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On the estimation of the remainder term in the CLT for the number of occupied cells in a multinomial random allocation scheme

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RESUME

In this paper the best estimate for the remainder term in the central limit theorem for the number of occupied cells in the multinomial random allocation scheme is established.

Key words: Random allocation, Poisson distribution, central limit theorem.

1. Introduction

We consider a model with n balls and N cells (urns) numbered 1, 2, ..., N, where $N = N(n) \to \infty$ as $n \to \infty$. Balls are allocated at random independently of each other. The probability of a ball falling into the mth cell is $p_m > 0$, where $p_1 + p_2 + \cdots + p_N = 1$. Let η_m be the number of balls in the mth cell after allocation of all balls, where $\eta(\eta_1, ..., \eta_N)$.

In specific applications, instead of cells, one has types or species of sampling units, and the sample array η is of interest because it reveals population frequencies p_j . Such species sampling problems arise in ecology, also in database query optimization, where the sampling units maybe entries in columns of a database while the species consist of all of distinct values appearing in the column; in literature, where the sampling units may be words appearing in a given author's known works while the species consist of all words known to that author; in disclosure risk limitation, where the sampling units may be people or firms listed in a microdata file, without names or other overtly identifying information, while the types are unique combinations of values of variables with which the people or firms might be implicitly identified; and in many other areas.

The subject of our interest here is the following important in practice count statistic: the number of occupied i.e.,

$$K_N = \sum_{m=1}^{N} \mathbf{I} \{ \eta_m > 0 \},$$

where $\mathbf{I}(A) = 1$ if event A occurred, otherwise zero.

The occupancy counts statistic K_N has been given many names, such as the "profile" (in information theory) or the "fingerprint" (in theoretical computer science) of the probability distribution $\{p_m\}$. Our goal in this paper is to establish the best estimate of the remainder term when approximating the distribution function of K_N with a normal distribution.

2. Results and Proof.

We adopt the following notation: ξ_1, ξ_2, \ldots is a sequence of independent random variables, where ξ_m is a Poisson random variable with parameter $\lambda_m := np_m$, $\Phi(x)$ is standard normal distribution function,

$$\Lambda_N = \sum_{m=1}^{N} (1 - e^{-\lambda_m}), \quad \gamma_N = n^{-1} \sum_{m=1}^{N} \lambda_m e^{-\lambda_m},$$

$$\sigma_N^2 = \sum_{m=1}^N e^{-\lambda_m} (1 - e^{-\lambda_m}) - n\gamma_N^2,$$

$$g_m(\xi_m) = \mathbf{I}\{\xi_m > 0\} - (1 - e^{-\lambda_m}) - \gamma_N(\xi_m - \lambda_m)$$

We assume that

$$N \max_{m} p_m \le C_0 \tag{6}$$

where, hereafter, C_k denotes a positive universal constant, which may vary across instances.

Theorem. There exists a constant C > 0 such that

$$\Delta_N := \sup_{-\infty < x < \infty} |P\{K_N < x\sigma_N + \Lambda_N\} - \Phi(x)| \le \frac{C}{\sigma_N}.$$

Proof. Write

$$\sigma_N^2 = \sum_{m=1}^N e^{-\lambda_m} \left(1 - (1 + \lambda_m) e^{-\lambda_m} \right) + \sum_{m=1}^N \lambda_m (\gamma_N - e^{-\lambda_m})^2,$$

$$g_m(\xi_m) = -(\mathbf{I}\{\xi_m = 0\} + \xi_m e^{-\lambda_m}) + (1 + \lambda_m) e^{-\lambda_m} - (\gamma_N - e^{-\lambda_m})(\xi_m - \lambda_m).$$

= -[(\mathbf{I}\{\xi_m = 0\} + \xi_m e^{-\lambda_m} - (1 + \lambda_m) e^{-\lambda_m}) + (\gamma_N - e^{-\lambda_m})(\xi_m - \lambda_m)].

