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## Some Probabilistic Aspects of Set Partitions

## Jim Pitman

**1. INTRODUCTION.** A partition of the set  $\mathbb{N}_n := \{1, 2, ..., n\}$  is an unordered collection of non-empty subsets of  $\mathbb{N}_n$ . Let  $\mathbb{P}_n$  denote the set of all such partitions, and let  $B_n = \#(\mathbb{P}_n)$ , the number of partitions of  $\mathbb{N}_n$ . The numbers  $B_n$  are known as Bell numbers after E.T. Bell [3, 4, 45]. See Rota [50] and Gardner [24, Chapter 2] for surveys of their properties and applications. The remarkable Dobiński formula [18]

$$B_n = e^{-1} \sum_{m=1}^{\infty} \frac{m^n}{m!} \qquad (n = 1, 2, \dots)$$
 (1)

leads [36, 1.9] to the asymptotic evaluation

$$B_n \sim \frac{1}{\sqrt{n}} \lambda(n)^{n+1/2} e^{\lambda(n)-n-1}$$
 as  $n \to \infty$ , (2)

where  $\lambda(n)\log(\lambda(n)) = n$ . As noted by Comtet [11], for each n the infinite sum in (1) can be evaluated as the least integer greater than the sum of the first 2n terms.

From a probabilistic perspective, the series on the right side of Dobiński's formula represents the nth moment of the Poisson distribution with mean 1. So the initially surprising fact that this series yields an integer for all n amounts to the fact that all positive integer moments of the Poisson(1) distribution are integers. As explained in Section 2, Dobiński's formula reduces to the fact that the factorial moments of the Poisson(1) distribution are identically equal to 1, and this identity can be understood probabilistically with essentially no calculation.

While such probabilistic interpretations of identities related to set partitions are the main theme of this paper, Section 1.2 recalls an elementary combinatorial proof of Dobiński's formula.

**1.1 Notation.** Following the notation of [27], let  $\binom{n}{k}$  denote the number of partitions of  $\mathbb{N}_n$  into exactly k distinct non-empty subsets, so that

$$B_n = \sum_{k=1}^n \binom{n}{k}. \tag{3}$$

The  $\binom{n}{k}$  are known as the Stirling numbers of the second kind. Let  $m^k$  denote the falling factorial with k factors

$$m^{\underline{k}} = m(m-1)\cdots(m-k+1), \tag{4}$$

which, for positive integers m and k, is the number of permutations of length k of

m distinct symbols. The formula

$$m^n = \sum_{k=1}^n \binom{n}{k} m^{\underline{k}} \tag{5}$$

decomposes the number  $m^n$  of sequences of m distinct symbols of length n as the sum over k of the number of such sequences that contain exactly k distinct symbols [54, p. 35]. As an identity of polynomials in m of degree n, this identity provides an alternative definition of the coefficients  $\binom{n}{k}$  for  $1 \le k \le n$ . See [11, 47, 48, 54] for background and a wealth of further information about Stirling numbers.

**1.2** A quick proof of Dobiński's formula. This argument is attributed to Schützenberger by Foata [22, p. 73]. Divide (5) by m! to obtain for positive integers m and n

$$\frac{m^n}{m!} = \sum_{k=1}^m \binom{n}{k} \frac{1}{(m-k)!}.$$
 (6)

This is the identity of coefficients of  $\lambda^m$  in the power series identity

$$\sum_{m=1}^{\infty} \frac{m^n}{m!} \lambda^m = \left(\sum_{k=1}^n \binom{n}{k} \lambda^k\right) \left(\sum_{j=0}^{\infty} \frac{\lambda^j}{j!}\right),\tag{7}$$

which, upon rearrangement, gives the following horizontal generating function for the Stirling numbers of the second kind:

$$\sum_{k=1}^{n} {n \choose k} \lambda^k = e^{-\lambda} \sum_{m=1}^{\infty} \frac{m^n}{m!} \lambda^m.$$
 (8)

Now take  $\lambda = 1$  and use (3) to obtain Dobiński's formula (1). The polynomial appearing in (8) is known as an *exponential polynomial*. Many other proofs of the generalization (8) of Dobiński's formula are known. See for instance Roman [49, p. 66] and Wilf [58, p. 106]. Closely related arguments appear in Rota [50], Berge [5, p. 44], Comtet [11, p. 211], Lupas [37], and Chen-Yeh [10].

