

**$C_p$ -minimal positive approximants**

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**§ 1. Introduction**

In [8], P. R. HALMOS initiated the study of positive operator approximation. Among other things he established the proximality of the convex set of positive operators on Hilbert space by producing a canonical best positive approximant. This approximant, hereafter referred to as the Halmos approximant was later shown by R. H. BOULDIN [2] to be maximal, in the sense of order, among all positive approximants to a given operator.

This paper originated in the attempt to find a canonical minimal approximant since canonical approximants shed much light on the structure of the set of best approximants [3], [4], [5]. As will be shown in 4, there need not be a positive approximant minimal in the sense of order. Nevertheless, we construct a positive approximant  $P_m$  that is minimal in a sense given by the following theorem, in which  $\|\cdot\|_p$  denotes the usual  $C_p$  norm on finite matrices.

*Theorem 1.1. Each operator  $A=B+iC$  on a finite dimensional complex Hilbert space  $\mathfrak{H}$  has a positive approximant  $P_m$  such that  $A-P_m$  is a normal operator and such that for each positive operator  $Q \neq P_m$  it follows that  $\|A-Q\|_p > \|A-P_m\|_p$  for all finite  $p$  sufficiently large. This operator  $P_m$  will be referred to as the  $C_p$ -minimal positive approximant of  $A$ .*

In section 2, relevant background information is given along with needed notation. Section 3 contains the proof of the main theorem, the heart of which involves an inductive construction. There are many open questions related to our result, and these questions along with some examples comprise section 4.

## § 2. Preliminaries

The term operator shall mean a bounded linear operator on a complex Hilbert space, and the operator norm of an operator  $X$  is denoted by  $\|X\| = \sup \{\|Xf\| : f \in \mathfrak{H}, \|f\| = 1\}$ . If  $\mathfrak{M}$  is a set of operators, then an operator  $Y_0 \in \mathfrak{M}$  is an  $\mathfrak{M}$ -approximant of  $X$  if  $\|X - Y_0\| = \inf \{\|X - Y\| : Y \in \mathfrak{M}\}$ ; approximants using other norms are defined similarly. We shall follow Halmos's convention of using "positive operator" as synonymous with "nonnegative operator" and "approximant" in place of "best approximant". For the reader's convenience we restate the following results proved by Halmos in [8].

**Theorem 2.1.** *If  $B+iC$  is the usual Cartesian representation for the operator  $A$ , then*

$$\inf \{A - P : P \geq 0\} = \inf \{r : r \geq \|C\|, B + (r^2 - C^2)^{1/2} \geq 0\}.$$

The first infimum shall henceforth be denoted  $\delta(A)$ .

**Theorem 2.2.** *If  $B+iC$  is the usual Cartesian representation for the operator  $A$  and if  $P_H = B + ((\delta(A))^2 - C^2)^{1/2}$ , then  $P_H$  is a positive approximant of  $A$ .*

The operator  $P_H$  is the Halmos approximant referred to in the introduction.

**Theorem 2.3.** *Any operator  $A$  has a representation of the form  $P + U\delta(A)$  where  $P \geq 0$  and  $U$  is unitary with negative real part. If  $A$  is not a positive operator, then the above representation is unique.*

In another direction, the notion of a *strict approximant* was introduced by J. R. RICE [10] in the course of his investigations into  $l_\infty$  approximation as a method of selecting one approximant among many. A full discussion of strict approximants would lead us too far astray but, roughly speaking, to find a strict approximant one minimizes as much as one can. The following example will serve to illustrate.

**Example.** Consider the vector  $v \equiv (2i, i, 0)$  viewed as an element of  $l_\infty(3)$ . The distance of  $v$  to the set of positive functions is 2, and there are clearly an infinite number of positive approximants. The vector  $(0, 0, 0)$ , however, is the unique strict approximant since 0 is the nearest nonnegative number to  $2i$ ,  $i$  and 0.

