On strong duality, theorems of the alternative, and projections in conic optimization

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Abstract

A conic program is the problem of optimizing a linear function over a closed convex cone intersected with an affine preimage of another cone. We analyse three constraint qualifications, namely a Closedness CQ, Slater CQ, and Boundedness CQ (also called Clark-Duffin theorem), that are sufficient for achieving strong duality and show that the first implies the second which implies the third, and also give a more general form of the third CQ for conic problems. Furthermore, two consequences of strong duality are presented, the first being a theorem of the alternative on almost feasibility (also called weak infeasibility), and the second being an explicit description of the projection of conic sets onto linear subspaces, akin to using projection cones for polyhedral sets.

Keywords. Closedness of adjoint image · Slater constraint qualification · Clark-Duffin theorem · Theorem of Alternative · Projection onto subspace · Bounded feasible set

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1 Introduction

Let $C \subseteq \mathbb{E}$ and $K \subseteq \mathbb{E}'$ be nonempty closed convex cones in Euclidean spaces $\mathbb{E}$ and $\mathbb{E}'$ having respective inner-products $\langle \cdot, \cdot \rangle_{\mathbb{E}}$ and $\langle \cdot, \cdot \rangle_{\mathbb{E}'}$. For a linear map $\mathcal{A}: \mathbb{E} \to \mathbb{E}'$ and vectors $b \in \mathbb{E}'$ and $c \in \mathbb{E}$, the conic optimization problem is

$$z^*_P = \sup \{ \langle c, x \rangle_{\mathbb{E}} : \mathcal{A}x \preceq_K b, \ x \in C \}.$$  \hfill (1a)

where the constraints $\mathcal{A}x \preceq_K b$ mean that $b - \mathcal{A}x \in K$. Because $K$ can be a Cartesian product of finitely many cones, the conic inequalities $\mathcal{A}x \preceq_K b$ can incorporate multiple constraints over different cones. Equality constraints can be represented as conic constraints with respect to the singleton cone $\{0\}$. The conic constraint $x \in C$ is kept separate in (1a) because it involves a special linear map (identity map) and constant vector being all zeros. For nontriviality, we assume that $C \neq \{0\}$ and that at least one of $C$ or $K$ is not equal to its ambient Euclidean space. The Lagrangian dual problem is the conic program

$$z^*_D = \inf \{ \langle b, y \rangle_{\mathbb{E}'} : \mathcal{A}^*y \succeq_{C^*} c, \ y \in K^* \}.$$  \hfill (1b)

for the adjoint linear map $\mathcal{A}^*: \mathbb{E}' \to \mathbb{E}$ and dual cones $K^* \subseteq \mathbb{E}'$ and $C^* \subset \mathbb{E}$. The primal-dual pair are symmetric since the conic dual of (1b) yields (1a).

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Duality gap is the nonnegative difference $z_D^* - z_P^*$. Strong duality means that $z_P^* = z_D^*$ (allowing for values $\pm \infty$), and if these values are finite then at least one of the two problems is solvable (i.e., has an optimal solution). When both $K$ and $C$ are polyhedral, feasibility of one of the problems guarantees strong duality, but this is not true for general $K$ and $C$. There is abundant literature on sufficient conditions for strong duality in conic programs, see [NN92; LSZ97; Ram97; RTW97; BN01; Gli01; Sha01; ART02; STO07; Zål08; TW12; Pat13; KT21].

The most well-known of these conditions is the generalised Slater constraint qualification (CQ) which requires that one of the two problems be strictly feasible (in the sense of satisfying the conic constraints through their relative interiors). There are many proofs of sufficiency of Slater CQ, and it also follows from the classical Fenchel duality theorems [Roc70, chap. 31]. There have been studies on the geometry of Slater CQ and degeneracy of solutions [PT01; DW17; DJS17]. Primal problem (1a) can be extended to what is called an abstract convex program in Banach spaces by considering a $K$-convex function $g(x)$ and replacing $b - Ax \in K$ with $-g(x) \in K$; strong duality results for these can be found in [BW81; JW90; BKW05; BGW07; JL08; BS09].

In this paper, we make two observations about sufficient conditions for strong duality between (1a) and (1b). We add to the vast literature on Slater CQ by showing that it follows from a more general Closedness CQ shown by [Sha01; Bar02] but which is probably not as well-known. Our argument relies on known sufficient conditions for closedness of linear image of a closed convex cone and noting a property about perspective images of a convex cone. We also give another proof for the sufficiency of the Boundedness CQ, which has been called the Clark-Duffin theorem [Duf78] for convex programs with inequality constraints. A restricted version of this with $K = \{0\}$ and $C$ being a proper cone appeared in [STO07, Theorem 3.15] in the context of universal duality. Our derivation relies on a conic version of the Gordan’s theorem of the alternative. Given the importance of these two CQs, we also present several necessary/sufficient conditions for them to hold.

Another set of contributions are obtaining consequences of strong duality through Slater CQ. We give a theorem of the alternative (Theorem 5.1) for almost feasibility of a problem, and give an explicit description of the projection of a conic set onto a linear subspace (Theorem 6.1). The former generalises [BN01, Proposition 1.7.1] which deals with a proper cone $K$ and $C = E$, and the latter extends the notion of projection cones for polyhedral sets.

§2 specifies the notation used in this paper, basic definitions and some fundamental results from convex analysis. §3 gives some detailed observations about conic optimization, in particular, a basis representation of the dual that implies any of the duality results can be translated to be in terms of the span of $C$, and characterisations of the recession cone of $X$ and its polar cone. The three sufficient conditions for strong duality — Closedness CQ, Slater CQ and Boundedness CQ, are analysed in §4 for their inter-relationships and conditions for them to hold. §5 gives a new theorem of the alternative and §6 projection of conic sets. Primal-dual symmetry means that results for one problem can be directly extended to the other problem, and so we generally prove our results only for the primal and omit analogous statements/proofs for the dual.

\section{Preliminaries}

\subsection{General Notation & Terminology}

A Euclidean space is a finite-dimensional inner-product space over the reals. Unless there is ambiguity, we drop the subscripts in the notation for inner-products. The vector of all zeros is written as $0$. The Minkowski sum of two sets $S_1, S_2 \subset E$ is $S_1 + S_2$, their Cartesian product (equiv. direct sum) is $S_1 \times S_2$, and we write the Minkowski difference as $S_1 - S_2$ and define it as
the Minkowski sum $S_1 + (-S_2)$. An ordered pair of $x \in \mathbf{E}$ and $y \in \mathbf{E}'$ is the tuple $(x, y)$ which belongs to the product space $\mathbf{E} \times \mathbf{E}'$. For a set $S \subseteq \mathbf{E}$, $\text{ri} S$ is the relative interior, $\text{int} S$ is the interior, $\text{cl} S$ is the closure, $\partial S := \text{cl} S \setminus \text{ri} S$ is the relative boundary, $\text{aff} S$ is the affine hull, $\text{sub} S$ is the linear subspace parallel to $\text{aff} S$ (equal to $\text{aff} S - x$ for any $x \in S$), and $\text{span} S$ is the linear hull (span). The orthogonal complement of a linear subspace $L \subseteq \mathbf{E}$, also called its annihilator, is denoted by $L^\perp$. The recession cone of a nonempty convex set $S \subseteq \mathbf{E}$ is the convex cone $0^+ S := \{ r \in \mathbf{E} : x + \mu r \in S, \forall x \in S, \mu \geq 0 \}$; when $S$ is also closed, it is known that $0^+ S = \{ r \in \mathbf{E} : \exists x \in S \text{ s.t. } x + \mu r \in S, \forall \mu \geq 0 \}$. The lineality space of $S$ is defined as $\text{lin} S := 0^+ S \cap -0^+ S$, and when this is trivial (equal to $\{0\}$) the set is called pointed.

The polar cone and the dual cone of a convex cone $C$ are denoted by $C^\circ$ and $C^*$. For a convex cone $C$ containing $0$, the binary relation $\preceq_C$ is a quasi-ordering which becomes a partial order when $C$ is pointed. It is additive equivariant and distributes over the Cartesian product of cones. Strict inequality for the binary relation $\preceq_C$ is written as $x \prec_C y$ and is defined as $y - x \in \text{ri} C$.

For a map $L: \mathbf{E} \rightarrow \mathbf{E}'$, the image of a set $S \subseteq \mathbf{E}$ is $L(S) := \{ L(x) : x \in S \}$, the image of $L$ is $\text{Im} L := L(\mathbf{E})$, the preimage of a set $S \subseteq \mathbf{E}'$ is $L^{-1}(S) := \{ x \in \mathbf{E} : L(x) \in S \}$. We work with linear and affine maps only. The kernel of a linear map $L$ is $\ker L := L^{-1}(\{0\})$. The adjoint of a linear map $L$ is a unique linear map $L^*: \mathbf{E}' \rightarrow \mathbf{E}$ which satisfies $\langle L(x), y \rangle_{\mathbf{E}} = \langle L^*(y), x \rangle_{\mathbf{E}}$ for all $x \in \mathbf{E}, y \in \mathbf{E}'$. Affine maps are translates of linear maps. For every affine map $G$ we have the associated linear map $L_G(x) := G(0) - G(x)$.

The feasible sets of the primal (1a) and dual (1b) are denoted by

\begin{align*}
X &= X(b) := \{ x \in C : Ax \preceq_K b \} = \{ x \in C : \exists s \in K \text{ s.t. } s = b - Ax \}, \quad (2a) \\
Y &= Y(c) := \{ y \in K^* : A^* y \succ_C c \} = \{ y \in K^* : \exists w \in C^* \text{ s.t. } w = A^* y - c \}. \quad (2b)
\end{align*}

These are parametrised by their respective right-hand sides because it will be sometimes be necessary to refer to them in the parametric form. We will use the parametric and non-parametric forms as appropriate depending on the context. The feasible sets can also be expressed as affine preimages of closed convex sets,

\begin{align*}
X &= G_p^{-1}(K \times C), \quad Y = G_d^{-1}(C^* \times K^*). \quad (2c)
\end{align*}

where $G_p: x \mapsto (b - Ax, x)$ and $G_d: y \mapsto (A^* y - c, y)$. The adjoints of the corresponding linear maps $L_{G_p} = (Ax, -x)$ and $L_{G_d} = (-A^* y, -y)$ are

\begin{align*}
L_{G_p}^*(y, w) &= A^* y - w, \quad L_{G_d}^*(x, s) = -Ax - s. \quad (2d)
\end{align*}

The sets of feasible right-hand sides for the primal and dual are represented by the following two convex cones

\begin{align*}
C_p := A(C) + K = -L_{G_d}^*(C \times K), \quad C_d := A^*(K^*) - C^* = L_{G_p}^*(K^* \times C^*). \quad (3)
\end{align*}

**Observation 2.1.** $C_p = \{ b \in \mathbf{E}' : X(b) \neq \emptyset \}$ and $\text{ri} C_p = \{ b \in \mathbf{E}' : \exists x \in \text{ri} C \text{ s.t. } Ax \prec_K b \}$. Similarly, $C_d = \{ c \in \mathbf{E} : Y(c) \neq \emptyset \}$ and $\text{ri} C_d = \{ c \in \mathbf{E} : \exists y \in \text{ri} K^* \text{ s.t. } A^* y \succ_C c \}$.

