

# **Beta distribution and associated Stirling numbers of the second kind**

Jakub Gismatullin, Patrick J C Tardivel

# **To cite this version:**

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

# **HAL Id: hal-04366123 <https://hal.science/hal-04366123v2>**

Submitted on 25 Jul 2024

**HAL** is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

# Beta distribution and associated Stirling numbers of the second kind

Jakub Gismatullin<sup>∗1</sup> and Patrick Tardivel<sup>†1,2</sup>

<sup>1</sup>Instytut Matematyczny Uniwersytetu Wrocławskiego, pl. Grunwaldzki 2/4, 50-384 Wrocław <sup>2</sup>Institut de Mathématiques de Bourgogne, UMR 5584 CNRS, Université de Bourgogne, Dijon, France

#### 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,](http://oeis.org/A008277) [A008299,](http://oeis.org/A008299) [A059022.](http://oeis.org/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,](http://oeis.org/A000478) [A058844.](https://oeis.org/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

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

Otherwise, according to Dobinski'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

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

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:

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

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  $\mathbb{E}\left[\left(X_1 + \cdots + X_m\right)^{p-rm}\right]$ . These bounds are sharp when one parameter m, r, or  $p-rm$  tends to +∞ and provides thus accurate approximations of r−associated Stirling numbers.

<sup>∗</sup> jakub.gismatullin@uwr.edu.pl orcid: 0000-0002-4711-3075

<sup>†</sup>Patrick.Tardivel@u-bourgogne.fr orcid: 0000-0002-8496-3909

# 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

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

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  $p \ge rm$ . The Stirling numbers of the second kind satisfy the following identity

$$
S_r(p,m) = \frac{p!}{m!(r!)^m(p-rm)!} \mathcal{M}_r(p-rm,m).
$$
\n(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  $Z \text{ is } \sum_{k\geq 0} \mathbb{E}(Z^k)t^k/k! = ((\exp(t) - 1)/t)^m$ 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

$$
\mathcal{M}_r(k,m) = m^k \mathbb{E}\left(\overline{X}_m^k\right) \ge m^k \mathbb{E}\left(\overline{X}_m\right)^k = \frac{m^k}{(r+1)^k}.
$$
\n(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  $x \in [0, 1]$ 

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

$$
x^{k} \leq q(x_0) + q'(x_0)(x - x_0) + c(x - x_0)^2. \tag{7}
$$

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

$$
\frac{m^k}{(r+1)^k} \le \mathcal{M}_r(k,m) \le \frac{m^k}{(r+1)^k} + m^{k-1} \frac{(r+1)^k - 1 - kr}{(r+1)^k r(r+2)}.
$$
\n(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  $k \in \mathbb{N}_{>0}$  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.

$$
\mathcal{M}_r(k,m) \leq \frac{(r!)^m}{(mr-1)!} \int_0^m x^k (m-x)^{mr-1} dx = \frac{k!(r!)^m m^{k+rm}}{(k+mr)!}.
$$
  

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

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

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

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

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

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

• Proposition 3.1 provides the following lower and upper bounds

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

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

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

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

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

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

$$
S_1(p,m) \leq \underbrace{\min\left\{\frac{m^p}{m!}, \binom{p}{m} \left(\frac{m}{2}\right)^{p-m} \left(1 + \frac{2^{p-m} + m - p - 1}{3m}\right)\right\}}_{U(p,m)}
$$
  

$$
S_1(p,m) \geq \underbrace{\max\left\{\frac{m^p}{m!} - \frac{(m-1)^{p-m}}{m!} \sum_{i=1}^m \binom{p}{m-i} (m-1)^i, \binom{p}{m} \left(\frac{m}{2}\right)^{p-m}\right\}}_{L(p,m)}.
$$

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:

$$
\underbrace{\frac{1}{2}(m^2 + m + 2)m^{p-m-1} - 1}_{L_{\rm rd}(p,m)} \le S_1(p,m) \le \underbrace{\frac{1}{2}\binom{p}{m}m^{p-m}}_{U_{\rm rd}(p,m)}.
$$
\n(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:  $B(p) \leq U_{\text{bt}}(p) = \left(\frac{0.792p}{\ln(p+1)}\right)^p$  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 Dobinski 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  $\ln(L(p, m)) - \ln(L_{\text{rd}}(p, m))$  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  $U_{\text{bt}}(p)$  (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

#### 5.1 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  $p \ge rm$ . 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  $r + i<sub>2</sub>$  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

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

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

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

 $\Box$ 

*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

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



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

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

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

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

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

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.  $\Box$ 

*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

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

$$
\leq \frac{1}{(r+1)^{k}} + \frac{k}{(r+1)^{k-1}} \left(x - \frac{1}{r+1}\right) + c \left(x - \frac{1}{r+1}\right)^{2}
$$

 $\Box$ 

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

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

#### 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

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

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

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

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

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

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

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

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]).

 $\Box$ 

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

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

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

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

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

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

#### 5.4 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

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

$$
= \mathbb{E} \left( \sum_{i=1}^m \mathcal{E}_i \right)^k = \frac{(m-1+k)!}{(m-1)!}.
$$

 $\Box$ 



## 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).

## References

- [1] D. Berend T. Tassa, Improved bounds on Bell numbers and on moments of sums of random variables, Probab. Math. Statist. 30 (2010), pp. 185–205.
- [2] W. E. Bleick P.C.C. Wang, Asymptotics of Stirling numbers of the second kind, Proc. Amer. Math. Soc. 42 (1974), pp. 575–580.
- [3] L. Comtet, Advanced combinatorics: The art of finite and infinite expansions, Springer, 1974.
- [4] E. Czabarka, P.L. Erdős, V. Johnson, A. Kupczok, L.A. Székely, Asymptotically normal distribution of some tree families relevant for phylogenetics, and of partitions without singletons, Mosc. J. Comb. Number Theory. 1 (2011), pp. 12–24.
- [5] A.K. Gupta S. Nadarajah, Handbook of beta distribution and its applications, 1st ed., CRC press, 2004.
- [6] L.H. Harper, Stirling behavior is asymptotically normal, Ann. Math. Statist. 38 (1967), pp. 410–414.
- [7] P. Hall, The distribution of means for samples of size n drawn from a population in which the variate takes values between 0 and 1, all such values being equally probable, Biometrika. 19 (1927), pp. 240-244.
- [8] J.O. Irwin, On the frequency distribution of the means of samples from a population having any law of frequency with finite moments, with special reference to Pearson's type II, Biometrika. 19 (1927), pp. 225- 239.
- [9] T. Mansour M. Schork, Commutation relations, normal ordering, and Stirling numbers, CRC Press, 2015.
- [10] B.C. Rennie A.J. Dobson, On Stirling numbers of the second kind, J. Combinatorial Theory, 7 (1969), pp. 116–121.
- [11] N.J.A. Sloane The OEIS Foundation Inc., The On-Line Encyclopedia of Integer Sequences, (2024).