

On Principal Angles between Subspaces in \mathbb{R}^n *

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Abstract

Let L, M be subspaces in \mathbb{R}^n , $\dim L = l \leq \dim M = m$. Then the **principal angles** between L and M ,

$$0 \leq \theta_1 \leq \theta_2 \leq \dots \leq \theta_l \leq \frac{\pi}{2}$$

are given by,

$$\cos \theta_i = \frac{\langle \mathbf{x}_i, \mathbf{y}_i \rangle}{\|\mathbf{x}_i\| \|\mathbf{y}_i\|} = \max \left\{ \frac{\langle \mathbf{x}, \mathbf{y} \rangle}{\|\mathbf{x}\| \|\mathbf{y}\|} : \begin{array}{l} \mathbf{x} \in L, \quad \mathbf{x} \perp \mathbf{x}_k, \\ \mathbf{y} \in M, \quad \mathbf{y} \perp \mathbf{y}_k, \end{array} k = 1, \dots, i-1 \right\}, \text{ see [1],}$$

where

$$(\mathbf{x}_i, \mathbf{y}_i) \in L \times M, i = 1, \dots, l,$$

are the corresponding pairs of **principal vectors**. We also define

$$\sin\{L, M\} := \prod_{i=1}^l \sin \theta_i, \quad \cos\{L, M\} := \prod_{i=1}^l \cos \theta_i.$$

We study relations between the principal angles and the **volume** of a matrix $A \in \mathbb{R}_r^{m \times n}$ defined by,

$$\text{vol } A := \sqrt{\sum \det^2 A_{IJ}}, \text{ see [2],}$$

summing over all $r \times r$ submatrices A_{IJ} of A . Sample results are the following generalizations of the Hadamard and Cauchy-Schwarz inequalities.

Theorem 4. Let $A = (A_1, A_2)$, $A_1 \in \mathbb{R}_l^{n \times n_1}$, $A_2 \in \mathbb{R}_m^{n \times n_2}$, $\text{rank } A = l + m$. Then

$$\text{vol } A = \text{vol } A_1 \text{ vol } A_2 \sin\{R(A_1), R(A_2)\}.$$

Theorem 5. Let $B, C \in \mathbb{R}_r^{n \times r}$. Then

$$|\det(B^T C)| = \text{vol } B \text{ vol } C \cos\{R(B), R(C)\}.$$

Key words: Principal angles. Singular values. Volume. Compound matrix.

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1 Introduction

Let L, M be subspaces in \mathbb{R}^n , and $\dim L = l \leq \dim M = m$. Then the **principal angles** between L and M ,

$$0 \leq \theta_1 \leq \theta_2 \leq \dots \leq \theta_l \leq \frac{\pi}{2} \quad (1.1)$$

are defined by

$$\begin{aligned} \cos \theta_i &:= \frac{\langle \mathbf{x}_i, \mathbf{y}_i \rangle}{\|\mathbf{x}_i\| \|\mathbf{y}_i\|} = \\ &= \max \left\{ \frac{\langle \mathbf{x}, \mathbf{y} \rangle}{\|\mathbf{x}\| \|\mathbf{y}\|} : \begin{array}{l} \mathbf{x} \in L, \\ \mathbf{y} \in M, \\ \mathbf{x} \perp \mathbf{x}_k, \\ \mathbf{y} \perp \mathbf{y}_k, \\ k = 1, \dots, i-1 \end{array} \right\}, \quad (1.2) \end{aligned}$$

where

$$(\mathbf{x}_i, \mathbf{y}_i) \in L \times M, \quad i = 1, \dots, l, \quad (1.3)$$

are the corresponding l pairs of **principal vectors**. Note that

$$\theta_1 = \dots = \theta_k = 0 < \theta_{k+1} \quad \text{iff} \quad \dim L \cap M = k, \quad (1.4)$$

and that if $\dim L = \dim M = 1$, θ_1 is the (nonobtuse) angle between the lines L and M .

We also denote the product of principal sines, and the product of principal cosines, by

$$\sin\{L, M\} := \sin \theta_1 \cdots \sin \theta_l, \quad (1.5)$$

$$\cos\{L, M\} := \cos \theta_1 \cdots \cos \theta_l, \quad (1.6)$$

Note that (1.5) and (1.6) are just notation, and not ordinary trigonometrical functions. In particular, $\sin^2\{L, M\} + \cos^2\{L, M\} \leq 1$.

Principal angles were introduced by Afriat in his study [1] of the geometry of subspaces in \mathbb{R}^n in terms of their orthogonal and oblique projectors. An important application of principal angles in Statistics is the canonical correlation theory of Hotelling [11], see also [5].

