



An analysis of the last round matching problem

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Abstract

The probability distribution of the number of players in the last round of a matching problem is analyzed and the existence of the limiting distribution is proved by using convolution method.

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1. Introduction

The matching problem or the “Hats Problem” goes back to at least 1713 when it was proposed by the French mathematician Pierre de Montmort in his book [9] on games of gambling and chance, *Essay d’Analyse sur les Jeux de Hazard* (see also [11]). Although it has many formulations, the most common one is as follows:

Suppose that each of n men at a party throws his hat into the center of the room. The hats are first mixed up, and then each man randomly selects a hat. What is the probability that exactly k of the men select their own hats? And what is the expected number of people that select their own hats?

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There are several approaches such as the inclusion–exclusion principle, recurrence relations, binomial inversion, to find the distribution of the so called Montmort random variable X_n , the number of matches with n men. These approaches can be found, for example, in [10, pp. 125–127] and [4]. See more details in the next section. A celebrated paper of Kaplansky [7] first showed that X_n has an approximate Poisson distribution. Takács [12] gave an extensive history for this single round version of the matching problem. The problem of a multi-round matching is as follows:

Suppose that those choosing their own hats depart, while the others put their selected hats in the center of the room, mixed them up, and then reselect. Also, suppose that this process continues until each individual has his own hat. Assume we start with n men. What is the expected number τ_n of rounds that are necessary?

Detailed analysis can be found in [10, p. 110], see also Section 2 for more detail. The purpose of this note is to analyze the following questions in the multi-round matching problem:

What is the expected number of men, L_n , on the last round? And what is the limiting distribution of L_n as $n \rightarrow \infty$?

Note that the last round is the round that all remaining players obtained their own hats. This requires us to determine the distribution of the stopped outcome and hence it can be viewed as conditioning on the future. There are many problems of interests that involve the moments (outcomes) of the last events at a stopping time, and they are in general hard problems even in determination of the expected values (see [13, p. 162], for a case of independent random walks).

We will show in Section 3 that the limiting distribution of L_n exists and the expectation $\mathbb{E}(L_n)$ converges to a constant $l \approx 2.26264703816$. The exact value of the constant l is still unknown. Let p_n be the probability that any particular individual man (among n men) is in the last round. Then we have $p_n = \mathbb{E}(L_n)/n$ by taking the expectation on the relation $L_n = \sum_{i=1}^n 1_{A_i}$, where A_i is the event that the i th man is in the last round. Thus $p_n = \mathbb{E}(L_n)/n \sim l/n$ as $n \rightarrow \infty$ is the asymptotic probability of winning (a winner is defined to be one in the last round), and the fair pay off for a winner is $1/p_n \sim n/l \approx n/2.262647$ unit if each man pays a unit to play.

The remaining of the paper is organized as follows. In Section 2, we provide mathematical formulation of the problem together with some remarkable properties of the Montmort random variable X_n , and we give the recursive relations for the expectation $\mathbb{E}(L_n)$ and estimate the distribution $q_{n,m} = \mathbb{P}(L_n = m)$. We prove our main result in Section 3 by using convolution method. In Section 4, we provide a table for the distribution of L_n for $2 \leq n \leq 15$ and numerical approximations of the limiting values. Related problems, questions and remarks are collected at the end of Section 4.

2. Recursive relations

Mathematically, we are dealing with random permutation of n elements and the Montmort random variable X_n is the number of fixed elements in a random permutation. Its distribution is well known and is given by

$$p_{n,k} = \mathbb{P}(X_n = k) = \frac{1}{k!} \sum_{i=0}^{n-k} (-1)^i \frac{1}{i!}, \quad k = 0, 1, \dots, n. \quad (2.1)$$

