

## MULTINOMIAL PROBABILITIES, PERMANENTS AND A CONJECTURE OF KARLIN AND RINOTT

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**ABSTRACT.** The probability density function of a multiparameter multinomial distribution can be expressed in terms of the permanent of a suitable matrix. This fact and certain results on conditionally negative definite matrices are used to prove a conjecture due to Karlin and Rinott.

**1. Introduction.** If  $A = ((a_{ij}))$  is an  $n \times n$  matrix, the permanent of  $A$ , denoted by  $\text{per } A$ , is defined as

$$\text{per } A = \sum_{\sigma \in S_n} \prod_{i=1}^n a_{i\sigma(i)},$$

where  $S_n$  is the group of permutations of  $1, 2, \dots, n$ .

The probability density function of a multinomial distribution can be expressed in terms of the permanent of a suitable matrix. To begin with a simple example, suppose  $X$  denotes the number of heads resulting from  $n$  tosses of a coin, with probability of heads equal to  $p$ ,  $0 \leq p \leq 1$ , on a single toss. Then  $X$  has the Binomial distribution and its probability density function is given by

$$(1) \quad P(X = x) = \frac{n!}{x!(n-x)!} p^x q^{n-x}, \quad x = 0, 1, \dots, n;$$

where  $P(X = x)$  denotes the probability that  $X$  equals  $x$ ,  $q = 1 - p$ , and, as usual,  $P(X = x)$  is understood to be zero for all values of  $x$  not specified in (1).

The probability in (1) can be expressed in terms of a permanent as follows:

$$(2) \quad P(X = x) = \frac{1}{x!(n-x)!} \text{per} \left[ \begin{array}{ccc} p & \cdots & p \\ \vdots & & \vdots \\ p & \cdots & p \\ q & \cdots & q \\ \vdots & & \vdots \\ q & \cdots & q \end{array} \right] \left. \begin{array}{l} \left. \vphantom{\begin{matrix} p \\ \vdots \\ p \\ q \\ \vdots \\ q \end{matrix}} \right\} x \text{ times,} \\ \left. \vphantom{\begin{matrix} p \\ \vdots \\ p \\ q \\ \vdots \\ q \end{matrix}} \right\} (n-x) \text{ times,} \end{array} \right\} x = 0, 1, \dots, n.$$

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The expression in (2) admits generalizations. For example, suppose  $n$  different coins are tossed once and let  $X$  be the number of heads obtained. If  $p_i$  is the probability of heads on a single toss of the  $i$ th coin and if  $q_i = 1 - p_i, i = 1, 2, \dots, n$ , then it can be verified that

$$(3) \quad P(X = x) = \frac{1}{x!(n-x)!} \text{per} \left\{ \begin{array}{l} \left[ \begin{array}{ccc} p_1 & \cdots & p_n \\ \vdots & & \vdots \\ p_1 & \cdots & p_n \\ q_1 & \cdots & q_n \\ \vdots & & \vdots \\ q_1 & \cdots & q_n \end{array} \right] \left. \vphantom{\begin{array}{ccc} p_1 & \cdots & p_n \\ \vdots & & \vdots \\ p_1 & \cdots & p_n \\ q_1 & \cdots & q_n \\ \vdots & & \vdots \\ q_1 & \cdots & q_n \end{array}} \right\} \begin{array}{l} x \text{ times,} \\ \\ \\ (n-x) \text{ times,} \end{array} \quad x = 0, 1, \dots, n.$$

More generally, instead of tossing  $n$  coins, it may be an experiment of rolling  $n$  dice, differently loaded, and if  $X_i$  denotes the number of times  $i$  spots are obtained,  $i = 1, 2, \dots, 6$ , the density function of  $(X_1, \dots, X_6)$  can be written in terms of a permanent.

