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# On the convex programming problem

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## Abstract

The nondegenerate Lagrange multipliers rule is proved under an explicit constraint qualification strictly weaker than the Slater condition. A first approach is based on direct exact penalization arguments, while a second one uses the theory of maximal monotone operators and allows a large class of applications.

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## 1. Introduction

Let  $H$  be a Banach space,  $g : H \rightarrow ]-\infty, +\infty]$  be a convex, lower semicontinuous proper function and  $h_i : H \rightarrow ]-\infty, +\infty]$ ,  $i = \overline{1, n}$ , be convex mappings. Precise hypotheses will be stated in the sequel, for each result. We consider the standard convex programming problem with inequality constraints:

$$(1.1) \quad \text{Minimize } \{g(x)\}$$

subject to

$$(1.2) \quad h_i(x) \leq 0, \quad i = \overline{1, n}.$$

We define the feasible (convex) set

$$(1.3) \quad C = \{x \in H; h_i(x) \leq 0, \quad i = \overline{1, n}\}$$

assumed to be nonvoid (admissibility) and we denote by  $\bar{x} \in C$  a solution of (1.1), (1.2), supposed to exist. We also introduce the convex mapping  $h : H \rightarrow ]-\infty, +\infty]$ ,  $h(x) = \max\{h_i(x); i = \overline{1, n}\}$  and the problem (1.1), (1.2) may be equivalently reformulated as (1.1) and

$$(1.4) \quad h(x) \leq 0.$$

This work discusses the classical question of the Lagrange multipliers associated to  $\bar{x}$  and of the related constraint qualification conditions if a "nondegenerate" (involving  $g$  essentially) characterization is obtained.

In the next section, we show by a direct argument a geometric property of convex functions (valid on  $H \setminus C$ ) which is a consequence of the Slater [11] condition. This allows to obtain a modified nondegenerate Lagrange multipliers rule involving  $h_+$ , the positive part of  $h$ .

A new explicit constraint qualification, strictly weaker than the Slater condition, is formulated and the standard Karush-Kuhn-Tucker [7], [8] conditions are reobtained (under this assumption) via subdifferential calculus.

In the last section the continuity hypotheses on  $g$  are relaxed by means of an alternative approach based on the Minty [9] theorem on maximal monotone operators in Banach spaces, Barbu [2].

Finally, we underline that the literature on the relaxation of constraint qualifications in the mathematical programming or of interiority conditions in optimal control is quite rich: Zowe and Kurcyusz [17], Clarke [5, Ch. VI], Troltzsch [16], Barbu and Pavel [3], Tiba [12], Neittaanmaki and Tiba [13, Ch. VI], Tiba and Bergounioux [14], Vinter and Ferreira [18]. Generally, qualification conditions on the feasible set are required, while our hypothesis is related to the behaviour of  $h$  on the complementary of  $C$ .

## 2. Slater condition and exact penalization

**Theorem 2.1.** *Let  $h$  be convex proper lower semicontinuous and let the Slater assumption*

$$(2.1) \quad \exists \hat{x} \in C : h(\hat{x}) < 0$$

*be fulfilled. Then,  $\forall r > 0, \forall x \in B(\hat{x}, r) \setminus C$ , we have*

$$(2.2) \quad h(x) \geq -\frac{h(\hat{x})}{r} \text{dist}(x, C).$$

*Proof.*

Take  $a = -h(\hat{x}) > 0$  and denote by  $S$  the line passing through  $\hat{x}, x$  in  $H$ . If  $h(x) = +\infty$ , then (2.2) is trivial. Therefore, we may assume that  $h$  is finite on the closed segment  $[\hat{x}x] \subset S$  since it is convex, proper. Then  $h|_S$  is continuous on the open segment  $] \hat{x}x[ \subset S$ . As  $x \in H \setminus C$  and  $C$  is closed, there is  $\varepsilon > 0$  such that  $h > 0$  on  $B(x, \varepsilon) \cap ] \hat{x}x[$  due to (1.3). Since  $h$  is convex, it follows that  $\lim_{y \rightarrow x^-} h(y)|_S > 0$ .

