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Sum Decompositions of Symmetric Matrices

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ABSTRACT

Given a symmetric $n \times n$ matrix A and n numbers r_1, \dots, r_n , necessary and sufficient conditions for the existence of a matrix B , with a given zero pattern, with row sums r_1, \dots, r_n , and such that $A = B + B^T$ are proven. If the pattern restriction

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is relaxed, then such a matrix B exists if and only if the sum $r_1 + \dots + r_n$ is equal to half the sum of the elements of A . The case where A and B are nonnegative matrices is solved as well.

1. INTRODUCTION

The question of the existence of an entrywise nonnegative $m \times n$ matrix B with row sums r_1, \dots, r_m and column sums c_1, \dots, c_n is of long standing, e.g. [1], [4], and [5]. In particular, it follows from [1] that such a matrix B exists if and only if $r_1 + \dots + r_m = c_1 + \dots + c_n$. In fact, the author in [1] goes further and studies the existence of such a matrix with a given zero pattern. Some obviously necessary conditions turn to be also sufficient. In this paper we study the case where the matrix B is not necessarily nonnegative and where the matrix $B + B^T$ is given, namely, given a symmetric $n \times n$ matrix A and n numbers r_1, \dots, r_n , we find necessary and sufficient conditions for the existence of a matrix B satisfying

$$A = B + B^T; \quad R_i(B) = r_i, \quad i = 1, 2, \dots, n, \quad (1.1)$$

where $R_i(B)$ denotes the row sum of the i th row of B . There are two versions of this problem. In one case we also prescribe the zero pattern of the required matrix B . In the other case, called the *graph-free* case, we have no restrictions on the zero pattern. Another interesting problem is where A is nonnegative and B is required to be nonnegative.

In the next section we discuss the general problem. Of course, a necessary condition for the existence of a matrix B satisfying (1.1) is that the sum $r_1 + \dots + r_n$ is equal to half the sum $R_1(A) + \dots + R_n(A)$, that is,

$$\sum_{i=1}^n r_i = \frac{1}{2} \sum_{i,j=1}^n a_{ij}. \quad (1.2)$$

Among other results, we show that in the graph-free case the condition (1.2) is also sufficient for the existence of a general matrix B , over an arbitrary field, satisfying (1.1). Under pattern restrictions, an extra graph theoretic condition is needed. The condition (1.2) is not sufficient also in the case where A is a nonnegative matrix and we require B to be a nonnegative matrix. This harder case, which has an interpretation in the theory of network flows, is solved in Section 3.

2. THE GENERAL CASE

For a positive integer n we denote by $\langle n \rangle$ the set $\{1, \dots, n\}$. For a subset S of $\langle n \rangle$ we denote by S^c the complement of S in $\langle n \rangle$. Finally, for a digraph D we denote by $E(D)$ and $V(D)$ the arc set and the vertex set of D respectively.

We start with a few graph theoretic definitions.

DEFINITION 2.1. Let $D = (V, E)$ be a digraph. A digraph $D' = (V', E')$ is said to be a *subdigraph* of D if $V' \subseteq V$ and $E' \subseteq E$. We write $D' \subseteq D$ to indicate that D' is a subdigraph of D .

DEFINITION 2.2. A set S of vertices in a digraph D is said to be *D-loose* if for every $i \in S$ and every $j \in V(D) \setminus S$ at least one of the arcs (i, j) and (j, i) is not present in D . By convention, \emptyset and $V(D)$ are *D-loose* sets.

DEFINITION 2.3. The *symmetric closure* \bar{D} of a digraph D is the digraph with $V(\bar{D}) = V(D)$, and where (i, j) is an arc in \bar{D} whenever (i, j) and/or (j, i) is an arc in D .

DEFINITION 2.4. Let A be an $n \times n$ matrix. The *digraph* $D(A)$ of A is defined as the digraph with vertex set $\langle n \rangle$, and where (i, j) is an arc in $D(A)$ if and only if $a_{ij} \neq 0$.

We can now state the main theorem of this section.

