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## The number of distinct part sizes of some multiplicity in compositions of an Integer. A probabilistic Analysis

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Random compositions of integers are used as theoretical models for many applications. The degree of distinctness of a composition is a natural and important parameter. A possible measure of distinctness is the number *X* of distinct parts (or components). This parameter has been analyzed in several papers. In this article we consider a variant of the distinctness: the number  $X(m)$  of distinct parts of multiplicity m that we call the m-distinctness. A first motivation is a question asked by Wilf for random compositions: what is the asymptotic value of the probability that a randomly chosen part size in a random composition of an integer v has multiplicity m. This is related to  $\mathbb{E}(X(m))$ , which has been analyzed by Hitczenko, Rousseau and Savage. Here, we investigate, from a probabilistic point of view, the first full part, the maximum part size and the distribution of  $X(m)$ . We obtain asymptotically, as  $v \to \infty$ , the moments and an expression for a continuous distribution  $\varphi$ , the (discrete) distribution of  $X(m, v)$  being computable from  $\varphi$ .

**Keywords:** Mellin transforms, urns models, Poissonization, saddle point method, generating functions

## 1 Introduction

Let us first recall some well-known results. Let us consider the composition of an integer  $v$ , i.e.  $v =$  $\sum_{i}^{N} x_i$ ,  $x_i$ : integer > 0. Considering all compositions as equiprobable, we know (see [HL01]) that the number of parts *N* is asymptotically Gaussian,  $v \rightarrow \infty$ :

$$
N \sim \mathcal{N}\left(\frac{\nu}{2}, \frac{\nu}{4}\right),\tag{1}
$$

and that the part sizes are asymptotically id  $GEOM(1/2)$  and *independent*. Consider now *n* random variables (R.V.), GEOM $(1/2)$  and define the indicator R.V.<sup>†</sup>

 $Y_i :=$  [value *i* appears among these *n* R.V.]

Then, asymptotically,  $n \to \infty$ , the  $Y_i$  are independent. The first empty part value, i.e the first *k* such that  $Y_k = 0$ , is of order  $O(\log n)$ . Here and in the sequel,  $\log := \log_2 L := \ln 2$ . Similarly, the maximum part

<sup>†</sup> Here we use the indicator function notation proposed by Knuth et al. [GKP89].

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size is also of order  $O(\log n)$ , as well as the number *Y* of distinct values (part sizes):  $Y = \sum_{i=1}^{\infty} Y_i$ . The asymptotic distributions and moments of these R.V. are also given in [HL01]. We know (see Hwang and Yeh [HY97]) that

$$
\mathbb{E}(Y) \sim \log n + \gamma/L - 1/2 + \beta(\log n) + O(1/n)
$$

where  $\beta$  is a small periodic function of log<sub>n</sub>, and the distribution of *Y* is highly concentrated around its mean, with a  $O(1)$  range. All these distributions depend on  $\log n$ . Hence, with (1), the same R.V. related to v are asymptotically equivalent by replacing  $\log n$  by  $\log \nu - 1$  (see [HL01]).

In this article we consider a variant of the distinctness: the number  $X(m)$  of distinct parts of multiplicity *m* that we call the *m*-distinctness. A first motivation is a question asked by Wilf for random compositions: what is the asymptotic value of the probability  $P(m, v)$  that a randomly chosen part size in a random composition of an integer ν has multiplicity *m*. (The corresponding problem for random partitions has been analyzed in Corteel et al. [CPSW99]). Of course, here,

$$
P(m, v) = \mathbb{E}(X(m, v)/Y(v)),
$$

where we explicitly show the dependence on v. But, as already mentioned,  $Y(v)$  has asymptotically the same distribution as *Y* (with log<sub>n</sub> replaced by logy  $-1$ ). On the other side, *Y* is highly concentrated around its mean . Hence, asymptotically, as shown in Hitczenko, Savage [HS99] and Hitczenko et al [HRS02], for  $m = O(1)$ ,