Feller [3, p. 231] showed that both the mean and the variance of X_n are equal to one for all $n \geq 2$. He also derived the Poisson approximation of rate $\lambda = 1$ for X_n . Some rather remarkable

properties of X_n were derived in [1], see also [5,6]. It is shown that the k th moment $\mathbb{E}(X_n^k)$ is the number of ways of putting k different things into at most $\min(k, n)$ indistinguishable cells, with no cell empty, i.e.,

$$\mathbb{E}(X_n^k) = \sum_{i=1}^{\min(k,n)} S(k, i)$$

where $S(k, i)$ is the Stirling number of the second kind, i.e., the number of ways of partitioning a set of k elements into i nonempty sets (i.e., set blocks). In particular, every moment of every X_n is a positive integer and $\mathbb{E}(X_n^k)$ are the same for all $k, 1 \leq k \leq n$, with n fixed. Also, it is easy to see that the k -factorial moment of X_n is one for $1 \leq k \leq n$ and zero for $k > n$, i.e.,

$$\mathbb{E}(X_n(X_n - 1) \cdots (X_n - k + 1)) = \begin{cases} 1 & \text{for } 1 \leq k \leq n, \\ 0 & \text{for } k > n. \end{cases}$$

More detailed and recent study on generalized matching, based on the Stein’s method of exchangeable pairs, is given in [2]. In particular, we know

$$\|\mathcal{L}(X_n) - \mathcal{L}(\text{Poisson}(1))\|_{\text{TV}} \leq \frac{2^n}{n!}$$

where $\mathcal{L}(\cdot)$ is the probability law for the random variable and $\|\cdot\|_{\text{TV}}$ is the total variation norm for measure. For the multi-round matching problem, the expected number of rounds τ_n is $\mathbb{E}(\tau_n) = n$, assuming we start with n men, see [10, p. 110].

Next we derive recursive relations for the expectation $\mathbb{E}(L_n)$ and the distribution $q_{n,m} = \mathbb{P}(L_n = m)$, where L_n is the number of people in the last round in the multi-round matching problem, assuming we start with n men. Let $l_n = \mathbb{E}(L_n)$ and X_n be the number of matchings in the first round. Then we have by using conditional expectation,

$$l_n = \sum_{k=0}^n \mathbb{E}(L_n | X_n = k) \cdot \mathbb{P}(X_n = k) = l_n p_{n,0} + n p_{n,n} + \sum_{k=1}^{n-1} l_{n-k} \cdot p_{n,k}$$

which can be rewritten as

$$l_n(1 - p_{n,0}) = n p_{n,n} + \sum_{k=1}^{n-1} l_k p_{n,n-k}.$$

From (2.1) we have the recursive relation

$$l_n \cdot \sum_{i=1}^n (-1)^{i+1} \frac{1}{i!} = \frac{1}{(n-1)!} + \sum_{k=1}^{n-1} \frac{l_k}{(n-k)!} \sum_{i=0}^k (-1)^i \frac{1}{i!} \tag{2.2}$$

with $l_0 = 0, l_1 = 1, l_2 = 2, l_3 = 9/4, l_4 = 34/15, l_5 = 43/19, l_6 = 3912/1729$. It is easy to show that l_n converges to $l = 2.2626470381671 \dots$ very quickly. See Section 4 for more discussion.

Next we turn to the distribution function of L_n . Let $q_{n,m} = \mathbb{P}(L_n = m), 2 \leq m \leq n$, be the distribution function of L_n . Then by conditioning on X_n , the number of matchings in the first round, we have for $2 \leq m < n$,

$$q_{n,m} = \sum_{k=0}^{n-m} \mathbb{P}(L_n = m | X_n = k) \cdot \mathbb{P}(X_n = k) = q_{n,m} \cdot p_{n,0} + \sum_{k=1}^{n-m} q_{n-k,m} \cdot p_{n,k}$$

and

$$\begin{aligned}
 q_{n,n} &= \mathbb{P}(L_n = n \mid X_n = 0) \cdot \mathbb{P}(X_n = 0) + \mathbb{P}(L_n = n \mid X_n = n) \cdot \mathbb{P}(X_n = n) \\
 &= q_{n,n} \cdot p_{n,0} + \frac{1}{n!}.
 \end{aligned}$$

Thus we obtain the recursive relation

$$\begin{aligned}
 q_{n,m} &= (1 - p_{n,0})^{-1} \sum_{k=m}^{n-1} q_{k,m} \cdot p_{n,n-k}, \quad 2 \leq m < n, \\
 q_{n,n} &= \frac{1}{(1 - p_{n,0})n!} = \frac{1}{n! \sum_{i=1}^n (-1)^{i-1} \frac{1}{i!}} \tag{2.3}
 \end{aligned}$$

with $q_{2,2} = 1, q_{3,2} = 3/4, q_{3,3} = 1/4, q_{4,2} = 4/5, q_{4,3} = 2/15, q_{4,4} = 1/15$.

