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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,  $v \to \infty$ :

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

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>lt;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  $\log n$ , 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  $\log v - 1$  (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 v has multiplicity m. (The corresponding problem for random partitions has been analyzed in Corteel et al. [CPSW99]). Of course, here,

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

where we explicitly show the dependence on v. But, as already mentioned, Y(v) has asymptotically the same distribution as Y (with log *n* replaced by log v - 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.:  $X_i(m) := [value \ i \ appears \ among \ the \ n \ GEOM(1/2) \ R.V.$  with multiplicity m, for fixed m = O(1)]. Of course,

$$\Pr[X_i(m) = 1] = \binom{n}{m} (1/2^i)^m (1 - 1/2^i)^{n-m}.$$
(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  $G^*(s)$  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>

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

This leads, by converse mapping, to

$$G(n,m) \sim \frac{1}{mL} + \beta_0(\log n) + O(1/n),$$
 (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  $n \sim \mathcal{N}(\frac{\nu}{2}, \frac{\nu}{4})$ , we just have to replace log *n* by log  $\nu - 1$ . 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  $Y_i$ , 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, [SČ64] 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

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

<sup>&</sup>lt;sup>‡</sup> The symbol  $\asymp$  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  $\Pi_n = \frac{n!}{2\pi i} \int_{\Gamma} \exp\{nf(\tau)\} d\tau/\tau$ , 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 fi xed 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,  $\Pi_n \sim \frac{n! e^{nf(\tau^*)}}{\sqrt{2\pi\tau^*}\sqrt{nf''(\tau^*)}}$ , and, by Stirling, we easily derive the theorem.

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,

$$\begin{aligned} \text{VAR}(X(m)) &\sim \quad \frac{1}{m!} \sum_{1}^{\infty} (n/2^i)^m \exp(-n/2^i) \left[ 1 - \frac{1}{m!} (n/2^i)^m \exp(-n/2^i) \right] \\ &\sim \quad \int_0^\infty e^{-y} \frac{y^m}{m!} (1 - e^{-y} \frac{y^m}{m!}) \frac{dy}{Ly} + \beta_1 (\log_2 n) \\ &= \quad \frac{1}{mL} - \frac{(2m-1)!}{Lm!^2 2^{2m}} + \beta_1 (\log_2 n). \end{aligned}$$

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

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

and fi nally, the centered moments are given by

$$\begin{split} \tilde{\sigma}^2 &:= \operatorname{VAR}(X(m)) \quad \sim \quad \frac{1}{mL} - \frac{(2m)!}{2Lm!^2 2^{2m}m}, \\ \tilde{\mu}_3 &:= \mu_3(X(m)) \quad \sim \quad \frac{1}{mL} - \frac{3(2m)!}{2Lm!^2 2^{2m}m} + \frac{2(3m)!}{3Lm!^3 3^{3m}m}, \\ \tilde{\mu}_4 &:= \mu_4(X(m)) \quad \sim \quad \frac{1}{mL} + \frac{3}{m^2 L^2} - \frac{3(4m)!}{2Lm!^4 4^{4m}m} + \frac{4(3m)!}{Lm!^3 3^{3m}m} \\ &- \quad \frac{7(2m)!}{2Lm!^2 2^{2m}m} - \frac{3(2m)!}{3L^2m!^2 2^{2m}m^2} + \frac{3(2m)!^2}{4L^2m!^4 2^{4m}m^2} \end{split}$$

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  $\mathcal{M}_n(m)$  of multiplicity *m* is such that

$$\Pr(\mathcal{M}_n(m) < k) \sim \prod_{i=k}^{\infty} \left[ 1 - \frac{1}{m!} (n/2^i)^m \exp(-n/2^i) \right].$$

	Theoretical asymptotic value	Observed value	Probability related value
mean	.7213	.7345	.7214
variance	.5861	.5945	.5863
$\mu_3$	.3750	.3752	.3752
$\mu_4$	1.1197	1.1341	1.1198