For completeness, a proof of this is given in Appendix A.
2.2 Convexity

Every nonempty convex set in $\mathbf{E}$ has a nonempty relative interior. The following well-known results from convex analysis about closure and relative interior will be used as standard facts throughout this paper without necessarily referencing them every time they are used.

Lemma 2.2 (cf. [Roc70]). Let $S$ be a nonempty convex set and $G$ be an affine map.

1. $\text{ri} S = \text{ri}(\text{cl} S) \subseteq \text{cl}(\text{ri} S) = \text{cl} S$.
2. $\text{ri} G(S) = G(\text{ri} S) \subseteq G(\text{cl} S) \subseteq G(S)$, with $G(\text{cl} S) = \text{cl} G(S)$ when $G(\text{cl} S)$ is closed.
3. $G^{-1}(\text{cl} S) \supseteq \text{cl} G^{-1}(S) \supseteq \text{ri} G^{-1}(S) \supseteq G^{-1}(\text{ri} S)$, with $\text{ri} G^{-1}(S) = G^{-1}(\text{ri} S)$ and $\text{cl} G^{-1}(S) = \text{cl} G^{-1}(\text{ri} S)$ when $G^{-1}(\text{ri} S) \neq \emptyset$.
4. For two convex sets $S_1, S_2 \neq \emptyset$, we have
   
   (a) $\text{ri} (S_1 \times S_2) = \text{ri} S_1 \times \text{ri} S_2$ and $\text{cl} (S_1 \times S_2) = \text{cl} S_1 \times \text{cl} S_2$.
   (b) if $\text{ri} S_1 \cap \text{ri} S_2 \neq \emptyset$, then $\text{ri} (S_1 \cap S_2) = \text{ri} S_1 \cap \text{ri} S_2$ and $\text{cl} (S_1 \cap S_2) = \text{cl} S_1 \cap \text{cl} S_2$.
   (c) $\text{ri} (S_1 \pm S_2) = \text{ri} S_1 \pm \text{ri} S_2$.

The Minkowski sum of closed sets is not closed in general, see [Roc70, chap. 9] for sufficient conditions for the sum to be closed.

Now let $C$ be a closed convex cone. Since $0 \in C$, the affine hull and span of a cone are equal and written as $\text{span} C$, the orthogonal complement of $\text{span} C$ is written as $C^\perp$. The following relationships can be used to obtain dual counterparts of the conditions/assertions on the primal problem.

Lemma 2.3 (cf. [LSZ97, Corollary 1]). $C^\perp = C^* \cap C^0$, and $\text{span} C^0 = (\text{lin} C)^\perp$.

A pointed (resp. full-dimensional) $C$ is equivalent to $C^0$ being full-dimensional (resp. pointed). Another basic result on polarity is an identity for the polar of a linear preimage of a convex cone. This is also related to the Farkas lemma for conic linear systems [cf. DJ14, Theorem 2.1] which states that for any $c \in \mathbf{E}$, exactly one of the following holds: either $c \in \text{cl} \mathbf{L}^*(C^*)$ or there exists $y \in \mathbf{E}$ such that $\langle c, y \rangle > 0$ and $\mathbf{L}(y) \preceq_C 0$.

Lemma 2.4 (cf. [LSZ97, Lemma 3]). A linear map $\mathbf{L}$ and closed convex cone $C$ satisfy\footnote{The first identity follows from the second identity after applying the Bipolar Theorem [Roc70, Theorem 14.5] to the set $\mathbf{L}^{-1}(-S)$ (which contains $0$ due to $0 \in S$) and using the fact that a convex set and its closure have the same polar.}

\[ \left( \mathbf{L}^{-1}(-C) \right)^\circ = \text{cl} \mathbf{L}^*(C^*), \quad \mathbf{L}^{-1}(-C^\perp) = \left( \mathbf{L}^*(C^*) \right)^\circ. \]

Linear subspaces are closed convex cones, but we may sometimes need to exclude such pathological cones, which can be characterised as follows:

$C$ is a linear subspace $\iff C = \text{lin} C \iff C^*$ is a linear subspace $\iff C^* = C^\perp$. \ (4)

For a non-subspace cone the origin is not contained in the relative interior of the cone or that of the dual cone\footnote{This is a generalization of $0 \notin \text{int} C^*$ when $C$ is pointed which is directly due to the fact that a closed convex set containing $0$ is unbounded if and only if $0$ is not in the interior of its polar set [Roc70, Corollary 14.5.1]}

Lemma 2.5. If $C$ is not a linear subspace and not equal to $\{0\}$, then
1. \( \mathcal{C}^\perp \subseteq \partial \mathcal{C}^* \) and \( \operatorname{lin} \mathcal{C} \subseteq \partial \mathcal{C} \),
2. \( \mathbf{0} \notin \operatorname{ri} \mathcal{C} \cup \operatorname{ri} \mathcal{C}^* \).

We also have a conic version of Gordan’s theorem of the alternative from linear programming.

**Lemma 2.6** (Conic version of Gordan’s theorem). Consider the following two statements:

1. there exists \( x \in C \setminus \{0\} \) such that \( -Ax \in K \),
2. there exists \( y \in \operatorname{ri} K^* \) such that \( A^*y \in \operatorname{ri} C^* \).

If (1) is not true, then (2) must be true. Furthermore, if at least one of \( A^{-1}(K) \) or \( C \) is not a subspace, then exactly one of the two statements is true.

Different versions of this theorem of the alternative have appeared for e.g. in [LSZ97, Corollary 2] and [STO07, Lemma 2.1], but our version is more general as it allows for two cones \( K \) and \( C \) and also notes the non-subspace requirement to obtain equivalence. Note that the non-subspace condition is needed to obtain the reverse direction \( (2) \implies \neg (1) \), because if this assumption does not hold, then the simple example from linear programming where the two systems are \( \{(x_1, x_2) \in \mathbb{R}^2 : x_1 \leq 0, -x_1 \leq 0\} \) and \( \{(y_1, y_2) \in \mathbb{R}_+^2 : y_1 = y_2\} \), tells us that both statements can be simultaneously true.

The proofs for the previous two lemmas are given in Appendix A for completeness.

## 3 Basic Results

Assuming feasibility, note that \( z^*_P = +\infty \) if \( c \notin (\operatorname{lin} X)^\perp \) and \( z^*_D = -\infty \) if \( b \notin (\operatorname{lin} Y)^\perp \). These conditions are trivially satisfied when the respective feasible sets are pointed, but they are not necessary for unboundedness. Because \( (\operatorname{lin} X)^\perp = (\operatorname{lin} X)^* \cap (\operatorname{lin} X)^\circ \subset (0^+ X)^\circ \) and similarly \( (\operatorname{lin} Y)^\perp \subset (0^+ Y)^\circ \), another sufficient condition for unboundedness that is also not necessary in general is \( c \notin (0^+ X)^\circ \) and \( b \notin (0^+ Y)^\circ \), respectively. Trivial cases for computing \( z^*_P \) and \( z^*_D \) would be when the objective functions are constant-valued over the respective feasible sets. For the primal, \( \langle c, x \rangle \) is constant-valued over \( X \) if and only if there exists some \( u \in E \) such that \( u + c^\perp \supseteq X \), and for the dual, \( \langle b, y \rangle \) is constant-valued over \( Y \) if and only if there exists some \( u \in E' \) such that \( u + b^\perp \supseteq Y \).

Now let us give some detailed observations on conic optimization.

### 3.1 Basis Representation of the Dual

Ordinarily, as seen in (1b), we use the adjoint of \( A \) to write the constraints of the dual. When given a basis for the span of \( C \), the dual problem can be written using a linear map that depends on the basis, instead of using the adjoint of \( A \).

**Proposition 3.1.** Let \( B = \{v^1, \ldots, v^m\} \) be an orthonormal basis of \( \operatorname{span} C \) and denote the linear map \( B : y \in E' \mapsto \sum_{j=1}^m \langle Av^j, y \rangle v^j \in \operatorname{span} C \). We have \( z^*_D = \inf \{\langle b, y \rangle : By \geq_C c, y \in K^* \} \).

**Proof.** Any \( x \in C \) can be written as \( x = \sum_{i=1}^m \alpha_i v^i \) for some \( \alpha \in \mathbb{R}^m \). We have
\[
\langle x, By \rangle_E = \left\langle \sum_{i=1}^m \alpha_i v^i, \sum_{j=1}^m \langle Av^j, y \rangle v^j \right\rangle_E = \sum_{i=1}^m \sum_{j=1}^m \alpha_i \langle Av^j, y \rangle v^i \langle v^j, v^i \rangle_E = \sum_{j=1}^m \alpha_j \langle Av^j, y \rangle v^j = \left\langle A \left( \sum_{j=1}^m \alpha_j v^j \right), y \right\rangle_{E'},
\]
5
where the penultimate equality is because of orthonormality of the basis. Thus,

\[
\langle Ax, y \rangle_{E'} = \langle x, B y \rangle_E, \quad x \in C, y \in E'.
\]

(5)

Since \( z^*_D \) is the Lagrangian dual of the primal, we have \( z^*_D = \inf_{y \in K^*} \sup_{x \in C} \langle c, x \rangle_E + \langle y, b - Ax \rangle_{E'} \). The inner maximization objective becomes \( \langle b, y \rangle_{E'} + \langle c, x \rangle_E - \langle y, Ax \rangle_{E'} \), and then equation (5) transforms this to \( \langle b, y \rangle_{E'} + \langle c - By, x \rangle_E \). Hence, the inner maximum is finite (and equal to zero) if and only if \( c - By \in C^\circ \), which is the dual constraint \( By \geq_{C^*} c \). □

The implication of this is that any of the results in this paper that make use of the dual constraints can be restated in terms of the basis form of the dual constraints using a basis for the span of \( C \).

### 3.2 Recession Cone

Our second basic result is characterising the recession cone of the feasible regions. This will also be helpful later in the paper when deriving conditions for boundedness since a nonempty closed convex set is bounded if and only if its recession cone contains nonzero elements.