It was later shown by B. MITIAGIN [9] and J. DESCLOUX [6] that the strict approximants have an additional approximation property.

**Theorem 2.4.** *Let  $l_p(n)$  denote  $n$ -dimensional complex Cartesian space endowed with the  $l_p$  norm and  $M$  a subspace of  $l_p(n)$ . If  $x \in l_p(n) \setminus M$ , let  $y_p$  denote an approximant from  $M$ . Then  $y = \lim_{p \rightarrow \infty} y_p$  exists, and  $y$  is the strict approximant of  $x$  in  $l_\infty(n)$ .*

The construction in the next section is modelled after the construction of the strict approximant, although the fact that the space of  $n \times n$  matrices is not a commutative algebra introduces some new twists into the construction.

§ 3. The Main Result

In this section the proof of Theorem 1.1 is given. The first lemma is stated in more generality than is needed, but it seems of interest in its own right.

Lemma 3.1. *Let  $\mathfrak{C}$  be a norm-closed convex set of compact operators on a uniformly convex Banach space  $\mathfrak{B} \neq 0$ . Define  $d = \inf \{\|X\| : X \in \mathfrak{C}\}$  and*

$$\mathfrak{D} = \{X \in \mathfrak{C} : \|X\| = d\}.$$

*If  $\mathfrak{D}$  is separable, then there exist unit vectors  $y, z \in \mathfrak{B}$  such that for every  $X \in \mathfrak{D}$  it follows that  $Xy = dz$ . In particular, if  $D$  is in  $\mathfrak{D}$ , then*

$$\bigcap_{X \in \mathfrak{D}} \ker(X - D) \neq \{0\}.$$

Proof. Let  $\{X_1, X_2, \dots\}$  be dense in  $\mathfrak{D}$ ; define operators  $Y_n \in \mathfrak{D}$  to be the corresponding Cesaro means, i.e.  $Y_n = (X_1 + \dots + X_n)/n$ . Because each  $Y_n$  is a compact operator, there exists a unit vector  $y_n \in \mathfrak{B}$  such that  $\|Y_n y_n\| = d$ ; define the unit vector  $z_n = Y_n y_n / d$ . Since  $\mathfrak{B}$  is reflexive, the sequences  $\{y_n\}$  and  $\{z_n\}$  have weak cluster points in the unit ball of  $\mathfrak{B}$ . Thus it is possible to find vectors  $y, z$  and subsequences  $\{y_{n,j}\}$  and  $\{z_{n,j}\}$  that converge weakly to  $y$  and to  $z$ . Fix  $k \geq 1$ . Because  $X_k$  is compact it follows that  $X_k(y_{n,j})$  converges to  $X_k(y)$  in norm, as  $j \rightarrow \infty$ . But  $X_k(y_{n,j}) = dz_{n,j}$  for all  $j$  sufficiently large, by the definition of  $Y_{n,j}$  and the fact that  $\mathfrak{B}$  is uniformly convex. Thus  $dz_{n,j}$  converges to  $X_k(y)$  in norm. Hence  $\|X_k(y)\| = d$ , which implies  $\|y\| = 1$  since  $\|X_k\| = d$ . Also,  $X_k y = dz$  since  $dz_{n,j}$  converges weakly to  $dz$ ; thus  $\|z\| = 1$ . Since  $\{X_k\}$  is dense in  $\mathfrak{D}$ , it follows that  $Xy = dz$  for each  $X \in \mathfrak{D}$ .

The next lemma is crucial in what follows. It is a slight generalization of a lemma appearing in [2].

Lemma 3.2. *If  $X = X^*, Y = Y^*, P = P^*$ , and  $d = \|X + iY - P\|$ , then  $P \leq X + \sqrt{d^2 I - Y^2}$ .*

Proof. As in [1], [8] it follows that  $(P - X)^2 + Y^2 \leq d^2 I$ . Because the square root function is order-preserving, it follows that  $P - X \leq \sqrt{(P - X)^2} \leq \sqrt{d^2 I - Y^2}$ .

