COORDINATE SHADOWS OF SEMIDEFINITE AND EUCLIDEAN DISTANCE MATRICES*

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Abstract. We consider the projected semidefinite and Euclidean distance cones onto a subset of the matrix entries. These two sets are precisely the input data defining feasible semidefinite and Euclidean distance completion problems. We classify when these sets are closed and use the boundary structure of these two sets to elucidate the Krislock–Wolkowicz facial reduction algorithm. In particular, we show that under a chordality assumption, the "minimal cones" of these problems admit combinatorial characterizations. As a by-product, we record a striking relationship between the complexity of the general facial reduction algorithm (singularity degree) and facial exposedness of conic images under a linear mapping.

Key words. matrix completion, semidefinite programming, Euclidean distance matrices, facial reduction, Slater condition, projection, closedness

AMS subject classifications. 90C22, 90C46, 52A99

DOI. 10.1137/140968318

1. Introduction. To motivate the discussion, consider an undirected graph G =(V, E) with vertex set $V = \{1, \ldots, n\}$ and edge set $E \subset V \times V$ possibly containing self-loops. The classical positive semidefinite (PSD) completion problem asks whether given a data vector a indexed by E, there exists an $n \times n$ positive semidefinite matrix X completing a, meaning $X_{ij} = a_{ij}$ for all $ij \in E$. Similarly, the Euclidean distance matrix (EDM) completion problem asks whether given such a data vector, there exists a Euclidean distance matrix completing it. For a survey of these two problems, see for example [22, 2, 24, 25]. The semidefinite and Euclidean distance completion problems are often mentioned in the same light due to a number of parallel results; see, e.g., [21]. Here, we consider a related construction: projections of the PSD cone \mathcal{S}^n_+ and the EDM cone \mathcal{E}^n onto matrix entries indexed by E. These "coordinate shadows," denoted by $\mathcal{P}(\mathcal{S}^n_+)$ and $\mathcal{P}(\mathcal{E}^n)$, respectively, appear naturally: they are precisely the sets of data vectors that render the corresponding completion problems feasible. We note that these sets are interesting types of "spectrahedral shadows"—an area of intensive research in recent years. For a representative sample of recent papers on spectrahedral shadows, we refer to [25, 11, 15, 16, 3] and references therein.

In this paper, our goal is twofold: we will (1) highlight the geometry of the two sets $\mathcal{P}(\mathcal{S}^n_+)$ and $\mathcal{P}(\mathcal{E}^n)$ and (2) illustrate how such geometric considerations yield a much simplified and transparent analysis of an EDM completion algorithm proposed in [18]. To this end, we begin by asking a basic question: Under what conditions are the coordinate shadows $\mathcal{P}(\mathcal{S}^n_+)$ and $\mathcal{P}(\mathcal{E}^n)$ closed?

^{*}Received by the editors May 8, 2014; accepted for publication (in revised form) April 7, 2015; published electronically June 18, 2015.

http://www.siam.org/journals/siopt/25-2/96831.html

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This question sits in a broader context still of deciding if a linear image of a closed convex set is itself closed—a thoroughly studied topic due to its fundamental connection to constraint qualifications and strong duality in convex optimization; see, e.g., [31, 10, 9, 27, 28] and references therein. In contrast to the general setting, a complete answer to this question in our circumstances is easy to obtain. An elementary argument¹ shows that $\mathcal{P}(\mathcal{E}^n)$ is always closed, whereas $\mathcal{P}(\mathcal{S}^n_+)$ is closed if, and only if, the set of vertices attached to self-loops $L = \{i \in V : ii \in E\}$ is disconnected from its complement L^c (Theorems 3.1, 3.2). Moreover, whenever there is an edge joining L and L^c , one can with ease exhibit vectors lying in the closure of $\mathcal{P}(\mathcal{S}^n_+)$ but not in the set $\mathcal{P}(\mathcal{S}^n_+)$ itself, thereby certifying that $\mathcal{P}(\mathcal{S}^n_+)$ is not closed.

To illustrate the algorithmic significance of the coordinate shadows $\mathcal{P}(\mathcal{S}^n_+)$ and $\mathcal{P}(\mathcal{E}^n)$, consider first the feasible region of the PSD completion problem:

$$\{X \in \mathcal{S}^n_+ : X_{ij} = a_{ij} \text{ for } ij \in E\}.$$

For this set to be nonempty, the data vector $a \in \mathbb{R}^E$ must be a partial PSD matrix, meaning all its principal submatrices are positive semidefinite. This, however, does not alone guarantee the inclusion $a \in \mathcal{P}(S^n_+)$, unless the restriction of G to L is chordal and L is disconnected from L^c (Theorem 2.1, [12, Theorem 7]). On the other hand, the authors of [18] noticed that even if the feasible set is nonempty, the Slater condition (i.e., existence of a positive definite completion) will often fail: small perturbations to any specified principal submatrix of a having deficient rank can render the semidefinite completion problem infeasible. In other words, in this case the partial matrix a lies on the boundary of $\mathcal{P}(S^n_+)$ —the focus of this paper. An entirely analogous situation occurs for EDM completions

$$\{X \in \mathcal{E}^n : X_{ij} = a_{ij} \text{ for } ij \in E\},\$$

with the rank of each principal submatrix of $a \in \mathbb{R}^{E}$ replaced by its "embedding dimension." In [18], the authors propose a preprocessing strategy utilizing the cliques in the graph G to systematically decrease the size of the EDM completion problem. Roughly speaking, the authors use each clique to find a face of the EDM cone containing the entire feasible region, and then iteratively intersect such faces. The numerical results in [18] were impressive. In the current work, we provide a much simplified and transparent geometric argument behind their algorithmic idea, with the boundary of $\mathcal{P}(\mathcal{E}^n)$ playing a key role. As a result, (a) we put their techniques in a broader setting unifying the PSD and EDM cases, and (b) the techniques developed here naturally lead to a robust variant of the method for noisy (inexact) EDM completion problems [8]—a more realistic setting. In particular, we show that when G is chordal and all cliques are considered, the preprocessing technique discovers the minimal face of \mathcal{E}^n (respectively, \mathcal{S}_{\perp}^{n}) containing the feasible region; see Theorems 4.7 and 4.14. This in part explains the observed success of the method in [18]. Thus in contrast to general semidefinite programming, the minimal face of the PSD cone containing the feasible region of the PSD completion problem under a chordality assumption (one of the simplest semidefinite programming problems) admits a purely combinatorial description.

As a byproduct, we record a striking relationship between the complexity of the general facial reduction algorithm (singularity degree) and facial exposedness of conic images under a linear mapping; see Theorem 4.1. To the best of our knowledge, this basic relationship either has gone unnoticed or has mostly been forgotten.

¹The elementary proofs of Theorem 3.1 and 3.2 were suggested to us by an anonymous referee, for which we are very grateful. Our original reasoning was based on the more general considerations in [27]; see also the discussion after Theorem 8 in [28].

The outline of the manuscript is as follows. In section 2 we record basic results on convex geometry and PSD and EDM completions. In section 3, we consider when the coordinate shadows $\mathcal{P}(\mathcal{S}^n_+)$ and $\mathcal{P}(\mathcal{E}^n)$ are closed, while in section 4 we discuss the aforementioned clique facial reduction strategy.

