MATHEMATICAL STATISTICS

Vol. 44, Fasc. 1 (2024), pp. 119–132 Published online 20.9.2024 doi:10.37190/0208-4147.00156

# BETA DISTRIBUTION AND ASSOCIATED STIRLING NUMBERS OF THE SECOND KIND

BY

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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 lower and upper bounds for these numbers, using a probabilistic approach.

**2020 Mathematics Subject Classification:** Primary 05A40; Secondary 44A60.

Key words and phrases: beta distribution, associated Stirling numbers.

# 1. INTRODUCTION

The classical Stirling numbers of the second kind S(p,m) count the number of 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 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 subsequences of the multi-sequence  $\{S_r(p,m) : p,m,r \in \mathbb{N}_{>0}, p \ge rm\}$  appear in the On-Line Encyclopedia of Integer Sequences (OEIS) [11]. Specifically, the arrays  $\{S_1(p,m)\}_{\{p,m\}}$ ,  $\{S_2(p,m)\}_{\{p,m\}}$ ,  $\{S_3(p,m)\}_{\{p,m\}}$  appear as A008277, A008299, A059022. Moreover, the 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 appear in the OEIS as A000478, A058844.

There are well-known connections between Stirling numbers of the second kind and probability theory. For example, the sequences  $S_1(p,m)$  and  $S_2(p,m)$ 

<sup>\*</sup> Jakub Gismatullin is supported by The National Science Centre, Poland, NCN grants no. 2014/13/D/ST1/03491 and 2017/27/B/ST1/01467.

<sup>\*\*</sup> The Institut de Mathématiques de Bourgogne (IMB) receives support from the EIPHI Graduate School (contract ANR-17-EURE-0002). Patrick Tardivel receives support from the region Bourgogne-Franche-Comté (EPADM project).

are asymptotically normal when p tends to  $+\infty$  [6, 4]. More precisely, when  $r \in \{1, 2\}$ , the following convergence in distribution holds:

$$\frac{Y_p - \mathbb{E}(Y_p)}{\sqrt{\operatorname{var}(Y_p)}} \xrightarrow{d} \mathcal{N}(0, 1)$$

where

$$\mathbb{P}(Y_p = m) = \frac{S_r(p, m)}{\sum_{k=1}^p S_r(p, k)} \quad \text{for all } m \in \mathbb{N}_{>0}.$$

Furthermore, according to Dobiński's formula, the moment of order p of a Poisson distribution with parameter  $\lambda \ge 0$  is  $\sum_{m=1}^{p} S_1(p,m)\lambda^m$  (see, e.g., [3, p. 211]). However, to our knowledge, there is no closed 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:

(1.1) 
$$S_r(p,m) = \frac{p!}{m!(r!)^m(p-rm)!} \mathbb{E}[(X_1 + \dots + X_m)^{p-rm}],$$

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

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

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

# 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

(2.1) 
$$g_r(x) = \begin{cases} r(1-x)^{r-1} & \text{if } x \in [0,1], \\ 0 & \text{otherwise.} \end{cases}$$

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

(2.2) 
$$\mathcal{M}_r(k,m) = \mathbb{E}[(X_1 + \dots + X_m)^k].$$

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

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

(2.3) 
$$S_r(p,m) = \frac{p!}{m!(r!)^m(p-rm)!} \mathcal{M}_r(p-rm,m).$$

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 distribution on [0,m] [8, 7]. Note that the moment generating function of Z is  $\sum_{k\geq 0} \mathbb{E}(Z^k)t^k/k! = ((\exp(t)-1)/t)^m$ , and 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 [9, Theorem 3.3, p. 52]). The above expression of  $S_r(p,m)$  is explicit up to the computation of  $\mathcal{M}_r(k,m)$ . Whereas computing it explicitly might be technically complicated, 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**

We will use a probabilistic approach 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  $\overline{X}_m = (X_1 + \cdots + X_m)/m$ . Jensen's inequality provides the following lower bound:

