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▶ To cite this version:

Jakub Gismatullin, Patrick J C Tardivel. Beta distribution and associated Stirling numbers of the second kind. Probability and Mathematical Statistics, in Press. hal-04366123v2

HAL Id: hal-04366123 https://hal.science/hal-04366123v2

Submitted on 25 Jul 2024

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Beta distribution and associated Stirling numbers of the second kind

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Abstract

This article gives a formula for associated Stirling numbers of the second kind based on the moment of a sum of independent random variables having a beta distribution. From this formula we deduce, using probabilistic approaches, lower and upper bounds for these numbers.

1 Introduction

Classical Stirling numbers of the second kind S(p,m) counts the number of all partitions of $\{1,\ldots,p\}$ into m nonempty subsets, for $p\in\mathbb{N}_{>0}$ and $m\in\mathbb{N}$. More generally, the r-associated Stirling number $S_r(p,m)$, with $r\in\mathbb{N}_{>0}$, is the number of all partitions of $\{1,\ldots,p\}$, into m subsets where each subset contains at least r elements [3, p. 221]. Obviously $S(p,m)=S_1(p,m)$. Some sub-sequences of the multi-sequence $\{S_r(p,m):p,m,r\in\mathbb{N}_{>0},\ p\geq rm\}$ are in the On-Line Encyclopedia of Integer Sequences (OEIS) [11]. Specifically, arrays $\{S_1(p,m)\}_{\{p,m\}},\ \{S_2(p,m)\}_{\{p,m\}},\ \{S_3(p,m)\}_{\{p,m\}}$ are identified in the OEIS by A008277, A008299, A059022. Moreover, sequences $\{S_2(k+6,3)\}_{\{k\}},\ \{S_2(k+8,4)\}_{\{k\}}$, representing the number of ways of placing k+6 or k+8 labelled balls into 3 or 4 indistinguishable boxes with at least 2 balls in each box are identified in the OEIS by A000478, A058844.

There are well-known connections between Stirling numbers of the second kind and probability theory. For example, sequences $S_1(p, m)$ and $S_2(p, m)$ are asymptotically normal when p tends to $+\infty$ [6, 4]. More precisely, when $r \in \{1, 2\}$, the following convergence in distribution occurs

$$\frac{Y_p - \mathbb{E}(Y_p)}{\sqrt{\mathrm{var}(Y_p)}} \xrightarrow[p \to +\infty]{d} \mathcal{N}(0,1) \text{ where } \mathbb{P}(Y_p = m) = \frac{S_r(p,m)}{\sum_{k=1}^p S_r(p,k)}, \text{ for all } m \in \mathbb{N}_{>0}.$$

Otherwise, according to Dobiński's formula, the moment of order p of a Poisson distribution with parameter $\lambda \geq 0$ is $\sum_{m=1}^{p} S_1(p,m)\lambda^m$ (see, e.g. [3, p. 211]). However, to our knowledge, there is no close formula in the literature for $S_r(k,m)$ based on moments of a sum of independent and identically distributed (i.i.d) random variables. The main result in this article is Theorem 2.1 providing the following new identity

$$S_r(p,m) = \frac{p!}{m!(r!)^m(p-rm)!} \mathbb{E}\left[(X_1 + \dots + X_m)^{p-rm} \right]$$
 (1)

where X_1, \ldots, X_p are i.i.d random variables having beta distribution with parameter (1, r). Note that the beta distribution with parameter (1, 1) is a uniform distribution on [0, 1]. Thus, when r = 1, the above formula is quite simple:

$$S_1(p,m) = \binom{p}{m} \mathbb{E}\left(Z^{p-m}\right)$$

where $Z = \sum_{i=1}^m X_i$ has the Irwin-Hall distribution on [0,m]. Propositions 3.1, 3.2 and 3.3 give upper and lower bounds for $\mathbb{E}\left[(X_1+\cdots+X_m)^{p-rm}\right]$. These bounds are sharp when one parameter m, r, or p-rm tends to $+\infty$ and provides thus accurate approximations of r-associated Stirling numbers.

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2 Closed formula for Stirling numbers and moments of random variables

The density g_r of a beta distribution with parameters (1,r) where $r \in \mathbb{N}_{>0}$ is

$$g_r(x) = \begin{cases} r(1-x)^{r-1} & \text{if } x \in [0,1] \\ 0 & \text{otherwise} \end{cases}$$
 (2)

Let X_1, \ldots, X_m be independent random variables having the same beta (1, r) distribution. The moment of order $k \in \mathbb{N}$ of the sum of these variables is defined as follows

$$\mathcal{M}_r(k,m) = \mathbb{E}\left[\left(X_1 + \dots + X_m \right)^k \right]. \tag{3}$$

Theorem 2.1 provides a closed formula for the Stirling numbers of the second kind based on the moment $\mathcal{M}_r(k,m)$.