2. Preliminaries.

2.1. Basic elements of convex geometry. We begin with some notation, following closely the classical text [31]. Consider a Euclidean space \mathbb{E} with the inner product $\langle \cdot, \cdot \rangle$. The adjoint of a linear mapping $\mathcal{M} \colon \mathbb{E} \to \mathbb{Y}$, between two Euclidean spaces \mathbb{E} and \mathbb{Y} , is written as \mathcal{M}^* , while the range and kernel of \mathcal{M} are denoted by rge \mathcal{M} and ker \mathcal{M} , respectively. We denote the closure, boundary, interior, and relative interior of a set Q in \mathbb{E} by cl Q, bnd Q, int Q, and ri Q, respectively. Consider a convex cone C in \mathbb{E} . The linear span and the orthogonal complement of the linear span of C will be denoted by span C and C^{\perp} , respectively. For a vector v, we let $v^{\perp} := \{v\}^{\perp}$. We associate with C the nonnegative polar cone

$$C^* = \{ y \in \mathbb{E} : \langle y, x \rangle \ge 0 \text{ for all } x \in C \}$$

The second polar $(C^*)^*$ coincides with the original C if, and only if, C is closed. A convex subset $F \subseteq C$ is a face of C, denoted $F \trianglelefteq C$, if F contains any line segment in C whose relative interior intersects F. The minimal face containing a set $S \subseteq C$, denoted face(S, C), is the intersection of all faces of C containing S. When S is itself a convex set, then face(S, C) is the smallest face of C intersecting the relative interior of S. A face F of C is an exposed face when there exists a vector $v \in C^*$ satisfying $F = C \cap v^{\perp}$. In this case, we say that v exposes F. The cone C is facially exposed when all faces of C are exposed. In particular, the cones of positive semidefinite and Euclidean distance matrices, which we will focus on shortly, are facially exposed. With any face $F \trianglelefteq C$, we associate a face of the polar C^* , called the conjugate face $F^{\perp} := C^* \cap F^{\perp}$. Equivalently, F^{\perp} is the face of C^* exposed by any point $x \in \mathrm{ri} F$, that is, $F^{\perp} := C^* \cap x^{\perp}$. Thus, in particular, conjugate faces are always exposed. Not surprisingly then equality $(F^{\perp})^{\perp} = F$ holds if, and only if, $F \trianglelefteq C$ is exposed.

2.2. Semidefinite and Euclidean distance matrices. We will focus on two particular realizations of the Euclidean space \mathbb{E} : the *n*-dimensional vector space \mathbb{R}^n with a fixed basis and the induced *dot-product* $\langle \cdot, \cdot \rangle$ and the vector space of $n \times n$ real symmetric matrices S^n with the *trace inner product* $\langle A, B \rangle := \text{trace } AB$. The symbols \mathbb{R}_+ and \mathbb{R}_{++} will stand for the nonnegative orthant and its interior in \mathbb{R}^n , while S^n_+ and S^n_{++} will stand for the cones of positive semidefinite and positive definite matrices in S^n (or *PSD* and *PD* for short), respectively. We let $e \in \mathbb{R}^n$ be the vector of all ones, and for any vector $v \in \mathbb{R}^n$, the symbol Diag(v) will denote the $n \times n$ diagonal matrix with v on the diagonal.

It is well known that all faces of \mathcal{S}^n_+ are convex cones that can be expressed as

$$F = \left\{ U \begin{bmatrix} A & 0 \\ 0 & 0 \end{bmatrix} U^T : A \in \mathcal{S}^r_+ \right\}$$

for some orthogonal matrix U and some integer r = 0, 1, ..., n. Such a face can equivalently be written as $F = \{X \in S^n_+ : \operatorname{rge} X \subset \operatorname{rge} \overline{U}\}$, where \overline{U} is formed from the first r columns of U. The conjugate face of such a face F is then

$$F^{\triangle} = \left\{ U \begin{bmatrix} 0 & 0 \\ 0 & A \end{bmatrix} U^T : A \in \mathcal{S}^{n-r}_+ \right\}.$$

For any convex set $Q \subset S^n_+$, the cone face (Q, S^n_+) coincides with face (X, S^n_+) , where X is any maximal rank matrix in Q.

A matrix $D \in S^n$ is an Euclidean distance matrix (or EDM for short) if there exist n points p_i (for i = 1, ..., n) in some Euclidean space \mathbb{R}^k satisfying $D_{ij} = ||p_i - p_j||^2$ for all indices i, j. These points are then said to realize D. The smallest integer k for which this realization of D by n points is possible is the embedding dimension of D and will be denoted by embdim D. We let \mathcal{E}^n be the set of $n \times n$ Euclidean distance matrices. There is a close relationship between PSD and EDM. Indeed, \mathcal{E}^n is a closed convex cone that is linearly isomorphic to S^{n-1}_+ . To state this precisely, consider the mapping

$$\mathcal{K}: \mathcal{S}^n \to \mathcal{S}^n$$

defined by

(2.1)
$$\mathcal{K}(X)_{ij} := X_{ii} + X_{jj} - 2X_{ij}.$$

Then the adjoint $\mathcal{K}^* \colon \mathcal{S}^n \to \mathcal{S}^n$ is given by

$$\mathcal{K}^*(D) = 2(\operatorname{Diag}(De) - D)$$

and the equations

(2.2)
$$\operatorname{rge} \mathcal{K} = \mathcal{S}_H, \quad \operatorname{rge} \mathcal{K}^* = \mathcal{S}_c$$

hold, where

(2.3)
$$\mathcal{S}_c := \{ X \in \mathcal{S}^n : Xe = 0 \}, \qquad \mathcal{S}_H := \{ D \in \mathcal{S}^n : \operatorname{diag}(D) = 0 \}$$

are the *centered* and *hollow matrices*, respectively. It is known that \mathcal{K} maps \mathcal{S}^n_+ onto \mathcal{E}^n , and moreover the restricted mapping

(2.4) $\mathcal{K}: \mathcal{S}_c \to \mathcal{S}_H$ is a linear isomorphism carrying $\mathcal{S}_c \cap \mathcal{S}_+^n$ onto \mathcal{E}^n .

In turn, it is easy to see that $S_c \cap S^n_+$ is a face of S^n_+ isomorphic to S^{n-1}_+ ; see the discussion after Lemma 4.12 for more details. These and other related results have appeared in a number of publications; see, for example, [1, 37, 14, 35, 13, 36, 20, 19, 38].

2.3. Semidefinite and Euclidean distance completions. The focus of the current work is on the PSD and EDM completion problems; see, e.g., [17, Chapter 49]. Throughout the rest of the manuscript, we fix an undirected graph G = (V, E) with a vertex set $V = \{1, ..., n\}$ and an edge set $E \subset V \times V$. As usual, we identify the symbols ij and ji with a single edge. Observe that we allow self-loops. These loops will play an important role in what follows, and hence we define L to be the set of all vertices i satisfying $ii \in E$, that is, those vertices that are attached to a loop.

Any vector $a = [a_{ij}]_{ij \in E}$ satisfying $a_{ij} = a_{ji}$ for all $ij \in E$ is called a *partial* matrix. We denote the vector space of partial matrices by \mathbb{R}^E . Note that for any edge $ij \in E$ with $i \neq j$, the pairs ij and ji index different (but equal in value) coordinates of any $a \in \mathbb{R}^E$. In particular, partial matrices have 2|E| - |L| entries. Define now the projection map $\mathcal{P}: S^n \to \mathbb{R}^E$ by setting

$$\mathcal{P}(A) = (A_{ij})_{ij \in E}$$

that is, $\mathcal{P}(A)$ is the vector of all the entries of A indexed by E. The adjoint map $\mathcal{P}^*: \mathbb{R}^E \to \mathcal{S}^n$ is found by setting

$$(\mathcal{P}^*(y))_{ij} = \begin{cases} y_{ij} & \text{if } ij \in E, \\ 0 & \text{otherwise.} \end{cases}$$

Define also the Laplacian operator $\mathcal{L} \colon \mathbb{R}^E \to \mathcal{S}^n$ by setting

$$\mathcal{L}(a) := \frac{1}{2} (\mathcal{P} \circ \mathcal{K})^*(a) = \text{Diag}(\mathcal{P}^*(a)e) - \mathcal{P}^*(a).$$

Consider a partial matrix $a \in \mathbb{R}^E$ whose components are all strictly positive. Classically, then the Laplacian matrix $\mathcal{L}(a)$ is positive semidefinite and moreover the kernel of $\mathcal{L}(a)$ is determined only by the connectivity of the graph G; see, for example, [7], [17, Chapter 47]. Consequently all partial matrices with strictly positive weights define the same minimal face of the positive semidefinite cone. In particular, when G is connected, we have the equalities

(2.5)
$$\ker \mathcal{L}(a) = \operatorname{span}\{e\}$$
 and $\operatorname{face}(\mathcal{L}(a), \mathcal{S}^n_+) = \mathcal{S}_c \cap \mathcal{S}^n_+$.