(3.1) 
$$\mathcal{M}_r(k,m) = m^k \mathbb{E}(\overline{X}_m^k) \ge m^k \mathbb{E}(\overline{X}_m)^k = \frac{m^k}{(r+1)^k}.$$

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  $x \in [0, 1]$ :

(3.2) 
$$x^{k} = q(x) \ge q(x_{0}) + q'(x_{0})(x - x_{0})$$
$$= \frac{1}{(r+1)^{k}} + \frac{k}{(r+1)^{k-1}} \left(x - \frac{1}{r+1}\right).$$

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

(3.3) 
$$x^{k} \leq q(x_{0}) + q'(x_{0})(x - x_{0}) + c(x - x_{0})^{2}.$$

Proposition 3.1 below is a consequence of inequalities (3.2) and (3.3).

**PROPOSITION 3.1.** Let  $r, m \in \mathbb{N}_{>0}$  and  $k \in \mathbb{N}$ . Then

(3.4) 
$$\frac{m^k}{(r+1)^k} \leq \mathcal{M}_r(k,m) \leq \frac{m^k}{(r+1)^k} + m^{k-1} \frac{(r+1)^k - 1 - kr}{(r+1)^k r(r+2)}.$$

When m is large, the leading term in both the lower and upper bounds is  $m^k/(r+1)^k$ . Therefore, the 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 sharp when m is large since  $\overline{X}_m$  converges to  $\mathbb{E}(\overline{X}_m)$  and both (3.2) and (3.3) are sharp on the neighbourhood of  $\mathbb{E}(\overline{X}_m)$ .

**3.2. Sharp upper and lower bounds when** k is large. The 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 prove the following inequality in Corollary 5.1:

(3.5) 
$$g_r^{*m}(x) \leq \frac{(r!)^m}{(mr-1)!} (m-x)^{mr-1}$$
 for all  $x \in [0,m]$ ,

where  $g_r$  is given by (2.1) and  $g_r^{*m}$  is the *m*th convolution power of  $g_r$ . Moreover, this inequality is an equality for  $x \in [m-1,m]$ . From this fact we derive lower and upper bounds for  $\mathcal{M}_r(k,m)$  in the proposition below.

**PROPOSITION 3.2.** For any  $r \in \mathbb{N}_{>0}$ , any  $k \in \mathbb{N}$  and any  $m \in \mathbb{N}_{>0}$ ,

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

These bounds are sharp when k is large since

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

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

**3.3. Sharp upper bound when** r is large. Proposition 3.3 below 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 let  $\mathcal{E}_1, \ldots, \mathcal{E}_m$  be *i.i.d.* random variables having the standard exponential distribution with density  $\exp(-x)$ .

(i) The following inequality holds:

(3.6) 
$$r^{k}\mathcal{M}_{r}(k,m) \leq \left(\frac{r}{r-1}\right)^{2k} \mathbb{E}(\mathcal{E}_{1}+\dots+\mathcal{E}_{m})^{k}$$
$$= \left(\frac{r}{r-1}\right)^{2k} \frac{(m-1+k)!}{(m-1)!}.$$

(ii) The upper bound in (i) is sharp since

(3.7) 
$$\lim_{r \to +\infty} r^k \mathcal{M}_r(k,m) = \mathbb{E}(\mathcal{E}_1 + \dots + \mathcal{E}_m)^k = \frac{(m-1+k)!}{(m-1)!}.$$

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

• Proposition 3.1 provides the following lower and upper bounds:

(3.8) 
$$\begin{cases} S_r(p,m) \ge \frac{p!m^{p-rm}}{m!(r!)^m(p-rm)!(r+1)^{p-rm}}, \\ S_r(p,m) \le \frac{p!m^{p-rm}}{m!(r!)^m(p-rm)!(r+1)^{p-rm}} \\ \times \left(1 + \frac{(r+1)^{p-rm} - 1 - r(p-rm)}{mr(r+2)}\right). \end{cases}$$

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

• Proposition 3.2 provides the following lower and upper bounds:

(3.9) 
$$\begin{cases} S_r(p,m) \ge \frac{m^p}{m!} - \frac{(m-1)^{p-rm}}{m!} \sum_{i=1}^{mr} \binom{p}{p-rm+i} (m-1)^i, \\ S_r(p,m) \le \frac{m^p}{m!}. \end{cases}$$

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

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

(3.10) 
$$S_r(p,m) \leq \frac{p! r^{2(p-rm)} (m-1+p-rm)!}{m! (r!)^m (p-rm)! (r-1)^{2(p-rm)} (m-1)!}$$

This upper bound is equivalent to  $S_r(p,m)$  when p - rm and 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 satisfy the following inequalities:

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

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

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

(4.1) 
$$\underbrace{\frac{1}{2}(m^2+m+2)m^{p-m-1}-1}_{L_{\mathrm{rd}}(p,m)} \leqslant S_1(p,m) \leqslant \underbrace{\frac{1}{2}\binom{p}{m}m^{p-m}}_{U_{\mathrm{rd}}(p,m)}.$$

Numerical comparison between U(p,m) and  $U_{rd}(p,m)$  is not needed since clearly

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

for m < p. Unlike the upper bound, the lower bound L(p,m) is not uniformly larger than the one given by Rennie and Dobson; for instance,  $31 = L_{rd}(6,2) > L(6,2) = 28.5$ . 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  $L_{rd}(p,m)$ .

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 partitions of  $\{1, \ldots, p\}$ . Since the Bell number is a sum of Stirling numbers of the second kind,  $B(p) = \sum_{m=1}^{p} S_1(p, m)$ , the following inequality holds:

(4.2) 
$$B(p) \leq \underbrace{\sum_{m=0}^{p} \min\left\{\frac{m^{p}}{m!}, \binom{p}{m} \left(\frac{m}{2}\right)^{p-m} \left(1 + \frac{2^{p-m} + m - p - 1}{3m}\right)\right\}}_{=U(p)}$$

for all  $p \in \mathbb{N}_{>0}$ . In Figure 3 we compare U(p) with the upper bound  $B(p) \leq U_{\mathrm{bt}}(p) = \left(\frac{0.792p}{\ln(p+1)}\right)^p$  given by 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.  $\ln(L(p,m)) - \ln(L_{rd}(p,m))$  as a function of m (on the x-axis) and p (on the y-axis). For most integers the lower bound L(p,m) is a better approximation of  $S_1(p,m)$  than  $L_{rd}(p,m)$  (as  $\ln(L(p,m)) - \ln(L_{rd}(p,m)) > 0$ ).



FIGURE 2.  $\ln(S_1(p, m)) - \ln(L(p, m))$  (left) and  $\ln(U(p, m)) - \ln(S_1(p, m))$  (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 sharply 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.  $\ln(U_{\rm bt}(p)) - \ln(U(p))$  as a function of p. When  $p \ge 13$ , U(p) is a more accurate upper bound for  $S_1(p,m)$  than  $U_{\rm bt}(p)$  (as  $\ln(U_{\rm bt}(p)) - \ln(U(p)) > 0$  for  $p \ge 13$ ).



FIGURE 4. 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

**5.1. Proof 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  $p \in \mathbb{N}$  with  $p \ge rm$ . The r-associated Stirling numbers of the second kind satisfy the equality

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

where the sum is taken 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  $\{1, \ldots, p\}$  into 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,  $\binom{p-r-i_1}{r+i_2}$  possibilities for the second part and so on. Therefore the relevant number is

$$\frac{p!}{(r+i_1)!\cdots(r+i_m)!}$$

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

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

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

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

*Proof of Theorem 2.1.* Let us recall the multinomial formula: for  $x_1, \ldots, x_m$  in  $\mathbb{R}$  and  $k \in \mathbb{N}$ ,

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

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

$$\mathbb{E}[(X_1 + \dots + X_m)^k] = \sum_{i_1 + \dots + i_m = k} \frac{k!}{i_1! \cdots i_m!} \mathbb{E}(X_1^{i_1}) \cdots \mathbb{E}(X_m^{i_m})$$
$$= (r!)^m k! \sum_{i_1 + \dots + i_m = k} \frac{1}{(r+i_1)! \cdots (r+i_m)!}$$
$$= \frac{m!(r!)^m k!}{(k+rm)!} S_r(k+rm,m)$$
$$= \frac{m!(r!)^m (p-rm)!}{p!} S_r(p,m),$$

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



FIGURE 5. Illustration of the inequality given in Lemma 5.2.

