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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.

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#### 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  $S(p, m) = S_1(p, m)$ .

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  $+\infty$  [4, 7]. More precisely, when  $r \in \{1, 2\}$ , the following convergence in distribution occurs

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

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 a beta distribution with parameter (1, 1) is a uniform distribution on [0, 1]. Thus, when r = 1, the above formula is quite simple:

$$S_1(p,m) = \binom{p}{m} \mathbb{E}\left(Z^{p-m}\right), \text{ where } Z = \sum_{i=1}^m X_i \text{ has the Irwin-Hall distribution on } [0,m]$$

Propositions 3.1, 3.2 and 3.3 give upper and lower bounds for  $\mathbb{E}\left[(X_1 + \dots + X_m)^{p-rm}\right]$ . 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.

#### 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

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

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

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] [6, 8]. 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

$$\mathcal{M}_{r}(k,m) = m^{k} \mathbb{E}\left(\overline{X}_{m}^{k}\right) \geq m^{k} \mathbb{E}\left(\overline{X}_{m}\right)^{k} = \frac{m^{k}}{(r+1)^{k}}.$$
(4)

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  $x \in [0, 1]$ 

$$x^{k} = q(x) \ge q(x_{0}) + q'(x_{0})(x - x_{0}) = \frac{1}{(r+1)^{k}} + \frac{k}{(r+1)^{k-1}} \left(x - \frac{1}{r+1}\right).$$
(5)

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

$$x^{k} \leq q(x_{0}) + q'(x_{0})(x - x_{0}) + c(x - x_{0})^{2}.$$
(6)

Proposition 3.1 below is a consequence of inequalities (5) and (6).

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

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

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 (5) and (6) 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:

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

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.

$$\begin{aligned} \mathbf{Upper \ bound:} \ \mathcal{M}_{r}(k,m) &\leq \frac{(r!)^{m}}{(mr-1)!} \int_{0}^{m} x^{k} (m-x)^{mr-1} dx = \frac{k! (r!)^{m} m^{k+rm}}{(k+mr)!}. \\ \mathbf{Lower \ bound:} \ \mathcal{M}_{r}(k,m) &\geq \frac{(r!)^{m}}{(mr-1)!} \int_{m-1}^{m} x^{k} (m-x)^{mr-1} dx \\ &\geq \frac{k! (r!)^{m} m^{k+mr}}{(k+mr)!} \left( 1 - \frac{(m-1)^{k}}{m^{k+mr}} \sum_{i=1}^{mr} \binom{k+mr}{k+i} (m-1)^{i} \right). \end{aligned}$$

These bounds are accurate 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, 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

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

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

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

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:

• Proposition 3.1 provides the following lower and upper bounds

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

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

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

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

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

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

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

#### 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:

$$S_{1}(p,m) \leq \underbrace{\min\left\{\frac{m^{p}}{m!}, \binom{p}{m} \left(\frac{m}{2}\right)^{p-m} \left(1 + \frac{2^{p-m} + m - p - 1}{3m}\right)\right\}}_{U(p,m)},\\S_{1}(p,m) \geq \underbrace{\max\left\{\frac{m^{p}}{m!} - \frac{(m-1)^{p-m}}{m!} \sum_{i=1}^{m} \binom{p}{m-i} (m-1)^{i}, \binom{p}{m} \left(\frac{m}{2}\right)^{p-m}\right\}}_{L(p,m)}.$$

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

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

Numerical comparison between U(p,m) and  $U_{\rm rd}(p,m)$  is not relevant since  $\frac{1}{2} {p \choose m} m^{p-m} \ge {p \choose m} \left(\frac{m}{2}\right)^{p-m} \left(1 + \frac{2^{p-m}+m-p-1}{3m}\right)^{p-m}$ 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  $L_{\rm rd}(p,m)$ .

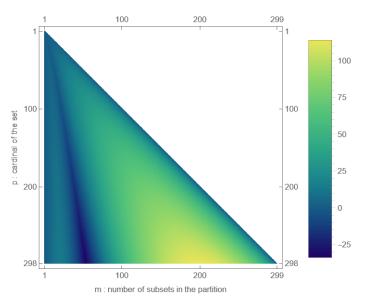


FIG 1. This figure report  $\ln(L(p,m)) - \ln(L_{rd}(p,m))$  as a function of m (on the x-axis) and p (on the y-axis). One may observe that for most integers the lower bound L(p,m) is a better approximation of  $S_1(p,m)$  than  $L_{rd}(p,m)$  (as  $\ln(L(p,m)) - \ln(L_{rd}(p,m)) > 0$ ).

