Strict Complementarity in MaxCut SDP

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Abstract. The MaxCut SDP is one of the most well-known semidefinite programs, and it has many favorable properties. One of its nicest geometric/duality properties is the fact that the vertices of its feasible region correspond exactly to the cuts of a graph, as proved by Laurent and Poljak in 1995. Recall that a boundary point \( x \) of a convex set \( C \) is called a vertex of \( C \) if the normal cone of \( C \) at \( x \) is full-dimensional.

We study how often strict complementarity holds or fails for the MaxCut SDP when a vertex of the feasible region is optimal, i.e., when the SDP relaxation is tight. While strict complementarity is known to hold when the objective function is in the interior of the normal cone at any vertex, we prove that it fails generically at the boundary of such normal cone. In this regard, the MaxCut SDP displays the nastiest behavior possible for a convex optimization problem.

We also study strict complementarity with respect to two classes of objective functions. We show that, when the objective functions are sampled uniformly from the negative semidefinite rank-one matrices in the boundary of the normal cone at any vertex, the probability that strict complementarity holds lies in \((0,1)\). We also extend a construction due to Laurent and Poljak of weighted Laplacian matrices for which strict complementarity fails. Their construction works for complete graphs, and we extend it to cosums of graphs under some mild conditions.

1. Introduction

Complementary slackness is a fundamental optimality condition, and hence ubiquitous in optimization. In the most general setting of nonlinear programming (formulated below in the classical language of nonlinear optimization), it requires a pair \((\bar{x}, \bar{y})\) of primal-dual feasible solutions to an optimization problem

\[
\min \{ f(x) : g_i(x) \geq 0 \forall i \in \{1, \ldots, m\} \}
\]

and its (Lagrangean) dual to satisfy \( \bar{y}_i g_i(\bar{x}) = 0 \) for every \( i \in [m] \) \( := \{1, \ldots, m\} \); that is, at least one of the feasibility conditions \( g_i(\bar{x}) \geq 0 \) (in the primal) and \( \bar{y}_i \geq 0 \) (in the dual) must be tight, i.e., they cannot both have a slack. In this case, we say that \((\bar{x}, \bar{y})\) is complementary. This condition can be stated very conveniently in structured convex optimization. For a linear program (LP)

\[
\max \{ c^T x : Ax = b, \ x \geq 0 \}
\]

and its dual \( \min \{ b^T y : s = A^T y - c, \ s \geq 0 \} \), where \( A \in \mathbb{R}^{m \times n} \) is a matrix, and \( b \in \mathbb{R}^m \) and \( c \in \mathbb{R}^n \) are vectors, a pair \((\bar{x}, \bar{y} + \bar{s})\) of primal-dual feasible solutions is complementary if \( \bar{x}_i \bar{s}_i = 0 \) for every \( i \in [n] \). One can similarly consider a semidefinite program (SDP)

\[
\max \{ \text{Tr}(CX) : A(X) = b, \ X \succeq 0 \}
\]

defined on symmetric \( n \)-by-\( n \) matrices. Here, as usual, we equip the space \( \mathbb{S}^n \) of symmetric \( n \)-by-\( n \) matrices with the trace inner-product \( \langle C, X \rangle := \text{Tr}(CX^T) = \sum_{i,j} C_{ij} X_{ij} \), the map \( A : \mathbb{S}^n \to \mathbb{R}^m \) is linear, and \( X \succeq 0 \) denotes that \( X \in \mathbb{S}^n \) is positive semidefinite; most of our notation can be found in Tables 1 to 5.

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Strict complementarity is a refinement of the notion of complementary slackness where we require \textit{precisely one} of the feasibility conditions involved to be tight, which forces the other one to have a slack. A pair \((\bar{x}, \bar{y})\) of primal-dual feasible solutions for the optimization problem in (1) and its dual is \textit{strictly complementary} if \(\bar{y}_ig_i(\bar{x}) = 0\) and \(\bar{y}_i + g_i(\bar{x}) > 0\) for every \(i \in [m]\). A pair \((\bar{x}, \bar{y} \oplus \bar{s})\) of primal-dual feasible solutions for the LP in (2) and its dual is \textit{strictly complementary} if \(\bar{x}_i\bar{s}_i = 0\) and \(\bar{x}_i + \bar{s}_i > 0\) for every \(i \in [n]\). Finally, a pair \((\bar{X}, \bar{Y} \oplus \bar{S})\) of primal-dual feasible solutions for the SDP in (3) and its dual is \textit{strictly complementary} if \(\bar{X}\bar{S} = 0\) and \(\bar{X} + \bar{S} > 0\), i.e., \(\bar{X} + \bar{S}\) is positive definite. The latter two notions can be neatly unified in the context of convex conic optimization via the concept of faces (see [17]).

Complementary slackness characterizes optimality whenever Strong Duality holds, in both LPs and SDPs: a primal-dual pair of feasible solutions is optimal if and only if it is complementary. This is sometimes described by saying that \textit{complementary slackness holds} for the (primal-dual pair of) programs. In the case of LPs, whenever primal and dual are both feasible, there exists a primal-dual pair of optimal solutions that is strictly complementary [7]; i.e., \textit{strict complementarity holds} for every primal-dual pair of feasible LPs. However, there exist primal-dual pairs of SDPs (which satisfy strong regularity conditions sufficient for SDP Strong Duality) that have no strictly complementary primal-dual pair of optimal solutions (see [22]); in such cases, we say that \textit{strict complementarity fails} for said primal-dual pair of SDPs. In fact, failure of strict complementarity is deeply related to failure of Strong Duality in the context of convex conic optimization [23].

Existence of a strictly complementary pair of optimal solutions for an SDP is crucial for some key properties of interior-point methods used to solve such an optimization problem; see, e.g., [2, 10, 15, 11] for superlinear convergence and [8] for convergence of the central path to the analytic center of the optimal face. Strict complementarity is also very useful in the identification of optimal faces (in the primal and dual problems), for detection of infeasibility and unboundedness as well as efficient recovery of certificates of these [25, 16]. Hence, it is important to determine whether strict complementarity holds for a given SDP.

It is known that strict complementarity holds generically for SDPs [1]; for a generalization to convex optimization problems, see [18]. However, there are some generic properties of LPs that fail in some natural, highly structured formulations arising in combinatorial optimization. For instance, whereas systems of linear inequalities are well-known to be generically nondegenerate, the natural description of many classical polytopes is degenerate (e.g., for the matching polytope, see [21, Theorem 25.4]), and “...most real-world LP problems are degenerate” according to [24]. Thus, one ought to be careful about strict complementarity when approaching combinatorial optimization problems via SDP relaxations.

In this paper, we study how often strict complementarity holds or fails for the MaxCut SDP and its dual, when an optimal solution of the primal occurs at a vertex of its feasible region. Recall that the \textit{MaxCut problem} for a given graph \(G = (V, E)\) on \(V = [n]\) and weight function \(w: E \to \mathbb{R}\) can be cast as the optimization problem \(\max\{x^TCx : x \in \{\pm 1\}^n\}\), where \(C \in \mathbb{S}^n\) is defined as

\[
4C := \mathcal{L}_G(w) := \sum_{(i,j) \in E} w_{i,j}(e_i - e_j)(e_i - e_j)^T
\]

and \(\{e_1, \ldots, e_n\}\) is the standard basis of \(\mathbb{R}^n\). The matrix \(\mathcal{L}_G(w)\) is known as a (weighted) Laplacian matrix of \(G\), and it is simple to check that \(\mathcal{L}_G(w) \succeq 0\) if \(w \succeq 0\). The natural SDP relaxation for this problem is the following \textit{MaxCut SDP}, which we write along with its dual:

\[
\begin{align*}
\max \quad & \quad \text{Tr}(CX) \\
\text{subject to} \quad & \quad \text{diag}(X) = \mathbb{1}, \quad S = \text{Diag}(y) - C, \quad X \succeq 0, \quad S \succeq 0;
\end{align*}
\]

here, \(\text{diag}: \mathbb{S}^n \to \mathbb{R}^n\) extracts the diagonal, \(\text{Diag}: \mathbb{R}^n \to \mathbb{S}^n\) is the adjoint of diag, and \(\mathbb{1}\) is the vector of all-ones. Strong Duality holds for every \(C \in \mathbb{S}^n\) since both SDPs have Slater points, i.e., feasible solutions that are positive definite.

The feasible region of the MaxCut SDP, called the \textit{elliptope} and denoted by \(\mathcal{E}_n\), is a compact convex set in \(\mathbb{S}^n\) and its vertices are precisely its elements that are rank-one matrices [13], i.e., matrices of the form \(xx^T\) with \(x \in \{\pm 1\}^n\). Thus, they correspond precisely to the \textit{exact} solutions of the MaxCut problem, for which the SDP is a relaxation. The vertices of \(\mathcal{E}_n\) are by definition the points of \(\mathcal{E}_n\) whose normal cones are full-dimensional (we postpone the definition of normal cone to Section 2.2). It is known [3] that strict complementarity holds in (5) precisely when \(C\) lies in the relative interior of the normal cone of \textit{some} \(X \in \mathcal{E}_n\). In particular, if \(\bar{X}\) is a vertex of \(\mathcal{E}_n\), then strict complementarity holds for (5) whenever \(C\) is in the interior.
of the normal cone of $\mathcal{E}_n$ at $\bar{X}$. However, when $C$ lies in the boundary of this normal cone, it is not clear whether strict complementarity holds.

In this paper, we prove that, when $C$ is chosen from the boundary of the normal cone at a vertex of the elliptope, strict complementarity almost always fails for (5); in this regard, surprisingly, the MaxCut SDP displays the worst possible behavior for a convex optimization problem. In order to make the statement “almost always fails” rigorous, we shall make use of Hausdorff measures. However, our treatment is self-contained and it does not require in-depth knowledge of the theory of Hausdorff measures.

We also focus on two classes of objective functions for (5). We prove that, when $C$ is sampled uniformly from (a normalization of) the negative semidefinite rank-one matrices in the normal cone at a vertex of the elliptope, the probability that strict complementarity fails for (5) is in $(0,1)$. Naturally, we shall also use Hausdorff measures to achieve this. Finally, we also extend a construction due to Laurent and Poljak [14], who proved that strict complementarity may fail for (5) when $C$ is a weighted Laplacian matrix. Their construction works for complete graphs, and we extend it to graphs which are cosums where one of the summands is connected and with some mild condition relating the maximum eigenvalues of their Laplacians.

The order in which our results are presented is different from what we described above. Since the weighted Laplacian construction generalized from Laurent and Poljak involves only matrix analysis and spectral graph theory, and no measure theory, we start with that result. Only then we shall delve into measure theory tools to prove the other results. Hence, the rest of this paper is organized as follows. Section 2 contains some preliminaries, such as notation and background results about the MaxCut SDP (5). In Section 3 we discuss failure of strict complementarity for (5) using previous results by Laurent and Poljak and we extend their Laplacian construction to cosums of graphs. In Section 4, we develop some Hausdorff measure basics and use them to prove that strict complementarity fails generically (“almost everywhere”) for the MaxCut SDP (5) when the objective function is in the boundary of the normal cone of a vertex of the elliptope. Finally, in Section 5, we zoom into the set of negative semidefinite rank-one matrices in the latter boundary, and prove that in this case the probability that strict complementarity holds is in $(0,1)$.