THEOREM 2.5. Let A be a symmetric $n \times n$ matrix over an arbitrary field \mathbf{F} with characteristic different from 2, let r_1, \dots, r_n be n numbers in \mathbf{F} , and let D be a digraph satisfying $D(A) \subseteq \bar{D}$. The following are equivalent:

- (i) There exists an $n \times n$ matrix B over \mathbf{F} , with $D(B) \subseteq D$, with row sums r_1, \dots, r_n , and such that $A = B + B^T$.
- (ii) We have

$$\sum_{i \in S} r_i = \frac{1}{2} \sum_{i, j \in S} a_{ij} + \sum_{(i, j) \in S \times S^c \cap E(D)} a_{ij} \quad \text{for every } D\text{-loose set } S. \quad (2.6)$$

Proof. (i) \Rightarrow (ii): It follows from (i) that for every subset S of $\langle n \rangle$ we have

$$\sum_{i \in S} r_i = \frac{1}{2} \sum_{i, j \in S} a_{ij} + \sum_{(i, j) \in S \times S^c \cap E(D)} b_{ij}.$$

Let S be a D -loose set, and let $(i, j) \in S \times S^c \cap E(D)$. Since S is a D -loose set, we have $(j, i) \notin E(D)$, and since $A = B + B^T$, it follows that $b_{ij} = a_{ij}$. Therefore, we have

$$\begin{aligned} \sum_{i \in S} r_i &= \frac{1}{2} \sum_{i, j \in S} a_{ij} + \sum_{(i, j) \in S \times S^c \cap E(D)} b_{ij} \\ &= \frac{1}{2} \sum_{i, j \in S} a_{ij} + \sum_{(i, j) \in S \times S^c \cap E(D)} a_{ij}. \end{aligned}$$

(ii) \Rightarrow (i): Define a set of numbers $\{X_S : S \subseteq \langle n \rangle\}$ by

$$X_{\emptyset} = 0,$$

$$X_{\{i\}} = r_i - \frac{1}{2}a_{ii}, \quad i \in \langle n \rangle, \tag{2.7}$$

$$X_{\{i, j\}} = X_{\{i\}} + X_{\{j\}} - a_{ij}, \quad i, j \in \langle n \rangle, \quad i \neq j, \tag{2.8}$$

$$X_S = \sum_{i \in S} X_{\{i\}} - \sum_{\substack{i, j \in S \\ i < j}} (X_{\{i\}} + X_{\{j\}} - X_{\{i, j\}}), \quad S \subseteq \langle n \rangle, \quad |S| > 2. \tag{2.9}$$

Assume that $(i, j), (j, i) \notin E(D)$. Since $D(A) \subseteq \bar{D}$, it follows that $(i, j) \notin E(D(A))$, and by (2.8) we have

$$X_{\{i, j\}} = X_{\{i\}} + X_{\{j\}} \quad \text{whenever } (i, j), (j, i) \notin E(D). \tag{2.10}$$

Let S be a D -loose set with $|S| > 2$. By (2.7) and (2.9) we have

$$\begin{aligned} X_S &= \sum_{i \in S} X_{\{i\}} - \sum_{\substack{i, j \in S \\ i < j}} (X_{\{i\}} + X_{\{j\}} - X_{\{i, j\}}) \\ &= \sum_{i \in S} r_i - \frac{1}{2} \sum_{i \in S} a_{ii} - \sum_{\substack{i, j \in S \\ i < j}} (X_{\{i\}} + X_{\{j\}} - X_{\{i, j\}}). \end{aligned}$$

By (2.6) we now obtain

$$\begin{aligned} X_S &= \frac{1}{2} \sum_{i,j \in S} a_{ij} + \sum_{(i,j) \in S \times S^c \cap E(D)} a_{ij} - \sum_{\substack{i,j \in S \\ i < j}} (X_{(i)} + X_{(j)} - X_{(i,j)}) \\ &= \sum_{\substack{i,j \in S \\ i < j}} a_{ij} + \sum_{(i,j) \in S \times S^c \cap E(D)} a_{ij} - \sum_{\substack{i,j \in S \\ i < j}} (X_{(i)} + X_{(j)} - X_{(i,j)}), \end{aligned}$$

which, in view of (2.8), yields that

$$X_S = \sum_{(i,j) \in S \times S^c \cap E(D)} (X_{(i)} + X_{(j)} - X_{(i,j)}) \quad \text{for every } D\text{-loose set } S. \tag{2.11}$$

In particular, it follows from (2.11) that $X_{\langle n \rangle} = 0$. By Theorem (3.6) of [2], it follows from (2.9), (2.10), and (2.11) that there exists an $n \times n$ matrix B , with $D(B) \subseteq D$, such that

$$\sum_{(i,j) \in S \times S^c} b_{ij} = X_S, \quad S \subseteq \langle n \rangle. \tag{2.12}$$

