

Necessary and Sufficient Conditions for Farkas' Lemma for Cone Systems and Second-Order Cone Programming Duality*

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We present conditions, which completely characterize Farkas' lemma for cone-convex systems, and obtain strong duality characterizations for convex optimization problems. In particular, we establish a necessary and sufficient closed cone condition for the Farkas lemma. As an application, we obtain necessary and sufficient conditions for the strong duality of convex second-order cone programming problems.

Keywords: Farkas lemma, cone-convex systems, strong duality, closed cone condition, second-order cone programming

1. Introduction

Farkas' lemma states that given any vectors a_1, a_2, \dots, a_m in \mathbb{R}^n and for each choice of vector $c \in \mathbb{R}^n$, the linear inequality $c^T x \geq 0$ is a consequence of the linear system $a_i^T x \geq 0$, $i = 1, 2, \dots, m$ if and only if $c = \sum_{i=1}^m \lambda_i a_i$, for some multipliers $\lambda_i \geq 0$. Applications by way of extensions of the celebrated Farkas lemma range from classical nonlinear programming to modern areas of optimization such as nonsmooth optimization and semidefinite programming. For a recent look at its extensions and applications see [3, 7, 12, 18]. However, it is well known that the Farkas lemma for linear systems involving non-polyhedral cones is, in general, not valid without a regularity condition, often called "closed cone condition" (see e.g. [6, 12])

The generalized Farkas lemma for a given cone-convex system $-g(x) \in S$ and for each choice of real-valued convex function f states that $f(x) \geq 0$ is a consequence of the system $-g(x) \in S$ if and only if there exists $\lambda \in S^+$ such that, for each $x \in \mathbb{R}^n$, $f(x) + (\lambda \circ g)(x) \geq 0$. Symbolically,

$$[-g(x) \in S \Rightarrow f(x) \geq 0] \Leftrightarrow (\exists \lambda \in S^+)(\forall x \in \mathbb{R}^n) f(x) + (\lambda \circ g)(x) \geq 0, \quad (1)$$

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where the set $S \subset \mathbb{R}^m$ is a closed convex cone with the dual cone S^+ and $g : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a continuous convex function with respect to S . Various sufficient conditions have been established for this equivalence in (1). For details, see e.g. [6, 10, 11, 12].

The purpose of this paper is to present necessary and sufficient conditions for the Farkas lemma with respect to the given cone-convex system $-g(x) \in S$ and to obtain corresponding duality properties for convex optimization problems. In particular, we establish a necessary and sufficient closed cone condition for the Farkas lemma. As an application, we obtain necessary and sufficient conditions for the strong duality for convex second-order cone programming problems [1, 2, 17], which have received a great deal of attention in recent years.

The outline of the paper is as follows. Section 2 provides background material on convex analysis that will be used later in the paper. Section 3 presents several characterizations of the Farkas lemma, including a necessary and sufficient closed cone (regularity) condition. Section 4 provides an application of the Farkas lemma and establishes characterizations of strong duality properties for second-order cone programming problems and for semidefinite programming problems.

2. Preliminaries

We recall in this section some notations and basic results which will be used in this paper. Let X be a normed space with X^* its dual endowed with weak*-topology. For a subset $D \subset X^*$, the w^* -closure of D will be denoted by $\text{cl } D$ and the convex cone generated by D by $\text{cocone } D$.

Let $h : X \rightarrow \mathbb{R} \cup \{-\infty, +\infty\}$. The conjugate function of h , $h^* : X^* \rightarrow \mathbb{R} \cup \{+\infty\}$, is defined by

$$h^*(v) := \sup\{v(x) - h(x) \mid x \in \text{dom } h\},$$

where $\text{dom } h := \{x \in X \mid h(x) < +\infty\}$ is the effective domain of h . The function h is said to be proper if h does not take on the value $-\infty$ and $\text{dom } h \neq \emptyset$. The **epigraph** of h is defined by

$$\text{epi } h := \{(x, r) \in X \times \mathbb{R} \mid x \in \text{dom } h, h(x) \leq r\}.$$