2. MOMENTS. For a non-negative integer-valued random variable X with

$$P(X = m) = p_m \qquad (m = 0, 1, ...)$$
 (9)

and a non-negative function f, let

$$E[f(X)] := \sum_{m} p_m f(m), \tag{10}$$

which is the *expected value of* f(X) for X with distribution (9). See [20, 43] for background. From (5) and linearity of the expectation operator E, we obtain the following well-known formula for  $E[X^n]$ , the *n*th moment of X, in terms of  $E[X^k]$ , the *k*th factorial moment of X for  $1 \le k \le n$  [47, 14]:

$$E[X^n] = \sum_{k=1}^n {n \choose k} E[X^{\underline{k}}]. \tag{11}$$

For  $\lambda > 0$ , let  $X_{\lambda}$  denote a random variable with the Poisson distribution

$$P(X_{\lambda} = m) = e^{-\lambda} \frac{\lambda^{m}}{m!}$$
  $(m = 0, 1, ...)$  (12)

so that

$$E[f(X_{\lambda})] = e^{-\lambda} \sum_{m=0}^{\infty} f(m) \frac{\lambda^{m}}{m!}.$$
 (13)

Take  $f(m) = m^n$  to see that the right side of (8) equals  $E[X_{\lambda}^n]$ , so the identity (8) amounts to the formula

$$E[X_{\lambda}^{n}] = \sum_{k=1}^{n} {n \choose k} \lambda^{k} \qquad (n = 1, 2, \dots)$$

$$(14)$$

for the moments of the  $Poisson(\lambda)$  distribution [46, 42]. This formula is the particular case of (11) for X with  $Poisson(\lambda)$  distribution, for it is known [46, 14] that

$$E[X_{\lambda}^{\underline{k}}] = \lambda^{k} \qquad (k = 1, 2, \dots). \tag{15}$$

Formula (15) follows easily from (13) with  $f(m) = m^k$  by change of summation variable from m to j = m - k. In particular, for  $\lambda = 1$  the factorial moments of the Poisson(1) distribution are identically equal to 1. So Dobiński's formula (1) can be read from (14) for  $\lambda = 1$ , which follows as indicated above from (11) and (15). In essence, this is Rota's [50] proof of Dobiński's formula cast in probabilistic notation. This argument differs from the proof in Section 1.2 in that it involves checking (15) for  $\lambda = 1$ .

Formula (15) has the following interpretation in terms of a *Poisson process* [33, 43]. Let

$$0 < U_{(1)} < \dots < U_{(X_{\lambda})} < 1 \tag{16}$$

denote the random locations in (0,1) of the points of a homogeneous Poisson process on (0,1) with mean intensity measure  $\lambda du$  for 0 < u < 1. For each  $k = 1, 2, \ldots$  define an associated k-tuple point process, with points in  $(0,1)^k$ , to have a point at each of the locations  $(U_{(\sigma_1)}, \ldots, U_{(\sigma_k)})$  as  $\sigma$  ranges over the  $X_{\lambda}^k$  different permutations of  $\{1, \ldots, X_{\lambda}\}$  of length k. For distinct  $u_i \in (0, 1)$ , independence properties of the basic Poisson process on (0, 1) imply that the mean intensity of the k-tuple point process at  $(u_1, \ldots, u_k) \in (0, 1)^k$  is

$$\frac{P(\text{some } U_{(\sigma_i)} \in du_i \text{ for each } 1 \le i \le k)}{du_1 \cdots du_k} = \frac{(\lambda du_1) \cdots (\lambda du_k)}{du_1 \cdots du_k} = \lambda^k, \quad (17)$$

so the expected number of points in the k-tuple point process is

$$E[X_{\lambda}^{\underline{k}}] = \lambda^k \int_0^1 du_1 \cdots \int_0^1 du_k = \lambda^k.$$
 (18)

Constantine and Savits [12] derive a generalization of Dobiński's formula by consideration of compound nonhomogeneous Poisson processes. See also Stam [52] and Di Bucchianico [8] for related results. For various applications of Stirling numbers and their generalizations to the computation of moments of probability distributions, see [47, 9]. Moments of the normal distribution also have interesting combinatorial interpretations [19, 25]. More generally, the idea of representing combinatorially defined numbers by an infinite sum or an integral, typically with a probabilistic interpretation, has proved to be a very fruitful one. Other examples are the representation of n! as a gamma integral, which leads to Stirling's formula [7, 16, 38], and Laplace's representation of kth differences of powers [35, 14, 30], which yields an asymptotic formula for the Stirling numbers of the second kind. See [41] for a recent survey of asymptotic enumeration methods.

**3. VARIATIONS OF DOBIŃSKI'S FORMULA.** The derivation of Dobiński's formula given in the previous section yields the following proposition:

**Proposition 1.** Let X be a random variable with values in  $\{0, 1, 2, ...\}$  and let n be a positive integer. The following two conditions are equivalent:

- (i) the first n factorial moments of X are identically equal to 1;
- (ii) the kth moment of X equals  $B_k$  for every  $1 \le k \le n$ .