2. Preliminaries

We refer the reader to Tables 1 to 5 for our mostly standard notation and terminology. In order to treat $\mathbb{R}^n$ and $\mathbb{S}^n$ uniformly, we adopt the language of Euclidean spaces, i.e., finite-dimensional real vectors spaces equipped with an inner product. We denote arbitrary Euclidean spaces by $E$ and $Y$. We adopt Minkowski’s notation; for instance, $\mathcal{C} + \Lambda \mathcal{D} := \{ x + \lambda y : x \in \mathcal{C}, \lambda \in \Lambda, y \in \mathcal{D} \}$ for $\mathcal{C}, \mathcal{D} \subseteq E$ and $\Lambda \subseteq \mathbb{R}$. Also, whenever possible we shorten singletons to their single elements, e.g., we write $\mathbb{R}_+ (1 \oplus \mathcal{C})$ to denote the conic homogenization of the set $\mathcal{C}$ in one higher dimensional space.

| Table 1. Notation for special sets. |
|------------------------------------|
| $[n]$                               := $\{1, \ldots, n\}$ for each $n \in \mathbb{N}$ |
| $\mathcal{P}(X)$                   := the power set of $X$ |
| $\mathbb{R}_+$                     := $\{ x \in \mathbb{R} : x \geq 0 \}$, the set of nonnegative reals |
| $\mathbb{R}_{++}$                  := $\{ x \in \mathbb{R} : x > 0 \}$, the set of positive reals |
| $\mathbb{R}^{n \times n}$         := the space of $n \times n$ real-valued matrices |
| $\mathbb{S}^n$                     := $\{ X \in \mathbb{R}^{n \times n} : X = X^T \}$, the space of symmetric $n \times n$ matrices |
| $\mathbb{S}^n_+$                   := $\{ X \in \mathbb{S}^n : h^T X h \geq 0 \forall h \in \mathbb{R}^n \}$, the cone of positive semidefinite matrices |
| $\mathbb{S}^{n,+}$                 := $\{ X \in \mathbb{S}^n : h^T X h > 0 \forall h \in \mathbb{R}^n \setminus \{0\} \}$, the cone of positive definite matrices |
| $\mathcal{E}_n$                    := the elliptope; see (10) |

2.1. Uniqueness of Dual Optimal Solutions. Delorme and Poljak [4] proved that the dual SDP in (5) has a unique optimal solution. We shall state a slightly generalized version of their result with some changes and include a proof for the sake of completeness.
Proposition 1 ([4, Theorem 2]). Consider the primal-dual pair of SDPs
\[
\max \{ \text{Tr}(CX) : \mathcal{A}(X) = b, X \succeq 0 \} \quad \text{and} \quad (6)
\]
\[
\min \{ b^T y : S = \mathcal{A}^*(y) - C, S \succeq 0 \}, \quad (7)
\]
Theorem 3

Also, set

\[ \tilde{y} := \frac{1}{2}(y_1 + y_2) \]

and note that \( b^T \tilde{y} = 0 \). Let \( \mu > 0 \) be the smallest positive eigenvalue of \( \tilde{S} \in S_+^n \) \( \setminus \{0\} \). Let \( \|\cdot\|_2 \) denote the operator 2-norm. If \( \beta \|A^*(\tilde{z})\|_2 = 0 \), set \( \varepsilon := 1 \); otherwise set

\[ \varepsilon := \frac{\mu}{\|\beta\|A^*(\tilde{z})\|_2} > 0. \]

Also, set \( \tilde{y} := \tilde{y} + \varepsilon d \) and \( \tilde{S} := A^*(\tilde{y}) - C \). Let \( h \in \mathbb{R}^n \). Write \( h = h_1 + h_2 \) with \( h_1 \in \text{Null}(\tilde{S}) \) and \( h_2 \in [\text{Null}(\tilde{S})]^\perp \). By (8) we have

\[ h^T \tilde{S} h = h^T \tilde{S} h + \varepsilon h^T A^*(d) h \]

\[ \geq \mu \|h_2\|^2 + \varepsilon h^T A^*(\tilde{y}) h + \varepsilon \beta \|A^*(\tilde{z})\|_2 \|h_2\|^2 \]

\[ \geq \mu \|h_2\|^2 + \varepsilon h^T A^*(\tilde{y}) h - \varepsilon \|\beta\|A^*(\tilde{z})\|_2 \|h_2\|^2 \]

\[ \geq \varepsilon h^T A^*(\tilde{y}) h. \]

Thus, \( \tilde{S} \geq \varepsilon A^*(\tilde{y}) \geq 0 \), so there exists a feasible solution for (7) with objective value strictly smaller than \( b^T \tilde{y} = b^T \tilde{y} \), a contradiction. \( \square \)

Corollary 2 ([4, Theorem 2]). The dual SDP in (5) has a unique optimal solution.

Proof. We shall apply Proposition 1 to (5). Let us see that the map \( A := \text{diag} \) satisfies the required properties. Take \( \tilde{X} := I \) and \( \tilde{y} := 1 \). Let \( y \in \mathbb{R}^n \) be nonzero. Define \( z \in \mathbb{R}^n \) as \( z_i := |y_i| \) for every \( i \in [n] \), and note that \( \text{Null}(\text{Diag}(y)) = \text{Null}(\text{Diag}(z)) \) and that \( 1^T z > 0 \) since \( y \neq 0 \). \( \square \)

2.2. Vertices of the Elliptope. Let \( \mathcal{C} \) be a convex set in an Euclidean space \( \mathbb{E} \). The normal cone of \( \mathcal{C} \) at \( \bar{x} \in \mathcal{C} \) is

\[ \text{Normal}(\mathcal{C}; \bar{x}) := \{ a \in \mathbb{E} : \langle a, x \rangle \leq \langle a, \bar{x} \rangle \forall x \in \mathcal{C} \}, \]

i.e., it is the set of all normals to supporting halfspaces of \( \mathcal{C} \) at \( \bar{x} \). Note that we are identifying the dual space \( \mathbb{E}^* \) of \( \mathbb{E} \) with \( \mathbb{E} \). We say that \( \bar{x} \in \mathcal{C} \) is a vertex of \( \mathcal{C} \) if \( \text{Normal}(\mathcal{C}; \bar{x}) \) is full-dimensional. The set of vertices of the elliptope

\[ \mathcal{E}_n := \{ X \in S_+^n : \text{diag}(X) = 1 \} \]

was determined by Laurent and Poljak [13]:

Theorem 3 ([13, Theorem 2.5]). The set vertices of \( \mathcal{E}_n \) is \( \{ xx^T : x \in \{ \pm 1 \}^n \} \).

Table 5. Notation for a graph \( G = (V, E) \).

| Symbol | Definition |
|--------|------------|
| \( V(G) \) | the vertex set of \( G \) |
| \( E(G) \) | the edge set of \( G \) |
| \( \mathcal{L}_G(w) \) | the weighted Laplacian matrix of \( G \) with weights \( w \in R^E \); see (4) |
| \( G \uplus H \) | the cosum of graphs \( G \) and \( H \); see (14) |
An automorphism of $\mathcal{E}_n$ is a nonsingular linear operator $T$ on $\mathbb{S}^n$ that preserves $\mathcal{E}_n$, i.e., $T(\mathcal{E}_n) = \mathcal{E}_n$. For $s \in \{\pm1\}^n$, the map $X \mapsto \text{Diag}(s)X\text{Diag}(s)$ is easily checked to be an automorphism of $\mathcal{E}_n$. If $x, y \in \{\pm1\}^n$, then $y = \text{Diag}(s)x$ for $s \in \{\pm1\}^n$ defined by $s_i := x_iy_i$ for each $i \in [n]$. Hence, any vertex of $\mathcal{E}_n$ can be mapped into the vertex $11^T$ by an automorphism of $\mathcal{E}_n$; i.e., the automorphism group of $\mathcal{E}_n$ acts transitively on the vertices of $\mathcal{E}_n$. This allows us to prove many linear properties about the vertices of $\mathcal{E}_n$ by just proving them for the vertex $11^T$. We shall make extensive use of this fact without further mention.

Laurent and Poljak [13] also provided formulas for the normal cones of the elliptope. Here we shall use slightly different formulas from [3, Proposition 2.1]:

$$\text{Normal}(\mathcal{E}_n; X) = \text{Im}(\text{Diag}) - (\mathbb{S}_+^n \cap \{X\}^\perp)$$

$$= \text{Im}(\text{Diag}) - \{Y \in \mathbb{S}_+^n : \text{Im}(Y) \subseteq \text{Null}(X)\} \quad \forall X \in \mathcal{E}_n. \quad (11)$$

When $\tilde{X}$ is a vertex of $\mathcal{E}_n$, every element of $\text{Normal}(\mathcal{E}_n; \tilde{X})$ can be described in a unique way as an element of the Minkowski sum at the RHS of (11):

**Lemma 4.** Let $\tilde{X}$ be a vertex of $\mathcal{E}_n$. Let $\tilde{y}, \tilde{y} \in \mathbb{R}^n$ and $\tilde{S}, \tilde{S} \in \mathbb{S}_+^n \cap \{\tilde{X}\}^\perp$ be such that $\text{Diag}(\tilde{y}) + \tilde{S} = \text{Diag}(\tilde{y}) + \tilde{S}$. Then $\tilde{y} = \tilde{y}$ and $\tilde{S} = \tilde{S}$.

**Proof.** We may assume that $\tilde{X} = 11^T$. Then $\tilde{S} \in \mathbb{S}_+^n \cap \{11^T\}^\perp$ implies that $\tilde{S}1 = 0$. Analogously, $\tilde{S}1 = 0$. Thus $\tilde{y} = (\text{Diag}(\tilde{y})1 = (\text{Diag}(\tilde{y}) + \tilde{S})1 = \text{Diag}(\tilde{y})1 = \tilde{y}$, so $\tilde{S} = \tilde{S}$. \hfill \square

### 3. Failure of Strict Complementarity with Laplacian Objectives

Existence of strictly complementary optimal solutions is known to be equivalent to membership of the objective vector in the relative interior of some normal cone:

**Proposition 5** ([3, Proposition 4.2]). If the feasible region $\mathcal{C}$ of the SDP (3) has a positive definite matrix, then strict complementarity holds for (3) and its dual if and only if $C \in \text{ri}(\text{Normal}(\mathcal{C}; X))$ for some $X \in \mathcal{C}$.

Hence, strict complementarity is locally generic when the objective function is chosen in the normal cone of a given feasible solution; see [3, Corollary 4.3].

