Inertia of Kraus matrices

Takashi Sano and Kazuki Takeuchi

Abstract. For positive real numbers r, p_0 , and $p_1 < \cdots < p_n$, let K_r be the $n \times n$ Kraus matrix whose (i, j) entry is equal to

$$\frac{1}{p_i - p_j} \Big(\frac{p_i^r - p_0^r}{p_i - p_0} - \frac{p_j^r - p_0^r}{p_j - p_0} \Big).$$

We determine the inertia of this matrix.

1. Introduction

In matrix analysis and operator theory, the notions of matrix monotone functions and matrix convex functions initiated by Löwner [19] and Kraus [18] are quite important. There have been several studies of these two classes of functions, see [1, 3, 8–11, 16] for instance. Let f be a real function defined on an interval I. The function f is said *matrix monotone of order n* if $A \leq B$ implies $f(A) \leq f(B)$ for all $n \times n$ Hermitian matrices A, B with eigenvalues in I; it is called *matrix convex of order n* if $f(tA + (1 - t)B) \leq tf(A) + (1 - t)f(B)$ for all $n \times n$ Hermitian matrices A, B with eigenvalues in I.

Let f be a C¹-function on I. For $\lambda_1, \ldots, \lambda_n \in I$, the $n \times n$ matrix

$$L_f(\lambda_1,\ldots,\lambda_n) := [[\lambda_i,\lambda_j]_f]$$

is called a *Loewner matrix associated with* f, where $[\lambda_i, \lambda_j]_f$ is the first divided difference of f; $[\lambda_i, \lambda_j]_f$ is defined as $\frac{f(\lambda_i) - f(\lambda_j)}{\lambda_i - \lambda_j}$ if $\lambda_i \neq \lambda_j$, and $f'(\lambda_i)$ if $\lambda_i = \lambda_j$.

Let f be a C^2 -function on I. For $\lambda_0, \lambda_1, \ldots, \lambda_n$ in I, the $n \times n$ matrix

$$K_f(\lambda_0; \lambda_1, \dots, \lambda_n) := [[\lambda_0, \lambda_i, \lambda_j]_f]$$
(1.1)

²⁰²⁰ Mathematics Subject Classification. Primary 15A39; Secondary 15A18. *Keywords.* Kraus matrix, inertia, power function.

is called a *Kraus matrix associated with* f, where $[\lambda_0, \lambda_i, \lambda_j]_f$ is the second divided difference of f; for distinct $\lambda_0, \lambda_i, \lambda_j$,

$$\begin{split} [\lambda_0, \lambda_i, \lambda_j]_f &:= \frac{[\lambda_i, \lambda_0]_f - [\lambda_j, \lambda_0]_f}{\lambda_i - \lambda_j} \\ &= \frac{1}{\lambda_i - \lambda_j} \Big(\frac{f(\lambda_i) - f(\lambda_0)}{\lambda_i - \lambda_0} - \frac{f(\lambda_j) - f(\lambda_0)}{\lambda_j - \lambda_0} \Big), \end{split}$$

and this can be extended continuously for any $\lambda_0, \lambda_i, \lambda_j \in I$. To be precise, if $\lambda_1, \ldots, \lambda_n$ are distinct and λ_0 is different from them, then the (i, i) entry $[\lambda_0, \lambda_i, \lambda_i]_f$ is

$$\frac{f'(\lambda_i)}{\lambda_i - \lambda_0} - \frac{f(\lambda_i) - f(\lambda_0)}{(\lambda_i - \lambda_0)^2}$$

If λ_0 coincides with some λ_j , then the (j, j) entry $[\lambda_j, \lambda_j, \lambda_j]_f$ is $f''(\lambda_j)/3!$. We refer to [4, 10, 16, 21] for divided differences.

It is known, thanks to Löwner [19], that for a C^1 -function f on I, f is matrix monotone of order n if and only if the Loewner matrix $L_f(\lambda_1, \ldots, \lambda_n)$ is positive semidefinite for any $\lambda_1, \ldots, \lambda_n \in I$ and, thanks to Kraus and Heinävaara [14, 18], that for a C^2 -function f on I, f is matrix convex of order n if and only if the Kraus matrix $K_f(\lambda_0; \lambda_1, \ldots, \lambda_n)$ is positive semidefinite for any $\lambda_0, \lambda_1, \ldots, \lambda_n \in I$.

Let *A* be an $n \times n$ Hermitian matrix. The inertia of *A* is the triple

$$In(A) := (\pi(A), \zeta(A), \nu(A))$$

where $\pi(A)$ is the number of positive eigenvalues of A, $\zeta(A)$ is the number of zero eigenvalues of A, and $\nu(A)$ is the number of negative eigenvalues of A.