The set (possibly empty)

$$\partial h(a) := \{v \in X^* \mid h(x) - h(a) \geq v(x - a), \forall x \in \text{dom } h\}$$

is the subdifferential of the convex function h at $a \in \text{dom } h$. For a closed convex subset D of X , the indicator function δ_D is defined as $\delta_D(x) = 0$ if $x \in D$ and $\delta_D(x) = +\infty$ if $x \notin D$. The **support function** δ_D^* is defined by $\delta_D^*(u) = \sup_{x \in D} u(x)$. Then $\partial \delta_D(x) = N_D(x)$, which is known as the normal cone of D of x . If h is proper lower semicontinuous and sublinear (i.e., convex and positively homogeneous of degree one), then $\text{epi } g^* = \partial g(0) \times \mathbb{R}_+$.

For proper lower semicontinuous convex functions $g, h : X \rightarrow \mathbb{R} \cup \{+\infty\}$, the infimal convolution of g with h , denoted $g \square h$, is defined by

$$(g \square h)(x) = \inf_{x_1 + x_2 = x} \{g(x_1) + h(x_2)\}.$$

The lower semicontinuous envelope and lower semicontinuous convex hull of g are denoted respectively by $\text{cl } g$ and $\text{clco } g$. That is, $\text{epi}(\text{cl } g) = \text{cl}(\text{epi } g)$ and $\text{epi}(\text{clco } g) = \text{clco}(\text{epi } g)$. For details, see [21].

Let g, h and $g_i, i \in I$ (where I is an arbitrary index set) be proper lower semicontinuous convex functions. It is well known from the dual operation (see [20], [21]) that if $\text{dom } g \cap \text{dom } h \neq \emptyset$, then

$$(g \square h)^* = g^* + h^*, \quad (g + h)^* = \text{cl}(g^* \square h^*)$$

and if $\sup_{i \in I} g_i$ is proper, then

$$(\sup_{i \in I} g_i)^* = \text{cl co}(\inf_{i \in I} g_i^*).$$

So, one can check that

$$\text{epi}(g + h)^* = \text{cl}(\text{epi } g^* + \text{epi } h^*) \quad \text{and} \quad \text{epi}(\sup_{i \in I} g_i)^* = \text{cl co}(\bigcup_{i \in I} \text{epi } g_i^*).$$

The closure in the first equation is superfluous if one of g and h is continuous at some $x_0 \in \text{dom } g \cap \text{dom } h$ (see [4, 21] for details).

Let Y be another normed linear space with topological dual Y^* and let S be a closed convex cone in Y . Denote by S^+ the dual cone of S , defined as

$$S^+ = \{y^* \in Y^* \mid y^*(y) \geq 0 \text{ for any } y \in S\}.$$

We say that the map $g : X \rightarrow Y$ is S -convex if for any $x_1, x_2 \in X$ and any $\lambda \in [0, 1]$,

$$g(\lambda x_1 + (1 - \lambda)x_2) \in \lambda g(x_1) + (1 - \lambda)g(x_2) - S$$

and that g is S -sublinear if g is S -convex and positively homogeneous of degree 1. Note that $g^{-1}(-S) := \{x \in X \mid -g(x) \in S\}$.

3. Necessary and Sufficient Conditions for Farkas' Lemma

In this section we present several characterizations of Farkas lemma for cone-convex systems. In particular, we show that the well known closed convex cone condition for sublinear systems completely characterize the Farkas lemma.

Theorem 3.1. *Suppose that $g : X \rightarrow Y$ is a continuous and S -convex function with $g^{-1}(-S) \neq \emptyset$. Let $\mathcal{A} = \{l \mid l : X \rightarrow \mathbb{R} \text{ is continuous and convex}\}$. Then the following statements are equivalent:*

(i) For each $f \in \mathcal{A}$,

$$[-g(x) \in S \Rightarrow f(x) \geq 0] \Leftrightarrow (\exists \lambda \in S^+)(\forall x \in X) f(x) + (\lambda \circ g)(x) \geq 0. \quad (1)$$

(ii) For each $f \in \mathcal{A}$,

$$[-g(x) \in S \Rightarrow f(x) \geq 0] \Leftrightarrow (\exists \lambda \in S^+) (0, 0) \in \text{epi } f^* + \text{epi}(\lambda \circ g)^*. \quad (2)$$