Theorem 2.1. Let $m, r \in \mathbb{N}_{>0}$ and $p \in \mathbb{N}$ where $p \geq rm$. The Stirling numbers of the second kind satisfy the following identity

$$S_r(p,m) = \frac{p!}{m!(r!)^m(p-rm)!} \mathcal{M}_r(p-rm,m).$$
 (4)

From Theorem 2.1, one may deduce that $\mathbb{E}(Z^k) = S_1(m+k,m)/\binom{m+k}{m}$ where Z has an Irwin-Hall's distribution on [0,m] [8, 7]. Note that the moment generating function of Z is $\sum_{k\geq 0} \mathbb{E}(Z^k)t^k/k! = ((\exp(t)-1)/t)^m$ therefore we recover the well-known exponential generating function of the Stirling numbers of the second kind $\sum_{p\geq m} S_1(p,m)t^p/p! = (\exp(t)-1)^m/m!$ (see Theorem 3.3 page 52 in [9]). The above expression of $S_r(p,m)$ is explicit up to the computation of the moment $\mathcal{M}_r(k,m)$. Whereas computing explicitly $\mathcal{M}_r(k,m)$ might be technical, lower bounds, upper bounds and approximations of $\mathcal{M}_r(k,m)$ are tractable as illustrated in the following section.

3 Upper and lower bounds

Hereafter, we will use probabilistic approaches to derive upper and lower bounds for the moment $\mathcal{M}_r(k,m)$.

3.1 Sharp upper and lower bounds when m is large

Let $\overline{X}_m = (X_1 + \cdots + X_m)/m$, the Jensen's inequality provides the following lower bound

$$\mathcal{M}_r(k,m) = m^k \mathbb{E}\left(\overline{X}_m^k\right) \ge m^k \mathbb{E}\left(\overline{X}_m\right)^k = \frac{m^k}{(r+1)^k}.$$
 (5)

This inequality relies on the linearization of the function $q(x) = x^k$ at $x_0 = \mathbb{E}(\overline{X}_m) = \frac{1}{r+1}$. Specifically, the following inequality holds for all $x \in [0,1]$

$$x^{k} = q(x) \ge q(x_0) + q'(x_0)(x - x_0) = \frac{1}{(r+1)^k} + \frac{k}{(r+1)^{k-1}} \left(x - \frac{1}{r+1} \right).$$
 (6)

Moreover one may choose $c \ge 0$ for which the following inequality is true for all $x \in [0,1]$ (see Lemma 5.2)

$$x^{k} \le q(x_0) + q'(x_0)(x - x_0) + c(x - x_0)^{2}.$$

$$(7)$$

Proposition 3.1 below is a consequence of inequalities (6) and (7).

Proposition 3.1. Let $r, m \in \mathbb{N}_{>0}$ and $k \in \mathbb{N}$. The following inequality holds

$$\frac{m^k}{(r+1)^k} \le \mathcal{M}_r(k,m) \le \frac{m^k}{(r+1)^k} + m^{k-1} \frac{(r+1)^k - 1 - kr}{(r+1)^k r(r+2)}.$$
 (8)

The leading term when m is large in both the lower and upper bounds is $m^k/(r+1)^k$. Therefore, lower and upper bounds are asymptotically equivalent when $k \in \mathbb{N}_{\geq 0}$ and $r \in \mathbb{N}_{> 0}$ are fixed and m tends to $+\infty$. These bounds are accurate when m is large since \overline{X}_m converges to $\mathbb{E}(\overline{X}_m)$ and both inequalities (6) and (7) are accurate on the neighbourhood of $\mathbb{E}(\overline{X}_m)$.

3.2 Sharp upper and lower bounds when k is large

Asymptotic behaviour of moments, when k is large, depends on the density of $X_1 + \cdots + X_m$ on the tail, *i.e.* on the neighbourhood of m. This motivates us to introduce the following inequality proved in Corollary 5.4:

$$g_r^{*m}(x) \le \frac{(r!)^m}{(mr-1)!} (m-x)^{mr-1}, \text{ for all } x \in [0, m]$$
 (9)

where g_r is given by (2) and g_r^{*m} is m-th convolution of g_r . Moreover, this inequality is an equality for $x \in [m-1, m]$. We derive from this fact a lower and upper bounds for $\mathcal{M}_r(k, m)$ given in Proposition 3.2 below.