$$
P(m, v) \sim \mathbb{E}(X(m, v))/\mathbb{E}(Y(v)).
$$

Here, we investigate, from a probabilistic point of view, the first full part, the maximum part size and the distribution of  $X(m, v)$ . We obtain asymptotically, as  $v \to \infty$ , the moments and an expression for a continuous distribution  $\varphi$ , the (discrete) distribution of  $X(m, v)$  being computable from  $\varphi$ . We will see that, again, all asymptotic distributions for some multiplicity *m* depend only on log*n*. Hence, the same R.V. related to v are again simply obtained by replacing  $\log n$  by  $\log v - 1$ . The paper is organized as follows: in Section 2, we consider a fixed multiplicity  $m = O(1)$ . We analyze the moments, the first full part, the maximum part size, and the distribution of  $X(m)$ . Section 3 is devoted to large multiplicity *m*. Section 4 concludes the paper. Due to length constraints, some proofs have been briefly presented.

In this section, we are interested in the properties of the R.V.:  $X_i(m) := \llbracket$  value *i* appears among the *n* GEOM(1/2) R.V. with multiplicity *m*, for fixed  $m = O(1) \llbracket$ . Of course,

$$
Pr[X_i(m) = 1] = {n \choose m} (1/2^i)^m (1 - 1/2^i)^{n-m}.
$$
 (2)

We immediately see that the dominant range is given by  $i = \log n + O(1)$ . To the left and the right of this range,  $Pr[X_i(m) = 1] \sim 0$ . Within the range,  $Pr[X_i(m) = 1]$  is asymptotically equivalent to a Poisson distribution:

$$
Pr[X_i(m) = 1] \sim \frac{1}{m!} (n/2^i)^m \exp(-n/2^i),
$$

and, with  $X(m) := \sum_{i=1}^{\infty} X_i(m)$ ,

$$
\mathbb{E}(X(m)) \sim G(n,m),
$$

where, using the "sum splitting technique" as described in Knuth [Knu73], p.131,

$$
G(n,m) := \frac{1}{m!} \sum_{i=1}^{\infty} (n/2^i)^m \exp(-n/2^i),
$$

which, for large *n*, can be analyzed using Mellin transforms: see Flajolet et al. [FGD95]. It is well known that the dominant value is given by some constant. The oscillatory part has a very small amplitude, usually of order  $10^{-5}$ . Indeed, set  $f(y) := y^m e^{-y}$ . We obtain

$$
G(n,m) = \frac{1}{m!} \sum_{i=1}^{\infty} f(n/2^i),
$$

the Mellin transform of which is

$$
G^*(s) = \frac{\Gamma(m+s)}{m!} \frac{2^s}{1-2^s},
$$

defined in the fundamental strip  $\langle -m, 0 \rangle$ . To the right of this strip, the poles of  $G^s(s)$  are a simple pole at  $s = 0$ , and simple poles at  $s = \chi_k := 2k\pi i/L(k \neq 0)$ . The singular expansion of  $G^*(s)$  is given by <sup>‡</sup>

$$
G^*(s) \approx \left[\frac{\Gamma(m)}{Lm!s}\right] + \sum_{k\neq 0} \frac{\Gamma(m+\chi_k)}{Lm!(s-\chi_k)}.
$$

This leads, by converse mapping, to

$$
G(n,m) \sim \frac{1}{mL} + \beta_0(\log n) + O(1/n),\tag{3}
$$

where  $\beta_0$  is a small periodic function of log *n*:

$$
\beta_0(\log_2 n) := \sum_{k\neq 0} \frac{\Gamma(m+\chi_k)}{Lm!} n^{-\chi_k} = \sum_{k\neq 0} \frac{\Gamma(m+\chi_k)}{Lm!} e^{-2\pi i k \log n}.
$$

In the sequel,  $\beta$ . (log*n*) will always denote (small) periodic functions. As *n* ~  $\mathcal{N}(\frac{v}{2}, \frac{v}{4})$ , we just have to replace  $\log n$  by  $\log y - 1$ . So we recover the mean already computed in Hitczenko and Savage, [HS99] and Hitczenko, Rousseau and Savage, [HRS02]. To compute all moments, we must check that the  $X_i$ are asymptotically independent. We could proceed as was done in [HL01] for the *Y<sup>i</sup>* , but we follow here another route. Let us consider  $\Pi_n = \mathbb{E}(z^X)$ . We obtain

