## Exploiting the Feller Coupling for the Ewens Sampling Formula

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We congratulate Harry Crane on a masterful survey, showing the universal character of the Ewens sampling formula.

There are two grand ways to get a simple handle on the Ewens sampling formula; one is the Chinese restaurant coupling, and the other is the Feller coupling. Since Crane has discussed the Chinese Restaurant process, but not the Feller coupling, we will give a brief survey of the latter.

The Ewens sampling formula, given in Crane's (1), has an interpretation in terms of the cycle type of a random permutation of *n* objects. For  $\theta = 1$ , it is just Cauchy's formula, expressed in terms of the *fraction* of permutations of *n* objects that have exactly  $m_i$  cycles of order *i*,  $1 \le i \le n$ . For general  $\theta$ , the power

$$\theta^{m_1+m_2+\cdots+m_n} = \theta^K$$

appearing in the formula, where *K* denotes the number of cycles, biases the uniform random choice of a permutation by weighting with the factor  $\theta^{K}$ , the remaining factors involving  $\theta$  merely reflecting the new normalization constant required to specify a probability distribution. We use the notation  $(C_1(n), \ldots, C_n(n))$ to denote a random object distributed according to the Ewens sampling formula, suppressing the parameter  $\theta$  but making explicit the parameter *n*, so that, with Crane's notation (1),

(1)  
$$\mathbb{P}(C_1(n) = m_1, \dots, C_n(n) = m_n)$$
$$= p(m_1, \dots, m_n; \theta).$$

The Feller coupling, motivated by the example in Feller ([6], page 815) is defined as follows. Take independent Bernoulli random variables  $\xi_i$ , i = 1, 2, 3, ..., with the simple odds ratios  $\mathbb{P}(\xi_i = 0)/\mathbb{P}(\xi_i = 1) = (i - 1)/\theta$ . Thus,  $\mathbb{E}\xi_i = \mathbb{P}(\xi_i = 1) = \theta/(\theta + i - 1)$ , and  $\mathbb{P}(\xi_i = 0) = (i - 1)/(\theta + i - 1)$ . Say that an  $\ell$ -spacing occurs in a sequence  $a_1, a_2, ...$ , of zeros and ones, starting at position  $i - \ell$  and ending at position i, if  $a_{i-\ell}a_{i-\ell+1}\cdots a_{i-1}a_i = 10^{\ell-1}1$ , a one followed by  $\ell - 1$  zeros followed by another one. Then if, for each  $\ell \ge 1$ , we define

 $C_{\ell}(n) :=$  the number of  $\ell$ -spacings in

 $\xi_1, \xi_2, \ldots, \xi_{n-1}, \xi_n, 1, 0, 0, \ldots,$ 

the joint distribution of  $C_1(n), \ldots, C_n(n)$  is the Ewens sampling formula, as per Crane's (1) and our (1). This can be seen directly, for the case  $\theta = 1$ : consider a random permutation of 1 to *n*, write the canonical cycle notation one symbol at a time, and let  $\xi_i$  indicate the decision to complete a cycle, when there is an *i*-way choice of which element to assign next. The general case  $\theta > 0$  follows by biasing, with respect to  $\theta^K$ : since  $K = \xi_1 + \cdots + \xi_n$ , and the  $\xi_1, \ldots, \xi_n$ are independent, biasing their joint distribution by  $\theta^{\xi_1 + \cdots + \xi_n} = \theta^{\xi_1} \cdots \theta^{\xi_n}$  preserves their independence and Bernoulli distributions, while changing the odds  $\mathbb{P}(\xi_i = 0)/\mathbb{P}(\xi_i = 1)$  from (i - 1)/1 to  $(i - 1)/\theta$ .

Now, the wonderful thing that happens is that, with  $Y_{\ell}$  *defined* to be the number of  $\ell$ -spacings in the infinite sequence  $\xi_1, \xi_2, \ldots$ , it turns out that  $Y_1, Y_2, \ldots$  are mutually independent, and that  $Y_{\ell}$  is Poisson distributed, with  $\mathbb{E}Y_{\ell} = \theta/\ell$ , as in formula (11) in Section 3.8. This shows that the Ewens sampling formula is closely related to the simpler independent process  $Y_1, Y_2, \ldots, Y_n$ . Explicitly, let  $R_n$  be the position of the rightmost one in  $\xi_1, \xi_2, \ldots, \xi_{n-1}, \xi_n$ —noting that always  $\xi_1 = 1$  so  $R_n$  is well-defined—and let  $J_n := (n+1) - R_n$ . We have

(2)  $C_{\ell}(n) \leq Y_{\ell} + 1(J_n = \ell), \qquad 1 \leq \ell \leq n,$ 

with contributions to strict inequality whenever, for some  $1 \le \ell \le n$ , an  $\ell$ -spacing occurred in  $\xi_1, \xi_2, \ldots$ starting at  $i - \ell$  and ending at i > n.