Principal angles and vectors generalize least squares solutions in the following sense: If $\dim L = 1$, say L is the line spanned by the vector \mathbf{a} , then the principal angle and vector between L and M are found by minimizing $\{\|\mathbf{a} - \mathbf{y}\|_2 : \mathbf{y} \in M\}$.

Björck and Golub [3] used the singular value decomposition to compute the principal angles as follows:

Lemma 1 Let the columns of $Q_L \in \mathbb{R}^{n \times l}$ and $Q_M \in \mathbb{R}^{n \times m}$ be orthonormal bases for L and M respectively, and let

$$\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_l \geq 0 \quad (1.7)$$

be the singular values of $Q_M^T Q_L$, then

$$\cos \theta_i = \sigma_i, \quad i = 1, \dots, l, \quad (1.8)$$

and

$$\sigma_1 = \dots = \sigma_k = 1 > \sigma_{k+1} \quad \text{iff} \quad \dim L \cap M = k. \quad \square \quad (1.9)$$

In this paper, we discuss some relations between principal angles and the matrix volume defined in [2]. In § 3, we express the principal cosines and principal sines in terms of the volume function. Principal angles allow us to “equalize” the Hadamard and Cauchy-Schwarz inequalities in § 4.

2 Preliminary results

Let $\mathbb{R}_r^{m \times n}$ be the set of $m \times n$ matrices of rank r . The **k -dimensional volume** of $A \in \mathbb{R}_r^{m \times n}$, $0 < k \leq r$, is defined as

$$\text{vol}_k A := \prod_{i=1}^k \sigma_i, \quad (2.1)$$

where

$$\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r > 0 \quad (2.2)$$

are the nonzero **singular values** of A . In particular, the **r -dimensional volume** of $A \in \mathbb{R}_r^{m \times n}$ is called its **volume**, and denoted by $\text{vol} A$, [2],

$$\text{vol} A := \begin{cases} 0 & \text{if } r = 0, \\ \prod_{i=1}^r \sigma_i & \text{if } r > 0, \end{cases} \quad (2.3)$$

or equivalently

$$\text{vol} A = \sqrt{\sum \det^2 A_{IJ}}, \quad (2.4)$$

summing over all $r \times r$ submatrices A_{IJ} of A . For consistency, set

$$\text{vol}_0 A := \min\{1, \text{rank} A\}. \quad (2.5)$$

In particular, if A has full column rank, then

$$\text{vol} A = \sqrt{\det(A^T A)}, \quad (2.6)$$

the volume of the parallelepiped spanned by the columns of A .

The volume function is closely related to compound matrices. The k^{th} -**compound matrix** of A , $C_k(A)$, is the $\binom{m}{k} \times \binom{n}{k}$ matrix whose elements are determinants of all $k \times k$ submatrices of A in lexicographic order. The singular values of $C_k(A)$ are all products $\sigma_{i_1} \sigma_{i_2} \cdots \sigma_{i_k}$ of singular values of A . Therefore the largest singular value of $C_k(A)$ (its **spectral norm**) is $\text{vol}_k A$. Some well known properties of compound matrices are collected below, (see e.g. [8]).

Proposition 1

- (a) If $A \in \mathbb{R}^{m \times p}$, $B \in \mathbb{R}^{p \times n}$, $\text{rank} AB = r$, then $C_k(AB) = C_k(A)C_k(B)$, $k \leq \min\{m, r, n\}$.
- (b) $C_k(A^T) = C_k(A)^T$.
- (c) $C_k(I) = I$, (appropriate size identity matrix).
- (d) If A has orthonormal columns, so does $C_k(A)$.
- (e) $\text{vol}_k A = \|C_k(A)\|_2$. \square

3 Principal angles and volume

The following lemma is used in the sequel.

Lemma 2 Let the columns of $Q_L \in \mathbb{R}^{n \times l}$ and $Q_M \in \mathbb{R}^{n \times m}$ be orthonormal bases for L and M respectively, and denote

$$Q := (Q_M, Q_L). \quad (3.1)$$

Let the singular values of $Q_M^T Q_L$ be

$$\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_l \geq 0. \quad (3.2)$$

Then the singular values of Q , in decreasing order, are

$$\sqrt{1 + \sigma_1}, \dots, \sqrt{1 + \sigma_l}, \underbrace{1, \dots, 1}_{m-l}, \sqrt{1 - \sigma_l}, \dots, \sqrt{1 - \sigma_1}. \quad (3.3)$$