#### **Theorem 1.1.**

$$
\Pi_n \sim \prod_{l=1}^{\infty} \left[ \left( 1 - \frac{1}{m!} (n/2^l)^m e^{-n/2^l} \right) + z \frac{1}{m!} (n/2^l)^m e^{-n/2^l} \right], n \to \infty.
$$

*Proof.* We use an urn model, as in Sevastyanov and Chistyakov, [SC64] and Chistyakov, [Chi67], and the Poissonization method (see, for instance Jacquet and Szpankowski [JS98] for a general survey). If we Poissonize, with parameter  $\tau$ , the number of balls (i.e the number *n* of R.V. here), the generating function of  $X_l$  is given from (2), by

$$
\left(1 - \frac{1}{m!}(\tau/2^l)^m e^{-\tau/2^l}\right) + z\frac{1}{m!}(\tau/2^l)^m e^{-\tau/2^l},
$$

 $\ddot{\ddot{\phi}}$ . The symbol  $\times$  is used to denote the fact that two functions are of the same asymptotic order.

and we have independency of cells occupation. This leads to

$$
e^{-\tau} \sum_{n} \frac{\tau^n}{n!} \Pi_n = \prod_{l=1}^{\infty} \left[ \left( 1 - \frac{1}{m!} (\tau/2^l)^m e^{-\tau/2^l} \right) + z \frac{1}{m!} (\tau/2^l)^m e^{-\tau/2^l} \right].
$$

Hence, by Cauchy, we obtain  $\Pi_n = \frac{n!}{2\pi i} \int \exp\{nf\}$  $\frac{\partial}{\partial \pi i} \int_{\Gamma} \exp\{nf(\tau)\} d\tau/\tau$ , where  $\Gamma$  is inside the analyticity domain of the integrand and encircles the origin, and

$$
f(\tau) := -\log \tau + \tau/n + \frac{1}{n} \sum_{l=1}^{\infty} \ln \left[ \left( 1 - \frac{1}{m!} (\tau/2^l)^m e^{-\tau/2^l} \right) + z \frac{1}{m!} (\tau/2^l)^m e^{-\tau/2^l} \right].
$$

By standard saddle-point method (see, for instance, Flajolet and Sedgewick, [FS94]), we look for  $\tau^*$  such that  $f'(\tau^*) = 0$ , with

$$
f'(\tau) = -1/\tau + 1/n - \frac{z-1}{n\tau} \sum_{l=1}^{\infty} \frac{(\tau/2^l)^{m+1} - m(\tau/2^l)^m}{m!\exp(\tau/2^l) - (\tau/2^l)^m + z(\tau/2^l)^m}.
$$

But, again by Mellin, for fixed  $z > 0$ ,

$$
\sum_{l=1}^{\infty} \frac{(\tau/2^l)^{m+1} - m(\tau/2^l)^m}{m! \exp(\tau/2^l) - (\tau/2^l)^m + z(\tau/2^l)^m} \sim C + \beta.(\log \tau).
$$

with

$$
C := \int_0^\infty \frac{y^{m+1} - my^m}{m! \exp(y) - y^m + zy^m} dy/L.
$$

Hence  $\tau^* \sim n + C$ . It is easily checked that  $C = 0$ . Finally,  $\Pi_n \sim \frac{n! e^{n f(\tau^*)}}{\sqrt{n!} \sqrt{n!} \sqrt{n!}}$  $\frac{n!}{\sqrt{2\pi}\tau^*\sqrt{n f''(\tau^*)}}$ , and, by Stirling, we easily derive the theorem.  $\Box$ 

Theorem 1.1 confirms the asymptotic independence assumption.