**Lemma 3.2.** For \( X \neq \emptyset \), we have \( 0^+X = X(0) = -A^{-1}(K) \cap C \) and

\[
(X(0))^\circ = \text{cl} \mathcal{C}_d, \quad \text{and} \quad \text{ri} (X(0))^\circ = \text{ri} \mathcal{C}_d.
\]

Furthermore, \( (X(0))^\circ = \mathcal{C}_d \) when there exists \( x \in \text{ri} C \) with \( -Ax \in \text{ri} K \).

**Proof.** We will use the fact that for an affine map \( \mathcal{G} \), the \( 0^+ \) operator commutes with the linear map associated with \( \mathcal{G} \), i.e., \( 0^+ \mathcal{G}^{-1}(S) = -\mathcal{G}^{-1}(0^+ S) \), where \( \mathcal{G} \) is an affine map with \( \mathcal{G}^{-1}(S) \neq \emptyset \) for a closed convex set \( S \), and \( \mathcal{L}_{\mathcal{G}} \) being cones leads to \( 0^+ X = -\mathcal{L}_{\mathcal{G}^{-1}}(0^+(K \times C)) \). Using the affine preimage expression for \( X \) from (2c), the above claim yields \( 0^+X = -\mathcal{L}_{\mathcal{G}^{-1}}(0^+(K \times C)) \). Distributivity of \( 0^+ \) over Cartesian product and \( K \) and \( C \) being cones leads to \( 0^+X = -\mathcal{L}_{\mathcal{G}^{-1}}(K \times C) \), which means that \( 0^+X = \{ x \in C : Ax \preceq_K 0 \} \). The set on the right-hand side is equal to \( X(0) \) from (2) and can be rewritten as \( -A^{-1}(K) \cap C \).

Since \( X(0) = \mathcal{L}_{\mathcal{G}^{-1}}(\mathcal{L}_{\mathcal{G}^{-1}}(-K \times -C)) \), Lemma 2.4 implies that the polar of \( X(0) \) is equal to the closure of \( \mathcal{L}_{\mathcal{G}^{-1}}(K^* \times C^*) \), and from (3) we know that this is the closure of \( \mathcal{C}_d \). The equality for relative interiors follows from the assertion on closure because of the first equality in Lemma 2.2 for arbitrary convex sets.

For the last part, it suffices to argue that \( \mathcal{C}_d \) is a closed set. The existence of a strict solution in the recession cone means that \( A^{-1}(\text{ri} K) \neq \emptyset \), and then known conditions for closeness of an adjoint image (cf. Lemma 4.5) give us that \( A^*(K^*) \) is a closed cone. Therefore, \( \mathcal{C}_d \) is a Minkowski difference of two closed cones. Lemma 2.4 tells us that \( \mathcal{C}_d \) is the Minkowski sum of the polars of \( A^{-1}(-K) \) and \( C \). For this sum to be closed, a sufficient condition from literature (cf. [Pat07]) is that the relative interiors of \( A^{-1}(-K) \) and \( C \) intersect. We have \( \text{ri} A^{-1}(-K) \supseteq A^{-1}(-\text{ri} K) \). Strict feasibility implies \( A^{-1}(-\text{ri} K) \cap \text{ri} C \neq \emptyset \), and therefore, \( \mathcal{C}_d \) is a closed set. □

**Corollary 3.3.** A nonempty \( X \) is bounded if and only if for any norm \( \| \cdot \| \) the convex maximization problem \( \max \{ \| x \| : Ax \preceq_K 0, \ x \in C \} \) has value greater than 0.

Henceforth, whenever we refer to a feasible set being bounded, it is assumed that the set is nonempty.
3.3 Objective Qualification

The third basic result to note here is a special case of strong duality that does not require any constraint qualification, instead it requires the objective vectors to belong to specific parts of feasibility cones of the other problem.

Observation 3.4. If the primal is feasible, then strong duality holds when \( c \in A^*(K^\perp) \).

Proof. Let \( c = A^*(y) \) for some \( y \in K^\perp \). Because \( A^*(K^\perp) \subseteq A^*(K^\star) \subseteq C_d \), we have \( c \in C_d \) and hence the dual is feasible with \( y \in Y \). Every \( x \in X \) has \( Ax + s = b \) for some \( s \in K \). Then, \( \langle c, x \rangle = \langle A^*(y), x \rangle = \langle y, Ax \rangle = \langle y, b - s \rangle = \langle y, b \rangle - \langle y, s \rangle = \langle y, b \rangle \), where the last equality is due to \( y \in K^\perp \) and \( s \in K \). Thus, there is zero duality gap and solvability of the dual. \( \square \)

After using Lemma 2.3, the analogous condition for the dual becomes \( b \in A(\text{lin } C) \).

4 Closedness CQ and its Consequences

Consider the linear maps

\[
L_p: (\alpha, \alpha_0) \in E \times \mathbb{R} \mapsto (A\alpha + \alpha_0 b, -\alpha) \in E' \times E, \quad L_p^*: (y, w) \mapsto (A^* y - w, \langle b, y \rangle). \tag{6}
\]

The convex cone \( L_p^*(K^\star \times C^\star) \) may not be a closed set in general because linear images of closed sets are not closed (cf. Lemma 4.5). Closedness of this adjoint image is known to be a sufficient condition for strong duality.

Theorem 4.1 (Shapiro [Sha01, Proposition 2.6], Barvinok [Bar02, Theorem 7.2]). Assume the primal is feasible. Strong duality holds, with the dual being solvable, when

\[
L_p^*(K^\star \times C^\star) \text{ is a closed set.} \tag{7a}
\]

Remark 1. When the dual is feasible, primal-dual symmetry implies that an analogous sufficient condition for strong duality is

\[
L_d^*(C \times K) \text{ being a closed set, where } L_d^*: (x, s) \mapsto (Ax + s, \langle c, x \rangle) \tag{7b}
\]

is the adjoint of the linear map \( L_d: (\beta, \beta_0) \in E' \times \mathbb{R} \mapsto (A^* \beta + \beta_0 c, \beta) \in E \times E' \). We refer to closedness of \( L_p^*(K^\star \times C^\star) \) as the dual condition because the adjoint image relates to dual feasibility and optimality, and closedness of \( L_d^*(C \times K) \) as the primal condition.

Remark 2. The Closedness CQs in (7) are different than another closedness CQ given by Boţ and Wanka [BW06, Theorem 3.2]; the latter was shown to be necessary and sufficient for strong duality in abstract convex programs. It is not clear whether (7) is also necessary for strong duality; we leave its relationship to the CQ of Boţ and Wanka as an open question.

The proofs of Theorem 4.1 by Shapiro and by Barvinok are different in their approaches — the former uses functional analysis arguments for the value function whereas the latter uses geometric arguments using the separation theorem for closed convex sets. For the sake of exposition, we give a sketch of the latter proof by breaking down its main components.

Remark 3 (On Barvinok’s proof of Theorem 4.1). The proof can be essentially broken down into three main steps. For convenience, denote \( S := L_p^*(K^\star \times C^\star) \).
1. First is to argue that when $z^*_D$ is finite, which happens when primal is assumed to be feasible, then $(c, z^*_D)$ belongs to the closure of $S$. This is shown by observing that $z^*_D$ is the infimum of values over a closed interval of the real line; in particular, $z^*_D = \inf_{t \in I} \{ t : t \in I \}$ where $I = \{ t \in \mathbb{R} : (c, t) \in \text{cl } S \}$.

2. Note that although $(c, z^*_D)$ belongs to $\text{cl } S$, there is no guarantee that this point belongs to $S$, and this inclusion is equivalent to solvability of the dual due to the definition of the adjoint image in $(6)$. In particular, note that if $(c, z) \in S$, then it must be that $c \in C_d$ and $z \geq z^*_D$; this is because of the definition of $C_d$ in $(3)$ and $z^*_D = \inf \{ \langle b, y \rangle : c \in C_d \}$. Assuming closedness of the adjoint image gives $(c, z^*_D) \in S$, and since this is dual solvability, it follows that $(c, z^*_D - \varepsilon) \notin S$ for every $\varepsilon > 0$, because otherwise we would have a contradiction to the optimality of the solution.

3. The third, and main, step uses separation theorem to show that the condition $(c, z^*_D - \varepsilon) \notin \text{cl } S$, for some $\varepsilon > 0$, is sufficient to guarantee an $\varepsilon$-gap between the primal and dual, i.e., $z^*_D - z^*_P \leq \varepsilon$. Using the closedness assumption and the second claim that $(c, z^*_D - \varepsilon) \notin S$ for every $\varepsilon > 0$, leads to $z^*_D - z^*_P \leq \inf \{ \varepsilon : \varepsilon > 0 \} = 0$, and then $z^*_D - z^*_P \geq 0$ from weak duality implies that $z^*_P = z^*_D$, thereby establishing strong duality. $\diamond$

The third claim in the above remark is a topological necessary condition for one problem to have finite value and the other problem to be infeasible (resulting in infinite duality gap).

**Corollary 4.2.** Suppose that the dual is solvable and the primal is infeasible. Then it must be that for every $\varepsilon > 0$ we have $(c, z^*_D - \varepsilon) \in \partial S$, where $\partial S := \text{cl } S \setminus S$ is the boundary of $S := \mathcal{L}_p^*(K^* \times C^*)$.

Let us illustrate this necessary condition with a family of examples that have infinite duality gap in every dimension.

**Example 1.** For arbitrary $n \geq 3$, let $A = \begin{bmatrix} I_{n-1} & e_1 \end{bmatrix}$ be an $(n - 1) \times n$ matrix, where $I_{n-1}$ is an identity matrix and $e_1$ is the column vector $(1, 0, \ldots, 0)^\top \in \mathbb{R}^{n-1}$, and $\mathcal{C}_n = \{ x \in \mathbb{R}^n : x_n \geq \|(x_1, \ldots, x_{n-1})\|_2 \}$ be the Lorentz cone in $\mathbb{R}^n$. Consider the primal-dual pair

$$z^*_P = \sup \{ 0 : Ax = 1 - e_1, x \in \mathcal{C}_n \}, \quad z^*_D = \inf \{ y_2 + \cdots + y_{n-1} : A^\top y = e_n, 0, y \in \mathbb{R}^{n-1} \},$$

With respect to the primal formulation in $(1a)$ we have $K = \{ 0 \}$, $C = \mathcal{C}_n$, $b = 1 - e_1$ and $c = 0$. The primal constraints $x_j = 1, j = 2, \ldots, n-1$ imply that for $x \in \mathcal{C}_n$, we must have $x_n \geq \sqrt{x_1^2 + n - 2}$, but then the first primal constraint $x_1 + x_n = 0$ makes the problem infeasible. Hence, $z^*_P = -\infty$. The dual is obviously feasible with $y = 0$. In fact, the dual optimum is $z^*_D = 0$ because $A^\top y = (y_1, \ldots, y_{n-1}, 1)^\top$, and so every dual feasible solution has $y_1 \geq \|(y_1, \ldots, y_{n-1})\|_2$, which implies that $y_j = 0, j = 2, \ldots, n-1$.

