# Prediction rules for exchangeable sequences related to species sampling<sup>1</sup>

by

Ben Hansen and Jim Pitman

Technical Report No. 520

Department of Statistics University of California 367 Evans Hall # 3860 Berkeley, CA 94720-3860

 $May\ 1998$ 

<sup>&</sup>lt;sup>1</sup>Research supported in part by N.S.F. Grant DMS 97-03961

## Prediction rules for exchangeable sequences related to species sampling<sup>†</sup>

Ben Hansen and Jim Pitman

May 18, 1998

#### Abstract

Suppose an exchangable sequence with values in a nice measurable space S admits a prediction rule of the following form: given the first n terms of the sequence, the next term equals the jth distinct value observed so far with probability  $p_{j,n}$ , for  $j=1,2,\ldots$ , and otherwise is a new value with distribution  $\nu$  for some probability measure  $\nu$  on S with no atoms. Then the  $p_{j,n}$  depend only on the partitition of the first n integers induced by the first n values of the sequence. All possible distributions for such an exchangeable sequence are characterized in terms of constraints on the  $p_{j,n}$  and in terms of their de Finetti representations.

#### 1 Introduction

There are very few models for exchangeable sequences  $(X_n)$  with an explicit  $prediction\ rule$ , that is a formula for the conditional distribution of  $X_{n+1}$  given  $X_1, \ldots, X_n$  for each  $n=0,1,\ldots$  The Blackwell-MacQueen urn scheme [3] provides an example: given a probability measure  $\nu(\cdot)$  on a nice measurable space  $(S, \mathcal{S})$  and  $\theta > 0$ , the prediction rule

$$\mathbb{P}(X_{n+1} \in \cdot \mid X_1, \dots, X_n) = \frac{1}{(n+\theta)} \sum_{i=1}^n 1(X_i \in \cdot) + \frac{\theta}{(n+\theta)} \nu(\cdot)$$
 (1)

determines an exchangeable sequence  $(X_n)$  whose directing random measure F has Dirichlet distribution with parameter  $\theta\nu(\cdot)$ . See [6] for background and applications of this model to non-parametric statistics. The subject of this paper is exchangeable sequences admitting a prediction rule of the more general form

$$\mathbb{P}(X_{n+1} \in \cdot \mid X_1, \dots, X_n) = \sum_{i=1}^n r_{i,n} 1(X_i \in \cdot) + q_n \, \nu(\cdot)$$
 (2)

for some  $r_{i,n}$  and  $q_n$  which are non-negative product-measurable functions of  $(X_1, \ldots, X_n)$ . As a minimal regularity condition on  $(S, \mathcal{S})$ , we suppose that the

 $<sup>^{\</sup>dagger} \rm Research$  supported in part by N.S.F. Grant DMS 97-03961

diagonal  $\{(x,y): x=y\}$  is a product-measurable subset of  $S\times S$ . The rule (2) can then be rewritten as follows, by grouping terms with equal values of  $X_i$ :

$$\mathbb{P}(X_{n+1} \in \cdot \mid X_1, \dots, X_n) = \sum_{j=1}^{K_n} p_{j,n} 1(\tilde{X}_j \in \cdot) + q_n \nu(\cdot)$$
 (3)

where the  $X_j$  for  $1 \leq j \leq K_n$  are the distinct values among  $X_1, \ldots, X_n$  in the order that they appear, and the  $p_{j,n}$  and  $q_n$  are some non-negative product-measurable functions of  $(X_1, \ldots, X_n)$ . This paper provides a description of all prediction rules of this form which generate exchangeable sequences, assuming that the probability measure  $\nu$  is diffuse, meaning  $\nu\{x\} = 0$  for all points x of S

Let  $\Pi$  denote the random partition of  $\{1, 2, \ldots, \}$  generated by  $X_1, X_2, \ldots$  So  $\Pi = \{\mathcal{A}_1, \mathcal{A}_2, \ldots \}$  where  $\mathcal{A}_j$  is the random set of indices m such that  $X_m = \tilde{X}_j$ . Let  $\Pi_n$  be the restriction of  $\Pi$  to  $\{1, \ldots, n\}$ . So  $\Pi_n$  is a measurable function of  $X_1, \ldots, X_n$  with values in the finite set of all partitions of the set  $\{1, \ldots, n\}$ . The main new result of this paper is the following theorem, which is proved in Section 2.