A symmetric matrix $A \in S^n$ is a completion of a partial matrix $a \in \mathbb{R}^E$ if it satisfies $\mathcal{P}(A) = a$. We say that a completion $A \in S^n$ of a partial matrix $a \in \mathbb{R}^E$ is a *PSD completion* if A is a PSD matrix. Thus the image $\mathcal{P}(S^n_+)$ is the set of all partial matrices that are PSD completable. A partial matrix $a \in \mathbb{R}^E$ is a *partial PSD matrix* if all existing principal submatrices, defined by a, are PSD matrices. Finally we call G itself a *PSD completable graph* if every partial PSD matrix $a \in \mathbb{R}^E$ is completable to a PSD matrix. *PD completions, partial PD matrices, and PD completable graphs* are defined similarly.

We call a graph *chordal* if any cycle of four or more nodes (vertices) has a chord, i.e., an edge exists joining any two nodes that are not adjacent in the cycle. Before we proceed, a few comments on completability are in order. In [12, Proposition 1], the authors claim that G is PSD completable (PD, respectively) if, and only if, the graph induced on L by G is PSD completable (PD, respectively). In light of this, the authors then reduce all their arguments to this induced subgraph. It is easy to see that the statement above does not hold for PSD completability (but is indeed valid for PD completability). Consider, for example, the partial PSD matrix $\begin{bmatrix} 0 & 1\\ 1 & -? \end{bmatrix}$, which is clearly not PSD completable. Taking this into account, the correct statement of their main result [12, Theorem 7] is as follows. See also the discussion in [23].

THEOREM 2.1 (PSD completable matrices and chordal graphs). The following are true.

- 1. The graph G is PD completable if, and only if, the graph induced by G on L is chordal.
- The graph G is PSD completable if, and only, if the graph induced by G on L is chordal and L is disconnected from L^c.

With regard to EDMs, we will always assume $L = \emptyset$ for the simple reason that the diagonal of an EDM is always fixed at zero. With this in mind, we say that a completion $A \in S^n$ of a partial matrix $a \in \mathbb{R}^E$ is an *EDM completion* if A is an EDM. Thus the image $\mathcal{P}(\mathcal{E}^n)$ (or equivalently $\mathcal{L}^*(S^n_+)$) is the set of all partial matrices that are EDM completable. We say that a partial matrix $a \in \mathbb{R}^E$ is a *partial EDM* if any existing principal submatrix, defined by a, is an EDM. Finally we say that G is an

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EDM completable graph if any partial EDM is completable to an EDM. The following theorem is analogous to Theorem 2.1. For a proof, see [4].

THEOREM 2.2 (Euclidean distance completability and chordal graphs). The graph G is EDM completable if, and only if, G is chordal.

3. Closedness of the projected PSD and EDM cones. We begin this section by characterizing when the projection of the PSD cone S^n_+ onto some subentries is closed. To illustrate, consider the simplest setting n = 2, namely,

$$\mathcal{S}^2_+ = \left\{ \begin{bmatrix} x & y \\ y & z \end{bmatrix} : x \ge 0, z \ge 0, xz \ge y^2 \right\}.$$

Abusing notation slightly, one can easily verify

$$\mathcal{P}_z(\mathcal{S}^2_+) = \mathbb{R}_+, \qquad \mathcal{P}_y(\mathcal{S}^2_+) \cong \mathbb{R}, \qquad \mathcal{P}_{x,z}(\mathcal{S}^2_+) = \mathbb{R}^2_+.$$

Here \cong refers to an obvious linear isomorphism. Clearly all these projected sets are closed. Projecting S^2_+ onto a single row (and corresponding column), on the other hand, yields a set that is not closed:

$$\mathcal{P}_{x,y}(\mathcal{S}^2_+) = \mathcal{P}_{z,y}(\mathcal{S}^2_+) \cong \{(0,0)\} \cup (\mathbb{R}_{++} \times \mathbb{R}).$$

In this case, the graph G has two vertices and two edges, and in particular, there are edges joining L with L^c . As we will now see, this connectivity property is the only obstacle to $\mathcal{P}(S^n_+)$ being closed. The elementary proof of the following two theorems was suggested to us by one of the anonymous referees, for which we are very grateful. Our original arguments were based on more general principles; see Remark 3.3.

THEOREM 3.1 (closedness of the projected PSD cone). The projected set $\mathcal{P}(\mathcal{S}^n_+)$ is closed if, and only if, the vertices in L are disconnected from those in the complement L^c . Moreover, if the latter condition fails, then for any edge $i^*j^* \in E$ joining a vertex in L with a vertex in L^c , any partial matrix $a \in \mathbb{R}^E$ satisfying

$$a_{i^*j^*} \neq 0$$
 and $a_{ij} = 0$ for all $ij \in E \cap (L \times L)$

lies in $(\operatorname{cl} \mathcal{P}(\mathcal{S}^n_+)) \setminus \mathcal{P}(\mathcal{S}^n_+)$.

Proof. Without loss of generality, assume $L = \{1, \ldots, r\}$ for some integer $r \ge 0$. Suppose first that the vertices in L are disconnected from those in the complement L^c . Let $\{a_i\} \subseteq \mathbb{R}^E$ be a sequence in $\mathcal{P}(\mathcal{S}^n_+)$ converging to a partial matrix $a \in \mathbb{R}^E$. We may now write $a_i = \mathcal{P}(X_i)$ for some matrices $X_i \in \mathcal{S}^n_+$. Denoting by $X_i[L]$ the restriction of X_i to the $L \times L$ block, we deduce that the diagonal elements of $X_i[L]$ are bounded and therefore the matrices $X_i[L]$ are themselves bounded. Hence there exists a subsequence of $X_i[L]$ converging to some PSD matrix X_L . Let $Y \in \mathcal{S}^{|L^c|}$ now be any completion of the restriction of a to L^c . Observe that for sufficiently large values λ the matrix $Y + \lambda I$ is positive definite and hence $\begin{bmatrix} X_L & 0\\ 0 & Y + \lambda I \end{bmatrix}$ is a positive semidefinite completion of a.

Conversely, suppose that L is not disconnected from L^c and consider the vertices i^*, j^* and a matrix a specified in the statement of the theorem. Since the block $\begin{bmatrix} 0 & a_{i^*j^*} \\ a_{i^*j^*} & ? \end{bmatrix}$ is not PSD completable, clearly a is not PSD completable. To see the inclusion $a \in \operatorname{cl} \mathcal{P}(\mathcal{S}^n_+)$, consider the matrix $X := \mathcal{P}^*(a) + \begin{bmatrix} \epsilon I & 0 \\ 0 & \lambda I \end{bmatrix}$. Using Schur's complement, we deduce that for any fixed ϵ there exists a sufficiently large λ such

that X is positive definite. On the other hand, clearly $\mathcal{P}(X)$ converges to a as ϵ tends to zero. This completes the proof. \Box

We next consider projections of the EDM cone.

THEOREM 3.2 (closedness of the projected EDM cone). The projected image $\mathcal{P}(\mathcal{E}^n)$ is always closed.

Proof. First, we can assume without loss of generality that the graph G is connected. To see this, let $G_i = (V_i, E_i)$ for $i = 1, \ldots, l$ be the connected components of G. Then one can easily verify that $\mathcal{P}(\mathcal{E}^n)$ coincides with the Cartesian product $P_{E_1}(\mathcal{E}^{|V_1|}) \times \cdots \times P_{E_l}(\mathcal{E}^{|V_l|})$. Thus if each image $\mathcal{P}_{E_i}(\mathcal{E}^{|V_i|})$ is closed, then so is the product $\mathcal{P}(\mathcal{E}^n)$. We may therefore assume that G is connected. Now suppose that for a sequence $D_i \in \mathcal{E}^n$ the vectors $a_i = \mathcal{P}(D_i)$ converge to some vector $a \in \mathbb{R}^E$. Let x_1^i, \ldots, x_n^i be the point realizing the matrices D_i . Translating the points, we may assume that one of the points is the origin. Since G is connected, all the points x_j^i are bounded in norm. Passing to a subsequence, we obtain a collection of points realizing the matrix a.

Remark 3.3. Theorems 3.1 and 3.2 are part of a broader theme. Indeed, a central (and classical) question in convex analysis is when a linear image of a closed convex cone is itself closed. In a recent paper [27], the author showed that there is a convenient characterization for "nice cones"—those cones C for which $C^* + F^{\perp}$ is closed for all faces $F \leq C$ [5, 27]. Reassuringly, most cones which we can efficiently optimize over are nice; see the discussion in [27]. For example, the cones of positive semidefinite and Euclidean distance matrices are nice. The results of Theorems 3.1 and 3.2 can be entirely recovered from the more general perspective; originally, the content of the aforementioned results were noticed exactly in this way. See version 2 of this paper on arxiv.org and the discussion after Theorem 8 in [28].