By, Proposition 3.1.2., Hiriart-Urruty and Lemarechal [6, Ch. I], we have that

$$\lim_{y \rightarrow \hat{x}^+} h(y)|_S \leq h(\hat{x}) < 0.$$

We suppose that a unit vector  $u, |u|_H = 1$  is chosen parallel to  $S$  and a parametrization of  $S$  with respect to  $u$  and some origin is given. The above discussion shows the existence of  $\tilde{x} \in ] \hat{x}x[$  such that  $h(\tilde{x}) = 0, \tilde{x} \in C$ . Let  $\hat{\lambda}, \tilde{\lambda}, \lambda$  be the "coordinates" of  $\hat{x}, \tilde{x}, x$  on  $S$ , respectively and assume that  $\hat{\lambda} < \tilde{\lambda} < \lambda$ .

For  $y \in S, y \cong \mu u, \mu \in R$ , we define

$$(2.3) \quad f(y) = \frac{a}{r}(\mu - \tilde{\lambda})$$

$f$  being an affine mapping on  $S$ . We notice by (2.3) that

$$(2.4) \quad f(\tilde{x}) = 0 = h(\tilde{x}),$$

$$(2.5) \quad f(\hat{x}) = -\frac{h(\hat{x})}{r}(\hat{\lambda} - \tilde{\lambda}) = h(\hat{x}) \frac{\tilde{\lambda} - \hat{\lambda}}{r} \geq h(\hat{x})$$

since  $h(\hat{x}) < 0$  and  $0 \leq \frac{\tilde{\lambda} - \hat{\lambda}}{r} = \frac{|\tilde{x} - \hat{x}|_H}{r} \leq 1$ . Due to the convexity of  $h|_S$  and the affine character of  $f$ , we get by (2.4), (2.5) that

$$\begin{aligned} h(x) &\geq f(x) = \frac{a}{r}(\lambda - \tilde{\lambda}) = \frac{a}{r} |x - \tilde{x}|_H \geq \\ &\geq \frac{a}{r} |x - \text{proj}_C x|_H = -\frac{h(\hat{x})}{r} \text{dist}(x, C) \end{aligned}$$

since  $\bar{x} \in C$ . This ends the proof.

**Remark.** Suggested by Theorem 2.1, we formulate the following explicit constraint qualification:

$$(2.6) \quad \forall M \subset H \text{ bounded} : M \setminus C \neq \emptyset,$$

$$\exists c_M > 0 : h(x) \geq c_M |x - \text{proj}_C x|_H, \forall x \in M \setminus C,$$

which is strictly weaker than (2.1) by the above result and by the example of  $h$  satisfying (2.1) and its positive part  $h_+$  for which (2.1) fails but not (2.6).

If  $C$  is bounded, then  $c_M$  may be chosen independent of  $M$ . Moreover, in (2.6) a neighbourhood of  $C$  may be taken into account instead of  $H$ , by the convexity of  $h$ .

**Remark.** By (2.6) and (1.3), we have

$$(2.6)' \quad h(x) - h(\text{proj}_C x) \geq c_M |x - \text{proj}_C x|_H$$

for  $x \in M \setminus C$ . Relation (2.6)' expresses that the subgradients of  $h$  in  $x$ ,  $\partial h(x)$ , are "far" from zero. This will play an essential role in the next section.

**Theorem 2.2.** *Let  $h : H \rightarrow ]-\infty, +\infty]$  be convex lower semicontinuous proper and  $g : H \rightarrow \mathbb{R}$  be convex continuous. Then, if  $\bar{x}$  is a solution of (1.1), (1.4) and (2.6) is satisfied, there is  $\lambda \geq 0$  such that  $\bar{x}$  is a minimum point of  $g + \lambda h_+$  over  $H$ .*

*Proof.*

Let  $B(\bar{x}, \varepsilon)$  be a "small" ball around  $\bar{x}$ . We show the minimum property of  $g + \lambda h_+$  (for some  $\lambda \geq 0$ ) on  $B(\bar{x}, \varepsilon)$  and it will follow on  $H$ , by convexity. For any  $y \in B(\bar{x}, \varepsilon)$ ,  $\text{proj}_C y \in B(\bar{x}, \varepsilon)$  since  $\bar{x} = \text{proj}_C \bar{x}$  and  $\text{proj}_C(\cdot)$  is nonexpansive. We have

$$g(y) - g(\text{proj}_C y) \geq (w, y - \text{proj}_C y)_{H^* \times H} \geq -L |y - \text{proj}_C y|_H$$

for any  $w \in \partial g(\text{proj}_C y)$  and for  $L \geq \|w\|_{H^*}$  given by the boundedness of  $\partial g(\cdot)$  on  $B(\bar{x}, \varepsilon)$  due to the continuity of  $g$ . Then

$$(2.7) \quad g(y) + L |y - \text{proj}_C y|_H \geq g(\text{proj}_C y) \geq g(\bar{x}).$$

