

Examples of Chebyshev Sets in Matrix Spaces

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Let A be a matrix in $\mathbb{C}^{n \times n}$ and let $U\Sigma V^*$ be its singular value decomposition. The authors prove that for each $1 \leq k \leq n$ the set $\mathcal{G}_1^{(k)} = \{S \in \mathbb{C}^{n \times n} : \sum_{1 \leq j_1 < \dots < j_k \leq n} \sigma_{j_1}(S) \sigma_{j_2}(S) \cdots \sigma_{j_k}(S) \leq 1\}$ is a Chebyshev set in $\mathbb{C}^{n \times n}$ endowed with the spectral norm and that the metric projection is globally Lipschitz-continuous.

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In the 1980s the senior author studied various problems on matrix approximation; he was, in particular, interested in Chebyshev sets and suns. Approximation in matrix spaces proved to be rich in providing simple, but impressive examples as the papers [1, 4, 5, 7] show. We want to take up the results in [3, 4], extend, and re-prove them.

We denote by $\mathbb{C}^{n \times m}$ the vector space of complex $n \times m$ matrices over \mathbb{C} , $n, m \in \mathbb{N}$, with elements A, B, \dots . For $A \in \mathbb{C}^{n \times p}$ and $B \in \mathbb{C}^{p \times m}$, $n, m, p \in \mathbb{N}$, A^* denotes the adjoint of A in $\mathbb{C}^{p \times n}$ and AB the matrix product of A and B in $\mathbb{C}^{n \times m}$. Instead of $\mathbb{C}^{n \times 1}$ we write \mathbb{C}^n the vector space of complex column vectors; we also write z, u, \dots to denote its elements.

By l_n^2 we denote \mathbb{C}^n endowed with the Euclidian norm $|\cdot|_2$, and by $\mathcal{L}(l_n^2)$ the vector space of linear transformations of l_n^2 into itself. $\mathcal{L}(l_n^2)$ can be identified with $\mathbb{C}^{n \times n}$ endowed with the l^2 -operator norm, where $A \in \mathbb{C}^{n \times n}$ acts on $z \in \mathbb{C}^n$ via the matrix product Az . Since it will not lead to

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any ambiguity, in the following we shall write \mathbb{C}^n and $\mathbb{C}^{n \times n}$ instead of l_n^2 and $\mathcal{L}(l_n^2)$, respectively.

For an $A \in \mathbb{C}^{n \times n}$, we denote its *singular value decomposition* (SVD) by $U\Sigma V^*$, U and V are unitary matrices in $\mathbb{C}^{n \times n}$, and $\Sigma = \text{diag}(\sigma_1, \sigma_2, \dots, \sigma_n)$, $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n \geq 0$. Σ is uniquely determined by A ; its elements are the so-called *singular values* of A ; to indicate the dependence on A , we will write $\Sigma(A)$ and $\sigma_1(A), \sigma_2(A), \dots, \sigma_n(A)$, respectively. If A is non-singular, then UV^* is uniquely defined.

Here we will be interested in best approximation on $\mathbb{C}^{n \times n}$ w.r.t. the spectral norm. It will, however, be appropriate to consider more generally *unitarily invariant norms* $\|\cdot\|$; in particular, we will consider the *Schatten p -norms*, where

$$\text{for all } A \in \mathbb{C}^{n \times n}, \quad \|A\|_p^p = \|\Sigma\|_p^p = \sum_{j=1}^n \sigma_j^p.$$

For $p=1$ we speak of the (classical) Schatten norm $\|\cdot\|_1$, while for $p = \infty$, $\|\cdot\|_\infty$ just denotes the spectral norm on $\mathbb{C}^{n \times n}$. Introducing on $\mathbb{C}^{n \times n}$ the *sesqui-linear form*

$$\text{for all } A, B \in \mathbb{C}^{n \times n}, \quad \langle A, B \rangle = \text{trace } A^*B,$$