(iii) For each $f \in \mathcal{A}$,

$$(\exists \lambda \in S^+) \inf_{x \in X} \{f(x) + (\lambda \circ g)(x)\} = \inf_{x \in g^{-1}(-S)} f(x). \quad (3)$$

(iv) The convex cone,

$$\bigcup_{\lambda \in S^+} \text{epi}(\lambda \circ g)^* \text{ is } w^* \text{-closed.} \quad (4)$$

Proof. [(ii) \iff (i)]. Assume that (ii) holds. Let $f \in \mathcal{A}$. Suppose that $f(x) \geq 0$ for each $x \in g^{-1}(-S)$. Then, by (ii), there exists $\lambda \in S^+$ such that $(0, 0) \in \text{epi } f^* + \text{epi}(\lambda \circ g)^*$. So, there exists $(v, r) \in \text{epi}(\lambda \circ g)^*$ such that $-(v, r) \in \text{epi } f^*$. Now, for each $x \in X$, $v(x) - (\lambda \circ g)(x) \leq r$, and for each $x \in X$, $-v(x) - f(x) \leq -r$. Adding the two inequalities, we get that for each $x \in X$, $f(x) + (\lambda \circ g)(x) \geq 0$. The converse implication in (1) always holds.

Conversely, assume that (i) holds. Let $f \in \mathcal{A}$. Suppose that $f(x) \geq 0$ for each $x \in g^{-1}(-S)$. Then, by (i), there exists $\lambda \in S^+$ such that for each $x \in X$, $f(x) + (\lambda \circ g)(x) \geq 0$. So, $(f + \lambda \circ g)^*(0) \leq 0$ and hence $(0, 0) \in \text{epi}(f + \lambda \circ g)^*$. Since f is continuous, $\text{epi}(f + \lambda \circ g)^* = \text{epi } f^* + \text{epi}(\lambda \circ g)^*$, and consequently $(0, 0) \in \text{epi } f^* + \text{epi}(\lambda \circ g)^*$.

[(iii) \iff (i)]. Suppose that (iii) holds. Let $f \in \mathcal{A}$. If $f(x) \geq 0$ for each $x \in g^{-1}(-S)$ then $\inf_{x \in g^{-1}(-S)} f(x) \geq 0$ and so, from (iii), there exists $\lambda \in S^+$ such that

$$\inf_{x \in X} \{f(x) + (\lambda \circ g)(x)\} \geq 0.$$

Hence (i) holds as the converse implication always holds.

Suppose now that (i) holds. Let $f \in \mathcal{A}$. If $\inf_{x \in g^{-1}(-S)} f(x) = -\infty$, then

$$\inf_{x \in X} \{f(x) + (0 \circ g)(x)\} = \inf_{x \in g^{-1}(-S)} f(x) = -\infty.$$

Since $g^{-1}(-S) \neq \emptyset$, we may assume that $r := \inf_{x \in g^{-1}(-S)} f(x)$ is finite. Since $f(\cdot) - r \in \mathcal{A}$, it follows from (i) that there exists $\lambda \in S^+$ such that $f(x) + (\lambda \circ g)(x) \geq r$ for each $x \in X$. Thus,

$$\inf_{x \in X} \{f(x) + (\lambda \circ g)(x)\} \geq \inf_{x \in g^{-1}(-S)} f(x).$$

Since $f(x) \geq f(x) + (\lambda \circ g)(x)$, for each $x \in g^{-1}(-S)$, we have

$$\inf_{x \in g^{-1}(-S)} f(x) \geq \inf_{x \in X} \{f(x) + (\lambda \circ g)(x)\}.$$

Hence, (iii) holds.