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

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

$$\Pr(\mathcal{M}_n(m) < k) \sim \varphi_1(m, \eta),$$

with

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

Figure 1 gives  $\varphi_1(m,\eta)$  for m = 1,...,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, ..., 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

$$\begin{split} \phi_3(m,\eta) &= \phi_4(m,\eta) \frac{1}{m!} e^{-m\eta} e^{-e^{-\eta}}, \\ \phi_4(m,\eta) &= \prod_{j=1}^{\infty} \left[ 1 - \frac{1}{m!} e^{-m(\eta - L_j)} e^{-e^{-(\eta - L_j)}} \right]. \end{split}$$

Again, for compositions, we replace  $\log n$  by  $\log v - 1$ . Figure 6 gives  $\varphi_4(2,\eta)$  and Figure 7 gives  $\varphi_4(2,\eta)$  for large values of  $\eta$ . 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  $\eta := Lk - \ln n$ ,

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

$$P(m,2) \sim \sum_{k} \varphi_{3}(\eta - L)\varphi_{1}(\eta) \sum_{r_{1} \geq k} \left\{ \frac{1}{m!} (n/2^{r_{1}})^{m} \exp(-n/2^{r_{1}}) \left/ \left[ 1 - \frac{1}{m!} (n/2^{r_{1}})^{m} \exp(-n/2^{r_{1}}) \right] \right\},$$

and more generally,

$$P(m, u+1) \sim \sum_{k} \varphi_{3}(\eta - L) \varphi_{1}(\eta).$$
  
$$\sum_{k} [r_{1} > r_{2} > ... > r_{u}, r_{j} \ge k] \prod_{i=1}^{u} \left\{ \frac{1}{m!} (n/2^{r_{i}})^{m} \exp(-n/2^{r_{i}}) \middle/ \left[ 1 - \frac{1}{m!} (n/2^{r_{i}})^{m} \exp(-n/2^{r_{i}}) \right] \right\}$$

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

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

$$\sum \left[ w_1 > w_2 > \ldots > w_u, w_j \ge 0 \right] \prod_{i=1}^u \left\{ \frac{1}{m!} e^{-m(\eta + Lw_i)} e^{-e^{-(\eta + Lw_i)}} \middle/ \left[ 1 - \frac{1}{m!} e^{-m(\eta + Lw_i)} e^{-e^{-(\eta + Lw_i)}} \right] \right\}$$

Note that, for compositions, we obtain asymptotically  $\psi(n) = \psi(v)$ . 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<sup>-7</sup> precision.

Figure 13 gives P(m, i), m = 1, ..., 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*

n

#### 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  $M_i$  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, ... are given by n/2, n/4, ..., separated by n/4, n/8, ... 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

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

is related to *m* and X(m) has only two possible values:  $\{0,1\}$ . The following events are equivalent:  $[X_i(m) = 1] \equiv [M_i = m]$ . 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,

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

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

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

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

#### 2.2 Composition of v.

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

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

We obtain

$$\mathbb{E}(M_k) = \frac{\nu}{2} \frac{1}{2^k}.$$
(6)

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

$$\begin{split} \mathbb{E}\left[\exp[iM_{k}\theta/\sqrt{\nu}]\right] &= \mathbb{E}\left[\exp[in\theta/(\sqrt{\nu}2^{k}) + i\tilde{M}_{k}\theta]\right] \\ &\sim \mathbb{E}\left[\exp[in\theta/(\sqrt{\nu}2^{k}) - \theta^{2}n/(2\nu2^{k})(1 - 1/2^{k})]\right] \\ &\sim \exp\left[i\nu\theta/(2\sqrt{\nu}2^{k}) - \nu\theta^{2}/(2\nu2.2^{k})(1 - 1/2^{k}) + \nu/8[i\theta/(\sqrt{\nu}2^{k}) - \theta^{2}/(2\nu2^{k})(1 - 1/2^{k})]^{2}\right] \\ &\sim \exp\left[i\theta\sqrt{\nu}/(2.2^{k}) - \theta^{2}/2[1/(4.4^{k}) + 1/(2.2^{k})(1 - 1/2^{k})]\right], \nu \to \infty. \end{split}$$

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

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

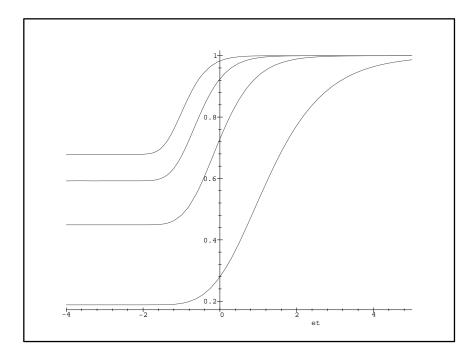
The conclusions of Sec. 2.2 are still valid.