The primal linear map and its adjoint from $(6)$ are $\mathcal{L}_p(\alpha, \alpha_0) = (A\alpha + \alpha_0(1 - e_1), -\alpha)$ and $\mathcal{L}_p^*(y, w) = (A^\top y - w, y_2 + \cdots + y_{n-1})$. Denote $S = \text{cl } \mathcal{L}_p^*(\{ 0 \}^* \times \mathcal{C}_n^*)$ and note that this is a closed convex cone. Because the Lorentz cone $\mathcal{C}_n$ is self-dual, we get that $S$ is equal to $\text{cl } \{ (A^\top y - w, y_2 + \cdots + y_{n-1}) : y \in \mathbb{R}^{n-1}, w \in \mathcal{C}_n \}$. Take any $\varepsilon > 0$. By the separation theorem, the point $(0, -\varepsilon)$, which is equal to $(c, z^*_D - \varepsilon)$, does not belong to $S$ if and only if there exists $(\alpha, \alpha_0) \in S^0$ such that $\langle 0, \alpha \rangle - \varepsilon \alpha_0 > 0$. Suppose there exists $(\alpha, \alpha_0) \in S^0$ with $\varepsilon \alpha_0 < 0$. **Lemma 2.4** tells us that $S^0 = \mathcal{L}_p^{-1}(\{ 0 \} \times -\mathcal{C}_n)$. Therefore, it must be that $\mathcal{L}_p(\alpha, \alpha_0) \in \{ 0 \} \times -\mathcal{C}_n$, which implies that $A\alpha + \alpha_0(1 - e_1) = 0$ and $\alpha \in \mathcal{C}_n$. The linear equation can be scaled by $-\alpha_0$ to get $A\alpha' = 1 - e_1$ for $\alpha' = -\alpha/\alpha_0$. Because $\alpha_0 < 0$ due to $\varepsilon > 0$ and $\varepsilon \alpha_0 < 0$, we have $\alpha' \in \mathcal{C}_n$ whenever $\alpha \in \mathcal{C}_n$. Hence, there must exist some $\alpha' \in \mathcal{C}_n$.
for which $A\alpha' = 1 - e_1$. However, this is exactly primal feasibility and we already argued that the primal is infeasible, therefore giving us a contradiction to the existence of $\alpha'$. Hence, $(0,-\varepsilon)$ cannot be separated from the closure of $\mathcal{L}_p^*(\mathbb{R}^{n-1} \times \mathcal{C}_n)$ for any $\varepsilon > 0$.

Now we show two sufficient conditions for strong duality that emerge as special cases of the closedness condition. Both of these conditions are known in literature through different proofs. The first such condition is the well-known constraint qualification related to strict feasibility of one of the problems, whereas the second condition is boundedness of the feasible region and is perhaps less well-known.

### 4.1 Slater CQ

A problem has the *generalized Slater’s constraint qualification* (Slater CQ) when it has strictly feasible solutions, where the sets of such solutions to the primal and dual problems are

$$
\text{strict } X = \text{strict } X(b) := \{x \in \text{ri } C : Ax \prec_K b\},
$$

$$
\text{strict } Y = \text{strict } Y(c) := \{y \in \text{ri } K^* : A^*y \succ_{C^*} c\}.
$$

Thus, the primal (resp. dual) has Slater CQ if and only if $\text{strict } X \neq \emptyset$ (resp. $\text{strict } Y \neq \emptyset$). From Observation 2.1 we know that a problem has Slater CQ if and only if the right-hand side of the conic inequality constraints belongs to the relative interior of the feasibility cone for that problem.

A related notion to strictly feasible points is that of points in the relative interior of $X$, so the question arises whether the two sets $\text{ri } X$ and $\text{strict } X$ are equal. We know from basic convex analysis that the relative interior of a set is a topological concept and is independent of algebraic representation of the set, and nonempty convex sets have nonempty relative interiors. However, strictly feasible solutions may exist for one algebraic formulation of the set but not for the other. For example, $X = \{x \in \mathbb{R}^n_+ : a^\top x \leq 1, -a^\top x \leq -1\}$, for some vector $a > 0$, has $\text{strict } X = \emptyset$, but writing the same set as $X = \{x \in \mathbb{R}^n_+ : a^\top x = 1\}$ gives $\text{strict } X = \{x > 0 : a^\top x = 1\}$. When strictly feasible solutions do exist, they indeed form the relative interior of the set.

**Observation 4.3.** For $X \neq \emptyset$, we have $\text{ri } X = \text{strict } X$ if and only if $\text{strict } X \neq \emptyset$.

The arguments for this are elementary and presented in Appendix A for sake of completeness and because we did not find an explicit reference in literature.

The significance of Slater CQ is that it is well-known to be a sufficient condition for achieving strong duality in convex optimization; there are many proofs of this in literature, for e.g. [Roc70, Theorem 28.2] and [Güll10, Theorem 11.15 and Remark 11.16]. For conic problems, there are different specialised proofs of strong duality under Slater CQ, such as [LSZ97, Theorem 7], [BN01, Theorem 1.7.1], [Rus06, Theorem 4.14], and [TW12, Corollary 4.8], and it can also be derived directly from the powerful Fenchel Duality Theorem in convex analysis [Roc70, Theorem 31.4]. We state the result explicitly since it will be used later in this paper.

**Theorem 4.4** (Slater CQ). If either $\text{strict } X \neq \emptyset$ or $\text{strict } Y \neq \emptyset$, then strong duality holds. Furthermore, the dual (resp. primal) has an optimal solution when the primal (resp. dual) has Slater CQ.

We now present another proof for sufficiency of Slater CQ for strong conic duality by showing that this CQ is a stronger condition than the closedness criteria stated in equations (7). The key ingredient of this argument is sufficient conditions for a linear image of a closed convex cone to be a closed set.
Lemma 4.5 (cf. [Pat07]). Given a linear map $\mathcal{L}$ and a non-polyhedral closed convex cone $\mathcal{C}$, the linear image $\mathcal{L}^*(\mathcal{C}^*)$ is a closed set if any of the following conditions hold:

1. $\text{Im} \mathcal{L} \cap \text{ri} \mathcal{C} \neq \emptyset$,
2. $\text{ker} \mathcal{L}^* \cap \text{ri} \mathcal{C}^* \neq \emptyset$,
3. $\text{Im} \mathcal{L} \cap \mathcal{C} \subseteq \text{lin} \mathcal{C}$,
4. $\text{ker} \mathcal{L}^* \cap \mathcal{C}^* \subseteq \mathcal{C}^\perp$.

Proposition 4.6. If a problem has Slater CQ then the corresponding closedness condition (either (7a) or (7b)) is satisfied and consequently, strong duality holds.

Proof. If closedness condition is satisfied (either for primal or dual), then Theorem 4.1 and Remark 1 tell us that strong duality holds. So we have to argue that if strict $X \neq \emptyset$ (primal Slater CQ) then $\mathcal{L}^*_p(\mathcal{K}^* \times \mathcal{C}^*)$ is a closed set. We will need the following result which says that the perspective image of the preimage of a cone under the linear map $\mathcal{L}_p$ is exactly the preimage of a cone under the affine map $G: x \mapsto (b - Ax, x)$.

Claim 1. Let $\mathcal{C}$ and $\mathcal{C}'$ be two nonempty convex cones, and $S_1 = \{ (\alpha, \alpha_0) \in \mathcal{L}^{-1}_p(\mathcal{C}' \times \mathcal{C}) : \alpha_0 > 0 \}$ and $S_2 = \mathcal{G}^{-1}(\mathcal{C}' \times \mathcal{C})$. Denote the perspective map by $\mathcal{P}: (\alpha, \alpha_0) \in \mathcal{E} \times \mathbb{R} \setminus \{0\} \mapsto \alpha/\alpha_0 \in \mathcal{E}$. Then, $-\mathcal{P}(S_1) \subseteq S_2$.

Proof of Claim. Take any $(\alpha, \alpha_0) \in S_1$. We have $-\alpha/\alpha_0 \in \mathcal{C}'$ due to $-\alpha \in \mathcal{C}$ and $\alpha_0 > 0$. Also, $\alpha_0(b - (\mathcal{A}(-\alpha/\alpha_0))) = \mathcal{A}x + \alpha_0b \in \mathcal{C}'$ and $\alpha_0 > 0$ implies that $b - (\mathcal{A}(-\alpha/\alpha_0)) \in \mathcal{C}'$. Hence, $-\mathcal{P}(S_1) \subseteq S_2$. The reverse inclusion $-\mathcal{P}(S_1) \supseteq S_2$ follows from $G(x) = \mathcal{L}_p(-x, 1)$.

Because $X = \mathcal{G}^{-1}_p(\mathcal{K} \times \mathcal{C})$ as per (2c), this claim implies that $X$ is the negative perspective image of the preimage of $\mathcal{K} \times \mathcal{C}$ under the map $\mathcal{L}_p$. Because strict $X = \mathcal{G}^{-1}(\text{ri} \mathcal{K} \times \text{ri} \mathcal{C})$, applying Claim 1 with $\mathcal{C}' = \text{ri} \mathcal{K}$ and $\mathcal{C} = \text{ri} \mathcal{C}$ gives us that strict $X \neq \emptyset$ implies $\mathcal{L}^{-1}_p(\text{ri} \mathcal{K} \times \text{ri} \mathcal{C}) \neq \emptyset$. Polarity and ri operators distribute over the Cartesian product, and so $(\mathcal{K} \times \mathcal{C})^* = \mathcal{K}^* \times \mathcal{C}^*$ and $\text{ri} (\mathcal{K} \times \mathcal{C}) = \text{ri} \mathcal{K} \times \text{ri} \mathcal{C}$. Using the first condition in Lemma 4.5 with $\mathcal{C} = \mathcal{K} \times \mathcal{C}$ implies that $\mathcal{L}^*_p(\mathcal{K}^* \times \mathcal{C}^*)$ is a closed set.

