

A New Method for Computing the Expected Hitting Time between Arbitrary Different Configurations of the Multiple-Urn Ehrenfest Model

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Abstract. We study a multiple-urn version of the Ehrenfest model. In this setting, we denote the n urns by Urn 1 to Urn n , where $n \geq 2$. Initially, M balls are randomly placed in the n urns. At each subsequent step, a ball is selected and put into the other $n-1$ urns with equal probability. The expected hitting time leading to a change of the M balls' status is computed using the method of stopping times. As a corollary, we obtain the expected hitting time of moving all the M balls from Urn 1 to Urn 2.

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1 Introduction

We extend the classical two-urn Ehrenfest model to the multiple-urn case. Label the n urns by Urn 1 to Urn n , where $n \geq 2$. At the beginning, M balls are arbitrarily placed in the n urns. Then at each time, one ball is chosen at random, taken from the current urn it resides in, and placed in one of the other $n-1$ urns with equal probability. This model can be treated as a symmetric simple random walk on the graph $G_M = (V_M, E_M)$, where $V_M = \{1, \dots, n\}^M$, and E contains edges connecting two vertices in V_M if exactly one of their components differs. Here the subscript “ M ” is to stress that the number of balls is M . Therefore, G_M is a transitive graph (that is, for any $e, e' \in E_M$, there is an automorphism of the graph that takes e to e') with n^M vertices, and each vertex has common degree $(n-1)M$. Strictly speaking, if we let $X_t = (X_t^{(1)}, \dots, X_t^{(M)})$ be the state at time $t = 0, 1, \dots$, where $X_t^{(i)}$ is the number of the urn in which the i th ball resides at

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time t , then $\{X_t : t=0,1,\dots\}$ is a time homogeneous Markov chain on V_M with transition probability

$$p_{(x_1,\dots,x_M),(y_1,\dots,y_M)} = \begin{cases} \frac{1}{(n-1)^M}, & \text{if there exists } i \text{ s.t. } x_i \neq y_i, \text{ and } x_j = y_j \text{ for } j \neq i; \\ 0, & \text{otherwise.} \end{cases} \tag{1.1}$$

For $x_1, \dots, x_M \in \{1, 2, \dots, n\}$, denote by

$$T_{(x_1,\dots,x_M)} = \inf\{t \geq 0 : X_t = (x_1, \dots, x_M)\}$$

the first time that $\{X_t\}$ hits state (x_1, \dots, x_M) . Our main result is described in the following theorem.

Theorem 1.1. For any two different configurations $(a_1, \dots, a_M), (b_1, \dots, b_M) \in \{1, \dots, n\}^M$, denote $L = \sum_{i=1}^M \mathbf{1}_{\{a_i=b_i\}}$. Then

$$\mathbb{E}(T_{(b_1,\dots,b_M)} \mid X_0 = (a_1, \dots, a_M)) = \sum_{k=L}^{M-1} \frac{(n-1)^{k+1}}{\binom{M-1}{k}} \sum_{i=0}^k \frac{\binom{M}{i}}{(n-1)^i},$$

where $\mathbf{1}_{\{\cdot\}}$ denotes the indicator function, and $\binom{n}{m} := \frac{n!}{m!(n-m)!}$ ($0 \leq m \leq n$) denotes the combinatorial number.

Note that when $L = M - 1$, the righthand side of Theorem 1.1 becomes

$$(n-1)^M \sum_{i=0}^{M-1} \frac{\binom{M}{i}}{(n-1)^i},$$

which equals to $n^M - 1$. This is a well known result in Markov chains.

As a special case, we obtain the following corollary, which provides the expected hitting time of moving all balls from Urn 1 to Urn 2.

Corollary 1.2. We have

$$\mathbb{E} \left(T_{\underbrace{(2,2,\dots,2)}_M} \mid X_0 = \underbrace{(1,1,\dots,1)}_M \right) = \frac{(n-1)M}{n} \sum_{k=1}^M \frac{n^k}{k}.$$

Remark 1.1. (1) Chen *et al.* [3] proved Corollary 1.2 for the special case $n=3$ by using the method of electric networks. They conjectured that the result for general multiple-urn case should be of the form as stated in Corollary 1.2.

(2) Corollary 1.2 is a special case of Theorem 1.1 by letting $L=0$. This is not a straightforward result. A key step is to establish the equality

$$\sum_{k=0}^{M-1} \frac{(n-1)^k}{M \binom{M-1}{k}} \sum_{i=0}^k \frac{\binom{M}{i}}{(n-1)^i} = \frac{1}{n} \sum_{k=1}^M \frac{n^k}{k}. \tag{1.2}$$