By (2.6), we have

$$(2.8) \quad h_+(x) \geq c_\varepsilon |x - \text{proj}_C x|_H, \forall x \in B(\bar{x}, \varepsilon) \setminus C.$$

Then, (2.7), (2.8) yield, for  $y \in B(\bar{x}, \varepsilon) \setminus C$ :

$$(2.9) \quad \begin{aligned} g(y) + \frac{L}{c_\varepsilon} h_+(y) &\geq g(y) + L |y - \text{proj}_C y|_H \geq \\ &\geq g(\bar{x}) = g(\bar{x}) + \frac{L}{c_\varepsilon} h_+(\bar{x}). \end{aligned}$$

Relation (2.9) remains valid for  $y \in B(\bar{x}, \varepsilon) \cap C$  since  $h_+|_C = 0$  and this finishes the proof with  $\lambda = \frac{L}{c_\varepsilon} \geq 0$ .

**Remark.** The above proof is based on direct exact penalization arguments as in Lemarechal and Hiriart-Urruty [6, Ch. VII 1.2]. We remark that only  $g$  locally Lipschitzian is necessary in this setting. On the contrary, based on subdifferential calculus rules (which are known to be equivalent with the separation theorem, Tichomirov [15], p. 52), we shall get the standard Karush-Kuhn-Tucker optimality conditions (Theorem 2.3 below).

*Example.* For  $H = R$ , if (2.6) fails, then the conclusion of Theorem 2.2 may be not valid. We take  $\tilde{g}(x) = x$ ,  $x \in R$  and

$$\tilde{h}(x) = \begin{cases} x^2, & x < 0 \\ 0, & x \geq 0 \end{cases}$$

and we consider the minimization of  $\tilde{g}$  subject to  $\tilde{h}(x) \leq 0$  (i.e.  $x \geq 0$ ). The solution is  $\bar{x} = 0$  and (2.6) is not fulfilled since  $\tilde{h}' \equiv 0$  for  $x \in C$ . For any  $\lambda \geq 0$ , we define the nondegenerate Lagrange function  $g_\lambda(x) = \tilde{g}(x) + \lambda \tilde{h}_+(x) = \tilde{g}(x) + \lambda \tilde{h}(x)$ . Obviously,  $x_\lambda = -\frac{1}{2\lambda}$  provides a global minimum for  $g_\lambda$  and  $x_\lambda \neq \bar{x}$ ,  $\forall \lambda > 0$ . For  $\lambda = 0$ , the infimum on  $R$  is not attained.

**Theorem 2.3.** Assume (2.6) and that  $g$  and  $h_i$ ,  $i = \overline{1, n}$ , are convex continuous on  $H$ . If  $\bar{x}$  is a solution of (1.1), (1.2) there are  $\lambda_i \geq 0$ ,  $i = \overline{1, n}$ , such that

$$(2.10) \quad 0 \in \partial g(\bar{x}) + \sum_{i=1}^n \lambda_i \partial h_i(\bar{x}),$$

$$(2.11) \quad \lambda_i h_i(\bar{x}) = 0, \quad i = \overline{1, n}$$

*Proof.*

This is a consequence of Theorem 2.2 and of the Dubovitskij-Milyutin theorem which is valid under the continuity assumptions, Tichomirov [15], p.52:

$$\partial h(x) = \left\{ \sum_{j \in T(x)} \lambda_j w_j; \lambda_j \in R_+, \sum_{j \in T(x)} \lambda_j = 1, w_j \in \partial h_j(x) \right\},$$

$$T(x) = \{j; h_j(x) = h(x)\}.$$

We also use that  $h_+ = \sup(0, h)$  and the additivity rule  $\partial(g + \lambda h_+) = \partial g + \lambda \partial h_+$  to finish the proof.