It is well known that for each  $\lambda > 0$  the Poisson( $\lambda$ ) distribution is uniquely determined by its moments; see for instance [6, Section 30]. The Poisson(1) distribution is therefore the unique probability distribution whose nth moment equals  $B_n$  for every n. But for each fixed n there are many probability distributions on  $\{0, 1, 2, ...\}$  that have the same first n moments as Poisson(1). It is obvious that there can be at most one such distribution of X with  $P(X \le n) = 1$ , because the moment conditions amount to a system of n linearly independent equations in n unknowns  $p_1, \ldots, p_n$ . Less obvious is the fact that the unique solution of these equations is such that  $p_i \ge 0$  for  $1 \le i \le n$  and  $\sum_{i=1}^n p_i \le 1$ , so that  $(p_1, \ldots, p_n)$  is the restriction to  $\{1, \ldots, n\}$  of a unique probability distribution on  $\{0, 1, \ldots, n\}$ . But this probability distribution on  $\{0, 1, 2, \dots, n\}$ , whose first n factorial moments are identically equal to one, is known to arise in the setting of the classical matching problem [31, 14, 20, 56]. If  $M_n$  is the number of fixed points of a uniformly distributed random permutation of  $\mathbb{N}_n$ , then it is easy to show by the method of indicators that the first n factorial moments of  $M_n$  are identically equal to 1; see [14]. The distribution of a random variable X with range  $\{0, 1, \ldots, n\}$  is recovered from its factorial moments by the classical sieve formula [14]

$$P(X=m) = \frac{1}{m!} \sum_{k=m}^{n} \frac{(-1)^{m-k} E[X^{\underline{k}}]}{(m-k)!} \quad (m=0,1,\ldots,n).$$
 (19)

For  $X = M_n$  with  $E[M_n^{\underline{k}}] \equiv 1$  for  $0 \le k \le n$ , this simplifies to

$$P(M_n = m) = \frac{1}{m!} \sum_{s=0}^{n-m} \frac{(-1)^s}{s!} \quad (m = 0, 1, ..., n).$$
 (20)

See [20, Section IV.4] for further discussion. According to Proposition 1, the kth moment of  $M_n$  equals  $B_k$  for every  $1 \le k \le n$ . That is to say,

$$B_k = \sum_{m=1}^n \frac{m^k}{m!} \sum_{s=0}^{n-m} \frac{(-1)^s}{s!} \qquad (1 \le k \le n).$$
 (21)

This variation of Dobiński's formula is derived in quite a different way by Wilf [58, p. 22] by substituting the classical formula

into (3). As observed by Wilf, Dobiński's formula (1) follows easily from (21) by letting  $n \to \infty$ . See also Lovász [36, 1.9] for a similar argument, James and Kerber [32, pp. 227–237] for connections with the representation theory of the symmetric group, and Diaconis and Shashahani [17] for various generalizations. In Dale and Skau [13] the Bell numbers appear as the factorial moments of a probability distribution on the non-negative integers.

**4. THE MOMENT GENERATING FUNCTION.** Consider now the *moment generating function* (m.g.f.) of the Poisson $(\lambda)$  distribution:

$$E\left[\exp(\theta X_{\lambda})\right] = E\left[\sum_{n=0}^{\infty} \frac{\theta^{n} X_{\lambda}^{n}}{n!}\right] = \sum_{n=0}^{\infty} E\left[X_{\lambda}^{n}\right] \frac{\theta^{n}}{n!},$$
 (23)

where the series converge for all real  $\theta$  and the interchange of E and  $\Sigma$  is easily justified. See [6] for a modern treatment of m.g.f's. From (13) with  $f(m) = e^{\theta m}$  there is the standard formula

$$E[\exp(\theta X_{\lambda})] = e^{-\lambda} \sum_{m=0}^{\infty} \frac{\left(\lambda e^{\theta}\right)^{m}}{m!} = \exp(\lambda(e^{\theta} - 1)).$$
 (24)

This combines with (8) to yield the following double generating function of the Stirling numbers of the second kind. This classical formula [11, p. 206] is an identity between two different expressions for the m.g.f. in  $\theta$  of the Poisson( $\lambda$ ) distribution:

$$1 + \sum_{n=1}^{\infty} \sum_{k=1}^{n} {n \choose k} \frac{\lambda^k \theta^n}{n!} = \exp(\lambda (e^{\theta} - 1)).$$
 (25)