In [5], Bhatia, Friedland, and Jain settled the conjecture about the inertia of Loewner matrices for the power functions t^r on $(0, \infty)$ which was proposed by Bhatia and Holbrook in [6]. In this article, we study the inertia of Kraus matrices for the power functions. We denote $K_{t^r}(p_0; p_1, \ldots, p_n)$ by $K_r(p_0; p_1, \ldots, p_n)$, and moreover simply by K_r when p_0, p_1, \ldots, p_n are easily inferred from the context:

$$K_r := \left[\frac{1}{p_i - p_j} \left(\frac{p_i^r - p_0^r}{p_i - p_0} - \frac{p_j^r - p_0^r}{p_j - p_0}\right)\right].$$
 (1.2)

Our main theorem is as follows:

Theorem 1.1. Let r, p_0 , and $p_1 < \cdots < p_n$, be positive real numbers. Let K_r be the $n \times n$ Kraus matrix defined in (1.2). Then

- (i) K_r is singular if and only if r = 1, ..., n;
- (ii) *if* r *is a positive integer and* $r \leq n + 1$ *, then*

$$r = 2k \implies \operatorname{In}(K_r) = (k, n+1-r, k-1),$$

and

$$r = 2k - 1 \implies \operatorname{In}(K_r) = (k - 1, n + 1 - r, k - 1)$$

for a positive integer k;

(iii) if r is not a positive integer and r < n, then

$$2(k-1) < r < 2k-1 \implies \ln(K_r) = (k-1,0,n+1-k),$$

and

$$2k - 1 < r < 2k \implies \ln(K_r) = (n + 1 - k, 0, k - 1)$$

for a positive integer k;

(iv) if r > n, then $In(K_r) = In(K_{n+1})$.

In Section 2, we give a proof of Theorem 1.1. We note that

$$\ln(K_{-r}) = \ln(K_{r+1})$$

for r > 0, since

$$K_{-r}(p_0; p_1, \dots, p_n) = q_0 D K_{r+1}(q_0; q_1, \dots, q_n) D$$

holds. Here

$$q_i := p_i^{-1}, \quad i = 0, \dots, n$$

and D is the $n \times n$ diagonal matrix diag (q_1, \ldots, q_n) ; hence, we just consider the case r > 0.

In the remainder of this section, we fix our notations and recall several notions. We refer the reader to [4, 17] for matrix analysis.

For an $n \times n$ Hermitian matrix A, all eigenvalues are real numbers and we denote them as

$$\lambda_1(A) \geq \lambda_2(A) \geq \cdots \geq \lambda_n(A)$$

An $n \times n$ Hermitian matrix A is said to be positive semidefinite or simply positive if

$$\langle Ax, x \rangle \ge 0$$
 for all $x \in \mathbb{C}^n$,

and positive definite or strictly positive if

$$\langle Ax, x \rangle > 0$$
 for all non-zero $x \in \mathbb{C}^n$.

For Hermitian matrices A and B, $A \ge B$ means that A - B is positive semidefinite. Let \mathcal{H}_1 be the subspace of \mathbb{C}^n defined as

$$\mathcal{H}_1 := \Big\{ x = (x_i) \in \mathbb{C}^n \colon \sum_{i=1}^n x_i = 0 \Big\},\$$

which is the kernel space of the $n \times n$ matrix E with all entries 1. An $n \times n$ Hermitian matrix A is said to be *conditionally positive definite* (*cpd* for short) or *almost positive* if

$$\langle Ax, x \rangle \ge 0$$
 for all $x \in \mathcal{H}_1$,

and *conditionally negative definite* (*cnd* for short) if -A is cpd. It is known that if A is cpd (resp. cnd), then $\lambda_{n-1}(A) \ge 0$ (resp. $\lambda_2(A) \le 0$). We refer the reader to [2, 10, 16, 21] for properties of these matrices. We also recall our study of the operator/matrix convexity by conditional negative/positive Loewner matrices in [7, 15].

2. Proof

In this section, we give a proof of Theorem 1.1. The following theorem and corollary are obtained similarly to those for Loewner matrices by Bhatia, Friedland, and Jain in [5]. All divided differences are associated with the power function t^r on $(0, \infty)$, so that we simply write them like $[p_0, p_i, p_j]$.

Let c_1, c_2, \ldots, c_n be real numbers, not all of which are zero. Let p_0 and $p_1 < \cdots < p_n$ be positive real numbers. Let us define the continuous function f on $(0, \infty)$ as

$$f(x) = \sum_{j=1}^{n} c_j[p_0, x, p_j] \quad \text{for } x \in (0, \infty).$$
(2.1)

Theorem 2.1. Let r be a positive real number not equal to 1, 2, ..., n. Then the function f defined in (2.1) has at most n - 1 zeros in $(0, \infty)$.

Proof. Let $r_1 < r_2 < \cdots < r_m$, and let a_1, a_2, \ldots, a_m be real numbers not all of which are zero. Then, the function

$$g(x) = \sum_{j=1}^{m} a_j x^{r_j}$$
(2.2)

has at most m - 1 zeros in $(0, \infty)$. This is a well known fact: for example, consult [20, p. 46]. For the function f, let

$$g(x) := f(x) \prod_{i=0}^{n} (x - p_i).$$

Then g can be expressed in the form of (2.2) with m = 2n + 1 and $\{r_1, ..., r_{2n+1}\} = \{0, 1, ..., n - 1, n, r, r + 1, ..., r + n - 1\}$. In fact, we have $g(x) = x^r h_1(x) - x^r h_1(x) + 1$.

