



ON EXISTENCE OF OPTIMAL SOLUTION FOR SUBLINEAR PROGRAMS*

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Dedicted to Professor Guang Ya Chen on his 70th birthday.

Abstract: A sublinear program (P), which involves a sublinear objective function and a constrained set defined by a cone-sublinear function and a closed convex cone, is considered. We show that the existence of optimal solutions for (P) is closely related to zero solution and that a condition for the existence can be expressed in terms of subdifferentials of the functions involved in (P).

 ${\bf Key \ words:}\ sublinear\ program,\ optimal\ solutions,\ existence,\ subdifferential$

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1 Introduction and Preliminaries

Consider the following sublinear program:

(P) Minimize f(x)subject to $g(x) \in -S$,

where X and Y are Banach spaces, S is a closed convex cone in Y, which does not necessarily have nonempty interior, and the mappings $f: X \to \mathbb{R}$ and $g: X \to Y$ are a continuous sublinear function and a continuous S-sublinear function, respectively. Recall that g is said to be S-sublinear if

(i) $\forall x \in X \ \forall \lambda \geq 0, \ g(\lambda x) = \lambda g(x),$

(ii) $\forall x, y \in X$, $g(x) + g(y) - g(x+y) \in S$.

We assume that the feasible set $A := \{x \in X \mid g(x) \in -S\}$ is nonempty. We denote the set of all solutions of (P) by sol(P). The continuous dual space of Y is denoted by Y^* and is endowed with the weak^{*} topology. The (positive) polar of the cone $S \subseteq Y$ is the cone $S^+ = \{\theta \in Y^* \mid \theta(k) \ge 0 \ \forall k \in S\}.$

Recently, many authors [2, 4–7] extended Farkas Lemma ([1]) to convex systems.

In this brief paper, we show that the existence of optimal solutions for (P) is closely related to zero solution and that a condition for the existence can be expressed in terms of subdifferentials of the functions involved in (P). The condition is obtained from a generalized Farkas Theorem in [6].

Now we give notations and preliminary results that will be used later.

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Definition 1.1 ([3,8]). Let X be a Banach space, let X^* be the continuous dual space of X and let $h: X \to \mathbb{R} \cup \{+\infty\}$ be a proper lower-semicontinuous convex function.

(1) The conjugate function of $h, h^* : X^* \to \mathbb{R} \cup \{+\infty\}$ is defined by

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

where the domain of h, domh, is given by

$$\operatorname{dom} h = \{ x \in X \mid h(x) < +\infty \}.$$

(2) The epigraph of h, epih, is defined by

$$epih = \{(x, r) \in X \times \mathbb{R} \mid x \in domh, h(x) \le r\}.$$

(3) The subdifferential of h at $a \in {\rm dom} h$ is defined as the non-empty weak* compact convex set

$$\partial h(a) := \{ v \in X^* \mid h(x) - h(a) \ge v(x - a), \quad \forall x \in \operatorname{dom} h \}.$$

It is well-known that if h is sublinear (i.e., convex and positively homogeneous of degree one), then

$$epih^* = \partial h(0) \times \mathbb{R}_+.$$

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.

Let $S \subseteq Y$ be a closed convex cone. Then we say that the mapping $g : X \to Y$ is S-convex if for any $x_1, x_2 \in X$ and any $\lambda \in [0, 1]$,

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

Now we give a generalized Farkas Lemma to convex systems.

Lemma 1.2 ([6]). Let $S \subseteq Y$ be a closed convex cone, let $u : X \to \mathbb{R}$ be a continuous linear mapping, and let $g : X \to Y$ be a continuous S-convex mapping. Suppose that the system $g(x) \in -S$ is consistent. Let $\alpha \in \mathbb{R}$. Then the following statements are equivalent: (i) $\{x \in X \mid g(x) \in -S\} \subseteq \{x \in X \mid u(x) \le \alpha\}$.

(ii)
$$\begin{pmatrix} u \\ \alpha \end{pmatrix}^T \in cl \Big(\bigcup_{\lambda \in S^+} \operatorname{epi}(\lambda \circ g)^*\Big).$$

2 Existence of Optimal Solution

The following proposition is needed to prove the main results.

Proposition 2.1. If $sol(P) \neq \emptyset$, then $sol(P) = \{x \in A \mid f(x) = 0\}$.

Proof. Suppose that $sol(\mathbf{P}) \neq \emptyset$. Let $\bar{x} \in A$ be a solution of (P). Then for any $x \in A$, $f(\bar{x}) \leq f(x)$. Since $\bar{x} \in A$ and g is S-sublinear,

$$g(\alpha \bar{x}) = \alpha g(\bar{x}) \in -S$$
 for any $\alpha \ge 0$.