[(ii) \iff (iv)]. Suppose that (ii) holds. Let $(v, r) \in \text{cl} \bigcup_{\lambda \in S^+} \text{epi}(\lambda \circ g)^*$. Then there exist nets $\{\lambda_\alpha\} \subset S^+$, $\{(v_\alpha, r_\alpha)\} \subset \text{epi}(\lambda_\alpha \circ g)^*$ such that $(v_\alpha, r_\alpha) \rightarrow (v, r)$. We can easily check that $v_\alpha(x) \leq r_\alpha$ for each $x \in g^{-1}(-S)$. Letting $\alpha \rightarrow \infty$, we get that $v(x) \leq r$, for each $x \in g^{-1}(-S)$. Let $f(x) = -v(x) + r$. Then for each $x \in g^{-1}(-S)$, $f(x) \geq 0$. Moreover,

$$\text{epi } f^* = \{(-v, -r + \alpha) \mid \alpha \geq 0\}.$$

So, it follows from (ii) that there exist $\lambda \in S^+$ and $\alpha \geq 0$ such that $(v, r - \alpha) \in \text{epi}(\lambda \circ g)^*$. This gives us that $(v, r) \in \text{epi}(\lambda \circ g)^*$. Hence $\bigcup_{\lambda \in S^+} \text{epi}(\lambda \circ g)^*$ is w^* -closed.

Conversely, assume that (iv) holds. Since $(S^+)^+ = S$, for each $x \in X$,

$$\sup_{\lambda \in S^+} (\lambda \circ g)(x) = \delta_{g^{-1}(-S)}(x).$$

So,

$$\text{epi } \delta_{g^{-1}(-S)}^* = \text{cl co} \left(\bigcup_{\lambda \in S^+} \text{epi}(\lambda \circ g)^* \right).$$

As the set $\bigcup_{\lambda \in S^+} \text{epi}(\lambda \circ g)^*$ is a convex cone [15], it follows from (iv),

$$\text{epi } \delta_{g^{-1}(-S)}^* = \bigcup_{\lambda \in S^+} \text{epi}(\lambda \circ g)^*.$$

Now,

$$\begin{aligned}
 & f(x) \geq 0 \text{ for each } x \in g^{-1}(-S) \\
 \iff & (f + \delta_{g^{-1}(-S)})^*(0) \leq 0 \\
 \iff & (0, 0) \in \text{epi}(f + \delta_{g^{-1}(-S)})^* \\
 \iff & (0, 0) \in \text{epi } f^* + \text{epi } \delta_{g^{-1}(-S)}^* \\
 \iff & (0, 0) \in \text{epi } f^* + \bigcup_{\lambda \in S^+} \text{epi}(\lambda \circ g)^*.
 \end{aligned}$$

Hence (ii) holds. □

The following examples illustrate the significance of the closed cone condition (iv) in Theorem 3.1 for Farkas lemma for cone-convex systems.

Example 3.2. Let $g(x) := \max\{0, x\}$, $S = \mathbb{R}_+$ and $\mathcal{A} = \{l \mid l : \mathbb{R} \rightarrow \mathbb{R} \text{ is convex}\}$. Then $g : \mathbb{R} \rightarrow \mathbb{R}$ is a convex function and $g^{-1}(-S) = (-\infty, 0]$. We can easily check that $\bigcup_{\lambda \in S^+} \text{epi}(\lambda \circ g)^* = \mathbb{R}_+^2$.

For each $f \in \mathcal{A}$, we can choose $\lambda \in \mathbb{R}$ such that $\lambda \geq \max\{-v, 0\}$, where $v \in \partial f(0)$. Now, if, for each $x \in g^{-1}(-S)$, $f(x) \geq 0$, then for each $x \in \mathbb{R}$, $f(x) + \lambda g(x) \geq 0$. Indeed, if $x \leq 0$, then $f(x) + \lambda g(x) = f(x) \geq 0$; if $x \geq 0$, then

$$f(x) + \lambda g(x) \geq vx + \lambda g(x) \geq (v + \lambda)x \geq 0.$$

On the other hand, if, for some $\lambda \geq 0$, $f(x) + \lambda g(x) \geq 0$, for each $x \in \mathbb{R}$, then, clearly, $f(x) \geq 0$, for each $x \in g^{-1}(-S)$. Now, let $r := \inf_{x \in g^{-1}(-S)} f(x)$. If r is finite, then $f - r \in \mathcal{A}$ and so, for each $x \in \mathbb{R}$, $f(x) + \lambda g(x) \geq r$. This gives us that $\inf_{x \in \mathbb{R}} f(x) + \lambda g(x) \geq r$. By the weak duality, $r = \max_{\mu \in \mathbb{R}^+} \inf_{x \in \mathbb{R}} f(x) + \mu g(x)$. If $r = -\infty$ then this equality trivially holds. Hence Theorem 3.1 holds. It is worth noting that the Slater condition that $g(x_0) < 0$ for some $x_0 \in \mathbb{R}$, does not hold for this Example.