#### 1.1 The moments of  $X(m)$

We now have all necessary ingredients to compute the moments. The variance of  $X(m)$  is now easily derived: we obtain, by Mellin,

$$
\text{VAR}(X(m)) \sim \frac{1}{m!} \sum_{1}^{\infty} (n/2^i)^m \exp(-n/2^i) \left[ 1 - \frac{1}{m!} (n/2^i)^m \exp(-n/2^i) \right]
$$

$$
\sim \int_0^{\infty} e^{-y} \frac{y^m}{m!} (1 - e^{-y} \frac{y^m}{m!}) \frac{dy}{Ly} + \beta_1 (\log_2 n)
$$

$$
= \frac{1}{mL} - \frac{(2m-1)!}{Lm!^2 2^{2m}} + \beta_1 (\log_2 n).
$$

The other moments can be derived as follows. We obtain, setting  $z = e^s$ ,

$$
\ln(\Pi_n) \sim S_2 = \sum_{l=1}^{\infty} \ln \left[ 1 + (e^s - 1) \frac{1}{m!} (n/2^l)^m \exp(-n/2^l) \right]
$$
  
= 
$$
\sum_{i=1}^{\infty} \frac{(-1)^{i+1} (e^s - 1)^i V_i}{i}, \text{ with}
$$
  

$$
V_i := \sum_{l=1}^{\infty} \left[ \frac{1}{m!} (n/2^l)^m \right]^i \exp(-in/2^l).
$$

The centered moments of  $X(m)$  can be obtained by analyzing

$$
S_3 := \exp[S_2 - sV_1].
$$

Again, by Mellin, we obtain

$$
V_i \sim B_i + \beta_i(\log n),
$$

with

$$
B_i = \int_0^\infty \left[\frac{y^m}{m!}\right]^i e^{-iy} \frac{dy}{Ly} = \frac{(im-1)!}{m!^i L^{im}},
$$

and finally, the centered moments are given by

$$
\tilde{\sigma}^2 := \text{VAR}(X(m)) \sim \frac{1}{mL} - \frac{(2m)!}{2Lm!^2 2^{2m} m!},
$$
  
\n
$$
\tilde{\mu}_3 := \mu_3(X(m)) \sim \frac{1}{mL} - \frac{3(2m)!}{2Lm!^2 2^{2m} m} + \frac{2(3m)!}{3Lm!^3 3^{3m} m!},
$$
  
\n
$$
\tilde{\mu}_4 := \mu_4(X(m)) \sim \frac{1}{mL} + \frac{3}{m^2 L^2} - \frac{3(4m)!}{2Lm!^4 4^{4m} m} + \frac{4(3m)!}{Lm!^3 3^{3m} m} - \frac{7(2m)!}{2Lm!^2 2^{2m} m} - \frac{3(2m)!}{3L^2 m!^2 2^{2m} m^2} + \frac{3(2m)!^2}{4L^2 m!^4 2^{4m} m^2}.
$$

The neglected terms are made of periodic functions  $\beta$  (log*n*) and of  $O(\frac{1}{n})$  contributions.

Again, the centered moments (of order  $\geq$  2) of *X* related to a composition of v are given by the same expressions.

For  $n = 20000$ ,  $m = 2$ , we have done a simulation (of  $T = 4000$  sets). We obtain the results of Table 1 (the probability related moments are explained later on). For an easy comparison, we give here only four significant digits.

#### 1.2 The maximum part size of multiplicity *m*

The maximum part size  $M_n(m)$  of multiplicity *m* is such that

$$
\Pr(\mathcal{M}_n(m)
$$



**Tab.** 1: Moments,  $n = 20000$ ,  $m = 2$ .

Set  $\eta := Lk - \ln n$ . This leads, with  $\eta = O(1)$ , to

$$
\Pr(\mathcal{M}_n(m)
$$

with

$$
\varphi_1(m,\eta) = \prod_{j=0}^{\infty} \left[ 1 - \frac{1}{m!} e^{-m(\eta + Lj)} e^{-e^{-(\eta + Lj)}} \right].
$$

Figure 1 gives  $\varphi_1(m,\eta)$  for  $m=1,\ldots,4$ , bottom to top. It appears that for  $\eta \to -\infty$ ,  $\varphi_1(m,\eta)$  seems to converge to some value, which of course corresponds to

$$
P(m,0) := \Pr(X(m) = 0),
$$

but a closer view reveals the usual fluctuations, shown in Figure 2, for  $m = 2$ . Set  $\psi(n) := \log n \lfloor \log n \rfloor$  (fractional part). With  $\eta = L(-6 - \psi(20000))$ , we obtain  $P(2,0) = .4489079864...$ , which will be compared later on with a direct expression.

