

A CANONICAL TRACE CLASS APPROXIMANT

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ABSTRACT. Let H be a finite-dimensional Hilbert space, $B(H)$ the space of bounded linear operators on H , and C a convex subset of $B(H)$. If A is a fixed operator in $B(H)$, then A has a unique best approximant from C in the C_p norm for $1 < p < \infty$. However, in the C_1 (trace) norm, A may have many best approximants from C . In this paper, it is shown that the best C_p approximants to A converge to a select trace class approximant A_1 as $p \rightarrow 1$. Furthermore, A_1 is the unique trace class approximant minimizing $\sum_{i=1}^n S_i(A - B) \ln S_i(A - B)$ over all trace class approximants B . The numbers $S_i(T)$ are the eigenvalues of the positive part $|T|$ of T .

1. Introduction. The question of approximating a finite-dimensional matrix by operators from some prescribed convex set in the C_p norm, $1 < p < \infty$, has a simple answer. Namely, there is always a unique approximant due to the convexity of the class of approximating operators and the uniform rotundity of the C_p norm. However, in general a given matrix has a “large” set of approximants in the C_p norm for $p = 1$ or ∞ .

The purpose of this paper is to show that the unique C_p approximants to a given finite-dimensional matrix A from a prescribed convex set converge to a “select” C_1 approximant as p tends to 1. This result answers, in the affirmative, questions 3 and 4 in [8]. In addition, the fact that the limit exists as $p \rightarrow 1$ establishes a canonical trace class approximant. Very often canonical approximants shed much light on the structure of the set of best approximants as seen in [2, 3 and 4].

2. A canonical trace class approximant. In what follows, the term operator shall refer to a bounded linear operator on a complex Hilbert space H of dimension $n < \infty$. The numbers $S_1(A), \dots, S_n(A)$ will denote the eigenvalues of $|A|$ listed in decreasing order, where A is a bounded linear operator on H having polar factorization $A = U|A|$. For $1 \leq p < \infty$, the C_p norm is defined by

$$\|A\|_p = \left(\sum_{i=1}^n S_i(A)^p \right)^{1/p}.$$

For a given matrix A , let A'_p denote a unique nearest point in the C_p norm from a closed convex set C and let $A_p = A - A'_p$. We establish that $\lim_{p \rightarrow 1} A_p$ (and hence $\lim_{p \rightarrow 1} A'_p$) exists in the trace norm.

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Motivated by the approach in [7] we let, for a fixed A ,

$$\phi(A, p) = \|A\|_p^p = \sum_{i=1}^n S_i(A)^p.$$

Then

$$\frac{d\phi}{dp}(A, p) = \sum_{i=1}^n S_i(A)^p \ln S_i(A),$$

and so

$$\frac{d\phi}{dp}(A, 1) = \sum_{i=1}^n S_i(A) \ln S_i(A),$$

where \ln denotes the natural log.

LEMMA 2.1. *If B is any cluster point of $\{A_p\}$ in the trace norm as $p \rightarrow 1$, then $A - B$ is a best C_1 approximant to A from C , and*

$$\sum_{i=1}^n S_i(B) \ln S_i(B) = \min_{D \in \mathcal{D}} \sum_{i=1}^n S_i(D) \ln S_i(D),$$

where $\mathcal{D} = \{A - D' : D' \text{ is a best } C_1 \text{ approximant to } A \text{ from } C\}$.

PROOF. Let $p_k \rightarrow 1$ with $A_{p_k} \rightarrow B$ in the trace norm. Clearly $A - B$ is a best C_1 approximant. Let $\Phi(x) = x \ln x$ for $x > 0$ with $\Phi(0) = 0$. Let $D_1 \in \mathcal{D}$ satisfy

$$\sum_{i=1}^n \Phi(S_i(D_1)) = \min_{D \in \mathcal{D}} \sum_{i=1}^n \Phi(S_i(D)).$$

It is easily checked that $\Phi(x) \leq (x^p - x)/(p - 1)$ for $x \geq 0$, $p > 1$, and so, for each $i = 1, \dots, n$, we have

$$\Phi(S_i(A_{p_k})) \leq \frac{S_i(A_{p_k})^{p_k} - S_i(A_{p_k})}{p_k - 1}.$$

Hence

$$\begin{aligned} \sum_{i=1}^n \Phi(S_i(A_{p_k})) &\leq \frac{1}{p_k - 1} \left[\sum_{i=1}^n S_i(A_{p_k})^{p_k} - \sum_{i=1}^n S_i(A_{p_k}) \right] \\ &\leq \frac{1}{p_k - 1} \left[\sum_{i=1}^n S_i(D_1)^{p_k} - \sum_{i=1}^n S_i(D_1) \right] \\ &= \frac{1}{p_k - 1} [\phi(D_1, p_k) - \phi(D_1, 1)]. \end{aligned}$$