Thus $\alpha \bar{x} \in A$ for any $\alpha \geq 0$, and hence $f(\bar{x}) \leq f(\alpha \bar{x}) = \alpha f(\bar{x})$ for any $\alpha \geq 0$. Since $0 \in A$, $f(\bar{x}) \leq f(0) = 0$. If $f(\bar{x}) < 0$, $f(\alpha \bar{x}) = \alpha f(\bar{x}) \to -\infty$ as $\alpha \to \infty$. This is impossible since (P) has a solution. Therefore $f(\bar{x}) = 0$. So, we have,

$$sol(P) = \{x \in A \mid f(x) = 0\}.$$

Now we give an example to illustrate Proposition 2.1.

Example 2.2. Let $f(x) = g(x) = \begin{cases} 0 & \text{if } x \leq 0 \\ x & \text{if } x > 0 \end{cases}$. Consider the following sublinear program:

(P) Minimize
$$f(x)$$

subject to $x \in A := \{x \mid g(x) \in -\mathbb{R}_+ \}.$

Then it is clear that (P) has a solution. Moreover,

$$sol(\mathbf{P}) = \{x \in A \mid f(x) = 0\} \\ = \{x \in (-\infty, 0] \mid f(x) = 0\} \\ = (-\infty, 0].$$

So, Proposition 2.1 holds.

Now we give a well-known formula for normal cone ([2]). For the completeness, we give a proof for the formula.

Proposition 2.3. Let $g : X \to Y$ be a continuous S-sublinear function and $A := \{x \in X \mid g(x) \in -S\}$. Then $N_A(0) = cl\Big(\bigcup_{\lambda \in S^+} \partial(\lambda \circ g)(0)\Big)$.

Proof. $v \in N_A(0)$

$$\begin{array}{ll} \Longleftrightarrow & \forall x \in A, \ v(x) \leq 0 \\ \Leftrightarrow & A \subset \{v \mid v(x) \leq 0\} \\ \Leftrightarrow & (\text{by Lemma 1.2}) \quad \begin{pmatrix} v \\ 0 \end{pmatrix}^T \in cl \Big(\bigcup_{\lambda \in S^+} \operatorname{epi}(\lambda \circ g)^*\Big) \end{array}$$

$$\begin{array}{ll} \Longleftrightarrow & (\text{since } g \text{ is } S - \text{sublinear}) \quad \begin{pmatrix} v \\ 0 \end{pmatrix}^T \in cl \bigcup_{\lambda \in S^+} \left(\partial(\lambda \circ g)(0) \times \mathbb{R}_+ \right) \\ \\ \Leftrightarrow & \begin{pmatrix} v \\ 0 \end{pmatrix}^T \in cl \Big(\bigcup_{\lambda \in S^+} \partial(\lambda \circ g)(0) \Big) \times \mathbb{R}_+ \\ \\ \\ \Leftrightarrow & v \in cl \Big(\bigcup_{\lambda \in S^+} \partial(\lambda \circ g)(0) \Big). \end{array}$$

We give a theorem which shows that the existence of optimal solutions for (P) is closely related to zero solution and that a condition for the existence can be expressed in terms of subdifferentials of the functions involved in (P).

Theorem 2.4. The following statements are equivalent:

(i)
$$sol(P) \neq \emptyset$$
.
(ii) $0 \in sol(P)$.
(iii) $0 \in \partial f(0) + cl\left(\bigcup_{\lambda \in S^+} \partial(\lambda \circ g)(0)\right)$.

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Proof. (i) \Rightarrow (ii): Suppose that $sol(P) \neq \emptyset$. Then by Proposition 2.1, $sol(P) = \{x \in \mathbb{R}\}$ $A \mid f(x) = 0$. Since f is sublinear, f(0) = 0 and hence $0 \in sol(P)$.

(ii) \Rightarrow (i): It is trivial.

(ii) \Rightarrow (iii): Suppose that $0 \in sol(P)$. Then $f(x) + \delta_A(x) \geq f(0) + \delta_A(0) + 0(x - 0)$ for any $x \in A$ and hence $0 \in \partial(f + \delta_A)(0) = \partial f(0) + \partial \delta_A(0) = \partial f(0) + N_A(0)$. Thus by Proposition 2.3, $0 \in \partial f(0) + cl \left(\bigcup_{\lambda \in S^+} \partial(\lambda \circ g)(0) \right).$

(iii) \Rightarrow (ii): Suppose that (iii) holds. Then it follows from Proposition 2.3 that there exists $v \in \partial f(0)$ such that $-v \in N_A(0) = \partial \delta_A(0)$. Hence $0 \in \partial f(0) + \partial \delta_A(0) = \partial (f + \delta_A)(0)$. So, $f(x) + \delta_A(x) \ge f(0) + \delta_A(0) + 0(x-0)$ for any $x \in X$ and hence $f(x) \ge f(0)$ for any $x \in A$. This means that $0 \in sol(\mathbf{P})$.