Example 3.3. Let $g(x) := [\max\{0, x\}]^2$, $S = \mathbb{R}_+$. Then $g : \mathbb{R} \rightarrow \mathbb{R}$ is a convex function and $g^{-1}(-S) = (-\infty, 0]$. Then if $\lambda = 0$, $\text{epi}(\lambda \circ g)^* = \{0\} \times \mathbb{R}_+$, and if $\lambda > 0$, $\text{epi}(\lambda \circ g)^* = \{(v, \alpha) \in \mathbb{R}^2 \mid v \geq 0, \frac{v^2}{4\lambda} \leq \alpha\}$, and hence $\bigcup_{\lambda \in S^+} \text{epi}(\lambda \circ g)^*$ is not closed. Let $f(x) = -x$. Then

$\text{epi } f^* = \{-1\} \times \mathbb{R}_+$, and hence we can not find $\lambda \geq 0$ such that $(0, 0) \in \text{epi } f^* + \text{epi}(\lambda \circ g)^*$. Thus, Theorem 3.1 fails to hold.

The generalized Farkas lemma for cone-sublinear systems has been well known under the closed cone condition that $\bigcup_{\lambda \in S^+} \partial(\lambda \circ g)(0)$ is w^* -closed. For details, see [8, 12] and other references therein. We now show that this closed cone condition completely characterizes the Farkas lemma for the cone-sublinear systems.

Corollary 3.4. Let $g : X \rightarrow Y$ be a continuous and S -sublinear function. Let $\mathcal{B} = \{l \mid l : X \rightarrow \mathbb{R} \text{ is continuous and sublinear}\}$. Then the following statements are equivalent:

(i) For each $f \in \mathcal{B}$,

$$[-g(x) \in S \Rightarrow f(x) \geq 0] \Leftrightarrow (\exists \lambda \in S^+) 0 \in \partial f(0) + \partial(\lambda \circ g)(0).$$

(ii) $\bigcup_{\lambda \in S^+} \partial(\lambda \circ g)(0)$ is w^* -closed.

Proof. Since g is S -sublinear,

$$\bigcup_{\lambda \in S^+} \text{epi}(\lambda \circ g)^* = \bigcup_{\lambda \in S^+} \partial(\lambda \circ g)(0) \times \mathbb{R}^+$$

and hence (iv) of Theorem 3.1 holds if and only if $\bigcup_{\lambda \in S^+} \partial(\lambda \circ g)(0)$ is w^* -closed. Since f is sublinear and g is S -sublinear,

$$f(x) + (\lambda \circ g)(x) \geq 0 \quad \forall x \in X \Leftrightarrow 0 \in \partial f(0) + \partial(\lambda \circ g)(0).$$

Thus the conclusion follows from Theorem 3.1. □

Corollary 3.5. *Let $A : X \rightarrow Y$ be continuous and linear. Then the following statements are equivalent:*

(i) $\forall c \in X^*, [-Ax \in S \Rightarrow c(x) \geq 0] \Leftrightarrow (\exists \lambda \in S^+) c + A^T \lambda = 0.$

(ii) $A^T(S^+)$ is w^* -closed.

Proof. Let $g(x) = Ax$. Then $\bigcup_{\lambda \in S^+} \partial(\lambda \circ g)(0) = A^T(S^+)$. Then, the conclusion follows from Corollary 3.4. □

Remark 3.6. When $Y = \mathbb{R}^m$ and S is a polyhedral convex cone in Y , then $A^T(S^+)$ is a finitely generated cone and hence $A^T(S^+)$ is closed. So, from Corollary 3.5, the original Farkas lemma follows (see [5]).

The following corollary, which can be used to study convex vector optimization problems, provides solvability conditions for systems involving convex vector functions f . See [16, 14] for recent results for convex vector optimization problems.