Similarly, we derive

$$
Pr(\mathcal{M}_n(m) = k - 1) \sim \varphi_2(m, \eta) = \varphi_1(m, \eta) e^{-m(\eta - L)} e^{-e^{-(\eta - L)}/m!}.
$$

Figure 3 gives  $\varphi_2(m,\eta)$  for  $m = 1, \ldots, 4$ , (more and more concentrated as *m* increases).

Our simulation for  $n = 20000$ ,  $m = 2$  of  $T = 4000$  sets leads to Figure 4 ( $\varphi_1$ , observed = circle, asymptotic = line) and Figure 5 ( $\varphi_2$ , observed = circle, asymptotic = line). Again, for compositions, we replace  $\log n$  by  $\log v - 1$ .

#### 1.3 First full part value of multiplicity *m*

Another variable of interest is the first *k* such that  $X_k = 1$ , i.e we are interested in the probability

$$
Pr[X_i = 0, i = 1 \cdots k - 1, X_k = 1].
$$

Note that this is the opposite situation of the  $Y_k$  case (see [HL01]), where we looked for the *first*  $k$  such that  $Y_k = 0$ . The probability is asymptotically given by

$$
\prod_{i=1}^{k-1} \left[ 1 - \frac{1}{m!} (n/2^i)^m \exp(-n/2^i) \right] \frac{1}{m!} (n/2^k)^m \exp(-n/2^k).
$$

Again, we set  $\eta := Lk - \ln n$ . This leads asymptotically, with  $\eta = O(1)$  to

$$
Pr[X_i = 0, i = 1 \cdots k - 1, X_k = 1] \sim \varphi_3(m, \eta),
$$

with

$$
\varphi_3(m,\eta) = \varphi_4(m,\eta) \frac{1}{m!} e^{-m\eta} e^{-e^{-\eta}},
$$
  

$$
\varphi_4(m,\eta) = \prod_{j=1}^{\infty} \left[ 1 - \frac{1}{m!} e^{-m(\eta - L_j)} e^{-e^{-(\eta - L_j)}} \right].
$$

Again, for compositions, we replace  $\log n$  by  $\log \nu - 1$ . Figure 6 gives  $\varphi_4(2, \eta)$  and Figure 7 gives  $\varphi_4(2, \eta)$ for large values of η. Again, this is oscillating and corresponds to  $P(2,0)$ .

#### 1.4 Asymptotic distribution of  $X(m)$

The analysis is rather similar to the one we used in [Lou87] and [HL01]. First of all we have, for any fixed  $k = O(\log n)$ ,

$$
P(m,0) \sim \varphi_4(\eta)\varphi_1(\eta).
$$

Let us choose  $k = \lfloor \log n \rfloor$ . This leads to  $\eta = -L\psi(n)$  and we obtain a periodic function of  $\psi$ :

$$
P(m,0) \sim \varphi_4[-L\psi(n)]\varphi_1[-L\psi(n)],
$$

shown in Figure 10 for  $m = 2$ . For  $n = 20000$ ,  $m = 2$ , the numerical value of  $P(2,0)$  is exactly the same as before. Now we turn to  $P(m, j) := Pr(X(m) = j)$ . We take advantage of the fact that all urns are empty *before* the first occupied urn,  $k-1$  say. Then, again with  $\eta := Lk - \ln n$ ,

$$
P(m,1) \sim \sum_{k} \varphi_3(\eta - L)\varphi_1(\eta),
$$
  

$$
P(m,2) \sim \sum_{k} \varphi_3(\eta - L)\varphi_1(\eta) \sum_{r_1 \ge k} \left\{ \frac{1}{m!} (n/2^{r_1})^m \exp(-n/2^{r_1}) \bigg/ \left[ 1 - \frac{1}{m!} (n/2^{r_1})^m \exp(-n/2^{r_1}) \right] \right\},
$$

and more generally,

$$
P(m, u+1) \sim \sum_{k} \varphi_{3}(\eta - L)\varphi_{1}(\eta).
$$
  

$$
\sum [r_{1} > r_{2} > .. > r_{u}, r_{j} \geq k] \prod_{i=1}^{u} \left\{ \frac{1}{m!} (n/2^{r_{i}})^{m} \exp(-n/2^{r_{i}}) / \left[ 1 - \frac{1}{m!} (n/2^{r_{i}})^{m} \exp(-n/2^{r_{i}}) \right] \right\}
$$

