

PROPOSITION 11.5. Let  $K \subseteq \mathbb{R}^n$  and  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  be differentiable at  $\bar{x} \in K$ . If  $\bar{x}$  is a local minimizer of  $f$  in  $K$ , then

$$\langle \nabla f(\bar{x}), v \rangle \geq 0, \quad \forall v \in T(K; \bar{x}).$$

In other words,

$$F_0 \cap T(K; \bar{x}) = \emptyset,$$

where  $F_0 = \{v \in \mathbb{R}^n : \langle \nabla f(\bar{x}), v \rangle < 0\}$ .

PROOF. Let  $v \in T(K; \bar{x})$ . Then, there exist  $\lambda_k > 0$ ,  $x_k \in K$ , such that  $x_k \rightarrow \bar{x}$  and  $\lambda_k(x_k - \bar{x}) \rightarrow v$ . By Taylor series

$$f(x_k) - f(\bar{x}) = \langle \nabla f(\bar{x}), x_k - \bar{x} \rangle + \|x_k - \bar{x}\| o(\|x_k - \bar{x}\|),$$

with  $o(t) \rightarrow 0$  as  $t \rightarrow 0$ . As  $\bar{x}$  is a local minimizer,  $f(\bar{x}) \leq f(x_k)$  for all  $k$  sufficiently large. This along with the previous equality after multiplying by  $\lambda_k$ , and letting  $k \rightarrow +\infty$ , we get  $\langle \nabla f(\bar{x}), v \rangle \geq 0$ .  $\square$

EXERCISE 11.4. Provide an instance where the converse implication in Proposition 11.5 does not hold.

REMARK 11.2. When  $K$  is convex, then by the preceding proposition we have

$$T(K; \bar{x}) = R(K; \bar{x}) = \overline{\bigcup_{t>0} t(K - \bar{x})} \quad \forall \bar{x} \in K.$$

Thus, the standard optimality condition is recovered:  $\langle \nabla f(\bar{x}), x - \bar{x} \rangle \geq 0$  for all  $x \in K$ . In such a case  $[T(K; \bar{x})]^* = -N(K; \bar{x})$  as one can check it directly.

In what follows, we use the equality:

$$\mathbb{R}_+^m + \text{int } \mathbb{R}_+^m = \text{int } \mathbb{R}_+^m,$$

which follows from

$$\mathbb{R}_+^m + \text{int } \mathbb{R}_+^m \subseteq \mathbb{R}_+^m + \mathbb{R}_+^m = \mathbb{R}_+^m,$$

and so

$$\mathbb{R}_+^m + \text{int } \mathbb{R}_+^m \subseteq \text{int } \mathbb{R}_+^m \subseteq \mathbb{R}_+^m + \text{int } \mathbb{R}_+^m.$$

More generally, one can check that

$$P + \text{int } P = \text{int } P,$$

provided  $P$  is a convex cone with nonempty interior.

THEOREM 11.6. (An alternative Gordan-type theorem) Let  $K \subseteq \mathbb{R}^n$  be convex;  $f_i : K \rightarrow \mathbb{R}$ ,  $i = 1, \dots, m$  be convex functions. Prove that exactly one of the following two systems admits a solution:

(I)  $f_i(x) < 0$ ,  $i = 1, \dots, m$ ,  $x \in K$ ;

(II)  $p \in \mathbb{R}_+^m \setminus \{0\}$ ,  $\sum_{i=1}^m p_i f_i(x) \geq 0$ ,  $\forall x \in K$ .

PROOF. Obviously both systems cannot have solutions simultaneously.