We have for arbitrary $\nu \geq 3$,

$$\sum_{m=1}^{N} E|g_{m}(\xi_{m})|^{\nu} \leq 2^{\nu-1} \sum_{m=1}^{N} E|\mathbf{I}\{\xi_{m}=0\} + \xi_{m}e^{-\lambda_{m}} - (1+\lambda_{m})e^{-\lambda_{m}}|^{\nu}$$

$$+2^{\nu-1} \sum_{m=1}^{N} |\gamma_{N} - e^{-\lambda_{m}}|^{\nu} E|\xi_{m} - \lambda_{m})|^{\nu} =: J_{1} + J_{2}.$$

$$(7)$$

Next,

$$J_1 = 2^{\nu - 1} \sum_{m=1}^{N} \left((1 - (1 + \lambda_m)e^{-\lambda_m})^{\nu} \right) e^{-\lambda_m} + 2^{\nu - 1} \sum_{m=1}^{N} \lambda_m^{\nu} e^{-\nu \lambda_m}$$

$$+2^{\nu-1} \sum_{m=1}^{N} \sum_{j=2}^{\infty} |j-1-\lambda_m|^{\nu} e^{-(\nu+1)\lambda_m} \frac{\lambda_m^j}{j!} =: J_{11} + J_{12} + J_{13}.$$
 (8)

We have

$$J_{11} \le 2^{\nu - 1} \sum_{m=1}^{N} \left(1 - (1 + \lambda_m) e^{-\lambda_m} \right) e^{-\lambda_m} \le 2^{\nu - 1} \sigma_N^2.$$
 (9)

By using inequalities $xe^{-x} \le e^{-1}$ and $2^{-1}x^2e^{-x} \le 1 - (1+x)e^{-x}$ we obtain

$$J_{12} \le 2^{\nu} e^{-(\nu-2)} \sum_{m=1}^{N} \frac{1}{2} \lambda_m^2 e^{-2\lambda_m} \le 2^{\nu} e^{-(\nu-2)} \sum_{m=1}^{N} \left(1 - (1+\lambda_m)e^{-\lambda_m} \right) e^{-\lambda_m} \le 2^{\nu} e^{-(\nu-2)} \sigma_N^2. \tag{10}$$

Assume $\nu \geq 4$ is an even integer. We have

$$2^{1-\nu}J_{13} = \sum_{m=1}^{N} \sum_{j=2}^{\infty} |j-1-\lambda_m|^{\nu} e^{-(\nu+1)\lambda_m} \frac{\lambda_m^j}{j!} = \sum_{m=1}^{N} e^{-\nu\lambda_m} \sum_{j=2}^{\infty} (j-1-\lambda_m)^{\nu} e^{-\lambda_m} \frac{\lambda_m^j}{j!}$$

$$= \sum_{m=1}^{N} e^{-\nu \lambda_m} \left[\sum_{j=0}^{\infty} (j - 1 - \lambda_m)^{\nu} e^{-\lambda_m} \frac{\lambda_m^j}{j!} - (1 + \lambda_m)^{\nu} e^{-\lambda_m} - \lambda_m^{\nu+1} e^{-\lambda_m} \right]$$

$$= \sum_{m=1}^{N} e^{-\nu \lambda_m} \left[E(\xi_m - \lambda_m - 1)^{\nu} - (1 + \lambda_m)^{\nu} e^{-\lambda_m} - \lambda_m^{\nu+1} e^{-\lambda_m} \right]$$

$$= \sum_{m=1}^{N} e^{-(\nu-1)\lambda_m} \left[E(\xi_m - \lambda_m - 1)^{\nu} e^{-\lambda_m} - (1 + \lambda_m)^{\nu} e^{-2\lambda_m} - \lambda_m^{\nu+1} e^{-2\lambda_m} \right]$$

$$\leq \sum_{m=1}^{N} e^{-(\nu-1)\lambda_m} \left[\sum_{k=0}^{\nu} (-1)^{\nu-k} C_{\nu}^k E(\xi_m - \lambda_m)^k e^{-\lambda_m} - (1+\lambda_m)^{\nu-1} e^{-\lambda_m} + (1+\lambda_m)^{\nu-1} e^{-\lambda} (1-(1+\lambda)e^{-\lambda}) \right] \\
= \sum_{m=1}^{N} e^{-(\nu-1)\lambda_m} \left[\sum_{k=0}^{\nu} (-1)^{\nu-k} C_{\nu}^k E(\xi_m - \lambda_m)^k e^{-\lambda_m} - \sum_{k=0}^{\nu-1} C_{\nu-1}^k \lambda_m^k e^{-\lambda_m} \right] \\
+ \sum_{m=1}^{N} \left[(1+\lambda_m) e^{-\lambda_m} \right]^{\nu-1} e^{-\lambda} (1-(1+\lambda)e^{-\lambda}) = : J'_{13} + J''_{13}. \tag{11}$$