Now we set  $r_i = k + w_i, l = k - \lfloor \log n \rfloor$  and we finally derive the following theorem **Theorem 1.2.** *Set*  $\psi(n) := \log n - \lfloor \log n \rfloor$ *, then* 

$$
P(m, u+1) \sim \sum_{l=-\infty}^{\infty} \varphi_5[L(l-\psi(n))],
$$

*with*

$$
\varphi_5(\eta)=\varphi_3(\eta-L)\varphi_1(\eta).
$$

$$
\sum [w1 > w2 > \ldots > w_u, w_j \ge 0] \prod_{i=1}^u \left\{ \frac{1}{m!} e^{-m(\eta + L w_i)} e^{-e^{-(\eta + L w_i)}} / \left[ 1 - \frac{1}{m!} e^{-m(\eta + L w_i)} e^{-e^{-(\eta + L w_i)}} \right] \right\}
$$

Note that, for compositions, we obtain asymptotically  $\psi(n) = \psi(\nu)$ . We get again periodic function of  $\psi(n)$ . We give in Figure 11 and Figure 12 the sums

 $\sum_{i=0}^{3} P(2,i), \sum_{i=0}^{4} P(2,i)$ . The effect of computing  $P(2,i)$  with bounded indices (we limit the values of  $w_u$ to 16) becomes apparent at the  $10^{-7}$  precision.

Figure 13 gives  $P(m, i)$ ,  $m = 1, \ldots, 4$ , (from top to bottom to the right of  $i = 2$ ). The distributions become more concentrated as m increases.

Finally, we compare the observed distribution of  $X(2)$  with the asymptotic one in Figure 14 (observed = circle, asymptotic = line). Apart from  $i = 0$  the fit is quite good. The "Probability related values" moments given in Table 1 are computed with the distribution  $P(2, i)$ .

#### 2 Large multiplicity *m*

#### 2.1 Fixed number of parts *n*

It is now clear that large *m* are related to small integer values *i*. More precisely, the number *M<sup>i</sup>* of integers equal to *i* is asymptotically given by a Gaussian:

$$
Pr(M_i = m) \sim \exp\{-(m - n/2^i)^2/[2n/2^i(1 - 1/2^i)]\}/\sqrt{2\pi n/2^i(1 - 1/2^i)}.
$$
 (4)

The means  $n/2^i$ ,  $i = 1, 2, \ldots$  are given by  $n/2, n/4, \ldots$ , separated by  $n/4, n/8, \ldots$  which shows that the Gaussians (4) are asymptotically exponentially distinct in the sense that some common intervals, for instance  $m \in [3n/2^{i+2} - n/2^{i+3}..3n/2^{i+2} + n/2^{i+3}]$  have asymptotically small probability measures. So for any large value *m* , *only one* value

$$
i = round[log(n/m)] \tag{5}
$$

is related to *m* and  $X(m)$  has only two possible values:  $\{0,1\}$ . The following events are equivalent:  $\llbracket X_i(m) = 1 \rrbracket \equiv \llbracket M_i = m \rrbracket$ . The probability (4) is small, of order at most  $O(1/\sqrt{m})$ . Figure 15 gives  $Pr(X_i(m) = 1)$  for  $n = 2000$  (first three ranges,  $i = 1, 2, 3$ ) and Figure 16 gives the corresponding distribution functions, together with the observed values provided by a simulation of  $T = 2000$  sets (observed  $=$  $circle, asymptotic = line).$ 

An interesting check would be to recover the dominant term of the mean of  $Y : \mathbb{E}(Y) \sim \log n$ . Choose  $\tilde{j} := \alpha \log n, 0 < \alpha < 1$  which corresponds, by (5), to  $\tilde{m} = n^{1-\alpha}$ . For each  $i \leq \tilde{j}$ , by Euler–McLaurin,

$$
\sum_{m=\lfloor 3n/2^{i+2} \rfloor}^{\lfloor 3n/2^{i+1} \rfloor} \exp\{-(m-n/2^i)^2/[2n/2^i(1-1/2^i)]\}/\sqrt{2\pi n/2^i(1-1/2^i)} \sim 1,
$$

and this contributes to  $\mathbb{E}(Y)$  by  $S_1 = \tilde{j}$ . On the other side, each  $m < \tilde{m}$  contributes, by (3), with  $\frac{1}{mL}$ , with a total contribution

$$
S_2 = \frac{1}{L} \sum_{1}^{m} 1/m \sim \frac{1}{L} \ln \tilde{m}.
$$

The quantity  $S_1 + S_2 \sim \log n$  as expected.