*Example.* As a byproduct of Theorem 2.1, we put into evidence a class of functionals on  $H$  which have the exact penalization property for convex closed sets  $C \subset H$  such that  $0 \in C$ . We denote by  $p_c : H \rightarrow ]-\infty, +\infty]$  the Minkowski (gauge) functional associated to  $C$  and we define, for  $\varepsilon > 0$ :

$$(2.10) \quad h^\varepsilon(x) = \varepsilon(p_c(x) - 1)_+.$$



The following properties are obvious:

$$\begin{aligned} C &= \{x \in H; h^\varepsilon(x) = 0\}, \quad \forall \varepsilon > 0, \\ h^\varepsilon(x) &\rightarrow +\infty, \text{ for } \varepsilon \rightarrow \infty, \quad \forall x \in H \setminus C, \\ h^\varepsilon(x) &\rightarrow 0 \text{ for } \varepsilon \rightarrow 0, \quad \forall x \in \text{aff}(C). \end{aligned}$$

Moreover, if  $\tilde{h}^\varepsilon(x) = \varepsilon(p_c(x) - 1)$ , then it satisfies the Slater condition  $\tilde{h}^\varepsilon(0) = -\varepsilon$ ,  $\forall \varepsilon > 0$  and Theorem 2.1 shows that  $h^\varepsilon$  satisfies (2.6). If  $g$  is a continuous convex mapping with a minimum in  $\bar{x} \in C$ , then (by the proof of Theorem 2.2) there is  $\varepsilon > 0$  such that  $\bar{x}$  is a minimum for  $g + h^\varepsilon$  on  $H$ .

It is well known that the distance function,  $\text{dist}(x, C)$ , has the exact penalization property, Hiriart-Urruty and Lemarechal [6, Ch. VII, 1.2], while (2.10) is an example of a different nature.

### 3. Maximal monotonicity

In this section we assume that  $H$  is a reflexive Banach space. Then, an equivalent norm may be defined such that  $H$  and the dual  $H^*$  are strict convex, Asplund [1].

By using a different technique, we prove the nondegenerate Lagrange multipliers rule under condition (2.6) and with general convex lower semicontinuous proper mapping  $g : H \rightarrow ]-\infty, +\infty]$ . This is important since it allows to add to the problem (1.1), (1.2) some abstract type constraints  $x \in A \subset H$  convex, closed nonvoid subset, by the standard trick of redefining  $g$  as  $+\infty$  outside  $A$ .

**Theorem 3.1.** *Let  $h : H \rightarrow R$  be convex continuous satisfying (2.6) and  $0 \in \partial h(0)$ . Then, the operator  $N \subset H \times H^*$  given by*

$$(3.1) \quad N(x) = \begin{cases} 0, & \text{if } h(x) < 0 \\ \lambda w, \lambda \geq 0, w \in \partial h(x), & \text{if } h(x) = 0 \\ \emptyset & \text{if } h(x) > 0 \end{cases}$$

*is maximal monotone and  $N(x) = \partial I_C(x)$ , the normal cone to  $C$  at  $x$ .*

In (3.1), it is possible that no  $x \in H$  satisfies  $h(x) < 0$  and, then, the first line disappears.

*Proof*

By the definition of the subdifferential, we have  $N(x) \subset N_C(x)$ ,  $\forall x \in H$ . This is obvious if  $x \notin C$  or  $h(x) < 0$ . If  $x \in C$  and  $h(x) = 0$ , we have

$$h(x) - h(y) \leq (w, x - y)_{H \times H^*}, \quad \forall x \in \partial h(x), \quad \forall y \in H$$

that is

$$(3.2) \quad 0 \leq \lambda(w, x - y)_{H \times H^*}, \quad \forall \lambda \geq 0, \quad \forall y \in C$$

since  $h(x) - h(y) = -h(y) \geq 0$ . Relation (3.2) proves the inclusion. It yields that  $N \subset H \times H^*$  is monotone and we have to show its maximality to get the desired



equality. Since  $\partial h \subset H \times H^*$  is maximal monotone, the Minty theorem, Barbu [2, Ch. II] gives the existence of  $x_\lambda \in H$  such that

$$(3.3) \quad F(x_\lambda - y) + \lambda \partial h(x_\lambda) \ni 0$$

where  $\lambda > 0$ ,  $F: H \rightarrow H^*$  is the duality mapping and  $y \in H$  is arbitrary.

We show that  $x_\lambda \in C$  for  $\lambda$  "sufficiently big". If  $y \in C$ , we may take  $x_0 = x_\lambda = y$  and  $\lambda = 0$ . Then (3.3) becomes (by using (3.1)):

$$(3.4) \quad F(x_0 - y) + N(x_0) \ni 0.$$

Therefore, we may assume  $h(y) > 0$ . By the definition and properties of the Moreau-Yosida regularization, Barbu and Precupanu [4, Ch. 2.3], we have

$$(3.5) \quad \begin{aligned} h(x_\lambda) &= h_\lambda(y) - \frac{\lambda}{2} |\partial h_x(y)|_{H^*}^2 = \\ &= h_\lambda(y) - \frac{\lambda}{2} |w_\lambda|_{H^*}^2 \end{aligned}$$

where  $w_\lambda$  is some element in  $\partial h(x_\lambda)$ .