In particular, for  $\lambda = 1$  this reduces by (3) to Bell's [3, 4] formula

$$1 + \sum_{n=1}^{\infty} B_n \frac{\theta^n}{n!} = \exp(e^{\theta} - 1), \tag{26}$$

which gives two expressions for the m.g.f. in  $\theta$  of the Poisson(1) distribution. Equating coefficients of  $\lambda^k$  in (25) yields the *vertical generating function* of the Stirling numbers of the second kind:

$$\sum_{n>k} {n \choose k} \frac{\theta^n}{n!} = \frac{1}{k!} (e^{\theta} - 1)^k \qquad (k = 1, 2, ...).$$
 (27)

See [11,47,54] for alternative derivations of these identities. There are similar identities for many other arrays of combinatorial numbers, such as the binomial coefficients and Stirling numbers of the first kind [11,58], [27, p. 351], most of which admit probabilistic interpretations. Formulae with binomial coefficients typically involve independent trials, while those with Stirling numbers of the first kind typically involve the cycle structure of random permutations [1]. See also [2] for probabilistic analysis of more general combinatorial structures and further references.

**5. RANDOM PARTITIONS.** A random partition of  $\mathbb{N}_n$  is a random variable  $\Pi$  with values in the set  $\mathbb{P}_n$  of partitions of  $\mathbb{N}_n$ . The distribution of  $\Pi$  then refers to the collection of probabilities  $P(\Pi = \pi)$  as  $\pi$  ranges over  $\mathbb{P}_n$ . Questions about enumeration of partitions of  $\mathbb{N}_n$  of various kinds can be phrased probabilistically in terms of a uniform random partition, that is, a random partition  $\Pi$  with the uniform distribution  $P(\Pi = \pi) = 1/B_n$  for each partition  $\pi \in \mathbb{P}_n$ . For developments of this idea see [29, 28, 51, 23]. Random partitions with non-uniform distribution also arise naturally in various contexts, so it is useful to have models for random partitions, both uniform and non-uniform.

The following random allocation scheme provides a basic method of generating a random partition of  $\mathbb{N}_n$ . See [14, 34, 57] for extensive study of this and related schemes, and further references. Throw n balls labelled by  $\mathbb{N}_n$  into m boxes

labelled by  $\mathbb{N}_m$ , and assume that all  $m^n$  possible allocations of balls into boxes are equally likely. Let  $\Pi_{nm}$  be the partition of balls by boxes. More formally, let  $X_i$  be the number of the box containing the ith ball for  $1 \le i \le n$ . Then the  $X_i$  are independent and uniformly distributed on  $\mathbb{N}_m$ , and  $\Pi_{nm}$  is the partition of  $\mathbb{N}_n$  induced by the random equivalence relation  $i \sim j$  if and only if  $X_i = X_j$ . Formally, the  $X_i$  can be regarded as coordinate maps defined on  $(\mathbb{N}_n)^m$ , and  $\Pi_{nm}$  is then defined as a map from  $(\mathbb{N}_n)^m$  to  $\mathbb{P}_n$ , the set of partitions of  $\mathbb{N}_n$ . Let  $\#(\pi)$  denote the number of subsets in a partition  $\pi \in \mathbb{P}_n$ . The distribution of  $\Pi_{nm}$  induced by the uniform distribution P on  $\mathbb{N}_m$  can be read from formula (5):

$$P(\Pi_{nm} = \pi) = \frac{m^{\underline{k}}}{m^n}$$
 if  $\#(\pi) = k$ . (28)

The distribution of  $\#(\Pi_{nm})$ , the number of occupied boxes when n balls are thrown into m boxes, is given by the following probabilistic equivalent of (5):

$$P[\#(\Pi_{nm}) = k] = {n \choose k} \frac{m^{\underline{k}}}{m^n} \qquad (1 \le k \le n).$$
 (29)

Because the probability displayed in (28) depends on the number of occupied boxes k, for  $n \geq 3$  this random partition  $\Pi$  of  $\mathbb{N}_n$  does not have uniform distribution on  $\mathbb{P}_n$  for any m. However, as observed by Stam [53], for each fixed n it is possible to generate a uniformly distributed random element of  $\mathbb{P}_n$  by a suitable randomization of m. The following proposition was suggested by Stam's construction, which is described in Corollary 3.