4. Boundaries of projected sets and facial reduction. To motivate the discussion, consider the general conic system

$$(4.1) F := \{ X \in C : \mathcal{M}(X) = b \},$$

where C is a proper (closed, with nonempty interior) convex cone in a Euclidean space \mathbb{E} and $\mathcal{M} \colon \mathbb{E} \to \mathbb{R}^m$ is a linear mapping. Classically we say that the Slater condition holds for this problem whenever there exists X in the interior of C satisfying the system $\mathcal{M}(X) = b$. In this section, we first relate properties of the image set $\mathcal{M}(C)$ to the *facial reduction algorithm* of Borwein and Wolkowicz [5, 6], and to more recent variants [40, 29, 30], and then specialize the discussion to the PSD and EDM completion problems we have been studying.

When strict feasibility fails, the *facial reduction* strategy aims to embed the feasibility system in a Euclidean space of smallest possible dimension. The starting point is the following elementary geometric observation: exactly one of the following alternatives holds [5, 6, 29, 30].

- 1. The conic system in (4.1) is strictly feasible.
- 2. There is a nonzero matrix $Y \in C^*$ so that the orthogonal complement Y^{\perp} contains the affine space $\{X : \mathcal{M}(X) = b\}$.

The matrix Y in the second alternative certifies that the entire feasible region F is contained in the slice $C \cap Y^{\perp}$. Determining which alternative is valid is yet another system that needs to be solved, namely, find a vector v satisfying the *auxiliary system*

$$0 \neq \mathcal{M}^* v \in C^*$$
 and $\langle v, b \rangle = 0$,

and set $Y := \mathcal{M}^* v$. One can now form an "equivalent" feasible region to (4.1) by replacing C with $C \cap Y^{\perp}$ and \mathbb{E} with the linear span of $C \cap Y^{\perp}$. One then repeats the procedure on this smaller problem, forming the alternative, and so on and so forth until strict feasibility holds. The number of steps for the procedure to terminate depends on the choices of the exposing vectors Y. The minimal number of steps needed is the *singularity degree*—an intriguing measure of complexity [34]. In general, the singularity degree is no greater than n-1, and there are instances of semidefinite programming that require exactly n-1 facial reduction iterations [39, section 2.6].

The following theorem provides an interesting perspective on facial reduction in terms of the image set $\mathcal{M}(C)$. In essence, the minimal face of $\mathcal{M}(C)$ containing *b* immediately yields the minimal face of *C* containing the feasible region *F*, that is, in principle no auxiliary sequence of problems for determining face(*F*, *C*) is needed. The difficulty is that geometry of $\mathcal{M}(C)$ is in general complex and so a simple description of face($b, \mathcal{M}(C)$) is unavailable. The auxiliary problem in the facial reduction iteration instead tries to represent face($b, \mathcal{M}(C)$) using some dual vector *v* exposing a face of $\mathcal{M}(C)$ containing *b*. The singularity degree is then exactly one if, and only if, the minimal face face($b, \mathcal{M}(C)$) is exposed. To the best of our knowledge, this relationship to exposed faces has either been overlooked in the literature or forgotten. In particular, an immediate consequence is that whenever the image cone $\mathcal{M}(C)$ is facially exposed, the feasibility problem (4.1) has singularity degree at most one for any right-hand-side vector *b*, for which the feasible region is nonempty.

THEOREM 4.1 (facial reduction and exposed faces). Consider a linear operator $\mathcal{M} \colon \mathbb{E} \to \mathbb{Y}$, between two Euclidean spaces \mathbb{E} and \mathbb{Y} , and let $C \subset \mathbb{E}$ be a proper convex cone. Consider a nonempty feasible set

(4.2)
$$F := \{X \in C : \mathcal{M}(X) = b\}$$

for some point $b \in \mathbb{Y}$. Then a vector v exposes a proper face of $\mathcal{M}(C)$ containing b if, and only if, v satisfies the auxiliary system

(4.3)
$$0 \neq \mathcal{M}^* v \in C^* \quad and \quad \langle v, b \rangle = 0.$$

For notational convenience, define $N := face(b, \mathcal{M}(C))$. Then the following are true:

- 1. We always have $C \cap \mathcal{M}^{-1}N = \text{face}(F, C)$.
- 2. For any vector $v \in \mathbb{R}^m$ the following equivalence holds:

$$v \ exposes \ N \iff \mathcal{M}^*v \ exposes \ face(F,C).$$

Consequently whenever the Slater condition fails, the singularity degree of the system (4.2) is equal to one if, and only if, the minimal face face($b, \mathcal{M}(C)$) is exposed.

Proof. First suppose that v exposes a proper face of $\mathcal{M}(C)$ containing b. Clearly we have $\langle v, b \rangle = 0$. Observe moreover

$$\langle \mathcal{M}^* v, X \rangle = \langle v, \mathcal{M}(X) \rangle \ge 0$$
 for any $X \in C$,

and hence the inclusion $\mathcal{M}^* v \in C^*$ holds. Finally, since v exposes a proper face of $\mathcal{M}(C)$, we deduce $v \notin (\operatorname{rge} \mathcal{M})^{\perp} = \ker \mathcal{M}^*$. We conclude that v satisfies the auxiliary system (4.3). The converse follows along the same lines.

We first prove claim 1. To this end, we first verify that $C \cap \mathcal{M}^{-1}N$ is a face of C. Observe for any $x, y \in C$ satisfying $\frac{1}{2}X + \frac{1}{2}Y \in C \cap \mathcal{M}^{-1}N$, we have $\frac{1}{2}\mathcal{M}(X) + \frac{1}{2}\mathcal{M}(X)$

 $\frac{1}{2}\mathcal{M}(Y) \in N$. Since N is a face of $\mathcal{M}(C)$, we deduce $X, Y \in C \cap \mathcal{M}^{-1}N$ as claimed. Now clearly $C \cap \mathcal{M}^{-1}N$ contains F. It is easy to verify now the equality

$$N = \mathcal{M}(C \cap \mathcal{M}^{-1}N).$$

Appealing to [31, Theorem 6.6], we deduce

$$\operatorname{ri} N = \mathcal{M}(\operatorname{ri}(C \cap \mathcal{M}^{-1}N)).$$

Thus b can be written as $\mathcal{M}(X)$ for some $X \in \operatorname{ri}(C \cap \mathcal{M}^{-1}N)$. We deduce that the intersection $F \cap \operatorname{ri}(C \cap \mathcal{M}^{-1}N)$ is nonempty. Appealing to [26, Proposition 2.2(ii)], we obtain the claimed equality $C \cap \mathcal{M}^{-1}N = \operatorname{face}(F, C)$.

Finally we prove 2. To this end, suppose first that a vector v exposes N. Then by what has already been proved, v satisfies the auxiliary system and therefore $C \cap (\mathcal{M}^* v)^{\perp}$ is an exposed face of C containing F. It is standard now to verify equality

(4.4)
$$\mathcal{M}(C \cap (\mathcal{M}^* v)^{\perp}) = \mathcal{M}(C) \cap v^{\perp} = N.$$

Indeed, for any $a \in \mathcal{M}(C) \cap v^{\perp}$, we may write $a = \mathcal{M}(X)$ for some $X \in C$ and consequently deduce $\langle X, \mathcal{M}^* v \rangle = \langle a, v \rangle = 0$. Conversely, for any $X \in C \cap (\mathcal{M}^* v)^{\perp}$, we have $\langle \mathcal{M}(X), v \rangle = \langle X, \mathcal{M}^* v \rangle = 0$, as claimed.

Combining (4.4) with [31, Theorem 6.6], we deduce

$$\operatorname{ri}(N) = \mathcal{M}(\operatorname{ri}(C \cap (\mathcal{M}^* v)^{\perp})).$$

Thus b can be written as $\mathcal{M}(X)$ for some $X \in \mathrm{ri}(C \cap (\mathcal{M}^* v)^{\perp})$. We deduce that the intersection $F \cap \mathrm{ri}(C \cap (\mathcal{M}^* v)^{\perp})$ is nonempty. Appealing to [26, Proposition 2.2(ii)], we conclude that $C \cap (\mathcal{M}^* v)^{\perp}$ is the minimal face of C containing F.

