ON A CERTAIN SET OF LINEAR INEQUALITIES

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1. Introduction. In this paper we shall discuss the following set of n+1 linear inequalities:

If we let $Y = (y_i)$, $C_n = (\binom{n}{i})$, and $Z = (z_i)$ (i = 0, 1, ..., n) be (n+1)-dimensional column vectors, and define the n+1 by n+1 tridiagonal matrix $D_n(\varphi)$ by

$$D_{\mathbf{n}}(\varphi) \ = \ \begin{pmatrix} \varphi & 1 & 0 & 0 & 0 & \dots & 0 & 0 & 0 \\ n & \varphi & 2 & 0 & 0 & \dots & 0 & 0 & 0 \\ 0 & n-1 & \varphi & 3 & 0 & \dots & 0 & 0 & 0 & 0 \\ 0 & 0 & n-2 & \varphi & 4 & \dots & 0 & 0 & 0 & 0 \\ \dots & \dots \\ 0 & 0 & 0 & 0 & 0 & \dots & \varphi & n-1 & 0 \\ 0 & 0 & 0 & 0 & 0 & \dots & 2 & \varphi & n \\ 0 & 0 & 0 & 0 & 0 & \dots & 0 & 1 & \varphi \end{pmatrix} \ ,$$

the set of inequalities (1) may be written

$$A_{n} Y = C_{n} + Z$$

where $A_n = D_n(1)$ and $z_i \ge 0$ (i = 0, 1, ..., n). In sections 2 and 3, we

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consider real solutions of (2), and give expressions for the solution Y corresponding to a specified vector Z of slack variables. The inequalities (1) arise in connection with a current investigation of some covering properties of groups [4], where it is necessary to find all solutions of (1) in non-negative integers y_i with Σy_i specified. In section 4, we give an efficient algorithm for obtaining such solutions.

2. Some properties of the matrix An.

We begin by considering the set of n+1 linear equations

$$D_{n}(\phi) X = 0$$

where $X = (x_i)$ (i = 0, 1, ..., n) is a column vector. (3) may be written

(4)
$$(n+1-i)x_{i-1} + \phi x_i + (i+1)x_{i+1} = 0, -\infty < i < \infty,$$

with boundary conditions

(5)
$$x_i = 0 \text{ if } i < 0 \text{ or } i > n.$$

Multiplying (4) by ti and summing over all i gives

(6)
$$\text{nt } G(t) - t^2 G'(t) + \phi G(t) + G'(t) = 0$$

where $G(t) = \sum_{i} t^{i}$. The solution of (6) with $G(0) = x_{0} = 1$ is

(7)
$$G(t) = (1+t)^{\frac{n-\phi}{2}} \underbrace{(1-t)^{\frac{n+\phi}{2}}}_{i=0} = \sum_{r=0}^{\infty} t^{i} \sum_{r=0}^{i} (-1)^{r} \binom{(n+\phi)/2}{r} \binom{(n-\phi)/2}{i-r}.$$

It follows that

(8)
$$x_{i} = \sum_{r=0}^{i} (-1)^{r} {\binom{(n+\phi)/2}{r}} {\binom{(n-\phi)/2}{i-r}}.$$

The boundary conditions (5) are satisfied if and only if $n+\varphi$ is an even integer and $-n \le \varphi \le n$; thus (3) has a nontrivial solution X if and only if

(9)
$$\phi = \phi_{j} = n - 2j \quad (j = 0, 1, ..., n).$$

The solution vector $X_{i} = (x_{i})$ corresponding to ϕ_{i} is given by

(10)
$$x_{ij} = \sum_{r=0}^{i} (-1)^{r} {n-j \choose r} {j \choose i-r}.$$

We may now determine the eigenvalues and eigenvectors of A_n ; for the equations $A_n X = \lambda X$ may be written $D_n(\phi) X = 0$ with $\phi = 1 - \lambda$. Thus A_n has eigenvalues

(11)
$$\lambda_{j} = 1 - \phi_{j} = 1 - n + 2j \quad (j = 0, 1, ..., n)$$

and the eigenvector corresponding to λ_j is $X_j = (x_{ij})$. It also follows that det $A_n = \pi \lambda_j$, which is zero for n odd, and equal to $(-1)^{n/2} (n+1) (n-1)^2 (n-3)^2 \dots 3^2 \dots 1^2$ for n even. Det A_n may also be obtained simply by direct expansion and recursion. In fact, the determinant of D_n (ϕ) was evaluated by J. J. Sylvester [5] as early as 1854.