By (11) and standard convex analysis,

$$\text{ri}(\text{Normal}(\mathcal{E}_n; X)) = \text{Im}(\text{Diag}) - \text{ri}(\mathbb{S}_+^n \cap \{X\}^\perp)$$

$$= \text{Im}(\text{Diag}) - \{ \{Y \in \mathbb{S}_+^n : \text{Im}(Y) = \text{Null}(X)\} \quad \forall X \in \mathcal{E}_n. \quad (12)$$

When $\tilde{X}$ is a vertex of $\mathcal{E}_n$, we may combine (11) with (12) and Lemma 4 to conclude that

$$\text{bd}(\text{Normal}(\mathcal{E}_n; \tilde{X})) = \text{Im}(\text{Diag}) - \text{rbd}(\mathbb{S}_+^n \cap \{\tilde{X}\}^\perp)$$

$$= \text{Im}(\text{Diag}) - \{ \{Y \in \mathbb{S}_+^n : \text{Im}(Y) \subseteq \text{Null}(\tilde{X})\} \quad \forall \tilde{X} \in \mathcal{E}_n. \quad (13)$$

In [3], we noted that strict complementarity holds for (5) for every $C$ of the form $C = \frac{1}{2}L_G(w)$ with $w \geq 0$ provided that the polar $\mathcal{E}_n^\perp := \{Y \in \mathbb{S}^n : \text{Tr}(YX) \leq 1 \forall X \in \mathcal{E}_n\}$ of the elliptope is facially exposed, and we (implicitly) asked whether the latter holds. It turns out, Laurent and Poljak [14, Example 5.10] showed, even before we raised the question, in a different context and using a slightly different terminology, that strict complementarity may fail for (5) for every $n \geq 3$, hence answering the question in the negative. For each complete graph $G = K_n$ with $n \geq 3$, they provided a weight function $w \geq 0$ for which strict complementarity fails for (5) with $C = \frac{1}{2}L_G(w)$.

We generalize their construction showing that strict complementarity may fail with a weighted Laplacian objective for graphs which are cosums, with mild conditions on the (co-)summands. Recall that, if $G = (V, E)$ and $H = (U, F)$ are graphs such that $V \cap U = \varnothing$, the cosum of $G$ and $H$ is the graph

$$G \mp H := (V \cup U, E \cup F \cup \{\{v, u\} : (v, u) \in V \times U\}). \quad (14)$$

We shall use a characterization of positive semidefinite matrices partitioned in blocks using Schur complements and the Moore-Penrose pseudoinverse:

**Lemma 6** (see [6, Theorem 4.3]). For $A \in \mathbb{S}^m$, $C \in \mathbb{S}^n$, and $B \in \mathbb{R}^{m \times n}$, we have

$$\begin{bmatrix} A & B \\ B^T & C \end{bmatrix} \succeq 0 \iff A \succeq 0, \quad (I - AA^\dagger)B = 0, \quad \text{and} \quad C \succeq B^T A^\dagger B. \quad (15)$$
Theorem 7. Let $G$ and $H$ be graphs with $n_G \geq 1$ and $n_H \geq 1$ vertices, respectively. Let $w_G: E(G) \to \mathbb{R}_{++}$ and $w_H: E(H) \to \mathbb{R}_{++}$ be weight functions, and denote the respective weighted Laplacians by $L_G := \mathcal{L}_G(w_G)$ and $L_H := \mathcal{L}_H(w_H)$. Set $\mu_G := \lambda_{\text{max}}(L_G)$ and $\mu_H := \lambda_{\text{max}}(L_H)$. Suppose that $n_G \mu_G > n_H \mu_H$ and that $H$ is connected. Define $\bar{w}: E(G \uplus H) \to \mathbb{R}_{++}$ as $\bar{w} := w_G \oplus w_H \oplus \alpha \mathbf{1}$ where $\alpha := \mu_G/n_H$. For enhanced clarity denote the vectors of all-ones in $\mathbb{R}^{V(G)}$ and $\mathbb{R}^{V(H)}$ by $\mathbf{1}_G$ and $\mathbf{1}_H$, respectively. Then the unique pair of primal-dual optimal solutions for (5) with $4C := \mathcal{L}_{G \uplus H}(\bar{w})$ is $(X^*, y^* \oplus S^*)$ where

$$
X^* := \begin{bmatrix} -\mathbf{1}_G & -\mathbf{1}_G \\ \mathbf{1}_H & \mathbf{1}_H \end{bmatrix}^T, \quad y^* := 2\alpha \begin{bmatrix} n_H \mathbf{1}_G \\ n_G \mathbf{1}_H \end{bmatrix}, \quad S^* := \begin{bmatrix} \mu_GI - L_G \\ \alpha \mathbf{1}_H \mathbf{1}_G^T \end{bmatrix} \mathbf{1}_H^T .
$$

(16)

In particular, since $(X^* + S^*)(\mathbf{1} \oplus 0) = 0$ for any $\mu_G$-eigenvector $h$ of $L_G$, there is no strictly complementary pair of primal-dual optimal solutions for (5).

Proof. It is easy to check that $X^*$ is feasible in the primal. We have

$$
S^* = 2 \begin{bmatrix} \mu_GI & 0 \\ 0 & \alpha n_GI \end{bmatrix} - \begin{bmatrix} L_G + \mu_GI & -\alpha \mathbf{1}_H \mathbf{1}_G^T \\ -\alpha \mathbf{1}_H \mathbf{1}_G^T & L_H + \alpha n_GI \end{bmatrix} = \text{Diag}(y^*) - L_{G \uplus H}(\bar{w}),
$$

and the condition $S^* \succeq 0$ is equivalent to the conditions

$$
\begin{align}
A &:= \mu_GI - L_G \succeq 0, \quad (17a) \\
(I - AA^T) &\geq 0, \quad (17b) \\
\alpha n_GI &\geq L_H + \alpha^2 \mathbf{1}_G^T \mathbf{1}_G \mathbf{1}_H^T. \quad (17c)
\end{align}
$$

Note that (17a) holds trivially. Also $AI_G = \mu_GI_G$, so $I_G \in \text{Im}(A)$ and (17b) holds since $I - AA^T$ is the orthogonal projector onto $\text{Null}(A) = \text{Im}(A)^\perp$. Finally, $A^T I_G = \mu_G^{-1}I_G$, so $\alpha n_GI \succeq L_H$ by assumption. It follows that $y^* \oplus S^*$ is feasible in the dual. It is easy to check that $\text{Tr}(X^* S^*) = 0$, so $X^*$ and $y^* \oplus S^*$ are optimal solutions. By Corollary 2, $y^* \oplus S^*$ is the unique optimal solution for the dual.

It remains to show that $X^*$ is the unique optimal solution for the primal. Let

$$
X = \begin{bmatrix} X_G & B \\ B^T & X_H \end{bmatrix}
$$

be an optimal solution for the primal. Complementary slackness yields

$$
0 = XS^* = \begin{bmatrix} X_G(\mu_GI - L_G) + \alpha B \mathbf{1}_H \mathbf{1}_G^T & \alpha X_G \mathbf{1}_H \mathbf{1}_G^T + B(\alpha n_GI - L_H) \\ B^T(\mu_GI - L_G) + \alpha X_H \mathbf{1}_H \mathbf{1}_G^T & \alpha B^T \mathbf{1}_G \mathbf{1}_H^T + X_H(\alpha n_GI - L_H) \end{bmatrix}.
$$

(18)

If $h \perp \mathbf{1}_H$ is an eigenvector of $L_H$, (left-)multiplying $h$ by the bottom right block in (18) yields $X_Hh = 0$, where we used the assumption that $\alpha n_GI \succeq L_H$. Since $H$ is connected, this implies that $X_H$ is a nonnegative scalar multiple of $\mathbf{1}_H \mathbf{1}_H^T$, and so

$$
X_H = \mathbf{1}_H \mathbf{1}_H^T.
$$

Next apply $\mathbf{1}_G^T \cdot \mathbf{1}_H$ and $\mathbf{1}_H^T \cdot \mathbf{1}_G$ to the top right block and bottom left block of (18), respectively, to get

$$
0 = n_H \mathbf{1}_G X_G \mathbf{1}_H + n_G \mathbf{1}_G^T B \mathbf{1}_H, \quad (19)
$$

$$
0 = n_H \mathbf{1}_H B^T \mathbf{1}_G + n_G \mathbf{1}_H^T X_H \mathbf{1}_G. \quad (20)
$$

Hence,

$$
\frac{\mathbf{1}_G^T X_G \mathbf{1}_G}{n_G} = \frac{\mathbf{1}_H^T X_H \mathbf{1}_H}{n_H} \quad \text{and} \quad X_G = \mathbf{1}_G \mathbf{1}_G^T.
$$

Finally, by (19) we get $\mathbf{1}_G^T B \mathbf{1}_H = -n_Gn_H$, and so $B = -\mathbf{1}_G \mathbf{1}_H^T$. Hence, $X = X^*$.

Note that the dimension of the $\lambda_{\text{max}}(\mathcal{L}_G(w))$-eigenspace controls the “degree” to which strict complementarity fails in Theorem 7. In particular, when $G$ is the complete graph and $w_G = 1$, we have rank($X^*$) + rank($S^*$) = 1 + $n_H$.

Theorem 7 shows that if $F$ is a graph which is a cosum (i.e., the complement of $F$ is not connected) $F = G \uplus H$, where $G$ has at least one edge and $H$ is connected, then there is a nonnegative weight function $w: E(F) \to \mathbb{R}_{++}$ such that strict complementarity fails for (5) with $C = \frac{1}{4}\mathcal{L}_F(w)$; one may just fix
$w_H \in \mathbb{R}_{++}^{E(H)}$ arbitrarily, e.g., $w_H = 1$, and set $w_G := M1$ for large enough $M$ so that $n_G \mu_G > n_H \mu_H$. A natural question following from this is:

**Problem 8.** Characterize the set of graphs for which there exists a positive weight function on the edges such that strict complementarity fails for (5) when $4C$ is the corresponding weighted Laplacian matrix.

4. **Generic Failure of Strict Complementarity on the Boundaries of Normal Cones**

In this section, we consider how often strict complementarity holds for (5) when $C$ lies in the (relative) boundary of $\text{Normal}(\mathcal{E}_n; X)$ for some vertex $X$ of $\mathcal{E}_n$. Note that this boundary is described as a Minkowski sum in (13).

We start by considering the case $n = 3$, where (13) simplifies to

$$\text{bd}(\text{Normal}(\mathcal{E}_3; \bar{x}^T)) = \text{Im}(\text{Diag}) - \{z \bar{z}^T : z \in \{\pm 1\}^3\}$$

for every $\bar{x} \in \{\pm 1\}^3$.

**Proposition 9.** Let $\bar{x} \in \{\pm 1\}^3$, and let $C = \text{Diag}(\bar{y}) - \bar{z} \bar{z}^T$ for some $\bar{y} \in \mathbb{R}^3$ and $\bar{z} \in \{\pm 1\}^3$, so that $C \in \text{bd}(\text{Normal}(\mathcal{E}_3; \bar{x}^T))$. Then strict complementarity holds for (5) if and only if $\bar{z}_i = 0$ for some $i \in [3]$.