Let  $F = (f_1, \dots, f_m)$  be the function defined by  $F(x) = (f_1(x), \dots, f_m(x))$ . Suppose that (I) has no solution. Then,  $F(x) \notin -\text{int } \mathbb{R}_+^m$ , for all  $x \in K$ , that is,  $F(K) \cap (-\text{int } \mathbb{R}_+^m) = \emptyset$ . This is equivalent to  $(F(K) + \mathbb{R}_+^m) \cap (-\text{int } \mathbb{R}_+^m) = \emptyset$ . Due to convexity of each  $f_i$ , we obtain the convexity of  $F(K) + \mathbb{R}_+^m$ . By applying a standard theorem on separation of convex sets, we get the existence of  $p \in \mathbb{R}^n \setminus \{0\}$  and  $\alpha \in \mathbb{R}$  such that

$$(11.3) \quad \langle p, z \rangle \geq \alpha, \quad \forall z \in F(K) + \mathbb{R}_+^m.$$

$$(11.4) \quad \langle p, w \rangle \leq \alpha, \quad \forall w \in -\text{int } \mathbb{R}_+^m.$$

From (11.4) it follows that  $\alpha \geq 0$ , and from (11.3), we get  $\langle p, F(x) + q \rangle \geq \alpha$  for all  $x \in K$  and all  $q \geq 0$ . Hence  $\langle p, F(x) \rangle \geq 0$  for all  $x \in K$  and  $p \geq 0$ , proving the desired result.  $\square$

REMARK 11.3. It is not difficult to check that for any set  $A \subseteq \mathbb{R}^m$ ,

$$A \cap (-\text{int } \mathbb{R}_+^m) = \emptyset \iff \overline{A} \cap (-\text{int } \mathbb{R}_+^m) = \emptyset \iff (A + \mathbb{R}_+^m) \cap (-\text{int } \mathbb{R}_+^m) = \emptyset$$

$$\iff \overline{\text{cone}}(A + \mathbb{R}_+^m) \cap (-\text{int } \mathbb{R}_+^m) = \emptyset.$$

As a consequence, the previous theorem remains valid under the convexity of  $\overline{\text{cone}}(F(K) + \mathbb{R}_+^m)$ .

EXERCISE 11.5. Provide functions  $F = (f_1, \dots, f_m) : K \rightarrow \mathbb{R}^m$  such that  $F(K) + \mathbb{R}_+^m$  is convex without being each  $f_i : K \rightarrow \mathbb{R}$ . Similarly, find a function  $F$  such that  $\overline{\text{cone}}(F(K) + \mathbb{R}_+^m)$  is convex but not the set  $F(K) + \mathbb{R}_+^m$ .

We recall the notion of pointedness of a cone: a cone  $P$  is pointed if  $\text{co } P \cap (-\text{co } P) = \{0\}$ . Further, we will use the fact that

$$\text{int } P = \{p \in P : \langle q, p \rangle > 0 \quad \forall q \in P^*, q \neq 0\},$$

and

$$P = P^{**} = \{p \in \mathbb{R}^n : \langle q, p \rangle \geq 0 \quad \forall q \in P^*\},$$

whenever  $P$  is any convex, closed cone with nonempty interior.

THEOREM 11.7. Let  $P \subseteq \mathbb{R}^m$  be a convex closed cone such that  $\text{int } P \neq \emptyset$ , and  $A \subseteq \mathbb{R}^m$  be any nonempty set. Then the following assertions are equivalent:

- (a)  $\exists \lambda^* \in P^* \setminus \{0\}, \langle \lambda^*, a \rangle \geq 0 \quad \forall a \in A$ ;
- (b)  $\text{cone}(A + \text{int } P)$  is pointed;
- (c)  $\text{co}(A) \cap (-\text{int } P) = \emptyset$ .