Define b_{ii} to be $\frac{1}{2}a_{ii}$, $i \in \langle n \rangle$. It then follows from (2.12) that the i th row sum of B is $b_{ii} + X_{(i)}$, which, by (2.7), is equal to r_i . Also, it follows from (2.12) that

$$b_{ij} + b_{ji} = X_{(i)} + X_{(j)} - X_{(i,j)}, \quad i, j \in \langle n \rangle, \quad i \neq j.$$

In view of (2.8), we have $b_{ij} + b_{ji} = a_{ij}$, $i, j \in \langle n \rangle$, $i \neq j$, and so $A = B + B^T$. ■

REMARK 2.13. A similar result holds for Hermitian complex matrices A and a matrix B that is required to satisfy $A = B + B^*$. In this case, (2.6) should be replaced by

$$\operatorname{Re} \left(\sum_{i \in S} r_i \right) = \frac{1}{2} \sum_{i,j \in S} a_{ij} + \sum_{(i,j) \in S \times S^c \cap E(D)} a_{ij}$$

whenever S is a D -loose set.

If we choose D to be complete digraph with n vertices, then the only D -loose sets are \emptyset $\langle n \rangle$, and we obtain the following graph-free version of Theorem 2.5.

THEOREM 2.14. *Let A be a symmetric $n \times n$ matrix over an arbitrary field \mathbf{F} with characteristic different from 2, and let r_1, \dots, r_n be n numbers in \mathbf{F} . The following are equivalent:*

- (i) *There exists an $n \times n$ matrix B over \mathbf{F} with row sums r_1, \dots, r_n and such that $A = B + B^T$.*
- (ii) *We have*

$$\sum_{i=1}^n r_i = \frac{1}{2} \sum_{i,j=1}^n a_{ij}.$$

As an interesting corollary of Theorem 2.14 we can obtain the following.

THEOREM 2.15. *Let r_1, \dots, r_n and c_1, \dots, c_n be real numbers. The following are equivalent:*

- (i) *There exists a real $n \times n$ matrix B , with row sums r_1, \dots, r_n and column sums c_1, \dots, c_n , such that $B + B^T$ is nonnegative entrywise.*
- (ii) *We have $r_1 + \dots + r_n = c_1 + \dots + c_n$ and $r_i + c_i \geq 0$, $i \in \langle n \rangle$.*

Proof. (i) \Rightarrow (ii): Clearly, since r_1, \dots, r_n are row sums and c_1, \dots, c_n are column sums of a matrix B , we have $r_1 + \dots + r_n = c_1 + \dots + c_n$. Since $r_i + c_i$ is the i th row sum of the nonnegative matrix $B + B^T$, it follows that $r_i + c_i \geq 0$.

(ii) \Rightarrow (i): Let $r_i = r_i + c_i$, $i \in \langle n \rangle$. By [1], there exists a nonnegative matrix C with row sums and column sums s_1, \dots, s_n . Hence, $A = \frac{1}{2}(C + C^T)$ is a nonnegative symmetric matrix with row sums s_1, \dots, s_n . We have

$$\sum_{i=1}^n r_i = \frac{1}{2} \left(\sum_{i=1}^n r_i + \sum_{i=1}^n c_i \right) = \frac{1}{2} \sum_{i=1}^n s_i = \frac{1}{2} \sum_{i,j} a_{ij}.$$

By Theorem 2.14, there exists a real matrix B with row sums r_1, \dots, r_n and such that $B + B^T = A$. Note that the i th column sum of B is $s_i - r_i = c_i$. ■

COROLLARY 2.16. *Let r_1, \dots, r_n be real numbers. The following are equivalent:*

- (i) *There exists a real $n \times n$ matrix B with row sums r_1, \dots, r_n and such that $B + B^T$ is nonnegative.*
- (ii) *We have $r_1 + \dots + r_n \geq 0$.*

Proof. (i) \Rightarrow (ii) follows because $r_1 + \cdots + r_n$ is equal to half the sum of the elements of $B + B^T$.