To make the concepts precise, consider an experiment which can result in any of  $r$  possible outcomes and suppose  $n$  trials of the experiment are performed. Let  $\pi_{ij}$  be the probability that the experiment results in the  $i$ th outcome at the  $j$ th trial,  $i = 1, 2, \dots, r; j = 1, 2, \dots, n$ . Let  $\Pi$  denote the  $r \times n$  matrix  $((\pi_{ij}))$ , which, of course, is column stochastic. Let  $X_i$  denote the number of times the  $i$ th outcome is obtained in the  $n$  trials,  $i = 1, 2, \dots, r$ , and let  $X = (X_1, \dots, X_r)$ . In this setup we will say that  $X$  has the multiparameter multinomial distribution with the  $r \times n$  parameter matrix  $\Pi$ . Let  $\pi^i$  denote the  $i$ th row of  $\Pi, i = 1, 2, \dots, r$ . The density function of  $X$  can again be expressed in terms of a permanent as

$$(4) \quad P(X = t) = \frac{1}{t_1! \cdots t_r!} \text{per} \left\{ \begin{array}{l} \left[ \begin{array}{c} \pi^1 \\ \vdots \\ \pi^1 \\ \vdots \\ \pi^r \\ \vdots \\ \pi^r \end{array} \right] \left. \vphantom{\begin{array}{c} \pi^1 \\ \vdots \\ \pi^1 \\ \vdots \\ \pi^r \\ \vdots \\ \pi^r \end{array}} \right\} \begin{array}{l} t_1 \text{ times,} \\ \\ \\ t_r \text{ times,} \end{array} \quad t \in \mathcal{K}_n,$$

where

$$\mathcal{K}_n = \left\{ k = (k_1, \dots, k_r): k_i \text{ nonnegative integers, } \sum_{i=1}^r k_i = n \right\}.$$

It seems that the representation (4) is important in understanding certain properties of the multinomial distribution. The purpose of this paper is to exploit (4) to settle a conjecture due to Karlin and Rinott [4].

**2. The problem.** Before stating the main problem, we need some notation.

Let

$$H^r = \left\{ x \in R^r: \sum_{i=1}^r x_i = 0 \right\}.$$

DEFINITION 1. A real, symmetric  $r \times r$  matrix  $A$  is said to be conditionally negative definite (c.n.d.) if for any  $x \in H^r$ ,

$$\sum_{i=1}^r \sum_{j=1}^r a_{ij} x_i x_j \leq 0.$$

Now suppose  $X = (X_1, \dots, X_r)$  follows the multiparameter multinomial distribution with the  $r \times n$  parameter matrix  $\Pi = ((\pi_{ij}))$ . We assume throughout that  $\Pi$  is positive and that  $n \geq 2, r \geq 2$ .

Fix  $k \in \mathcal{K}_{n-2}$  and let

$$(5) \quad k_{ij} = (k_1, \dots, k_{i-1}, k_i + 1, k_{i+1}, \dots, k_{j-1}, k_j + 1, k_{j+1}, \dots, k_r), \quad 1 \leq i \neq j \leq r,$$

$$k_{ii} = (k_1, \dots, k_{i-1}, k_i + 2, k_{i+1}, \dots, k_r), \quad 1 \leq i \leq r.$$

Clearly,  $k_{ij} \in \mathcal{K}_n, i, j = 1, 2, \dots, r$ . Our main result is that  $((\log P(X = k_{ij})))$  is c.n.d. This appears as Conjecture 2.1 in Karlin and Rinott [4], where it has been confirmed for  $r = 2, 3$  and for any  $n$ . We refer to [4] for a discussion concerning the relevance of the problem to multivariate majorization and inequalities as well as for certain consequences of proving the conjecture. In particular, Conjecture 2.2 of [4] is also verified once Conjecture 2.1 is established. It must be remarked that we have used a notation different than that of [4] at various places in the paper. Also, we find it more convenient to work with conditionally negative definite matrices rather than conditionally positive definite matrices.

**3. Permanents of positive matrices.** In this section we give certain results on permanents that will be used. Let  $A$  be a real matrix of order  $(n - 2) \times n, n \geq 2$ . Let  $e^j$  denote the unit row vector  $(0, \dots, 0, 1, 0, \dots, 0)$ , where the 1 occurs at the  $j$ th place,  $j = 1, 2, \dots, n$ . Define an  $n \times n$  matrix  $\hat{A}$  as follows. The  $(i, j)$ th entry of  $\hat{A}$  is the permanent of the augmented matrix

$$\begin{bmatrix} A \\ e^i \\ e^j \end{bmatrix}, \quad i, j = 1, 2, \dots, n.$$

Note that the diagonal entries of  $\hat{A}$  are all zero and  $\hat{A}$  is symmetric. Of course if  $n = 2$ , then  $A$  is absent and

$$\hat{A} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}.$$

The following theorem is a deep result in the theory of permanents. It was originally proved by Alexandroff [1] in a more general form. It serves as a crucial step in the proof of the well-known van der Waerden conjecture due to Egorychev [2, 10] and Falikman [3]. For a proof of the result, see Theorem 2.8 of [10]. Again, beware of differences in notation.