Finally we can state our main results.

Theorem 2.1. *The random variables L_n converges in distribution, i.e.,*

$$\lim_{n \rightarrow \infty} q_{n,m} = q_m > 0, \quad m \geq 2,$$

and $\lim_{n \rightarrow \infty} \mathbb{E}(L_n) = l = \sum_{m=2}^{\infty} m q_m$.

3. Convolution limit

We first show the following lemma for the existence of a limit of convolutions.

Lemma 3.1. *Assume the sequence h_n converges to a finite real number h as $n \rightarrow \infty$ and the series $\sum_{i=1}^{\infty} |g_i|$ converges. Then the convolution sequence $\sum_{i=1}^n g_i h_{n-i}$ converges.*

Proof. For any given $\varepsilon > 0$, there exists a positive integer N such that for any $m > n \geq N$, $|h_n - h| \leq \varepsilon, |h_m - h_n| \leq \varepsilon, \sum_{i=n+1}^m |g_i| \leq \varepsilon$. Hence for any $m > n \geq 2N$,

$$\begin{aligned}
 &\left| \sum_{i=1}^m g_i h_{m-i} - \sum_{i=1}^n g_i h_{n-i} \right| \\
 &\leq \sum_{i=1}^{n-N} |g_i| \cdot |h_{m-i} - h_{n-i}| + \sum_{i=n-N+1}^n |g_i| \cdot |h_{m-i} - h_{n-i}| + \sum_{i=n+1}^m |g_i| \cdot |h_{m-i}| \\
 &\leq \varepsilon \cdot \sum_{i=1}^{n-N} |g_i| + 2M \sum_{i=n-N+1}^n |g_i| + M \sum_{i=n+1}^m |g_i| \leq \varepsilon \cdot \sum_{i=1}^{\infty} |g_i| + 3M\varepsilon \tag{3.1}
 \end{aligned}$$

where $M = \max_{i \geq 0} |h_i| < \infty$. \square

Proof of Theorem 2.1. Let l_n be recursively defined as in (2.2). We first show l_n converges. The goal is to rewrite l_n as a convolution. We start with writing our recursive equation (2.2) into a convolution form with error terms. Denote the tail of the Taylor series expansion of the function e^{-x} at $x = 1$ by

$$a_n = \sum_{i=n+1}^{\infty} (-1)^i \frac{1}{i!} = \int_0^1 \frac{(-1)^{n+1} (1-t)^n}{n!} e^{-t} dt.$$

Then for any integer $n \geq 1$,

$$|a_n| \leq \frac{1}{(n+1)!}. \tag{3.2}$$

Using the Taylor expansion $e^{-1} = \sum_{i=0}^{\infty} (-1)^i \frac{1}{i!}$, we can rewrite (2.2) as

$$l_n(1 - e^{-1} + a_n) = \frac{1}{(n-1)!} + \sum_{k=1}^{n-1} \frac{l_k}{(n-k)!} \cdot (e^{-1} - a_k)$$

which gives us

$$l_n = \frac{1}{e-1} \sum_{k=1}^{n-1} \frac{l_k}{(n-k)!} + g_n, \quad n \geq 2, \tag{3.3}$$

where

$$g_n = \frac{e}{e-1} \left(\frac{1}{(n-1)!} - \sum_{k=1}^n \frac{a_k l_k}{(n-k)!} \right).$$

Next we show $|g_n|$ is very small for large n . Obviously, we know $l_k \leq k$ for any k since the number of people on the last round cannot be more than the total number of people we started with. Using (3.2), it holds for any $n \geq 1$,

$$\begin{aligned} (1 - e^{-1})|g_n| &\leq \frac{1}{(n-1)!} + \sum_{k=1}^n \frac{k}{(k+1)!(n-k)!} \\ &= \frac{1}{(n-1)!} + \frac{1}{n!} \sum_{k=0}^n \frac{n!}{k!(n-k)!} - \frac{1}{(n+1)!} \sum_{k=0}^n \frac{(n+1)!}{(k+1)!(n-k)!} \\ &= \frac{1}{(n-1)!} + \frac{2^n}{n!} - \frac{2^{n+1} - 1}{(n+1)!}. \end{aligned} \tag{3.4}$$