PROOF. (a)  $\implies$  (b): Suppose that  $0 = \sum_{i=1}^l x_i$  with  $x_i \in \text{cone}(A + \text{int } P)$ , we shall prove that  $x_i = 0$  for all  $i$ . By choice,  $x_i = t_i(a_i + p_i)$  with  $t_i \geq 0$ ,  $a_i \in A$ ,  $p_i \in \text{int } P$  for  $i = 1, \dots, l$ . This implies that  $\sum_{i=1}^l t_i a_i \in -\text{int } P$ . This yields a contradiction if  $\sum_{i=1}^l t_i > 0$  under (a), since the inequality in (a) also holds for all  $a \in \text{co}(A)$ , and

$$\text{int } P = \{p \in P : \langle q, p \rangle > 0 \quad \forall q \in P^*, q \neq 0\}.$$

(b)  $\implies$  (c): Assume on the contrary that  $\text{co}(A) \cap (-\text{int } P) \neq \emptyset$ . Then, there exist  $a_i \in A$ ,  $p_0 \in \text{int } P$ ,  $\alpha_i \geq 0$ , satisfying  $\sum_{i=1}^l \alpha_i = 1$  and

$0 = \sum_{i=1}^l \alpha_i a_i + p_0$ . Thus,  $0 = \sum_{i=1}^l \alpha_i (a_i + p_0)$ . By pointedness, we get  $\alpha_i (a_i + p_0) = 0$  for all  $i = 1, \dots, l$ . Hence,  $0 = a_j + p_0 \in A + \text{int } P$  for some  $j$ , which implies that  $\text{cone}_+(A + \text{int } P) = \mathbb{R}^m$ , contradicting (b).

(c)  $\implies$  (a) It follows from a simple use of a standard theorem on separation of convex sets, we get the existence of  $p \in \mathbb{R}^n \setminus \{0\}$  and  $\alpha \in \mathbb{R}$  such that

$$(11.5) \quad \langle p, z \rangle \geq \alpha, \forall z \in \text{co } A, \text{ and } \langle p, w \rangle \leq \alpha, \forall w \in -\text{cl}(\text{int } P) = -P.$$

From the first inequality of (11.5) it follows that  $\langle p, a \rangle \geq \alpha$  for all  $a \in A$ , and from the second inequality we get  $\alpha \geq 0$ . Hence  $p \in P^*$ , proving the desired result.  $\square$

We can go further when specializing  $A$ . We consider  $A$  to be the image of a subset  $C \subseteq \mathbb{R}^n$  through a linear transformation  $\mathcal{F} : \mathbb{R}^n \rightarrow \mathbb{R}^m$ .

**PROPOSITION 11.8.** *Let  $\mathcal{F}$  be a real matrix of order  $m \times n$ , and write  $\mathcal{F}^\top = (\mathcal{F}_1^\top, \dots, \mathcal{F}_m^\top)$ , where  $\mathcal{F}_i$  is the  $i$ -th row of  $\mathcal{F}$ . Let  $C \subseteq \mathbb{R}^n$  be any nonempty set. Then*

$$\mathcal{F}(C) \cap (-\text{int } \mathbb{R}_+^m) = \emptyset \iff \max_{1 \leq i \leq m} \langle \mathcal{F}_i^\top, v \rangle \geq 0 \forall v \in \overline{C},$$

and the following statements are equivalent:

- (a)  $\text{cone}(\mathcal{F}(C) + \text{int } \mathbb{R}_+^m)$  is pointed;
- (b)  $\mathcal{F}(\overline{\text{co}}(C)) \cap (-\text{int } \mathbb{R}_+^m) = \emptyset$ ;
- (c)  $\mathcal{F}(\text{co}(C)) \cap (-\text{int } \mathbb{R}_+^m) = \emptyset$ ;
- (d)  $\max_{1 \leq i \leq m} \langle \mathcal{F}_i^\top, v \rangle \geq 0 \forall v \in \overline{\text{co}}(C)$ ;
- (e)  $\text{co}(\{\mathcal{F}_i^\top : i = 1, \dots, m\}) \cap (C)^* \neq \emptyset$ .

**PROOF.** The first part is straightforward. By the previous theorem  $\text{cone}(\mathcal{F}(C) + \text{int } \mathbb{R}_+^m)$  is pointed  $\iff \text{co}(\mathcal{F}(C)) \cap (-\text{int } \mathbb{R}_+^m) = \emptyset$ .