**THEOREM 2.** *If  $A$  is a positive  $(n - 2) \times n$  matrix,  $n \geq 2$ , then the matrix  $\hat{A}$  is nonsingular and has exactly one, simple, positive eigenvalue.*

The following result will be deduced from Theorem 2.

**THEOREM 3.** Let  $A$  be a positive  $(n-2) \times n$  matrix,  $n \geq 2$ , and let  $x^1, \dots, x^r$  be row vectors in  $R^n$ . Define the  $r \times r$  matrix  $B = ((b_{ij}))$  as

$$b_{ij} = \text{per} \begin{bmatrix} A \\ x^i \\ x^j \end{bmatrix}, \quad i, j = 1, 2, \dots, r.$$

Then  $B$  has at most one, simple, positive eigenvalue.

**PROOF.** The permanent admits a Laplace expansion in terms of a set of rows [7, p. 16].

Expand

$$\text{per} \begin{bmatrix} A \\ x^i \\ x^j \end{bmatrix}$$

in terms of the last two rows and observe that

$$(6) \quad b_{ij} = \langle x^i \hat{A}, x^j \rangle, \quad i, j = 1, 2, \dots, r.$$

Let  $X$  be the  $r \times n$  matrix whose  $i$ th row is  $x^i$ ,  $i = 1, 2, \dots, r$ . Then it follows from (6) that  $B = X \hat{A} X'$ . By Theorem 2,  $\hat{A}$  has exactly one positive eigenvalue and hence (essentially) by Sylvester's law,  $B$  has at most one positive eigenvalue. This completes the proof.

**4. Conditionally negative definite matrices.** The definition of a c.n.d. matrix was given in §2. The following result will be useful (see, for example, Parthasarathy and Schmidt [8, p. 3]).

**LEMMA 4.** A real, symmetric matrix  $A$  is c.n.d. if and only if for each  $\alpha > 0$ , the matrix  $((e^{-\alpha a_{ij}}))$  is positive semidefinite.

Recall that a function  $F: (0, \infty) \rightarrow R$  is said to be completely monotonic if it is in  $C^\infty(0, \infty)$  and  $(-1)^k F^{(k)}(x) \geq 0$ ,  $x \in (0, \infty)$ ,  $k = 0, 1, 2, \dots$ , where, by definition,  $F^{(0)} = F$ .

The following result appears in a recent paper by Micchelli [6]. A proof is included for completeness.

**LEMMA 5.** Let  $A$  be a c.n.d.  $n \times n$  matrix with positive entries and let  $F: (0, \infty) \rightarrow R$  be completely monotonic. Then the matrix  $((F(a_{ij})))$  is positive semidefinite.

**PROOF.** By a well-known theorem of Bernstein (see, for example, [11, p. 160]),  $F$  admits the representation

$$F(t) = \int_0^\infty e^{-t\sigma} d\mu(\sigma), \quad t > 0,$$

where  $d\mu(\sigma)$  is a Borel measure on  $(0, \infty)$ . The result follows by Lemma 4.

Now we have the following.

**LEMMA 6.** Let  $A$  be a symmetric, positive  $r \times r$  matrix with exactly one, simple, positive eigenvalue. Then, for any  $x \in H^r$ ,

$$\prod_{i,j=1}^r a_{ij}^{x_i x_j} \leq 1.$$

PROOF. It is known that if  $A$  is a symmetric, positive  $r \times r$  matrix, then there exists a positive vector  $z$  such that the matrix  $B = ((b_{ij})) = ((a_{ij}z_i z_j))$  is doubly stochastic (see, for example, [5, 9]). Clearly, if  $A$  satisfies the hypothesis of the lemma, then  $B$  also has only one positive eigenvalue by Sylvester's law, and, furthermore,

$$\begin{aligned} \prod_{i,j=1}^r b_{ij}^{x_i x_j} &= \prod_{i,j=1}^r (a_{ij} z_i z_j)^{x_i x_j} \\ &= \prod_{i,j=1}^r a_{ij}^{x_i x_j} \quad \text{for any } x \in H^r. \end{aligned}$$