4.2 Existence of Slater CQ

We present some necessary and sufficient conditions for Slater CQ to hold. The definition of strict feasibility tells us that strictly feasible solutions exist only when $X \cap \text{ri} \mathcal{C} \neq \emptyset$ and so $X \subseteq \partial \mathcal{C}$ is necessary for their existence. But this is far from being a sufficient condition. The example $X = \{ x \in \mathbb{R}^+_n : a^T x \leq 1, -a^T x \leq -1 \}$ mentioned earlier satisfies $X \subseteq \bigcup_{j=1}^n \{ x : x_j = 0 \} = \partial \mathbb{R}^+_n$, but it does not have any strictly feasible solutions because the slacks of the two inequality constraints are negative of each other, implying that $X$ is contained in the subspace formed by the slack variables equal to zero. We show that the slacks of conic constraints forming a full-dimensional set is a sufficient condition for strict feasibility, and a necessary condition, under a technicality, is that $X$ have the same dimension as $\mathcal{C}$. Define the set of slack values for $X$ as

\[
\text{slack } X := b - \mathcal{A}(X) = \{ b - \mathcal{A} x : x \in X \}
\]

and let $\text{dim} \cdot$ denote the affine dimension of a set.

Proposition 4.7. We have the following.

\footnote{Since $\mathcal{C}^\perp = \mathcal{C}^\circ \cap \mathcal{C}^*$, a stronger version of the last condition would require that $\text{ker} \mathcal{L}^* \subseteq \mathcal{C}^\circ$.}
Lemma 8 to be the recession cone $\text{ri}(\text{slack} X) \subseteq \text{span} K$. Hence, $b - \mathcal{A}(\text{span} C) \subseteq \text{span} K$ and so $x \in \text{span} K$ for a nonempty convex set $K$. This implies that whenever the primal is feasible, it is also strictly feasible, and that the primal is feasible for all right-hand sides in the span of $K$.

1. strict $X \neq \emptyset$ if $\text{aff}(\text{slack} X) = \text{span} K$ and $\emptyset \neq X \subset \partial C$.

2. When $\mathcal{A}(\text{span} C) \subseteq \text{span} K$, strict $X \neq \emptyset$ only if $\dim X = \dim C$.

Proof. (1) For a nonempty convex set $X$, we have that $\text{slack} X$, which is the affine image $b - \mathcal{A}(X)$, is also a nonempty convex set, and therefore $\text{ri}(\text{slack} X) \neq \emptyset$. Because $\text{slack} X \subset K$, the assumption of equal affine hulls for $K$ and $\text{slack} X$ implies that $\text{ri}(\text{slack} X) \subseteq \text{ri} K$. The commutativity of $\text{ri}$ and affine images of convex sets gives us $\text{ri}(\text{slack} X) = b - \mathcal{A}(\text{ri} X)$. Hence, $b - \mathcal{A}(\text{ri} X) \subseteq \text{ri} K$. When $X \subset \partial C$, we have $\text{ri} X \cap \text{ri} C \neq \emptyset$ because otherwise $X \subset C = \text{ri} C \cup \partial C$, $\text{cl}(\text{ri} X) = X$, and $\partial C$ being a closed set implies the contradiction $X \subset \partial C$. Therefore, there exists some $x \in \text{ri} C$ with $b - \mathcal{A}x \in \text{ri} K$, and hence strict $X \neq \emptyset$.

(2) Clearly, $\dim X \leq \dim C$ because $X \subseteq C$. Suppose strict $X \neq \emptyset$. Take any $x \in \text{strict} X$ and $y \in \text{span} C$. We have $\mathcal{A}y \in \text{span} K$ due to the assumption that $\mathcal{A}(\text{span} C) \subseteq \text{span} K$. Because strict $X = \mathcal{A}^{-1}(b - \text{ri} K) \cap \text{ri} C$, there exists some $\varepsilon > 0$ such that $x + \varepsilon y \in \text{span} K$ and $b - \mathcal{A}x + \varepsilon Ay \in K$. The linearity of $\mathcal{A}$ makes $b - \mathcal{A}x + \varepsilon Ay \in K$ equivalent to $b - \mathcal{A}(x + \varepsilon y) \in K$, and so $x + \varepsilon y \in X$. Therefore, for any basis $\{y^1, \ldots, y^k\}$ of the subspace $\text{span} C$, the points $\{x, x + \varepsilon y^1, \ldots, x + \varepsilon y^k\}$ are affinely independent in $X$, implying that $\dim X \geq \dim C$, as desired.

This leads to a case where full-dimensionality of $X$ is necessary and sufficient for the existence of strictly feasible solutions.

Proposition 4.8. Suppose $C$ is full-dimensional and $\mathcal{A}$ is an injective map with $\text{Im} \mathcal{A} \subseteq \text{span} K$. Then strict $X \neq \emptyset$ if and only if $X \not\subset \partial C$ and $X$ is a full-dimensional set.

Proof. The only if direction is directly from the second claim in Proposition 4.7, whereas the if direction is from the first claim in the proposition. To see the if direction, suppose $X$ is a full-dimensional set. We have $\text{aff}(\text{slack} X) = \text{aff}(b - \mathcal{A}(X)) = b - \text{aff} \mathcal{A}(X) = b - \mathcal{A}(\text{aff} X)$, which implies $\dim(\text{slack} X) = \dim(\mathcal{A}(\text{aff} X))$. Recall the basic fact from linear algebra that taking a linear image of a set is a dimension-reducing operation in general, but the dimension is preserved when the linear map is injective. Therefore, when $\mathcal{A}$ is injective, we have $\dim(\mathcal{A}(\text{aff} X)) = \dim(\text{aff} X) = \dim X$. Now, $\dim(\text{slack} X) = \dim(\mathcal{A}(\text{aff} X))$ argued earlier and full-dimensionality of $X$ implies that $\text{slack} X$ is also full-dimensional. Since $\text{slack} X \subseteq K$, we have $\text{aff}(\text{slack} X) \subseteq \text{span} K$ and therefore, full-dimensionality of $\text{slack} X$ leads to $\text{aff}(\text{slack} X) = \text{span} K = \text{E}'$.

Our last sufficient condition for existence of Slater CQ is in terms of the homogenous set (cf. (2)) $X(0) := \{x \in C : Ax \preceq K 0\}$, which we know from Lemma 3.2 to be the recession cone whenever the right-hand side gives feasibility. We argue that when this set is strictly feasible, which by applying equation (8) with $b = 0$ means that the strictly feasible solutions in the homogenous set can be described as

$$\text{strict } X(0) = -\mathcal{A}^{-1}(\text{ri} K) \cap \text{ri} C = \{x \in \text{ri} C : Ax \preceq K 0\},$$

the convex cone $\mathcal{C}_p$ of all feasible right-hand sides for the primal is open and contains the span of $K$. This implies that whenever the primal is feasible, it is also strictly feasible, and that the primal is feasible for all right-hand sides in the span of $K$.

Proposition 4.9. Suppose that $\text{strict } X(0) \neq \emptyset$. Then $\mathcal{C}_p = \text{ri} \mathcal{C}_p \supseteq \text{span} K$; in particular, strict $X \neq \emptyset$ if and only if $X$ is feasible, and $X$ is strictly feasible for $b \in \text{span} K$. 

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\textbf{Proof.} For the equality we have to argue the \(\subseteq\)-inclusion. Take \(y \in \text{strict}\, \text{aff}\, X(0)\) and \(x \in X(b)\) for some \(b \in C_p\). We have \(x + y \in C + \text{ri}\, C\) and \(b - A(x + y) = b - Ax - Ay \in K + \text{ri}\, K\). This leads to \(x + y \in \text{strict}\, X(b)\) because of the fundamental fact about convex cones that their relative interiors are invariant to addition by the cone (i.e., a closed convex cone \(C\) obeys \(\text{ri}\, C + C = \text{ri}\, C\)).

For the inclusion of \(\text{span}\, K\), we use the following technical result which says that existence of preimage of relative interior of a cone under a linear map implies that the preimage exists for all affine maps that are a suitable shift of the linear map.

\textbf{Claim 2.} Let \(C\) be a closed convex cone. If a linear map \(L\) has \(L^{-1}(\text{ri}\, C) \neq \emptyset\), then \(L_v^{-1}(\text{ri}\, C) \neq \emptyset\) for the affine map \(G_v(\cdot) = v + L(\cdot)\) with \(v \in \text{aff}\, C\). On the contrary, if an affine map \(G\) has \(G^{-1}(\text{ri}\, C) \neq \emptyset\), then \(G_v^{-1}(\text{ri}\, C) \neq \emptyset\) for all affine maps \(G_v(\cdot) = (1 - \lambda)G(0) + \lambda G(\cdot)\) where \(\lambda \neq 0\).

\textbf{Proof of Claim.} We consider two cases for the first claim. First, suppose that \(L^{-1}(\text{ri}\, C) \subseteq \ker\, L\). Since \(\ker\, L = L^{-1}(\{0\})\), we have \(0 \in \text{ri}\, C\) in this case. It follows from Lemma 2.5 that \(C\) must be a linear subspace. This implies \(\text{aff}\, C = \text{ri}\, C\), and so for every \(x \in L^{-1}(\text{ri}\, C)\) and \(v \in \text{aff}\, C\) we have \(G_v x = v + L x = v + 0 = v \in \text{aff}\, C = \text{ri}\, C\), implying that \(L_v^{-1}(\text{ri}\, C) \supseteq L^{-1}(\text{ri}\, C) \neq \emptyset\). Now suppose that \(L^{-1}(\text{ri}\, C) \notin \ker\, L\). Therefore, there exists \(x \in L^{-1}(\text{ri}\, C)\) such that \(L x \neq 0\). Take any \(v \in \text{aff}\, C\). Then, \(L x \in \text{ri}\, C\) implies that there exists some \(\varepsilon > 0\) for which \(L x + \varepsilon v \in \text{ri}\, C\). Scaling by \(\varepsilon\) and using the fact that \(\text{ri}\, C\) is an open convex cone, gives us \(L(x/\varepsilon) + v \in \text{ri}\, C\). Hence, \(G_v(x/\varepsilon) \in \text{ri}\, C\), which implies that \(x/\varepsilon \in G_v^{-1}(\text{ri}\, C)\).

For the second claim, take any \(x \in G^{-1}(\text{ri}\, C)\). Denote \(G_v(\cdot) = G_0 - \lambda G(\cdot)\), which is a linear map. Then, \(G_v(\cdot) = G_0 - \lambda G(\cdot)\) is a linear map. Then, \(G_v(\cdot) = G_0 - \lambda G(\cdot)\). Because \(G x = G_0 - L G x \in \text{ri}\, C\), linearity of \(G_v\) implies that for any \(\lambda \neq 0\), we have \(G x = G_0 - \lambda G(x/\lambda) = G_0(x/\lambda)\), and so \(G_v^{-1}(\text{ri}\, C) \neq \emptyset\). ☐

\(X(0)\) is the preimage of the cone \(K \times C\) under the linear map \(L : x \mapsto (-Ax, x)\), and \(X(b)\) is the preimage of \(K \times C\) under the affine map \((b, 0) + L(x)\). The definition of strict feasibility is that \(\text{strict}\, X(0) = L^{-1}(\text{ri}\, K \times \text{ri}\, C)\). The claim \(\text{strict}\, X(b) \neq \emptyset\) follows from Claim 2 due to \((b, 0) \in \text{span}\, K \times \text{span}\, C\). ☐

Strictly feasible solutions may not always exist for the cone \(X(0)\); we can apply the conditions presented earlier in this section with \(b = 0\) to certify when \(\text{strict}\, X(0) \neq \emptyset\). Note that if \(\text{strict}\, X(0) = \{0\}\), then it means that for each of \(K\) and \(C\), either the cone is equal to \(\{0\}\) or is a subspace, but even in this case the above proof goes through because we have the cones equal to their relative interiors.