Let $X = (X_0 | X_1 | \dots | X_n)$ be the n+1 by n+1 modal matrix whose columns are the eigenvectors X_0, X_1, \dots, X_n . The matrix X was computed for small values of n, and it was noted that each row of X is orthogonal to all but one of the columns of X. Consequently the following lemma was obtained.

LEMMA. Let
$$S_{ij} = \sum_{\alpha=0}^{n} x_{i\alpha} x_{\alpha j}$$
, where x_{ij} is defined by (10). Then

(12)
$$S_{ij} = 0 \text{ if } j \neq n-i; S_{i, n-i} = (-1)^{i} 2^{n}.$$

<u>Proof</u>: Put $H_j(u) = \sum_i S_{ij} u^i$. Then, using the generating function

$$G_{j}(t) = \sum_{i} x_{ij} t^{i} = (1+t)^{j} (1-t)^{n-j}$$

corresponding to ϕ_i we get

$$H_{j}(u) = \sum_{i} \sum_{k} x_{ik} x_{kj} u^{i} = \sum_{k} x_{kj} (1+u)^{j} (1-u)^{n-j}$$

$$= (1-u)^{n} (1 + \frac{1+u}{1-u})^{j} (1 - \frac{1+u}{1-u})^{n-j} = 2^{n} (-u)^{n-j}.$$

By comparing coefficients of powers of u we obtain (12).

Thus the i^{th} row of X is orthogonal to every column of X except the $(n-1)^{th}$ $(i=0,1,\ldots,n)$, and

(13)
$$X^{-1} = 2^{-n} (X_n | -X_{n-1} | X_{n-2} | -X_{n-3} | \dots | (-1)^n | X_0).$$

Let Λ be the diagonal matrix with diagonal entries $\lambda_0, \lambda_1, \dots, \lambda_n$, so that $A_n X = X \Lambda$. Then if n is even,

(14)
$$A_n^{-1} = X \Lambda^{-1} X^{-1}.$$

We are indebted to Professor D.A. Sprott for pointing out a connection with probability theory. If n is even, $\frac{1}{n} D_n(0)$ is the matrix of transition probabilities for the Ehrenfest Model. A discussion of some of the properties of this matrix appears in [1] and [3]. The arguments given here are somewhat simpler than theirs because the Lemma makes the derivation of X^{-1} almost trivial.

3. Real Solutions of $A_n Y = C_n + Z$. If n is even, A_n is nonsingular, and there will be a unique solution Y of (2) corresponding to each vector Z. Since $C_n = X_n$, the eigenvector corresponding to $\lambda_n = n+1$, we have

(15)
$$Y = A_n^{-1}(C_n + Z) = \frac{1}{n+1} C_n + A_n^{-1} Z.$$

If the slack variables z_i are all non-negative, Y is a solution of (1).

If n is odd, say n = 2k + 1, A_n is singular and the situation is slightly more complicated. Since X_0, X_1, \ldots, X_n , are linearly independent, every vector Y may be expressed as a linear combination

of them. If $Y = \sum a_i X_i$ is a solution,

$$A_n Y = \sum a_j A_n X_j = \sum a_j \lambda_j X_j = C_n + Z.$$

Since $C_n = X_n$ and $\lambda_k = 0$, Z must lie in the subspace spanned by $X_0, X_1, \dots, X_{k-1}, X_{k+1}, \dots, X_n$. But, by the Lemma,

$$\sum_{\alpha} \mathbf{x}_{k+1, \alpha} \mathbf{x}_{\alpha i} = 0, i \neq k,$$

and therefore the slack variables z_{α} must satisfy

(16)
$$\sum_{\alpha} \mathbf{z}_{\alpha} \mathbf{x}_{k+1, \alpha} = 0.$$

If Z is any vector satisfying (16), then $\mathbf{Z} = \sum \mathbf{b}_{j} \mathbf{X}_{j}$ with $\mathbf{b}_{k} = \mathbf{0}$, and corresponding to \mathbf{Z} are solutions

(17)
$$Y = \frac{1}{n+1} C_n + \sum_{j \neq k} \frac{b_j}{\lambda_j} X_j + t X_k$$

where t is an arbitrary real number. Let Λ^* be the diagonal matrix with diagonal entries $\lambda_0, \lambda_1, \ldots, \lambda_{k-1}, 1, \lambda_{k+1}, \ldots, \lambda_n$, and let B be the column vector (b_j). Then Z = XB, and thus $B = X^{-1}Z$. We may now rewrite (17) in the form

(18)
$$Y = \frac{1}{n+1} C_n + X \Lambda *^{-1} X^{-1} Z + t X_k.$$

If all the slack variables z_{α} are non-negative, Y is a solution of (1).

