + I_3 + I_4$, where

$$\begin{aligned} I_1 &= \int_{-A}^A e^{-izt} [\varphi_N(t) - e^{-t^2/2}] dt, \\ I_2 &= \int_{A < |t| < a\sigma\sqrt{N}} e^{-izt} \varphi_N(t) dt, \\ I_3 &= \int_{a\sigma\sqrt{N} \leq |t| \leq \pi\sigma\sqrt{N}} e^{-izt} \varphi_N(t) dt, \\ I_4 &= - \int_{A < |t|} e^{-izt - t^2/2} dt, \end{aligned} \quad (25)$$

$$\begin{aligned} I_3 &= \int_{a\sigma\sqrt{N} \leq |t| \leq \pi\sigma\sqrt{N}} e^{-izt} \varphi_N(t) dt, \\ I_4 &= - \int_{A < |t|} e^{-izt - t^2/2} dt, \end{aligned}$$

$$B(\lambda, \tau) = \begin{cases} (1 - \lambda)^{-\tau+4+\varepsilon}, & 5/2 < \tau \leq 4; \\ (-\ln(1 - \lambda))(1 - \lambda)^{3/2+\varepsilon}, & \tau = 5/2; \\ (1 - \lambda)^{3/2+\varepsilon}, & 1 < \tau < 5/2. \end{cases} \quad (27)$$

From (13) we get that

$$\begin{aligned} &\ln \varphi \left(\frac{t}{\sigma\sqrt{N}} \right) = \\ &= \frac{itm}{\sigma\sqrt{N}} - \frac{t^2}{2N} + \frac{t^3}{6\sigma^3 N^{3/2}} Q \left(\frac{t}{\sigma\sqrt{N}} \right), \end{aligned}$$

where

$$|Q(t/(\sigma\sqrt{N}))| \leq 2 \max_{|u| \leq |t/(\sigma\sqrt{N})|} |\ln''' \varphi(u)|.$$

In the integration domains of the integral I_2' $t/(\sigma\sqrt{N}) \rightarrow 0$, then from Lemma 1, (17), (20) and (27) we obtain:

$$\left| \frac{t}{(\sigma^3\sqrt{N})} Q \left(\frac{t}{\sigma\sqrt{N}} \right) \right| \leq a.$$

the positive constants A and a will be chosen later. Lemma 2 will be proved if we show that by choosing sufficiently large n, N the difference R can be made arbitrarily small. From (14) we get that $I_1 \rightarrow 0$. Moreover,

$$|I_4| \leq \int_{A < |t|} e^{-t^2/2} dt, \quad (26)$$

and the integral I_4 is as small as desired, provided that A is large enough.

Let us estimate the integral I_2 . From (23) and (12) we obtain that for sufficiently small a $|\varphi_N(t)| \leq e^{-C_{11}t^2}$ as $n/N \rightarrow 1$, therefore the next estimate is true $|I_2| \leq \int_{A < |t|} e^{-C_{11}t^2} dt$, and the integral I_2 is small for large enough A . From (13), (17) and (20) we obtain the same estimate as $n/N \nearrow \Sigma(1, \tau - 1)/\Sigma(1, \tau)$, $\tau > 4$.

Let $n/N \nearrow \Sigma(1, \tau - 1)/\Sigma(1, \tau)$, $\tau \leq 4$. We divide I_2 into integrals I_2' and I_2'' , where the integration domains are

$$\{t : A < |t| \leq aB(\lambda, \tau)\sigma\sqrt{N}\}$$

and

$$\{t : aB(\lambda, \tau)\sigma\sqrt{N} < |t| \leq a\sigma\sqrt{N}\},$$

where

It follows that for small enough a $|\varphi_N(t)| \leq \exp\{-C_{12}t^2\}$. Therefore

$$|I_2'| \leq 2 \int_{A < |t|} e^{-C_{12}t^2} dt,$$

and the integral I_2' is as small as desired, provided that A is large enough. To estimate the integral I_2'' we expand the function $\Sigma(\lambda z, \tau)$, where $z = e^{it/(\sigma\sqrt{N})}$ in the Taylor series near the point $z = 1$. Then

$$\begin{aligned} \varphi(t) &= 1 - (1 + o(1))\Sigma(\lambda, \tau - 1)\Sigma^{-1}(\lambda, \tau) \times \\ &\quad \times (1 - \cos(t/(\sigma\sqrt{N})) - i \sin(t/(\sigma\sqrt{N}))). \end{aligned}$$