(ii) \Rightarrow (i): Let $r = r_1 + \cdots + r_n$, and define $c_i = -r_i + 2r/n$. Observe that we have both $r_1 + \cdots + r_n = c_1 + \cdots + c_n$ and $r_i + c_i \geq 0$, $i \in \langle n \rangle$. By Theorem 2.15, there exists a real $n \times n$ matrix B , with row sums r_1, \dots, r_n and column sums c_1, \dots, c_n , such that $B + B^T$ nonnegative. ■

3. THE NONNEGATIVE CASE

Theorems 2.5 and 2.14 do not hold in general if A is a nonnegative matrix and B is required to be a nonnegative matrix, as is demonstrated by the following example.

EXAMPLE 3.1. Let

$$A = \begin{pmatrix} 1 & 1 \\ 1 & 5 \end{pmatrix},$$

and let $r_1 = 3$ and $r_2 = 1$. Observe that for every nonnegative matrix B such that $A = B + B^T$ we must have $b_{22} = \frac{1}{2}a_{22} = 2.5$ and $b_{21} \geq 0$. Hence, $R_2(B) \geq 2.5$, and so, although the condition (1.2) is satisfied, there exists no nonnegative matrix B with row sums r_1, r_2 and such that $A = B + B^T$.

As is observed in Example 3.1, a necessary condition in the nonnegative case is

$$r_i \geq \frac{1}{2}a_{ii}, \quad i \in \langle n \rangle. \quad (3.2)$$

In the sequel we shall show that for $n \leq 2$ the conditions (1.2) and (3.2) form a necessary and sufficient condition for the existence of a nonnegative matrix B , with no pattern restrictions, satisfying (1.1); see Corollary 3.23. However, we shall show that for $n > 2$ the conditions (1.2) and (3.2) are not sufficient. In the latter case we need a generalized version of (3.2), that is, (3.4) in the case that the pattern of B is prescribed, and (3.20) in the graph-free case.

Our main result here is the following.

THEOREM 3.3. *Let A be a nonnegative symmetric $n \times n$ matrix, let D be a digraph, with loops on all vertices, satisfying $D(A) \subseteq \bar{D}$, and let r_1, \dots, r_n be nonnegative numbers. The following are equivalent:*

(i) *There exists a nonnegative $n \times n$ matrix B , with $D(B) \subseteq D$, with row sums r_1, \dots, r_n , and such that $A = B + B^T$.*

(ii) We have (2.6) and

$$\sum_{i \in S} r_i \geq \frac{1}{2} \sum_{i,j \in S} a_{ij} + \sum_{\substack{(i,j) \in S \times S^c \\ (j,i) \notin E(D)}} a_{ij} \quad \text{for every } S \subseteq \langle n \rangle. \quad (3.4)$$

Proof. (i) \Rightarrow (ii): By Theorem 2.5, (i) implies (2.6). Let S be a subset of $\langle n \rangle$. We have

$$\begin{aligned} \sum_{i \in S} r_i &= \sum_{(i,j) \in S \times \langle n \rangle} b_{ij} = \sum_{i,j \in S} b_{ij} + \sum_{(i,j) \in S \times S^c} b_{ij} \\ &\geq \frac{1}{2} \sum_{i,j \in S} (b_{ij} + b_{ji}) + \sum_{\substack{(i,j) \in S \times S^c \\ (j,i) \notin E(D)}} b_{ij}. \end{aligned} \quad (3.5)$$

Since $b_{ij} + b_{ji} = a_{ij}$, $i, j \in \langle n \rangle$, and since $D(B) \subseteq D$, it follows that whenever $(j, i) \notin E(D)$ we have $b_{ij} = a_{ij}$, and so (3.4) follows from (3.5).

(ii) \Rightarrow (i): Define an $n \times n$ matrix \tilde{A} by

$$\tilde{a}_{ij} = \begin{cases} a_{ij}, & (i, j), (j, i) \in E(D), \\ 0 & \text{otherwise,} \end{cases}$$

and define the numbers $\tilde{r}_1, \dots, \tilde{r}_n$ by

$$\tilde{r}_i = r_i - \sum_{\substack{j=1 \\ (j,i) \notin E(D)}}^n a_{ij}.$$

First, observe that by (3.4) we have

$$\begin{aligned} \sum_{i \in S} r_i &\geq \frac{1}{2} \sum_{i,j \in S} a_{ij} + \sum_{\substack{(i,j) \in S \times S^c \\ (j,i) \notin E(D)}} a_{ij} \\ &= \frac{1}{2} \sum_{\substack{i,j \in S \\ (i,j), (j,i) \in E(D)}} a_{ij} + \sum_{\substack{i,j \in S \\ (i,j) \notin E(D)}} a_{ij} + \sum_{\substack{(i,j) \in S \times S^c \\ (j,i) \notin E(D)}} a_{ij} \\ &= \frac{1}{2} \sum_{i,j \in S} \tilde{a}_{ij} + \sum_{\substack{(i,j) \in S \times \langle n \rangle \\ i \neq j, (j,i) \notin E(D)}} a_{ij}, \end{aligned}$$

which implies

$$\sum_{i \in S} \tilde{r}_i \geq \frac{1}{2} \sum_{i, j \in S} \tilde{a}_{ij}. \tag{3.6}$$