*Proof.* First note that for all  $x \in [0, 1]$  the condition  $f(x) \ge x^k$  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

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

Let  $d(x) = ka^{k-1} + c(x - a)$  and  $p(x) = x^{k-1} + ax^{k-2} + \dots + a^{k-1}$ . Then 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  $d(x) \leq p(x)$  if  $x \in [0, a]$ , while  $d(x) \geq p(x)$  if  $x \in [a, 1]$ , which completes the proof.

*Proof of 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

$$x^{k} \leq \mathbb{E}(\overline{X}_{m})^{k} + k\mathbb{E}(\overline{X}_{m})^{k-1}(x - \mathbb{E}(\overline{X}_{m})) + c(x - \mathbb{E}(\overline{X}_{m}))^{2}$$
$$\leq \frac{1}{(r+1)^{k}} + \frac{k}{(r+1)^{k-1}}\left(x - \frac{1}{r+1}\right) + c\left(x - \frac{1}{r+1}\right)^{2}$$

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

$$\mathcal{M}_{r}(k,m) = m^{k} \mathbb{E}(\overline{X}_{m}^{k}) \leqslant m^{k} \left( \mathbb{E}(\overline{X}_{m})^{k} + c \operatorname{var}(\overline{X}_{m}) \right)$$
$$\leqslant \frac{m^{k}}{(1+r)^{k}} + c \frac{rm^{k-1}}{(1+r)^{2}(2+r)}$$
$$\leqslant \frac{m^{k}}{(r+1)^{k}} + \frac{(r+1)^{k} - 1 - kr}{(r+1)^{k}r(r+2)} m^{k-1}.$$

**5.3. Proof of Proposition 3.2.** We use a well-known beta integral (see, for example, [5]): if  $a, b \in \mathbb{N}$  and  $x \in \mathbb{R}_{>0}$  then

(5.1) 
$$\int_{0}^{x} (x-t)^{a} t^{b} dt = \frac{a!b!}{(a+b+1)!} x^{a+b+1}.$$

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

(5.2) 
$$h_r(x) = \begin{cases} rx^{r-1} & \text{if } x \in [0,1], \\ 0 & \text{otherwise.} \end{cases}$$

Then  $h_r$  is the density of 1-X when the density of X is the  $g_r$  of (2.1). Convolution computations are slightly easier to handle with  $h_r$  than  $g_r$ .

LEMMA 5.3. Let  $m \in \mathbb{N}_{>0}$ . Then

(5.3) 
$$h_r^{*m}(x) = \frac{(r!)^m}{(mr-1)!} x^{mr-1} \quad \text{for all } x \in [0,1],$$

(5.4) 
$$h_r^{*m}(x) \leq \frac{(r!)^m}{(mr-1)!} x^{mr-1} \quad \text{for all } x \in \mathbb{R}_{\geq 0}.$$

*Proof.* We prove (5.3) by induction. When m = 1, we notice that  $h_r^{*m}(x) = h_r(x)$  for all  $x \in [0, 1]$ . Let  $m \in \mathbb{N}_{>0}$  be such that

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

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

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

The proof of (5.4) by induction is quite similar. When m = 1, the result is straightforward. Let  $m \in \mathbb{N}_{>0}$  be such that

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

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

$$\begin{split} h_r^{*m+1}(x) &= \int_{\mathbb{R}} h_r^{*m}(x-t)h_r(t) \, dt = \int_0^x h_r^{*m}(x-t)h_r(t) \, dt \\ &\leqslant \frac{r(r!)^m}{(mr-1)!} \int_0^x (x-t)^{mr-1} t^{r-1} \, dt \\ &\leqslant \frac{r(r!)^m}{(mr-1)!} \frac{(mr-1)!(r-1)!}{((m+1)r-1)!} x^{(m+1)r-1} \\ &\leqslant \frac{(r!)^{m+1}}{((m+1)r-1)!} x^{(m+1)r-1}. \quad \bullet \end{split}$$

Note that the only difference between the proofs of (5.3) and (5.4) 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 power  $g_r^{*m}$  on [m-1,m] and an upper bound on [0,m].