Proposition 3.2. For any $r \in \mathbb{N}_{>0}$, for any $k \in \mathbb{N}$ for any $m \in \mathbb{N}_{>0}$ the following inequalities hold.

$$\mathcal{M}_{r}(k,m) \leq \frac{(r!)^{m}}{(mr-1)!} \int_{0}^{m} x^{k} (m-x)^{mr-1} dx = \frac{k!(r!)^{m} m^{k+rm}}{(k+mr)!}.$$

$$\mathcal{M}_{r}(k,m) \geq \frac{(r!)^{m}}{(mr-1)!} \int_{m-1}^{m} x^{k} (m-x)^{mr-1} dx$$

$$\geq \frac{k!(r!)^{m} m^{k+mr}}{(k+mr)!} \left(1 - \frac{(m-1)^{k}}{m^{k+mr}} \sum_{i=1}^{mr} {k+mr \choose k+i} (m-1)^{i} \right).$$

These bounds are accurate when k is large since

$$\lim_{k \to +\infty} \frac{(m-1)^k}{m^{k+mr}} \sum_{i=1}^{mr} {k+mr \choose k+i} (m-1)^i = 0.$$

As a consequence of Proposition 3.2, we observe that $S_r(p,m) \leq m^p/m!$. Moreover, lower and upper bounds are asymptotically equivalent when k tends to $+\infty$ therefore $S_r(p,m) \sim m^p/m!$ when p is large. This approximation, well known when r = 1 (see [2]), remains true when r > 1.

3.3 Sharp upper bound when r is large

Proposition 3.3 proves that the moment $\mathcal{M}_r(k,m)$ is bounded, up to an explicit expression, by the moment of a sum of independent random variables having the same standard exponential distribution.

Proposition 3.3. Let $k \in \mathbb{N}$, $m \in \mathbb{N}_{>0}$, $r \in \mathbb{N}_{>1}$ and $\mathcal{E}_1, \ldots, \mathcal{E}_m$ be i.i.d random variables having standard exponential distribution with density $\exp(-x)$.

i) The following inequality holds

$$r^k \mathcal{M}_r(k,m) \le \left(\frac{r}{r-1}\right)^{2k} \mathbb{E} \left(\mathcal{E}_1 + \dots + \mathcal{E}_m\right)^k = \left(\frac{r}{r-1}\right)^{2k} \frac{(m-1+k)!}{(m-1)!}.$$
 (10)

ii) The upper bound given in i) is sharp since the following limit holds

$$\lim_{r \to +\infty} r^k \mathcal{M}_r(k, m) = \mathbb{E} \left(\mathcal{E}_1 + \dots + \mathcal{E}_m \right)^k = \frac{(m - 1 + k)!}{(m - 1)!}.$$
 (11)

It seems difficult for authors to find lower bounds which are sharp when r tends to $+\infty$. Finally, we recap hereafter lower and upper bounds for r-associated Stirling number of the second kind:

 $\bullet\,$ Proposition 3.1 provides the following lower and upper bounds

$$\begin{cases}
S_r(p,m) \ge \frac{p!m^{p-rm}}{m!(r!)^m(p-rm)!(r+1)^{p-rm}} \\
S_r(p,m) \le \frac{p!m^{p-rm}}{m!(r!)^m(p-rm)!(r+1)^{p-rm}} \left(1 + \frac{(r+1)^{p-rm} - 1 - r(p-rm)}{mr(r+2)}\right).
\end{cases}$$
(12)

These bound are equivalent when p-rm, r are fixed and when m tends to $+\infty$.

• Proposition 3.2 provides the following lower and upper bounds

$$\begin{cases} S_r(p,m) \ge \frac{m^p}{m!} - \frac{(m-1)^{p-rm}}{m!} \sum_{i=1}^{mr} {p \choose p-rm+i} (m-1)^i \\ S_r(p,m) \le \frac{m^p}{m!}. \end{cases}$$
(13)

These bound are equivalent when m, r are fixed and when p tends to $+\infty$.

• Proposition 3.3 provides the following upper bound when $r \geq 2$

$$S_r(p,m) \le \frac{p! r^{2(p-rm)} (m-1+p-rm)!}{m! (r!)^m (p-rm)! (r-1)^{2(p-rm)} (m-1)!}.$$
(14)

This upper bound is equivalent to $S_r(p,m)$ when p-rm,m are fixed and when r tends to $+\infty$.