Therefore, we have

$$\sum_{n=1}^{\infty} |g_n| < \infty. \tag{3.5}$$

To write l_n explicitly as a convolution, we use the generating function $\sum_{n=1}^{\infty} l_n x^n$ which converges for $|x| < 1$ due to the fact that $l_n \leq n$. From (3.3), we have

$$\begin{aligned} \sum_{n=1}^{\infty} l_n x^n &= \frac{1}{e-1} \sum_{n=1}^{\infty} x^n \sum_{k=1}^{n-1} \frac{l_k}{(n-k)!} + \sum_{n=1}^{\infty} g_n x^n \\ &= \frac{1}{e-1} \sum_{k=1}^{\infty} l_k \sum_{n=k+1}^{\infty} \frac{1}{(n-k)!} x^n + \sum_{n=1}^{\infty} g_n x^n \\ &= \frac{e^x - 1}{e-1} \sum_{k=1}^{\infty} l_k x^k + \sum_{n=1}^{\infty} g_n x^n. \end{aligned}$$

Solving for the generating function $\sum_{n=1}^{\infty} l_n x^n$ from above, we obtain

$$\begin{aligned} \sum_{n=1}^{\infty} l_n x^n &= \frac{1 - e^{-1}}{1 - e^{-x-1}} \sum_{n=1}^{\infty} g_n x^n = (1 - e^{-1}) \sum_{k=0}^{\infty} e^{k(x-1)} \sum_{j=1}^{\infty} g_j x^j \\ &= (1 - e^{-1}) \sum_{k=0}^{\infty} e^{-k} \sum_{i=0}^{\infty} \frac{x^i k^i}{i!} \sum_{j=1}^{\infty} g_j x^j \\ &= (1 - e^{-1}) \sum_{i=0}^{\infty} h_i x^i \sum_{j=1}^{\infty} g_j x^j = (1 - e^{-1}) \sum_{n=1}^{\infty} \sum_{i=1}^n g_i h_{n-i} x^n, \end{aligned}$$

where

$$h_n = \frac{1}{n!} \sum_{k=0}^{\infty} k^n e^{-k}. \tag{3.6}$$

Thus, we obtain

$$l_n = (1 - e^{-1}) \sum_{i=1}^n g_i h_{n-i}, \tag{3.7}$$

which is in a convolution form.

To show the convergence of the sequence h_n defined in (3.6), we note that $f(x) = x^n e^{-x}$ is strictly increasing in the interval $[0, n]$ and strictly decreasing in the interval $[n, \infty)$, for any integer $n \geq 1$. Thus using Stirling’s formula for $n^n e^{-n}$, we have for n large,

$$\begin{aligned} \sum_{k=0}^{\infty} k^n e^{-k} &= \sum_{k=0}^{n-1} k^n e^{-k} + n^n e^{-n} + \sum_{k=n+1}^{\infty} k^n e^{-k} \\ &\leq \int_0^{\infty} x^n e^{-x} dx + \frac{n!}{\sqrt{2n\pi}} = n!(1 + (2n\pi)^{-1/2}). \end{aligned} \tag{3.8}$$

Similarly, we have

$$\begin{aligned} \sum_{k=0}^{\infty} k^n e^{-k} &= \sum_{k=0}^n k^n e^{-k} - n^n e^{-n} + \sum_{k=n}^{\infty} k^n e^{-k} \\ &\geq \int_0^{\infty} x^n e^{-x} dx - \frac{n!}{\sqrt{2n\pi}} = n!(1 - (2n\pi)^{-1/2}). \end{aligned} \tag{3.9}$$

Combining the above estimates together, we see

$$\lim_{n \rightarrow \infty} h_n = 1. \tag{3.10}$$

And thus the limit of l_n exists by Lemma 3.1, (3.7), (3.5) and (3.10).

