

## Projecting a Simplex onto Another One

M. DEZA AND H. MAEHARA

For a given  $n$ -simplex  $T$  in  $\mathbb{R}^{2n-1}$ , the size of a regular  $n$ -simplex in  $\mathbb{R}^{2n-1}$  which orthogonally projects onto  $T$ , and the size of a regular  $n$ -simplex onto which  $T$  projects orthogonally, are described in terms of spectra of a matrix associated with  $T$ .

### 1. INTRODUCTION

For two  $n$ -simplices  $S$  and  $T$  in Euclidean space, we write

$$(S \rightarrow T)_N$$

if we can place  $S$  and  $T$  in  $\mathbb{R}^N$  so that  $S$  is mapped to  $T$  by the orthogonal projection from  $\mathbb{R}^N$  onto an  $n$ -dimensional subspace. The symbol  $\Delta$  denotes the regular  $n$ -simplex of unit edge-length, and  $s\Delta$  denotes the regular  $n$ -simplex of edge-length  $s$ .

For any  $n$ -simplex  $T$ , there exists an  $N > 0$  and an  $s > 0$  such that  $(s\Delta \rightarrow T)_N$ . To see this, consider, for example, a triangle  $T$  in  $\mathbb{R}^2$  with vertices the row vector of the matrix

$$\begin{pmatrix} 0 & 0 \\ 1 & 0 \\ 2 & 1 \end{pmatrix}.$$

Putting, instead of each column, all of its permutations, we obtain

$$\begin{pmatrix} 0 & 0 & 1 & 1 & 2 & 2 & 0 & 0 & 1 \\ 1 & 2 & 0 & 2 & 1 & 0 & 0 & 1 & 0 \\ 2 & 1 & 2 & 0 & 0 & 1 & 1 & 0 & 0 \end{pmatrix}.$$

Then, the triangle in  $\mathbb{R}^9$  with vertices the row vectors of this matrix is equilateral and it clearly projects  $S$  onto a triangle congruent to  $T$ . Thus  $(s\Delta \rightarrow T)_9$  for some  $s$ . A similar argument will clearly work for any simplex.

Now, for a given  $n$ -simplex  $T$ , how small  $s$  can be and how small  $N$  can be?

We are going to answer this question. Let  $\mathbf{I}$  and  $\mathbf{J}$  respectively denote the  $(n + 1) \times (n + 1)$  identity matrix and the  $(n + 1) \times (n + 1)$  matrix with all entries 1. Then

$$\mathbf{P} := \mathbf{I} - (1/(n + 1))\mathbf{J}$$

is the matrix of the projection  $\mathbb{R}^{n+1} \rightarrow V$ , where

$$V = \{\mathbf{x} \in \mathbb{R}^{n+1} : \mathbf{x} \cdot (1, 1, \dots, 1) = 0\}.$$

The *square-distance matrix* of an  $m$ -point set is an  $m \times m$  matrix, the  $(i, j)$ -entry of which is the square of the distance between the points  $i$  and  $j$ .

**THEOREM 1.** *Let  $T$  be an  $n$ -simplex and let  $\mathbf{D}$  be the square-distance matrix of the vertex set of  $T$ . Let  $\lambda_{\max}$  be the maximum eigenvalue of  $-\mathbf{PDP}$  with multiplicity  $\nu$ . Then:*

- (1)  $(s\Delta \rightarrow T)_N$  for some  $s > 0 \Leftrightarrow N \geq 2n - \nu$ ;
- (2)  $N \geq 2n$  and  $s \geq (\lambda_{\max})^{\frac{1}{2}} \Rightarrow (s\Delta \rightarrow T)_N$ ;
- (3)  $2n - \nu \leq N \leq 2n - 1$  and  $(s\Delta \rightarrow T)_N \Rightarrow s = (\lambda_{\max})^{\frac{1}{2}}$ .

**THEOREM 2.** Let  $T$  be an  $n$ -simplex and let  $\mathbf{D}$  be the square-distance matrix of the vertex set of  $T$ . Let  $\lambda_{\min}$  be the minimum positive eigenvalue of  $-\mathbf{PDP}$  with multiplicity  $\nu'$ . Then:

- (1)  $(T \rightarrow s\Delta)_n$  for some  $s > 0 \Leftrightarrow N \geq 2n - \nu'$ ;
- (2)  $N \geq 2n$  and  $0 < s \leq (\lambda_{\min})^{\frac{1}{2}} \Rightarrow (T \rightarrow s\Delta)_N$ ;
- (3)  $2n - \nu' \leq N \leq 2n - 1$  and  $(T \rightarrow s\Delta)_N \Rightarrow s = (\lambda_{\min})^{\frac{1}{2}}$ .