4 Numerical experiments

4.1 Upper and lower bounds of Stirling numbers of the second kind

According to Propositions 3.1 and 3.2, for all $m \in \mathbb{N}$ and all $p \in \mathbb{N}_{>0}$, Stirling numbers of the second kind do satisfy the following inequalities:

$$S_{1}(p,m) \leq \min \left\{ \frac{m^{p}}{m!}, \binom{p}{m} \left(\frac{m}{2} \right)^{p-m} \left(1 + \frac{2^{p-m} + m - p - 1}{3m} \right) \right\}$$

$$U(p,m)$$

$$S_{1}(p,m) \geq \max \left\{ \frac{m^{p}}{m!} - \frac{(m-1)^{p-m}}{m!} \sum_{i=1}^{m} \binom{p}{m-i} (m-1)^{i}, \binom{p}{m} \left(\frac{m}{2} \right)^{p-m} \right\}.$$

$$L(p,m)$$

First of all we are going to compare these bounds U(m,p) and L(p,m) to bounds given in Rennie and Dobson [10] reported below:

$$\underbrace{\frac{1}{2}(m^2 + m + 2)m^{p - m - 1} - 1}_{L_{\mathrm{rd}}(p, m)} \le S_1(p, m) \le \underbrace{\frac{1}{2}\binom{p}{m}m^{p - m}}_{U_{\mathrm{rd}}(p, m)}.$$
 (15)

Numerical comparison between U(p, m) and $U_{\rm rd}(p, m)$ is not relevant since

$$\frac{1}{2} \binom{p}{m} m^{p-m} \ge \binom{p}{m} \left(\frac{m}{2}\right)^{p-m} \left(1 + \frac{2^{p-m} + m - p - 1}{3m}\right)$$

holds for m < p. Contrarily to the upper bound, the lower bound L(p, m) is not uniformly larger than the one given by Rennie and Dobson; for instance, $31 = L_{\rm rd}(6,2) > L(6,2) = 28.5$. Numerical experiments in Figure 1 illustrate that for most integers p, m, L(p, m) is a better approximation of $S_1(p, m)$ than $L_{\rm rd}(p, m)$.

Figure 2 provides a comparison between L(p, m), U(p, m) and $S_1(p, m)$.

4.2 Upper bounds of Bell numbers

The Bell number B(p), where $p \in \mathbb{N}_{>0}$, represents the number of partition of $\{1,\ldots,p\}$. Since a Bell number is a sum of Stirling numbers of the second kind $B(p) = \sum_{m=1}^{p} S_1(p,m)$ then, the following inequality occurs:

$$B(p) \le \underbrace{\sum_{m=0}^{p} \min\left\{\frac{m^{p}}{m!}, \binom{p}{m} \left(\frac{m}{2}\right)^{p-m} \left(1 + \frac{2^{p-m} + m - p - 1}{3m}\right)\right\}}_{=U(p)}$$

$$(16)$$

for all $p \in \mathbb{N}_{>0}$. In Figure 3 we compare U(p) with the upper bound: $B(p) \leq U_{\rm bt}(p) = \left(\frac{0.792p}{\ln(p+1)}\right)^p$ given in Berend and Tassa [1].

Note that $U(p) \leq \sum_{m=0}^{+\infty} m^p/m! = eB(p)$ (the last equality is due to the Dobiński formula). In Figure 4 we show that U(p)/B(p) is very close to e when p is large.

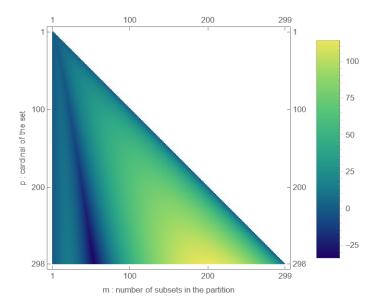


Figure 1: This figure report $\ln(L(p,m)) - \ln(L_{\rm rd}(p,m))$ as a function of m (on the x-axis) and p (on the y-axis). One may observe that for most integers the lower bound L(p,m) is a better approximation of $S_1(p,m)$ than $L_{\rm rd}(p,m)$ (as $\ln(L(p,m)) - \ln(L_{\rm rd}(p,m)) > 0$).