$\mathbb{C}^{n \times n}$ becomes an inner product space isomorphic to \mathbb{C}^{n^2} . The associated norm $\|\cdot\|_2$, the so-called Frobenius norm, is nothing but the 2-norm

$$\text{for all } A \in \mathbb{C}^{n \times n}, \quad \sum_{j,k=1}^n |a_{jk}|^2 = \|A\|_2^2 = \|\Sigma\|_2^2 = \sum_{j=1}^n \sigma_j^2.$$

For $k = 1, 2, \dots, n$, we define

$$\mathcal{S}_1^{(k)} = \left\{ S \in \mathbb{C}^{n \times n} : \sum_{1 \leq j_1 < j_2 < \dots < j_k \leq n} \sigma_{j_1}(S) \sigma_{j_2}(S) \dots \sigma_{j_k}(S) \leq 1 \right\}.$$

If $k = 1$, $\mathcal{S}_1^{(1)}$ coincides with the l_1 -unit ball in $\mathbb{C}^{n \times n}$, i.e., $\mathcal{S}_1^{(1)} = \{S \in \mathbb{C}^{n \times n} : \|S\|_1 \leq 1\}$, while for $k = n$ $\mathcal{S}_1^{(n)} = \{S \in \mathbb{C}^{n \times n} : |\det S| \leq 1\}$. It was proved in [4], that $\mathcal{S}_1^{(n)}$ is a Chebyshev set in $\mathbb{C}^{n \times n}$ w.r.t. $\|\cdot\|_\infty$ and that the metric projection $P_{\mathcal{S}_1^{(n)}}^\infty$ is globally Lipschitz-continuous. We will re-prove the result and extend it; our proof is different and more elementary.

THEOREM. *For $1 \leq k \leq n$, the set $\mathcal{S}_1^{(k)}$ is a Chebyshev set in $\mathbb{C}^{n \times n}$ w.r.t. the spectral norm.*

More precisely, if $A \in \mathbb{C}^{n \times n} \setminus \mathcal{S}_1^{(k)}$ and if $U\Sigma V^*$ is a SVD of A , then the distance of A from $\mathcal{S}_1^{(k)}$ is given by

$$d^{(k)} = \min \left\{ t > 0 : \sum_{1 \leq j_1 < j_2 < \dots < j_k \leq n} (\sigma_{j_1} - t)_+ (\sigma_{j_2} - t)_+ \cdots (\sigma_{j_k} - t)_+ = 1 \right\}.$$

With $r \in \mathbb{N}$, $k \leq r \leq n$, such that $\sigma_{r+1} < d^{(k)} \leq \sigma_r$ ($\sigma_{n+1} = 0$), and

$$\tilde{\Sigma}^{(k)} = \text{diag}[\sigma_1 - d^{(k)}, \dots, \sigma_r - d^{(k)}, 0, \dots, 0],$$

$\tilde{A}^{(k)} = U\tilde{\Sigma}^{(k)}V^*$ is the unique element of best approximation of A in $\mathcal{S}_1^{(k)}$.

Moreover, the metric projection onto $\mathcal{S}_1^{(k)}$ is globally Lipschitz-continuous.

Before we get into proving the theorem, let us remark that the element of best approximation (briefly, el. b. appr.) is well-defined. It follows from the result stated below; although it is well known, we consider it worthwhile to point this out.

Let $A \in \mathbb{C}^{n \times n}$, $A \neq 0$, and let $U\Sigma V^*$ and $X\Sigma Y^*$ be two SVDs of A , where

$$\Sigma = \text{diag}(\underbrace{\sigma_1, \dots, \sigma_1}_{n_1}, \underbrace{\sigma_2, \dots, \sigma_2}_{n_2}, \dots, \underbrace{\sigma_r, \dots, \sigma_r}_{n_r}, 0, \dots, 0),$$

$n_1 + \dots + n_r = \text{rank}(A)$. Then there exist unitary matrices P and Q in $\mathbb{C}^{n \times n}$, such that $X = UP$ and $Y = VQ$, and

$$P = \text{diag}(W_1, \dots, W_r, P_0) \quad \text{and} \quad Q = \text{diag}(W_1, \dots, W_r, Q_0),$$

where P_0, Q_0 , and W_j , $1 \leq j \leq r$, are unitary matrices in $\mathbb{C}^{n_0 \times n_0}$, $n_0 + \text{rank}(A) = n$, and in $\mathbb{C}^{n_j \times n_j}$, respectively. If A is nonsingular then $UV^* = XY^*$.