Now conversely suppose that $\mathcal{M}^* v$ exposes face (F, C). Then clearly v exposes a face of $\mathcal{M}(C)$ containing b. On the other hand, by claim 1, we have

$$C \cap \mathcal{M}^{-1}N = \text{face}(F, C) = C \cap (\mathcal{M}^*v)^{\perp}.$$

Hence a point $\mathcal{M}(X)$ with $X \in C$ lies in $\mathcal{M}(C) \cap v^{\perp}$ if, and only if, it satisfies $0 = \langle v, \mathcal{M}(X) \rangle = \langle \mathcal{M}^* v, X \rangle$, which by the above equation is equivalent to the inclusion $\mathcal{M}(X) \in N$. This completes the proof. \square

The following example illustrates Theorem 4.1.

Example 4.2 (singularity degree and facially exposed faces). Consider the cone $C := S^3_+$. Define now the mapping $\mathcal{M} : S^3 \to \mathbb{R}^2$ and the vector $b \in \mathbb{R}^2$ to be

$$\mathcal{M}(X) = \begin{pmatrix} X_{11} \\ X_{33} \end{pmatrix}$$
 and $b = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$.

Then the singularity degree of the system (4.1), namely,

$${X \in \mathcal{S}^3_+ : X_{11} = 1, X_{33} = 0},$$

is one. Indeed $v = (0,1)^T$ is a solution to the auxiliary system, and hence the Slater condition fails. On the other hand, the feasible matrix $X = I - e_3 e_3^T$ shows that the maximal solution rank of the system is two. This also follows immediately from Theorem 4.1 since the image set $\mathcal{M}(\mathcal{S}^3_+) = \mathbb{R}^2_+$ is facially exposed.

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We next slightly change this example by adding a coordinate. Namely, define $\mathcal{M}: S^3 \to \mathbb{R}^3$ and b to be

$$\mathcal{M}(X) = \begin{pmatrix} X_{11} \\ X_{33} \\ X_{22} + X_{13} \end{pmatrix} \quad \text{and} \quad b = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}.$$

Observe that the singularity degree of the system (4.1) is at most two, since it admits a rank one feasible solution $X = e_1 e_1^T$. On the other hand, one can see directly from Theorem 4.1 that the singularity degree is exactly two. Indeed, one easily checks

$$\mathcal{M}(\mathcal{S}^3_+) \,=\, \mathbb{R}^3_+ \cup \{ (x, y, z) \,|\, x \ge 0, \, y \ge 0, \, z \le 0, \, xy \ge z^2 \, \},$$

i.e., we obtain $\mathcal{M}(\mathcal{S}^3_+)$ by taking the union of \mathbb{R}^3_+ with a rotated copy of \mathcal{S}^2_+ . The set $\mathcal{M}(\mathcal{S}^3_+)$ has a nonexposed face which contains *b* in its relative interior—this is easy to see by intersecting $\mathcal{M}(\mathcal{S}^3_+)$ with the hyperplane x = 1 and graphing.

An interesting consequence of Theorem 4.1 above is that it is the lack of facial exposedness of the image set $\mathcal{M}(C)$ that is responsible for a potentially large singularity degree and hence for serious numeric instability, i.e., weak Hölderian error bounds [34].

COROLLARY 4.3 (Hölderian error bounds and facial exposedness). Consider a linear mapping $\mathcal{M}: S^n \to \mathbb{R}^m$ having the property that $\mathcal{M}(S^n_+)$ is facially exposed. For any vector $b \in \mathbb{R}^m$, define the affine space

$$\mathcal{V} = \{ X : \mathcal{M}(X) = b \}.$$

Then the intersection $S^n_+ \cap \mathcal{V}$, when nonempty, admits a $\frac{1}{2}$ -Hölder error bound: for any compact set U, there is a constant c > 0 so that

$$\operatorname{dist}_{\mathcal{S}^n_+ \cap \mathcal{V}}(X) \le c \cdot \sqrt{\max\left\{\operatorname{dist}_{\mathcal{S}^n_+}(X), \operatorname{dist}_{\mathcal{V}}(X)\right\}} \qquad for \ all \ x \in U.$$

Proof. This follows immediately from Theorem 4.1 and [34].

4.1. Facial reduction for completion problems. For those problems with highly structured constraints, one can hope to solve the auxiliary problems directly. For example, the following simple idea can be fruitful: fix a subset $I \subset \{1, \ldots, m\}$ and let $\mathcal{M}_I(X)$ and b_I , respectively, denote restrictions of $\mathcal{M}(X)$ and b to coordinates indexed by I. Consider then the relaxation

$$F_I := \{ X \in C : \mathcal{M}_I(X) = b_I \}.$$

If the index set I is chosen so that the image $\mathcal{M}_I(C)$ is "simple," then we may find the minimal face face(F_I, C), as discussed above. Intersecting such faces for varying index sets I may yield a drastic dimensional decrease. Moreover, observe that this preprocessing step is entirely parallelizable.

Interpreting this technique in the context of matrix completion problems, we recover the Krislock–Wolkowicz algorithm [18]. Namely, note that when \mathcal{M} is simply the projection \mathcal{P} and we set $C = S^n_+$ or $C = \mathcal{E}^n$, we obtain the PSD and EDM completion problems,

$$F := \{ X \in C : \mathcal{P}(X) = a \} = \{ X \in C : X_{ij} = a_{ij} \text{ for all } ij \in E \},\$$

where $a \in \mathbb{R}^E$ is a partial matrix. It is then natural to consider indices $I \subset E$ describing clique edges in the graph since then the images $\mathcal{P}_I(C)$ are the smaller dimensional PSD and EDM cones, respectively—sets that are well understood. This algorithmic strategy becomes increasingly effective when the rank (for the PSD case) or the embedding dimension (for the EDM case) of the specified principal minors are all small. Moreover, we will show that under a chordality assumption, the minimal face of C containing the feasible region is guaranteed to be discovered if all the maximal cliques were to be considered; see Theorems 4.7 and 4.14. This, in part, explains why the EDM completion algorithm of [18] works so well. Understanding the geometry of $\mathcal{P}_I(C)$ for a wider class of index sets I would yield an even better preprocessing strategy. We defer to [18] for extensive numerical results and implementation issues showing that the discussed algorithmic idea is extremely effective for EDM completions.

In what follows, by the term "clique χ in G" we will mean a collection of k pairwise connected vertices of G. The symbol $|\chi|$ will indicate the cardinality of χ (i.e., the number of vertices), while $E(\chi)$ will denote the edge set in the subgraph induced by G on χ . For a partial matrix $a \in \mathbb{R}^E$, the symbol a_{χ} will mean the restriction of ato $E(\chi)$, whereas \mathcal{P}_{χ} will be the projection of \mathcal{S}^n onto $E(\chi)$. The symbol \mathcal{S}^{χ} will indicate the set of $|\chi| \times |\chi|$ symmetric matrices whose rows and columns are indexed by χ . Similar notation will be reserved for \mathcal{S}^{χ}_+ . If χ is contained in L, then we may equivalently think of a_{χ} as a vector lying in $\mathbb{R}^{E(\chi)}$ or as a matrix lying in \mathcal{S}^{χ} . Thus the adjoint \mathcal{P}^*_{χ} assigns to a partial matrix $a_{\chi} \in \mathcal{S}^{\chi}$ an $n \times n$ matrix whose principal submatrix indexed by χ coincides with a_{χ} and whose all other entries are zero.

THEOREM 4.4 (clique facial reduction for PSD completions). Let $\chi \subseteq L$ be any k-clique in the graph G. Let $a \in \mathbb{R}^E$ be a partial PSD matrix and define

$$F_{\chi} := \{ X \in \mathcal{S}^n_+ : X_{ij} = a_{ij} \text{ for all } ij \in E(\chi) \}.$$

Then for any matrix v_{χ} exposing face $(a_{\chi}, \mathcal{S}^{\chi}_{+})$, the matrix

 $\mathcal{P}^*_{\chi} v_{\chi} \quad exposes \quad face(F_{\chi}, \mathcal{S}^n_+).$

Proof. Simply apply Theorem 4.1 with $C = \mathcal{S}^n_+$, $\mathcal{M} = \mathcal{P}_{\chi}$, and $b = a_{\chi}$.