Thus, we may assume, without loss of generality, that  $A$  is doubly stochastic, so that the vector  $(1, \dots, 1)$  is an eigenvector of  $A$  corresponding to the eigenvalue 1. Since  $A$  has only one positive eigenvalue, any vector  $x \in H^r$  lies in the span of eigenvectors of  $A$  corresponding to only nonpositive eigenvalues and, hence,

$$\sum_{i=1}^r \sum_{j=1}^r a_{ij} x_i x_j \leq 0.$$

Thus,  $A$  is c.n.d. The function  $F(t) = t^{-\alpha}$ ,  $t > 0$ , is completely monotonic for any  $\alpha > 0$  and by Lemma 5,  $((a_{ij}^{-\alpha}))$  is positive semidefinite.

Hence, for any  $\alpha > 0$ ,  $((e^{-\alpha \log a_{ij}}))$  is positive semidefinite and by Lemma 4,  $((\log a_{ij}))$  is c.n.d. This completes the proof.

**5. The main result.** We are now in a position to prove the following statement, conjectured by Karlin and Rinott [4].

**THEOREM 7.** *Let  $X = (X_1, \dots, X_r)$  have the multiparameter multinomial distribution with the  $r \times n$  parameter matrix  $\Pi$ . Let  $k \in K_{n-2}$ , let  $k_{ij}$  be defined as in (5), and let  $m_{ij} = P(X = k_{ij})$ ,  $i, j = 1, 2, \dots, r$ . Then the matrix  $((\log m_{ij}))$  is c.n.d.*

PROOF. We have to prove that for any  $x \in H^r$ ,

$$(7) \quad \prod_{i,j=1}^r m_{ij}^{x_i x_j} \leq 1.$$

Let  $A$  be the  $(n-2) \times n$  matrix formed by taking  $k_i$  copies of  $\pi^i$ , the  $i$ th row of  $\Pi$ ,  $i = 1, 2, \dots, n$ . Define the  $r \times r$  matrix  $B = ((b_{ij}))$  as

$$b_{ij} = \text{per} \begin{bmatrix} A \\ \pi^i \\ \pi^j \end{bmatrix}, \quad i, j = 1, 2, \dots, r.$$

By Theorem 3,  $B$  has at most one positive eigenvalue, and since  $B$  is a positive matrix, it must have exactly one positive eigenvalue. So by Lemma 6,

$$(8) \quad \prod_{i,j=1}^r b_{ij}^{x_i x_j} \leq 1 \quad \text{for any } x \in H^r.$$

By the relationship (4),

$$(9) \quad m_{ij} = P(X = k_{ij}) = \begin{cases} \frac{b_{ij}}{(k_1!, \dots, k_r!)(k_i + 1)(k_j + 1)}, & 1 \leq i \neq j \leq r, \\ \frac{b_{ij}}{(k_1! \cdots k_r!)(k_i + 1)(k_i + 2)}, & 1 \leq i \leq r. \end{cases}$$

Let  $D$  be the  $r \times r$  diagonal matrix with its  $i$ th diagonal entry equal to  $(k_i + 1)^{-1}$ ,  $i = 1, 2, \dots, r$ , and let

$$(10) \quad C = \frac{1}{k_1! \cdots k_r!} DBD.$$

It follows from (8) and (10) that for any  $x \in H^r$ ,

$$(11) \quad \prod_{i,j=1}^r c_{ij}^{x_i x_j} \leq 1.$$

From (9) and (10),

$$(12) \quad m_{ij} = \begin{cases} c_{ij}, & 1 \leq i \neq j \leq r, \\ ((k_i + 1)/(k_i + 2))c_{ii}, & 1 \leq i \leq r, \end{cases}$$

Since  $(k_i + 1)/(k_i + 2) < 1$ ,  $i = 1, 2, \dots, r$ , it follows from (12) and (11) that

$$\prod_{i,j=1}^r m_{ij}^{x_i x_j} \leq \prod_{i,j=1}^r c_{ij}^{x_i x_j} \leq 1 \quad \text{for any } x \in H^r,$$

and the proof is complete.

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