To prove the theorem it suffices to consider Σ instead of A and to prove that $\tilde{\Sigma}^{(k)}$ is the el. b. appr. of Σ in $\mathcal{S}_1^{(k)}$. Concerning the proof itself, we will prove more strongly that the *strict Kolmogorov condition*, namely,

$$\text{for all } S \in \mathcal{S}_1^{(k)}, S \neq \tilde{\Sigma}^{(k)}, \quad \min_{\substack{|q|_2 = 1; \\ q \in \mathbb{C}^r \times \{0\}^{n-r}}} \Re q^*(S - \tilde{\Sigma}^{(k)})q < 0 \quad (1)$$

holds, which implies uniqueness.

The following remark is in order. Let X be real or complex, normed vector space with norm $|\cdot|_X$, and let K be a (closed) subset of X . We denote the metric projection of X onto K by P_K . For $x \in X \setminus K$ and $k \in P_K(x)$, k is said to be a *solar point* of x in K , if $k \in P_K(x_t)$ for each point x_t on the ray from k through x ; clearly, $x_t = k + t(x - k)$, $t \in \mathbb{R}_+$. If for some subset K of X each $x \in X \setminus K$ has a solar point, then K is said to be a *sun* in X in the

sense of L. P. Vlasov. Solar points are best described by use of the semi-inner product $\langle \cdot, \cdot \rangle_s$ defined on $X \times X$;

for $x, y \in X$,

$$\langle y, x \rangle_s = \lim_{t \rightarrow 0^+} \frac{|x + ty|_X^2 - |x|_X^2}{2t} = \max\{ \Re \langle w, y \rangle : w \in \Phi(x) \},$$

Φ is the *duality map* on X ; i.e., for each $x \in X$: $\Phi(x) = \{ w \in X^* : \Re \langle w, x \rangle = |x|_X^2 = |w|_{X^*}^2 \}$. In the definition we consider X to be a vector space over the reals; this is often done in approximation theory. To distinguish the complex space X from the space X considered as a real vector space, one usually writes X_r —the subscript, however, will be dropped, when there will be no confusion. If X is an inner product space, the semi-inner products reduces to the inner product. The semi-inner product has many properties similar to the inner product, but it is in general far more restrictive, see, e.g., [2] for details.

To get back to the characterization of solar points, under the conditions given above, a point $k \in K$ is a solar point for $x \in X \setminus K$ if and only if

$$\text{for all } k' \in K, \quad 0 \leq \langle k - k', x - k \rangle_s; \tag{2}$$

the condition can be, and will be, interpreted as a *Kolmogorov condition*. All this is well-known and well documented.

To conclude the remark, in the finite dimensional setting a Chebyshev set is a Chebyshev sun, and the metric projection is continuous.

The proof of the theorem is based upon the following lemmas.

LEMMA 1. Let $\Sigma = \text{diag}(\sigma_1, \dots, \sigma_n)$, $\sigma_1 = \dots = \sigma_r > \sigma_{r+1} \geq \dots \geq \sigma_n \geq 0$. Then

$$\Phi_\infty(\Sigma) = \text{co} \{ qq^* \in \mathbb{C}^{r \times r} : q \in \mathbb{C}^r \times \{0\}^{n-r} \text{ and } |q|_2^2 = \sigma_1 \},$$

“co” means the (closed) convex hull, and consequently,

$$\text{for all } B \in \mathbb{C}^{n \times n}, \quad \langle B, \Sigma \rangle_\infty = \max_{\substack{|q|_2^2 = \sigma_1 \\ q \in \mathbb{C}^r \times \{0\}^{n-r}}} \Re q^* B q.$$

If $A \in \mathbb{C}^{n \times n}$ has $U \Sigma V^*$ to be its SVD, then $\Phi_\infty(A) = U \Phi_\infty(\Sigma) V^*$.