**Proof.** Set $\bar{S} := \text{Diag}(\bar{y}) - C = \bar{z} \bar{z}^T$ and $\bar{X} := \bar{x} \bar{x}^T$. Clearly, $\bar{y} \oplus \bar{S}$ is feasible in the dual and $\text{Tr}(\bar{S} \bar{X}) = (\bar{z} \bar{z}^T)^2 = 0$, so $(\bar{X}, \bar{y} \oplus \bar{S})$ is a pair of primal-dual optimal solutions. By Corollary 2, $\bar{y} \oplus \bar{S}$ is the unique optimal solution in the dual.

Suppose that $\bar{z}_i \neq 0$ for every $i \in [3]$. We claim that $\bar{X}$ is the unique optimal solution in the primal. Indeed, let $X \in \mathcal{E}_3$ be optimal in the primal. Then $0 = \text{Tr}(\bar{S} \bar{X}) = \bar{z} \bar{z}^T \bar{x} = 0$. Thus,

$$0 = \begin{bmatrix} 1 & X_{12} & X_{13} \\ X_{12} & 1 & X_{23} \\ X_{13} & X_{23} & 1 \end{bmatrix} \begin{bmatrix} \bar{z}_1 \\ \bar{z}_2 \\ \bar{z}_3 \end{bmatrix} = \begin{bmatrix} \bar{z}_1 + 2\bar{z}_2 X_{12} + \bar{z}_3 X_{13} \\ \bar{z}_1 X_{12} + \bar{z}_2 + 2\bar{z}_3 X_{23} \\ \bar{z}_1 X_{13} + \bar{z}_2 X_{23} + \bar{z}_3 \end{bmatrix},$$

so

$$\begin{bmatrix} \bar{z}_2 & \bar{z}_3 & 0 \\ \bar{z}_1 & 0 & \bar{z}_3 \\ 0 & \bar{z}_1 & \bar{z}_2 \end{bmatrix} \begin{bmatrix} X_{12} \\ X_{13} \\ X_{23} \end{bmatrix} = -\bar{z}.$$

The determinant of the matrix defining the latter linear system is $-2\bar{z}_1 \bar{z}_2 \bar{z}_3 \neq 0$, so the unique solution is given by the off-diagonal entries of $\bar{X}$.

Suppose now that $\bar{z}_i = 0$ for some $i \in [3]$. If $\bar{z} = 0$ then $(I, \bar{y} \oplus 0)$ satisfies strict complementarity, so assume $\bar{z} \neq 0$. Set $\bar{x} := \text{Diag}(I - e_i) \bar{x}$ and $\bar{X} := \bar{x} \bar{x}^T + e_i e_i^T \in \mathcal{E}_3$. Then $\text{Tr}(\bar{S} \bar{X}) = \bar{z} \bar{z}^T (\bar{\bar{x}} \bar{\bar{x}}^T + e_i e_i^T) \bar{z} = (\bar{z} \bar{z}^T)^2 + \bar{z}_i^2 = 0$ since $\bar{z} \bar{z}^T \bar{x} = \bar{z} \bar{z}^T \bar{x} = 0$. Hence, $(\bar{X}, \bar{y} \oplus \bar{S})$ is a strictly complementarity pair of primal-dual optimal solutions for (5).

For $n \geq 4$, characterization of strict complementarity in (5) is not as easily described. However, we can prove the following condition sufficient for the failure of strict complementarity, which will turn out to be sufficient for our purposes.

**Theorem 10.** Let $n \geq 3$. Let $C = \text{Diag}(\bar{y}) - \tilde{S}$ for some $\bar{y} \in \mathbb{R}^n$ and $\tilde{S} \in \mathbb{S}_n^+$, so that $C \in \text{bd}(\text{Normal}(\mathcal{E}_n; 11^T))$. Suppose that $\text{Null}(\tilde{S}) = \text{span}\{1, h\}$ for some $h \in \{1\}^+$ and that $h$ has at least three distinct coordinates. Then strict complementarity fails for (5).

**Proof.** Set $y^* := \bar{y}$ and $S^* := \text{Diag}(y^*) - C = \bar{S}$. Set $\bar{X} := 11^T$. Clearly, $y^* \oplus S^*$ is feasible in the dual and $\text{Tr}(S^* X^*) = 0$, so $(\bar{X}, y^* \oplus S^*)$ is a pair of primal-dual optimal solutions. By Corollary 2, $y^* \oplus S^*$ is the unique optimal solution in the dual. We shall prove that $\bar{X}$ is the unique optimal solution in the primal.

Let $X \in \mathcal{E}_n$ be an optimal solution in the primal. By complementary slackness, $\text{Tr}(X S^*) = 0$, so $S^* X = 0$ and $\text{Im}(X) \subseteq \text{Null}(S^*) = \text{span}\{1, h\}$. Hence, $X = \alpha_1 11^T + \alpha_2 h h^T + \alpha_3 (h 1^T + 1 h^T)$ for some $\alpha \in \mathbb{R}^3$. Since $\text{diag}(X) = 1$, we find that $\alpha_1 + \alpha_2 h_i^2 + 2 \alpha_3 h_i = 1$ for every $i \in [n]$. Let $i, j, k \in [n]$ such that $\{|h_i, h_j, h_k\}| = 3$. Then

$$\begin{bmatrix} 1 & 2h_i & h_i^2 \\ 1 & 2h_j & h_j^2 \\ 1 & 2h_k & h_k^2 \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{bmatrix} = 1.$$
Theorem 10 seems to indicate that strictly complementarity fails “almost everywhere” on the boundary of Normal(ε_r; 11T), since the high rank matrices make up the bulk of the boundary (consider that the set of nonsingular matrices is open and dense) and for “most” of them the extra vector h in the nullspace has at least three distinct coordinates. Unfortunately, we are dealing with somewhat complicated sets (e.g., the high rank matrices in the boundary of a normal cone). In order to make our previous statements precise, we shall make use of the theory of Hausdorff measures, which we introduce next.

4.1. Preliminaries on Hausdorff Measures. We refer the reader to [20], though we use different notation and more standard terminology. See also [5, 18] for a somewhat similar presentation. We focus our presentation on finite-dimensional normed spaces (over the reals) but most of it could be developed for arbitrary metric spaces. Our main normed spaces are (subspaces of) \( \mathbb{R}^n \) and \( \mathbb{S}^n \). Since these are Euclidean spaces, they are equipped with a norm induced by their inner-products, and that is the norm that we will consider unless explicitly stated otherwise. We shall only use other norms in Section 5.

Let \( \mathcal{Y} \) be a finite-dimensional normed space. Let \( d \in \mathbb{R}_+ \) and \( \varepsilon \in \mathbb{R}_{++} \). For each \( \mathcal{X} \subseteq \mathcal{Y} \), define

\[
H^d_\varepsilon(\mathcal{X}) := \inf \left\{ \sum_{i=0}^{\infty} [\text{diam}(\mathcal{U}_i)]^d : \{\mathcal{U}_i\}_{i \in \mathbb{N}} \subseteq \mathcal{P}(\mathcal{Y}), \mathcal{X} \subseteq \bigcup_{i=0}^{\infty} \mathcal{U}_i, \text{diam}(\mathcal{U}_i) < \varepsilon \forall i \in \mathbb{N} \right\},
\]

where the diameter of \( \mathcal{U} \subseteq \mathcal{Y} \) is \( \text{diam}(\mathcal{U}) := \sup_{x,y \in \mathcal{U}} \|x - y\| \). The function \( H_d : \mathcal{P}(\mathcal{Y}) \to [0, +\infty] \) defined by

\[
H_d(\mathcal{X}) := \sup_{\varepsilon > 0} H^d_\varepsilon(\mathcal{X}) = \lim_{\varepsilon \downarrow 0} H^d_\varepsilon(\mathcal{X}) \quad \forall \mathcal{X} \subseteq \mathcal{Y}
\]

is an outer measure on \( \mathcal{Y} \). Hence, the restriction of \( H_d \) to the \( H_d \)-measurable subsets of \( \mathcal{Y} \) is a complete measure on \( \mathcal{Y} \), called the \( d \)-dimensional Hausdorff measure on \( \mathcal{Y} \). The 0-dimensional Hausdorff measure \( H_0 \) is the cardinality of a set, \( H_1 \) is its length, \( H_2 \) is its area, and so on.

Let \( d \) be a positive integer and set \( \mathcal{Y} := \mathbb{R}^d \). Let \( \lambda_d : \mathcal{P}(\mathbb{R}^d) \to [0, +\infty] \) denote the \( d \)-dimensional Lebesgue outer measure on \( \mathbb{R}^d \). It can be proved [20, Theorem 30] that

\[
\frac{\lambda_d(\mathcal{X})}{\lambda_d(\mathcal{B})} = \frac{H_d(\mathcal{X})}{2^d} \quad \forall \mathcal{X} \subseteq \mathbb{R}^d.
\]

In particular, the \( H_d \)-measurable subsets of \( \mathbb{R}^d \) are the same as the \( \lambda_d \)-measurable sets.

Let \( a, b \in \mathbb{R}_+ \) with \( a < b \) and let \( \mathcal{X} \subseteq \mathcal{Y} \). It is not hard to prove from the definition that

\[
H_a(\mathcal{X}) < \infty \implies H_b(\mathcal{X}) = 0,
\]

\[
H_b(\mathcal{X}) > 0 \implies H_a(\mathcal{X}) = \infty.
\]

Hence,

\[
\sup\{ d \in \mathbb{R}_+ : H_d(\mathcal{X}) = \infty \} = \inf\{ d \in \mathbb{R}_+ : H_d(\mathcal{X}) = 0 \},
\]

and the common value in (26) is the Hausdorff dimension of \( \mathcal{X} \), denoted by \( \dim_H(\mathcal{X}) \). In particular, if \( d \in \mathbb{R}_+ \) and \( \mathcal{X} \subseteq \mathcal{Y} \) satisfy \( H_d(\mathcal{X}) \in (0, \infty) \), then \( \dim_H(\mathcal{X}) = d \).

We may now define genericity precisely. Let \( \mathcal{X} \) be a subset of a finite-dimensional normed space \( \mathcal{Y} \). Let \( P \) be a property that may hold or fail for points in \( \mathcal{X} \), i.e., \( P(x) \) is either true or false for each \( x \in \mathcal{X} \). We say that \( P \) holds generically on \( \mathcal{X} \) if \( H_d(\{ x \in \mathcal{X} : P(x) \text{ is false} \}) = 0 \) for \( d := \dim_H(\mathcal{X}) \). We say that \( P \) fails generically on \( \mathcal{X} \) if the negation of \( P \) holds generically on \( \mathcal{X} \). In Section 4.3, we will use Theorem 10 to prove that strict complementarity fails generically at the boundary of the normal cone of any vertex of \( \mathcal{C}_n \), for \( n \geq 3 \), modulo some qualification on the ambient space. In the remainder of this section and in the next one, we will describe a few more measure-theoretic tools that we shall use towards this goal.