It is easy to check that  $\text{co}(\mathcal{F}(C)) = \mathcal{F}(\text{co}(C))$  and

$$\begin{aligned} \mathcal{F}(\text{co}(C)) \cap (-\text{int } \mathbb{R}_+^m) = \emptyset &\iff \mathcal{F}(\overline{\text{co}}(C)) \cap (-\text{int } \mathbb{R}_+^m) = \emptyset \\ &\iff \overline{\mathcal{F}(\text{co}(C))} \cap (-\text{int } \mathbb{R}_+^m) = \emptyset. \end{aligned}$$

Both relations along with the fact that (a) of Theorem 11.7 amounts to writing

$$\text{co}(\{\mathcal{F}_i^\top : i = 1, \dots, m\}) \cap (C)^* \neq \emptyset,$$

we get all the remaining equivalences.  $\square$

## 11.2. Optimization with inequality constraints

Let us consider the minimization problem with inequality constraints:

$$(11.6) \quad \begin{cases} \min f(x) \\ g_i(x) \leq 0, \quad i = 1, \dots, m \\ x \in X, \end{cases}$$

where  $f, g_i : \mathbb{R}^n \rightarrow \mathbb{R}$ ,  $i = 1, \dots, m$ , are given functions. Define the feasible set to (11.6) as

$$K \doteq \{x \in X : g_i(x) \leq 0, \quad i = 1, \dots, m\}.$$

For fixed  $\bar{x} \in K$ , we associate its active indexes set,

$$(11.7) \quad I \doteq I(\bar{x}) \doteq \{i : g_i(\bar{x}) = 0\}.$$

Based on Proposition 11.8 we establish a new version, as an alternative-type result, of the Fritz John optimality conditions at points not necessarily optimal, which is new in the literature.

**THEOREM 11.9.** (Fritz John necessary optimality conditions of alternative-type) *Let us consider problem (11.6) and  $\bar{x} \in K$ , with  $X \subseteq \mathbb{R}^n$ . Let  $f, g_i$ ,  $i \in I$ , be differentiable at  $\bar{x}$ . Then, exactly one of the following two assertions holds:*

(a) *there exists  $v \in \mathbb{R}^n$  such that*

$$(11.8) \quad \begin{aligned} \langle \nabla f(\bar{x}), v \rangle &< 0, & v &\in \overline{\text{co}}[T(X; \bar{x})]; \\ \langle \nabla g_i(\bar{x}), v \rangle &< 0, & i &\in I, \end{aligned}$$

(b) *there exist  $\lambda_0 \geq 0$ ,  $\lambda_i \geq 0$ ,  $i \in I$ , not all zero, satisfying*

$$(11.9) \quad \lambda_0 \nabla f(\bar{x}) + \sum_{i \in I} \lambda_i \nabla g_i(\bar{x}) \in [T(X; \bar{x})]^*,$$

*or equivalently,  $\max_{i \in I} \{\langle \nabla f(\bar{x}), v \rangle, \langle \nabla g_i(\bar{x}), v \rangle\} \geq 0$ ,  $\forall v \in T(X; \bar{x})$ .*

Furthermore, if each  $g_i$  is differentiable at  $\bar{x}$ , condition (11.9) can be written as

$$(11.10) \quad \begin{cases} \lambda_0 \nabla f(\bar{x}) + \sum_{i=1}^m \lambda_i \nabla g_i(\bar{x}) \in [T(X; \bar{x})]^*; \\ \lambda_i g_i(\bar{x}) = 0, \quad i = 1, \dots, m. \end{cases}$$

**PROOF.** This is a direct application of Proposition 11.8 to

$$\mathcal{F} \doteq \begin{pmatrix} \nabla f(\bar{x})^\top \\ \nabla g_I(\bar{x})^\top \end{pmatrix},$$

where  $\nabla g_I(\bar{x})^\top$  is the matrix having as rows  $\nabla g_i(\bar{x})^\top$  for  $i \in I$ . □

The first relation in (11.10) is called the Fritz John optimality conditions and the real numbers  $\lambda_0, \lambda_1, \dots, \lambda_m$ , are named the Lagrange multipliers, while the second relation are termed the complementarity conditions.