For a proof see [3]. Note that we used the subscript ∞ to indicate that we consider the spectral norm on $\mathbb{C}^{n \times n}$.

Our second lemma is concerned with unitarily invariant norms in general.

LEMMA 2. *For any unitarily invariant norm $\|\cdot\|$ on $\mathbb{C}^{n \times n}$*

$$\text{for all } A, B \in \mathbb{C}^{n \times n}, \quad \|\Sigma(A) - \Sigma(B)\| \leq \|A - B\|.$$

A proof of the basic inequality is, e.g. given in the monograph of R. A. Horn and Ch. R. Johnson [6, Theorem 7.4.51]. For the Frobenius norm the inequality is known as the inequality of Hoffmann and Wielandt.

Proof of the Theorem. For the proof we drop the superscript (k) ; this will not lead to any confusion.

We shall first prove uniqueness. Under the assumption of the theorem we have that

$$\Sigma - \tilde{\Sigma} = \text{diag}[d, \dots, d, \sigma_{r+1}, \dots, \sigma_n], \quad \sigma_{r+1} < d \leq \sigma_r,$$

where d denotes the distance of Σ from \mathcal{S}_1 . By Lemma 1, Kolmogorov's condition (2) reads

for all $S \in \mathcal{S}_1, S \neq \tilde{\Sigma}$,

$$0 < \langle \tilde{\Sigma} - S, \Sigma - \tilde{\Sigma} \rangle_\infty = d \cdot \max_{\substack{q \in \mathbb{C}^r \times \{0\}^{n-r} \\ |q|_2 = 1}} \Re q^* (\tilde{\Sigma} - S) q. \quad (3)$$

We will use Lemma 2 to conclude that

$$\langle \tilde{\Sigma} - \Sigma(S), \Sigma - \tilde{\Sigma} \rangle_\infty \leq \langle \tilde{\Sigma} - S, \Sigma - \tilde{\Sigma} \rangle_\infty. \quad (4)$$

Indeed, $\|\Sigma - \tilde{\Sigma} + t(\tilde{\Sigma} - \Sigma(S))\|_\infty \leq \|\Sigma - \tilde{\Sigma} + t(\tilde{\Sigma} - S)\|_\infty$ for all $t \in \mathbb{R}_+$, and the inequality follows.

Let us assume that $S \in \mathcal{S}_1$ is such that $\Sigma(S) \neq \tilde{\Sigma}$. We claim that

$$0 < \max_{\substack{q \in \mathbb{C}^r \times \{0\}^{n-r} \\ |q|_2 = 1}} \Re q^* (\tilde{\Sigma} - \Sigma(S)) q. \quad (5)$$

Since

$$\sum_{1 \leq j_1 < \dots < j_k \leq n} \sigma_{j_1}(S) \cdots \sigma_{j_k}(S) \leq 1 \quad \text{and} \quad \sum_{1 \leq j_1 < \dots < j_k \leq n} \tilde{\sigma}_{j_1} \cdots \tilde{\sigma}_{j_k} = 1,$$

there is at least one index in $\{1, \dots, r\}$, say, j_0 for which $\sigma_{j_0}(S) < \tilde{\sigma}_{j_0}$. For otherwise, for all j $\tilde{\sigma}_j \leq \sigma_j(S)$, but then

$$1 = \sum_{1 \leq j_1 < \dots < j_k \leq n} \tilde{\sigma}_{j_1} \cdots \tilde{\sigma}_{j_k} \leq \sum_{1 \leq j_1 < \dots < j_k \leq n} \sigma_{j_1}(S) \cdots \sigma_{j_k}(S) \leq 1,$$

implies that for all j $\sigma_j(S) = \tilde{\sigma}_j$, contradicting the condition $\Sigma(S) \neq \tilde{\Sigma}$. Hence $\sigma_{j_0}(S) < \tilde{\sigma}_{j_0}$ and for $q = e_{j_0}$, the j_0 's natural basis element in \mathbb{C}^r , $e_{j_0}^*(\tilde{\Sigma} - \Sigma(S)) e_{j_0} = \tilde{\sigma}_{j_0} - \sigma_{j_0}(S) > 0$, which proves (5). From (5) and (4) we obtain (3).