 $xh_2(x) + h_3(x)$, where

$$h_1(x) := \sum_{j=1}^n c_j \prod_{i=1, i \neq j}^n (x - p_i),$$

$$h_2(x) := \sum_{j=1}^n c_j [p_j, p_0] \prod_{i=1, i \neq j}^n (x - p_i),$$

$$h_3(x) := \sum_{j=1}^n c_j (p_0[p_j, p_0] - p_0^r) \prod_{i=1, i \neq j}^n (x - p_i).$$

Note that

$$[p_0, x, p_j](x - p_j) = [x, p_0] - [p_j, p_0],$$

and

$$[p_0, x, p_j](x - p_j)(x - p_0) = x^r - p_0^r - [p_j, p_0]x + p_0[p_j, p_0].$$

These polynomials $h_1(x)$, $h_2(x)$, and $h_3(x)$ are of degree at most n - 1. Since $h_1(p_i) \neq 0$ for some *i* with $c_i \neq 0$, if $r \neq 1, 2, ..., n$, then *g* is not identically zero, and, by the fact mentioned above, the function *g* has at most 2n zeros in $(0, \infty)$. It is clear that n + 1 zeros occur at $x = p_0$, p_j $(1 \leq j \leq n)$, so *f* has at most n - 1 zeros in $(0, \infty)$, and the proof is complete.

Corollary 2.2. Let r be a positive real number different from 1, 2, ..., n. Then, the $n \times n$ Kraus matrix K_r defined in (1.2) is nonsingular.

Proof. If the matrix K_r were singular, then there would be a non-zero vector $c = (c_1, \ldots, c_n)$ such that $K_r c = 0$; that is,

$$\sum_{j=1}^{n} c_j [p_0, p_i, p_j] = 0$$

for i = 1, 2, ..., n. This means that the function f(x) in (2.1) would have *n* zeros: $x = p_1, ..., p_n$. But this contradicts Theorem 2.1.

Proposition 2.3. Let p_0 and $p_1 < p_2$ be in $(0, \infty)$. If r > 2, then the 2×2 Kraus matrix $K_r(p_0; p_1, p_2)$ has a positive eigenvalue and a negative eigenvalue.

Proof. Let $f(t) = t^r$. Since the function $(p_0, p_1, p_2) \mapsto \det K_r(p_0; p_1, p_2)$ is continuous and the matrix $K_r := K_r(p_0; p_1, p_2)$ is nonsingular by Corollary 2.2, either

det $K_r > 0$ for any p_0 , $p_1 < p_2$ or det $K_r < 0$ for any p_0 , $p_1 < p_2$. Suppose that det $K_r > 0$. Note that

$$\begin{vmatrix} [p_0, p_1, p_1] & [p_0, p_1, p_2] \\ [p_0, p_2, p_1] & [p_0, p_2, p_2] \end{vmatrix}$$

$$= \begin{vmatrix} [p_0, p_1, p_1] & [p_0, p_1, p_2] - [p_0, p_1, p_1] \\ [p_0, p_2, p_1] & [p_0, p_2, p_2] - [p_0, p_2, p_1] \end{vmatrix}$$

$$= \begin{vmatrix} [p_0, p_1, p_1] & (p_2 - p_1)[p_0, p_1, p_2, p_1] \\ [p_0, p_2, p_1] & (p_2 - p_1)[p_0, p_2, p_2, p_1] \end{vmatrix}$$

$$= (p_2 - p_1) \begin{vmatrix} [p_0, p_1, p_1] & [p_0, p_1, p_1, p_2] \\ [p_0, p_2, p_1] & [p_0, p_1, p_2, p_2] \end{vmatrix}$$

$$= (p_2 - p_1) \begin{vmatrix} [p_0, p_1, p_1] & [p_0, p_1, p_1, p_2] \\ [p_0, p_2, p_1] - [p_0, p_1, p_1] & [p_0, p_1, p_2, p_2] \end{vmatrix}$$

$$= (p_2 - p_1)^2 \begin{vmatrix} [p_0, p_1, p_1] & [p_0, p_1, p_1, p_2] \\ [p_0, p_1, p_2, p_1] & [p_0, p_1, p_1, p_2] \\ [p_0, p_1, p_2, p_1] & [p_0, p_1, p_1, p_2] \end{vmatrix} ;$$

that is,

det
$$K_r = (p_2 - p_1)^2 \begin{vmatrix} [p_0, p_1, p_1] & [p_0, p_1, p_1, p_2] \\ [p_0, p_1, p_1, p_2] & [p_0, p_1, p_1, p_2, p_2] \end{vmatrix}$$
.