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We view (2) as saying that the Ewens sampling formula distributed  $(C_1(n), \ldots, C_n(n))$  can be constructed from the independent Poisson *Y*'s using at most one insertion, together with a random number of deletions. The expected number of deletions is  $O_{\theta}(1)$ , that is, bounded over all *n*, with the upper bound depending on the value of  $\theta$ . A concrete upper bound is given in [3], but the limit value, call it  $c(\theta)$ , is cleaner. This limit *is* the expected number of spacings of length at most 1, with right end greater than 1, in the scale invariant Poisson process on  $(0, \infty)$  with intensity  $\theta/x dx$ ; see [1]. We have

$$c(\theta) = \int_{x>1} \frac{\theta}{x} \mathbb{P}(\text{at least one arrival in } (x - 1, x)) dx$$
$$= \int_{x>1} \frac{\theta}{x} \left( 1 - \exp\left(-\int_{x-1}^{x} \frac{\theta}{y} dy\right) \right) dx$$
$$= \int_{x>1} \frac{\theta}{x} \left( 1 - \left(1 - \frac{1}{x}\right)^{\theta} \right) dx$$

and, using the substitution v = 1 - 1/x, we get

$$\begin{aligned} c(\theta) &= \theta \int_0^1 (1-v)^{-1} (1-v^\theta) \, dv \\ &= \theta \sum_{n \ge 0} \left( \frac{1}{n+1} - \frac{1}{n+1+\theta} \right) \\ &= \theta \left( \frac{1}{\theta} + \sum_{n \ge 0} \left( \frac{1}{n+1} - \frac{1}{n+\theta} \right) \right) \\ &= 1 + \theta \left( \gamma + \psi(\theta) \right), \end{aligned}$$

where  $\gamma$  is Euler's constant and  $\psi$  is the digamma function.

The simple fact that one can transform the Ewens sampling formula into the highly tractable Poisson process  $Y_1, Y_2, \ldots, Y_n$  using a bounded (in expectation) number of insertions and deletions is, in itself, quite powerful, since there are interesting aspects of the joint distribution which are insensitive to a bounded number of insertions and deletions. For example, consider the Erdős-Turán law for the order of a random permutation. The order of a permutation is the least common multiple of the lengths of its cycles, and the Erdős-Turán law is the statement of convergence to the standard normal distribution, for the log of the order, centered by subtracting an asymptotic mean  $\log^2 n/2$ , and scaling by dividing by an asymptotic standard deviation,  $\log^{3/2} n/3$ . The effect of a finite number of cycle lengths is washed away by the scaling; see [5] for details.

In a similar spirit, and modeled after the Feller coupling for the Ewens sampling formula, [2] shows that for a random integer chosen uniformly from 1 to n, the counts  $C_p(n)$  of prime factors, including multiplicity, can be coupled to independent  $Z_2, Z_3, Z_5, ...$ with  $\mathbb{P}(Z_p \ge k) = p^{-k}$  for prime p and k = 0, 1, 2, ...in such a way that  $\mathbb{E} \sum_{p \le n} |C_p(n) - Z_p| \le 2 +$  $O((\log \log n)^2 / \log n)$ ; informally, the prime factorization can be converted into the process of independent geometric random variables, using on average no more than  $2 + \varepsilon_n$  insertions and deletions. The fact of being able to convert with  $o(\log \log n)$  insertions and deletions already easily implies the Hardy-Ramanujan theorem for the normal order of the number of prime divisors, and the fact of being able to convert with  $o(\sqrt{\log \log n})$  insertions and deletions readily implies the Erdős-Kac central limit theorem for the number of prime divisors.

The Feller coupling expresses the Ewens sampling formula in terms of the spacings of the independent Bernoulli sequence  $\xi_1, \xi_2, \ldots, \xi_n$ . The conditioning relation, described in Crane's article at the start of Section 3.8, expresses the Ewens sampling formula in terms of the independent Poisson  $Y_1, Y_2, \ldots, Y_n$ . Both these independent processes have the same limit upon rescaling, namely, the scale invariant Poisson process on  $(0, \infty)$  with intensity  $\theta/x \, dx$ . This leads to a property of the scale invariant Poisson process: the set of its spacings has the same distribution as the set of its arrivals. This property can be exploited to bound the distance to the Poisson-Dirichlet limit, which is mentioned in Crane's Section 4.2. Write  $(X_1, X_2, \ldots)$  for the random vector distributed according to the Poisson–Dirichlet( $\theta = 1$ ). For random permutations, writing  $L_i(n)$  for the size of the *i*th largest cycle, [4] shows that there are couplings which achieve  $\mathbb{E}\sum_{i\geq 1} |L_i(n) - nX_i| \sim \frac{1}{4}\log n$ , and that no coupling can achieve a constant smaller than 1/4. For prime factorizations, writing  $P_i(n)$  for the *i*th largest prime factor of a random integer distributed uniformly from 1 to n, [2] shows that there is a coupling of random integers to Poisson-Dirichlet such that  $\mathbb{E}\sum_{i\geq 1} |\log P_i(n) - (\log n)X_i| = O(\log \log n), \text{ and the}$ conjecture that O(1) can be achieved remains open.

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