Corollary 3.7. *Let $g : X \rightarrow Y$ be a continuous and S -convex function with $g^{-1}(-S) \neq \emptyset$. Let $\mathcal{C} = \{l \mid l : X \rightarrow \mathbb{R}^p \text{ is continuous and } \mathbb{R}_+^p\text{-convex}\}$. Let $\Lambda_p = \{x \in \mathbb{R}^p \mid \sum_{i=1}^p x_i = 1\}$. Then the following statements are equivalent:*

(i) For each $f \in \mathcal{C}$,

$$\begin{aligned} & [-g(x) \in S \Rightarrow f(x) \notin -\text{int } \mathbb{R}_+^p] \\ & \Leftrightarrow (\exists \theta \in \Lambda_p)(\exists \lambda \in S^+)(\forall x \in X) \theta^T f(x) + (\lambda \circ g)(x) \geq 0. \end{aligned}$$

(ii) For each $f \in \mathcal{C}$,

$$\begin{aligned} & [-g(x) \in S \Rightarrow f(x) \notin -\text{int } \mathbb{R}_+^p] \\ & \Leftrightarrow (\exists \theta \in \Lambda_p)(\exists \lambda \in S^+) (0, 0) \in \text{epi}(\theta^T f)^* + \text{epi}(\lambda \circ g)^*. \end{aligned}$$

(iii) $\bigcup_{\lambda \in S^+} \text{epi}(\lambda \circ g)^*$ is w^* -closed.

Proof. Let $f \in \mathcal{C}$. The Hahn-Banach separation theorem [20] ensures that

$$[-g(x) \in S \Rightarrow f(x) \notin -\text{int } \mathbb{R}_+^p] \Leftrightarrow (\exists \theta \in \Lambda_p) [-g(x) \in S \Rightarrow \theta^T f(x) \geq 0].$$

Now, from Theorem 3.1, (i) is equivalent to (ii), and (iii) implies (ii).

To establish that (ii) implies (iii) suppose that (ii) holds. Let $(v, r) \in \bigcup_{\lambda \in S^+} \text{epi}(\lambda \circ g)^*$. Then $v(x) \leq r$ for each $x \in g^{-1}(-S)$. Let $f(x) = (r - v(x), \dots, r - v(x))$. Then $f(x) \notin -\text{int } \mathbb{R}_+^p$, and from (ii), there exist $\theta \in \Lambda_p$ and $\lambda \in S^+$ such that $(0, 0) \in \text{epi}(\theta^T f)^* + \text{epi}(\lambda \circ g)^*$. Noting that $\theta^T f(x) = r - v(x)$, the same arguments of the proof of Theorem 3.1 lead to the condition that $(v, r) \in \text{epi}(\lambda \circ g)^*$. Hence (iii) holds. □

4. Characterizations of Duality in Cone Programming

In this section, we examine second-order cone programming problems which arise in various application areas (see e.g. [2, 17]). As an application of our Farkas lemma, we first present a constraint qualification which completely characterizes the strong duality of a second-order cone programming problem. We then derive a necessary and sufficient condition for the strong min-max duality whenever the given problem attains its infimum.

Consider the following convex second-order cone program:

$$(SOCP) \quad \inf f(x) \\ \text{subject to } x \in M := \{x \in \mathbb{R}^n \mid \|Hx + b\| \leq c^T x + d\},$$

where $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is a convex function and H is an $(m - 1) \times n$ matrix, $b \in \mathbb{R}^{m-1}$, $c \in \mathbb{R}^n$, $d \in \mathbb{R}$ and $\|z\| = \sqrt{z^T z}$, $z \in \mathbb{R}^{m-1}$.

Suppose that $M \neq \emptyset$. Let $K = \{(y, t)^T \in \mathbb{R}^{m-1} \times \mathbb{R} \mid \|y\| \leq t\}$, that is, K is a second-order cone in \mathbb{R}^m . Then K is self-dual, that is, $K = K^+$.