**Theorem 1** Suppose that an S-valued exchangeable sequence  $(X_n)$  admits a prediction rule of the form (3) for  $p_{j,n}$  and  $q_n$  some product-measurable functions of  $(X_1, \ldots, X_n)$ , and  $\nu$  a diffuse measure on S. Then for each n and  $1 \le j \le K_n$  the  $p_{j,n}$  and  $q_n$  are almost surely equal to some functions of  $\Pi_n$ , the partition of  $\{1, \ldots, n\}$  generated by  $(X_1, \ldots, X_n)$ .

While the focus of this paper is exchangeable sequences subject to a prediction rule of the form (3) for a diffuse measure  $\nu$ , we note that a weakening of Theorem 1 holds for  $\nu$  that is a mixture of diffuse and atomic measures. Then the  $p_{j,n}$  and  $q_n$  are almost surely equal to some functions of  $\Pi_n$  and the collection of random sets

$$\{\{i \le n : X_i = a\} : a \text{ an atom of } \nu\}. \tag{4}$$

This can be established by a slight variation of the proof of Theorem 1 given in Section 2.

The rest of this introduction shows how Theorem 1 combines with results obtained previously in [14] to yield a description of all possible functions  $p_{j,n}$  and  $q_n$  that could be used to generate an exchangeable sequence  $(X_n)$  by a prediction rule of the form (3) for diffuse  $\nu$ , and a corresponding description of the de Finetti representation of  $(X_n)$  in terms of sampling from a random distribution.

The assumption that  $(X_n)$  is exchangeable implies that  $\Pi$  is an exchangeable random partition of the set of positive integers, as considered by Kingman [9, 10] and subsequent authors [1, 12]. That is to say, for each  $n = 1, 2, \ldots$  and each partition  $\{A_1, \ldots, A_k\}$  of  $\{1, \ldots, n\}$ ,

$$\mathbb{P}(\Pi_n = \{A_1, \dots, A_k\}) = p(\#A_1, \dots, \#A_k)$$
(5)

for some non-negative symmetric function p of finite sequences of positive integers  $\mathbf{n} := (n_1, \ldots, n_k)$ . Here #A is the number of elements of A. Following [12, 14], call p the exchangeable partition probability function (EPPF) determined by  $\Pi$ . Write  $k(\mathbf{n})$  for the length k of  $\mathbf{n} := (n_1, \ldots, n_k)$ . For each finite sequence  $\mathbf{n}$  of positive integers and each  $1 \le j \le k(\mathbf{n}) + 1$ , a finite sequence  $\mathbf{n}^{j+}$  of positive integers is defined by incrementing  $n_j$  by 1. From (5) and the addition rule of probability, an EPPF must satisfy

$$p(1) = 1 \text{ and } p(\boldsymbol{n}) = \sum_{j=1}^{k(\boldsymbol{n})+1} p(\boldsymbol{n}^{j+}), \text{ for all } \boldsymbol{n}.$$
 (6)

Let

$$N_{j,n} := \sum_{m=1}^{n} 1[X_m = \tilde{X}_j] \tag{7}$$

which is the number of times that the jth distinct value  $\tilde{X}_j$  appears among  $X_1, \ldots, X_n$ . So  $N_{j,n}$  is the number of elements in the jth class of  $\Pi_n$  when classes are ordered by their least elements. If  $(X_n)$  is exchangeable and subject to a prediction rule of the form (3), with  $p_{j,n}$  and  $q_n$  functions of  $\Pi_n$ , it is easily seen that almost surely for all  $j \leq K_n$ 

$$p_{j,n} = p_j(N_{1,n}, \dots, N_{K_{n,n}}); \quad q_n = q(N_{1,n}, \dots, N_{K_{n,n}})$$
(8)

for some non-negative functions  $p_j$  and q of finite sequences of positive integers. These functions  $p_j$  and q can be characterized as follows:

**Theorem 2** [14, Prop. 13 and Thm. 14] Suppose  $(X_n)$  is exchangeable and subject to a prediction rule of the form (3), with  $p_{j,n}$  and  $q_n$  as in (8). Then the functions  $p_j$  and q can be expressed as follows in terms of the EPPF associated with the random partition  $\Pi$  generated by  $(X_n)$ : provided p(n) > 0,

$$p_j(\mathbf{n}) = \frac{p(\mathbf{n}^{j+})}{p(\mathbf{n})} \text{ for } 1 \le j \le k(\mathbf{n}); \ q(\mathbf{n}) = \frac{p(\mathbf{n}^{\ell+})}{p(\mathbf{n})} \text{ for } \ell = k(\mathbf{n}) + 1.$$
 (9)

Conversely, given a diffuse measure  $\nu$  on  $(S, \mathcal{S})$  and a non-negative symmetric function of finite sequences of positive integers subject to (6), the prediction rule (3) determined via (8) and (9) defines an exchangeable sequence  $(X_n)$ . Such a sequence  $(X_n)$  may be constructed by first generating an exchangeable random partition  $\Pi = \{A_1, A_2, \ldots\}$  whose EPPF is p, then setting  $X_n = \tilde{X}_j$  for  $n \in \mathcal{A}_j$  where the  $\tilde{X}_j$  are i.i.d. with distribution  $\nu$ , independent of  $\Pi$ .

Following [14], call such an exchangeable sequence  $(X_n)$  a species sampling sequence. This terminology is used to suggest the interpretation of  $(X_n)$  as the sequence of species of individuals in a process of sequential random sampling from some hypothetical infinite population of individuals of various species. The species of the first individual to be observed is assigned a random tag  $X_1 = \tilde{X}_1$  distributed according to  $\nu$ . Given the tags  $X_1, \ldots, X_n$  of the first n individuals

observed, it is supposed that the next individual is one of the jth species observed so far with probability  $p_{j,n}$ , and one of a new species with probability  $q_n$ . Each distinct species is assigned an independent random tag with distribution  $\nu$  as it appears in the sampling process. In this interpretation the random partition  $\Pi$  generated by the species sampling process is of primary importance: the allocation of i.i.d. random tags to distinct species is just a device to encode  $\Pi$  in a sequence of exchangeable random variables  $(X_n)$ . As shown by Aldous [1], this device allows Kingman's representation of exchangeable random partitions to be immediately deduced from de Finetti's representation of exchangeable sequences. For this purpose, the choice of the space S of species tags and the diffuse measure  $\nu$  on S is of no importance: one may as well take S = [0, 1] with Borel sets and  $\nu$  the uniform distribution on [0, 1].

The de Finetti representation of a species sampling sequence  $(X_n)$  can be described as follows:

**Theorem 3** [14] Write  $\tilde{P}_j$  for the limiting frequency of the jth species to appear in a species sampling sequence  $(X_n)$ :

$$\tilde{P}_j := \lim_{n \to \infty} \frac{N_{j,n}}{n} \tag{10}$$

which exists almost surely. Let  $F_n$  denote the conditional distribution of  $X_{n+1}$  given  $X_1, \ldots, X_n$ , as displayed in (3). Then  $F_n$  converges in total variation norm almost surely as  $n \to \infty$  to the random measure

$$F(\cdot) := \sum_{j} \tilde{P}_{j} 1(\tilde{X}_{j} \in \cdot) + (1 - \sum_{j} \tilde{P}_{j}) \nu(\cdot). \tag{11}$$

Conditionally given F the  $X_n$  are independent and identically distributed according to F.