Let \( \mathcal{Y} \) and \( \mathcal{W} \) be finite-dimensional normed spaces. Let \( \mathcal{X} \subseteq \mathcal{Y} \). Recall that a function \( \varphi : \mathcal{X} \to \mathcal{W} \) is Lipschitz continuous with Lipschitz constant \( L > 0 \) if

\[
\|\varphi(x) - \varphi(y)\| \leq L\|x - y\| \quad \forall x, y \in \mathcal{X}.
\]

The following is well known and easy to prove:

**Theorem 11.** Let \( \mathcal{Y} \) and \( \mathcal{W} \) be finite-dimensional normed spaces. Let \( \mathcal{X} \subseteq \mathcal{Y} \) and \( d \in \mathbb{R}_+ \). Let \( \varphi : \mathcal{X} \to \mathcal{W} \) be Lipschitz continuous with Lipschitz constant \( L \). Then

\[
H_d(\varphi(\mathcal{X})) \leq L^d H_d(\mathcal{X}).
\]
Theorem 11 is especially useful to determine some Hausdorff dimensions via bi-Lipschitz maps. We recall the definition here. Let \( \mathcal{V} \) and \( \mathcal{W} \) be finite-dimensional normed spaces. Let \( \mathcal{X} \subseteq \mathcal{V} \), and let \( \varphi: \mathcal{X} \to \mathcal{W} \) be a one-to-one function with range \( \mathcal{Y} := \varphi(\mathcal{X}) \). We say that \( \varphi \) is bi-Lipschitz continuous with Lipschitz constants \( L_1 > 0 \) and \( L_2 > 0 \) if \( \varphi \) is Lipschitz continuous with Lipschitz constant \( L_1 \) and \( \varphi^{-1}: \mathcal{W} \to \mathcal{V} \) is Lipschitz continuous with Lipschitz constant \( L_2 \).

Corollary 12. Let \( \mathcal{V} \) and \( \mathcal{W} \) be finite-dimensional normed spaces. Let \( \mathcal{X} \subseteq \mathcal{V} \) and \( d \in \mathbb{R}_+ \). Let \( \varphi: \mathcal{X} \to \mathcal{W} \) be bi-Lipschitz continuous with Lipschitz constants \( L_1 \) and \( L_2 \). Then

\[
L_2^{-d} H_d(\mathcal{X}) \leq H_d(\varphi(\mathcal{X})) \leq L_1^d H_d(\mathcal{X}).
\]

In particular, if \( H_d(\mathcal{X}) \in (0, \infty) \), then \( \dim_H(\varphi(\mathcal{X})) = d \).

This corollary may be used, for instance, to regard any \( d \)-dimensional Euclidean space \( \mathcal{V} \) as \( \mathbb{R}^d \) by considering the coordinate map \( \varphi: \mathcal{V} \to \mathbb{R}_d \) with respect to a fixed orthonormal basis of \( \mathcal{V} \). Another frequent use of Corollary 12 goes as follows. Equip the space \( S^n \) with the trace inner-product. If \( Q \in \mathbb{R}^{n \times n} \) is an orthogonal matrix, the map \( X \in S^n \to QXQ^T \) preserves inner-products, and hence norms and distances; hence, the map is Lipschitz continuous with Lipschitz constant 1. Its inverse is \( X \in S^n \to Q^TXQ \) and so the map \( X \in S^n \to QXQ^T \) is bi-Lipschitz continuous with Lipschitz constants 1 and 1.

The next result is useful for determining the Hausdorff dimension of some simple unbounded sets in the \( \sigma \)-finite case, when (27) is not directly applicable:

Proposition 13. Let \( \mathcal{X} \) be a subset of a finite-dimensional normed \( \mathcal{V} \). For each \( i \in \mathbb{N} \), let \( \mathcal{Y}_i \) be a subset of a finite-dimensional normed space \( \mathcal{W}_i \), and let \( \varphi_i: \mathcal{Y}_i \to \mathcal{V} \) be a Lipschitz continuous function with Lipschitz constant \( L_i \). If \( \mathcal{X} \subseteq \bigcup_{i \in \mathbb{N}} \varphi_i(\mathcal{Y}_i) \), then \( \dim_H(\mathcal{X}) \leq \sup_{i \in \mathbb{N}} \dim_H(\mathcal{Y}_i) \).

Proof. Set \( d := \sup_{i \in \mathbb{N}} \dim_H(\mathcal{Y}_i) \). Let \( d \geq d \). Then (26) yields \( H_d(\mathcal{Y}_i) = 0 \) for each \( i \in \mathbb{N} \), so by Theorem 11 we have \( H_d(\mathcal{X}) \leq \sum_{i \in \mathbb{N}} L_i^d H_d(\mathcal{Y}_i) = 0 \).

For instance, \( \mathbb{R}^d = \bigcup_{M \in \mathbb{N}} MB \) and the ball \( MB \subseteq \mathbb{R}^d \) with nonzero \( M \) has Hausdorff dimension \( d \) by (27) and (23), so Proposition 13 shows that \( \dim_H(\mathbb{R}^d) \leq d \). Since \( \mathbb{R}^d \supseteq B \) shows that \( H_d(\mathbb{R}^d) \geq H_d(B) > 0 \) by (23), we conclude by (27) that \( \dim_H(\mathbb{R}^d) = d \). Together with Corollary 12, this shows that Hausdorff dimension and the usual (linear) dimension coincide on linear subspaces, and hence also for convex sets by translation invariance.

4.2. Hausdorff Measures and the Boundary Structure of Convex Sets. In this section we collect some results relating Hausdorff measures and the boundary structure of convex sets, including a quick review of basic facts about faces.

The following result is well known:

Theorem 14. Let \( \mathbb{E} \) be an Euclidean space. If \( \mathcal{C} \subseteq \mathbb{E} \) is a compact convex set with dimension \( d \geq 1 \), then \( \dim_H(\text{rbd}(\mathcal{C})) = d - 1 \).

Proof. We may assume that \( \dim(\mathbb{E}) = d \) so that \( \mathcal{C} \) has nonempty interior. By choosing an orthonormal basis for \( \mathbb{E} \), we may assume that \( \mathbb{E} = \mathbb{R}^d \). We may also assume that \( 0 \in \text{int}(\mathcal{C}) \) by translation invariance of Hausdorff measure. Set \( X := \text{bd}(\mathcal{B}_\infty) \), and note that \( H_{d-1}(X) \in (0, +\infty) \) by (23) and Corollary 12. Let \( \varepsilon, M \in \mathbb{R}^+ \) such that \( 2\varepsilon\mathcal{B}_\infty \subseteq \mathcal{C} \subseteq \frac{1}{2}M\mathcal{B}_\infty \). Let \( p_\mathcal{E}: \mathbb{R}^d \to \mathcal{C} \) be the metric projection onto \( \mathcal{C} \), i.e., \( \{p_\mathcal{E}(x)\} = \text{arg min}_{p_\mathcal{E}(x)} \|y - x\| \) for each \( x \in \mathbb{R}^d \). Then \( p_\mathcal{E} \) is Lipschitz continuous (with Lipschitz constant 1). Theorem 11 applied to \( p_\mathcal{E} \rvert_{\mathcal{X}} \) and positive homogeneity of \( H_{d-1} \) (of degree \( d - 1 \)) yield \( H_{d-1}(\text{bd}(\mathcal{C})) < \infty \). Similarly, applying Theorem 11 to the restriction to \( \text{bd}(\mathcal{C}) \) of metric projection onto \( \varepsilon\mathcal{B}_\infty \) yields \( H_{d-1}(\text{bd}(\mathcal{C})) > 0 \). The theorem now follows from (27).

Since we are dealing with convex cones, the previous result will be more useful to us when stated in a lifted form about pointed closed convex cones:

Corollary 15. Let \( \mathbb{E} \) be an Euclidean space. If \( \mathcal{K} \subseteq \mathbb{E} \) is a pointed closed convex cone with dimension \( d \geq 1 \), then \( \dim_H(\text{rbd}(\mathcal{K})) = d - 1 \).
Theorem 17. Let \( \mathcal{H} \) be pointed, after applying some rotation, which preserves Hausdorff measures by Corollary 12, we may assume that \( \mathcal{H} = \mathbb{R}^d \) for some compact convex set \( \mathcal{C} \subseteq \mathbb{R}^d \) where \( d := d - 1 \). For each \( N \in \mathbb{N} \), define the compact convex set \( \mathcal{H}_N := \mathcal{H} \cap [N, N + 1] \oplus \mathbb{R}^d \). Since

\[
\text{rbd}(\mathcal{H}) \subseteq \bigcup_{N=0}^{\infty} \text{rbd}(\mathcal{H}_N),
\]

the result follows from Proposition 13 and Theorem 14.

\[ \square \]

The next result refers to faces of a convex set, so before we state it we shall briefly recall the basic theory; see [19, Sec. 18]. Let \( E \) be an Euclidean space. Let \( C \subseteq E \) be a compact convex set. A convex subset \( \mathcal{F} \) of \( C \) is a face of \( C \) if, for each \( x, y \in C \) such that the open line segment \( (x, y) := \{(1 - \lambda)x + \lambda y : \lambda \in (0, 1)\} \) between \( x \) and \( y \) meets \( \mathcal{F} \), we have \( x, y \in \mathcal{F} \). We use the notation \( \mathcal{F} \subseteq C \) to denote that \( \mathcal{F} \) is a face of \( C \), and \( \mathcal{F} \triangleleft C \) to denote that \( \mathcal{F} \) is a proper face of \( C \), i.e., \( \mathcal{F} \subseteq C \) and \( \mathcal{F} \neq C \). Denote \( \text{Faces}(C) := \{ \mathcal{F} : \mathcal{F} \subseteq C \} \).

Faces of closed convex sets are closed, and faces of convex cones are convex cones. An arbitrary intersection of faces of \( C \) is a face of \( C \) and, since the faces of a convex set are partially ordered by inclusion and \( C \subseteq C \), every point \( x \) of \( C \) lies in a unique minimal face \( \mathcal{F} \) of \( C \); this face \( \mathcal{F} \) is characterized by the property \( x \in \text{ri}(\mathcal{F}) \). Also, it can be proved that \( \{ \text{ri}(\mathcal{F}) : \emptyset \neq \mathcal{F} \subseteq C \} \) is a partition of \( C \). If \( C \) is a compact convex set, it is not hard to prove that the faces of the homogenization of \( C \) are described by:

\[
\text{Faces}(\mathbb{R}^+ (1 + C)) = \{ \emptyset, \{0\} \} \cup \{ \mathbb{R}^+ (1 + \mathcal{F}) : \emptyset \neq \mathcal{F} \subseteq C \}.
\]

Theorem 16 (Larman [12]). Let \( E \) be an Euclidean space. If \( C \subseteq E \) is a compact convex set with dimension \( d \geq 1 \), then

\[
H_{d-1} \left( \bigcup_{\mathcal{F} \triangleleft C} \text{rbd}(\mathcal{F}) \right) = 0.
\]

As before, we shall need a conic version of Larman’s Theorem. We apply tools similar to the ones used to lift Theorem 14 to Corollary 15:

Theorem 17. Let \( E \) be an Euclidean space. If \( \mathcal{H} \subseteq E \) is a pointed closed convex cone with dimension \( d \geq 1 \), then

\[
H_{d-1} \left( \bigcup_{\mathcal{F} \triangleleft \mathcal{H}} \text{rbd}(\mathcal{F}) \right) = 0.
\]

\[ \square \]

4.3. Generic Failure of Strict Complementarity. In this section, we prove one of our main results: strict complementarity fails generically in the relative boundary of the normal cone of the ellipope at any of its vertices.