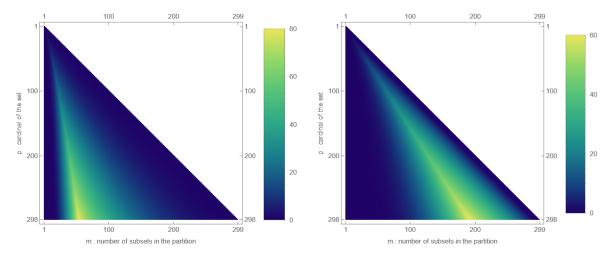


Figure 2: This figure report $\ln(S_1(p,m)) - \ln(L(p,m))$ (on the left) and $\ln(U(p,m)) - \ln(S_1(p,m))$ (on the right) as a function of m and p. These numerical experiments comply with Propositions 3.1 and 3.2 since both lower and upper bounds accurately approximate $S_1(p,m)$ when p is large and m is small or when m is large and p-m is small.

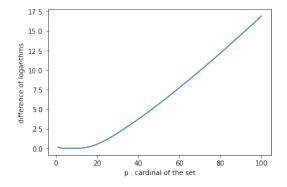


Figure 3: This figure report $\ln(U_{bt}(p)) - \ln(U(p))$ as a function of p. One may observe that when $p \ge 13$, U(p) is more accurate upper bound for $S_1(p,m)$ than $U_{bt}(p)$ (as $\ln(U_{bt}(p)) - \ln(U(p)) > 0$ for $p \ge 13$).

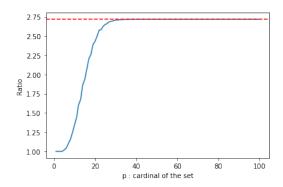


Figure 4: This figure report U(p)/B(p) as a function of p. One may observe that U(p)/B(p) is approximately equal to e when p is large.

5 **Proofs**

5.1Proof of Theorem 2.1

The identity given in Lemma 5.1 below, combined with the multinomial formula, allows us to complete the proof of Theorem 2.1.

Lemma 5.1. Let $m,r \in \mathbb{N}_{>0}$ and $p \in \mathbb{N}$ where $p \geq rm$. The r-associated Stirling numbers of the second kind satisfy the following equality

$$S_r(p,m) = \frac{p!}{m!} \sum_{i_1 + \dots + i_m = p - rm} \frac{1}{(r + i_1)! \times \dots \times (r + i_m)!}$$

where the sum is computed over all the integers $i_1, \ldots, i_m \in \{0, \ldots, p-rm\}$ satisfying $i_1 + \cdots + i_m = p-rm$.

Proof. Given $i_1, \ldots, i_m \in \mathbb{N}$ such that $i_1 + \cdots + i_m = p - rm$, let us count the number of ordered partitions of

 $\{1,\ldots,p\}$ in m parts where the first part has $r+i_1$ elements, the second part has $r+i_2$ elements and so on. There are $\binom{p}{r+i_1}$ possibilities for the first part. There are $\binom{p-r-i_1}{r+i_2}$ possibilities for the second part and so on. Therefore the number of ordered partitions where the first part has $r+i_1$ elements, the second part has $r + i_2$ elements and so on is

$$\frac{p!}{(r+i_1)! \times \ldots \times (r+i_m)!}.$$

Consequently, the number of ordered partitions of $\{1, \ldots, p\}$ in m parts having at least r elements is

$$\sum_{i_1+\dots+i_m=p-rm} \frac{p!}{(r+i_1)! \times \dots \times (r+i_m)!}.$$

Finally, when the order is not taken into account, by dividing by m!, one may deduce that

$$S_r(p,m) = \frac{p!}{m!} \sum_{i_1 + \dots + i_m = p - rm} \frac{1}{(r+i_1)! \times \dots \times (r+i_m)!}.$$

Proof. Theorem 2.1Let us recall the multinomial formula. Given $x_1, \ldots, x_m \in \mathbb{R}$ and $k \in \mathbb{N}$, we have

$$(x_1 + \dots + x_m)^k = \sum_{i_1 + \dots + i_m = k} \frac{k!}{i_1! \times \dots \times i_m!} x_1^{i_1} \times \dots \times x_m^{i_m}.$$

6

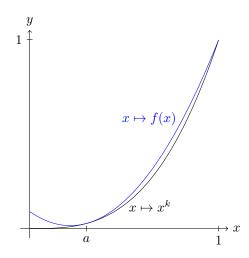


Figure 5: Illustration of the inequality given in Lemma 5.2.

Let k=p-rm. Since $\mathbb{E}(X_1^s)=\frac{s!r!}{(s+r)!}$, the multinomial formula and Lemma 5.1 give

$$\mathbb{E}\left[(X_{1} + \dots + X_{m})^{k} \right] = \sum_{i_{1} + \dots + i_{m} = k} \frac{k!}{i_{1}! \times \dots \times i_{m}!} \mathbb{E}(X_{1}^{i_{1}}) \times \dots \times \mathbb{E}(X_{m}^{i_{m}})$$

$$= (r!)^{m} k! \sum_{i_{1} + \dots + i_{m} = k} \frac{1}{(r + i_{1})! \times \dots \times (r + i_{m})!}$$

$$= \frac{m!(r!)^{m} k!}{(k + rm)!} S_{r}(k + rm, m)$$

$$= \frac{m!(r!)^{m} (p - rm)!}{p!} S_{r}(p, m)$$

which finishes the proof.