The joint law of the  $\tilde{P}_j$  is determined by the EPPF of the partition  $\Pi$  generated by  $(X_n)$  via formulae described in [14]. See [12, 14] regarding the conditional distribution of  $\Pi$  given the sequence  $(\tilde{P}_j)$ , which is the same for all species sampling sequences. See [14] regarding the conditional distribution of F given  $(X_1, \ldots, X_n)$ . Theorem 3 yields also:

Corollary 4 [14] A sequence  $(X_n)$  is a species sampling sequence with marginal distributions equal to  $\nu$  if and only if  $(X_n)$  is conditionally i.i.d. (F) given some random probability distribution F on S of the form

$$F := \sum_{j} P_{j} 1(\hat{X}_{j} \in \cdot) + (1 - \sum_{j} P_{j}) \nu(\cdot).$$
 (12)

for some sequence of random variables  $P_j \geq 0$  with  $\sum_j P_j \leq 1$ , and given  $(P_j)$  the  $\hat{X}_j$  corresponding to j with  $P_j > 0$  are i.i.d.  $(\nu)$ .

**Example.** The Two-Parameter Model [12]. Consider the prediction rule (3) defined by some diffuse measure  $\nu$  and

$$p_{j,n} = \frac{N_{j,n} - \alpha}{n + \theta} \text{ for } 1 \le j \le K_n; \quad q_n = \frac{\theta + K_n \alpha}{n + \theta}$$
 (13)

where  $\alpha$  and  $\theta$  are two real parameters and as before the  $N_{j,n}$ ,  $1 \leq j \leq K_n$  are the numbers of representatives of the various distinct species  $\tilde{X}_j$ ,  $1 \leq j \leq K_n$  among  $X_1, \ldots, X_n$ . To ensure that all relevant probabilities are non-negative and that the rule is not degenerate, it must be supposed that either

$$\alpha = -\kappa < 0 \text{ and } \theta = m\kappa \text{ for some } \kappa > 0 \text{ and } m = 2, 3 \dots$$
 (14)

or

$$0 \le \alpha < 1 \text{ and } \theta > -\alpha.$$
 (15)

This prediction rule (13) is that determined by (9) for the function  $p = p_{(\alpha,\theta)}$  defined by the formula

$$p_{(\alpha,\theta)}(n_1,\dots,n_k) = \frac{\left(\prod_{\ell=1}^{k-1} (\theta + \ell\alpha)\right) \left(\prod_{i=1}^{k} [1-\alpha]_{n_i-1}\right)}{[1+\theta]_{n-1}}$$
(16)

where  $n = \sum_i n_i$  and  $[x]_m = \prod_{j=1}^m (x+j-1)$ . It is easily checked that  $p_{(\alpha,\theta)}$  is an EPPF. So a sequence  $(X_1,X_2,\ldots)$  defined by the prediction rule (13) is exchangeable, hence a species sampling sequence. The case with  $\alpha=0$  is the Blackwell-McQueen scheme. Then (16) is a variation of the Ewens sampling formula [4,2,5]. In the case (14), the distribution of the exchangeable sequence  $(X_n)$  is identical to that generated by sampling from  $F:=\sum_{i=1}^m P_i 1(\hat{X}_i \in \cdot)$ , where  $(P_1,\ldots,P_m)$  has a symmetric Dirichlet distribution with m parameters equal to  $\kappa$ , and the  $\hat{X}_i$  are i.i.d. with distribution  $\nu$ . This is Fisher's model for species sampling [7] with m species identified by i.i.d. $(\nu)$  tags. See [13, 16, 15, 17, 8, 18, 11] for further characterizations and applications of the two-parameter model.

#### 2 Proof of Theorem 1

Suppose throughout this section that  $(X_n)$  is an S-valued exchangeable sequence subject to a prediction rule of the form (3) for  $p_{j,n}$  and  $q_n$  some arbitrary measurable functions of  $(X_1, \ldots, X_n)$ , and  $\nu$  a diffuse measure on S. Let  $\Pi_n$  be the partition of  $\{1, \ldots, n\}$  generated by  $X_1, \ldots, X_n$ . In view of the last sentence of Theorem 2, to establish the conclusion of Theorem 1 that modulo null sets the  $p_{j,n}$  and  $q_n$  depend only on  $\Pi_n$ , it suffices to show that conditionally given  $\Pi$ , the partition of all positive integers generated by  $(X_n)$ , the random variables  $\tilde{X}_j$  for  $j = 1, 2, \ldots$  are independent and identically distributed according to  $\nu$ . The following lemma provides a convenient reformulation of this condition: **Lemma 5** For all  $1 \le k \le n$ , all partitions  $\pi$  of  $\{1, ..., n\}$  with k classes, and for all choices of measurable  $B_i \subseteq S, 1 \le j \le k$ 

$$\mathbb{P}(\Pi_n = \pi; \tilde{X}_j \in B_j, 1 \le j \le k) = \left(\prod_{j=1}^k \nu(B_j)\right) \mathbb{P}(\Pi_n = \pi)$$
 (17)