**EXERCISE 11.6.** Provide a nonconvex set  $X$  in such a way that  $T(X; \bar{x})$  is convex.

In connection to the previous theorem, we will see that a stronger system than (11.8) is implied by optimality: To be more precise, besides differentiability of  $g_i$  at  $\bar{x}$ , for  $i \in I$ ; continuity of  $g_i$  at  $\bar{x}$  for  $i \notin I$ , local minimality of  $\bar{x}$  imply that the system

$$(11.11) \quad \begin{aligned} \langle \nabla f(\bar{x}), v \rangle &< 0, & v &\in T(X; \bar{x}); \\ \langle \nabla g_i(\bar{x}), v \rangle &< 0, & i &\in I, \end{aligned}$$

has no solution. Thus, if  $T(X; \bar{x})$  is convex, we have the impossibility of system (11.8), and so by Theorem 11.9, (b) holds. This will be expressed in Corollary 11.11, whose formulation is very versatile as we shall show

in Section 11.3, and it encompasses many recent results appearing in the literature. Thus, the interesting case is when  $T(X; \bar{x})$  is nonconvex.

We start by proving a result about the impossibility of system (11.11) under local optimality.

**PROPOSITION 11.10.** *Let us consider problem (11.6) and  $\bar{x} \in K$ . In addition, let  $X \subseteq \mathbb{R}^n$  be any nonempty set;  $f, g_i, i \in I$ , be differentiable at  $\bar{x}$ ;  $g_i, i \notin I$ , be continuous at  $\bar{x}$ . If  $\bar{x}$  is a local solution to (11.6), then system (11.11) has no solution.*

**PROOF.** If on the contrary,  $v$  is a solution to (11.11) there exist sequences  $\lambda_k > 0$ ,  $x_k \in X$ ,  $x_k \rightarrow \bar{x}$ , satisfying  $\lambda_k(x_k - \bar{x}) \rightarrow v$ . By Taylor series,

$$f(x_k) - f(\bar{x}) = \langle \nabla f(\bar{x}), x_k - \bar{x} \rangle + \|x_k - \bar{x}\| o(\|x_k - \bar{x}\|)$$

with  $o(t) \rightarrow 0$  as  $t \rightarrow 0$ . On multiplying this equality by  $\lambda_k$ , letting  $k \rightarrow +\infty$  and using the first inequality of (11.11), we get the existence of  $k_1$  such that

$$(11.12) \quad f(x_k) < f(\bar{x}), \quad \forall k \geq k_1.$$

It only remains to check that  $x_k$  is feasible for all  $k$  sufficiently large to reach a contradiction.

Let  $i \in I$ . We get similarly as for  $f$

$$g_i(x_k) - g_i(\bar{x}) = \langle \nabla g_i(\bar{x}), x_k - \bar{x} \rangle + \|x_k - \bar{x}\| o(\|x_k - \bar{x}\|)$$

with  $o(t) \rightarrow 0$  as  $t \rightarrow 0$ . On multiplying by  $\lambda_k$  this equality and letting  $k \rightarrow +\infty$ , we obtain, for some  $k_2$ ,

$$(11.13) \quad g_i(x_k) < 0, \quad \forall i \in I, \quad \forall k \geq k_2.$$

Since  $g_i$  is continuous for  $i \notin I$ , there exists  $k_3$  such that

$$(11.14) \quad g_i(x_k) < 0, \quad \forall i \notin I, \quad \forall k \geq k_3.$$

On combining (11.13) and (11.14), we conclude that  $x_k$  is feasible for  $k$  sufficiently large. Hence  $\bar{x}$  cannot be a local solution to (11.6), and the proof is completed.  $\square$

We are now ready to establish the result which is suitable for our purposes.

