Part 5 Functions and Matrices

We study functions on matrices related to the Löwner (positive semidefinite) ordering on positive semidefinite matrices that $A \ge B$ means A - B is positive semi-definite. Note that every function $f : \mathbf{C} \to \mathbf{C}$ can be extended to diagonalizable matrices using the spectral decomposition

$$A = S^{-1} \operatorname{diag} (a_1, \dots, a_n) S \mapsto S^{-1} \operatorname{diag} (f(a_1), \dots, f(a_n)) S.$$

More generally, if A has minimal polynomial $(z - a_1)^{r_1} \cdots (z - a_k)^{r_k}$, and f has derivatives at a neighborhoods of a_j up to order $r_j - 1$, then f(A) can be defined, and has the same value as p(A) for a polynomial interpolating the values and derivatives of $f(a_j)$ up to order $r_j - 1$ for $j = 1, \ldots, k$.

1 Positive and completely positive linear maps

Definition 1.1 A function $\Phi : M_n \to M_m$ is a positive linear map if $\Phi(A) \ge \Phi(B)$ whenever $A \ge B$. It is unital if $\Phi(I) = I$. It k-positive if

$$\Phi \otimes I_k(A_{ij})_{1 \le i,j \le k}) = (\Phi(A_{ij}))_{1 \le i,j \le k}$$

is positive whenever $(A_{ij}) \in M_k(M_n)$ is positive. If Φ is k-positive for all k = 1, 2, ..., then Φ is completely positive.

Lemma 1.2 Let $A, B \in H_n$.

- (a) Then $A \leq B$ if and only if $X^*AX \leq X^*BX$ for any nonzero $n \times m$ matrix X.
- (b) Suppose $0 \le A, B$ and B is invertible. Then $A \le B$ is and only if $||A^{1/2}B^{-1/2}|| \le 1$.

Lemma 1.3 A matrix $C = \begin{pmatrix} A & X \\ X^* & B \end{pmatrix} \in H_n$ is positive semi-definite if and only if $B - X^*A^{-1}X$ is positive semi-definite.

Lemma 1.4 Suppose Φ is a positive linear map. If $A, B, C \in H_n$ are such that $\begin{pmatrix} A & C \\ C & B \end{pmatrix} \ge 0$, then $\begin{pmatrix} \Phi(A) & \Phi(C) \\ \Phi(C) & \Phi(B) \end{pmatrix} \ge 0$.

Corollary 1.5 Suppose Φ is a positive linear map. If $A \ge 0$, then

$$\Phi(A^2) \ge \phi(A)^2 \quad and \quad \Phi(A^{-1}) \ge \Phi(A)^{-1}.$$

Remark 1.6 For a fixed $n \times m$ matrix X such that $X^*X = I_m$, the map defined by $\Phi(A) = X^*AX$ is unital and completely positive.

Suppose $P_1, \ldots, P_k \in M_n$ are orthogonal projections such that $P_1 + \cdots + P_k = I$. The map $A \mapsto P_1AP_1 + \cdots + P_kAP_k$ is unital and completely positive.

Theorem 1.7 Let $\Phi : M_n \to M_m$. The following are equivalent.

- (a) Φ is completely positive.
- (b) $(\Phi(E_{ij}))_{1 \le i,j \le n}$ is positive semidefinite.
- (c) There are $n \times m$ matrices X_1, \ldots, X_k such that

$$\Phi(A) = \sum_{j=1}^{k} X_j^* A X_j.$$

2 Operator monotone and operator convex functions

Definition 2.1 Let $f : \mathbf{R} \to \mathbf{R}$.

(a) The function f is matrix monotone of order \mathbf{n} if for all $A, B \in H_n$ and all $t \in [0, 1]$,

$$f(A) \le f(B)$$
 whenever $A \le B$.

If this is true for all orders, then f is operator monotone.

(b) The function f is matrix convex of order **n** if for all $A, B \in H_n$ and all $t \in [0, 1]$,

$$f((1-t)A + tB) \le (1-t)f(A) + tf(B).$$

If this is true for all orders, then f is operator monotone.

If the condition on f holds for t = 1/2, then f is mid-point matrix/operator convex. For continuous functions, mid-point convex implies matrix/operator convex.

(c) The function f is matrix concave of order n if -f is matrix convex of order n.

Example 2.2 Here are some basic examples.

- 1. The function f(t) = c + dt is operator monotone and operator convex if $d \ge 0$ and $c \in \mathbf{R}$. It is operator convex for all $c, d \in \mathbf{R}$.
- 2. The function $f(t) = t^2$ is operator convex, but not operator monotone.
- 3. The function $f(t) = t^3$ is not operator convex, and not operator monotone.
- 4. The function f(t) = |t| is not operator convex.

Theorem 2.3 (a) $f(t) = t^{-1}$ is operator monotone on $(0, \infty)$.

(b) If $r \in [0,1]$, then $f(t) = t^r$ is operator monotone on $[0,\infty)$.