Therefore

$$|\varphi_N(t)| \leq |\varphi^N(t/(\sigma\sqrt{N}))| \leq$$

$$\leq \exp\{C_{13}N(1 - \cos(t/(\sigma\sqrt{N})))\}.$$

Using (27), the conditions (A1) – (A5) and the inequality

$$1 - \cos(t/(\sigma\sqrt{N})) \geq 1 - (1 - C_{14}t^2/(\sigma^2N)),$$

$$|t| < a\sigma\sqrt{N},$$

we can show that

$$\begin{aligned} |I_2''| &\leq \int_{aB(\lambda,\tau)\sigma\sqrt{N}}^{\infty} e^{-C_{14}t^2/\sigma^2} dt \leq \\ &\leq C_{15} \frac{\sigma}{aB(\lambda,\tau)\sqrt{N}} e^{-C_{14}a^2B^2(\lambda,\tau)N} \rightarrow 0. \end{aligned}$$

Let us consider the integral I_3 . For $\varepsilon \leq |t| \leq \pi$ the inequality

$$|\varphi(t)| \leq e^{-C_{16}} \quad (28)$$

is valid. Then under the condition that $n/N \rightarrow 1$ it can be shown that

$$\varphi(t) = e^{it} (1 + \lambda p_2(e^{it} - 1)/p_1) + o(\lambda^2).$$

From this and Lemma 1 we get that for $\varepsilon \leq |t|/(\sigma\sqrt{N}) \leq \pi$

$$|\varphi(t/(\sigma\sqrt{N}))| \leq e^{-C_{17}\lambda}.$$

Therefore using relations (12) and (25) it is not hard to see that

$$|I_3| \leq C_{18}\sqrt{n - N} \exp\{-C_{19}(n - N)\} \rightarrow 0.$$

Let $n/N \rightarrow \Sigma(1, \tau - 1)/\Sigma(1, \tau)$. From the conditions (A1) – (A5), (21) and (28) we get that

$$|I_3| \leq C_{20}\sigma\sqrt{N}e^{-C_{21}N} \rightarrow 0. \quad (29)$$

Thus Lemma 2 is proved. \square

Let $\varphi_r(t)$ be the characteristic function of the random variable $(\zeta_N^{(r)} - n)/(\sigma\sqrt{N})$.

Lemma 3. *Let $n, N \rightarrow \infty$. Then uniformly with respect to t in any fixed finite interval the next conclusions are true*

1. if $n/N \rightarrow 1, (n - N)^3/N^2 \rightarrow \infty, NP_{r-1} \rightarrow \infty, NP_r \rightarrow \gamma$, where γ is a nonnegative constant, then for $s = 0, \pm 1$ $\varphi_{r+s}(t) \rightarrow e^{-t^2/2}$;
2. if $n/N \rightarrow \Sigma(1, \tau - 1)/\Sigma(1, \tau), NP_r \rightarrow \gamma$, where γ is a positive constant, parameters τ, N, n are determined by the conditions (A1) – (A5) then $\varphi_r(t) \rightarrow e^{-t^2/2}$.

Proof. From (7) and (10) it is easy to see that

$$\begin{aligned} \varphi_r(t) &= \quad (30) \\ &= \exp\left\{-\frac{itn}{\sigma\sqrt{N}}\right\} (1 - P_r)^{-N} \varphi^N\left(\frac{t}{\sigma\sqrt{N}}\right) \times \\ &\times \left(1 - (1 + o(1)) \sum_{k=r+1}^{\infty} p_k(\lambda) \exp\left\{\frac{itk}{\sigma\sqrt{N}}\right\}\right)^N. \end{aligned}$$

It is not hard to get that

$$\begin{aligned} \sum_{k=r+1}^{\infty} p_k(\lambda) \exp\left\{tk/(\sigma\sqrt{N})\right\} &= \\ &= P_r + R(t), \quad (31) \end{aligned}$$

where $R(t) \leq |t/\sigma\sqrt{N}| \sum_{k=r+1}^{\infty} p_k(\lambda)k$.