Proof of Theorem 1.1. We proceed with constructing the operator  $P_m$  by defining numbers  $\{\delta_k\}$  and subspaces  $\{M_k\}$  that reduce  $C$ . If  $C(k)$  denotes the part of  $C$  on  $M_k$  and  $I(k)$  denotes the orthogonal projection from  $H$  onto  $M_k$ , then  $P_m = B + \varepsilon \sqrt{\delta_k^2 I(k) - C^2(k)}$ . The construction of the sequences  $\{\delta_k\}$  and  $\{M_k\}$  is by induction.

Define  $\delta_1 = \delta(A)$  (recall the definition immediately following Theorem 2.1) and  $M_1 = \bigcap \ker(B + \sqrt{\delta_1^2 - C^2} - Q)$  where this intersection is taken over all positive

approximants  $Q$ . Lemma 3.1 can be applied to the convex sets  $\mathfrak{C}_1 = \{A - P : P \geq 0\}$  and  $\mathfrak{D}_1 = \{A - Q : Q \text{ is a positive approximant of } A\}$  using  $d = \delta_1$  and  $D = A - (B + \sqrt{\delta_1^2 - C^2})$  to show  $M_1 \neq \{0\}$ .

The fact that  $M_1$  reduces  $C$  is shown in [1, proof of Lemma 4.1]; a different proof is given here. Let  $f$  be a unit vector in  $M_1$  and let  $Q$  be a positive approximant of  $A$ . Then  $(B - Q)f = -\sqrt{\delta_1^2 - C^2}f$  by the definition of  $M_1$ ; thus both  $A - Q$  and  $(A - Q)^*$  attain their norm at  $f$ . Hence  $|A - Q|^2 f = |(A - Q)^*|^2 f$ , and this implies that  $(B - Q)Cf = C(B - Q)f$ . Thus  $(B - Q)Cf = C(B - Q)f = C(-\sqrt{\delta_1^2 - C^2})f = -\sqrt{\delta_1^2 - C^2}(Cf)$ . Hence  $(B + \sqrt{\delta_1^2 - C^2} - Q)(Cf) = 0$ , so that  $Cf \in M_1$ .

Thus  $M_1$  reduces  $C$ , and it also reduces  $A - Q$  for each approximant  $Q$ . Clearly  $A$  has a unique approximant if and only if  $M_1 = H$ . Define the subspace  $H_1 = H$  and the projection  $E_1 = I$ .

Let  $H_1, E_1, \mathfrak{C}_1, \delta_1, \mathfrak{D}_1, M_1$  be as defined above. Define  $H_2 = H \ominus M_1$  with orthogonal projection  $E_2 : H \rightarrow H_2$ . Put  $\mathfrak{C}_2 = \{(A - Q)E_2 : Q \geq 0 \text{ and } (A - Q)E_1 \in \mathfrak{D}_1\}$ ; this set  $\mathfrak{C}_2$  is convex because  $\mathfrak{D}_1$  is convex. Define  $\delta_2 = \min \{\|X\| : X \in \mathfrak{C}_2\}$  and  $\mathfrak{D}_2 = \{X \in \mathfrak{C}_2 : \|X\| = \delta_2\}$ ; this set  $\mathfrak{D}_2$  is convex because  $\mathfrak{C}_2$  is convex.

