

# CONVEX OPTIMIZATION VIA LINEARIZATION

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## Notation

$X$  denotes a l.c. Hausdorff t.v.s and  $X^*$  its topological dual endowed with the weak\*-topology.

$$\mathbb{R}_+^{(T)} = \{\lambda : T \rightarrow \mathbb{R}_+ \mid |\text{supp } \lambda| < +\infty\}, \text{ with } \text{supp } \lambda = \{t \in T \mid \lambda_t \neq 0\}.$$

The asymptotic (or recession) cone of  $C \subset X$  is

$$\begin{aligned} C^\infty &= \{z \in X \mid C + z \subset C\} \\ &= \left\{ z \in X \mid \begin{array}{l} \exists c \in C \text{ such that} \\ c + \lambda z \in C \quad \forall \lambda \geq 0 \end{array} \right\} \\ &= \{z \in X \mid c + \lambda z \in C \quad \forall c \in C \text{ and } \forall \lambda \geq 0\} \end{aligned}$$

For a set  $D \subset X$ , the **normal cone** of  $D$  at  $x$  is

$$N_D(x) = \begin{cases} \{u \in X^* \mid u(y - x) \leq 0 \text{ for all } y \in D\}, & \text{if } x \in D, \\ N_D(x) = \emptyset, & \text{if } x \notin D. \end{cases}$$

The effective domain, the graph, and the epigraph of  $h : X \rightarrow \mathbb{R} \cup \{+\infty\}$  are denoted by

$\text{dom}h$ ,  $\text{gph}h$  and  $\text{epi}h$

The **subdifferential** of  $h$  at a point  $x \in \text{dom}h$  is

$$\partial h(x) = \{u \in X^* \mid h(y) \geq h(x) + u(y - x) \quad \forall y \in X\}$$

The **conjugate** of  $h$  is

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

$h^*$  is also a proper l.s.c. convex function and its conjugate (**biconjugate** of  $h$ ) is  $h^{**} = h$ .

In particular, if  $f(x) = a'x + b$ , then

$$f^*(u) = \sup_{x \in \mathbb{R}^n} \{(u - a)'x - b\} = \delta_{\{a\}}(u) - b,$$

i.e.,  $f^* = \delta_{\{a\}} - b$ .

The *asymptotic function* of  $h$  is  $h^\infty$  such that

$$\text{epi } h^\infty = (\text{epi } h)^\infty.$$

The *indicator function* of  $D \subset X$  is

$$\delta_D(x) = \begin{cases} 0, & \text{if } x \in D \\ +\infty, & \text{if } x \notin D. \end{cases}$$

If  $D \neq \emptyset$  is closed and convex, then  $\delta_D$  is a proper l.s.c. convex function.

The *support function* of  $D$  is

$$\delta_D^*(u) = \delta_{\text{cl}(\text{conv}D)}^*(u) = \sup_{x \in D} u(x), \quad u \in X^*.$$

In particular,

$$\delta_{\mathbb{R}^n}^*(u) = \sup_{x \in \mathbb{R}^n} u'x = \delta_{\{0_n\}}(u) \Rightarrow \delta_{\mathbb{R}^n}^* = \delta_{\{0_n\}}.$$

# LINEARIZING CONVEX SYSTEMS

We consider

$$\sigma := \{f_t(x) \leq 0, t \in T; x \in C\},$$

where

- ◆  $T$  is an arbitrary (possibly infinite) index set,
- ◆  $C$  is a nonempty closed convex subset of  $X$ , and
- ◆  $f_t : X \rightarrow \mathbb{R} \cup \{+\infty\}$  is a proper l.s.c. convex function,  $\forall t \in T$ .