5.2 Proof of Proposition 3.1

Proposition 3.1 is a consequence of the following lemma.

Lemma 5.2. Let $k \geq 2$, $a \in (0,1)$ and $f: x \in [0,1] \rightarrow a^k + ka^{k-1}(x-a) + c(x-a)^2$ where $c \geq 0$ is such that f(1) = 1 (namely $c = \frac{a^{k-1}(ak-a-k)+1}{(1-a)^2}$). Then $f(x) \geq x^k$ for all $x \in [0,1]$.

Proof. First note that for all $x \in [0,1]$ the condition $f(x) \geq x^k$ is equivalent to

$$ka^{k-1}(x-a) + c(x-a)^2 \ge x^k - a^k = (x-a)(x^{k-1} + ax^{k-2} + \dots + a^{k-1}).$$

Note that this inequality holds if and only if

$$ka^{k-1} + c(x-a) \ge x^{k-1} + ax^{k-2} + \dots + a^{k-1}$$
 for all $x \in [a,1]$ and $ka^{k-1} + c(x-a) \le x^{k-1} + ax^{k-2} + \dots + a^{k-1}$ for all $x \in [0,a]$.

Let $d(x) = ka^{k-1} + c(x-a)$ and $p(x) = x^{k-1} + ax^{k-2} + \cdots + a^{k-1}$. Thus p(a) = d(a) and p(1) = d(1), by construction of c. Because p is convex and d is affine, one may deduce that $d(x) \leq p(x)$ once $x \in [0, a]$ and $d(x) \geq p(x)$ once $x \in [a, 1]$, which completes the proof.

Proof. Proposition 3.1 By Lemma 5.2, for $a = \mathbb{E}(\overline{X}_m) = \frac{1}{r+1}$, r > 0, for all $x \in [0,1]$ we get that

$$\begin{array}{ll} x^k & \leq & \mathbb{E}(\overline{X}_m)^k + k\mathbb{E}(\overline{X}_m)^{k-1}\left(x - \mathbb{E}(\overline{X}_m)\right) + c\left(x - \mathbb{E}(\overline{X}_m)\right)^2 \\ & \leq & \frac{1}{(r+1)^k} + \frac{k}{(r+1)^{k-1}}\left(x - \frac{1}{r+1}\right) + c\left(x - \frac{1}{r+1}\right)^2 \end{array}$$

where $c = (1 + \frac{1}{r})^2 \left(1 - \frac{kr+1}{(r+1)^k}\right)$. This inequality implies that

$$\begin{split} \mathcal{M}_r(k,m) &= m^k \mathbb{E}(\overline{X}_m^k) & \leq & m^k \left(\mathbb{E}(\overline{X}_m)^k + c \text{var}(\overline{X}_m) \right) \\ & \leq & \frac{m^k}{(1+r)^k} + c \frac{rm^{k-1}}{(1+r)^2(2+r)} \\ & \leq & \frac{m^k}{(r+1)^k} + \frac{(r+1)^k - 1 - kr}{(r+1)^k r(r+2)} m^{k-1}. \end{split}$$

5.3 Proof of Proposition 3.2

We use in the proof a well-known beta integral (see, for example, [5]): let $a, b \in \mathbb{N}$ and $x \in \mathbb{R}_{>0}$ then we have

$$\int_0^x (x-t)^a t^b dt = \frac{a!b!}{(a+b+1)!} x^{a+b+1}.$$
 (17)

To compute explicitly the density of $X_1 + \cdots + X_m$ on the tail [m-1, m], we use the following technical lemma. Let h_r be the following density

$$h_r(x) = \begin{cases} rx^{r-1} & \text{if } x \in [0,1] \\ 0 & \text{otherwise} \end{cases}$$
 (18)

Actually, h_r is the density of 1 - X where the density X is g_r (2). Convolution computations are slightly easier to handle with h_r than g_r .