We shall now prove that the following linear system of $m = (n^2 + 3n)/2$ equations with n^2 variables x_{ij} , $i, j \in \langle n \rangle$, has a nonnegative solution:

$$\begin{aligned} x_{ij} + x_{ji} &= \tilde{a}_{ij} & i, j \in \langle n \rangle, \quad i \leq j, \\ \sum_{j=1}^n x_{ij} &= \tilde{r}_i & i \in \langle n \rangle. \end{aligned} \tag{3.7}$$

Let C be the coefficient matrix of the system (3.7), and let E be the m -vector whose elements are the right hand sides of the equations in (3.7). By the Farkas lemma, e.g. Corollary 7.1d in [6, p. 89], the system (3.7) has a nonnegative solution if and only if for every m -vector F satisfying $F^T C \geq 0$ we have $F^T E \geq 0$. Denote the element of F that corresponds to (that is, is in the same place as) \tilde{a}_{ij} by f_{ij} , and the element of F that corresponds to \tilde{r}_i by f_i . Since each variable x_{ij} appears in exactly two equations (in the equation $x_{ij} + x_{ji} = \tilde{a}_{ij}$ and in the i th row sum equation), it follows that each column of C contains exactly two nonzero elements (two 1's if it corresponds to x_{ij} where $i \neq j$, and a 2 and a 1 if it corresponds to x_{ii}) and the rest zeros, and that F satisfies $F^T C \geq 0$ if and only if

$$f_i + 2f_{ii} \geq 0, \quad f_i + f_{ij} \geq 0, \quad f_j + f_{ij} \geq 0, \quad i, j \in \langle n \rangle. \tag{3.8}$$

Let F be a vector satisfying (3.8). If F is a nonnegative vector, then clearly we have $F^T E \geq 0$. If F has some negative elements, then we shall show that $F^T E \geq \bar{F}^T E$ for some vector \bar{F} satisfying

$$\bar{f}_i + 2\bar{f}_{ii} \geq 0, \quad \bar{f}_i + \bar{f}_i \bar{f}_{ij} \geq 0, \quad \bar{f}_j + \bar{f}_{ij} \geq 0, \quad i, j \in \langle n \rangle \tag{3.9}$$

and such that \bar{F} has less negative elements than F has. Repeating this procedure, we end up with a nonnegative vector \bar{F} such that $\bar{F}^T E \geq F^T E$. Since we have $\bar{F}^T E \geq 0$, it follows that $F^T E \geq 0$, and our claim follows by the Farkas lemma.

Assume first that some of the f_i 's are negative, and let $T = \{i : f_i < 0\}$. Let $f = \min_{i \in T} |f_i|$. By (3.8) we have

$$f_{ij} \geq f \quad \forall i, j, \quad \{i, j\} \cap T \neq \emptyset, \tag{3.10}$$

and

$$f_{ii} \geq \frac{1}{2}f, \quad i \in T. \tag{3.11}$$

Since $\langle n \rangle$ is a D -loose set, by (2.6) we have

$$\sum_{i \in \langle n \rangle} r_i = \frac{1}{2} \sum_{i, j \in \langle n \rangle} a_{ij}.$$

Hence,

$$\sum_{i \in \langle n \rangle} r_i = \frac{1}{2} \sum_{i, j \in \langle n \rangle} \tilde{a}_{ij} + \sum_{\substack{i, j \in \langle n \rangle \\ i \neq j, (j, i) \notin E}} a_{ij},$$

which implies

$$\sum_{i \in \langle n \rangle} \tilde{r}_i = \frac{1}{2} \sum_{i, j \in \langle n \rangle} \tilde{a}_{ij}. \tag{3.12}$$