Proof. Since

$$Q^T Q = \begin{pmatrix} I & Q_M^T Q_L \\ Q_L^T Q_M & I \end{pmatrix} = I + \begin{pmatrix} O & Q_M^T Q_L \\ Q_L^T Q_M & O \end{pmatrix},$$

the eigenvalues of $\begin{pmatrix} O & Q_M^T Q_L \\ Q_L^T Q_M & O \end{pmatrix}$ are (see [9, p. 418])

$$\pm \sigma_1, \pm \sigma_2, \dots, \pm \sigma_l, \underbrace{0, \dots, 0}_{m-l}.$$

Thus the eigenvalues of $Q^T Q$ are

$$1 \pm \sigma_1, \dots, 1 \pm \sigma_l, \underbrace{1, \dots, 1}_{m-l}. \quad \square$$

Theorem 1 Let the columns of $Q_L \in \mathbb{R}^{n \times l}$ and $Q_M \in \mathbb{R}^{n \times m}$ be orthonormal bases for L and M respectively, and denote

$$Q := (Q_M, Q_L). \quad (3.4)$$

Let

$$0 \leq \theta_1 \leq \theta_2 \leq \dots \leq \theta_l \leq \frac{\pi}{2} \quad (3.5)$$

be the principal angles between L and M . Then for $i = 0, 1, \dots, l - k - 1$,

$$\cos \theta_{l-i} = 1 - \frac{\text{vol}_{m+i+1}^2(Q)}{\text{vol}_{m+i}^2(Q)} = \frac{\text{vol}_{l-i}^2(Q)}{\text{vol}_{l-i-1}^2(Q)} - 1, \quad (3.6)$$

$$\sin \theta_{l-i} = \frac{\text{vol}_{m+l+1}(Q)}{\text{vol}_{l-i-1}(Q)} (\sin \theta_{l-i+1} \cdots \sin \theta_l)^{-1}, \quad (3.7)$$

where

$$k = \dim L \cap M. \quad (3.8)$$

Proof. By Lemmas 1 and 2,

$$\begin{aligned} \text{vol}_{m+i+1}(Q) &= \text{vol}_{m+i}(Q) \sqrt{1 - \cos \theta_{l-i}}, \quad i = 0, 1, \dots, l-1, \\ \text{vol}_{l-i}(Q) &= \text{vol}_{l-i-1}(Q) \sqrt{1 + \cos \theta_{l-i}}, \quad i = 0, 1, \dots, l-1, \\ \text{vol}_{m+i+1}(Q) &= \text{vol}_{l-i-1}(Q) \sin \theta_{l-i} \cdots \sin \theta_l, \quad i = 0, 1, \dots, l-1, \end{aligned}$$

which, after some arithmetic calculations, prove (3.6) and (3.7). \square

The following theorem gives the analogous results in terms of the orthogonal projectors P_L and P_M on L and M respectively.

Theorem 2 Let P_L and P_M be the orthogonal projectors on L and M respectively,

$$P := (P_M, P_L), \quad (3.9)$$

and let

$$0 \leq \theta_1 \leq \theta_2 \leq \dots \leq \theta_l \leq \frac{\pi}{2} \quad (3.10)$$

be the principal angles between L and M . Then for $i = 0, 1, \dots, l - k - 1$,

$$\cos \theta_{l-i} = 1 - \frac{\text{vol}_{m+i+1}^2(P)}{\text{vol}_{m+i}^2(P)} = \frac{\text{vol}_{l-i}^2(P)}{\text{vol}_{l-i-1}^2(P)} - 1, \quad (3.11)$$

$$\sin \theta_{l-i} = \frac{\text{vol}_{m+l+1}(P)}{\text{vol}_{l-i-1}(P)} (\sin \theta_{l-i+1} \cdots \sin \theta_l)^{-1}, \quad (3.12)$$

where

$$k = \dim L \cap M. \quad (3.13)$$

Proof. Let Q_L, Q_M be as in Lemma 2. Then

$$P_L = Q_L Q_L^T, \quad P_M = Q_M Q_M^T.$$

Therefore

$$P = (Q_M, Q_L) \begin{pmatrix} Q_M^T & O \\ O & Q_L^T \end{pmatrix} = Q U^T,$$

where

$$Q = (Q_M, Q_L),$$

and

$$U = \begin{pmatrix} Q_M & O \\ O & Q_L \end{pmatrix}.$$

Thus U has orthonormal columns, and

$$\begin{aligned} \text{vol}_i(P) &= \|C_i(Q)\|_2, \quad \text{by Proposition 1} \\ &= \text{vol}_i(Q), \quad i = 0, 1, \dots, m+l. \end{aligned}$$