We refer the reader to [10, 12, 21] for this computation. It follows from our assumption det $K_r > 0$ that

$$\begin{vmatrix} [p_0, p_1, p_1] & [p_0, p_1, p_1, p_2] \\ [p_0, p_1, p_1, p_2] & [p_0, p_1, p_1, p_2, p_2] \end{vmatrix} > 0$$

so that

$$\lim_{p_1 \to p_0, p_2 \to p_0} \begin{vmatrix} [p_0, p_1, p_1] & [p_0, p_1, p_1, p_2] \\ [p_0, p_1, p_1, p_2] & [p_0, p_1, p_1, p_2, p_2] \end{vmatrix} \\
= \begin{vmatrix} f^{(2)}(p_0)/2! & f^{(3)}(p_0)/3! \\ f^{(3)}(p_0)/3! & f^{(4)}(p_0)/4! \end{vmatrix} \ge 0.$$
(2.3)

It is known in [13, Proposition 3.1] for $f(t) = t^r$ on $(0, \infty)$ that

$$\begin{vmatrix} f^{(2)}(t)/2! & f^{(3)}(t)/3! \\ f^{(3)}(t)/3! & f^{(4)}(t)/4! \end{vmatrix} = -\frac{1}{144}r^2(r-1)^2(r-2)(r+1)t^{2r-6},$$

which is negative if r > 2. This contradicts (2.3); therefore, det $K_r < 0$ and we get the conclusion.

Corollary 2.4. Let p_0 and $p_1 < \cdots < p_n$ be in $(0, \infty)$. If r > 2, then the $n \times n$ Kraus matrix $K_r(p_0; p_1, \ldots, p_n) = K_r$ admits both a positive eigenvalue and a negative eigenvalue; that is,

$$\lambda_1(K_r) > 0 > \lambda_n(K_r).$$

Proof. Since

$$\lambda_1(K_r(p_0; p_1, p_2)) > 0 > \lambda_2(K_r(p_0; p_1, p_2))$$

by Proposition 2.3, using Cauchy's interlacing principle, we have the conclusion.

Proof of Theorem 1.1 (iv). If the inertia of K_r (r > 0) were to change, then one of the eigenvalues of K_r had to change sign, but this contradicts Corollary 2.2.

Proof of Theorem 1.1 (iii). For t > 0 and 0 < r < 1 the following formula is well known [4, p. 116]:

$$t^{r} = \frac{\sin r\pi}{\pi} \int_{0}^{\infty} \frac{t}{\lambda + t} \,\lambda^{r-1} \,d\lambda.$$

We write this as

$$t^{r} = \int_{0}^{\infty} \frac{t}{\lambda + t} \, d\mu(\lambda), \tag{2.4}$$

where μ is a positive measure on $(0, \infty)$. For each $\lambda > 0$ let

$$k_{\lambda}(t) = \frac{t}{\lambda + t}.$$

Since

$$\frac{1}{p_i-p_j}\Big(\frac{k_\lambda(p_i)-k_\lambda(p_0)}{p_i-p_0}-\frac{k_\lambda(p_j)-k_\lambda(p_0)}{p_j-p_0}\Big)=-\frac{\lambda}{(\lambda+p_0)(\lambda+p_i)(\lambda+p_j)},$$

the Kraus matrix of k_{λ} is expressed as

$$K_{k_{\lambda}}(p_0; p_1, \ldots, p_n) = -\frac{\lambda}{\lambda + p_0} D_{\lambda} E D_{\lambda},$$

where *E* is the matrix with all entries equal to 1: $E = [1] \ge 0$, and D_{λ} is the diagonal matrix diag $(\frac{1}{\lambda+p_1}, \ldots, \frac{1}{\lambda+p_n})$. It follows that $K_{k_{\lambda}} \le 0$; hence, $K_r \le 0$. To be precise, by Corollary 2.2 or a direct computation, K_r is negative definite, and

$$\ln(K_r) = (0, 0, n).$$

For 1 < r < 2, we get from (2.4) that

$$t^{r} = \int_{0}^{\infty} \frac{t^{2}}{\lambda + t} d\mu(\lambda).$$

For each $\lambda > 0$ let

$$h_{\lambda}(t) = \frac{t^2}{\lambda + t}.$$

Since

$$\frac{1}{p_i-p_j}\Big(\frac{h_\lambda(p_i)-h_\lambda(p_0)}{p_i-p_0}-\frac{h_\lambda(p_j)-h_\lambda(p_0)}{p_j-p_0}\Big)=\frac{\lambda^2}{(\lambda+p_0)(\lambda+p_i)(\lambda+p_j)},$$

the Kraus matrix of h_{λ} is of the form

$$K_{h_{\lambda}}(p_0; p_1, \ldots, p_n) = \frac{\lambda^2}{\lambda + p_0} D_{\lambda} E D_{\lambda} \ge 0,$$