COROLLARY 5.1. For all  $x \in [0, m]$  we have

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

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. Since the density of  $(1 - X_1) + \cdots + (1 - X_m)$  is  $h_r^{*m}$ , we have

$$\frac{\partial}{\partial z} \mathbb{P}((1 - X_1) + \dots + (1 - X_m) \leq z) = h_r^{*m}(z)$$
$$\frac{\partial}{\partial z} \mathbb{P}(X_1 + \dots + X_m \geq m - z) = h_r^{*m}(z)$$
$$g_r^{*m}(m - z) = h_r^{*m}(z)$$
$$g_r^{*m}(x) = h_r^{*m}(m - x). \quad \blacksquare$$

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]).

The upper and lower bounds given in Proposition 3.2 are straightforward consequences of Corollary 5.1. Indeed, the upper bound is just the beta integral

$$\int_{0}^{m} \frac{(r!)^{m}}{(mr-1)!} (m-x)^{mr-1} x^{k} \, dx.$$

*Proof of Proposition 3.2.* The lower bound is derived from the following computation:

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

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

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

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

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

Because  $\mathcal{M}_r(k,m) \ge \frac{(r!)^m}{(mr-1)!} \int_{m-1}^m x^k (m-x)^{mr-1} dx$  one may deduce that

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

Assume that k > mr. Then  $\binom{k+mr}{k+i} \leq \binom{k+mr}{k}$  for  $1 \leq i \leq mr$ , hence

$$\begin{aligned} \frac{(m-1)^k}{m^{k+mr}} \sum_{i=1}^{mr} \binom{k+mr}{k+i} (m-1)^i \\ &\leqslant \frac{1}{m^{mr}} \left(1 - \frac{1}{m}\right)^k \sum_{i=1}^{mr} \binom{k+mr}{k} (m-1)^i \\ &\leqslant \left(1 - \frac{1}{m}\right)^k \binom{k+mr}{k} \frac{((m-1)^{mr+1} - (m-1))}{m-2} \\ &\leqslant \left(1 - \frac{1}{m}\right)^k (k+mr)^{mr} \frac{(m-1)^{2mr}}{(mr)!} \xrightarrow[k \to +\infty]{} 0. \end{aligned}$$

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

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

Since for all  $x \in [0, r-1]$ ,

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

one may deduce that

$$(r-1)^{k} \mathbb{E}\left(\sum_{i=1}^{m} X_{i}\right)^{k} = \mathbb{E}\left(\sum_{i=1}^{m} Y_{i}\right)^{k} \leqslant \left(\frac{r}{r-1}\right)^{k} \mathbb{E}\left(\sum_{i=1}^{m} \mathcal{E}_{i}\right)^{k}$$
$$\leqslant \left(\frac{r}{r-1}\right)^{k} \frac{(m-1+k)!}{(m-1)!}$$

because  $\sum_{i=1}^{m} \mathcal{E}_i$  has an Erlang distribution (with density  $h(x) = \frac{x^{m-1} \exp(-x)}{(m-1)!}$ ) whose moment of order k is  $\frac{(m-1+k)!}{(m-1)!}$ . This inequality completes the proof of (i). (ii) The density of  $(r-1)X_1$  converges pointwise to the density of  $\mathcal{E}_1$ , namely

 $\lim_{x\to+\infty} f(x) = \exp(-x)$  for all  $x \in [0, +\infty)$ . Therefore,

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

Finally, the dominated convergence theorem gives

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

Acknowledgments. The authors thank the anonymous referee for the suggestion concerning the on-line encyclopedia of integer sequences.

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Received 5.1.2024; accepted 31.7.2024