Letting $k \rightarrow \infty$, we obtain

$$\sum_{i=1}^n \Phi(S_i(B)) \leq \sum_{i=1}^n \Phi(S_i(D_1)).$$

Hence

$$\sum_{i=1}^n \Phi(S_i(B)) = \min_{D \in \mathcal{D}} \sum_{i=1}^n \Phi(S_i(D)).$$

THEOREM 2.2 *Let C be a closed convex set of $n \times n$ matrices and let A_p be the matrix in C with smallest C_p norm. Then $\lim_{p \rightarrow 1} A_p$ exists.*

PROOF. Since the underlying Hilbert space is finite dimensional, every sequence $\{A_{p_k}\}$, $p_k \rightarrow 1$, contains a convergent subsequence in the trace norm. Hence it suffices to show that $\{A_p\}$ has a unique cluster point as $p \rightarrow 1$.

Suppose B_1 and B_2 are cluster points of $\{A_p\}$ as $p \rightarrow 1$. If we multiply B_1 and B_2 by a sufficiently large positive scaling factor, we may assume that all the numbers

$$S_i(B_1/2), \quad S_i(B_2/2), \quad S_i(\tfrac{1}{2}(B_1 + B_2)),$$

$i = 1, \dots, n$, are either 0 or greater than 2.

Let $f(x)$ be a strictly increasing convex function such that $f(0) = 0$ and $f(x) = \Phi(x)$ for $x \geq 2$. By [5, Chapter 2, Theorem 4.1], we have

$$\begin{aligned} \sum_{i=1}^n \Phi\left(S_i\left(\frac{1}{2}(B_1 + B_2)\right)\right) &= \sum_{i=1}^n f\left(S_i\left(\frac{1}{2}(B_1 + B_2)\right)\right) \\ (1) \quad &\leq \sum_{i=1}^n f\left(S_i\left(\frac{B_1}{2}\right) + S_i\left(\frac{B_2}{2}\right)\right) \\ (2) \quad &\leq \sum_{i=1}^n \left[\frac{1}{2}f(S_i(B_1)) + \frac{1}{2}f(S_i(B_2))\right] \\ &= \frac{1}{2} \sum_{i=1}^n \Phi(S_i(B_1)) + \frac{1}{2} \sum_{i=1}^n \Phi(S_i(B_2)). \end{aligned}$$

Hence by Lemma 2.1, since B_1 and B_2 are the cluster points of $\{A_p\}$ as $p \rightarrow 1$, we have

$$\sum_{i=1}^n \Phi\left(S_i\left(\frac{1}{2}(B_1 + B_2)\right)\right) \leq \min_{D \in \mathcal{D}} \sum_{i=1}^n \Phi(S_i(D)).$$

Since $\frac{1}{2}(B_1 + B_2) \in \mathcal{D}$, it follows that

$$\sum_{i=1}^n \Phi\left(S_i\left(\frac{1}{2}(B_1 + B_2)\right)\right) = \min_{D \in \mathcal{D}} \sum_{i=1}^n \Phi(S_i(D)).$$

Hence both inequality signs in (1) and (2) must actually be equal signs.

The inequality (1) implies that

$$S_i(\tfrac{1}{2}(B_1 + B_2)) = S_i(B_1/2) + S_i(B_2/2)$$

for each $i = 1, \dots, n$ since f is strictly convex, while (2) implies that $S_i(B_1) = S_i(B_2) = S_i((B_1 + B_2)/2)$ for all i ; i.e., $\|B_1\|_p = \|B_2\|_p = \|(B_1 + B_2)/2\|_p$. From the strict convexity of the C_p norm, it follows that $B_1 = B_2$. Hence there exists a unique cluster point of $\{A_p\}$ as $p \rightarrow 1$ and this completes the proof.

REMARK. A similar result seems likely in the case $p = \infty$ if the approximating class is the set of positive operators. We conjecture that the best C_p approximants converge in the operator norm to the approximant P_m as $p \rightarrow \infty$. See [8].

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