In many applications  $C = X$ , in which case we write

$$\sigma := \{f_t(x) \leq 0, t \in T\}$$

Given  $t \in T$ ,

$$f_t(x) \leq 0$$

$$\iff f_t^{**}(x) \leq 0$$

$$\iff u_t(x) - f_t^*(u_t) \leq 0, \quad \forall u_t \in \text{dom } f_t^*$$

$$\iff u_t(x) \leq f_t^*(u_t), \quad \forall u_t \in \text{dom } f_t^*$$

$$\iff u_t(x) \leq f_t^*(u_t) + \alpha, \\ \forall u_t \in \text{dom } f_t^*, \quad \forall \alpha \in \mathbb{R}_+$$

Analogously,

$$x \in C \iff \delta_C(x) \leq 0$$

$$\iff u(x) \leq \delta_C^*(u), \quad \forall u \in \text{dom } \delta_C^*$$

$$\iff u(x) \leq \delta_C^*(u) + \beta, \\ \forall u \in \text{dom } \delta_C^*, \quad \forall \beta \in \mathbb{R}_+$$

Consequently, the following linear systems are equivalent to  $\sigma$  :

$$\left\{ \begin{array}{l} u_t(x) \leq f_t^*(u_t), \quad u_t \in \text{dom} f_t^*, \quad t \in T \\ u(x) \leq \delta_C^*(u), \quad u \in \text{dom} \delta_C^* \end{array} \right\}$$

and

$$\left\{ \begin{array}{l} u_t(x) \leq f_t^*(u_t) + \alpha, \quad u_t \in \text{dom} f_t^*, \quad t \in T, \quad \alpha \in \mathbb{R}_+ \\ u(x) \leq \delta_C^*(u) + \beta, \quad u \in \text{dom} \delta_C^*, \quad \beta \in \mathbb{R}_+ \end{array} \right\}$$



## EXISTENCE THEOREMS

For linear systems [(Chu, 1966), Goberna et al. (1995)]:

(i)  $\{a_t(x) \leq b_t, t \in T\}$  is consistent

$\Leftrightarrow$

(ii)  $(0, -1) \notin \text{cl cone}\{(a_t, b_t), t \in T\}$

$\Leftrightarrow$

(iii)  $\text{cl cone}\{(a_t, b_t), t \in T; (0, 1)\} \neq \text{cl cone}\{a_t, t \in T\} \times \mathbb{R}$ .

Associating with  $\sigma$  the convex cones

$$\begin{aligned}M &= \text{cone} \left\{ \bigcup_{t \in T} \text{dom } f_t^* \cup \text{dom } \delta_C^* \right\} \\N &= \text{cone} \left\{ \bigcup_{t \in T} \text{gph } f_t^* \cup \text{gph } \delta_C^* \right\} \\K &= \text{cone} \left\{ \bigcup_{t \in T} \text{epi } f_t^* \cup \text{epi } \delta_C^* \right\} \\P &= \text{cone} \left\{ \bigcup_{t \in T} \text{epi } f_t^* + \text{epi } \delta_C^* \right\}\end{aligned}$$

we get

(i)  $\sigma$  is consistent

$\Leftrightarrow$

(ii)  $(0, -1) \notin \text{cl } K (\text{cl } N, \text{cl } P)$

$\Leftrightarrow$

(iv)  $\text{cl } K \neq \text{cl } M \times \mathbb{R}$

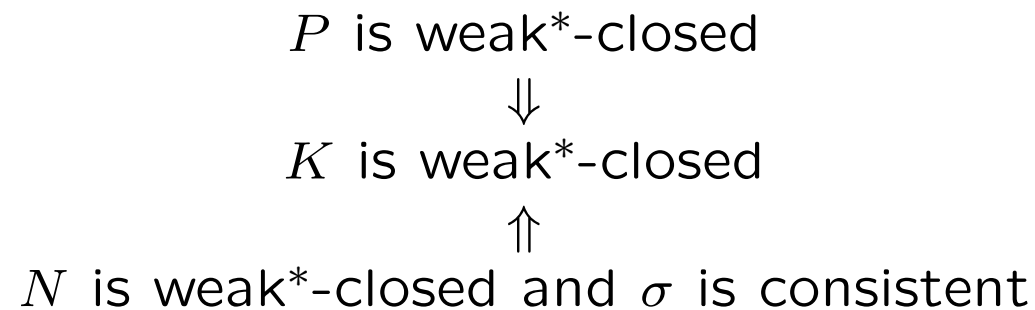
## 1. Short history of these cones

$K$ : Chu (1966), in LISs.