**Proof.** This is the result of repeated application of the following formula, which is claimed to hold for all choices of  $1 \le k \le n, \pi$  and  $B_j, 1 \le j \le k$ , as above, and all choices of i with  $1 \le i \le n$ :

$$\mathbb{P}(\Pi_n = \pi; \tilde{X}_j \in B_j \text{ all } j \le k) = \nu(B_i) \, \mathbb{P}(\Pi_n = \pi; \tilde{X}_j \in B_j \text{ all } j \le k, j \ne i)$$
(18)

If  $\pi$  is a partition of  $\{1,\ldots,n\}$  into k classes, write  $A_1^{\pi},\ldots,A_k^{\pi}$  for the k classes, ordered such that  $1=\min A_1^{\pi}<\min A_2^{\pi}<\cdots<\min A_k^{\pi}$ . Let  $n,\pi,k,B_1,\ldots,B_k$  be as in (18). It follows immediately from the prediction rule (3) and the assumption that  $\nu$  is diffuse that (18) holds if i=k and  $\#A_k^{\pi}=1$ . The assumed exchangeability of  $(X_n)$  then yields (18) for any  $1\leq i\leq k$  with  $\#A_i^{\pi}=1$ .

Now consider the inductive hypothesis, call it  $H_m$ , that (18) holds for all choices of  $1 \le k \le n$ ,  $\pi$ ,  $B_j$ ,  $1 \le j \le k$  and  $1 \le i \le k$  with  $\#A_i^{\pi} = m$ . We have just shown that  $H_1$  holds. We now assume  $H_m$  for some  $m = 1, 2, \ldots$ , and will complete the proof of the lemma by deducing  $H_{m+1}$ . As in the argument for m = 1, we first obtain a special case of  $H_{m+1}$ ; but by exchangeability, the special case implies the general case of  $H_{m+1}$ . So consider partitions  $\pi'$  of  $\{1, \ldots, n+1\}$  for which  $\#A_1^{\pi'} = m+1$  and  $n+1 \in A_1^{\pi'}$ . We prove  $H_{m+1}$  for these  $\pi'$  and for i = 1.

Fix such a  $\pi'$  partitioning  $\{1,\ldots,n+1\}$ , and measurable  $B_1,\ldots,B_k\subseteq S$ , and to avoid trivialities assume  $B_1,\ldots,B_k$  all have positive  $\nu$ -measure. Write  $\pi=\{A_1^\pi,\ldots,A_k^\pi\}$  for the restriction of  $\pi'$  to  $\{1,\ldots,n\}$ . For  $\ell=1,\ldots,k$ , write  $\pi^\ell$  for the partition  $\{A_1^\pi,\ldots,A_\ell^\pi\cup\{n+1\},\ldots,A_k^\pi\}$  of  $\{1,\ldots,n+1\}$ . Note that  $\pi'=\pi^1$ . Write  $\pi^{k+1}$  for the partition  $\{A_1^\pi,\ldots,A_k^\pi,\{n+1\}\}$  of  $\{1,\ldots,n+1\}$ . By  $H_m$ , for each  $\ell=2,\ldots,k+1$ ,

 $\mathbb{P}(\Pi_{n+1} = \pi^{\ell}, \tilde{X}_j \in B_j \text{ all } j \leq k) = \nu(B_1)\mathbb{P}(\Pi_{n+1} = \pi^{\ell}; \tilde{X}_j \in B_j \text{ all } 2 \leq j \leq k)$ since in each of the partitions  $\pi^2, \ldots, \pi^{k+1}$  the first class has size m. Similarly,

$$\mathbb{P}(\Pi_n = \pi, \tilde{X}_j \in B_j \text{ all } j \leq k) = \nu(B_1)\mathbb{P}(\Pi_n = \pi; \tilde{X}_j \in B_j \text{ all } 2 \leq j \leq k).$$