Next we follow the same idea as above to show the existence of the limit of $q_{n,m}$ as $n \rightarrow \infty$ for fixed $m \geq 2$. The recursive relation (2.3) can be rewritten as

$$q_{n,m} = (e - 1)^{-1} \sum_{k=m}^{n-1} \frac{q_{k,m}}{(n - k)!} + g_{n,m}, \quad 2 \leq m \leq n, \tag{3.11}$$

where

$$g_{n,m} = -(1 - e^{-1})^{-1} \sum_{k=m}^n \frac{a_k q_{k,m}}{(n - k)!}.$$

Using (3.2) and $q_{n,m} \leq 1$, it holds for $n \geq m \geq 2$,

$$(1 - e^{-1})|g_{n,m}| \leq \sum_{k=m}^n \frac{1}{(k + 1)!(n - k)!} \leq \sum_{k=0}^{n+1} \frac{1}{k!(n + 1 - k)!} = \frac{2^{n+1}}{(n + 1)!}.$$

Therefore, for any fixed $m \geq 2$, the series $\sum_{n=2}^{\infty} |g_{n,m}|$ converges. To write $q_{n,m}$ explicitly as a convolution for fixed $m \geq 2$, we use the generating function $\sum_{n=m}^{\infty} q_{n,m} x^n$ which converges for $|x| < 1$ since $q_{n,m} \leq 1$. From (3.11), we have

$$\begin{aligned} \sum_{n=m}^{\infty} q_{n,m} x^n &= \frac{1}{e - 1} \sum_{n=m}^{\infty} x^n \sum_{k=m}^{n-1} \frac{q_{k,m}}{(n - k)!} + \sum_{n=m}^{\infty} g_{n,m} x^n \\ &= \frac{1}{e - 1} \sum_{k=m}^{\infty} \sum_{n=k+1}^{\infty} \frac{q_{k,m}}{(n - k)!} x^n + \sum_{n=m}^{\infty} g_{n,m} x^n \\ &= \frac{e^x - 1}{e - 1} \sum_{k=m}^{\infty} q_{k,m} x^k + \sum_{n=m}^{\infty} g_{n,m} x^n. \end{aligned}$$

Solving for the generating function $\sum_{n=m}^{\infty} q_{n,m} x^n$ and then expanding it similar to the arguments used for l_n , we obtain

$$\sum_{n=m}^{\infty} q_{n,m} x^n = \frac{1 - e^{-1}}{1 - e^{x-1}} \sum_{n=m}^{\infty} g_{n,m} x^n = (1 - e^{-1}) \sum_{n=m}^{\infty} \sum_{i=m}^n g_{i,m} h_{n-i} x^n,$$

where h_i is defined in (3.6). Thus, we obtain for $n \geq m \geq 2$,

$$q_{n,m} = (1 - e^{-1}) \sum_{i=m}^n g_{i,m} h_{n-i}.$$

Therefore, similar to the argument for l_n , $q_{n,m}$ has a limit as $n \rightarrow \infty$ for each $m \geq 2$. \square

4. Remarks and related problems

Our argument of showing the existence of the limit can also provide an estimate of order $n^{-1/2}$ on the speed of convergence. We only consider the speed of $l_n = \mathbb{E}(L_n)$ to its limit l here since the other cases are similar.

Proposition 4.1. $|l_n - l| = O(n^{-1/2})$.

Proof. By taking $m \rightarrow \infty$ in the estimates (3.1) and using the basic facts (3.4), (3.8) and (3.9), we have

$$\begin{aligned}
 |l_n - l| &= \left| \sum_{i=1}^n g_i h_{n-i} - l \right| \\
 &\leq \sum_{i=1}^{n-N} |g_i| |h_{n-i} - 1| + \sum_{i=n-N+1}^n |g_i| |h_{n-i} - 1| + M \sum_{i=n+1}^{\infty} |g_i| \\
 &\leq \max_{N \leq k \leq n-1} |h_k - 1| \cdot \sum_{i=1}^{n-N} |g_i| + e^{-1} \sum_{i=n-N+1}^n |g_i| + (1 + e^{-1}) \sum_{i=n+1}^{\infty} |g_i| \\
 &\leq (2N\pi)^{-1/2} \cdot \sum_{i=1}^{\infty} \frac{2^i}{i!} + (1 + 2e^{-1}) \sum_{i=n-N+1}^{\infty} \frac{2^i}{i!} \\
 &\leq (e^2 - 1)(2N\pi)^{-1/2} + (1 + 2e^{-1}) \frac{2^{n-N}}{(n - N)!}
 \end{aligned}$$

where $M = \max_{i \geq 0} h_i \leq 1 + e^{-1}$. Taking $N = n - \log n$, we see that

$$|l_n - l| = O(n^{-1/2}). \quad \square$$

Furthermore, from the computation for l_n , see also Table 1 for $q_{n,m}$, we conjecture that the convergence is at geometric rate.