**COROLLARY 11.11.** (*Fritz John necessary optimality conditions*) *Let us consider problem (11.6) and  $\bar{x} \in K$ . In addition, let  $X \subseteq \mathbb{R}^n$  be a set such that  $T(X; \bar{x})$  is convex;  $f, g_i, i \in I$ , be differentiable at  $\bar{x}$ ;  $g_i, i \notin I$ , be continuous at  $\bar{x}$ . If  $\bar{x}$  is a local solution to (11.6), then there exist  $\lambda_0 \geq 0$ ,  $\lambda_i \geq 0, i \in I$ , not all null at the same time, satisfying*

$$\lambda_0 \nabla f(\bar{x}) + \sum_{i \in I} \lambda_i \nabla g_i(\bar{x}) \in [T(X; \bar{x})]^*.$$

**PROOF.** By the previous proposition, system (11.11) has no solution, but such a system is exactly that expressed in (a) of Theorem 11.9. Therefore (b) of the same theorem holds, and the corollary follows.  $\square$

### 11.3. Some Particular Situations

Let us show some interesting specializations appearing in applications where  $T(X; \bar{x})$  is convex.

#### 11.3.1. The set $X$ is not convex with $T(X; \bar{x})$ being convex.

Let us consider the problem with an additional quadratic equality constraint:

$$(11.15) \quad \min f(x) \text{ s. t. } g_i(x) \leq 0, \ i = 1, \dots, m; \ q(x) = 0; \ x \in \mathbb{R}^n,$$

where  $q$  is a function of the form

$$q(x) \doteq \frac{1}{2} x^\top A x + a^\top x + \alpha,$$

with  $A$  being a (real) symmetric matrix,  $a \in \mathbb{R}^n$ ,  $\alpha \in \mathbb{R}$ . Clearly the set

$$X \doteq \{x \in \mathbb{R}^n : q(x) = 0\}$$

is not necessarily convex even if  $q$  is convex. Let  $\bar{x}$  feasible for Problem (11.15). It is not difficult to find that (see for instance [26, Theorem 2.1])

$$T(X; \bar{x}) = \left\{ v \in \mathbb{R}^n : (A\bar{x} + a)^\top v = 0 \right\} \text{ if } A\bar{x} + a \neq 0;$$

whereas

$$T(X; \bar{x}) = \left\{ v \in \mathbb{R}^n : v^\top A v = 0 \right\} \text{ if } A\bar{x} + a = 0.$$

This set, in general, is nonconvex. If additionally,  $q$  is convex, that is,  $A$  is positive semidefinite, a more precise formulation may be obtained since  $v^\top A v = 0 \iff A v = 0$ :

$$(11.16) \quad T(X; \bar{x}) = \begin{cases} (A\bar{x} + a)^\perp & \text{if } A\bar{x} + a \neq 0; \\ \ker A & \text{if } A\bar{x} + a = 0. \end{cases}$$

Thus,

$$[T(X; \bar{x})]^* = \begin{cases} \mathbb{R}(A\bar{x} + a) & \text{if } A\bar{x} + a \neq 0; \\ (\ker A)^\perp = A(\mathbb{R}^n) & \text{if } A\bar{x} + a = 0. \end{cases}$$

Hence, the Fritz John conditions (11.11) reduces to the existence of  $\lambda_0$ ,  $\lambda_i$ ,  $i \in I$ , not all zero,  $\lambda \in \mathbb{R}$ ,  $y \in \mathbb{R}^n$ , such that

$$\lambda_0 \nabla f(\bar{x}) + \sum_{i \in I} \lambda_i \nabla g_i(\bar{x}) = \begin{cases} \lambda(A\bar{x} + a) & \text{if } A\bar{x} + a \neq 0; \\ A y & \text{if } A\bar{x} + a = 0. \end{cases}$$

**11.3.2. The set  $X$  is convex.** In this case,  $T(X; \bar{x})$  is convex, and since  $T(X; \bar{x}) = \bigcup_{t \geq 0} t(X - \bar{x})$ , we get  $[T(X; \bar{x})]^* = -N(X; \bar{x})$ , and so (11.11) can be written as

$$\lambda_0 \nabla f(\bar{x}) + \sum_{i \in I} \lambda_i \nabla g_i(\bar{x}) \in -N(X; \bar{x}).$$

Thus, Theorem 3.2.2 in Bector et al. (2005) is obtained.