We shall apply Theorem 17 to \( S^*_n \). Let us briefly recall some well-known descriptions of the faces of the positive semidefinite cone \( S^*_n \). Let \( \mathcal{L}_n \) denote the set of linear subspaces of \( \mathbb{R}^n \). For each \( \mathcal{L} \in \mathcal{L}_n \), denote

\[
\mathcal{F}_\mathcal{L} := \{ X \in S^*_n : \text{Null}(X) \supseteq \mathcal{L} \}
\]

and note that

\[
\text{ri}(\mathcal{F}_\mathcal{L}) = \{ X \in S^*_n : \text{Null}(X) = \mathcal{L} \}.
\]
Theorem 19. Let $r := n - \dim(\mathcal{L})$.

Note that, for $\mathcal{L} \in \mathcal{L}_n$ such that $\mathcal{L} \neq \mathbb{R}^n$, there is an orthogonal matrix $Q \in \mathbb{R}^{\times n}$ such that

$$\mathcal{F}_\mathcal{L} = \left\{ Q \begin{bmatrix} U & 0 \\ 0 & 0 \end{bmatrix} Q^T : U \in S_n^+ \right\},$$

where $r := n - \dim(\mathcal{L})$.

Lemma 18. Let $n \geq 2$ be an integer. Then the property “$C \mapsto \text{rank}(C) = n - 1$” holds generically in $\text{bd}(S_n^+)$. 

Proof. Set $d := \dim_H(S_n^+)$. Note that $d - 1 = \dim_H(\text{bd}(S_n^+))$ by Corollary 15. Let $X \in \text{bd}(S_n^+)$ such that $\text{rank}(X) = n - 1$ fails. Then $\text{rank}(X) \leq n - 2$. For each nonzero $h \in \text{Null}(X)$, let $\mathcal{L}$ be the linear subspace of $\mathbb{R}^n$ spanned by $h$ and note that $X \in \text{rd}(\mathcal{F}_\mathcal{L})$, following the notation from (34). Hence,

$$\left\{ X \in \text{bd}(S_n^+) : \text{rank}(X) \neq n - 1 \right\} = \left\{ X \in S_n^+ : \text{rank}(X) \leq n - 2 \right\} \subseteq \bigcup_{\mathcal{F} \subset S_n^+} \text{rd}(\mathcal{F}).$$

The $(d - 1)$-dimensional Hausdorff measure of the set on the RHS above is zero by Theorem 17.

We are ready to prove one of our main results:

Theorem 19. Let $n \geq 3$, and let $\bar{X}$ be a vertex of $\mathcal{E}_n$. Then the property “$C \mapsto \text{strict complementarity}$ holds for (5)” fails generically on $\text{rd}(S_n^+ \cap \{\bar{X}\}^\perp)$.

Proof. By Theorem 3 and the discussion of linear automorphisms of $\mathcal{E}_n$ from Section 2.2, we may assume that $\bar{X} = \mathbf{1} \mathbf{1}^T$. Set

$$m := n - 1.$$ 

Let $Q \in \mathbb{R}^{\times n \times n}$ be an orthogonal matrix such that $Q^T e_n = n^{-1/2} \mathbf{1}$ and $Q^T e_m = 2^{-1/2}(e_1 - e_2)$. Using the map $M \in \mathbb{S}^n \mapsto QMQ^T$ and Corollary 12, we find that $\text{rd}(S_n^+ \cap \{\bar{X}\}^\perp)$ and $\text{rd}(S_n^+ \cap \{e_n e_n^T\}^\perp)$ have the same Hausdorff dimension. Since the cone $S_n^+ \cap \{e_n e_n^T\}^\perp$ is an embedding of $S_n^+$ into $S_n^+$, the Hausdorff dimension of $\text{rd}(S_n^+ \cap \{e_n e_n^T\}^\perp)$ is $\dim_H(S_n^+) - 1$ by Corollary 15. Hence,

$$d := \dim_H \left( \text{rd}(S_n^+ \cap \{\bar{X}\}^\perp) \right) = \binom{n}{2} - 1. \quad (38)$$

Set $\mathcal{C} := \{ C \in \text{rd}(S_n^+ \cap \{\bar{X}\}^\perp) : \text{strict complementarity holds in (5)} \}$. By Theorem 10,

$$\mathcal{C} \subseteq \mathcal{D}_0 \cup \mathcal{D}_{12} \cup \mathcal{D}_{13} \cup \mathcal{D}_{23} \quad (39)$$

where

$$\mathcal{D}_0 := \{ C \in S_n^+ \cap \{\bar{X}\}^\perp : \text{rank}(C) \leq n - 3 \},$$

$$\mathcal{D}_{ij} := \{ C \in S_n^+ : \exists h \in \{1, e_i - e_j\}^\perp, h \neq 0, \text{Null}(C) = \text{span}\{1, h\} \},$$

for each $i, j \in [n]$. Clearly all the sets $\mathcal{D}_{ij}$ have the same $d$-dimensional Hausdorff measures, so it suffices to prove that

$$H_d(\mathcal{D}_0) = 0, \quad (40)$$

$$H_d(\mathcal{D}_{12}) = 0. \quad (41)$$

By using the map $M \in \mathbb{S}^n \mapsto QMQ^T$ and Corollary 12, $\mathcal{D}_0$ and $\{ C \in S_n^+ : \text{rank}(C) \leq m - 2 \}$ have the same $d$-dimensional Hausdorff measure. Hence, (40) follows from Lemma 18 and Corollary 15. Again using the map $M \in \mathbb{S}^n \mapsto QMQ^T$ and Corollary 12, we find that $H_d(\mathcal{D}_{12}) = H_d(\mathcal{D}')$ where

$$\mathcal{D}' := \{ U \in S_n^+ : \text{rank}(U) = m - 1, e_m \in \text{Im}(U) \}.$$ 

Hence, to prove (41) and thus the theorem, it suffices to prove that

$$H_d(\mathcal{D}') = 0. \quad (42)$$
For each $k \in [m - 1]$ define the permutation matrix $P_k := \sum_{i \in [m] \setminus \{k, m\}} e_i e_i^T + e_k e_m^T + e_m e_k^T \in \mathbb{S}^m$. Set $P_m := I$. For each $k \in [m]$ define the map $\varphi_k : \mathbb{S}_{++}^m \oplus \mathbb{R}^{m-1} \to \mathbb{S}^m$ by setting

$$ \varphi_k(A \oplus c) := P_k^T \begin{bmatrix} A & Ac \\ c^T A & c^T Ac \end{bmatrix} P_k. $$

It is easy to verify that

$$ \{ U \in \mathbb{S}_+^m : \text{rank}(U) = m - 1 \} = \bigcup_{k \in [m]} \varphi_k(\mathbb{S}_{++}^m \oplus \mathbb{R}^{m-1}), \quad (43) $$

\[ \text{Null}(\varphi_k(A \oplus c)) = P_k \text{span}\{-c \oplus 1\} \quad \forall A \oplus c \in \mathbb{S}_{++}^m \oplus \mathbb{R}^{m-1}. \quad (44) \]

Let $U \in \mathbb{S}_+^m$ with $\text{rank}(U) = m - 1$, and let $k \in [m]$ and $A \oplus c \in \mathbb{S}_{++}^m \oplus \mathbb{R}^{m-1}$ such that $U = \varphi_k(A \oplus c)$. Then $e_m \in \text{Im}(U)$ is equivalent to $e_m \perp P_k(-c \oplus 1)$, which is equivalent to $k \in [m - 1]$ and $c \perp e_k$. Hence, 

$$ \mathcal{G}' = \bigcup_{k \in [m - 1]} \varphi_k(\mathbb{S}_{++}^m \oplus \{e_k\}). \quad (45) $$

Let $k \in [m - 1]$. Since each entry of $\varphi_k(A \oplus c)$ is (component-wise) polynomial function of the input, the map $\varphi_k$ is Lipschitz continuous on any compact subset of the domain. It follows from Proposition 13 that

$$ \dim_H(\varphi(\mathbb{S}_{++}^m \oplus \{e_k\})) \leq \left(\frac{m}{2}\right) + m = m - 2 = d - 1; \quad (46) $$

note that the subspace $\{e_k\}$ in the LHS is $(m-2)$-dimensional, as this subspace is the set of vectors in $\mathbb{R}^{m-1}$ orthogonal to $e_k$. Now (42) follows from (45) and (46). \hfill \square

## 5. Failure of Strict Complementarity for Rank-One Objectives

In Section 4, we zoomed into the boundary of the normal cone of an arbitrary vertex of the ellipotope and proved that strict complementarity fails generically there. Informally, we might say that with zero “probability” a “uniformly chosen” objective function in the boundary of such normal cone yields an SDP that satisfies strict complementarity. Now we zoom in even further in that boundary, into the set of negative semidefinite rank-one objectives, and consider again how often strict complementarity holds. We will state and prove a self-contained result in Theorem 24 below. However, in order to motivate the objects of the construction and the intermediate results, we start with an informal discussion.

Assume throughout this discussion that $n \geq 4$. We will normalize the “sample space” so that we can have a probability space. For the sake of discussion, let us focus our attention on the vertex $11^T$ of $\mathcal{E}_n$ and consider the sample space to be

$$ \Omega_M := \{ C \in \text{bd(Normal}(\mathcal{E}_n; 11^T)) : C \preceq 0, \text{rank}(C) = 1, \|\text{vec}(C)\|_\infty = 1 \}. \quad (47) $$

Accordingly, equip $\mathbb{S}^n$ with the norm $X \in \mathbb{S}^n \mapsto \|\text{vec}(X)\|_\infty$. Set $d := \dim_H(\Omega_M)$. In order to obtain a probability space on $\Omega_M$, we will define a probability measure

$$ \mathbb{P}_M(\mathcal{A}_M) := \frac{H_d(\mathcal{A}_M)}{H_d(\Omega_M)} \quad (48) $$

over all $H_d$-measurable subsets $\mathcal{A}_M$ of $\Omega_M$; we shall prove that $H_{n-2}(\Omega_M) \in (0, \infty)$, so that (48) is properly defined and $d = n - 2$. Our goal is to prove that the probability of the event

$$ \mathcal{G}_M := \{ C \in \Omega_M : \text{strict complementarity holds for (5) with } C \}. \quad (49) $$

lies in (0, 1).

In order to achieve this, we shall reduce the problem to the space of vectors that generate the rank-one tensors in $\Omega_M$ and $\mathcal{G}_M$, which lie in the matrix space. In order to carry results back and forth between these spaces, we rely on Corollary 12. For each $s \in \{\pm 1\}^n$, define

$$ \mathbb{R}^n_s := \text{Diag}(s) \mathbb{R}^n_+, \quad (50) $$

$$ \varphi_s : b \in \mathbb{R}^n_s \cap \text{bd} ( \mathbb{B}^\infty ) \mapsto -bb^T. \quad (51) $$

Equip $\mathbb{R}^n$ with the norm $x \in \mathbb{R}^n \mapsto \|x\|_\infty$. We shall split our analysis to each of the $2^n$ bi-Lipschitz maps $\varphi_s$, one for each chamber/orthant of $\mathbb{R}^n$, according to their sign vectors:
We claim that such that get that the determinant of Suppose for the sake of contradiction that If

Proposition 21. For

We first prove the inclusion ‘. If

We consider 4 cases, according to which of or is largest, and according to their signs; note that both and have the same sign.