$M$ ,  $N$  and  $K$ : Charnes, Cooper & Kortanek (1965-1969), in LSIP.

$P$ : Jeyakumar, Dinh & Lee (2004), in CP.

## 2. Closedness



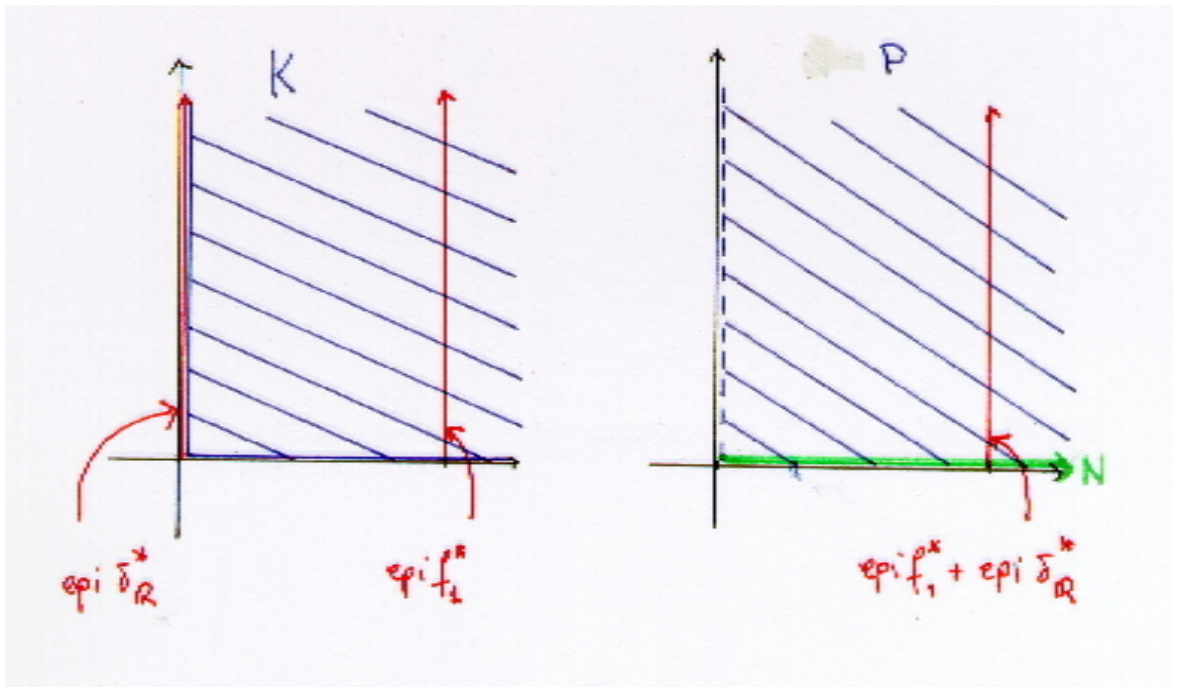
The converse statements are not true and the consistency of  $\sigma$  is not superfluous.

**Example 1:**  $C = X = \mathbb{R}$  and  $\sigma = \{f_1(x) = x \leq 0\}$ .

Since  $f_1^* = \delta_{\{1\}}$  and  $\delta_{\mathbb{R}}^* = \delta_{\{0\}}$ ,  
 $\text{epi } \delta_{\mathbb{R}}^* = \mathbb{R}_+ (0, 1)$