Next assume that $S \in \mathcal{S}_1$ is such that $S \neq \tilde{\Sigma}$, but $\Sigma(S) = \tilde{\Sigma}$. Assume Kolmogorov's condition (3) does not hold; i.e.,

$$\text{for all } q \in \mathbb{C}^r \times \{0\}^{n-r}, \quad \Re q^*(\tilde{\Sigma} - S) q \leq 0. \tag{6}$$

Setting $S = B + iC$ with B and C Hermitian ($B = (S + S^*)/2$ and $C = -i(S - S^*)/2$), then (6) just reads

for all $q \in \mathbb{C}^r$, $q^*(\tilde{\Sigma}_r - B_r) q \leq 0$ or $q^*(B_r - \tilde{\Sigma}_r) q \geq 0$; i.e., the matrix $B_r - \tilde{\Sigma}_r$ is positive semi-definite, where the subscript r indicates that we restrict the matrix to the r th principal submatrix. This forces $\tilde{\sigma}_j \leq b_{jj}$ for $1 \leq j \leq r$, and consequently,

$$\|\tilde{\Sigma}\|_2^2 = \sum_{j=1}^r \tilde{\sigma}_j^2 \leq \sum_{j=1}^r \tilde{\sigma}_j^2 + \sum_{j=r+1}^n b_{jj}^2 + \sum_{j \neq k} |b_{jk}|^2 \leq \|B\|_2^2 \leq \|S\|_2^2 = \|\tilde{\Sigma}\|_2^2.$$

Hence, all b_{jj} vanish for $r + 1 \leq j \leq n$ as well as all $b_{jk}, j \neq k$; i.e., $B = \tilde{\Sigma}$.

Note that $\Re \langle \Sigma, iC \rangle = 0$, since C is Hermitian. It follows that $\|S\|_2^2 = \|\tilde{\Sigma} + iC\|_2^2 = \|\tilde{\Sigma}\|_2^2 + \|C\|_2^2 = \|\Sigma(S)\|_2^2 = \|\tilde{\Sigma}\|_2^2$, hence $C = 0$, and consequently $S = \tilde{\Sigma}$ a contradiction.

Thus, the strict Kolmogorov condition (1) is satisfied in both cases, proving the first part of the theorem. It remains to prove that the metric projection is globally Lipschitz-continuous. As we remarked above, it is continuous, but we have more.

Let us at first consider diagonal matrices only, say, Σ and Σ' not in \mathcal{S}_1 with el. b. appr. $\tilde{\Sigma}$ and $\tilde{\Sigma}'$ and distances d and d' , respectively. Note that for any $1 \leq j \leq n$

$$|\tilde{\sigma}_j - \tilde{\sigma}'_j| = |(\sigma_j - d)_+ - (\sigma'_j - d')_+| \leq |\sigma_j - \sigma'_j| + |d - d'|,$$

and consequently,

$$\|\tilde{\Sigma} - \tilde{\Sigma}'\|_2 \leq \|\Sigma - \Sigma'\|_2 + \sqrt{n}|d - d'|,$$

but $|d - d'| \leq \|\Sigma - \Sigma'\|_\infty \leq \|\Sigma - \Sigma'\|_2$, giving

$$\|\tilde{\Sigma} - \tilde{\Sigma}'\|_2 \leq 2\sqrt{n} \|\Sigma - \Sigma'\|_2. \quad (7)$$