By (3.6) we have

$$\sum_{i \in T^c} \tilde{r}_i \geq \frac{1}{2} \sum_{i, j \in T^c} \tilde{a}_{ij}. \tag{3.13}$$

Subtracting (3.13) from (3.12) yields

$$\sum_{i \in T} \tilde{r}_i \leq \frac{1}{2} \sum_{i \in T \text{ and/or } j \in T} \tilde{a}_{ij}.$$

Therefore,

$$\frac{1}{2}f \sum_{i \in T \text{ and/or } j \in T} \tilde{a}_{ij} - f \sum_{i \in T} \tilde{r}_i \geq 0,$$

and hence

$$F^T E \geq F^T E - \left(\frac{1}{2}f \sum_{i \in T \text{ and/or } j \in T} \tilde{a}_{ij} - f \sum_{i \in T} \tilde{r}_i \right) = \bar{F}^T E. \tag{3.14}$$

In order to find the elements of \bar{F} , observe that if $i \in T, i \neq j$, then both \tilde{a}_{ij} and \tilde{a}_{ji} appear in the corresponding sum in (3.14), and so \tilde{a}_{ij} actually appears twice. Therefore, the elements of \bar{F} are

$$\bar{f}_i = \begin{cases} f_i, & i \notin T, \\ f_i + f, & i \in T, \end{cases}$$

and

$$\bar{f}_{ij} = \begin{cases} f_{ij}, & \{i, j\} \cap T = \emptyset, \\ f_{ij} - f, & \{i, j\} \cap T \neq \emptyset, \quad i \neq j, \\ f_{ii} - \frac{1}{2}f, & i \in T, \quad i = j. \end{cases}$$

Obviously, we have $f_i \geq 0 \Rightarrow \bar{f}_i \geq 0$. Also, it follows by (3.10) and (3.11) that $f_{ij} \geq 0 \Rightarrow \bar{f}_{ij} \geq 0$. Furthermore, for at least one $i \in T$ we have $f_i = -f$ and now $\bar{f}_i = 0$. Therefore, \bar{F} has less negative elements than F has.

Now, for $i \neq j$ we have

$$\bar{f}_i + \bar{f}_{ij} = \begin{cases} f_i + f_{ij}, & i \in T, \\ f_i + f_{ij} - f, & i \notin T, \quad j \in T, \\ f_i + f_{ij}, & \{i, j\} \cap T = \emptyset. \end{cases} \tag{3.15}$$

By (3.8) we have $f_i + f_{ij} \geq 0$. By definition of T , whenever $i \notin T$ we have $f_i \geq 0$. Also, by (3.10) we have $f_{ij} \geq f$ whenever $j \in T$. Hence it follows from (3.15) that $\bar{f}_i + \bar{f}_{ij} \geq 0$. Quite similarly we show that $\bar{f}_j + \bar{f}_{ij} \geq 0$. Also, for every i we have $f_i + 2\bar{f}_{ii} = f_i + 2f_{ii}$, and so the vector \bar{F} satisfies (3.9).

Now assume that the f_i 's are all nonnegative but some of the f_{ij} 's are negative. Let $T = \{i : f_{ij} < 0 \text{ or } f_{ji} < 0 \text{ for some } j\}$, and let

$$f = \min\{\{|f_{ij}| : f_{ij} < 0, i \neq j\}, \{2|f_{ii}| : f_{ii} < 0\}\}.$$

By (3.8) we have

$$f_i \geq f \quad \forall i \in T. \tag{3.16}$$

By (3.6) we have

$$f \sum_{i \in T} \tilde{r}_i - \frac{1}{2}f \sum_{i, j \in T} \tilde{a}_{ij} \geq 0.$$

Thus,

$$F^T E \geq F^T E - \left(f \sum_{i \in T} \bar{r}_i - \frac{1}{2} f \sum_{i, j \in T} \bar{a}_{ij} \right) = \bar{F}^T E,$$

where the elements of \bar{F} are

$$\bar{f}_i = \begin{cases} f_i, & i \notin T, \\ f_i - f, & i \in T, \end{cases}$$

and

$$\bar{f}_{ij} = \begin{cases} f_{ij} + f, & i, j \in T, \quad i \neq j, \\ f_{ii} + \frac{1}{2} f, & i \in T, \quad i = j, \\ f_{ij}, & \text{otherwise.} \end{cases}$$

Obviously, we have $f_{ij} \geq 0 \Rightarrow \bar{f}_{ij} \geq 0$. Also, it follows by (3.16) that $f_i \geq 0 \Rightarrow \bar{f}_i \geq 0$. Furthermore, either we have at least one pair of indices $i, j, i \neq j$, such that $f_{ij} = -f$ and now $\bar{f}_{ij} = 0$, or we have at least one index i such that $f_{ii} = -\frac{1}{2}f$ and now $\bar{f}_{ii} = 0$. Therefore, \bar{F} has less negative elements than F has.