Theorem 4.1. *Let $\mathcal{A} = \{l \mid l : \mathbb{R}^n \rightarrow \mathbb{R} \text{ is convex}\}$. Then the following statements are equivalent:*

(i) For each $f \in \mathcal{A}$,

$$\max_{\lambda \in K} \inf_{x \in \mathbb{R}^n} \left\{ f(x) - \lambda^T \begin{pmatrix} Hx + b \\ c^T x + d \end{pmatrix} \right\} = \inf_{x \in M} f(x).$$

(ii) $\bigcup_{\lambda \in K} \left\{ \left(- \begin{pmatrix} H \\ c^T \end{pmatrix}^T \lambda, \begin{pmatrix} b \\ d \end{pmatrix}^T \lambda \right) \right\} + \{0\} \times \mathbb{R}^+$ is closed.

Proof. Note that

$$\|Hx + b\| \leq c^T x + d \Leftrightarrow \begin{pmatrix} Hx + b \\ c^T x + d \end{pmatrix} \in K.$$

Let $g(x) = - \begin{pmatrix} Hx + b \\ c^T x + d \end{pmatrix}$. Then we have,

$$\bigcup_{\lambda \in K} (\lambda \circ g)^* = \bigcup_{\lambda \in K} \left\{ \left(- \begin{pmatrix} H \\ c^T \end{pmatrix}^T \lambda, \begin{pmatrix} b \\ d \end{pmatrix}^T \lambda \right) \right\} + \{0\} \times \mathbb{R}^+.$$

Hence the conclusions follow from Theorem 3.1. □

Remark 4.2. In Theorem 4.1, if $b = 0$ and $d = 0$, then (ii) is equivalent to the fact that $(H^T, c)(K)$ is closed. Moreover, if $m = n$ and the $n \times n$ matrix (H^T, c) is nonsingular, then the set $(H^T, c)(K)$ is closed.

Now we derive a new condition for the min-max Lagrangian duality for (SOCP), where $\inf_{x \in M} f(x)$ is attained.

Theorem 4.3. *Let $\mathcal{A} = \{l \mid l : \mathbb{R}^n \rightarrow \mathbb{R} \text{ is convex}\}$. Then the following statements are equivalent:*

(i) For each $x \in M$,

$$N_M(x) = \left\{ - \begin{pmatrix} H \\ c^T \end{pmatrix}^T \lambda \mid \lambda \in K, \lambda^T \begin{pmatrix} Hx + b \\ c^T x + d \end{pmatrix} = 0 \right\}.$$

(ii) For each $f \in \mathcal{A}$,

$$\min_{x \in M} f(x) = \max_{\lambda \in K} \inf_{x \in \mathbb{R}^n} \left\{ f(x) - \lambda^T \begin{pmatrix} Hx + b \\ c^T x + d \end{pmatrix} \right\}.$$

Proof. Suppose that (i) holds. Let $f \in \mathcal{A}$. Assume that $x \in M$ and that $f(x) = \min_{y \in M} f(y)$. Then, by optimality condition, $0 \in \partial f(x) + N_M(x)$ and so from (i), there exists $\lambda \in K$ such that

$$0 \in \partial f(x) - \begin{pmatrix} H \\ c^T \end{pmatrix}^T \lambda \quad \text{and} \quad \lambda^T \begin{pmatrix} Hx + b \\ c^T x + d \end{pmatrix} = 0.$$

This gives us that

$$\inf_{y \in \mathbb{R}^n} \left\{ f(y) - \lambda^T \begin{pmatrix} Hy + b \\ c^T y + d \end{pmatrix} \right\} \geq f(x).$$

Hence, by weak duality, (ii) holds.

Conversely, assume that (ii) holds. Let Ω be the right side of the equality in (i). Let $x \in M$ and $u \in N_M(x)$. Then $-u^T x = \min_{y \in M} (-u^T y)$. So, it follows from (ii) that there exists $\lambda \in K$ such that, for each $y \in \mathbb{R}^n$

$$-u^T x \leq -u^T y - \lambda^T \begin{pmatrix} Hy + b \\ c^T y + d \end{pmatrix}.$$

This gives us that

$$\lambda^T \begin{pmatrix} Hx + b \\ c^T x + d \end{pmatrix} = 0 \quad \text{and} \quad u = - \begin{pmatrix} H \\ c^T \end{pmatrix}^T \lambda.$$

Thus $u \in \Omega$. On the other hand, the inclusion $\Omega \subset N_M(x)$ always holds. Hence (i) holds. \square

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