To consider the general situation, let A, A' be two matrices in $\mathbb{C}^{n \times n} \setminus \mathcal{S}_1$. W.l.o.g., we may assume that $A = \Sigma, \Sigma$ as given above, and that $A' = XE'Y^*$. Then the metric projection of Σ and A' are $\tilde{\Sigma}$ and $X\tilde{\Sigma}'Y^*$, respectively. We have to estimate the difference

$$\|\tilde{\Sigma} - X\tilde{\Sigma}'Y^*\|$$

w.r.t. some (unitarily invariant) norm, say the 2-norm again. By the triangular inequality

$$\|\tilde{\Sigma} - X\tilde{\Sigma}'Y^*\|_2 \leq \|\tilde{\Sigma} - X\tilde{\Sigma}Y^*\|_2 + \|X(\tilde{\Sigma} - \tilde{\Sigma}')Y^*\|_2.$$

For the second term of the right-hand side we have

$$\|X(\tilde{\Sigma} - \tilde{\Sigma}')Y^*\|_2 = \|\tilde{\Sigma} - \tilde{\Sigma}'\|_2 \leq 2\sqrt{n} \|\Sigma - \Sigma'\|_2 \leq 2\sqrt{n} \|\Sigma - A'\|_2.$$

The two estimates follow from (7) and Lemma 2, respectively.

To estimate the first term, we take advantage of an estimate obtained in [3].

Let Σ and $\tilde{\Sigma}$ and d be given as above, and let $\Sigma + E = X\Sigma Y^$ be a perturbation of Σ so that its SVD leaves the singular values unchanged. Then*

$$\|\tilde{\Sigma} - X\tilde{\Sigma}Y^*\|_2 \leq \|E\|_2.$$

The estimate holds for any $\tilde{\Sigma} = \text{diag}(\tilde{\sigma}_1, \dots, \tilde{\sigma}_n)$, $\tilde{\sigma}_1 \geq \dots \geq \tilde{\sigma}_n \geq 0$, for which for $1 \leq j \leq l \leq n$ the inequality $\tilde{\sigma}_j \pm \tilde{\sigma}_l \leq \sigma_j \pm \sigma_l$ holds. In our case, $\tilde{\sigma}_j = \sigma_j - d$ for $1 \leq j \leq r$ and $\tilde{\sigma}_j = 0$ for $r + 1 \leq j \leq n$, where $\sigma_{r+1} < d \leq \sigma_r$.

As a consequence, we obtain

$$\begin{aligned} \|\tilde{\Sigma} - X\tilde{\Sigma}Y^*\|_2 &\leq \|\Sigma - X\Sigma Y^*\|_2 \leq \|\Sigma - X\Sigma'Y^*\|_2 + \|X(\Sigma' - \Sigma)Y^*\|_2 \\ &\leq 2\|\Sigma - A'\|_2. \end{aligned}$$

Adding the two estimates, proves Lipschitz-continuity and completes the proof of the theorem.

Instead of the class $\mathcal{S}_1^{(k)}$, $1 \leq k \leq n$, let us consider, more generally, the class

for $\delta > 0$,

$$\mathcal{S}_\delta^{(k)} = \left\{ S \in \mathbb{C}^{n \times n} : \sum_{1 \leq j_1 < j_2 < \dots < j_k \leq n} \sigma_{j_1}(S) \sigma_{j_2}(S) \dots \sigma_{j_k}(S) \leq \delta \right\}.$$

It follows from above that $\mathcal{S}_\delta^{(k)}$ is a Chebyshev set in $\mathbb{C}^{n \times n}$ w.r.t. the spectral norm for each $\delta > 0$. Moreover, repeating the proof of Lipschitz-continuity of the metric projection, we see that it remains unchanged when we replace $P_{\mathcal{S}_1^{(k)}}^\infty$ by $P_{\mathcal{S}_\delta^{(k)}}^\infty$; i.e., $P_{\mathcal{S}_1^{(k)}}^\infty : \mathbb{C}^{n \times n} \rightarrow \mathcal{S}_\delta^{(k)}$ is globally Lipschitz-continuous with Lipschitz constant independent of $1 \leq k \leq n$ and $\delta > 0$.