Theorem 4.4 is transparent and easy. Consequently it is natural to ask whether the minimal face of S^n_+ containing the feasible region of a PSD completion problem can be found using solely faces arising from cliques, that is those faces described in Theorem 4.4. The answer is no in general: the following example exhibits a PSD completion problem that fails the Slater condition but for which all specified principal submatrices are definite, and hence all faces arising from Theorem 4.4 are trivial.

Example 4.5 (Slater condition and nonchordal graphs). Let G = (V, E) be a cycle on four vertices with each vertex attached to a loop, that is, $V = \{1, 2, 3, 4\}$ and $E = \{12, 23, 34, 14\} \cup \{11, 22, 33, 44\}$. Define the following PSD completion problems $C(\epsilon)$, parametrized by $\epsilon \ge 0$:

$$C(\epsilon): \begin{bmatrix} 1+\epsilon & 1 & ? & -1\\ 1 & 1+\epsilon & 1 & ?\\ ? & 1 & 1+\epsilon & 1\\ -1 & ? & 1 & 1+\epsilon \end{bmatrix}.$$

Let $a(\epsilon) \in \mathbb{R}^E$ denote the corresponding partial matrices. According to [12, Lemma 6] there is a unique positive semidefinite matrix A satisfying $A_{ij} = 1$ for all $|i-j| \leq 1$,

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namely, the matrix of all 1's. (We will prove a generalization of this result shortly in Corollary 4.11.) Hence the PSD completion problem C(0) is infeasible, that is, a(0) lies outside of $\mathcal{P}(\mathcal{S}^4_+)$. On the other hand, for all sufficiently large ϵ , the partial matrices $a(\epsilon)$ do lie in $\mathcal{P}(\mathcal{S}^4_+)$ due to the diagonal dominance. Taking into account that $\mathcal{P}(\mathcal{S}^4_+)$ is closed (by Theorem 3.1), we deduce that there exists $\hat{\epsilon} > 0$, so that $a(\hat{\epsilon})$ lies on the boundary of $\mathcal{P}(\mathcal{S}^4_+)$, that is, the Slater condition fails for the completion problem $C(\hat{\epsilon})$. On the other hand, $a(\epsilon)$ are clearly partial PD matrices for all $\epsilon > 0$. Hence $a(\hat{\epsilon})$ is a partial PD matrix and the faces arising from Theorem 4.4 are trivial. In light of this observation, consider solving the semidefinite program

(4.5)
$$\min \epsilon \\ \text{s.t.} \begin{bmatrix} 1+\epsilon & 1 & \alpha & -1 \\ 1 & 1+\epsilon & 1 & \beta \\ \alpha & 1 & 1+\epsilon & 1 \\ -1 & \beta & 1 & 1+\epsilon \end{bmatrix} \succeq 0.$$

Doing so, we deduce that $\hat{\epsilon} = \sqrt{2} - 1$, $\hat{\alpha} = \hat{\beta} = 0$ is optimal. Formally, we can verify this by finding the dual of (4.5) and checking feasibility and complementary slackness for the primal-dual optimal pair \hat{X} and \hat{Z} ,

$$\widehat{X} = \begin{bmatrix} \sqrt{2} & 1 & 0 & -1 \\ 1 & \sqrt{2} & 1 & 0 \\ 0 & 1 & \sqrt{2} & 1 \\ -1 & 0 & 1 & \sqrt{2} \end{bmatrix}, \qquad \widehat{Z} = \frac{1}{4} \begin{bmatrix} 1 & -\frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & 1 & -\frac{1}{\sqrt{2}} & 0 \\ 0 & -\frac{1}{\sqrt{2}} & 1 & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & 0 & -\frac{1}{\sqrt{2}} & 1 \end{bmatrix}.$$

Despite this pathological example, we now show that at least for chordal graphs, the minimal face of the PSD completion problem can be found solely from faces corresponding to cliques in the graph. We begin with the following simple lemma.

LEMMA 4.6 (maximal rank completions). Suppose without loss of generality $L = \{1, \ldots, r\}$ and let $G_L := (L, E_L)$ be the graph induced on L by G. Let $a \in \mathbb{R}^E$ be a partial matrix and a_{E_L} the restriction of a to E_L . Suppose that $X_L \in S^r_+$ is a maximum rank PSD completion of a_{E_L} and

$$X = \begin{bmatrix} A & B \\ B^T & C \end{bmatrix}$$

is an arbitrary PSD completion of a. Then

$$X_{\mu} := \begin{bmatrix} X_L & B \\ B^T & C + \mu I \end{bmatrix}$$

is a maximal rank PSD completion of $a \in \mathbb{R}^E$ for all sufficiently large μ .

Proof. We construct the maximal rank PSD completion from the arbitrary PSD completion X by moving from A to X_L and from C to $C + \mu I$ while staying in the same minimal face for the completions. To this end, define the sets

$$F = \{X \in \mathcal{S}_{+}^{n} : X_{ij} = a_{ij} \text{ for all } ij \in E\},\$$

$$F_{L} = \{X \in \mathcal{S}_{+}^{r} : X_{ij} = a_{ij} \text{ for all } ij \in E_{L}\},\$$

$$\widehat{F} = \{X \in \mathcal{S}_{+}^{n} : X_{ij} = a_{ij} \text{ for all } ij \in E_{L}\}.$$

Then X_L is a maximum rank PSD matrix in F_L . Observe that the rank of any PSD matrix $\begin{bmatrix} P & Q \\ Q^T & R \end{bmatrix}$ is bounded by rank P + rank R. Consequently the rank of any PSD

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matrix in F and also in \widehat{F} is bounded by rank $X_L + (n-r)$, and the matrix

$$\bar{X} = \begin{bmatrix} X_L & 0\\ 0 & I \end{bmatrix}$$

has maximal rank in \widehat{F} , i.e.,

(4.6)
$$\bar{X} \in \operatorname{ri}(\hat{F}).$$

Let U be a matrix of eigenvectors of X_L , with eigenvectors corresponding to 0 eigenvalues coming first. Then

$$U^T X_L U = \begin{bmatrix} 0 & 0 \\ 0 & \Lambda \end{bmatrix},$$

where $0\prec\Lambda\in\mathcal{S}^k_+$ is a diagonal matrix with all positive diagonal elements. Define

$$Q = \begin{bmatrix} U & 0 \\ 0 & I \end{bmatrix}.$$

Let X be as in the statement of the lemma; then clearly $X \in \widehat{F}$ and we deduce using (4.6) that

(4.7)
$$\bar{X} \pm \epsilon(\bar{X} - X) \in \mathcal{S}^n_+ \Leftrightarrow Q^T \bar{X} Q \pm \epsilon Q^T (\bar{X} - X) Q \in \mathcal{S}^n_+$$

for some small $\epsilon > 0$. We now have

$$Q^{T}\bar{X}Q = \begin{bmatrix} U^{T}X_{L}U & 0\\ 0 & I \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0\\ 0 & \Lambda & 0\\ 0 & 0 & I \end{bmatrix},$$
$$Q^{T}XQ = \begin{bmatrix} U^{T}AU & U^{T}B\\ B^{T}U & C \end{bmatrix} = \begin{bmatrix} V_{11} & V_{12} & V_{13}\\ V_{12}^{T} & V_{22} & V_{23}\\ V_{13}^{T} & V_{23}^{T} & V_{33} \end{bmatrix},$$

where $V_{11} \in S^{r-k}$, $V_{22} \in S^k$, $V_{33} \in S^{n-r}$. From (4.7) we deduce $V_{11} = 0$, $V_{12} = 0$, $V_{13} = 0$. Therefore

$$Q^{T}X_{\mu}Q = \begin{bmatrix} U^{T}X_{L}U & U^{T}B\\ B^{T}U & \mu I + C \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0\\ 0 & \Lambda & V_{23}\\ 0 & V_{23}^{T} & \mu I + C \end{bmatrix}.$$

By the Schur complement condition for positive semidefiniteness we have that for sufficiently large μ the matrix X_{μ} is PSD, and rank $X_{\mu} = \operatorname{rank} X_L + (n-r)$; hence it is a maximal rank PSD matrix in F.