The construction of  $M_2$  is as follows. For an arbitrary operator  $X$  on  $H$  let  $X_2 = E_2 X E_2$ ; clearly  $M_1 \subseteq \ker X_2$  and  $M_1$  reduces  $X_2$ . Choose  $Q \geq 0$  such that  $(A - Q)E_2 \in \mathfrak{D}_2$ . Then  $0 \leq Q_2 \leq B_2 + \sqrt{\delta_2^2 E_2 - C_2^2}$  because  $M_1$  reduces  $A - Q$ ; this inequality follows from Lemma 3.2 with  $X = B_2, Y = C_2, P = Q_2$  and  $d = \delta_2$ . Notice that for each such  $Q$  it follows that  $Q|M_1 = (B + \sqrt{\delta_1^2 I(1) - C(1)^2})|M_1$  by the definition of  $M_1$ . Hence the operator  $Z = B + \sqrt{\delta_1^2 I(1) - C(1)^2} + \sqrt{\delta_2^2 E_2 - C_2^2}$  satisfies  $Z \geq Q \geq 0$  for each such  $Q$ . Thus the operator  $D_2 = iC_2 - \sqrt{\delta_2^2 E_2 - C_2^2} \in \mathfrak{D}_2$  because the operator  $Z = B + \sqrt{\delta_1^2 I(1) - C(1)^2} + \sqrt{\delta_2^2 E_2 - C_2^2}$  is a positive operator such that  $(A - Z)E_2 \in \mathfrak{C}_2$ . Define  $M_2 = \bigcap_X \ker(X - D_2) \cap H_2$  where the intersection is over all  $X \in \mathfrak{D}_2$ . From Lemma 3.1 with  $\mathfrak{C} = \mathfrak{C}_2, d = \delta_2, \mathfrak{D} = \mathfrak{D}_2$  and  $D = D_2$ , considered as operators from  $H_2$  to itself, it follows that  $M_2 \neq \{0\}$  if  $H_2 \neq \{0\}$ . If  $H_2 = \{0\}$ , then  $M_2 = \{0\}$ .

The fact that  $M_2$  reduces the operator  $C_2 = C|H_2$  is shown by a proof similar to that used for  $M_1$ . Let  $f$  be a unit vector in  $M_2$  and let  $Q \geq 0$  be such that  $(A - Q)E_2 \in \mathfrak{D}_2$ . Then  $H_2$  reduces  $(A - Q)E_2$  and  $(B - Q)f = -\sqrt{\delta_2^2 E_2 - C_2^2}f$  by the definition of  $M_2$ ; thus both  $(A - Q)E_2$  and  $(A - Q)^*E_2$  attain their norm at  $f$ . Hence  $|A_2 - Q_2|^2 f = |(A_2 - Q_2)^*|^2 f = \delta_2^2 f$ ; this implies  $(B_2 + \sqrt{\delta_2^2 E_2 - C_2^2} - Q)(C_2 f) = 0$  as before. In other words,  $C_2 f \in M_2$ , and thus  $M_2$  reduces  $C_2$ .

In general, once  $H_k, E_k, \mathfrak{C}_k, \delta_k, \mathfrak{D}_k, M_k$  have been defined, put  $H_{k+1} = H \ominus (M_1 \oplus \dots \oplus M_k)$  with orthogonal projection  $E_{k+1} : H \rightarrow H_{k+1}$ . Let  $\mathfrak{C}_{k+1} = \{(A - Q)E_{k+1} : Q \geq 0 \text{ and } (A - Q)E_k \in \mathfrak{D}_k\}$ ; this set  $\mathfrak{C}_{k+1}$  is convex because  $\mathfrak{D}_k$  is convex. Define  $\delta_{k+1} = \min \{\|X\| : X \in \mathfrak{C}_{k+1}\}$  and  $\mathfrak{D}_{k+1} = \{X \in \mathfrak{C}_{k+1} : \|X\| = \delta_{k+1}\}$ ; this set  $\mathfrak{D}_{k+1}$  is convex because  $\mathfrak{C}_{k+1}$  is convex.