and

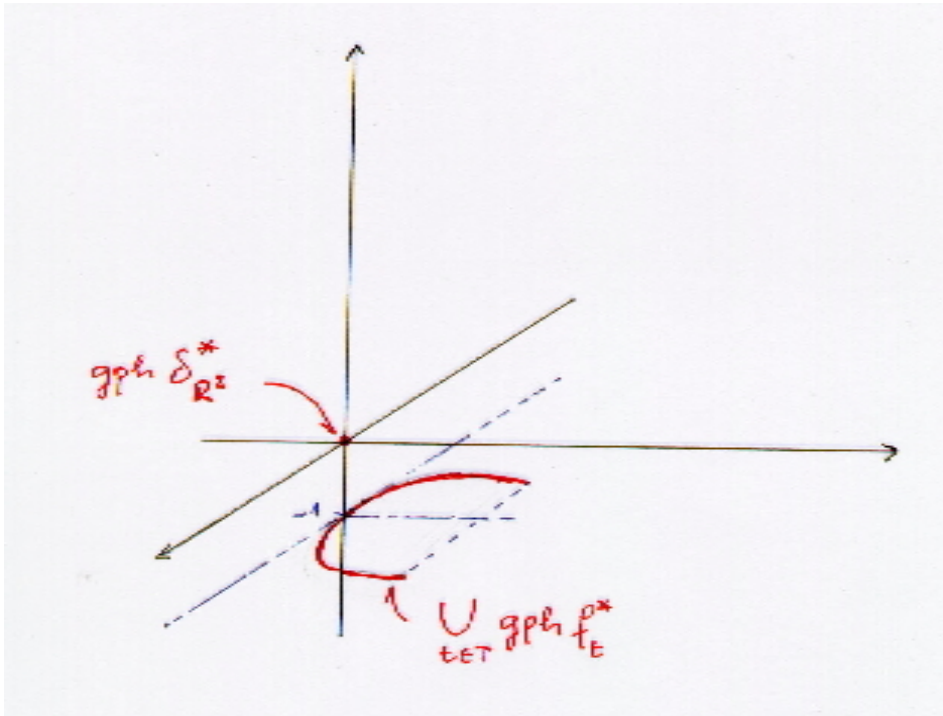
$\text{epi } f_1^* = \text{epi } f_1^* + \text{epi } \delta_{\mathbb{R}}^* = (1, 0) + \mathbb{R}_+ (0, 1)$ .



**Example 2:**  $C = X = \mathbb{R}^2$  and

$$\sigma = \{f_t(x) = tx_1 + t^2x_2 + 1 \leq 0, t \in [-1, 1]\}$$

Since  $\text{gph } \delta_{\mathbb{R}^2}^* = \{(0, 0, 0)\}$  and  $\text{gph } f_t^* = \{(t, t^2, -1)\} \forall t$ ,  $N = \text{cone} \left\{ \bigcup_{t \in T} \text{gph } f_t^* \right\}$  is closed whereas  $K$  is non-closed.



The following **recession condition** was introduced by Borwein (1981):

**(RC)**

$$C^\infty \cap \{x \in X \mid f_t^\infty(x) \leq 0, t \in T\} = \{0\}$$

**Generalized Fan's theorem:** if either

(a)  $K(N)$  is weak\*-closed, or

(b) **(RC)** holds and  $K(N)$  is solid if  $X$  is infinite dimensional

then  $\sigma$  is consistent iff

$\forall \lambda \in \mathbb{R}_+^{(T)}, \exists x_\lambda \in C$  such that

$$\sum_{t \in T} \lambda_t f_t(x_\lambda) \leq 0$$

## Some previous versions

### Under closedness conditions

Bohnenblust, Karlin & Shapley (1950), with  $X = \mathbb{R}^n$  and  $C$  compact.

Fan (1957), assuming that  $f_t : X \rightarrow \mathbb{R} \forall t \in T$  and  $C$  is compact.

Shioji & Takahashi (1988), with  $C$  compact.

### Under recession conditions

Rockafellar (1970), with  $X = \mathbb{R}^n$ .

## ASYMPTOTIC FARKAS LEMMA

From now on we assume that  $\sigma$  is consistent with solution set  $A \neq \emptyset$ .