If we allow δ to converge towards zero, the class $\mathcal{S}_\delta^{(k)}$ converges to

$$\mathcal{S}_0^{(k)} = \{ S \in \mathbb{C}^{n \times n} : \text{rank } S \leq k - 1 \}.$$

Trivially, $\mathcal{S}_0^{(1)}$ just reduces to $\{0\}$, while $\mathcal{S}_0^{(n)}$ is the class of singular matrices.

Let $A \in \mathbb{C}^{n \times n} \setminus \mathcal{S}_0^{(k)}$, and let $U\Sigma V^*$ be its SVD. Then, its singular value $\sigma_k \neq 0$ and $A \notin \mathcal{S}_\delta^{(k)}$ for $\delta > 0$ sufficiently small. By the theorem the distance $\text{dist}_\infty(A; \mathcal{S}_\delta^{(k)})$ is given by

$$d_\delta^{(k)} = \min \left\{ t > 0 : \sum_{1 \leq j_1 \leq \dots \leq j_k \leq n} (\sigma_{j_1} - t)_+ \dots (\sigma_{j_k} - t)_+ = \delta \right\},$$

and $U\tilde{\Sigma}_\delta^{(k)}V^*$ is the unique el. b. appr. of A from $\mathcal{S}_\delta^{(k)}$ with

$$\tilde{\Sigma}_\delta^{(k)} = \text{diag}[\sigma_1 - d_\delta^{(k)}, \dots, \sigma_k - d_\delta^{(k)}, 0, \dots, 0].$$

Since $d_\delta^{(k)}$ converges to σ_k as $\delta \rightarrow 0+$, $\text{dist}(A; \mathcal{S}_0^{(k)}) = \sigma_k$ and $U\tilde{\Sigma}_0^{(k)}V^*$ is an el. b. appr. of A in $\mathcal{S}_0^{(k)}$, where $\tilde{\Sigma}_0^{(k)} = \text{diag}[\sigma_1 - \sigma_k, \dots, \sigma_{k-1} - \sigma_k, 0, \dots, 0]$. More precisely,

COROLLARY. For $1 \leq k \leq n$ $\mathcal{S}_0^{(k)}$ is a sun in $\mathbb{C}^{n \times n}$ w.r.t. the spectral norm. If $A \in \mathbb{C}^{n \times n} \setminus \mathcal{S}_0^{(k)}$ and if $U\Sigma V^*$ is its SVD, then

$$\text{dist}_\infty(A; \mathcal{S}_0^{(k)}) = \sigma_k \quad \text{and} \quad U\tilde{\Sigma}_0^{(k)}V^* \text{ is a solar point of } A \text{ in } \mathcal{S}_0^{(k)},$$

$\tilde{\Sigma}_0^{(k)}$ being $\text{diag}[\sigma_1 - \sigma_k, \dots, \sigma_{k-1} - \sigma_k, 0, \dots, 0]$.

In addition, $\mathbb{C}^{n \times n} \ni A = U\Sigma V^* \mapsto U\tilde{\Sigma}_0^{(k)}V^*$ is a Lipschitz-continuous selection of $P_{\mathcal{S}_0^{(k)}}^\infty$.

This re-proves results in [3, 4]. In this context we would like to add the following observation.

As we remarked above, $\mathcal{S}_1^{(1)}$ is just equal to the l_1 -unit ball $\bar{b}_1^{(1)}(0)$ of $\mathbb{C}^{n \times n}$. $\mathcal{S}_1^{(1)}$ being a Chebyshev set in $\mathbb{C}^{n \times n}$ w.r.t. $\|\cdot\|_\infty$ then means that for each $A \in \mathbb{C}^{n \times n} \setminus \mathcal{S}_1^{(1)}$ the distance ball $b_{\mathcal{S}_1^{(1)}}^{(\infty)}(A)$ of A touches $\bar{b}_1^{(1)}(0)$ exactly at $P_{\mathcal{S}_1^{(1)}}(A) - b_{\mathcal{S}_1^{(1)}}^{(\infty)}(A)$ denotes the closed ball centered at A with radius $\text{dist}_\infty(A; \mathcal{S}_1^{(1)})$. Interpreting this statement right, leads to the following.