The result follows from Theorem 1. \square

In applications to linear equations the subspaces are the null spaces of the coefficient matrices. Let

$$A_i \mathbf{x} = \mathbf{b}_i, \quad i = 1, 2, \quad (3.14)$$

be two linear systems, with nonempty solution sets

$$S_i = \{\mathbf{x} : A_i \mathbf{x} = \mathbf{b}_i\}, \quad i = 1, 2. \quad (3.15)$$

Then the principal angles between S_1 and S_2 are the principal angles between $L = N(A_1)$ and $M = N(A_2)$. The respective projectors are

$$P_L = I - A_1^\dagger A_1, \quad P_M = I - A_2^\dagger A_2. \quad (3.16)$$

The following theorem suggests that instead of using P_L and P_M , we can use

$$P_{L^\perp} = A_1^\dagger A_1, \quad P_{M^\perp} = A_2^\dagger A_2, \quad (3.17)$$

to compute the principal angles between L, M .

Theorem 3 The non-zero principal angles between L, M equal to the non-zero principal angles between L^\perp, M^\perp .

Proof. Let $(\mathbf{x}_i, \mathbf{y}_i)$ be a pair of principal vectors corresponding to the i th non-zero principal angle θ_i between L and M . Then $P_L \mathbf{y}_i = \alpha \mathbf{x}_i$, for some scalar α . Let the two vectors \mathbf{x}_i^\perp and \mathbf{y}_i^\perp be obtained from \mathbf{x}_i and \mathbf{y}_i respectively, by a $\pi/2$ -rotation in $\{\mathbf{x}_i, \mathbf{y}_i\}$ -plane. Then $\mathbf{x}_i^\perp \in L^\perp, \mathbf{y}_i^\perp \in M^\perp$, and the angle between \mathbf{x}_i^\perp and \mathbf{y}_i^\perp is also θ_i . Therefore the i th non-zero principal angle between L^\perp, M^\perp is $\leq \theta_i$. The result follows by interchanging L, M with L^\perp, M^\perp . \square

4 Hadamard and Cauchy-Schwarz equalities

Let $\mathbf{a}_1, \dots, \mathbf{a}_m$ be a basis for M , and let L be 1-dimensional, say $L = \text{span}\{\mathbf{a}\}$, where $\mathbf{a} = \mathbf{a}_M + \mathbf{a}_{M^\perp}$, $\mathbf{a}_M \in M$ and $\mathbf{a}_{M^\perp} \in M^\perp$. Then ([8])

$$\|\mathbf{a}_{M^\perp}\|_2 = \frac{\text{vol}_{m+1}(\mathbf{a}_1, \dots, \mathbf{a}_m, \mathbf{a})}{\text{vol}(\mathbf{a}_1, \dots, \mathbf{a}_m)}. \quad (4.1)$$

That is

$$\text{vol}_{m+1}(\mathbf{a}_1, \dots, \mathbf{a}_m, \mathbf{a}) = \text{vol}(\mathbf{a}_1, \dots, \mathbf{a}_m) \|\mathbf{a}\|_2 \sin \theta, \quad (4.2)$$

where θ is the principal angle between \mathbf{a}, M . For the general case, Afriat [1] gave the following ‘‘equalized’’ Hadamard inequality.

Lemma 3 Let $A = (A_1, A_2), A_1 \in \mathbb{R}_l^{n \times l}, A_2 \in \mathbb{R}_m^{n \times m}$. Then

$$\text{vol}_{l+m} A = \text{vol} A_1 \text{vol} A_2 \sin\{R(A_1), R(A_2)\}, \quad (4.3)$$

where $\sin\{R(A_1), R(A_2)\}$ is the product of principal sines between $R(A_1)$ and $R(A_2)$, see (1.5). \square

In Lemma 3 the matrices A_1, A_2 are of full column-rank. A further generalization of the Hadamard inequality is:

Theorem 4 Let $A = (A_1, A_2), A_1 \in \mathbb{R}_l^{n \times n_1}, A_2 \in \mathbb{R}_m^{n \times n_2}$, $\text{rank} A = l + m$. Then

$$\text{vol} A = \text{vol} A_1 \text{vol} A_2 \sin\{R(A_1), R(A_2)\}. \quad (4.4)$$

Proof.

$$\text{vol}^2 A = \sum_J \text{vol}^2 A_{*J},$$

where the summation is over all $n \times (l + m)$ submatrices of rank $l + m$. Since every $n \times (l + m)$ submatrix of rank $l + m$ has l columns A_{*J_1} from A_1 and m columns A_{*J_2} from A_2 , then

$$\begin{aligned} \text{vol}^2 A &= \sum_{J_1} \sum_{J_2} \text{vol}^2(A_{*J_1}, A_{*J_2}), \\ &= \sum_{J_1} \sum_{J_2} \text{vol}^2 A_{*J_1} \text{vol}^2 A_{*J_2} \sin^2\{R(A_1), R(A_2)\}, \\ &\quad \text{by Lemma 3,} \\ &= \text{vol}^2 A_1 \text{vol}^2 A_2 \sin^2\{R(A_1), R(A_2)\}. \quad \square \end{aligned}$$