**11.3.3. The set  $X$  is open or  $\bar{x} \in \text{int } X$ .** In this situation,  $T(X; \bar{x}) = \mathbb{R}^n$ , and therefore condition (11.11) reduces to

$$\lambda_0 \nabla f(\bar{x}) + \sum_{i \in I} \lambda_i \nabla g_i(\bar{x}) = 0.$$

This is nothing else than Theorem 4.2.8 in Bazaraa et al. (2006), and Theorem 3.2.1 in Bector et al. (2005) when  $X = \mathbb{R}^n$ .

**11.3.4. The set  $X$  is an affine subspace.** This case deals with  $X = \{x \in \mathbb{R}^n : Hx = d\} = \bar{x} + \ker H$ , with  $H$  being a real  $p \times n$  matrix and  $\bar{x} \in X$ . Thus, we obtain  $T(X; \bar{x}) = \ker H$  and  $[T(X; \bar{x})]^* = (\ker H)^* = (\ker H)^\perp$ . Hence (11.11) is expressed as

$$\lambda_0 \nabla f(\bar{x}) + \sum_{i \in I} \lambda_i \nabla g_i(\bar{x}) \in (\ker H)^\perp = H^\top(\mathbb{R}^p),$$

that is, there exists  $y_i \in \mathbb{R}$ ,  $i = 1, \dots, p$ , such that

$$\lambda_0 \nabla f(\bar{x}) + \sum_{i \in I} \lambda_i \nabla g_i(\bar{x}) + H^\top y = 0.$$

**11.3.5. The set  $X$  is polyhedral.** Let us consider (see for instance Birbil et al., 2007)

$$(11.17) \quad \begin{cases} \min f(x) \\ g_i(x) \leq 0, \quad i = 1, \dots, m, \\ h_j^\top x \leq d_j, \quad j = 1, \dots, p, \\ x \in \mathbb{R}^n. \end{cases}$$

In this case, we can take  $X = \{x \in \mathbb{R}^n : Hx \leq d\}$  with  $H$  being a  $p \times n$  matrix, and refine (11.11). More precisely, by denoting  $h_j^\top$  to be the rows of  $H$  and setting  $J \doteq \{j : h_j^\top \bar{x} = d_j\}$ , the conclusion of Corollary 11.11 reduces to the existence of  $\lambda_0 \geq 0$ ,  $\lambda_i \geq 0$ ,  $i \in I$ , not all zero, and  $u_j \geq 0$ ,  $j \in J$ , such that

$$(11.18) \quad \lambda_0 \nabla f(\bar{x}) + \sum_{i \in I} \lambda_i \nabla g_i(\bar{x}) + \sum_{j \in J} u_j h_j = 0.$$

Indeed, by setting  $C \doteq \{v \in \mathbb{R}^n : h_j^\top v \leq 0, \quad j \in J\}$ , which is a closed convex cone, one can easily check that  $T(X; \bar{x}) = C$  (see Lemma 5.1.4 in Bazaraa et al., 2006). By applying Corollary 11.11, we get the existence of  $\lambda_0 \geq 0$ ,  $\lambda_i \geq 0$ ,  $i \in I$ , not all zero, such that

$$\lambda_0 \nabla f(\bar{x}) + \sum_{i \in I} \lambda_i \nabla g_i(\bar{x}) \in C^*.$$

The conclusion follows once we notice that  $C^* = \{H_J u : u \leq 0\}$ , with  $H_J$  is the matrix with columns  $h_j$  for  $j \in J$ . Thus we recovered the Fritz-John conditions as appears in Birbil et al. (2007).

We now discuss an example where  $T(X; \bar{x})$  is nonconvex. Our theorem will be applicable to such an example, and no result existing in the literature is.