#### 2.2 Composition of ν.

Now the number of parts  $N$  is such that (see(1))

$$
N \sim \mathcal{N}\left(\frac{\nu}{2}, \frac{\nu}{4}\right).
$$

We obtain

$$
\mathbb{E}(M_k) = \frac{v}{2} \frac{1}{2^k}.
$$
\n<sup>(6)</sup>

The asymptotic distribution of  $M_k$  is obtained as follows. We derive, setting  $\tilde{M}_k := (M_k - n/2^k)/\sqrt{v}$ ,

$$
\mathbb{E}\left[\exp[iM_k\theta/\sqrt{v}]\right] = \mathbb{E}\left[\exp[in\theta/(\sqrt{v}2^k) + i\tilde{M}_k\theta]\right]
$$
  
\n
$$
\sim \mathbb{E}\left[\exp[in\theta/(\sqrt{v}2^k) - \theta^2n/(2v2^k)(1-1/2^k)]\right]
$$
  
\n
$$
\sim \exp\left[iv\theta/(2\sqrt{v}2^k) - v\theta^2/(2v2.2^k)(1-1/2^k) + v/8[i\theta/(\sqrt{v}2^k) - \theta^2/(2v2^k)(1-1/2^k)]^2\right]
$$
  
\n
$$
\sim \exp\left[i\theta\sqrt{v}/(2.2^k) - \theta^2/2[1/(4.4^k) + 1/(2.2^k)(1-1/2^k)]\right], v \to \infty.
$$

The first term confirms (6). The second term shows that

$$
M_k \sim \mathcal{N}\left(\frac{\nu}{2}\frac{1}{2^k}, \nu \sigma_m^2\right),
$$

with

$$
\sigma_m^2 = 1/(4.4^k) + 1/(2.2^k)(1 - 1/2^k).
$$

The conclusions of Sec. 2.2 are still valid.

## 3 Conclusion

Using various techniques from analysis and probability theory, we have analyzed the stochastic properties of the *m*-distinctness of random compositions. An interesting open problem would be be to extend our results to the Carlitz compositions, where two successive parts are different (see [LP02]).

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**Fig. 1:**  $\varphi_1(m,\eta)$  for  $m = 1, \ldots, 4$ , bottom to top



**Fig.** 2:  $\varphi_1(2, \eta)$  for large negative values of  $\eta$ 



**Fig. 3:**  $\varphi_2(m,\eta)$  for  $m = 1, ..., 4$ 



**Fig. 4:** Maximum part size distribution function ( $m = 2$ , observed = circle, asymptotic = line)



**Fig. 5:** Maximum part size distribution ( $m = 2$ , observed = circle, asymptotic = line)



**Fig.** 6:  $\varphi_4(2,\eta)$ 



**Fig.** 7:  $\varphi_4(2, \eta)$  for large values of  $\eta$ 



**Fig. 8:**  $\varphi_3(2, \eta)$  for  $m = 1, ..., 4$ .







**Fig.** 11:  $\sum_{i=0}^{3} P(2, i)$ 

**Fig.** 12:  $\sum_{i=0}^{4} P(2,i)$ 





**Fig. 15:**  $Pr(X_i(m) = 1), i = 1, ..., 3$ 

0 400 600 800 1000<br>x

0.1

<sub>0</sub>+

200

0.005

0.01

0.015

0.02

0.025

2

**Fig. 13:**  $P(m, i), m = 1, ..., 4$ 

 $0.2$ 0.3

0.4

0.5

 $0.6$ 

**Fig. 16:** Distribution function of  $M_i$ ,  $i = 1,...,3$  (observed = circle, asymptotic = line)  $asymptotic = line)$ 

 400 600 800 1000 x

200

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