For a square matrix A , the above results imply:

Corollary 1 Let $A = (A_1, A_2)$ be a square matrix, and $A_1 \in \mathbb{R}^{n \times l}, A_2 \in \mathbb{R}^{n \times m}$. Then

$$|\det(A)| = \text{vol}_l(A_1) \text{vol}_m(A_2) \sin\{R(A_1), R(A_2)\}. \quad \square \quad (4.5)$$

Now we ‘‘equalize’’ the Cauchy-Schwarz inequality.

Theorem 5 Let $B, C \in \mathbb{R}_r^{n \times r}$. Then

$$|\det(B^T C)| = \text{vol} B \text{vol} C \cos\{R(B), R(C)\}, \quad (4.6)$$

where $\cos\{R(B), R(C)\}$ is the product of principal cosines between $R(B)$ and $R(C)$, see (1.6).

Proof. Let Q_B and Q_C be orthonormal bases for $R(B)$ and $R(C)$, respectively, so that,

$$B = Q_B R_B, \quad C = Q_C R_C,$$

for some matrices $R_B, R_C \in \mathbb{R}_r^{r \times r}$. Then

$$\begin{aligned} |\det(B^T C)| &= |\det(R_B^T)| |\det(R_C)| |\det(Q_B^T Q_C)|, \\ &= \text{vol} B \text{vol} C \cos\{R(B), R(C)\}, \quad \text{by Lemma 1.} \end{aligned}$$

This completes the proof. \square

Denote the set of strictly increasing sequences of k elements from $1, \dots, n$, by

$$Q_{k,n} := \{I = (i_1, \dots, i_k) : 1 \leq i_1 < \dots < i_k \leq n\}.$$

For any sets $I \subset \{1, 2, \dots, m\}, J \subset \{1, 2, \dots, n\}$, let A_{I*}, A_{*J}, A_{IJ} denote the submatrices of A lying in rows

indexed by I , in columns indexed by J , and in their intersection, respectively. For $A \in \mathbb{R}_r^{m \times n}$, let

$$\begin{aligned} \mathcal{I}(A) &:= \{I \in Q_{r,m} : \text{rank } A_{I*} = r\}, \\ \mathcal{J}(A) &:= \{J \in Q_{r,n} : \text{rank } A_{*J} = r\}. \end{aligned}$$

For $I \subset \{1, 2, \dots, m\}$ define the subspace \mathbb{R}_I^m by

$$\mathbb{R}_I^m := \{x = (x_1, \dots, x_m) \in \mathbb{R}^m : x_i = 0 \text{ if } i \notin I\}.$$

An interesting interpretation of the determinants of maximal nonsingular submatrices is the following

Corollary 2 Let $A \in \mathbb{R}_r^{m \times n}$, $I \in Q_{r,m}$. Then

$$\cos\{R(A), \mathbb{R}_I^m\} = \frac{|\det A_{IJ}|}{\text{vol } A_{*J}} \quad (4.7)$$

for any $J \in \mathcal{J}(A)$.

Proof. Let $I = \{i_1, i_2, \dots, i_r\}$, $B = (e_{i_1}, \dots, e_{i_r})$, and for any $J \in \mathcal{J}(A)$ let $C = A_{*J}$. Then

$$R(B) = \mathbb{R}_I^m, \quad R(C) = R(A), \quad \text{and} \quad B^T C = A_{IJ}.$$

By Theorem 5

$$\cos\{R(A), \mathbb{R}_I^m\} = \cos\{R(B), R(C)\} = \frac{|\det A_{IJ}|}{\text{vol } A_{*J}}. \quad \square$$

Note that for any $I \in Q_{r,m}$, the ratio $\frac{|\det A_{IJ}|}{\text{vol } A_{*J}}$ is independent of the choice of $J \in \mathcal{J}(A)$.