To define  $M_{k+1}$ , write  $X_{k+1} = E_{k+1} X E_{k+1}$  for each  $X$ ; clearly  $M_j \subset \ker X_{k+1}$  for  $1 \leq j \leq k$ . The operator  $D_{k+1} = iC_{k+1} - \sqrt{\delta_{k+1}^2 E_{k+1} - C_{k+1}^2}$  is in  $\mathfrak{D}_{k+1}$  because the operator  $Z = B + \sqrt{\delta_1^2 I(1) - C(1)^2} + \dots + \sqrt{\delta_k^2 I(k) - C(k)^2} + \sqrt{\delta_{k+1}^2 E_{k+1} - C_{k+1}^2}$  is a positive operator such that  $(A - Z)E_{k+1}$  is in  $\mathfrak{C}_{k+1}$ . Define the subspace  $M_{k+1}$  by  $M_{k+1} = \bigcap_X \ker (X - D_{k+1}) \cap H_{k+1}$ ; this intersection is taken over all  $X \in \mathfrak{D}_{k+1}$ .

Lemma 3.1 shows that  $M_{k+1} \neq \{0\}$  if  $H_{k+1} \neq \{0\}$ , and the operator  $D_{k+1}$  can be used to show  $M_{k+1}$  reduces  $C_{k+1}$ . This completes the inductive definition.

Thus for each integer  $k$  it is possible to define  $M_k$  and  $\delta_k$ . Because  $H$  is finite-dimensional, the subspaces  $H_{k+1}$  will be  $\{0\}$  for all  $k$  sufficiently large. Thus it is possible to define the positive operator  $P_m$  by

$$P_m = B + \Sigma \sqrt{\delta_k^2 I(k) - C(k)^2}.$$

Clearly  $A - P_m$  is a normal operator.

It remains to establish the minimality of  $P_m$ . If  $Q$  is a positive operator different from  $P_m$ , then there exists a least integer  $k \geq 1$  such that  $(A - Q)E_k \notin \mathfrak{D}_k$ . If  $k = 1$ , then  $\|A - Q\| > \delta_1$ . Hence if  $h$  denotes the dimension of  $H$ , then for all  $p$  sufficiently large it follows that  $\|A - Q\|_p^p \geq \|A - Q\|^p > h\delta_1^p \geq \|A - P_m\|_p^p$ . If  $k > 1$ , then let  $\Delta_k = \|(A - Q)E_k\|$ . Then  $\Delta_k > \delta_k$  because  $(A - Q)E_j$  is in  $\mathfrak{D}_j$  for each  $j \leq k - 1$ . For each  $j \leq k - 1$  the subspace  $M_j$  reduces  $A - Q$ , and the part of  $A - Q$  on  $M_j$  is equal to the part of  $A - P_m$  on  $M_j$ , which is  $iC(j) - \sqrt{\delta_j^2 I(j) - C(j)^2}$  and is  $\delta_j$  times a unitary operator. Thus for all  $p$  sufficiently large and  $m_j = \text{dimension of } M_j$ , it follows that

$$\begin{aligned} \|A - Q\|_p^p &\geq m_1 \delta_1^p + \dots + m_{k-1} \delta_{k-1}^p + \Delta_k^p > m_1 \delta_1^p + \dots + m_{k-1} \delta_{k-1}^p + \\ &+ (h - m_1 - \dots - m_{k-1}) \delta_k^p \geq \|A - P_m\|_p^p. \end{aligned}$$

This proves Theorem 1.1. ]

### § 4. Examples and Open Questions

Example 4.1. There does not always exist a positive approximant that is minimal in the sense of order.

Let  $A$  be the self-adjoint  $3 \times 3$  matrix given by  $A = \begin{pmatrix} -1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 2 \end{pmatrix}$ . It is easily seen that  $\delta(A) = 1$ , and that no positive approximant is smaller than  $P_0 = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ . For if there were such an approximant  $P_1$ , then  $P_0 - P_1 \geq 0$  and  $P_1$

necessarily would have the form  $\begin{pmatrix} \overline{0} & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & \alpha \end{pmatrix}$ ,  $\alpha < 1$ . But then  $P_1$  would no longer be an approximant of  $A$ , so  $P_0$  is the only candidate to be minimal. On the other hand, it is easily checked that  $P_2 = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1/8 & 1/2 \\ 0 & 1/2 & 5/2 \end{pmatrix}$  is an approximant of  $A$  and clearly  $P_2 - P_0 \neq 0$ .