Now, for $i \neq j$ we have

$$\bar{f}_i + \bar{f}_{ij} = \begin{cases} f_i + f_{ij}, & i, j \in T, \\ f_i - f + f_{ij}, & i \in T, \quad j \notin T, \\ f_i + f_{ij}, & i \notin T. \end{cases} \tag{3.17}$$

By (3.8) we have $f_i + f_{ij} \geq 0$. By (3.16), whenever $i \in T$ we have $f_i \geq f$. Also, by the definition of T we have $f_{ij} \geq 0$ whenever $j \notin T$. Hence it follows from (3.17) that $\bar{f}_i + \bar{f}_{ij} \geq 0$. Quite similarly we show that $\bar{f}_j + \bar{f}_{ij} \geq 0$. Also, for every i we have $\bar{f}_i + 2\bar{f}_{ii} = f_i + 2f_{ii}$, and so the vector \bar{F} satisfies (3.9).

As is explained above, it now follows by the Farkas lemma that the system (3.7) has a nonnegative solution $\{x_{ij} : i, j \in \langle n \rangle\}$. Let B be the $n \times n$ matrix defined by

$$b_{ij} = \begin{cases} x_{ij}, & (i, j), (j, i) \in E(D), \\ a_{ij}, & (j, i) \notin E(D), \\ 0, & (i, j) \notin E(D). \end{cases} \tag{3.18}$$

Observe that since $D(A) \subseteq \bar{D}$, if $(i, j), (j, i) \notin E(D)$ then $(i, j) \notin D(A)$ and so $a_{ij} = 0$. Therefore, the matrix B is well defined by (3.18). Let $i, j \in \langle n \rangle$. If $(i, j), (j, i) \in E(D)$ then, by (3.18), $b_{ij} + b_{ji} = x_{ij} + x_{ji} = \tilde{a}_{ij} = a_{ij}$. If $(i, j) \in E(D)$ and $(j, i) \notin E(D)$ then, by (3.18), $b_{ij} + b_{ji} = a_{ij} + 0 = a_{ij}$. If $(i, j) \notin E(D)$ and $(j, i) \in E(D)$ then, by (3.18), $b_{ij} + b_{ji} = 0 + a_{ji} = a_{ij}$. If $(i, j), (j, i) \notin E(D)$ then, by (3.18), $b_{ij} + b_{ji} = 0 = a_{ij}$. Thus, in any case, $b_{ij} + b_{ji} = a_{ij}$, and so $A = B + B^T$. Also, if $(i, j) \notin E(D)$ then, by (3.18), $b_{ij} = 0$, and so $(i, j) \notin D(B)$, implying that $D(B) \subseteq D$. Finally, we have

$$\begin{aligned} R_i(B) &= \sum_{j=1}^n b_{ij} = \sum_{\substack{j \in \langle n \rangle \\ (i, j), (j, i) \in E(D)}} x_{ij} + \sum_{\substack{j=1 \\ (j, i) \notin E(D)}}^n a_{ij} \\ &= \tilde{r}_i + \sum_{\substack{j=1 \\ (j, i) \notin E(D)}}^n a_{ij} = r_i, \quad i \in \langle n \rangle, \end{aligned}$$

and hence B is the required matrix. ■

The solution of the graph-free case is obtained when D is chosen to be the complete digraph with n vertices.