Example 1 Let

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}, \quad (4.8)$$

$I = \{1, 2\}$, $J = \{1, 2\}$. Then A is of rank 2, and

$$|\det A_{IJ}| = 3, \quad \text{vol } A_{*J} = \sqrt{3^2 + 6^2 + 3^2} = 3\sqrt{6},$$

and, by (4.7),

$$\cos\{R(A), \mathbb{R}_{\{1,2\}}^3\} = \frac{1}{\sqrt{6}}.$$

Corollary 3 Let $L \subset \mathbb{R}^m$ be a subspace of dimension r . Then

$$\sum_{I \in Q_{r,m}} \cos^2\{L, \mathbb{R}_I^m\} = 1. \quad (4.9)$$

Proof. Follows from (4.7) since

$$\text{vol}^2 A_{*J} = \sum_{I \in \mathcal{I}(A)} \det^2 A_{IJ}. \quad \square$$

The corresponding result for the complementary orthogonal subspace L^\perp is

$$\sum_{I \in Q_{r,m}} \cos^2\{L^\perp, \mathbb{R}_{I^c}^m\} = 1, \quad (4.10)$$

where I^c is the **complement** of I in $\{1, 2, \dots, m\}$. The equivalence of (4.9) and (4.10) is by

$$\cos\{L, \mathbb{R}_I^m\} = \cos\{L^\perp, \mathbb{R}_{I^c}^m\}, \quad (4.11)$$

see Theorem 3.

Example 2 Let $L = R(A)$ for A of (4.8). Then $L^\perp = N(A^T)$ is the line spanned by $(1, -2, 1)$. The complement of $I = \{1, 2\}$ is $I^c = \{3\}$, and $\mathbb{R}_{I^c}^3$ is the x_3 -axis. The angle between the lines L^\perp and $\mathbb{R}_{I^c}^3$, $\angle(L^\perp, \mathbb{R}_{I^c}^3)$, is given by

$$\cos \angle(L^\perp, \mathbb{R}_{I^c}^3) = \cos \angle((1, -2, 1), (0, 0, 1)) = \frac{1}{\sqrt{1^2 + 2^2 + 1^2}} = \frac{1}{\sqrt{6}},$$

in agreement with (4.11) and Example 1.

Theorem 6 Let $B \in \mathbb{R}_r^{n \times r}$, $C \in \mathbb{R}^{n \times s}$ and $\text{rank}(C) \geq r$. Then

$$\cos\{R(B), R(C)\} = \cos \angle\{C_r(B), R(C_r(C))\}, \quad (4.12)$$

where $\angle\{C_r(B), R(C_r(C))\}$ is the angle between the vector $C_r(B)$ and the subspace $R(C_r(C))$.

Proof. Let the columns of

$$E = (e_1, \dots, e_r) \quad \text{and} \quad F = (f_1, \dots, f_t)$$

be orthonormal bases for $R(B)$ and $R(C)$ respectively, and let $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r \geq 0$ be the singular values of $F^T E$. Then by Lemma 1

$$\cos\{R(B), R(C)\} = \sigma_1 \cdots \sigma_r.$$

On the other hand, the columns of $C_r(E)$ and $C_r(F)$ form orthonormal bases for $C_r(B)$ and $R(C_r(C))$ respectively. Moreover, the singular value of $C_r(F)^T C_r(E)$ ($= C_r(F^T E)$) is

$$\sigma_1 \cdots \sigma_r.$$

By the same lemma,

$$\cos \angle\{C_r(B), R(C_r(C))\} = \sigma_1 \cdots \sigma_r. \quad \square$$

We now use the above results (the equalized Hadamard and Cauchy-Schwarz inequalities) to equalize a determinantal inequality of Thompson.

Theorem 7¹ Let $A \in \mathbb{R}^{kr \times kr}$ be partitioned as

$$A = (B_1, B_2, \dots, B_k), \quad \text{where } B_i \in \mathbb{R}_r^{kr \times r}.$$

Then

$$\begin{aligned} \det \begin{pmatrix} B_1^T B_1 & \cdots & B_1^T B_k \\ \vdots & \ddots & \vdots \\ B_k^T B_1 & \cdots & B_k^T B_k \end{pmatrix} &= \\ &= \det \begin{pmatrix} \det(B_1^T B_1) & \cdots & \det(B_1^T B_k) \\ \vdots & \ddots & \vdots \\ \det(B_k^T B_1) & \cdots & \det(B_k^T B_k) \end{pmatrix} \times \\ &\times \prod_{i=1}^{k-1} \frac{\sin^2 \{R(B_i), R(B_{i+1}, \dots, B_k)\}}{1 - \cos^2 \{C_r(B_i), R(C_r(B_{i+1}), \dots, C_r(B_k))\}}. \end{aligned} \quad (4.13)$$

Proof. We prove the case $k = 2$.