\subsection{4.3 Boundedness CQ}

For a general convex program

\[
\inf \left\{ f_0(x) : f_i(x) \leq 0 \text{ for } i = 1, \ldots, m, \ x \in S \right\},
\]

Duffin [Duf78, Theorems 1 and 2] established that if the primal has a bounded feasible set then the Lagrangian dual of (10) has an unbounded feasible set and strong duality holds. We term this Clark-Duffin theorem as the ‘Boundedness CQ’ for strong duality; see Ernst and Volle [EV13] for generalisations of it where CQs are given in terms of recession directions satisfying certain properties. Since the conic program (1a) can be reformulated in the form (10) as

\[
\inf_{x,s} \{ c^T x : Ax + s = b, \ x \in C, s \in K \},
\]

and it is easy to see that the set \(X := \{(x, s) \in C \times K : Ax + s = b\}\) is bounded if and only if its projection \(X\) is bounded, we have the Boundedness CQ for conic programs. Nevertheless, this CQ can also be argued independently
Lemma 4.11. Suppose X is feasible and \( A \) is injective. Then, X is bounded if and only if \( A(C) \cap -K = \{0\} \). In particular, X is bounded if \( A(C) \subseteq K \) and K is pointed, or \( A(C) \subseteq K^* \).

Proof. The only if part is from Lemma 2.6. For the if part, Lemma 3.2 tells us that it suffices to show \( A(C \setminus \{0\}) \cap -K = \emptyset \). Suppose there exists a nonzero \( x \in C \) for which \( Ax \in -K \). Set \( y = Ax \). Then \( y \in A(C) \cap -K \). Because \( A(C) \cap -K = \{0\} \), it must be that \( y = 0 \) and therefore \( x \in \ker A \). The assumption that \( A \) is injective is equivalent to \( \ker A = \{0\} \), and so \( x = 0 \), which is a contradiction. Therefore, \( X' \) must be bounded. The particular conditions each imply \( A(C) \cap -K = \{0\} \) because \( K \cap -K = \{0\} \) for a pointed \( K \) and any closed convex cone \( \mathcal{C} \) satisfies \( \mathcal{C}^* \cap -\mathcal{C} = \{0\} \).

An unbounded set may or may not have strictly feasible solutions to its recession cone. A pointed set is unbounded if and only if the recession cone of the other problem does not have any strictly feasible solutions. We also present a sufficient condition for boundedness in terms of a basis for the span of \( C \).

Proposition 4.12. Let \( X \neq \emptyset \) be pointed. It is bounded if and only if \( \text{strict } Y(0) \neq \emptyset \).

Furthermore, it is bounded if there exists an orthonormal basis \( B = \{v^1, \ldots, v^m\} \) of \( \text{span } C \) such that \( B \subseteq C^* \) and \( A(B) \subseteq K \) with \( v^j \in \text{ri } C^* \) and \( Av^j \in K \setminus \text{lin } K \) for some \( j \).
Lemma 4 tells us that we can replace the adjoint map in the dual with the linear map $LP_{17}$. Theorem 5.1: $\{y \in K^*: BY \supseteq C^* \Rightarrow 0\}$. Lemma 2.6 applied to this recession cone tells us that to show that $X$ is bounded, it suffices to show that there exists $y \in ri K^*$ such that $BY \in ri C^*$. The assumption $A\nu_j \in K \setminus C \neq \emptyset$ implies that $K \setminus C \neq \emptyset$, which is equivalent to $K$ not being a linear subspace. Then Lemma 2.5 implies that there exists a nonzero $y \in ri K^*$. The second claim in this lemma combined with the assumption $A\nu_j \in K \setminus C \neq \emptyset$ gives us $\langle A\nu_j, y \rangle > 0$. This leads to $\langle A\nu, y \rangle \nu \in ri C^*$ due to $\nu \in ri C^*$ and the fact that $ri C^*$ is a cone. The assumptions $\nu \in C^*$ and $A\nu_i \in K$ for $i \neq j$ give us $\langle A\nu_i, y \rangle \geq 0$ and $\langle A\nu_i, y \rangle \nu \in C^*$, thereby leading to $\sum_{i \neq j} \langle A\nu_i, y \rangle \nu \in C^*$. Since $BY = \langle A\nu, y \rangle \nu + \sum_{i \neq j} \langle A\nu_i, y \rangle \nu$, and the relative interior of a cone is invariant to addition with the cone, we obtain the desired claim $BY \in ri C^*$.

5 A Theorem of the Alternative on Almost Feasibility

Besides strict feasibility, which guarantees existence of Slater CQ, another notion on feasibility is that of almost feasibility which is defined for $X(b)$ with respect to some chosen norm $\| \cdot \|$ in $E'$ as follows:

\[
\text{(Almost feasibility of } X \text{): } \forall \varepsilon > 0, \exists b' \in E' \text{ s.t. } \|b'\| \leq \varepsilon \text{ and } X(b + b') \neq 0. \tag{11}
\]

The definition for $Y(c)$ is analogous. Our definitions are inspired by Ben-Tal and Nemirovski [BN01, §1.7.2] who call it almost solvability. A set that is infeasible but almost feasible is called weakly infeasible by Luo et al. [LSZ97]. Algorithmic aspects and other issues surrounding almost feasibility are discussed by Pólik and Terlaky [PT09] and Liu and Pataki [LP17]. Clearly, Feasibility $\implies$ Almost feasibility.

Our main result here is to show that under some conditions, almost feasibility of the dual is equivalent to membership in the polar of the recession cone of the primal.

Theorem 5.1. $\langle X(0) \rangle^* = \{c \in E: Y(c) \text{ is almost feasible}\}$ when either $K$ is a subspace or $A^*(ri K^*) \cap (\text{span } C)^\perp \neq \emptyset$.

This result can be interpreted as a theorem of the alternative because the assertion tells us that for every $c \in E$, we have exactly one of the following two statements being true:

1. either $Y(c)$ is almost feasible, or
2. $\langle c, x \rangle > 0$ for some $x \in C'$ with $Ax \preceq_K 0$.

Remark 4. The analogous version of Theorem 5.1 for the dual problem is that

\[
\langle Y(0) \rangle^* = \{b \in E': X(b) \text{ is almost feasible}\} \tag{12}
\]

when either $C$ is a subspace or $A(ri C) \cap \text{span } K \neq \emptyset$. This is a generalization of Ben-Tal and Nemirovski [BN01, Proposition 1.7.1] which deals with $C = E$ and a full-dimensional $K$ and is therefore a special case of our result because it satisfies both of our conditions\(^4\).

\(^4\)Since [BN01] deal with the primal as a minimisation problem, the obvious modifications have to be made when comparing their results to ours.
Before proving our theorem, we derive a consequence of it which is the equivalence of feasibility and almost feasibility for bounded sets, and infeasibility of one problem implying an unbounded optimum for the other problem.

**Corollary 5.2.** Assume $X(0) = \{0\}$ and either $C$ is a subspace or $\mathcal{A}(\text{ri} C) \cap \text{span} \, K \neq \emptyset$. The following are equivalent for any $b \in E'$,

1. $b \in (Y(0))^*$,
2. $X(b)$ is feasible,
3. $X(b)$ is almost feasible.

Furthermore, if we also have that the primal is infeasible, then $z_{\text{p}}^* = -\infty$ when $c \in (\text{lin} \, C)^\perp$, otherwise the dual is either infeasible or has unbounded optimum when $c \notin (\text{lin} \, C)^\perp$.

**Proof.** The first and third are equivalent by (12), and second implying the third is trivially true without any assumptions. Take any $b \in E'$ is such that $X(b)$ is almost feasible. Lemma 2.6 tells us that $X(0) = \{0\}$ implies strict $Y(0) \neq \emptyset$. The dual analogue of the last part of Lemma 3.2 gives us $(Y(0))^* = C_p$, and then from (12) it follows that $b \in C_p$ and so $X(b)$ is feasible.

Now suppose that the primal is infeasible. The first part of this proof tells us that $b \notin (Y(0))^*$, which means that there exists some $r \in Y(0)$ such that $\langle b, r \rangle < 0$. Since the dual is a minimisation, when it is feasible, the ray $r$ leads to an optimum of $-\infty$. By Lemma 2.6, strict $Y(0) \neq \emptyset$ and then the dual analogue of Proposition 4.9 asserts dual feasibility for $c \in \text{span} \, C^* = (\text{lin} \, C)^\perp$. □

Without the above conditions it is possible to have infeasibility for one and finite optimum for the other as seen in Example 1 which has $X(0) \neq \{0\}$.

Now we proceed to prove Theorem 5.1 by first establishing some technical results.

### 5.1 Lemmata

The set of all right-hand sides for which the dual problem is almost feasible is sandwiched between the polar cone of $X(0)$ and the relative interior of this polar, and therefore one direction of inclusion in Theorem 5.1 is always true without requiring any assumptions.

**Lemma 5.3.** $(X(0))^\circ \supseteq \{c \in E : Y(c) \text{ is almost feasible}\} \supseteq \text{ri} \, (X(0))^\circ$.