The identity

$$\mathbb{P}(\Pi_n = \pi, \tilde{X}_j \in B_j \text{ all } j \le k) = \sum_{\ell=1}^{k+1} \mathbb{P}(\Pi_{n+1} = \pi^{\ell}, \tilde{X}_j \in B_j \text{ all } j \le k)$$

now implies that

 $\mathbb{P}(\Pi_{n+1} = \pi^1, \tilde{X}_j \in B_j \text{ all } j \leq k) = \nu(B_1)\mathbb{P}(\Pi_{n+1} = \pi^1, \tilde{X}_j \in B_j \text{ all } 2 \leq j \leq k),$ which is the identity required to establish  $H_{m+1}$ .  $\square$ 

### References

- [1] D.J. Aldous. Exchangeability and related topics. In P.L. Hennequin, editor, École d'Été de Probabilités de Saint-Flour XII, Springer Lecture Notes in Mathematics, Vol. 1117. Springer-Verlag, 1985.
- [2] C. Antoniak. Mixtures of Dirichlet processes with applications to Bayesian nonparametric problems. *Ann. Statist.*, 2:1152-1174, 1974.
- [3] D. Blackwell and J.B. MacQueen. Ferguson distributions via Pólya urn schemes. *Ann. Statist.*, 1:353-355, 1973.
- [4] W.J. Ewens. The sampling theory of selectively neutral alleles. *Theor. Popul. Biol.*, 3:87 112, 1972.
- [5] W.J. Ewens and S. Tavaré. The Ewens sampling formula. To appear in Multivariate Discrete Distributions edited by N.S. Johnson, S. Kotz, and N. Balakrishnan, 1995.
- [6] T.S. Ferguson. A Bayesian analysis of some nonparametric problems. *Ann. Statist.*, 1:209-230, 1973.
- [7] R.A. Fisher, A.S. Corbet, and C.B. Williams. The relation between the number of species and the number of individuals in a random sample of an animal population. *J. Animal Ecol.*, 12:42-58, 1943.
- [8] S. Kerov. Coherent random allocations and the Ewens-Pitman formula. PDMI Preprint, Steklov Math. Institute, St. Petersburg, 1995.
- [9] J. F. C. Kingman. The representation of partition structures. J. London Math. Soc., 18:374-380, 1978.
- [10] J. F. C. Kingman. The coalescent. Stochastic Processes and their Applications, 13:235-248, 1982.
- [11] A.Z. Mekjian and K.C. Chase. Disordered systems, power laws and random processes. *Phys. Letters A*, 229:340–346, 1997.
- [12] J. Pitman. Exchangeable and partially exchangeable random partitions. *Probab. Th. Rel. Fields*, 102:145-158, 1995.
- [13] J. Pitman. Random discrete distributions invariant under size-biased permutation. Adv. Appl. Prob., 28:525-539, 1996.
- [14] J. Pitman. Some developments of the Blackwell-MacQueen urn scheme. In T.S. Ferguson et al., editor, Statistics, Probability and Game Theory; Papers in honor of David Blackwell, volume 30 of Lecture Notes-Monograph Series, pages 245-267. Institute of Mathematical Statistics, Hayward, California, 1996.

- [15] J. Pitman. Coalescents with multiple collisions. Technical Report 495, Dept. Statistics, U.C. Berkeley, 1997. Available via http://www.stat.berkeley.edu/users/pitman.
- [16] J. Pitman. Partition structures derived from Brownian motion and stable subordinators. *Bernoulli*, 3:79-96, 1997.
- [17] J. Pitman and M. Yor. The two-parameter Poisson-Dirichlet distribution derived from a stable subordinator. *Ann. Probab.*, 25:855-900, 1997.
- [18] S.L. Zabell. The continuum of inductive methods revisited. In J. Earman and J. D. Norton, editors, *The Cosmos of Science*, Pittsburgh-Konstanz Series in the Philosophy and History of Science, pages 351-385. University of Pittsburgh Press/Universitätsverlag Konstanz, 1997.