so $K_r \ge 0$. Moreover, K_r is positive definite by Corollary 2.2, so that

$$In(K_r) = (n, 0, 0)$$

We pause the proof with a remark. Since $f(t) = t^r$ is operator convex for 1 < r < 2, that is, matrix convex of any order *n*, the corresponding Kraus matrix is known to be positive semidefinite for any p_0, p_1, \ldots, p_n ; see [18]. The above argument for K_r is already in [7], and for 2 < r < 3 or 3 < r < 4 the functions $g_{\lambda}(t) := \frac{t^3}{\lambda + t}$ and $f_{\lambda}(t) := \frac{t^4}{\lambda + t}$ work equally well. Actually, in terms of $D := \text{diag}(p_1, \ldots, p_n)$ and D_{λ} , we see that

$$K_{g_{\lambda}}(p_0; p_1, \ldots, p_n) = E - \frac{\lambda^3}{\lambda + p_0} D_{\lambda} E D_{\lambda},$$

and

$$K_{f_{\lambda}}(p_0; p_1, \dots, p_n) = DE + ED + p_0E - \lambda E + \frac{\lambda^4}{\lambda + p_0}D_{\lambda}ED_{\lambda};$$

thus, K_r is end for 2 < r < 3 and epd for 3 < r < 4, and we could determine its inertia by [2, Lemma 4.3.5] with Corollary 2.4.

To continue the proof in the general case, we take an alternative approach, following the argument by Bhatia, Friedland, and Jain as in the proof of [5, Theorem 1.1] for Loewner matrices, to determine the inertia of the Kraus matrix for the power function t^r .

Due to the identity

$$p_{i}\left(\frac{p_{i}^{r-2}-p_{0}^{r-2}}{p_{i}-p_{0}}-\frac{p_{j}^{r-2}-p_{0}^{r-2}}{p_{j}-p_{0}}\right)p_{j}$$

$$=-(p_{i}-p_{j})\left(p_{i}\frac{p_{i}^{r-2}-p_{0}^{r-2}}{p_{i}-p_{0}}+p_{j}\frac{p_{j}^{r-2}-p_{0}^{r-2}}{p_{j}-p_{0}}+p_{0}^{r-2}\right)$$

$$+\frac{p_{i}^{r}-p_{0}^{r}}{p_{i}-p_{0}}-\frac{p_{j}^{r}-p_{0}^{r}}{p_{j}-p_{0}},$$

the Kraus matrices K_r and K_{r-2} are related as

$$K_r = DK_{r-2}D + D^{\sim}DE + EDD^{\sim} + p_0^{r-2}E,$$

where $E = [1], D = \operatorname{diag}(p_1, \ldots, p_n)$ and

$$D^{\sim} := \operatorname{diag}\Big(\frac{p_1^{r-2} - p_0^{r-2}}{p_1 - p_0}, \dots, \frac{p_n^{r-2} - p_0^{r-2}}{p_n - p_0}\Big).$$

Suppose 2 < r < 3. Then K_r is end. In fact, for $x \in \mathcal{H}_1$ or Ex = 0,

$$\langle K_r x, x \rangle = \langle DK_{r-2}Dx, x \rangle + \langle D^{\sim}DEx, x \rangle + \langle x, D^{\sim}DEx \rangle + p_0^{r-2} \langle Ex, x \rangle$$

= $\langle DK_{r-2}Dx, x \rangle.$

We know that K_{r-2} is negative definite for 0 < r - 2 < 1, so $\langle K_r x, x \rangle \leq 0$ for $x \in \mathcal{H}_1$ or K_r is cnd; hence, $\lambda_2(K_r) \leq 0$. Since K_r is nonsingular by Corollary 2.2 and $\lambda_1(K_r) > 0$ by Corollary 2.4, we conclude that

$$In(K_r) = (1, 0, n - 1).$$

Especially, for n = 2 and r > 2

$$In(K_r) = (1, 0, 1) = In(K_3).$$

Let n > 2 and suppose 3 < r < 4. Since K_{r-2} is positive definite for 1 < r - 2 < 2,

$$\langle K_r x, x \rangle = \langle D K_{r-2} D x, x \rangle \ge 0$$

for $x \in \mathcal{H}_1$; hence, K_r is cpd, and $\lambda_{n-1}(K_r) \ge 0$. As K_r is nonsingular by Corollary 2.2 and $\lambda_n(K_r) < 0$ by Corollary 2.4, we have

$$In(K_r) = (n - 1, 0, 1).$$

In particular, for n = 3 and r > 3

$$In(K_r) = (2, 0, 1) = In(K_4).$$

Let us define the subspace \mathcal{H}_2 by

$$\mathcal{H}_{2} := \left\{ x = (x_{i}) \in \mathbb{C}^{n} : \sum_{i=1}^{n} x_{i} = 0 = \sum_{i=1}^{n} p_{i} x_{i} \right\}$$
$$= \{ x \in \mathbb{C}^{n} : Ex = 0 = EDx \},$$

where E = [1] and $D = \text{diag}(p_1, \ldots, p_n)$, being the orthogonal complement of the span of the vectors $(1, \ldots, 1)$ and (p_1, \ldots, p_n) .