Let $n/N \rightarrow 1$. It is clear that

$$NP_r = N \sum_{i \geq 0} p_{r+i+1}(\lambda) =$$

$$= N \left(\sum_{i=0}^M p_{r+i+1}(\lambda) + \sum_{i \geq M+1} p_{r+i+1}(\lambda) \right), \quad (32)$$

the positive constant M will be chosen later. For the fixed integer r we get from Lemma 1, (2)–(4) and (15) that

$$\sum_{i=0}^M p_{r+i+1}(\lambda) = \frac{N\lambda^{r+1}h(r+1)}{\Sigma(\lambda, \tau)(r+1)^\tau} (1 + o(1))$$

and for large enough N

$$\sum_{i \geq M+1} p_{r+i+1}(\lambda) = O\left(\sum_{i \geq 0}^M p_{r+i+1}(\lambda)\right).$$

Therefore

$$NP_r = \frac{N\lambda^{r+1}h(r+1)}{\Sigma(\lambda, \tau)(r+1)^\tau} (1 + o(1)). \quad (33)$$

Using $NP_r \rightarrow \gamma$ we obtain that for fixed integer r

$$(\sigma\sqrt{N})^{-1} \sum_{k>r} kp_k(\lambda) = o(N^{-1}). \quad (34)$$

As $r \rightarrow \infty$ we can deduce from Lemma 1, (2)–(4) and (15) the relation (33) is valid.

From (2)–(4), (15) and the relation $NP_r \rightarrow \gamma$ we can get that as $r \rightarrow \infty$

$$(\sigma\sqrt{N})^{-1} \sum_{k>r} kp_k(\lambda) \leq C_{22}t(r+1)p_r(\lambda)/(\sigma\sqrt{N}).$$

Since $NP_{r-1} \rightarrow \infty$ it is not hard to show that $r = o(\sqrt{n-N})$. From this, (31), (35) it follows that

$$(\sigma\sqrt{N})^{-1} \sum_{k>r} kp_k(\lambda) = o(N^{-1}).$$

Therefore, for $n/N \rightarrow 1$ the relation $\varphi_r(t) \rightarrow e^{-t^2/2}$ holds.

For $s = 1$ we get that $NP_{r+1} \rightarrow 0$. Therefore in this case the assertion of Lemma 3 follows from (30) by substituting r with $r + 1$.

Let $s = -1$. By analogy with the estimate (34) we can obtain that as $r \rightarrow \infty$

$$(\sigma\sqrt{N})^{-1} \sum_{k \geq r} p_k(\lambda)k \leq C_{23}rp_r(\lambda)/(\sigma\sqrt{N}).$$

Using (15) and the condition $(n-N)^3/N^2 \rightarrow \infty$ the relation (34) follows from this and (2)–(4). By analogy with this estimate for fixed integer r we can get that

$$\begin{aligned} & \frac{1}{\sigma\sqrt{N}} \sum_{k \geq r} p_k(\lambda)k = \\ & = \frac{trp_r(\lambda)}{\sigma\sqrt{N}} \sum_{k \geq r} \lambda^{k-r} \left(\frac{r}{k}\right)^{\tau-1} \frac{h(k)}{h(r)} \leq \\ & \leq C_{24} \frac{trp_r(\lambda)}{\sigma\sqrt{N}} = o\left(\frac{1}{N}\right). \end{aligned}$$

Therefore, as $n/N \rightarrow 1$, the relation $\varphi_{r-1}(t) \rightarrow e^{-t^2/2}$ holds.