Remark 2.5. The equivalence between (ii) and (iii) in Theorem 2.4 was already established by Glover ([2]). However, we give its proof in order to show that a condition for the existence of solution of (P) can be expressed in terms of subdifferentials of the functions involved in (P).

Consider the following linear program:

(LP) Minimize
$$c(x)$$

subject to $A(x) \in -S$

where X and Y are Banach spaces, S a closed convex cone in Y, which does not necessarily have nonempty interior, and the mappings $c: X \to \mathbb{R}$ and $A: X \to Y$ are continuous and linear functions. We denote the adjoint operator of the linear mapping A by A^T . We denote the set of all solutions of (LP) by sol(LP). Then we can easily obtain the following corollary from Theorem 2.4.

Corollary 2.6. The following statements are equivalent:

(i) $sol(LP) \neq \emptyset$. (ii) $0 \in sol(LP)$. (iii) $0 \in c + cl(A^T(S^+))$.

Now we give two examples illustrating Theorem 2.4. **Example 2.7.** Let $f(x) = \begin{cases} x & \text{if } x \leq 0 \\ 2x & \text{if } x > 0 \end{cases}$ and $g(x) = \begin{cases} 0 & \text{if } x \leq 0 \\ x & \text{if } x > 0 \end{cases}$. Consider the following sublinear program:

(P) Minimize
$$f(x)$$

subject to $x \in A := \{x \in \mathbb{R} \mid g(x) \in -\mathbb{R}_+\}.$

Then f(0) = 0, but (P) has no solution. Moreover, $\partial(\lambda g)(0) = [0, \lambda]$ for any $\lambda \geq 0$ and hence $\bigcup_{\lambda \ge 0} \partial(\lambda g)(0) = \mathbb{R}_+$. But $\partial f(0) = [1, 2]$ and hence $\partial f(0) + cl \left(\bigcup_{\lambda \ge 0} \partial(\lambda g)(0)\right) = [1, \infty)$.

So, Theorem 2.4 does not hold.

Example 2.8. Let f(x,y) = x and $g(x,y) = (x^2 + y^2)^{\frac{1}{2}} - y$. Consider the following sublinear program:

(P) Minimize
$$f(x, y)$$

subject to $(x, y) \in A := \{(x, y) \in \mathbb{R}^2 \mid g(x, y) \in -\mathbb{R}_+ \}.$

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Then $A = \{(0, y) \mid y \geq 0\}$ and sol(P) = A. Let $(\bar{x}, \bar{y}) = (0, 1)$. Then $(\bar{x}, \bar{y}) \in sol(P)$ and $f(\bar{x}, \bar{y}) = 0$. Moreover, $\partial f(0, 0) = \{(1, 0)\}$ and $cl(\bigcup_{\lambda \in \mathbb{R}_+} \partial(\lambda g)(0, 0)) = \{(v_1, v_2) \mid v_1 \in \mathbb{R}, v_2 \leq 0\}$. Thus

$$(0,0) \in \partial f(0,0) + cl \Big(\bigcup_{\lambda \in \mathbb{R}^+} \partial(\lambda g)(0,0)\Big).$$

So, Theorem 2.4 holds.

References

- J. Farkas, Uber die Theorie der einfachen Ungleichungen, Journal f
 ür die Reine und Angewandte Mathematik 124 (1902) 1–24.
- [2] B.M. Glover, A generalized Farkas lemma with applications to quasidifferentiable programming, Zeitschrift für Operations Research, Series A-B 26 (1982) 125–141.
- [3] J.B. Hiriart-Urruty and C. Lemarechal, Convex Analysis and Minimization Algorithms, Volumes I and II, Springer-Verlag, Berlin, Heidelberg, 1993.
- [4] V. Jeyakumar, Asymptotic dual conditions characterizing optimality for convex programs, J. Optim. Th. Appl. 93 (1997) 153–165.
- [5] V. Jeyakumar, G.M. Lee and N. Dinh, Characterization of solution sets of convex vector minimization problems, *Eur. J. Oper. Res.* 174 (2006) 1380–1395.
- [6] V. Jeyakumar, G.M. Lee and N. Dinh, New sequential Lagrange multiplier conditions characterizing optimality without constraint qualification for convex program, SIAM J. Optim. 14 (2003) 534–547.
- [7] V. Jeyakumar and G.M. Lee, Complete characterizations of stable Farkas' lemma and cone-convex programming duality, Math. Programming 114 (2008) 335–347.
- [8] C. Zalinescu, Convex Analysis in General Vector Space, World Scientific Pub., Singapore, 2002.

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