Let n > 3 and suppose 4 < r < 5. For $x \in \mathcal{H}_2$,

$$\langle K_r x, x \rangle = \langle K_{r-2} y, y \rangle,$$

where $y := Dx \in \mathcal{H}_1$. Since K_{r-2} is end for 2 < r - 2 < 3, $\langle K_{r-2}y, y \rangle \leq 0$ or

 $\langle K_r x, x \rangle \leq 0$

for $x \in \mathcal{H}_2$. The minmax principle implies that $\lambda_3(K_r) \leq 0$. We already proved that a 3 × 3 principal submatrix of K_r has two positive eigenvalues, so $\lambda_2(K_r) > 0$ by Cauchy's interlacing principle. Since K_r is nonsingular by Corollary 2.2, one has

$$In(K_r) = (2, 0, n-2);$$

especially, for n = 4 and r > 4

$$In(K_r) = (2, 0, 2) = In(K_5).$$

Let n > 4 and suppose 5 < r < 6. For $x \in \mathcal{H}_2$

$$\langle K_r x, x \rangle = \langle K_{r-2} y, y \rangle,$$

where $y := Dx \in \mathcal{H}_1$. Since K_{r-2} is cpd for 3 < r-2 < 4, $\langle K_{r-2}y, y \rangle \ge 0$:

$$\langle K_r x, x \rangle \ge 0$$

for $x \in \mathcal{H}_2$. By the minmax principle and the nonsingularity of $K_r \lambda_{n-2}(K_r) > 0$, and since a 4 × 4 principal submatrix of K_r has two negative eigenvalues, $\lambda_{n-1}(K_r) < 0$ by Cauchy's interlacing principle. Hence, we have

$$In(K_r) = (n - 2, 0, 2),$$

so for n = 5 and r > 5

$$In(K_r) = (3, 0, 2) = In(K_6).$$

We define the subspace \mathcal{H}_3 by

$$\mathcal{H}_3 := \{ x = (x_i) \in \mathbb{C}^n \colon \sum_{i=1}^n x_i = \sum_{i=1}^n p_i x_i = \sum_{i=1}^n p_i^2 x_i = 0 \}$$
$$= \{ x \in \mathbb{C}^n \colon Ex = EDx = ED^2 x = 0 \},$$

which is the orthogonal complement of the span of the vectors $(1, ..., 1), (p_1, ..., p_n)$, and $(p_1^2, ..., p_n^2)$.

Let n > 5 and suppose 6 < r < 7. For $x \in \mathcal{H}_3$, since Ex = EDx = 0,

$$\langle K_r x, x \rangle = \langle D K_{r-2} D x, x \rangle = \langle D^2 K_{r-4} D^2 x, x \rangle = \langle K_{r-4} y, y \rangle,$$

where $y := D^2 x \in \mathcal{H}_1$. Since K_{r-4} is end for 2 < r - 4 < 3, $\langle K_{r-4}y, y \rangle \leq 0$, that is,

$$\langle K_r x, x \rangle \leq 0 \quad \text{for } x \in \mathcal{H}_3$$

The minmax principle with the nonsingularity of K_r implies that $\lambda_4(K_r) < 0$. Since a 5 × 5 principal submatrix of K_r has three positive eigenvalues, $\lambda_3(K_r) > 0$ by Cauchy's interlacing principle. We conclude that

$$In(K_r) = (3, 0, n - 3);$$

in particular, for n = 6 and r > 6

$$In(K_r) = (3, 0, 3) = In(K_7).$$

Let n > 6 and suppose 7 < r < 8. For $x \in \mathcal{H}_3$,

$$\langle K_r x, x \rangle = \langle K_{r-4} y, y \rangle,$$

where $y := D^2 x \in \mathcal{H}_1$. Since K_{r-4} is cpd for 3 < r - 4 < 4, $\langle K_{r-4}y, y \rangle \ge 0$:

$$\langle K_r x, x \rangle \geq 0$$

for $x \in \mathcal{H}_3$. By the minmax principle and the nonsingularity of K_r , $\lambda_{n-3}(K_r) > 0$, and a 6×6 principal submatrix of K_r has three negative eigenvalues so $\lambda_{n-2}(K_r) < 0$ by Cauchy's interlacing principle; hence,

$$In(K_r) = (n - 3, 0, 3);$$

particularly, for n = 7 and r > 7,

$$In(K_r) = (4, 0, 3) = In(K_8).$$

In this way, we have a proof by induction. Let n > 2k - 3. If 2(k - 1) < r < 2k - 1, then 2 < r - 2(k - 2) < 3. Let us define the subspace \mathcal{H}_{k-1} as the orthogonal complement of the span of the vectors $(1, \ldots, 1), (p_1, \ldots, p_n), \ldots, (p_1^{k-2}, \ldots, p_n^{k-2})$; that is, $\mathcal{H}_{k-1} := \ker E \cap \ker ED \cap \cdots \cap \ker ED^{k-2}$. For $x \in \mathcal{H}_{k-1}$,

$$\langle K_r x, x \rangle = \langle K_{r-2(k-2)} y, y \rangle,$$

where $y := D^{k-2}x \in \mathcal{H}_1$. Since $K_{r-2(k-2)}$ is end for 2 < r - 2(k-2) < 3,