Let $n/N \nearrow \Sigma(1, \tau - 1)/\Sigma(1, \tau)$. Using (2)–(4), the properties of the slowly varying function (15), Lemma 1 and (32) we can deduce that

$$\sum_{i=0}^M p_{r+i+1}(\lambda) = p_{r+1}(\lambda) \sum_{i=0}^M \lambda^i (1 + o(1)). \quad (36)$$

From (15) it is not hard to get

$$\sum_{i \geq M+1} p_{r+i+1}(\lambda) = o\left(\sum_{i=0}^M p_{r+i+1}(\lambda)\right). \quad (37)$$

From the condition $NP_r \rightarrow \gamma$ it follows that $r \rightarrow \infty$. Then from (32), (36), (37) we get that

$$NP_r = \frac{N\lambda^{r+1}h(r+1)}{\Sigma(\lambda, \tau)(r+1)^\tau(1-\lambda)}(1 + o(1)). \quad (38)$$

Using Lemma 1, (15), (38) and the condition $NP_r \rightarrow \gamma$ it is not hard to see that

$$\frac{t}{\sigma\sqrt{N}} \sum_{k \geq r+1} p_k(\lambda)k \leq C_{25} \frac{t(r+1)p_{r+1}(\lambda)}{\sigma\sqrt{N}(1-\lambda)} \leq$$

$$\leq C_{26} \frac{t(r+1)\gamma}{\sigma N^{3/2}}. \quad (39)$$

From (38) and the condition $NP_r \rightarrow \gamma$ it is easy to see that

$$\frac{N\lambda^{r+1}h(r+1)}{(\Sigma(\lambda, \tau)(r+1)^\tau(1-\lambda))} \rightarrow \gamma > 0. \quad (40)$$

Using the conditions (A1)–(A5), (20) and (40) we get that $(r+1)/(\sigma\sqrt{N}) = o(1)$. From this and (39) we can obtain that

$$t(\sigma\sqrt{N})^{-1} \sum_{k \geq r+1} p_k(\lambda)k = o(N^{-1}).$$

Then the assertion of Lemma 3 follows from (30) and (31). \square

Lemma 4. *Let $n, N \rightarrow \infty$ and one of the following conditions be fulfilled*

1. $n/N \rightarrow 1, (n-N)^3/N^2 \rightarrow \infty, NP_{r-1} \rightarrow \infty, NP_r \rightarrow \gamma$, where γ is a nonnegative constant;
2. $n/N \nearrow \Sigma(1, \tau - 1)/\Sigma(1, \tau), NP_r \rightarrow \gamma$, where γ is a positive constant and parameters τ, N, n are determined by one of the conditions (A1)–(A5).

Then

$$\mathbf{P}\{\zeta_N^{(r)} = k\} = \frac{1 + o(1)}{\sigma\sqrt{2\pi N}} \exp\left\{-\frac{(k-n)^2}{2\sigma^2 N}\right\}$$

uniformly with respect to integers k such that $(k-n)/(\sigma\sqrt{N})$ lies in any fixed finite interval.

Proof. We follow the scheme of proving Lemma 2. We represent the probability $\mathbf{P}\{\zeta_N^{(r)} = k\}$ as the integral

$$\mathbf{P}\{\zeta_N^{(r)} = k\} = \frac{1}{2\pi\sigma\sqrt{N}} \int_{-\pi\sigma\sqrt{N}}^{\pi\sigma\sqrt{N}} e^{-izt} \varphi_r(t) dt,$$

where $z = (k-n)/(\sigma\sqrt{N})$ and $\varphi_r(t)$ is the characteristic function of the random variable $(\zeta_N^{(r)} - n)/(\sigma\sqrt{N})$. Using (24) the difference

$$R = 2\pi[\sigma\sqrt{N}\mathbf{P}\{\zeta_N^{(r)} = k\} - (2\pi)^{-1/2}e^{-t^2/2}]$$

can be rewritten as the sum of four integrals: $R = I_1^{(r)} + I_2^{(r)} + I_3^{(r)} + I_4$, where I_4 is given by (25) and $I_1^{(r)} - I_3^{(r)}$ are constructed similarly to $I_1 - I_3$ by substituting $\varphi_r(t)$ instead of $\varphi_N(t)$ in (25).