If  $x_\lambda \notin C$  for some  $\lambda > 0$ , then  $h(x_\lambda) > 0$  and (3.5) gives

$$(3.6) \quad |w_\lambda|_{H^*}^2 < \frac{2h_\lambda(y)}{\lambda} \leq \frac{2h(y)}{\lambda}.$$

Taking into account hypothesis (2.6), we get:

$$(3.7) \quad \begin{aligned} c_* |x_\lambda - \text{proj}_C x_\lambda|_H &\leq h(x_\lambda) - h(\text{proj}_C x_\lambda) \leq \\ &\leq (\partial h(x_\lambda), x_\lambda - \text{proj}_C x_\lambda)_{H \times H^*} \leq |\partial h(x_\lambda)|_{H^*} \cdot |x_\lambda - \text{proj}_C x_\lambda|_H, \end{aligned}$$

where  $c_* > 0$  is a constant depending on  $y^*$  since  $|x_\lambda|_H \leq |y^*|_{H^*}$  by  $0 \in \partial h(0)$ . Combining (3.6), (3.7) we get:

$$c_*^2 < \frac{2h(y)}{\lambda}$$

which is a contradiction for  $\lambda$  big enough.

By this discussion and since  $N(x)$  is a cone, we conclude that for any  $y \in H$ , the equation

$$(3.8) \quad F(x_* - y) + N(x_*) \ni 0$$

has a solution  $x_* \in C$ .

Assume now that  $N \subset H \times H^*$  is not maximal monotone. This means that there is  $x' \in H$ ,  $y' \in H^*$  such that  $y' \notin N(x')$  and

$$(3.9) \quad (x - x', y - y')_{H \times H^*} \geq 0$$

for any  $x \in C$ , any  $y \in N(x)$ .

By (3.8) there is  $x^0 \in C$ , solution for the equation

$$(3.10) \quad F(x^0 - x' - F^{-1}(y')) + y^0 = 0$$

with some  $y^0 \in N(x^0)$ . Choosing  $x = x^0$ ,  $y = y^0$  in (3.9), we get

$$(x^0 - x', -F(x^0 - x' - F^{-1}(y')) - y')_{H \times H^*} \geq 0$$

that is

$$0 \geq (x^0 - x' - F^{-1}(y') - (-F^{-1}(y')), F(x^0 - x' - F^{-1}(y')) - F(-F^{-1}(y')))_{H \times H^*}.$$

Since  $F$  is strictly monotone in  $H \times H'$ , we get  $x^0 = x'$  and  $y^0 = y'$  by (3.10) and the proof is finished.

**Remark.** The inclusion  $N \subset N_C$  is well known and the equality  $N = N_C$  is an abstract regularity condition, necessary and sufficient for the nondegenerate Lagrange multipliers rule to hold (Hiriart-Urruty and Lemarechal [6, Ch.VII.2]). Another "basic constraint qualification" may be formulated via tangent cones as well, Rockafellar [10].

**Corollary 3.2.** Let  $g : H \rightarrow ]-\infty, +\infty]$  be convex lower semicontinuous proper, continuous in some point of  $C$  and  $\bar{x}$  be a solution of (1.1), (1.2). If  $h$  satisfies the assumptions of Theorem 3.1, then there are  $\lambda_i \geq 0$  such that

$$0 \in \partial g(\bar{x}) + \sum_{i=1}^n \lambda_i \partial h_i(\bar{x}), \quad \lambda_i h_i(\bar{x}) = 0, \quad i = \overline{1, n}.$$

*Proof*

Since  $g$  is continuous in a point of  $C$ , we have the additivity rule  $0 \in \partial g(\bar{x}) + N_C(\bar{x})$ . The proof is finished by Theorem 3.1 and the Dubovitskij-Milyutin theorem.

**Remark.** The hypothesis  $0 \in \partial h(0)$  is not restrictive since if the Slater condition does not hold, then  $\forall x \in C$  is a minimum point for  $h$  by  $C = \{x \in H; h(x) = 0\}$  and  $h \geq 0$  on  $H$ . Therefore a simple shifting on  $h$  gives  $0 \in \partial h(0)$ . If the Slater condition is fulfilled the results are wellknown.

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