**Proof.** The second inclusion is due to $\text{ri} \, (X(0))^\circ = \text{ri} \, C_d$ from Lemma 3.2 and the fact that strict feasibility implies feasibility which implies almost feasibility. Let us prove the first inclusion by contraposition. Let $c \notin (X(0))^\circ$. Then there exists some nonzero $r \in X(0)$ for which $\langle c, r \rangle > 0$. Pick any norm $\| \cdot \|$ in $E'$ and set $\delta = \langle c, r \rangle$ and $\varepsilon = \frac{\delta}{2\|r\|}$. For any $c^\varepsilon \in E'$ such that $\|c^\varepsilon\| \leq \varepsilon$, we have

$$\langle c + c^\varepsilon, r \rangle \geq \langle c, r \rangle - \|c^\varepsilon\| \|r\| \geq \langle c, r \rangle - \varepsilon \|r\| = \delta - \frac{\delta}{2} = \frac{\delta}{2} > 0,$$

where the second inequality is by the Cauchy-Schwarz inequality. Therefore, $c + c^\varepsilon \notin (X(0))^\circ$, and then $(X(0))^\circ \supseteq C_d$ from Lemma 3.2 implies $c + c^\varepsilon \notin C_d$, which means that $Y(c + c^\varepsilon) = \emptyset$. Since $c^\varepsilon$ was arbitrary up to $\|c^\varepsilon\| \leq \varepsilon$, it follows from the definition in (11) that $Y(c)$ is not almost feasible. □

Define

$$\Omega := \{c \in E : \exists y \in K^*, w \in K^* \setminus \text{lin} \, K^* \text{ s.t. } A^*(y - w) \succcurlyeq c, c\}.$$
We will need a technical lemma on $\Omega$ for proving Theorem 5.1. Let us recall the fact that relative interior of a non-subspace convex cone is exactly the points in its span that have a positive product with every point in the dual cone that is not in the orthogonal complement.

**Lemma 5.4** (cf. [LSZ97, Theorem 2]). Suppose $\mathcal{C}$ is a closed convex cone that is not a linear subspace.\(^5\) We have $\text{ri} \mathcal{C} = \{ x \in \text{aff} \mathcal{C} : \langle x, y \rangle > 0, \forall y \in \mathcal{C}^* \setminus \mathcal{C}^\perp \}$, and so $\text{ri} \mathcal{C} \not\subset \mathcal{C}^* \setminus \mathcal{C}^\perp$.

Similarly, $\text{ri} \mathcal{C}^* = \{ y \in (\text{lin} \mathcal{C})^\perp : \langle y, x \rangle > 0, \forall x \in \mathcal{C} \setminus \text{lin} \mathcal{C} \}$ and $\text{ri} \mathcal{C}^* \not\subset \mathcal{C} \setminus \text{lin} \mathcal{C}$.

**Lemma 5.5.** $(0^+ X)^\circ \cap \Omega = \{ c \in \Omega : Y(c) \text{ is almost feasible} \}$.

**Proof.** The $\supseteq$ inclusion is from **Lemma 5.3**. Take $c \in \Omega$ and suppose $Y(c)$ is not almost feasible. We have to prove that $c \notin (0^+ X)^\circ$. Choose any $\xi \in \text{ri} C^*$ and $\gamma \in \text{ri} K$ and consider the conic problem

$$z^* = \inf \{ t_1 + t_2 + \langle \gamma, w \rangle : \mathcal{A}^* y - \mathcal{A}^* w + t_1 c + t_2 \xi \succ_c, y \in K^*, w \in K^*, t_1, t_2 \geq 0 \}.$$ 

Denote the feasible set by $Y_\xi$. Observe that $(\bar{y}, \bar{y}, 1, 1) \in \text{strict} Y_\xi$ for any $\bar{y} \in \text{ri} K^*$. The dual problem to $z^*$ is

$$\sup \{ \langle c, x \rangle : \mathcal{A} x \preceq_K 0, -Ax \preceq_K \gamma, \langle c, x \rangle \leq 1, \langle \xi, x \rangle \leq 1, x \in C \}.$$ 

This has a feasible solution $x = 0$ due to $\gamma \in \text{ri} K$. Then strong duality from Theorem 4.4 implies that there exists a feasible $x^*$ to the dual problem with $\langle c, x^* \rangle = z^*$. It is clear that $z^* \geq 0$. We claim that $z^* > 0$. Because the feasible set of the dual problem is a subset of $\{ x \in C : \mathcal{A} x \preceq_K 0 \} = 0^+ X$, we get $x^* \in 0^+ X$, and so $z^* > 0$ implies $c \notin (0^+ X)^\circ$, thereby finishing our proof.

Suppose for the sake of contradiction that $z^* = 0$. By **Lemma 5.4** and $\gamma \in \text{ri} K$, the term $\langle \gamma, w \rangle$ is equal to zero for any $w \in K^*$ if and only if $w \in K^\perp$. **Lemma 2.3** with $\mathcal{C} = K^*$ and the Bipolar Theorem gives us $K^\perp = \text{lin} K^*$. Because $(\bar{y}, \bar{y}, 1, 1) \in \text{strict} Y_\xi$ for any $\bar{y} \in \text{ri} K^*$, we know that the objective function is not identically equal to zero over $Y_\xi$. Because the infimum is equal to zero, for arbitrarily small $\varepsilon > 0$ there exist feasible solutions $(y, w, t) \in Y_\xi$ with $t_1 + t_2 + \langle \gamma, w \rangle = \varepsilon$. Then one of the following two things must happen: (i) at least one of $t_1$ or $t_2$ is strictly positive and $w \in \text{lin} K^*$, or (ii) $t_1 = t_2 = 0$ and $w \in K^* \setminus \text{lin} K^*$. However, $c \in \Omega$ means that the second case is not possible. In the first case, set $\delta = \varepsilon(\|c\| + \|\xi\|)$ and $c^\delta = -t_1 c - t_2 \xi$. Because $t_1, t_2 \leq \varepsilon$, we have $\|c^\delta\| \leq \delta$. Because $\varepsilon$ is arbitrary close to zero, $\delta$ is also arbitrary close to zero. Then $\mathcal{A}^*(y - w) \succ_K c + c^\delta$ and $y - w \in K^*$ due to $y \in K^*$, $w \in \text{lin} K^*$ implies that for every $\delta > 0$, we have $y - w \in Y(c + c^\delta)$, thereby making $Y(c)$ almost feasible, which is a contradiction.

We will need the following geometric property of cones that one can enter (resp. exit) a cone in some suitable directions starting from points outside (resp. inside) the cone; this is proved in Appendix A for completeness.

**Lemma 5.6.** Let $\mathcal{C}$ be a nonempty closed convex cone and $x \in \mathcal{C}$ and $y \in \text{span} \mathcal{C}$.

1. If $y \notin \mathcal{C}$, there exists a finite $t' \geq 0$ such that $x + t y \in \mathcal{C}$ for all $0 \leq t \leq t'$ and $x + t y \notin \mathcal{C}$ for all $t > t'$. Also, $t' = 0$ only if $x \in \partial \mathcal{C}$.

Now suppose $x \in \text{ri} \mathcal{C}$.

2. There exists a finite $\delta^* > 0$ such that $y + \delta x \in \text{ri} \mathcal{C}$ for all $\delta > \delta^*$.

3. If $y \notin \mathcal{C}$, then $y + \delta^* x \in \partial \mathcal{C}$ and $y + \delta x \notin \mathcal{C}$ for all $0 \leq \delta < \delta^*$.

\(^5\)The assertion here also works when $\mathcal{C}$ is a linear subspace, if we interpret the product being positive to be a vacuously true statement and drop it so that $\text{ri} \mathcal{C}$ becomes equal to $\text{aff} \mathcal{C}$, which is obvious for a subspace.
5.2 Proof of Theorem 5.1

The dual cone of a subspace is also a subspace and so \( K \) being a subspace implies \( K^* = \text{lin} \cdot K^* \), which makes \( \Omega = E \) in Lemma 5.5. The equality for the polar cone follows from the lemma.

Now let \( A^*(\text{ri} \cdot K^*) \cap \text{aff} \cdot C^* \neq \emptyset \). The \( \supseteq \) inclusion is from Lemma 3.2. We argue the \( \subseteq \) inclusion by contraposition. Suppose \( Y(c) \) is not almost feasible. Choose any \( \xi \in \text{ri} \cdot C^* \) and consider the conic problem

\[
z^* = \inf \left\{ t_1 + t_2 : A^*y + t_1c + t_2\xi \succcurlyeq_{C^*} c, \ y \in K^*, t \geq 0 \right\}.
\]

Denote the feasible set \( Y(\xi) \). The dual problem to \( z^* \) is

\[
\sup \left\{ \langle c, x \rangle : Ax \preccurlyeq_K 0, \langle c, x \rangle \leq 1, \langle \xi, x \rangle \leq 1, x \in C \right\}.
\]

This has a feasible solution \( x = 0 \). By assumption, there exists some \( \bar{y} \in \text{ri} \cdot K^* \) with \( A^*\bar{y} \in \text{aff} \cdot C^* \). Lemma 5.6 with \( \mathcal{C} = C^* \) and \( \xi \in \text{ri} \cdot C^* \) imply that \( (\bar{y}, 1, t_2) \in \text{strict} \cdot Y(\xi) \) for some \( t_2 > 0 \). Then strong duality under Slater CQ from Theorem 4.4 implies that there exists a feasible \( x^* \) to the dual problem with \( \langle c, x^* \rangle = z^* \). It is clear that \( z^* \geq 0 \). We claim that \( z^* > 0 \). Because the feasible set of the dual problem is a subset of \( \{ x \in C : Ax \preccurlyeq_K 0 \} = 0^+X \), we get \( x^* \in 0^+X \), and so \( z^* > 0 \) implies that \( c \notin (0^+X)^0 \), which finishes our proof by contraposition for the \( \subseteq \) inclusion. To argue the claim \( z^* > 0 \), suppose that \( z^* = 0 \). This means that there exist feasible solutions to \( Y(\xi) \) with both \( t_1 \) and \( t_2 \) arbitrarily close to zero. Then for any \( \varepsilon > 0 \), we can choose \( c^\varepsilon = -t_1c - t_2\xi \) for small values of \( t_1 \) and \( t_2 \) to get \( ||c^\varepsilon|| \leq \varepsilon \), and we would have some \( y \in K^* \) such that \( A^*y \succcurlyeq_{C^*} c + c^\varepsilon \). But this leads to the contradiction that \( Y(c) \) is almost feasible. \( \square \)

6 Projecting a Conic Set onto a Subspace

A well-known result for linear programming is that the orthogonal projection of a polyhedron onto a subspace of variables can be obtained by multiplying the algebraic linear description of the polyhedron by every extreme ray of a specific cone called the projection cone (cf. [Bal05, Theorem 1.1]). This can be proven using LP strong duality and it is very useful for deriving valid inequalities to mixed-integer programs. It is natural to expect that a similar result holds for sets described using conic inequalities because conic programs also exhibit strong duality, albeit under stronger assumptions than those required for linear programming. We describe the projection of conic sets onto arbitrary linear subspaces by giving a straightforward proof that is motivated by the polyhedral case. Note that our question is different than that of projecting a point onto a conic set, for which algorithms are given by Henrion and Malick [HM12].