Next we mention a martingale associated with the problem. Let R_k be the number of men after the k th round. Then $R_0 = n$ and $\mathbb{E}(R_{k+1} | R_k) = R_k - 1$. Thus

$$M_k(n) = R_k - (n - k)$$

is a martingale. Let

$$\tau_n = \inf\{k: R_k = 0\}$$

be the number of rounds played. Then optimal stopping theorem implies $\mathbb{E}(M_{\tau}) = \mathbb{E}(M_0) = n$, i.e., $\mathbb{E}(\tau_n) = n$. Note that the martingale $M_k(n)$ is in triangular arrays. It is nature to consider a functional central limit theorem and the techniques used in Li and Pritchard [8] are helpful.

Table 1
Table of $q_{n,m}$

n	$m = 2$	3	4	5	6	7	8
2	1						
3	0.75	0.25					
4	0.8	0.133333	0.0666667				
5	0.802632	0.144737	0.0394737	0.0131579			
6	0.801966	0.145518	0.0426836	0.00763447	0.0021978		
7	0.801877	0.145377	0.0428913	0.00826109	0.00127964	0.000313873	
8	0.801888	0.145353	0.0428509	0.00830217	0.00138452	0.000182657	3.92357e-005
9	0.801890	0.145355	0.0428442	0.00829432	0.00139138	0.00019763	2.28344e-005
10	0.801890	0.145356	0.0428448	0.00829301	0.00139006	0.00019861	2.47063e-005
11	0.801890	0.145356	0.0428450	0.00829313	0.00138985	0.000198422	2.48288e-005
12	0.801890	0.145356	0.0428450	0.00829316	0.00138987	0.000198391	2.48053e-005
13	0.801890	0.145356	0.0428450	0.00829316	0.00138987	0.000198394	2.48014e-005
14	0.801890	0.145356	0.0428450	0.00829316	0.00138987	0.000198395	2.48017e-005
15	0.801890	0.145356	0.0428450	0.00829316	0.00138987	0.000198395	2.48018e-005

It would be also useful to find other martingales associated with the problem. Other problems along this line are central limit theorems for τ_n , upper and lower tail probabilities for τ_n , etc.

There are also many other generalizations of the one round matching, see Feller [3, p. 102] for several multi-round matchings. All of them can be generalized to the last round problem also and our convolution method seems apply to these problems.

Finally, we mention a sort of complementary problem, the so called *Christmas gift problem*. Consider the Christmas party game where each person from 1 to n brings a gift. Slips with numbers $1, 2, \dots, n$ are placed in a hat. Everyone randomly selects a slip at the same time and then receives the gift brought by the person corresponding to the number. However, if someone draws her/his own number, she/he puts the slip back into the hat, and redraw with those remain. Also, suppose that this process continues until each individual has a gift from someone else or there is only one slip in the hat. What is the probability Q_n that there is only one slip remains in the hat? Let E_n be the event that there is only one slip in the hat (the case that the game cannot finish). Then we have the following recursive formula:

$$Q_n = \mathbb{P}(E_n) = \sum_{k=0}^n \mathbb{P}(E_n \mid X_n = k) p_{n,k} = \sum_{k=1}^n Q_k p_{n,k} \tag{4.1}$$

with $Q_1 = 1, Q_2 = 0, Q_3 = 3/5, Q_4 = 6/23$. It is easy to see the limit

$$Q := \lim_{n \rightarrow \infty} Q_n = \lim_{n \rightarrow \infty} \sum_{k=1}^n \frac{Q_k}{k!} (e^{-1} - a_{n-k}) = e^{-1} \sum_{k=1}^{\infty} \frac{Q_k}{k!},$$

and the convergence rate

$$|Q_n - Q| \leq \frac{2^{n+1} + 1}{(n + 1)!}$$

since

$$\left| \sum_{k=1}^n \frac{Q_k a_{n-k}}{k!} \right| \leq \sum_{k=1}^n \frac{1}{(n - k + 1)! k!} \leq \frac{2^{n+1}}{(n + 1)!}.$$

Note also that Q_n in (4.1) satisfies the generating functions relation

$$\sum_{n=1}^{\infty} Q_n x^n = (1 - x)^{-1} e^{-x} \sum_{n=1}^{\infty} \frac{Q_n}{n!} x^n.$$

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