**Proposition 2.** Let M be a random variable with values in  $\{1, 2, ...\}$ , and suppose given M = m that n balls labelled by  $\mathbb{N}_n$  are thrown independently and uniformly at random into m boxes. Let  $\Pi_{nM}$  denote the random partition of  $\mathbb{N}_n$  so generated. The following two conditions are equivalent:

- (i)  $\Pi_{nM}$  has uniform distribution over the set  $\mathbb{P}_n$  of all partitions of  $\mathbb{N}_n$ ;
- (ii) the distribution of M is of the form

$$P(M=m) = \frac{m^n p_m}{B_n}$$
  $(m=1,2,...)$  (30)

for some probability distribution  $(p_m)$  on  $\{0, 1, 2, ...\}$  whose first n factorial moments are identically equal to 1.

Before the proof, here are two corollaries which follow immediately from the Proposition and the discussion in Sections 2 and 3:

**Corollary 3.** [53] If M has the distribution (30) for  $p_m = e^{-1}/m!$ , then  $\Pi_{nM}$  has uniform distribution on  $\mathbb{P}_n$ .

Corollary 4. For each n there is a unique distribution of M such that

$$P(M \le n) = 1$$
 and  $\Pi_{nM}$  has uniform distribution on  $\mathbb{P}_n$ .

This distribution of M is defined by (30) for  $p_m = P(M_n = m)$  as in (20) with  $M_n$  the number of fixed points of a uniform random permutation of  $\mathbb{N}_n$ .

*Proof of Proposition 2.* By conditioning on M and using (28),

$$P(\Pi_{nM} = \pi) = \sum_{m=1}^{\infty} \frac{m^{\underline{k}}}{m^n} P(M = m) \quad \text{if } \#(\pi) = k,$$
 (31)

so the distribution of  $\Pi_{nM}$  is uniform on  $\mathbb{P}_n$  if and only if

$$\sum_{m=1}^{\infty} \frac{m^k}{m^n} P(M=m) = \frac{1}{B_n} \qquad (1 \le k \le n).$$
 (32)

Define

$$p_m = B_n m^{-n} P(M = m) \qquad (m = 1, 2, ...),$$
 (33)

so that (32) becomes

$$\sum_{m=1}^{\infty} m^{k} p_{m} = 1 \qquad (1 \le k \le n), \tag{34}$$

which for k=1 implies that  $\sum_{m=1}^{\infty} p_m \leq \sum_{m=1}^{\infty} m p_m = 1$ . It follows that  $\Pi_{nM}$  is uniform if and only if  $(p_m)$  derived from the distribution of M via (33) is the restriction to  $\{1,2,\ldots\}$  of a probability distribution on  $\{0,1,2,\ldots\}$  whose first n factorial moments are equal to 1. This is condition (ii).

As shown by Stam, Corollary 3 allows numerous results regarding the asymptotic distribution for large n of a uniform random partition of  $\mathbb{N}_n$  to be deduced from corresponding results for the classical occupancy problem defined by random allocations of balls in boxes, for which see [34,57]. See also [2,15,23,26,28,29,51] for a more detailed account of the asymptotics of uniform random partitions of  $\mathbb{N}_n$ .

As a variation, the following corollary is easily obtained by a similar argument:

## **Corollary 5.** Suppose that M has the distribution

$$P(M=m) = \frac{m^n P(X_{\lambda}=m)}{\mu_n(\lambda)} \qquad (m=1,2,\dots), \tag{35}$$

where  $X_{\lambda}$  has the Poisson( $\lambda$ ) distribution (12), and  $\mu_n(\lambda) = E(X_{\lambda}^n)$ . Then the distribution of  $\Pi_{nM}$  is given by

$$P(\Pi_{nM} = \pi) = \frac{\lambda^k}{\mu_n(\lambda)} \quad \text{if } \#(\pi) = k. \tag{36}$$

As a check, (36) implies

$$P[\#(\Pi_{nM}) = k] = {n \choose k} \frac{\lambda^k}{\mu_n(\lambda)} \qquad (1 \le k \le n).$$
 (37)

The fact that these probabilities sum to 1 amounts to formula (14) for  $\mu_n(\lambda)$ . The distribution of  $\Pi_{nM}$  defined by formula (36) defines a *Gibbs distribution* on partitions of  $\mathbb{N}_n$ . See [55, 44] for further discussion of such Gibbs distributions on sets of combinatorial objects. See Nijenhuis and Wilf [39] for a recursive algorithm to construct a uniform random partition of  $\mathbb{N}_n$  based on the recurrence

$$B_n = 1 + \sum_{k=1}^{n-1} {n-1 \choose k} B_k, \tag{38}$$

where the right side counts the number of partitions  $\pi$  of  $\mathbb{N}_n$  according to the size k of the subset in  $\pi$  that contains n [36, Problem 1.10]. See [40] for related

combinatorial algorithms, and [21] for a recent systematic approach to the random generation of labelled combinatorial structures and further references on this topic.

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