## 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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The pertinent comments of the referees led to substantial improvements in the presentation.



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

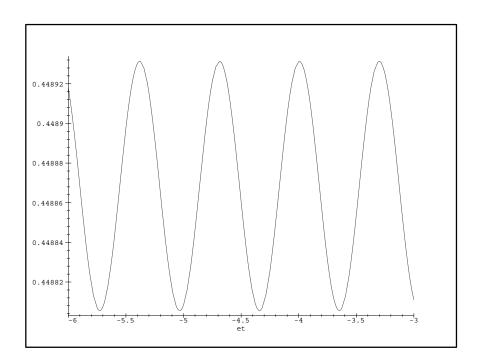
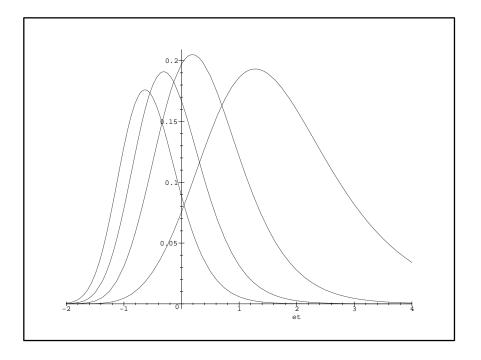


Fig. 2:  $\phi_1(2,\eta)$  for large negative values of  $\eta$ 



**Fig. 3:**  $\phi_2(m, \eta)$  for m = 1, ..., 4

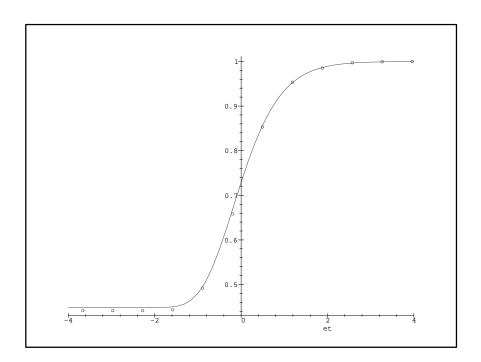
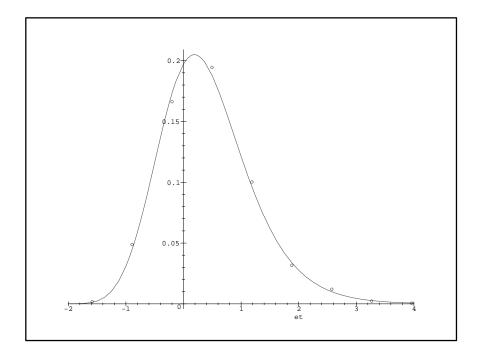


Fig. 4: Maximum part size distribution function (m = 2, observed = circle, asymptotic = line)