$$\langle K_{r-2(k-2)}y, y \rangle \leq 0$$

or

$$\langle K_r x, x \rangle \leq 0 \quad \text{for } x \in \mathcal{H}_{k-1}.$$

The minmax principle with the nonsingularity of K_r implies that $\lambda_k(K_r) < 0$. Since a $(2k-3) \times (2k-3)$ principal submatrix of K_r has (k-1) positive eigenvalues by induction, $\lambda_{k-1}(K_r) > 0$ by Cauchy's interlacing principle; hence, we conclude that

$$In(K_r) = (k - 1, 0, n + 1 - k);$$

in particular, for n = 2(k - 1) and r > n,

$$In(K_r) = (k - 1, 0, k - 1) = In(K_{n+1}).$$

Similarly, let n > 2k - 2. If 2k - 1 < r < 2k, then 3 < r - 2(k - 2) < 4 and for $x \in \mathcal{H}_{k-1}$

$$\langle K_r x, x \rangle = \langle K_{r-2(k-2)} y, y \rangle,$$

where

$$y := D^{k-2}x \in \mathcal{H}_1.$$

Since $K_{r-2(k-2)}$ is cpd for 3 < r - 2(k-2) < 4, $\langle K_{r-2(k-2)}y, y \rangle \ge 0$:

 $\langle K_r x, x \rangle \geqq 0$

for $x \in \mathcal{H}_{k-1}$, and by the minmax principle, $\lambda_{n-(k-1)}(K_r) \geq 0$. Since a 2(k-1)-square principal submatrix of K_r has (k-1) negative eigenvalues by induction, $\lambda_{n-k+2}(K_r) < 0$ by Cauchy's interlacing principle. We conclude by the nonsingularity of K_r that

$$In(K_r) = (n + 1 - k, 0, k - 1),$$

so that for n = 2k - 1 and r > n

$$\ln(K_r) = (k, 0, k - 1) = \ln(K_{n+1}).$$

The proof of (iii) is complete.

Proof of Theorem 1.1 (i) and Theorem 1.1 (ii). For r = 1, the (i, j) entry of K_1 is

$$\frac{1}{p_i - p_j} \left(\frac{p_i - p_0}{p_i - p_0} - \frac{p_j - p_0}{p_j - p_0} \right) = 0,$$

that is, $K_1 = 0$ and $In(K_1) = (0, n, 0)$.

For r = 2,

$$\frac{1}{p_i - p_j} \left(\frac{p_i^2 - p_0^2}{p_i - p_0} - \frac{p_j^2 - p_0^2}{p_j - p_0} \right) = \frac{1}{p_i - p_j} \left((p_i + p_0) - (p_j + p_0) \right) = 1,$$

so that $K_2 = E = [1]$ and $In(K_2) = (1, n - 1, 0)$.

In general, for a positive integer $r \leq n + 1$, let $\mathcal{H}_r := \ker E \cap \ker ED \cap \cdots \cap \ker ED^{r-1}$. It is easy to see that K_r is of the form

$$K_r = \sum_{k=1}^{r-1} p_0^{k-1} \sum_{l=1}^{r-k} D^{r-k-l} E D^{l-1},$$

so that

 $\ker K_r \supset \ker E \cap \ker ED \cap \ker ED^2 \cap \cdots \cap \ker ED^{r-2} = \mathcal{H}_{r-1},$

and

$$\dim \ker K_r \ge \dim \mathcal{H}_{r-1} = n - (r-1) = n + 1 - r$$

For r = 3 and $n \ge 2$, dim ker $K_3 \ge n - 2$. By Corollary 2.4, $\lambda_1(K_3) > 0 > \lambda_n(K_3)$, so that $In(K_3) = (1, n - 2, 1)$ for $n \ge 2$.

For r = 4 and $n \ge 3$, dim ker $K_4 \ge n - 3$. Since a 3×3 principal submatrix of K_4 has the inertia (2, 0, 1) as in the proof of (iii), $\lambda_2(K_4) > 0 > \lambda_n(K_4)$ by Cauchy's interlacing principle; hence, $\ln(K_4) = (2, n - 3, 1)$ for $n \ge 3$.

For r = 5 and $n \ge 4$, dim ker $K_5 \ge n - 4$. A 4×4 principal submatrix of K_5 has the inertia (2, 0, 2) as in the proof of (iii), we conclude by Cauchy's interlacing principle that $\ln(K_5) = (2, n - 4, 2)$ for $n \ge 4$.

We can continue this argument; if r = 2k, then a (r - 1)-square principal submatrix of K_r has the inertia (k, 0, k - 1) as in the proof of (iii), so $In(K_r) = (k, n + 1 - r, k - 1)$ by Cauchy's interlacing principle; if r = 2k - 1, then a (r - 1)square principal submatrix of K_r has the inertia (k - 1, 0, k - 1) as in the proof of (iii), so $In(K_r) = (k - 1, n + 1 - r, k - 1)$ by Cauchy's interlacing principle.