Theorem 2.4 If f is operator monotone on $[0, \infty)$, then there are $a, b \in \mathbf{R}$ with $b \ge 0$ and a positive measure μ on $[0, \infty)$ such that

$$f(t) = a + bt + \int_0^\infty \frac{st}{s+t} d\mu(s).$$

If f is operator convex on $[0, \infty)$, then there are $a, b, c \in \mathbf{R}$ with $c \ge 0$ and a positive measure μ on $[0, \infty)$ such that

$$f(t) = a + bt + ct^2 + \int_0^\infty \frac{st^2}{s+t} d\mu(s).$$

Theorem 2.5 Let $A, B \in H_n$ be positive semidefinite, and $\|\cdot\|$ be a UI norm. If f(t) is a nonnegative operator monotone function f(t) on $[0, \infty)$, then

$$||f(A+B)|| \le ||f(A) + f(B)||.$$

Furthermore, if f^{-1} exists such that f(0) = 0 and $f(\infty) = \infty$, then

$$||f^{-1}(A+B)|| \ge ||f^{-1}(A) + f^{-1}(B)||.$$

Corollary 2.6 Let $\|\cdot\|$ be a UI norm on M_n . If $A, B \in H_n$ are positive semidefinite and $r \in (0, 1]$ then

$$||(A+B)^r|| \le ||A^r+B^r||$$
 $||(A+B)^{1/r}|| \ge ||A^{1/r}+B^{1/r}||,$

and

$$\|\log(A+B+I)\| \le \|\log(A+I) + \log(B+I)\|$$
 and $\|e^A + e^B\| \le \|e^{A+B} + I\|$

It is known that a function g(t) on $[0, \infty)$ with g(0) = 0 is operator convex if and only if g(t)/t is operator monotone on $(0, \infty)$. We have the following corollary.

Corollary 2.7 Let g(t) be a nonnegative operator convex function on $[0, \infty)$ with g(0) = 0. If $\|\cdot\|$ is a UI norm on M_n , then

$$||g(A+B)|| \ge ||g(A) + g(B)||$$

for any positive semidefinite $A, B \in H_n$.

Theorem 2.8 Let f(t) be a nonnegative operator monotone function on $[0, \infty)$ and $\|\cdot\|$ be a UI norm on M_n . Let $A \in M_n$.

(a) If $\|\text{diag}(1, 0, \dots, 0)\| = 1$, then

 $f(||A||) \le f(|A|)||.$

(b) If ||I|| = 1, then

 $f(||A||) \ge f(|A|)||.$

3 Extremal representation techniques

Proposition 3.1 Let $A \in H_n$ be positive definite, and $B \in M_n$. Then

$$B^*A^{-1}B = \min\left\{C: \begin{pmatrix} A & B\\ B^* & C \end{pmatrix} \ge 0\right\}.$$

In particular,

$$A^{-1} = \min \left\{ C : \begin{pmatrix} A & I \\ I & C \end{pmatrix} \ge 0 \right\}.$$

Corollary 3.2 If $A \ge B$ then $A^{-1} \le B^{-1}$.

Proposition 3.3 Note that

$$A^{1/2} = \max\left\{C: \begin{pmatrix} A & C\\ C & I \end{pmatrix} \ge 0\right\}.$$

Thus, $A \ge B$ implies that $A^{1/2} \ge B^{1/2}$.

Proposition 3.4 Suppose $A = (A_{ij})_{1 \le i,j \le 2} \in H_n$ is positive definite. Then the Schurcomplement of A with respect to A_{11} given by $S(A) = A_{22} - A_{21}^* A_{11}^{-1} A_{12}$ satisfies

$$S(A) = \max\{C : A \ge 0_k \oplus C \text{ with } C \in H_{n-k}\}$$

and

$$S(A) = \min\{[Z|I_{n-k}]A[Z|I_{n-k}]^* : Z \text{ is } (n-k) \times k\}.$$

Corollary 3.5 Suppose f is operator monotone on $[0, \infty)$ and $f(0) \ge 0$. Then for any positive linear map from M_n to M_{n-k} , and any positive definite matrix A,

 $f(\Phi(A)) \ge \Phi(f(A)) \ge S(f(A)) \ge f(S(A)).$

In particular,

$$[\Phi(A^p)]^{1/p} \ge \Phi(A) \ge S(A) \ge [S(A^p)]^{1/p} \quad for \ p \ge 1,$$

and

$$[\Phi(A^p)]^{1/p} \le \Phi(A) \le S(A) \le [S(A^p)]^{1/p} \quad for \ p \le -1.$$

Corollary 3.6 Suppose $A \in H_n$ is positive definite and $B \in M_n$. Then for any $1 \le i_1 < \cdots < i_m \le k$ and $1 \le j_1 < \cdots < j_m \le n$. If $i_m + j_m \le m + k$, then

$$\prod_{s=1}^{m} \lambda_{i_s+j_s-s}(S(BAB^*)) \le \prod_{s=1}^{m} \lambda_{i_s}(S(BB^*))\lambda_{j_s}(A)$$

4 Current research

Theorem 4.1 Suppose $f : [0, \infty) \to [0, \infty)$ is concave. If $A, B \in H_n$ are positive semidefinite, then

$$||f(A+B)|| \le ||f(A) + f(B)||$$

for any UI norm.

Problem 4.2 Suppose $f : [0, \infty) \to [0, \infty)$ is concave. If $A, B \in M_n$ are normal. Is it true that

$$||f(|A+B|)|| \le ||f(|A|) + f(|B|)||$$

5 Exercises

1. Suppose Φ is a positive linear map. Show that

$$\|\Phi\| = \max\{\|\Phi(A)\| : s_1(A) \le 1\} = \|\Phi(I)\|.$$

2. Show that $\|\Phi\|$ is attained at a unitary, and show that

$$\begin{pmatrix} \Phi(I) & \Phi(U) \\ \Phi(U)^* & \Phi(I) \end{pmatrix} \ge 0$$

to conclude that $\|\Phi(U)\| \le \|\Phi(I)\|$ for any unitary U.

- 3. Suppose Φ is a positive linear map such that $\Phi(I) \leq I$. Show that $\|\Phi(A)\| \leq \|A\|$ for any UI norm $\|\cdot\|$.
- 4. Show that the scalar function f(x) = |x| on M_1 is 2-positive but not 3-positive.
- 5. Show that the scalar function f(x) = |x| is not operator convex.