Consider the conic set \( X = \{ x \in C : Ax \preccurlyeq_K b \} \) from (1a) and let \( L \subset E \) be a linear subspace. The projection of \( X \) onto \( L \) is defined as

\[
\text{proj}_L X := \left\{ x \in L : \exists u \in L^\perp \text{ s.t. } x + u \in X \right\}.
\]

Denote

\[
\mathcal{C} := \left\{ (y, w) \in K^* \times C^* : A^*y - w \in L \right\}.
\]

Clearly, this is a convex cone. It is also a closed set because it is equal to the intersection of two closed sets — (i) \( K^* \times C^* \), and (ii) the preimage of the subspace \( L \) under the linear map \( (y, w) \mapsto A^*y - w \). The latter is a closed set because the preimage of any closed set under a continuous map is a closed set. Thus, \( \mathcal{C} \) is a closed convex cone and so it is generated by its extreme rays.
Theorem 6.1. Suppose there exists \( \hat{y} \in \text{ri } K^* \) and \( \hat{w} \in \text{ri } C^* \) such that \( A^* \hat{y} - \hat{w} \in L \). Then,

\[
\text{proj}_L X = \left\{ x \in L : \langle A^* y - w, x \rangle \leq \langle b, y \rangle, \forall \text{ extreme rays } (y, w) \in \mathcal{C} \right\}.
\]

Proof. The definition of projection tells us that

\[
\text{proj}_L X = \left\{ x \in L : \exists u \in L^\perp \text{ s.t. } Au \preceq_K b - Ax, u \in C - x \right\}.
\]  

(13)

Consider the conic program

\[
\sup_u \left\{ \langle 0, u \rangle : Au \preceq_K b - Ax, u \in C - x, u \in L^\perp \right\}.
\]

Using the fact that \((L^\perp)^\perp = L\), the dual of the above primal can be easily derived to be

\[
\inf_{y,w} \left\{ \langle b - Ax, y \rangle + \langle x, w \rangle : (y, w) \in \mathcal{C} \right\}.
\]

The existence of \((\hat{y}, \hat{w})\) guarantees strict feasibility of \(\mathcal{C}\) because \(\text{ri } L = L\) for a subspace \(L\), and so the dual problem has Slater CQ, and then Theorem 4.4 implies that we have strong duality. Because the dual optimizes over a cone, its value is either 0 or \(-\infty\), and so strong duality implies that the dual value is 0 if and only if the primal is feasible. Hence, the existence of a \(u\) in equation (13) can be equivalently replaced with the infimum of the dual being zero, which is equivalent to \(\langle b - Ax, y \rangle + \langle x, w \rangle \geq 0\) for all \((y, w) \in \mathcal{C}\). Because \(\mathcal{C}\) is a cone, it suffices to enforce this nonnegativity for only its extreme rays. Substituting this in equation (13) yields

\[
\text{proj}_L X = \left\{ x \in L : \langle b - Ax, y \rangle + \langle x, w \rangle \geq 0, \forall \text{ extreme rays } (y, w) \in \mathcal{C} \right\}.
\]

The claimed expression follows after rearranging terms and using the adjoint. \(\square\)

The cone \(\mathcal{C}\) could have uncountably many extreme rays, in which case the projection is defined by an infinite system of linear inequalities.

Applying Theorem 6.1 to a high-dimensional set gives us its orthogonal projection.

Corollary 6.2. The orthogonal projection of \(S = \{(x, x') \in C \times C': Ax + Bx' \preceq_K b\}\) onto the \(x\)-space is equal to

\[
\left\{ x \in C : \langle A^* y, x \rangle \leq \langle b, y \rangle, \forall \text{ extreme rays } y \in \mathcal{C} \right\},
\]

for the cone \(\mathcal{C} = \{y \in K^*: B^* y \succeq_{C^*} 0\}\), when there exists a \(y \in \text{ri } K^*\) with \(B^* y \in \text{ri } C^*\).

7 Conclusions

Of the three CQs addressed in this paper, we showed that a Closedness CQ is the weakest condition since it implies the Slater CQ which in-turn implies the Boundedness CQ. We also gave a generalised version of the last CQ using a conic theorem of the alternative on strict feasibility. Another theorem of the alternative on almost feasibility was presented in its most general form. Subspace projections of conic sets were characterised using extreme rays of a projection cone, thereby extending the well-known idea for polyhedral sets.

An interesting direction of future research would be to establish more connections between the different sufficient conditions for strong duality in literature, particularly for abstract convex programs. In particular, it would be good to know whether the Closedness CQ used in this paper is also necessary for strong duality, and how it relates to another known CQ (cf. Remark 2).
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Appendix A  Missing Proofs

Proof of Observation 2.1. The equalities for \( C_p \) and \( C_d \) are straightforward from their definitions and equations (2). The relative interiors follow from basic properties of \( \text{ri} \) given in Lemma 2.2. Distributivity over sum gives us \( \text{ri} C_p = \text{ri} A(C) + \text{ri} K \), and commutativity with a linear map leads to \( \text{ri} C_p = A(\text{ri} C) + \text{ri} K \). The definition of strict feasibility makes it easy to see that strict \( X(b) \neq \emptyset \) if and only if \( b \in A(\text{ri} C) + \text{ri} K \).

Proof of Lemma 2.5. Lemma 5.4 implies that \( \text{ri} C^* \cap C^\perp = \emptyset \). Because \( C^\perp \subseteq C^* \) and \( C^* \) is the disjoint union \( \text{ri} C^* \cup \partial C^* \), it follows that \( C^\perp \subseteq \partial C^* \). Substituting \( C \) with \( C^* \) and using the Bipolar Theorem and Lemma 2.3 transforms \( C^\perp \subseteq \partial C^* \) to \( \text{lin} C \subseteq \partial C \). The second claim follows from the first claim due to \( 0 \in \text{lin} C \cap C^\perp \), \( \text{ri} C \cap C^\perp = \emptyset \), and \( \text{ri} C^* \cap \partial C^* = \emptyset \).

Proof of Lemma 2.6. The first statement being false means that \( X(0) = \{0\} \). Then, \( (X(0))^\circ = E \), and so Lemma 3.2 gives us \( \text{cl} C_d = E \), which implies that \( 0 \in \text{ri} (\text{cl} C_d) = \text{ri} C_d \). In fact, it is easy to argue that the second statement is equivalent to \( 0 \in \text{ri} C_d \). Hence, the second statement holds because of the description of \( \text{ri} C_d \) from Observation 2.1. If \( X(0) \) is not a subspace and second statement is true, we have \( 0 \in \text{ri} C_d = \text{ri} X(0)^\circ \). Lemma 2.5 and the non-subspace assumption implies that \( X(0) \) must be equal to \( \{0\} \).

Proof of Observation 4.3. The only if direction is due to the feasibility of a convex set being equivalent to the feasibility of its relative interior. Now suppose strict \( X \neq \emptyset \). The strict conic inequality \( Ax <_K b \) is defined as \( b - Ax \in \text{ri} K \), which is equivalent to \( x \in A^{-1}(b - \text{ri} K) \). We will use the basic properties from Lemma 2.2. The map \( K \mapsto b - K \) is affine and then commutativity of an affine map with \( \text{ri} \) implies that \( b - \text{ri} K = \text{ri}(b - K) \), leading to \( Ax <_K b \) if and only if \( x \in A^{-1}(b - K) \). Thus, strict \( X = A^{-1}(b - K) \cap \text{ri} C \), and so \( A^{-1}(b - K) \) is nonempty. Using the third claim from the lemma gives us strict \( X = A^{-1}(b - K) \cap \text{ri} C \). Because \( X = A^{-1}(b - K) \cap \text{ri} C \), the distributivity of \( \text{ri} \) over nonempty intersection implies that \( \text{ri} X = \text{ri} A^{-1}(b - K) \cap \text{ri} C \), giving us the desired equality \( \text{ri} X = \text{strict} X \).

Proof of Lemma 5.6. Suppose \( y \notin C \) and \( x + ty \in C \) for all \( t \geq 0 \). Because \( C \) is a cone, for \( t > 0 \), we have \( \frac{1}{t}(x + ty) = \frac{x}{t} + y \in C \). Because \( C \) is closed, \( \lim_{t \to \infty} \frac{x}{t} + y \in C \). But this limit is equal to \( y \), giving a contradiction to \( y \notin C \). Hence, there exists some \( \tilde{t} \geq 0 \) for which \( x + ty \notin C \). Convexity of \( C \) and \( x \in C \) implies that \( x + ty \in C \) is not possible for any \( t > \tilde{t} \). Taking \( t' := \inf \{ t : x + ty \notin C \} \), which is equal to \( \sup \{ t : x + ty \in C \} \), finishes our first claim on existence of a \( t' \). The second claim on \( t' = 0 \) is obvious from \( \text{span} C \) being a subspace and \( x + \varepsilon y \in C \) for some small \( \varepsilon > 0 \) when \( x \in \text{ri} C \) and \( y \in \text{span} C \).

We consider two cases: \( y \in C \) and \( y \in \text{span} C \setminus C \). In the first case, \( \delta^* \) can be any positive scalar because \( \text{ri} C + C = \text{ri} C \). Now consider the second case \( y \in \text{span} C \setminus C \). First we argue the existence of \( \delta^* > 0 \) such that \( y + \delta x \in C \) for all \( \delta \geq \delta^* \). Consider \( t' \) from the first claim in this proof. The value of \( t' \) is equal to 0 if and only if \( x + ty \notin C \) for all \( t > 0 \). When \( x \in \text{ri} C \), there exists some small enough \( \varepsilon > 0 \) such that \( x + \varepsilon d \in C \) for all \( d \in \text{aff} C \). Taking \( d = y \) implies \( x + \varepsilon y \in C \), and therefore \( t' > 0 \). Set \( \delta^* := 1/t' \); we have \( \delta^* \in (0, \infty) \) due to \( t' \in (0, \infty) \). Take any \( \delta \geq \delta^* \). Because \( 1/\delta \in (0, t') \), we have \( x + \frac{1}{\delta} y \in C \). This leads to \( \delta(x + \frac{1}{\delta} y) = y + \delta x \in C \) because \( C \) is a cone. If \( y + \delta x \in C \) for some \( 0 \leq \delta < \delta^* \), then \( \frac{1}{\delta} y + x \in C \), a contradiction to the first claim for \( t = 1/\delta \) which is larger than \( t' := 1/\delta^* \). Hence, we have \( y + \delta x \notin C \) for all \( 0 \leq \delta < \delta^* \). It follows then that \( y + \delta^* x \in \partial C \) because \( y \notin C \). Finally, consider \( y + \delta x \) for \( \delta > \delta^* \). Because \( y + \delta x = y + \delta^* x + (\delta - \delta^*) x \), and \( y + \delta^* x \in \partial C \) and \( (\delta - \delta^*) x \in \text{ri} C \) due to \( \delta > \delta^* \), \( \text{ri} C + C = \text{ri} C \) gives us \( y + \delta x \in \text{ri} C \).