If $r \in \{1, 2, ..., n\}$, then the number of zero eigenvalues is n + 1 - r, which is non-zero; so K_r is singular. The other implication follows from Corollary 2.2; therefore, the proof is complete.

Acknowledgement. The authors thank the referee for helpfull comments.

References

- T. Ando, *Topics on operator inequalities*. Hokkaido University, Sapporo, 1978 Zbl 0388.47024 MR 0482378
- [2] R. B. Bapat and T. E. S. Raghavan, *Nonnegative matrices and applications*. Encycl. Math. Appl. 64, Cambridge University Press, Cambridge, 1997 Zbl 0879.15015 MR 1449393
- [3] J. Bendat and S. Sherman, Monotone and convex operator functions. *Trans. Amer. Math. Soc.* **79** (1955), 58–71 Zbl 0064.36901 MR 82655

- [4] R. Bhatia, *Matrix analysis*. Grad. Texts Math. 169, Springer, New York, 1997 Zbl 0863.15001 MR 1477662.
- [5] R. Bhatia, S. Friedland, and T. Jain, Inertia of Loewner matrices. *Indiana Univ. Math. J.* 65 (2016), no. 4, 1251–1261 Zbl 1354.15005 MR 3549200
- [6] R. Bhatia and J. A. Holbrook, Fréchet derivatives of the power function. *Indiana Univ. Math. J.* 49 (2000), no. 3, 1155–1173 Zbl 0988.47011 MR 1803224
- [7] R. Bhatia and T. Sano, Loewner matrices and operator convexity. *Math. Ann.* 344 (2009), no. 3, 703–716 Zbl 1172.15010 MR 2501306
- [8] C. Davis, Notions generalizing convexity for functions defined on spaces of matrices. In Proc. Sympos. Pure Math., Vol. VII, pp. 187–201, Amer. Math. Soc., Providence, R.I., 1963 Zbl 0196.30303 MR 0155837
- [9] O. Dobsch, Matrixfunktionen beschränkter Schwankung. Math. Z. 43 (1938), no. 1, 353–388 Zbl 63.0848.02 MR 1545729
- [10] W. F. Donoghue, Jr., Monotone matrix functions and analytic continuation. Grundlehren Math. Wiss. 207, Springer, Berlin etc., 1974 Zbl 0278.30004 MR 0486556
- [11] F. Hansen and G. K. Pedersen, Jensen's inequality for operators and Löwner's theorem. *Math. Ann.* 258 (1982), no. 3, 229–241 Zbl 0473.47011 MR 1513286
- [12] F. Hansen and J. Tomiyama, Differential analysis of matrix convex functions. *Linear Algebra Appl.* 420 (2007), no. 1, 102–116 Zbl 1116.26006 MR 2277632
- F. Hansen and J. Tomiyama, Differential analysis of matrix convex functions. II. *JIPAM*. *J. Inequal. Pure Appl. Math.* **10** (2009), no. 2, article no. 32 Zbl 1167.26307 MR 2511925
- O. Heinävaara, Local characterizations for the matrix monotonicity and convexity of fixed order. *Proc. Amer. Math. Soc.* 146 (2018), no. 9, 3791–3799 Zbl 1408.26013 MR 3825834
- [15] F. Hiai and T. Sano, Loewner matrices of matrix convex and monotone functions. J. Math. Soc. Japan 64 (2012), no. 2, 343–364 Zbl 1261.15026 MR 2916071
- [16] R. A. Horn and C. R. Johnson, *Topics in matrix analysis*. Cambridge University Press, Cambridge, 1991 Zbl 0729.15001 MR 1091716
- [17] R. A. Horn and C. R. Johnson, *Matrix analysis*. Second edn., Cambridge University Press, Cambridge, 2013 Zbl 1267.15001 MR 2978290
- [18] F. Kraus, Über konvexe Matrixfunktionen. *Math. Z.* 41 (1936), no. 1, 18–42
 JFM 62.1079.02 Zbl 0013.39701 MR 1545602
- [19] K. Löwner, Über monotone Matrixfunktionen. Math. Z. 38 (1934), no. 1, 177–216
 Zbl 60.0055.01 MR 1545446
- [20] G. Pólya and G. Szegő, Problems and theorems in analysis. Vol. II. Theory of functions, zeros, polynomials, determinants, number theory, geometry. Springer Study Edition, Springer, Berlin etc., 1976 Zbl 0359.00003 MR 0465631
- [21] B. Simon, Loewner's theorem on monotone matrix functions. Grundlehren Math. Wiss. 354, Springer, Cham, 2019 Zbl 1428.26002 MR 3969971

Received 18 March 2022; revised 29 March 2022.

Takashi Sano

Faculty of Science, Yamagata University, Yamagata 990-8560, Japan; sano@sci.kj.yamagata-u.ac.jp

Kazuki Takeuchi

Graduate School of Science and Engineering, Yamagata University, Yamagata 990-8560, Japan; s211070m@st.yamagata-u.ac.jp