$$\begin{aligned} \det \begin{pmatrix} B_1^T B_1 & B_1^T B_2 \\ B_2^T B_1 & B_2^T B_2 \end{pmatrix} &= \det A^T A = \\ &= \text{vol}_r^2 B_1 \text{vol}_r^2 B_2 \sin^2 \{R(B_1), R(B_2)\}, \end{aligned}$$

by Corollary 1. Now let

$$\tilde{A} = (C_r(B_1), C_r(B_2)).$$

Then

$$\begin{aligned} \det \begin{pmatrix} \det(B_1^T B_1) & \det(B_1^T B_2) \\ \det(B_2^T B_1) & \det(B_2^T B_2) \end{pmatrix} &= \det(\tilde{A})^T \tilde{A} = \\ &= \|C_r(B_1)\|_2^2 \|C_r(B_2)\|_2^2 \sin^2 \{C_r(B_1), C_r(B_2)\} = \\ &= \text{vol}_r^2 B_1 \text{vol}_r^2 B_2 (1 - \cos^2 \{C_r(B_1), C_r(B_2)\}). \end{aligned}$$

The general case $k \geq 2$ is proved analogously. \square

Remark. If the denominator in (4.13) is zero, then the numerator is also zero. In this case, (4.13) is also valid if we let the ratio to be 1. Note that

$$R(C_r(B_{i+1}), \dots, C_r(B_k)) \subseteq R(C_r(B_{i+1}, \dots, B_k)),$$

and therefore

$$\begin{aligned} \cos \angle \{C_r(B_i), R(C_r(B_{i+1}), \dots, C_r(B_k))\} &\leq \\ &\leq \cos \angle \{C_r(B_i), R(C_r(B_{i+1}, \dots, B_k))\}, \\ &= \cos \angle \{R(B_i), R(B_{i+1}, \dots, B_k)\}. \end{aligned} \quad (4.14)$$

The last equality follows from Theorem 6. If $\theta_1, \dots, \theta_r$ are the principal angles between two subspaces L and M , then we always have

$$\frac{\sin^2 \{L, M\}}{1 - \cos^2 \{L, M\}} = \frac{\prod_{i=1}^r (1 - \cos^2 \theta_i)}{1 - \prod_{i=1}^r \cos^2 \theta_i} \leq 1. \quad (4.15)$$

Thus from (4.14) and (4.15), we have

$$\begin{aligned} \prod_{i=1}^{k-1} \frac{\sin^2 \{R(B_i), R(B_{i+1}, \dots, B_k)\}}{1 - \cos^2 \angle \{C_r(B_i), R(C_r(B_{i+1}), \dots, C_r(B_k))\}} &\leq \\ &\leq \prod_{i=1}^{k-1} \frac{\sin^2 \{R(B_i), R(B_{i+1}, \dots, B_k)\}}{1 - \cos^2 \{R(B_i), R(B_{i+1}, \dots, B_k)\}} \leq 1. \end{aligned} \quad (4.16)$$

Therefore Theorem 7 implies the following inequalities of Davis [6], Everitt [7] ($k = 2$) and Thompson [13] ($k \geq 2$).

Corollary 4 Let a p.s.d. matrix A of order kr be partitioned as

$$A = \begin{pmatrix} A_{11} & \cdots & A_{1k} \\ \vdots & \ddots & \vdots \\ A_{k1} & \cdots & A_{kk} \end{pmatrix}, \quad \text{where } A_{ij} \in \mathbb{R}^{r \times r}. \quad (4.17)$$

Then

$$\det \begin{pmatrix} A_{11} & \cdots & A_{1k} \\ \vdots & \ddots & \vdots \\ A_{k1} & \cdots & A_{kk} \end{pmatrix} \leq \det \begin{pmatrix} \det A_{11} & \cdots & \det A_{1k} \\ \vdots & \ddots & \vdots \\ \det A_{k1} & \cdots & \det A_{kk} \end{pmatrix}. \quad \square \quad (4.18)$$

A matrix $A \in \mathbb{R}_r^{n \times n}$ is called **range-Hermitian** (or **EP matrix**, **EP_r matrix**) if

$$R(A^T) = R(A). \quad (4.19)$$

Note that the class of range-Hermitian matrices includes normal matrices and nonsingular matrices. A good reference on range-Hermitian matrices is [12].

A characterization of range-Hermitian matrices follows.

Theorem 8 Let $A \in \mathbb{R}_r^{n \times n}$. Then A is range-Hermitian if and only if A has r nonzero eigenvalues $\lambda_1, \dots, \lambda_r$, and $\text{vol } A = \prod_{i=1}^r |\lambda_i|$.