THEOREM 4.7 (finding the minimal face on chordal graphs). Suppose that the graph induced by G on L is chordal. Consider a partial PSD matrix $a \in \mathbb{R}^E$ and the region

$$F = \{ X \in \mathcal{S}^n_+ : X_{ij} = a_{ij} \text{ for all } ij \in E \}.$$

Then the equality

$$face(F, \mathcal{S}^n_+) = \bigcap_{\chi \in \Theta} face(F_{\chi}, \mathcal{S}^n_+) \qquad holds,$$

where Θ denotes the set of all maximal cliques in the restriction of G to L, and for each $\chi \in \Theta$ we define the relaxation

$$F_{\chi} := \{ X \in \mathcal{S}^n_+ : X_{ij} = a_{ij} \text{ for all } ij \in E(\chi) \}.$$

Proof. For brevity, set

$$H = \bigcap_{\chi \in \Theta} \text{face}(F_{\chi}, \mathcal{S}^n_+)$$

We first prove the theorem under the assumption that L is disconnected from L^c . To this end, for each clique $\chi \in \Theta$, let $v_{\chi} \in S^{\chi}_+$ denote the exposing vector of face (a_{χ}, S^{χ}_+) . Then by Theorem 4.4, we have

$$\operatorname{face}(F_{\chi}, \mathcal{S}^n_+) = \mathcal{S}^n_+ \cap (\mathcal{P}^*_{\chi} v_{\chi})^{\perp}.$$

It is straightforward to see that $\mathcal{P}_{\chi}^* v_{\chi}$ is simply the $n \times n$ matrix whose principal submatrix indexed by χ coincides with v_{χ} and whose other entries are all zero. Letting $Y[\chi]$ denote the principal submatrix indexed by χ of any matrix $Y \in \mathcal{S}_{+}^{n}$, we successively deduce

$$\mathcal{P}(H) = \mathcal{P}\Big(\{Y \succeq 0 : Y[\chi] \in v_{\chi}^{\perp} \text{ for all } \chi \in \Theta\}\Big)$$
$$= \mathcal{P}(\mathcal{S}^n_+) \cap \{b \in \mathbb{R}^E : b_{\chi} \in v_{\chi}^{\perp} \text{ for all } \chi \in \Theta\}.$$

On the other hand, since the restriction of G to L is chordal and L is disconnected from L^c , Theorem 2.1 implies that G is PSD completable. Hence we have the representation $\mathcal{P}(\mathcal{S}^n_+) = \{b \in \mathbb{R}^E : b_{\chi} \in \mathcal{S}^{\chi}_+ \text{ for all } \chi \in \Theta\}$. Combining this with the equations above, we obtain

$$\mathcal{P}(H) = \{ b \in \mathbb{R}^E : b_{\chi} \in \mathcal{S}^{\chi}_+ \cap v_{\chi}^{\perp} \text{ for all } \chi \in \Theta \}$$
$$= \{ b \in \mathbb{R}^E : b_{\chi} \in \text{face}(a_{\chi}, \mathcal{S}^{\chi}_+) \text{ for all } \chi \in \Theta \}$$
$$= \bigcap_{\chi \in \Theta} \{ b \in \mathbb{R}^E : b_{\chi} \in \text{face}(a_{\chi}, \mathcal{S}^{\chi}_+) \}.$$

Clearly a lies in the relative interior of each set $\{b \in \mathbb{R}^E : b_{\chi} \in \text{face}(a_{\chi}, \mathcal{S}^{\chi}_+)\}$. Using [31, Theorems 6.5, 6.6], we deduce

$$a \in \operatorname{ri} \mathcal{P}(H) = \mathcal{P}(\operatorname{ri} H)$$

Thus the intersection $F \cap \operatorname{ri} H$ is nonempty. Taking into account that F is contained in H, and appealing to [26, Proposition 2.2(ii)], we conclude that H is the minimal face of S^n_+ containing F, as claimed.

We now prove the theorem in full generality, that is, when there may exist an edge joining L and L^c . To this end, let $\hat{G}_L = (V, E_L)$ be the graph obtained from G by deleting all edges adjacent to L^c . Clearly, L and L^c are disconnected in \hat{G}_L . Applying the special case of the theorem that we have just proved, we deduce that in terms of the set

$$\overline{F} = \{ X \in \mathcal{S}^n_+ : X_{ij} = a_{ij} \text{ for all } ij \in E_L \},\$$

we have

$$face(F, \mathcal{S}^n_+) = H$$

The X_{μ} matrix of Lemma 4.6 is a maximum rank PSD matrix in F, and also in \widehat{F} . Since $F \subseteq \widehat{F}$, we deduce face $(F, \mathcal{S}^n_+) = \text{face}(\widehat{F}, \mathcal{S}^n_+)$, and this completes the proof. \square

Remark 4.8 (finding maximal cliques on chordal graphs). In light of the theorem above, it is noteworthy that finding maximal cliques on chordal graphs is polynomially solvable; see, e.g., [23] or more generally [32].

The following is an immediate consequence.

COROLLARY 4.9 (singularity degree of chordal completions). If the restriction of G to L is chordal, then the PSD completion problem has singularity degree at most one.

Proof. In the notation of Theorem 4.7, the sum $Y := \sum_{\chi \in \Theta} \mathcal{P}_{\chi}^* v_{\chi}$ exposes the minimal face face (F, \mathcal{S}_n^+) . If the Slater condition fails, then Y is feasible for the first auxiliary problem in the facial reduction sequence. \square

Example 4.10 (finding the minimal face on chordal graphs). Let Ω consist of all matrices $X \in S^4_+$ solving the PSD completion problem

$$\begin{bmatrix} 1 & 1 & ? & ? \\ 1 & 1 & 1 & ? \\ ? & 1 & 1 & -1 \\ ? & ? & -1 & 2 \end{bmatrix}.$$

There are three nontrivial cliques in the graph. Observe that the minimal face of S^2_+ containing the matrix

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} -\frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

is exposed by

$$\begin{bmatrix} -\frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 4 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} -\frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} = \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}.$$

Moreover, the matrix $\begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix}$ is definite and hence the minimal face of S^2_+ containing this matrix is exposed by the all-zero matrix.

Classically, an intersection of exposed faces is exposed by the sum of their exposing vectors. Using Theorem 4.7, we deduce that the minimal face of S^4_+ containing Ω is the one exposed by the sum

Diagonalizing this matrix, we obtain

face(
$$\Omega, \mathcal{S}^4_+$$
) = $\begin{bmatrix} 0 & 1 \\ 0 & 1 \\ 0 & 1 \\ 3 & 0 \end{bmatrix} \mathcal{S}^2_+ \begin{bmatrix} 0 & 1 \\ 0 & 1 \\ 0 & 1 \\ 3 & 0 \end{bmatrix}^T$.

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The following is an interesting consequence of Theorem 4.4, generalizing [12, Lemma 6]. The latter was already used in Example 4.5, and here we give an elementary and self-contained proof using the developed techniques.

COROLLARY 4.11 (tridiagonal matrices with rank one blocks). Consider a path G = (V, E) with a loop attached to every vertex, that is, $E = \{ij : |i-j| \leq 1\}$. Fix any partial PSD matrix $a \in \mathbb{R}^E$, with each specified principal submatrix having rank one, and satisfying $a_{ii} \neq 0$ for each index i. Then $a \in \mathbb{R}^E$ has a unique PSD completion, which must also have rank one.

Proof. Fix a partial PSD matrix $a \in \mathbb{R}^E$ satisfying the assumed properties. Consider the 2 × 2 specified principal submatrix of a given by $a(k) := \begin{bmatrix} a_{k,k} & a_{k,k+1} \\ a_{k,k+1} & a_{k+1,k+1} \end{bmatrix}$. Since this matrix has rank one, there is a vector $0 \neq v_k \in \mathbb{R}^2$ such that $a(k) \in (v_k v_k^T)^{\perp}$. Define now the matrix

$$V_k := \begin{bmatrix} 0 & 0 & 0 \\ 0 & v_k v_k^T & 0 \\ 0 & 0 & 0 \end{bmatrix},$$

with the upper left diagonal block of order k - 1. By Theorem 4.4 all the PSD completions of a lie in the face of S^n_+ exposed by V_k . Since a has no zeros on the diagonal, both coordinates of v_k are nonzero. In particular, this implies that for each index kthe vector $[0, v_k^T, 0]^T$, in which the length of the first 0 block is k, lies outside of the range of the sum $\sum_{i=1}^{k-1} V_i$. By induction then we deduce that the sum $V = \sum_{k=1}^{n-1} V_k$ has rank n-1. Since all PSD completions of a lie in the face of S^n_+ exposed by V, we conclude that there is a unique completion of a and it has rank one, as claimed.