Lemma 5.3. Let $m \in \mathbb{N}_{>0}$ then the following equality and inequality hold

$$h_r^{*m}(x) = \frac{(r!)^m}{(mr-1)!} x^{mr-1}, \text{ for all } x \in [0,1].$$
 (19)

$$h_r^{*m}(x) \le \frac{(r!)^m}{(mr-1)!} x^{mr-1}, \text{ for all } x \in \mathbb{R}_{\ge 0}.$$
 (20)

Proof. Let us prove (19) by induction. When m=1, one may notice that whatever $x\in [0,1]$ we have $h_r^{*m}(x)=h_r(x)$. Let $m\in\mathbb{N}_{>0}$ such that

$$h_r^{*m}(x) = \frac{(r!)^m}{(mr-1)!} x^{mr-1} \quad \forall x \in [0,1].$$

Therefore for $x \in [0,1]$ we have that

$$h_r^{*m+1}(x) = \int_{\mathbb{R}} h_r^{*m}(x-t)h_r(t)dt = \int_0^x h_r^{*m}(x-t)rt^{r-1}dt$$

$$= \frac{r(r!)^m}{(mr-1)!} \int_0^x (x-t)^{mr-1}t^{r-1}dt$$

$$= \frac{r(r!)^m}{(mr-1)!} \frac{(mr-1)!(r-1)!}{((m+1)r-1)!} x^{(m+1)r-1}$$

$$= \frac{(r!)^{m+1}}{((m+1)r-1)!} x^{(m+1)r-1}.$$

The proof of (20) by induction is quite similar than the proof of (19). When m = 1, the result is straightforward. Let $m \in \mathbb{N}_{>0}$ such that

$$h_r^{*m}(x) \le \frac{(r!)^m}{(mr-1)!} x^{mr-1} \quad \forall x \in \mathbb{R}_{\ge 0}.$$

Therefore for $x \in \mathbb{R}_{\geq 0}$ we have

$$h_r^{*m+1}(x) = \int_{\mathbb{R}} h_r^{*m}(x-t)h_r(t)dt = \int_0^x h_r^{*m}(x-t)h_r(t)dt$$

$$\leq \frac{r(r!)^m}{(mr-1)!} \int_0^x (x-t)^{mr-1}t^{r-1}dt$$

$$\leq \frac{r(r!)^m}{(mr-1)!} \frac{(mr-1)!(r-1)!}{((m+1)r-1)!} x^{(m+1)r-1}$$

$$\leq \frac{(r!)^{m+1}}{((m+1)r-1)!} x^{(m+1)r-1}.$$

Note that the only difference between the proof of (19) and (20) is the majorization of $h_r(t)$ by rt^{r-1} . We are now ready to prove the explicit formula for the m-th convolution g_r^{*m} on [m-1,m] and an upper bound on [0,m].

Corollary 5.4. For all $x \in [0, m]$ we have that

$$g_r^{*m}(x) \le \frac{(r!)^m}{(mr-1)!} (m-x)^{mr-1}.$$

Moreover if $x \in [m-1, m]$ then

$$g_r^{*m}(x) = \frac{(r!)^m}{(mr-1)!}(m-x)^{mr-1}.$$

Proof. By Lemma 5.3, it suffices to prove that $g_r^{*m}(x) = h_r^{*m}(m-x)$ for all $x \in [0, m]$. Let $x \in [0, m]$ and set z = m - x, because the density of $(1 - X_1) + \cdots + (1 - X_m)$ is h_r^{*m} then the following equalities occur

$$\frac{\partial}{\partial z} \mathbb{P}((1-X_1) + \dots + (1-X_m) \le z) = h_r^{*m}(z)$$

$$\frac{\partial}{\partial z} \mathbb{P}(X_1 + \dots + X_m \ge m-z) = h_r^{*m}(z)$$

$$g_r^{*m}(m-z) = h_r^{*m}(z)$$

$$g_r^{*m}(x) = h_r^{*m}(m-x).$$

One may notice that if r = 1 and $x \in [m-1, m]$, then $g_r^{*m}(x) = (m-x)^{m-1}/(m-1)!$ which is the density in the tail of the Irvin-Hall distribution (see [7, 8]).