A convenient algorithm for numerical computation of real solutions Y of the equations AY = B, where Y and B are column vectors and A is an arbitrary nonsingular tridiagonal matrix, is described by Henrici [2, page 350]. This algorithm depends upon a rather interesting factorization of a tridiagonal matrix into two "bidiagonal" matrices.

4. Solutions of (1) in non-negative integers. The inequalities (1) arise in connection with a covering problem [4] where y_{α} represents the number of elements of a certain type in a covering set. Consequently

the y 's and z 's must be non-negative integers. The problem is to construct a covering set with as few members as possible - that is, with Σ y as small as possible. Adding the equations (2) gives

$$(n+1) \sum y_{\alpha} = 2^{n} + \sum z_{\alpha}$$

so that $\Sigma y_{\alpha} \geq 2^n/(n+1)$. This is not, however, a sufficient condition for the existence of a solution of (1). Furthermore not every solution Y corresponds to a covering set. Thus it is often necessary to consider several totals Σy_{α} , beginning with the least integer greater than or equal to $2^n/(n+1)$. It is, however, necessary to consider only solutions with $y_0 = 1$ since any covering set will be isomorphic to one with $y_0 = 1$.

In this section we give an efficient algorithm for finding all solutions of (1) with $y_0 = 1$ and $\Sigma y_\alpha \leq m$. The latter condition is equivalent to insisting that the total slack Σz_α be at most $T = 2^n - m(n+1)$. The algorithm is easily programmed for a computer, and may be generalized to yield the non-negative integer solutions of many sets of linear inequalities whose matrices of coefficients are tridiagonal.

The algorithm is represented pictorially by the directed graph in Figure 1. Vertices represent operations, and edges indicate the order in which they are performed. First, we give the operations corresponding to the vertices and the rules for moving from one to another. Then we shall explain the algorithm and give an example.

S (start): Put
$$y_{-1} = 0$$
, $y_0 = 1$, $z_0 = T$, $k = 0$. Go to A.

A: Calculate
$$R_k = {n \choose k} + z_k - y_k - (n-k+1)y_{k-1}$$
. If $R_k < 0$, go to C ; if $R_k \ge 0$, go to B .

B: Calculate
$$y_{k+1} = [R_k/(k+1)]$$
, $z_k = (k+1)y_{k+1} + y_k + (n-k+1)y_{k-1} - \binom{n}{k}$
and $z_{k+1} = T - z_1 - z_2 - \dots - z_k$. If $z_k < 0$, go to C; if $z_k \ge 0$ and $k < n-1$, go to G; if $z_k \ge 0$ and $k = n-1$, go to F.

C: Select the largest j < k for which $z_j \neq 0$ and go to D. If $z_j = 0$ for all j < k, go to E.

- D: Decrease z_j by 1 and put $z_{j+1} = T z_0 z_1 \dots z_j$. Put k = j and go to A.
- E: Algorithm terminates.
- F: Put $z_n = y_{n-1} + y_n 1$. Y is a solution with slack Z. Go to C.
- G: Increase k by 1 and go to A.

The algorithm examines all possible slack vectors \mathbf{Z} with $\Sigma\,\mathbf{z}_{\,\alpha}\!\leq \mathbf{T}$. We begin with $\,\mathbf{Z}$ = (T, 0, 0, ..., 0), and change $\,\mathbf{Z}\,$ in such a way that

$$||z|| = z_0(T+1)^n + z_1(T+1)^{n-1} + ... + z_{n-1}(T+1) + z_n$$

is steadily decreasing. Suppose that after a number of steps we reach A with $Z = (z_0, z_1, \ldots, z_k, 0, \ldots, 0)$, and have solved the first k equations of (2) for y_0, y_1, \ldots, y_k . The k^{th} equation (numbering from 0) is