We have

Hence, \( \|x - y\|_{\infty} \leq \|\text{vec}(xx^T - yy^T)\|_{\infty} \). □

Note that restricting the domain of \( \varphi_s \) in Theorem 20 to chambers of \( \mathbb{R}^n \) is necessary. Indeed, consider \( x := (1, -1, \varepsilon)^T \) and \( y := (-1, 1, 0)^T \), for an arbitrary \( \varepsilon \in (0, 1) \). Then \( \|x - y\|_{\infty} = 2 \) but \( \|\text{vec}(xx^T - yy^T)\|_{\infty} = \varepsilon \).

Next we relate the description for \( \Omega_M \) to the vectors that appear in the rank-one tensors:

Proposition 21. For \( n \geq 3 \), we have

\[
\Omega_M = \left\{ -bb^T : b \in \mathbb{R}^n \text{ and either } b = e_i - \alpha e_j \text{ for some distinct } i, j \in [n] \text{ and } \alpha \in [0, 1], \text{ or } (b \perp \mathbb{1}) \text{ and } |b|_{\infty} = 1 \right\}
\] (52)

Proof. We first prove the inclusion ‘\( \supseteq \)’. If \( b \perp \mathbb{1} \) and \( |b|_{\infty} = 1 \), it follows from (11) that \( -bb^T \in \Omega_M \). Suppose that \( b = e_i - \alpha e_j \) for distinct \( i, j \in [n] \) and \( \alpha \in [0, 1] \). Set \( \beta := 1 - \alpha \in [0, 1] \) and \( y := -\beta b \). It is easy to verify that \( S := \text{Diag}(y) + bb^T \succeq 0 \) and \( \mathbb{1}S = 0 \); now \( -bb^T = \text{Diag}(y) - S \in \Omega_M \) follows from (11). In both cases, we rely on \( n \geq 3 \) to ensure that \( -bb^T \) lies in the boundary.

Now we prove the inclusion ‘\( \subseteq \)’. Let \( b \in \mathbb{R}^n \) such that \( -bb^T \in C \in \Omega_M \). Clearly \( |b|_{\infty} = 1 \). We may assume that \( \beta \geq 1 \geq 0 \) and that \( b_1 > 0 \). Use (11) to write \( C = \text{Diag}(y) - S \) for some \( y \in \mathbb{R}^n \) and \( S \in \mathbb{S}_+^n \) such that \( \mathbb{1}S = 0 \). Then \( -\beta b = -bb^T \mathbb{1} = C\mathbb{1} = y - \mathbb{1}S = y \), so

\[
0 \preceq S = \text{Diag}(y) + bb^T = -\beta \text{Diag}(b) + bb^T.
\] (53)

We claim that

\[
b_i < 0 \quad \forall i \in \text{supp}(b) \setminus \{1\}.
\] (54)

Indeed, by restricting (53) to a principal submatrix we get

\[
\begin{bmatrix}
b_1^2 & b_1b_i & b_i^2 \\ b_1b_i & b_i^2 & b_i^2 \\
\end{bmatrix} \succeq \beta \begin{bmatrix}
b_1 & 0 \\ 0 & b_i \\
\end{bmatrix}.
\] (55)

If \( b_i > 0 \), then the RHS is positive definite, whereas the LHS is singular. This proves (54).

Suppose first that \( |\text{supp}(b)| \leq 2 \). Then \( b = e_i - \alpha e_j \) for some \( j \in \text{supp}(b) \setminus \{1\} \) and \( \alpha \in [-1, 1] \). By (54), we have \( \alpha \in [0, 1] \), and so \( -bb^T \) lies in the RHS of (52).

Suppose next that \( |\text{supp}(b)| \geq 3 \). We must prove that

\[
b \perp \mathbb{1}.
\] (56)

Suppose for the sake of contradiction that \( \beta > 0 \). Next let \( i, j \in \text{supp}(b) \setminus \{1\} \) be distinct. Again by (53) we get that the determinant of

\[
\begin{bmatrix}
b_1(b_1 - \beta) & b_1b_i \\ b_1b_i & b_i(b_i - \beta) \\
\end{bmatrix} \succeq 0,
\] (57)
is nonnegative. This yields \( b_1 + b_i \leq \beta \) using \( \beta > 0 \). But (54) implies that \( \beta \leq b_1 + b_i + b_j < b_1 + b_i \), contradiction. This concludes the proof of (56), and hence \(-bb^T\) lies in the RHS of (52). \( \square \)

Finally, we need to relate \( \mathcal{G}_M \) with the vectors that appear in the rank-one tensors. A vector \( b \in \mathbb{R}^n \) is strictly balanced if \( |b_i| < \sum_{j \in [n] \setminus \{i\}} |b_j| \) for every \( i \in [n] \). It is easy to verify that,

if \( b \in \mathbb{R}^n \) and \( i \in [n] \) is such that \( |b_i| = \|b\|_{\infty} \), then \( b \) is strictly balanced \( \iff |b_i| < \sum_{j \in [n] \setminus \{i\}} |b_j| \).

We shall rely on yet another result by Laurent and Poljak:

**Theorem 22** ([14, Theorem 2.6]). Let \( b \in \mathbb{R}^n \) such that \( b \perp 1 \) and \( \text{supp}(b) = [n] \). Then there exists \( X \in \mathcal{E}_n \) such that \( \text{Null}(X) = \text{span}\{b\} \) if and only if \( b \) is strictly balanced.

**Proposition 23.** Let \( b \in \mathbb{R}^n \) such that \( b \perp 1 \) and \( \text{supp}(b) = [n] \). Then strict complementarity holds for (5) with \( C = -bb^T \) if and only if \( b \) is strictly balanced.

**Proof.** Note that \( 11^T \) is an optimal solution for (5) if \( C = -bb^T \). By Proposition 5, we must show that existence of \( X \in \mathcal{E}_n \) such that \(-bb^T \in \text{ri}(\text{Normal}(\mathcal{E}_n; X))\) is equivalent to strict balancedness of \( b \). We will show that, for each \( X \in \mathcal{E}_n \),

\[
- bb^T \in \text{ri}(\text{Normal}(\mathcal{E}_n; X)) \iff bb^T \in \{ Z \in S_n^+ : \text{Im}(Z) = \text{Null}(X) \}.
\]

Since existence of \( X \in \mathcal{E}_n \) such that the RHS of (59) holds is equivalent to \( b \) being strictly balanced by Theorem 22, the result will follow.

The proof of sufficiency in (59) follows from (12) and \( \text{ri}(S_n^+ \cap \{X\}) = \{ Z \in S_n^+ : \text{Im}(Z) = \text{Null}(X) \} \). For the proof of necessity, recall (12) and suppose that there exists \( X \in \mathcal{E}_n \) such that \(-bb^T = \text{Diag}(1) - S \) for some \( y \in \mathbb{R}^n \) and \( S \in \text{ri}(S_n^+ \cap \{X\}) \). Then \( 0 = -bb^T 1 = (\text{Diag}(1) - S) 1 = y - S1 \) shows that

\[
y = S1.
\]

Since \( X \) and \( 11^T \) are optimal solutions for (5), we find that \( 0 = \text{Tr}(-bb^T11^T) = \text{Tr}(-bb^TX) = y^T \text{diag}(X) - \text{Tr}(SX) \) so \( 1^Ty = \text{Tr}(SX) = 0 \). By (60), \( 11^T 1 = 1^Ty = 0 \), so \( 1 \in \text{Null}(S) \) and \( y = 0 \).

We are now in position to present the main result of this section:

**Theorem 24.** Let \( n \geq 4 \) be an integer. Equip \( S^n \) with the norm \( \|\text{vec}(\cdot)\|_{\infty} \). Set

\[
\Omega_M := \{ C \in \text{bd}(\text{Normal}(\mathcal{E}_n; 11^T)) : C \leq 0, \text{rank}(C) = 1, \|\text{vec}(C)\|_{\infty} = 1 \} \subseteq S^n,
\]

\[
\mathcal{G}_M := \{ C \in \Omega_M : \text{strict complementarity holds for (5) with } C \},
\]

\[
d := \dim_H(\Omega_M).
\]

Let \( \Sigma_d \) be the \( \sigma \)-algebra of \( H_d \)-measurable subsets of \( S^n \) and set \( \Sigma_M := \{ \mathcal{A}_M \in \Sigma_d : \mathcal{A}_M \subseteq \Omega_M \} \). Then

(i) \( \Omega_M \in \Sigma_d \) and \( \mathcal{G}_M \in \Sigma_M \),

(ii) \( H_{n-2}(\Omega_M) = (0, \infty) \), so \( d = n - 2 \),

(iii) \( H_d(\mathcal{G}_M) > 0 \) and \( H_d(\overline{\mathcal{G}_M}) > 0 \), where \( \overline{\mathcal{G}_M} := \Omega_M \setminus \mathcal{G}_M \).

In particular, if we set

\[
\mathbb{P}_M(\mathcal{A}_M) := \frac{H_d(\mathcal{A}_M)}{H_d(\Omega_M)} \quad \forall \mathcal{A}_M \in \Sigma_M,
\]

then \( (\Omega_M, \Sigma_M, \mathbb{P}_M) \) is a probability space and the event \( \mathcal{G}_M \) satisfies \( \mathbb{P}_M(\mathcal{G}_M) \in (0, 1) \).

**Proof.** We start by proving that

\[
\Omega_M \in \Sigma_d.
\]

By standard Hausdorff measure theory, \( \Sigma_d \) contains every Borel set of \( S^n \); see, e.g., [20, Theorem 27]. Recall that the Borel sets of \( S^n \) are the elements of the smallest \( \sigma \)-algebra on \( S^n \) that contains all the open subsets of \( S^n \). For distinct \( i, j \in [n] \), set \( \mathcal{B}_{ij} := e_i - [0,1]e_j \). For each \( S \in \binom{[n]}{3} \) and \( m \in \mathbb{N} \setminus \{0\} \), define

\[
\mathcal{B}_{S,m} := \{ b \in \mathbb{R}^n : b \perp 1, \|b\|_{\infty} = 1, |b_i| \geq \frac{1}{m} \forall i \in S \}.
\]

Clearly, each \( \mathcal{B}_{ij} \) and \( \mathcal{B}_{S,m} \) is compact. Let \( \varphi : b \in \mathbb{R}^n \mapsto -bb^T \in S^n \). By Proposition 21,

\[
\Omega_M = \bigcup_{i \in [n]} \bigcup_{j \in [n] \setminus \{i\}} \varphi(\mathcal{B}_{ij}) \cup \bigcup_{m=1}^{\infty} \bigcup_{S \in \binom{[n]}{3}} \varphi(\mathcal{B}_{S,m}).
\]
Since each \( \varphi(\mathcal{B}_{ij}) \) and each \( \varphi(\mathcal{B}_{S,m}) \) is compact, (63) shows that \( \Omega_M \) is an \( F_\sigma \), i.e., a countable union of closed sets, and hence a Borel set. This proves (62).