**EXAMPLE.** Let  $T$  be the isosceles right triangle with hypotenuse  $2^{\frac{1}{2}}$ . Then  $-\mathbf{PDP}$  is

$$(1/3)^2 \begin{pmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{pmatrix} \begin{pmatrix} 0 & 1 & 2 \\ 1 & 0 & 1 \\ 2 & 1 & 0 \end{pmatrix} \begin{pmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{pmatrix} = (2/9) \begin{pmatrix} -5 & 1 & 4 \\ 1 & -2 & 1 \\ 4 & 1 & -5 \end{pmatrix}.$$

The eigenvalues of this matrix are  $2, 2/3, 0$ . Hence

$$(2^{\frac{1}{2}}\Delta \rightarrow T)_3 \quad \text{and} \quad (T \rightarrow (2/3)^{\frac{1}{2}}\Delta)_3.$$

## 2. LEMMAS

First we recall the following classical result (see, e.g., Schoenberg [2]).

**THEOREM S.** A non-negative symmetric  $(n+1) \times (n+1)$ -matrix  $\mathbf{M} = (M_{ij})$  with  $M_{ii} = 0$  for all  $i$ , is the square-distance matrix of an  $(n+1)$ -point set in Euclidean space iff  $\mathbf{M}$  is of negative type (i.e.  $-\mathbf{PMP}$  is positive semi-definite). Furthermore, the  $(n+1)$ -point set spans a  $k$ -dimensional flat iff  $\text{rank}(\mathbf{M}) = k$ .

**LEMMA 1.** Let  $T$  be an  $n$ -simplex with maximum edge-length  $d_{\max}$  and minimum edge-length  $d_{\min}$ . Let  $\mathbf{D} = (D_{ij})$  be the square-distance matrix of the vertex set of  $T$ . Then  $-\mathbf{PDP}$  is positive semi-definite of rank  $n$  and

$$\lambda_{\max} \geq (d_{\max})^2 \quad \text{and} \quad (d_{\min})^2 \geq \lambda_{\min} > 0,$$

where  $\lambda_{\max}$  and  $\lambda_{\min}$  are the maximum eigenvalue and the minimum positive eigenvalue of  $-\mathbf{PDP}$ , respectively.

**PROOF OF LEMMA 1.** Let  $\mathbf{j} = (1, 1, \dots, 1)$ . Since  $-\mathbf{jPDP} = 0, 0$  is an eigenvalue of  $\mathbf{M} := -\mathbf{PDP}$  with associated eigenvector  $\mathbf{j}$ . Hence any eigenvector  $\mathbf{x}$  associated to any other eigenvalue is orthogonal to  $\mathbf{j}$ ; that is  $\mathbf{x} = \mathbf{xP}$ . Furthermore, since  $T$  cannot be contained in  $(n-1)$ -space,  $\mathbf{M}$  is positive semi-definite of rank  $n$ . Hence, the other eigenvalues are all positive, and it follows from the property of Rayleigh quotient (see, e.g., Bellman [1, Ch. 7]) that

$$(*) \quad \lambda_{\max} = \max_{\mathbf{vP} \neq 0} \frac{-\mathbf{vPDPv}^t}{\mathbf{vPv}^t}, \quad \lambda_{\min} = \min_{\mathbf{vP} \neq 0} \frac{-\mathbf{vPDPv}^t}{\mathbf{vPv}^t}.$$

Now, to show that  $\lambda_{\max} \geq (d_{\max})^2$ , suppose that  $d := d_{\max}$  is the distance between the vertices 1 and 2, i.e.

$$d^2 := (d_{\max})^2 = D_{12} = D_{21}.$$

Then, since

$$\begin{aligned} (1, -1, 0, \dots, 0)\mathbf{M}(1, -1, 0, \dots, 0)^t &= -(-d^2, d^2, *, \dots, *) (1, -1, 0, \dots, 0)^t \\ &= 2(d_{\max})^2, \end{aligned}$$

and

$$(1, -1, 0, \dots, 0)\mathbf{P}(1, -1, 0, \dots, 0)^t = 2,$$

we have  $\lambda_{\max} \geq (d_{\max})^2$  by (\*). Similarly,  $(d_{\min})^2 \geq \lambda_{\min}$  follows.  $\square$

LEMMA 2. Let  $T$  be an  $n$ -simplex,  $\mathbf{D}$  its square-distance matrix, and let  $\lambda_{\max}$  and  $\lambda_{\min}$  be the maximum eigenvalue (with multiplicity  $\nu$ ) and the minimum positive eigenvalue (with multiplicity  $\nu'$ ) of  $-\mathbf{PDP}$ . Then:

(1)  $s^2\mathbf{P} - (-\mathbf{PDP})$  is positive semi-definite  $\Leftrightarrow s^2 \geq \lambda_{\max}$ ,  $(-\mathbf{PDP}) - s^2\mathbf{P}$  is positive semi-definite  $\Leftrightarrow s^2 \leq \lambda_{\min}$ ; and

$$(2) \quad \begin{aligned} \text{rank}(s^2\mathbf{P} - (-\mathbf{PDP})) &= \begin{cases} n & \text{for } s^2 > \lambda_{\max}, \\ n - \nu & \text{for } s^2 = \lambda_{\max}, \end{cases} \\ \text{rank}(-\mathbf{PDP} - s^2\mathbf{P}) &= \begin{cases} n & \text{for } s^2 < \lambda_{\min}, \\ n - \nu' & \text{for } s^2 = \lambda_{\min}. \end{cases} \end{aligned}$$