We have,

$$J'_{13} = \sum_{m=1}^{N} e^{-\nu\lambda_m} \left[\sum_{k=0}^{\nu} (-1)^{\nu-k} C_{\nu}^k E(\xi_m - \lambda_m)^k - \sum_{k=0}^{\nu-1} C_{\nu-1}^k \lambda_m^k \right]$$

$$= \sum_{m=1}^{N} e^{-\nu\lambda_m} \left[-\frac{1}{6} (\nu - 1)(\nu - 2)(\nu - 3) \lambda_m + \sum_{k=4}^{\nu} (-1)^{\nu-k} C_{\nu}^k E(\xi_m - \lambda_m)^k - \sum_{k=2}^{\nu-1} C_{\nu-1}^k \lambda_m^k \right].$$

It is known that $E(\xi_m - \lambda_m)^k = \lambda_m + \sum_{l=2}^{\lfloor k/2 \rfloor} c(l) \lambda_m^l$. Furthermore, $\sum_{k=0}^3 (-1)^{\nu-k} C_{\nu}^k - (\nu-1) = (\nu-1)(\nu-2)(\nu-3)/6$. Applying these facts at the last formula we obtain

$$J'_{13} \le c(\nu) \sum_{m=1}^{N} e^{-\nu \lambda_m} (\lambda_m^{\nu/2} + \lambda_m^2) \le c_1(\nu) \sum_{m=1}^{N} \lambda_m^2 e^{-2\lambda_m} \le c_2(\nu) \sigma^2(n)$$

since $2^{-1}x^2e^{-x} \le 1 - (1+x)e^{-x}$. Also $J_{13}'' \le c(\nu)\sigma^2(n)$ since $(1+\lambda_m)e^{-\lambda_m} \le 1$. So, by (6) $J_{13} \le c_3(\nu)\sigma^2(n)$. This together with (4), (5) imply

$$J_1 \le c_3(\nu)\sigma_N^2. \tag{12}$$

Under condition (1) by reasoning similar to those of [2] one can show that $J_2 \leq c(\nu)\sigma^2(n)$. Thus by (2), (7)

$$\sum_{m=1}^{N} E|g_m(\xi_m)|^{\nu} \le c(\nu)\sigma_N^2.$$

On using this inequality with $\nu=4$ and well-known inequality between Liyapunov's ratio we obtain

$$\frac{1}{\sigma_N^3} \sum_{m=1}^N E|g_m(\xi_m)|^3 \le \left[\frac{1}{\sigma_N^4} \sum_{m=1}^N E|g_m(\xi_m)|^4 \right]^{1/2} \le \frac{c}{\sigma_N}. \tag{13}$$

On the other hand it follows from Theorem 1 of [1] that

$$\Delta_N \le c \frac{1}{\sigma_N^3} \sum_{m=1}^N E|g_m(\xi_m)|^3.$$

Applying here (8) we complete the proof of Theorem.

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REZYUME

Ushbu maqolada zarrachalarni qutichalarga tasodifiy joylashtirishning multinomial modelida band bo'lgan qutichalar uchun markaziy limit teoremasidagi qoldiq hadga eng yaxshi baho olingan.

Kalit soʻzlar: tasodifiy joylash, Puasson taqsimoti, normal taqsimot, markaziy limit teorema.

РЕЗЮМЕ

В работе устанавливается наилучшая оценка остаточного члена в центральной предельной теореме для числа занятых ячеек в полиномиальной схеме случайного размещения.

Ключевые слова: случайное размещение, распределение Пуассона, нормальное распределение, центральная предельная теорема.