Next we prove that

\[
H_{n-2}(\Omega_M) \in (0, \infty)
\]  

(64)

from which it will follow via (27) that

\[
d = n - 2.
\]  

(65)

Again we shall use Proposition 21. By Corollary 12 and Theorem 20,

\[
H_1\left(\bigcup_{i \in [n]} \bigcup_{j \in [n] \setminus \{i\}} \varphi(\mathcal{B}_{ij})\right) \in (0, \infty).
\]  

(66)

Moreover,

\[
\Omega_M \supseteq \{-bb^T : b = -1 \oplus c, c \in \mathbb{R}_{++}^{n-1}, 1^T c = 1\} \implies H_{n-2}(\Omega_M) > 0.
\]

For each \( s \in \{\pm 1\}^n \) and \( i \in [n] \), the polytope \( \mathcal{B}_{s,i} := \{ b \in \mathbb{R}^n : b \perp 1, -1 \leq b \leq 1, b_i = s_i \} \) has dimension less than or equal to \( n - 2 \). Since

\[
\Omega_M \subseteq \mathcal{N} \cup \bigcup_{s \in \{\pm 1\}^n} \bigcup_{i \in [n]} \varphi(\mathcal{B}_{s,i})
\]

for some \( \mathcal{N} \) of zero \( d \)-dimensional Hausdorff measure, and each \( \varphi(\mathcal{B}_{s,i}) \) has finite \( d \)-dimensional Hausdorff measure by Corollary 12 and Theorem 20, the proof of (64) is complete.

In the remainder of the proof we shall use subsets of \( \mathbb{R}^n \) with constraints on the coordinates that are zero:

\[
\mathcal{Z}_i := \{ b \in \mathbb{R}^n : b_i = 0 \} \quad \forall i \in [n], \quad \text{and} \quad \mathcal{Z}_\emptyset := \mathbb{R}^n \setminus \bigcup_{i \in [n]} \mathcal{Z}_i = \{ b \in \mathbb{R}^n : \text{supp}(b) = [n] \}.
\]

Define also

\[
\Omega_V := \{ b \in \mathbb{R}^n : b \perp 1, |\text{supp}(b)| \geq 3, \|b\|_\infty = 1 \},
\]

\[
\mathcal{R}_V := \{ b \in \Omega_V : -bb^T \in \mathcal{B}_M \},
\]

\[
\mathcal{R}_V^\perp := \Omega_V \setminus \mathcal{R}_V,
\]

\[
\mathcal{B}_{\text{bal}} := \{ b \in \Omega_V : b \text{ is strictly balanced} \},
\]

\[
\mathcal{B}_{\text{bal}}^\perp := \Omega_V \setminus \mathcal{B}_{\text{bal}}.
\]

Proposition 23 implies that

\[
\mathcal{R}_V \cap \mathcal{Z}_\emptyset = \mathcal{B}_{\text{bal}} \cap \mathcal{Z}_\emptyset,
\]  

(67)

\[
\mathcal{R}_V \cap \mathcal{Z}_i = \mathcal{B}_{\text{bal}} \cap \mathcal{Z}_i.
\]  

(68)

For each \( i \in [n] \), we have \( \mathcal{R}_V \cap \mathcal{Z}_i \subseteq \Omega_V \cap \mathcal{Z}_i \) and the set on the RHS has zero \( d \)-dimensional Hausdorff measure. Hence,

\[
H_d(\mathcal{R}_V \cap \mathcal{Z}_i) = 0 \quad \forall i \in [n].
\]  

(69)

Define \( \varphi_s \) as in (51) for each \( s \in \{\pm 1\}^n \). By putting together (66), (69), and (67), we find that

\[
\mathcal{B}_M = \mathcal{N} \cup \bigcup_{s \in \{\pm 1\}^n} \varphi_s(\mathcal{R}_V \cap \mathcal{Z}_\emptyset \cap \mathbb{R}_s^n) = \mathcal{N} \cup \bigcup_{s \in \{\pm 1\}^n} \varphi_s(\mathcal{B}_{\text{bal}} \cap \mathcal{Z}_\emptyset \cap \mathbb{R}_s^n)
\]  

(70)

for some subset \( \mathcal{N} \subseteq \Omega_M \) such that \( H_d(\mathcal{N}) = 0 \).

Let us prove that

\[
\mathcal{B}_M \in \Sigma_M.
\]  

(71)

For each \( m \in \mathbb{N} \setminus \{0\} \) and each \( U \in \binom{[n]}{3} \), define

\[
\mathcal{B}_{\text{bal},m,U} := \left\{ b \in \mathbb{R}^n : b \perp 1, \|b\|_\infty = 1, |b_i| \geq \frac{1}{m} \forall i \in U, |b_i| + \frac{1}{m} \leq \sum_{j \in [n] \setminus \{i\}} |b_j| \forall i \in [n] \right\}.
\]
Clearly, $\mathcal{R}_{\text{bal}} = \bigcup_{m=1}^{\infty} \bigcup_{U \in \binom{[n]}{3}} \mathcal{R}_{\text{bal},m,U}$. Hence, by (70),

$$
\mathcal{G}_M = \mathcal{N} \cup \bigcup_{m=1}^{\infty} \bigcup_{U \in \binom{[n]}{3}} \varphi_s(\mathcal{R}_{\text{bal},m,U} \cap \mathcal{Z}_\mathcal{G} \cap \mathbb{R}^n_s).
$$

(72)

Since each $\varphi_s(\mathcal{R}_{\text{bal},m,U} \cap \mathcal{Z}_\mathcal{G} \cap \mathbb{R}^n_s)$ is compact, it follows that $\mathcal{G}_M$ is the union of a null set with an $F_x$, and hence $\mathcal{G}_M \in \Sigma_d$. This proves (71).

Set

$$
\hat{x} := 1 \oplus \frac{1}{n-1} \oplus \frac{-n}{(n-1)(n-2)} \mathbb{1} \in \mathbb{R}^n, \quad \varepsilon := \frac{3}{4(n-1)(n-2)}, \quad \text{and} \quad s(x) := 1 \oplus 1 \oplus -\mathbb{1} \in \{\pm 1\}^n.
$$

It is not hard to verify that

$$
\hat{x} + \varepsilon (\mathbb{B}_{\infty} \cap \{e_1, 1\}^\perp) \subseteq \mathcal{R}_{\text{bal}} \cap \mathcal{Z}_\mathcal{G} \cap \mathbb{R}^n_s(x).
$$

(73)

Since the set in the LHS of (73) has positive $d$-dimensional measure, so does the set in the RHS of (73), whence

$$
H_d(\mathcal{G}_M) > 0
$$

(74)

by Corollary 12, Theorem 20, and (70).

Set

$$
\hat{y} := 1 \oplus -\frac{1}{n-1} \mathbb{1} \in \mathbb{R}^n, \quad \delta := \frac{1}{2(n-1)}, \quad \text{and} \quad s(y) := 1 \oplus -\mathbb{1} \in \{\pm 1\}^n.
$$

It is not hard to verify that

$$
\hat{y} + \delta (\mathbb{B}_{\infty} \cap \{e_1, 1\}^\perp) \subseteq \mathcal{R}_{\text{bal}} \cap \mathcal{Z}_\mathcal{G} \cap \mathbb{R}^n_s(y).
$$

(75)

Hence,

$$
\mathcal{G}_M \supseteq \varphi(\mathcal{N} \cap \mathcal{Z}_\mathcal{G}) = \varphi(\mathcal{R}_{\text{bal}} \cap \mathcal{Z}_\mathcal{G}) \supseteq \varphi_s(\mathcal{R}_{\text{bal}} \cap \mathcal{Z}_\mathcal{G} \cap \mathbb{R}^n_s(y)) \supseteq \varphi_s(y) (\hat{y} + \delta (\mathbb{B}_{\infty} \cap \{e_1, 1\}^\perp)).
$$

Thus,

$$
H_d(\mathcal{G}_M) > 0
$$

by Corollary 12 and Theorem 20.

6. Conclusion

We proved in Section 4 that the MaxCut SDP (5) has the worst possible behavior with respect to strict complementarity when the objective function is in the boundary of the normal cone of the ellipotope at any of its vertices. At a first glance, this may seem surprising since the MaxCut SDP is so elementary and has so many favorable properties. However, as we explain next, from a properly chosen viewpoint this bad behavior is not so surprising.

Consider, for instance, the convex set $\mathcal{C} \subseteq \mathbb{R}^2$ in Figure 1. For concreteness, an explicit description of $\mathcal{C}$ is given by

$$
\mathcal{C} := \{ x \in \mathbb{R}^2 : \|x\| + |x_1| \leq 1 \} = \{ x \in \mathbb{R}^2 : |x_1| \leq 1/2, |x_2| \leq \sqrt{1 - 1/2|x_1|} \},
$$

(76)

and it is not hard to show that $\mathcal{C}$ is the projection of the feasible region of an SDP. It is intuitive and simple to verify that $\mathbb{1}$ lies in (the boundary of) the normal cone of $\mathcal{C}$ at its vertex $e_2$, but $\mathbb{1}$ is not in the relative interior of any normal cone of $\mathcal{C}$. We can trace this phenomenon to the smooth, nonpolyhedral boundary of $\mathcal{C}$ around $e_2$. It is straightforward to extend this example to $\mathbb{R}^3$ by considering the solid of revolution obtained by rotating $\mathcal{C}$ around the $e_2$ axis, i.e., an American football.

The ellipotope looks somewhat similar to $\mathcal{C}$ in the following sense. Let us consider the projection $\mathcal{C}'_{n} \subseteq \mathbb{R}^{n}_2$ of the ellipotope $\mathcal{E}_n$ into its off-diagonal entries. For $n \geq 3$, the set $\mathcal{C}'_{n}$ is a compact nonpolyhedral convex set with $2^{n-1}$ vertices by Theorem 3. Intuitively, $\mathcal{C}'_{n}$ can be thought of as being obtained from the polytope which is the convex hull of these $2^{n-1}$ vertices by inflating it like a balloon, while preserving the vertices fixed. (In fact, by [13, Proposition 2.9], the line segments between the $2^{n-1}$ vertices are also kept fixed.) In this way, $\mathcal{C}'_{n}$ is a round, plump convex set, whose boundary is smooth almost everywhere, and the neighborhood of $\mathcal{C}'_{n}$ around any vertex looks like (a generalization of) what is depicted by the set $\mathcal{C}$ from the previous paragraph. Thus, when one considers that the ellipotope around a vertex “locally” looks like $\mathcal{C}$ around $e_2$, the poor behavior of the MaxCut SDP described in Section 4 makes more intuitive sense. The discussion above
indicates a natural direction for future research. Namely, to extend Theorem 19 to more general SDPs, by requiring the feasible region to be “locally nonpolyhedral” around its vertices.

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