$$(k+1)y_{k+1} + y_k + (n-k+1)y_{k-1} = {n \choose k} + z_k$$

and we wish to solve this for y_{k+1} . Since z_k contains all the available slack, $(k+1)y_{k+1}$ cannot be larger than R_k . Thus if R_k is negative, we cannot solve for y_{k+1} , and must alter the values of z_{k-1} , z_{k-2} , ... (step C). If $R_k \geq 0$, we must choose y_{k+1} so that ||z|| will be as large as possible (step B). This means that we want as much slack as possible in the k equation, and so y_{k+1} must be chosen as large as possible. We must now recalculate z_k , which becomes the actual slack in the k equation, and place all remaining slack at z_{k+1} .

If the new value of z_k is negative, we have reached a contradiction and must go to C where the values of z_{k-1}, z_{k-2}, \ldots will be altered. If $z_k \geq 0$ and k=n-1, we have a solution Y (step F). The actual slack in the n^{th} equation will be $y_{n-1} + y_n - 1$, and the solution has total slack $T - z_n + y_{n-1} + y_n - 1$. To obtain other solutions we go to C where z_{n-1}, z_{n-2}, \ldots are changed. Finally, if $z_k \geq 0$ and k < n-1, we replace k by k+1 (step G), and then return to A to solve the next equation for the next Y value.

When we arrive at C from A, B, or F, we must alter the first k slack variables $z_0, z_1, \ldots, z_{k-1}$ in such a way that ||z|| decreases, but does so as little as possible. This is accomplished by decreasing by one the last nonzero z_j . Thus in step C we select the largest j < k with $z_j \neq 0$. If no such j exists (as must eventually happen) the algorithm terminates at E. Otherwise (step D) we decrease z_j by one, put all remaining slack at z_{j+1} , and return to A. Since $z_0, z_1, \ldots, z_{j-1}$ have not been changed, y_0, y_1, \ldots, y_j also remain fixed, and we begin our calculations with y_{j+1} - that is, with k = j in step A.

As an example, we take n=5, $\Sigma\,y_{\alpha}\leq 7$, and total allowable slack T=10. The first column of Table 1 names the step, and the remaining columns give the values of k, R_k , j, Z, and Y at that step. Opposite a step we have entered only the values which are calculated at that step, and other variable values remain unchanged from the preceding step. The first few iterations are rather uninteresting and no solutions are obtained. We omit them, and begin at A with $z_0=2$, $z_1=8$, $y_0=1$, $y_1=2$, and k=1. We give several steps during which a solution Y=(1,2,2,0,1,1), Z=(2,6,0,0,1,1) is obtained (F). If the algorithm is continued to completion, a total of seventeen solutions of (1) with $y_0=1$ and $\Sigma\,y_{\alpha}\leq 7$ are obtained, two with $\Sigma\,y_{\alpha}=6$, and fifteen with $\Sigma\,y_{\alpha}=7$. The computations required only a few seconds on the IBM 7040 computer.

Step	k	R _k	j	^z 0	z ₁	^z 2	^z 3	^z 4	^z 5	^у 0	у ₁	у ₂	у ₃	у ₄	у ₅
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A		6				_				İ					ł
B G					8	0						3			
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А С		- 1													
C			1												
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G G	3					U	2						0		
A	5	6													
В		O					0	2						1	1
C	4						U	4		1				1	
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D	1				5	3									1
A		3													ļ

Table 1

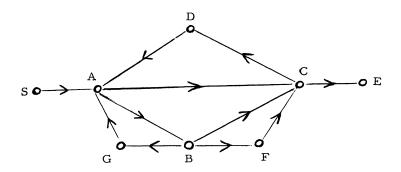


Figure 1

REFERENCES

- 1. D.R. Cox and H.D. Miller, The theory of stochastic processes. (Methuen and Co. Ltd., London, 1965) 129-132.
- 2. P. Henrici, Discrete variable methods in ordinary differential equations. (Wiley, 1962).
- 3. Mark Kac, Random walk and the theory of Brownian motion. American Mathematical Monthly 54 (1947), 369-391.
- 4. R.G. Stanton and J.G. Kalbfleisch, Covering problems for dichotomized matchings. Aequationes Mathematicae 1 (1968) 94-103.
- 5. J.J. Sylvester, Théorème sur les déterminants de M. Sylvester, Nouv. Annales de Math. 13 (1854) 305.

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