From Lemma 3 it follows that $I_1^{(r)} \rightarrow 0$. From, (12), (20)–(23) and (30) we get that

$$|\varphi_r(t)| \leq (1 - P_r)^{-N} (\exp \{C_{27}t^2/N\} + C_{28}/N)^N.$$

Therefore, $|I_2^{(r)}| \leq C_{29} \int_A^\infty e^{-C_{30}t^2} dt$. It is clear that $I_2^{(r)}$ can be made arbitrarily small by choosing a sufficiently large A .

It is easy to show that we can estimate the integral $I_3^{(r)}$ by analogy with I_3 in Lemma 2. For I_4 we can use the estimation (26). This completes the proof of Lemma 4.

It is not hard to see that the conclusion of Lemma 4 is true when replacing r with $r - 1$ and $r + 1$ as $n/N \rightarrow 1$. In these cases the proofs are similar to the proof of Lemma 4. \square

From (4) and (7) we have that

$$\begin{aligned} m_r &= \mathbf{E}\xi_1^{\tilde{r}(r)} = (m - rp_r(\lambda))/(1 - p_r(\lambda)), \\ \sigma_r^2 &= \mathbf{D}\xi_1^{\tilde{r}(r)} = \frac{\sigma^2}{(1 - p_r(\lambda))^2} \times \\ &\times \left(1 - p_r(\lambda) - \frac{(m - r)^2}{\sigma^2} p_r(\lambda)\right). \end{aligned} \quad (41)$$

Let $\tilde{\varphi}_r(t)$ be the characteristic function of the random variable $(\tilde{\zeta}_N^{(r)} - Sm_r)/(\sigma_r\sqrt{S})$. By analogy with Lemmas 2–4 it is not hard to prove the following assertions.

Lemma 5. *Let $n, N \rightarrow \infty$ and one of the following conditions be fulfilled*

1. $n/N \rightarrow 1, r = 1, (n - N)^2/N \rightarrow \infty$;
2. $n/N \rightarrow 1, r = 2, (n - N)^2/N \rightarrow \infty$;
3. $n/N \rightarrow 1, r \geq 3, n - N \rightarrow \infty$;
4. $n/N \rightarrow \Sigma(1, \tau - 1)/\Sigma(1, \tau)$ and parameters τ, N, n are determined by one of the conditions (A1)–(A5).

Then for $S = N(1 - p_r(\lambda))(1 + o(1))$

$$\tilde{\varphi}_r(t) \rightarrow e^{-t^2/2}$$

uniformly with respect to t in any fixed finite interval.

Lemma 6. *Under the conditions of Lemma 5 for $S = N(1 - p_r(\lambda))(1 + o(1))$*

$$\mathbf{P} \left\{ \tilde{\zeta}_S^{(r)} = k \right\} = \frac{1}{\sigma_r\sqrt{2\pi S}} e^{-z^2/2} (1 + o(1))$$

uniformly with respect to integers k such that $z = (k - Sm_r)/(\sigma_r\sqrt{S})$ lies in any fixed finite interval.

PROOFS OF THEOREMS

We are now ready to prove Theorems 1–4. Using Lemmas 2 and 4 we obtain

$$\mathbf{P}\{\zeta_N^{(r)} = n\}/\mathbf{P}\{\zeta_N = n\} \rightarrow 1. \quad (42)$$

The assertion of Theorem 1 follows from $NP_r \rightarrow \gamma, NP_{r-1} \rightarrow \infty$, (8), (33) and (42). The assertion of Theorem 2 follows from (8), (38) and (42).

According to the normal approximation of the binomial distribution under the condition $Np_r(\lambda)(1 - p_r(\lambda)) \rightarrow \infty$ Theorem 3 follows from (9), (41) and Lemmas 2, 6.

Using Poisson approximation of the binomial distribution as $p_r(\lambda) \rightarrow 0$, Lemmas 2, 6 and relations (9), (41) we can obtain the assertion of Theorem 4.

The study was carried out under state order to the Karelian Research Centre of the Russian Academy of Sciences (Institute of Applied Mathematical Research KarRC RAS) and supported by the Russian Foundation for Basic Research, grant 16-01-00005a.

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Received January 31, 2018

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