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

#### 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:

$$B(p) \leq \underbrace{\sum_{m=0}^{p} \min\left\{\frac{m^{p}}{m!}, \binom{p}{m} \left(\frac{m}{2}\right)^{p-m} \left(1 + \frac{2^{p-m} + m - p - 1}{3m}\right)\right\}}_{=U(p)} \quad \forall p \in \mathbb{N}_{>0}.$$

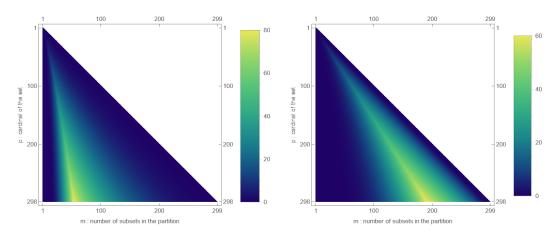


FIG 2. This figure report  $\ln(S_1(p,m)) - \ln(L(p,m))$  (on the left) and  $\ln(U(p,m)) - \ln(S_1(p,m))$  (on the right) as a function of m and p. These numerical experiments comply with Propositions 3.1 and 3.2 since both lower and upper bounds accurately approximate  $S_1(p,m)$  when p is large and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

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

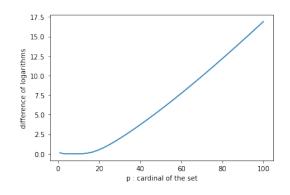


FIG 3. This figure report  $\ln(U_{bt}(p)) - \ln(U(p))$  as a function of p. One may observe that when  $p \ge 13$ , U(p) is more accurate upper bound for  $S_1(p,m)$  than  $U_{bt}(p)$  (as  $\ln(U_{bt}(p)) - \ln(U(p)) > 0$  for  $p \ge 13$ ).

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 shows that U(p)/B(p) is very close to e when p is large.

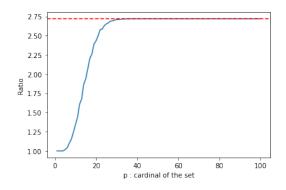


FIG 4. This figure report U(p)/B(p) as a function of p. One may observe that U(p)/B(p) is approximately equal to e when p is large.

#### 5. Proofs

#### 5.1. 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  $p \in \mathbb{N}$  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  $i_1, \ldots, i_m$  parts where the first part has  $n + i_n$  elements the second part has  $n + i_n$  elements and  $a_n$  are a superpart.

 $\{1, \ldots, p\}$  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

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

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

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

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 of Theorem 2.1.

#### 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  $f: x \in [0,1] \to a^k + ka^{k-1}(x-a) + c(x-a)^2$  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  $x \in [0,1]$ .

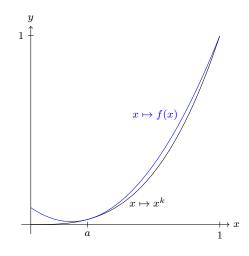


FIG 5. Illustration of the inequality given in Lemma 5.2.

*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

$$\begin{aligned} ka^{k-1} + c(x-a) &\geq x^{k-1} + ax^{k-2} + \dots + a^{k-1} \text{ for all } x \in [a,1] \text{ and} \\ ka^{k-1} + c(x-a) &\leq x^{k-1} + ax^{k-2} + \dots + a^{k-1} \text{ for all } x \in [0,a]. \end{aligned}$$

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

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

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

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

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

#### 5.3. Proof of Proposition 3.2

We use 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

$$\int_0^x (x-t)^a t^b dt = \frac{a!b!}{(a+b+1)!} x^{a+b+1}.$$
(7)

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

$$h_r(x) = \begin{cases} rx^{r-1} & \text{if } x \in [0,1] \\ 0 & \text{otherwise.} \end{cases}$$
(8)

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

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

$$h_r^{*m}(x) \leq \frac{(r!)^m}{(mr-1)!} x^{mr-1}, \text{ for all } x \in \mathbb{R}_{\geq 0}.$$
 (10)

*Proof.* Let us prove (9) 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

 $h_r^{*m}$ 

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

The proof of (10) by induction is quite similar than the proof of (9). 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

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

Note that the only difference between the proof of (9) and (10) 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 [6, 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

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

The lower bound is derived from the following computations

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

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

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

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

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 the following inequality

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

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

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

#### 5.4. Proof of Proposition 3.3

*Proof.* 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

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

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

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

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