For a given matrix  $A = B + iC$  let  $\|A\|_p$  denote the  $C_p$  norm of  $A$ . It is well known (and follows easily from [7, p. 94]) that  $B$  is a self-adjoint approximant of  $A$  in the  $C_p$  norm for all  $p$ , and it is unique in case  $1 < p < \infty$ . Thus if  $S_p$  denotes the self-adjoint  $C_p$  approximant to  $A$ , then  $S_p = B$  so  $\lim_{p \rightarrow \infty} \|S_p - B\| = 0$ . Let  $R_p$  denote a positive approximant to  $A$  in the  $C_p$  norm which again is unique if  $1 < p < \infty$ .

Q1. For a given matrix  $A$  and corresponding  $C_p$  minimal positive approximant  $P_m$ , does  $\lim_{p \rightarrow \infty} \|R_p - P_m\| = 0$ ?

A weaker question is:

Q2. For a given  $A$ , does the corresponding net  $\{R_p\}$  have a limit in the uniform norm as  $p \rightarrow \infty$ ?

Note that the  $C_p$ -minimal positive approximant  $P_m$  seems to be the operator analogue of the strict approximant mentioned in section 2. Since the strict approximant of Rice is a limit of  $l_p$  approximants by Theorem 2.4, the answer to Q1 could likewise be yes. Moreover Q1 and Q2 both have affirmative answers in the case  $A$  is a  $2 \times 2$  matrix. This follows from the fact that for a given  $2 \times 2$  matrix  $A$  and any positive approximant  $P$ ,  $A - P$  is normal; each convergent subnet of  $\{R_p\}$  must converge to a uniform positive approximant, which can in this case be shown to be  $P_m$  by using the minimality condition defining  $P_m$ . To establish that  $A - P$  is normal, note that one of two cases occurs:

- i)  $P_H$  is the unique approximant so that  $A - P_H$  is a multiple of a unitary by Theorem 2.3.
- ii) The subspace  $M_1$  mentioned in the proof of Theorem 1.1 is 1-dimensional. In this case for any approximant  $P$  the errors  $A - P$  and  $A - P_H$  can differ only in the  $(2, 2)$  entry (when viewed as matrices with respect to the subspaces  $M_1$  and  $M_1^\perp$ ). Thus  $A - P$  is normal.

Questions analogous to Q1 and Q2 may be asked for  $p \rightarrow 1$ :

Q3. Does  $\lim_{p \rightarrow 1} R_p$  exist?

If the answer to Q3 is yes, then

Q4. Can the limit in Q3 be identified by any characteristics?

An affirmative answer to Q4 would yield a canonical approximant for positive approximation in the trace norm.

Finally it seems as if Theorem 1.1 must have some extension at least to the compact operator case. Relevant to this problem is the following

**Example 4.2.** There exists a compact operator with no compact positive operator approximant.

Indeed, let  $\{e_1, e_2, \dots\}$  denote an orthonormal basis and let  $f$  be the vector  $f \equiv \sum e_k/k$ . Define  $Q$  to be the rank one orthogonal projection onto  $\text{sp}\{f\}$ ,  $C$  the compact operator given by  $C(e_k) = e_k/k$ ,  $B \equiv (1-Q) - \sqrt{1-C^2}$ , and finally set  $A = B + iC$ . Then  $A$  is a compact operator and has a unique positive approximant  $P_H$  [1, p. 282]. Now  $P_H$  is not compact since  $A - P_H$  is a multiple of a unitary.

**Q5.** Which compact operators admit compact positive approximants; is there a "minimal" approximant in this case?

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