Upper and lower bounds given in Proposition 3.2 are straightforward consequences of Corollary 5.4. Indeed, the upper bound is just the beta integral of

$$\int_0^m \frac{(r!)^m}{(mr-1)!} (m-x)^{mr-1} x^k dx.$$

Proof. Proposition 3.2 The lower bound is derived from the following computations

$$\frac{(r!)^m}{(mr-1)!} \int_0^{m-1} x^k (m-x)^{mr-1} dx$$

$$= \frac{(r!)^m}{(mr-1)!} \int_0^{m-1} x^k (1+(m-1)-x)^{mr-1} dx$$

$$= \frac{(r!)^m}{(mr-1)!} \int_0^{m-1} x^k \sum_{i=0}^{mr-1} {mr-1 \choose i} (m-1-x)^i dx$$

$$= \frac{(r!)^m}{(mr-1)!} \sum_{i=0}^{mr-1} {mr-1 \choose i} \frac{k!i!(m-1)^{k+i+1}}{(k+i+1)!}$$

$$= \frac{(r!)^m k!}{(k+mr)!} \sum_{i=1}^{mr} {k+mr \choose k+i} (m-1)^{k+i}.$$

□ **X**/o Because $\mathcal{M}_r(k,m) \geq \frac{(r!)^m}{(mr-1)!} \int_{m-1}^m x^k (m-x)^{mr-1} dx$ one may deduce the following inequality

$$\mathcal{M}_r(k,m) \ge \frac{k!(r!)^m m^{k+mr}}{(k+mr)!} \left(1 - \frac{(m-1)^k}{m^{k+mr}} \sum_{i=1}^{mr} {k+mr \choose k+i} (m-1)^i \right).$$

Let us assume that k > mr, then $\binom{k+mr}{k+i} \le \binom{k+mr}{k}$, for $1 \le i \le mr$, hence

$$\frac{(m-1)^k}{m^{k+mr}} \sum_{i=1}^{mr} \binom{k+mr}{k+i} (m-1)^i$$

$$\leq \frac{1}{m^{mr}} \left(1 - \frac{1}{m}\right)^k \sum_{i=1}^{mr} \binom{k+mr}{k} (m-1)^i$$

$$\leq \left(1 - \frac{1}{m}\right)^k \binom{k+mr}{k} \frac{((m-1)^{mr+1} - (m-1)}{m-2}$$

$$\leq \left(1 - \frac{1}{m}\right)^k (k+mr)^{mr} \frac{(m-1)^{2mr}}{(mr)!} \xrightarrow[k \to +\infty]{} 0.$$

Proof of Proposition 3.3

Proof. Proposition 3.3 i) Note that the density of the random variable $Y_i = (r-1)X_i$ on [0, r-1] is

$$f(x) = \frac{r}{r-1} \left(1 - \frac{x}{r-1} \right)^{r-1}.$$

Due to the following inequality for all $x \in [0, r-1]$

$$\frac{r}{r-1} \left(1 - \frac{x}{r-1} \right)^{r-1} \le \frac{r}{r-1} \exp(-x)$$

one may deduce that

$$(r-1)^k \mathbb{E}\left(\sum_{i=1}^m X_i\right)^k = \mathbb{E}\left(\sum_{i=1}^m Y_i\right)^k \leq \left(\frac{r}{r-1}\right)^k \mathbb{E}\left(\sum_{i=1}^m \mathcal{E}_i\right)^k$$

$$\leq \left(\frac{r}{r-1}\right)^k \frac{(m-1+k)!}{(m-1)!}$$

since $\sum_{i=1}^{m} \mathcal{E}_i$ has an Erlang distribution (the density of $\sum_{i=1}^{m} \mathcal{E}_i$ is $h(x) = \frac{x^{m-1} \exp(-x)}{(m-1)!}$) whose moment of order k is $\frac{(m-1+k)!}{(m-1)!}$. This inequality completes the proof of i).

ii) The density of the random variable $(r-1)X_1$ converges pointwise to the density of \mathcal{E}_1 namely $\lim_{r\to+\infty} f(x) =$ $\exp(-x)$, for all $x \in [0, +\infty)$. Therefore, the following limit holds

$$\lim_{r \to +\infty} x^k f^{*m}(x) = \frac{x^{m+k-1} \exp(-x)}{(m-1)!}.$$

Finally, the Dominated Convergence Theorem gives that

$$\lim_{r \to +\infty} (r-1)^k \mathbb{E}\left(\sum_{i=1}^m X_i\right)^k = \lim_{r \to +\infty} \int_0^{+\infty} \frac{x^{m+k-1} \exp(-x)}{(m-1)!} dx$$
$$= \mathbb{E}\left(\sum_{i=1}^m \mathcal{E}_i\right)^k = \frac{(m-1+k)!}{(m-1)!}.$$

Funding

The first author is supported by The National Science Centre, Poland NCN grants no. 2014/13/D/ST1/03491 and 2017/27/B/ST1/01467. The Institut de Mathématiques de Bourgogne (IMB) receives support from the EIPHI Graduate School (contract ANR-17-EURE-0002). The second author receives support from the region Bourgogne-Franche-Comté (EPADM project).

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