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

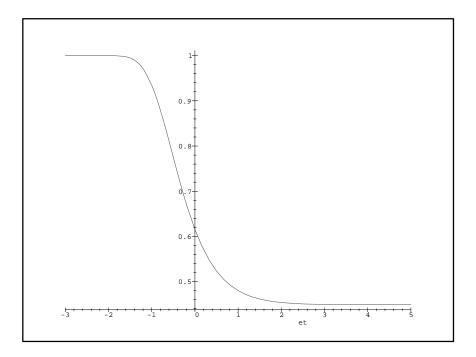


Fig. 6:  $\phi_4(2,\eta)$ 

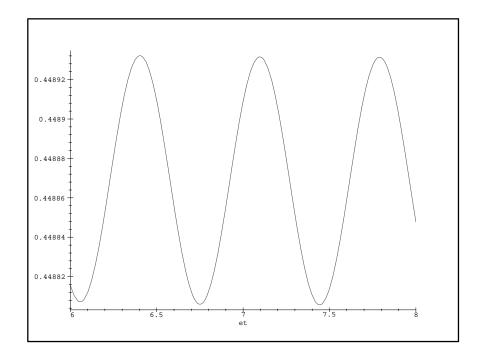
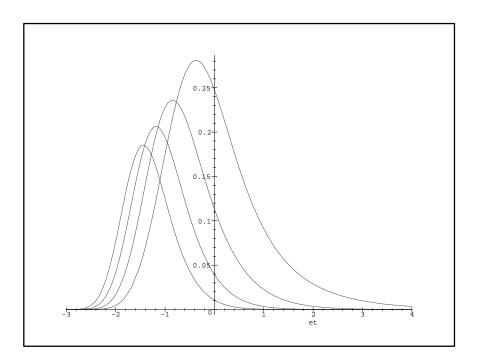
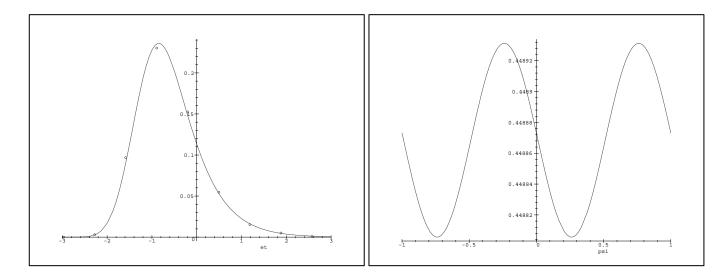


Fig. 7:  $\phi_4(2,\eta)$  for large values of  $\eta$ 

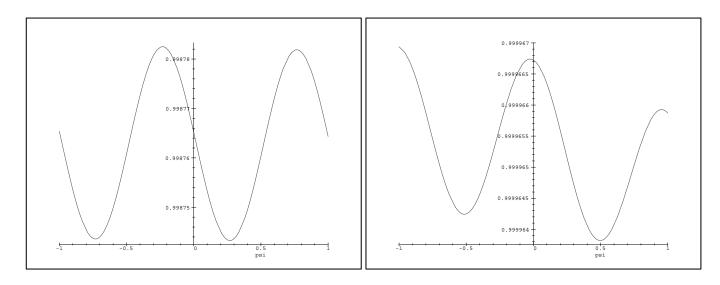


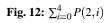
**Fig. 8:**  $\phi_3(2,\eta)$  for m = 1, ..., 4.

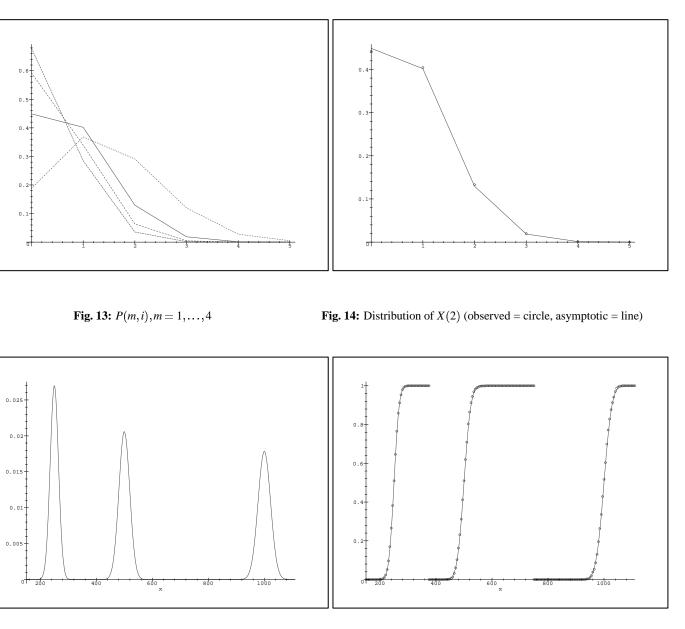


**Fig. 9:** First full part distribution (m = 2, observed = circle, asymptotic = line)

**Fig. 10:** P(2,0) as a function of  $\psi$ 







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

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