COROLLARY. *The l_∞ -unit ball $\bar{b}_1^{(\infty)}(0) = \{S \in \mathbb{C}^{n \times n} : \sigma_1(S) \leq 1\}$ is a Chebyshev set in $\mathbb{C}^{n \times n}$ w.r.t. the Schatten 1-norm. Moreover, the metric projection is globally Lipschitz-continuous.*

We conclude our investigation with the following remarks. Let $A \in \mathbb{C}^{n \times n}$ be Hermitian; i.e., $A = UAU^*$, $A = \text{diag}[\lambda_1, \lambda_2, \dots, \lambda_n]$ and U unitary, where the λ_j 's are the eigenvalues of A counting their multiplicities and ordered such that $|\lambda_1| \geq |\lambda_2| \geq \dots \geq |\lambda_n| \geq 0$. The selected el.'s b. appr. of A in the classes $\mathcal{S}_1^{(k)}$ and $\mathcal{S}_0^{(k)}$, $1 \leq k \leq n$, are then given by

$$\tilde{A}^{(k)} = U\tilde{\Lambda}^{(k)}U^* \quad \text{and} \quad \tilde{A}_0^{(k)} = U\tilde{\Lambda}_0^{(k)}U^*, \text{ respectively.}$$

Indeed, if we set $D_A = \text{diag}[\text{sign } \lambda_1, \text{sign } \lambda_2, \dots, \text{sign } \lambda_n]$ it follows that $UD_A\Sigma_A U^*$ is the SVD of A with $\Sigma_A = D_A A$, and consequently, $\tilde{\Lambda}^{(k)} = D_A \tilde{\Sigma}^{(k)}$ and $\tilde{\Lambda}_0^{(k)} = D_A \tilde{\Sigma}_0^{(k)}$, respectively, which proves our claim. A similar statement holds true if we assume $A \in \mathbb{C}^{n \times n}$ to be (complex) symmetric, but let us stick to the Hermitian case.

As we pointed out in our general discussion an approximation in a complex normed vector space X , approximation in X means approximation in X_r . For $\mathbb{C}_r^{n \times n}$ the subset of Hermitian matrices then forms a linear subspace, say, $\mathbb{H}^{n \times n}$. It follows from above that *rank approximation* in $\mathbb{H}^{n \times n}$, endowed with the spectral norm, means first rank approximation in $\mathbb{C}^{n \times n}$ and then restricting the results to $\mathbb{H}^{n \times n}$.

On the other hand, instead of dealing with the class $\mathcal{S}_1^{(k)} \cap \mathbb{H}^{n \times n}$, $1 \leq k \leq n$, it seemed to us more natural to replace it by

$$\mathcal{H}_1^{(k)} = \left\{ S \in \mathbb{H}^{n \times n} : \sum_{1 \leq j_1 \leq j_2 < \dots < j_k \leq n} \lambda_{j_1}(S) \lambda_{j_2}(S) \dots \lambda_{j_k}(S) \leq 1 \right\},$$

$$1 \leq k \leq n.$$

For $k=1$, $\mathcal{H}_1^{(1)}$ is just the subset $\{S \in \mathbb{H}^{n \times n} : \sum_{j=1}^n \lambda_j(S) \leq 1\}$, while for $k=n$, $\mathcal{H}_1^{(n)} = \{S \in \mathbb{H}^{n \times n} : \det S \leq 1\}$. For these two cases we were able to prove that the sets are Chebyshevian; we determined the el. b. appr. for a matrix A in $\mathbb{H}^{n \times n} \setminus \mathcal{H}_1^{(n)}$ and $\mathbb{H}^{n \times n} \setminus \mathcal{H}_1^{(1)}$, respectively, and verified that the metric projection is globally Lipschitz-continuous. We will spare the reader with details.

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