Proof. Let $A = CB^T$ be a full-rank factorization of A . Since

$$R(A^T) = R(B), \quad R(A) = R(C),$$

$$R(A^T) = R(A) \quad \text{if and only if} \quad \cos \angle \{R(B), R(C)\} = 1.$$

By Theorem 6

$$\cos \angle \{C_r(B), C_r(C)\} = \frac{|C_r(B)^T C_r(C)|}{\|C_r(B)\|_2 \|C_r(C)\|_2} =$$

¹We proved this originally for $k = 2$, and we thank the referee for suggesting extension to $k > 2$.

$$= \cos\{R(B), R(C)\} = 1 .$$

And using the fact , [2],

$$\text{vol } A = \text{vol } B \text{ vol } C .$$

Then

$$\frac{|\sum \det A_{II}|}{\text{vol } A} = 1 ,$$

where the summation is over all $r \times r$ principal submatrices of A . And finally note that the r th elementary symmetric function of the eigenvalues of A is the sum of all the $r \times r$ principal minors of A . \square

Example 3 The matrix A of (4.8) is range-Hermitian. Its nonzero eigenvalues are

$$\lambda_1 = \frac{15 + 3\sqrt{33}}{2} , \quad \lambda_2 = \frac{15 - 3\sqrt{33}}{2} ,$$

and its nonzero singular values are, correct to 6 decimals,

$$\sigma_1 = 16.848103 , \quad \sigma_2 = 1.068369 .$$

Therefore $\text{vol } A = \sigma_1 \sigma_2 = |\lambda_1| |\lambda_2| = 18$.

To put Theorem 8 in perspective recall that for any square matrix $A \in \mathbb{R}^{n \times n}$ with singular values $\sigma_1 \geq \dots \geq \sigma_n \geq 0$, and eigenvalues $|\lambda_1| \geq \dots \geq |\lambda_n|$,

$$\prod_{i=1}^k |\lambda_i| \leq \prod_{i=1}^k \sigma_i , \quad k = 1, \dots, n, \quad (4.20)$$

with equality for $k = n$, [10, Theorem 3.3.2]. For $k = n$ the common value in (4.20) is nonzero iff A is nonsingular, in which case it is range-Hermitian. Therefore, the equality in (4.20) for $k = r = n$ is a special case of Theorem 8 which states that, for range-Hermitian matrices of rank r ,

$$\prod_{i=1}^r |\lambda_i| = \prod_{i=1}^r \sigma_i . \quad (4.21)$$

Conversely, Theorem 8 follows from the nonsingular case, since a matrix A is range-Hermitian iff

$$A = U \begin{pmatrix} B & O \\ O & O \end{pmatrix} U^T , \quad U \text{ orthogonal, } B \text{ nonsingular ,}$$

see [4, Theorem 4.3.1]. Written analogously to (4.20), a characterization of normal matrices is, [10, Problem 3.3.14]

$$\prod_{i=1}^k |\lambda_i| = \prod_{i=1}^k \sigma_i , \quad k = 1, \dots, n. \quad (4.22)$$

References

- [1] S.N. Afriat, "Orthogonal and oblique projectors and the characteristics of pairs of vector spaces", *Proc. Cambridge Phil. Soc.* **53** (1957), 800-816
- [2] A. Ben-Israel, "A volume associated with $m \times n$ matrices", *Lin. Algeb. Appl.* **167**(1992), 87-111
- [3] A. Björck and G.H. Golub, "Numerical methods for computing the angles between linear subspaces", *Math. Comp.* **27** (1973), 579-594
- [4] S.L. Campbell and C.D. Meyer, Jr., *Generalized Inverses of Linear Transformations*, Pitman, 1979
- [5] C. Cohen and A. Ben-Israel, "On the computation of canonical correlations", *Cahiers Centre Études Recherche Opér.* **11** (1969),121-132
- [6] C. Davis, "A device for studying Hausdorff moments", *Trans. Amer. Math. Soc.* **87** (1958), 144-158
- [7] W.N. Everitt, "A note on positive definite matrices", *Proc. Glasgow Math. Assoc.* **3** (1958), 173-175.
- [8] F.R. Gantmacher, *The Theory of Matrices*, Vol. I, Chelsea, New York, 1959
- [9] R.A. Horn and C.R. Johnson, *Matrix Analysis*, Cambridge University Press, Cambridge, 1985
- [10] R.A. Horn and C.R. Johnson, *Topics in Matrix Analysis*, Cambridge University Press, Cambridge, to appear.
- [11] H. Hotelling, "Relations between two sets of variates", *Biometrika* **28**(1935), 321-377
- [12] I.J. Katz and M.H. Pearl, "On EP_r and normal EP_r matrices", *J. Res. Nat. Bur. Standards Sect. B* **70B**(1966), 47-44.
- [13] R.C. Thompson, "A determinantal inequality for positive definite matrices", *Canad. Math. Bull.* **4** (1961), 57-62.