THEOREM 3.19. *Let A be a nonnegative symmetric $n \times n$ matrix, and let r_1, \dots, r_n be nonnegative numbers. The following are equivalent:*

- (i) *There exists a nonnegative $n \times n$ matrix B with row sums r_1, \dots, r_n and such that $A = B + B^T$.*
- (ii) *We have (1.2), and for every $S \subseteq \langle n \rangle$ we have*

$$\sum_{i \in S} r_i \geq \frac{1}{2} \sum_{i, j \in S} a_{ij}. \tag{3.20}$$

REMARK 3.21. Let S be a subset of $\langle n \rangle$ with $|S| > k$. Then the complement S^c of S satisfies $|S^c| \leq n - k - 1$. By subtracting the inequality (3.2) from (1.2) we obtain

$$\sum_{i \in S^c} r_i \leq \frac{1}{2} \sum_{i \in S^c \text{ and/or } j \in S^c} a_{ij}.$$

Since $|S^c| = n - |S|$, condition (ii) in Theorem 3.19 is equivalent to each of the following conditions:

(iii) The equality (1.2) holds, and for some nonnegative integer k , $k < n$, the inequality (3.20) holds for every subset S of $\langle n \rangle$ with $1 \leq |S| \leq k$, and the inequality

$$\sum_{i \in S} r_i \leq \frac{1}{2} \sum_{i \in S \text{ and/or } j \in S} a_{ij} \quad (3.22)$$

holds for every subset S of $\langle n \rangle$ with $1 \leq |S| \leq n - k - 1$.

(iv) The equality (1.2) holds, and the inequalities (3.20) and (3.22) hold for every subset S of $\langle n \rangle$.

In the special case of $n = 2$ we have the following immediate corollary of Theorem (3.19).

COROLLARY 3.23. *Let A be a nonnegative symmetric 2×2 matrix, and let r_1, r_2 be nonnegative numbers. The following are equivalent:*

(i) *There exists a nonnegative 2×2 matrix B with row sums r_1, r_2 and such that $A = B + B^T$.*

(ii) *We have (1.2) and (3.2).*

It is natural to ask whether (1.2) and (3.2) form a necessary and sufficient condition for the existence of a nonnegative $n \times n$ matrix B satisfying (1.1) also in the case $n \geq 3$. The answer to this question is negative, as demonstrated by the following example.

EXAMPLE 3.24. Let

$$A = \begin{pmatrix} 0 & 8 & 2 \\ 8 & 4 & 4 \\ 2 & 4 & 4 \end{pmatrix},$$

and let $r_1 = 4$, $r_2 = 5$, and $r_3 = 9$. It is easy to verify that we have both (1.2) and (3.2). Nevertheless, since (3.20) is not satisfied for the set $S = \{1, 2\}$, it follows from Theorem 3.19 that there exists no nonnegative 3×3 matrix B with row sums r_1, r_2, r_3 and such that $A = B + B^T$.

We conclude the paper with a characterization of the positive decomposition case.

THEOREM 3.25. *Let A be a positive (entrywise) symmetric $n \times n$ matrix, and let r_1, \dots, r_n be positive numbers. The following are equivalents:*

(i) *There exists a positive $n \times n$ matrix B with row sums r_1, \dots, r_n and such that $A = B + B^T$.*

(ii) *We have (1.2), as well as a strict inequality in (3.20) for every subset S of $\langle n \rangle$.*

Proof. (i) \Rightarrow (ii): Since $b_{ij} > 0$ for all $i, j \in \langle n \rangle$, using the very same proof as for the corresponding implication in Theorem 3.19, we obtain strict inequalities in (3.20).

(ii) \Rightarrow (i): Define a new symmetric matrix \bar{A} by $\bar{a}_{ij} = a_{ij} - \epsilon$, $i, j \in \langle n \rangle$, and n new numbers $\bar{r}_1, \dots, \bar{r}_n$ by $\bar{r}_i = r_i - n\epsilon/2$, $i \in \langle n \rangle$. If we choose $\epsilon > 0$ sufficiently small, then the matrix \bar{A} is positive and the numbers $\bar{r}_1, \dots, \bar{r}_n$ are positive. Furthermore, since we have strict inequalities in (3.20), the inequalities there still hold if we replace a_{ij} by \bar{a}_{ij} and r_i by \bar{r}_i . By Theorem 3.19, there exists a nonnegative matrix \bar{B} with row sums $\bar{f}_1, \dots, \bar{r}_n$ and such that $\bar{A} = \bar{B} + \bar{B}^T$. The matrix B defined by $b_{ij} = \bar{b}_{ij} + \epsilon/2$, $i, j \in \langle n \rangle$, is a positive matrix with row sums r_1, \dots, r_n and such that $A = B + B^T$. ■

Professor Alan Hoffman and a referee have pointed out to us that network flow techniques which were used to solve a related problem on tournaments in [7] and [3] could be used to derive the results of Section 3. We hope to exploit network flow techniques in some generalizations in a forthcoming paper.

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