PROOF. Regard the matrices  $\mathbf{P}$  and  $-\mathbf{PDP}$  as linear transformations of the vector space

$$V = \{\mathbf{x} \in \mathbb{R}^{n+1} : \mathbf{x} \cdot (1, 1, \dots, 1) = 0\}.$$

Then, (1) follows from the fact that  $\mathbf{P}$  is the identity transformation of  $V$ , and (2) follows from  $\text{rank}(-\mathbf{PDP}) = n$ .  $\square$

### 3. PROOFS OF THEOREMS 1 AND 2

PROOF OF THEOREM 1. (1) Suppose that there is an  $s > 0$  such that  $(s\Delta \rightarrow T)_N$ . Then we can place  $s\Delta$  and  $T$  in  $\mathbb{R}^N$  so that  $T = p(s\Delta)$ , where

$$p: \mathbb{R}^N = \mathbb{R}^n \times \mathbb{R}^{N-n} \rightarrow \mathbb{R}^n$$

is the projection onto the first component. Projecting the vertices of  $s\Delta$  onto the second component  $\mathbb{R}^{N-n}$ , we have  $n+1$  (not necessarily all distinct) points with square-distance matrix  $\mathbf{D}' = s^2(\mathbf{J} - \mathbf{I}) - \mathbf{D}$ . Hence the matrix  $\mathbf{D}'$  must be of negative type, i.e.  $-\mathbf{PD}'\mathbf{P} = s^2\mathbf{P} - (-\mathbf{PDP})$  is positive semi-definite. Hence  $s^2 \geq \lambda_{\max}$  by Lemma 2(1), and hence  $-\mathbf{PD}'\mathbf{P}$  has rank  $\geq n - \nu$ . Therefore,  $N - n \geq n - \nu$  by Theorem S. Note that if  $s^2 > \lambda_{\max}$ , then  $-\mathbf{PD}'\mathbf{P}$  has rank  $n$  by Lemma 2(2), and hence  $N - n \geq n$  by Theorem S.

(2) Now suppose that  $N \geq 2n$  and  $s \geq (\lambda_{\max})^{\frac{1}{2}}$ . Then since  $\lambda_{\max} \geq (d_{\max})^2$  by Lemma 1,  $\mathbf{D}' := s^2(\mathbf{J} - \mathbf{I}) - \mathbf{D}$  is a non-negative matrix, and since  $-\mathbf{PD}'\mathbf{P} = s^2\mathbf{P} - (-\mathbf{PDP})$  is positive semi-definite with rank  $\leq n$  by Lemma 2, there exist  $n+1$  points  $y_1, \dots, y_{n+1}$  in  $\mathbb{R}^n$  with the square-distance matrix

$$s^2(\mathbf{J} - \mathbf{I}) - \mathbf{D}.$$

If we denote the vertices of  $T$  by  $x_1, \dots, x_{n+1}$ , then the  $n+1$  points  $(x_1, y_1), \dots, (x_{n+1}, y_{n+1})$  in  $\mathbb{R}^n \times \mathbb{R}^n = \mathbb{R}^{2n}$  span a regular simplex of edge-length  $s$  which clearly projects onto  $T$ . Hence  $(s\Delta \rightarrow T)_{2n}$ , and hence  $(s\Delta \rightarrow T)_N$ .

(3) Finally, suppose that  $s > (\lambda_{\max})^{\frac{1}{2}}$  and  $(s\Delta \rightarrow T)_N$ . Then, as noted in the proof of (1), we must have  $N - n \geq n$ , i.e.  $N \geq 2n$ . Hence, if

$$2n - \nu' \leq N \leq 2n - 1 \quad \text{and} \quad (s\Delta \rightarrow T)_{2n-1}$$

then  $s = (\lambda_{\max})^{\frac{1}{2}}$ .  $\square$

The proof of Theorem 2 is almost identical, and is omitted.

## 4. CONCLUDING REMARK

Let  $S$  and  $T$  be two  $n$ -simplices. Let  $sT$  denote the enlargement of  $T$  with ratio  $s > 0$ . Then the extremal values of  $s$ , and the minimum value of  $N$  such that

$$(sT \rightarrow S)_N \quad \text{or} \quad (S \rightarrow sT)_N$$

can be presented in terms of the maximum root or the minimum positive root (and their multiplicities) of the equation

$$\det(s\mathbf{PD}_T\mathbf{P} - \mathbf{PD}_S\mathbf{P}) = 0,$$

where  $\mathbf{D}_T$  and  $\mathbf{d}_S$  are the square-distance matrices of  $T$  and  $S$ , respectively.

## REFERENCES

1. R. Bellman, *Introduction to Matrix Analysis*, McGraw-Hill, New York, 1970.
2. I. J. Schoenberg, Remarks to Maurice Frechet's article, *Ann. Math.*, **36** (1935), 724–732.

*Received 4 August 1991*

M. DEZA  
CNRS, Paris, France

H. MAEHARA  
Ryukyu University, Okinawa, Japan