We now turn to an analogous development for the EDM completion problem. To this end, recall from (2.4) that the mapping $\mathcal{K}: \mathcal{S}^n \to \mathcal{S}^n$ restricts to an isomorphism $\mathcal{K}: \mathcal{S}_c \to \mathcal{S}_H$ carrying $\mathcal{S}_c \cap \mathcal{S}_+^n$ onto \mathcal{E}^n . Moreover, it turns out that the Moore– Penrose pseudoinverse \mathcal{K}^{\dagger} restricts to the inverse of this isomorphism $\mathcal{K}^{\dagger}: \mathcal{S}_H \to \mathcal{S}_c$. As a result, it is convenient to study the faces of \mathcal{E}^n using the faces of $\mathcal{S}_c \cap \mathcal{S}_+^n$. This is elucidated by the following standard result.

LEMMA 4.12 (faces under isomorphism). Consider a linear isomorphism $\mathcal{M} \colon \mathbb{E} \to \mathbb{Y}$ between linear spaces \mathbb{E} and \mathbb{Y} , and let $C \subset \mathbb{E}$ be a closed convex cone. Then the following are true:

1. $F \trianglelefteq C \iff \mathcal{M}F \trianglelefteq \mathcal{M}C.$

2. $(\mathcal{M}C)^* = (\mathcal{M}^{-1})^* C^*$.

3. For any face $F \trianglelefteq C$, we have $(\mathcal{M}F)^{\bigtriangleup} = (\mathcal{M}^{-1})^* F^{\bigtriangleup}$.

In turn, it is easy to see that $\mathcal{S}_c \cap \mathcal{S}_+^n$ is a face of \mathcal{S}_+^n isomorphic to \mathcal{S}_+^{n-1} . More specifically, for any $n \times n$ orthogonal matrix $\begin{bmatrix} \frac{1}{\sqrt{n}}e & U \end{bmatrix}$, we have the representation

$$\mathcal{S}_c \cap \mathcal{S}^n_+ = U \mathcal{S}^{n-1}_+ U$$

Consequently, with respect to the ambient space S_c , the cone $S_c \cap S^n_+$ is self-dual and for any face $F \leq S^{n-1}_+$ we have

$$UFU^T \leq \mathcal{S}_c \cap \mathcal{S}^n_{\perp}$$
 and $(UFU^T)^{\triangle} = UF^{\triangle}U^T.$

As a result of these observations, we make the following important convention: the *ambient spaces* of $S_c \cap S^n_+$ and of \mathcal{E}^n will always be taken as S_c and S_H , respectively. Thus the facial conjugacy operations of these two cones will always be taken with respect to these ambient spaces and *not* with respect to the entire S^n .

Given a clique χ in G, we let \mathcal{E}^{χ} denote the set of $|\chi| \times |\chi|$ Euclidean distance matrices indexed by χ . In what follows, given a partial matrix $a \in \mathbb{R}^E$, the restriction a_{χ} can then be thought of either as a vector in $\mathbb{R}^{E(\chi)}$ or as a hollow matrix in \mathcal{S}^{χ} . We will also use the symbol $\mathcal{K}_{\chi} : \mathcal{S}^{\chi} \to \mathcal{S}^{\chi}$ to indicate the mapping \mathcal{K} acting on \mathcal{S}^{χ} .

THEOREM 4.13 (clique facial reduction for EDM completions). Let χ be any k-clique in the graph G. Let $a \in \mathbb{R}^E$ be a partial EDM and define

$$F_{\chi} := \{ X \in \mathcal{S}^n_+ \cap \mathcal{S}_c : [\mathcal{K}(X)]_{ij} = a_{ij} \text{ for all } ij \in E(\chi) \}.$$

Then for any matrix v_{χ} exposing face $(\mathcal{K}^{\dagger}(a_{\chi}), \mathcal{S}^{\chi}_{+} \cap \mathcal{S}_{c})$, the matrix

$$\mathcal{P}^*_{\chi} v_{\chi}$$
 exposes face $(F, \mathcal{S}^n_+ \cap \mathcal{S}_c)$.

Proof. The proof proceeds by applying Theorem 4.1 with

$$C := \mathcal{S}^n_+ \cap \mathcal{S}_c, \quad \mathcal{M} := P_\chi \circ \mathcal{K}, \quad b := a_\chi$$

To this end, first observe $\mathcal{M}(C) = (P_{\chi} \circ \mathcal{K})(\mathcal{S}^n_+ \cap \mathcal{S}_c) = \mathcal{E}^{\chi}$. By Lemma 4.12, the matrix $\mathcal{K}^{\dagger *}_{\chi}(v_{\chi})$ exposes face $(a_{\chi}, \mathcal{E}^{\chi})$. Thus the minimal face of $\mathcal{S}^n_+ \cap \mathcal{S}_c$ containing F is the one exposed by the matrix

$$(P_{\chi} \circ \mathcal{K})^*(\mathcal{K}_{\chi}^{\dagger *}(v_{\chi})) = \mathcal{K}^* P_{\chi}^* \mathcal{K}_{\chi}^{\dagger *}(v_{\chi}) = P_{\chi}^* \mathcal{K}_{\chi}^* \mathcal{K}_{\chi}^{\dagger *}(v_{\chi}) = P_{\chi}^* v_{\chi}.$$

The result follows. \Box

THEOREM 4.14 (clique facial reduction for EDM is sufficient). Suppose that G is chordal, and consider a partial EDM $a \in \mathbb{R}^E$ and the region

$$F := \{ X \in \mathcal{S}_c \cap \mathcal{S}^n_+ : [\mathcal{K}(X)]_{ij} = a_{ij} \text{ for all } ij \in E \}$$

Let Θ denote the set of all maximal cliques in G, and for each $\chi \in \Theta$ define

$$F_{\chi} := \{ X \in \mathcal{S}_c \cap \mathcal{S}_+^n : [\mathcal{K}(X)]_{ij} = a_{ij} \text{ for all } ij \in E(\chi) \}.$$

Then the equality

$$face(F, \mathcal{S}_c \cap \mathcal{S}_+^n) = \bigcap_{\chi \in \Theta} face(F_{\chi}, \mathcal{S}_c \cap \mathcal{S}_+^n) \qquad holds$$

Proof. The proof follows entirely along the same lines as the first part of the proof of Theorem 4.7. We omit the details for the sake of brevity. \Box

COROLLARY 4.15 (singularity degree of chordal completions). If the graph G = (V, E) is chordal, then the EDM completion problem has singularity degree at most one, when feasible.

Conclusion. In this manuscript, we considered properties of the coordinate shadows of the PSD and EDM cones: $\mathcal{P}(\mathcal{S}^n_+)$ and $\mathcal{P}(\mathcal{E})$. We characterized when these sets are closed, related their boundary structure to a facial reduction algorithm of [18], and explained that the nonexposed faces of these sets directly impact the complexity of the facial reduction algorithm. In particular, under a chordality assumption, the "minimal face" of the feasible region admits a combinatorial description, the singularity degree of the completion problems are at most one, and the coordinate shadows, $\mathcal{P}(\mathcal{S}^n_+)$ and $\mathcal{P}(\mathcal{E})$, are facially exposed. This brings up an intriguing follow-up research agenda:

Classify graphs G for which the images $\mathcal{P}(\mathcal{S}^n_+)$ and $\mathcal{P}(\mathcal{E})$ are facially exposed, or equivalently those for which the corresponding completion formulations have singularity degree at most one irrespective of the known matrix entries.

Acknowledgments. We thank the anonymous referees for their insightful comments. We are indebted to one of the referees for pointing out the elementary proofs of Theorems 3.1 and 3.2 presented in the beginning of section 3.

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