Given  $v \in X^*$  and  $\alpha \in \mathbb{R}$ , then  $v(x) \leq \alpha$  is a consequence of the consistent system  $\{a_t(x) \leq b_t, t \in T\}$  iff

$$(v, \alpha) \in \text{cl cone} \{(a_t, b_t), t \in T; (0, 1)\}$$

Applying this result (Chu, 1966) to the linearization of  $\sigma$  we get the

**Asymptotic Farkas Lemma for linear inequalities:** given  $v \in X^*$  and  $\alpha \in \mathbb{R}$ ,  $v(x) \leq \alpha$  is consequence of  $\sigma$  iff  $(v, \alpha) \in \text{cl } K$ .



From now on  $f : X \rightarrow \mathbb{R} \cup \{+\infty\}$  denotes a proper l.s.c. convex function.

Another consequence is the

**Asymptotic Farkas Lemma for convex inequalities:**  $f(x) \leq \alpha$  is a consequence of  $\sigma$  iff  $(0, \alpha) + \text{epi} f^* \in \text{cl} K$ .

From here we get the following

**Characterization of the set containment convex-convex:**  $A \subset \{x \in X \mid h_w(x) \leq 0, w \in W\}$  ( $h_w$  as  $f_t$ ) iff

$$\bigcup_{w \in W} \text{epi} h_w^* \subset \text{cl} K$$

## Precedents:

For Farkas' Lemma: see Jeyakumar (2001).

For set containment:

Goberna & López (1998), with  $C = X = \mathbb{R}^n$  and  $f_t$  and  $h_w$  affine  $\forall t \in T$ ,  $\forall w \in W$ .

Mangasarian (2002), with  $C = X = \mathbb{R}^n$  and  $|T| < \infty$  and  $|W| < \infty$ .

Jeyakumar (2003), with  $C = X = \mathbb{R}^n$  and  $h_w$  affine  $\forall w \in W$ .

Goberna, Jeyakumar & Dinh (2006), with  $C = X = \mathbb{R}^n$ .

## FARKAS-MINKOWSKI SYSTEMS

The following concept was introduced by Charnes, Cooper & Kortanek (1965), in LSIP:

$\sigma$  is **FM** if  $K$  is weak\*-closed.

Since  $\text{cl } K = \text{epi } \delta_A^*$ ,  $\{\delta_A(x) \leq 0\}$  is a FM representation of  $A$ .

If  $\sigma$  is FM, then every continuous linear consequence of  $\sigma$  is also consequence of a finite subsystem of  $\sigma$ . The converse statement holds if  $\sigma$  is linear (but not if  $\sigma$  is convex).

**Example 3:** Let  $X = C = \mathbb{R}^n$  and  $\sigma = \{f_1(x) := \frac{1}{2} \|x\|^2 \leq 0\}$ .

Since  $f_1^*(v) = \frac{1}{2} \|v\|^2$ ,  $K = (\mathbb{R}^n \times \mathbb{R}_{++}) \cup \{0\}$  is not closed.

**Non-asymptotic Farkas lemma for linear inequalities:** let  $\sigma$  be FM,  $v \in X^* \setminus \{0\}$  and  $\alpha \in \mathbb{R}$ . Then:

(i)  $v(x) \geq \alpha$  is consequence of  $\sigma$

$\Leftrightarrow$

(ii)  $-(v, \alpha) \in K$

$\Leftrightarrow$

(iii)  $\exists \bar{\lambda} \in \mathbb{R}_+^{(T)}$  such that

$$v(x) + \sum_{t \in T} \bar{\lambda}_t f_t(x) \geq \alpha, \quad \forall x \in C$$

**Non-asymptotic Farkas Lemma for convex inequalities:** if  $\sigma$  is FM, then  $f(x) \leq \alpha$  is consequence of  $\sigma$  iff  $(0, \alpha) + \text{epi} f^* \subset K$ .

**Asymptotic Farkas Lemma for reverse-convex inequalities:** If  $\sigma$  is FM, then  $f(x) \geq \alpha$  is consequence of  $\sigma$  iff

$$(0, -\alpha) \in \text{cl}(\text{epi} f^* + K).$$

From here we get the following **characterization of the set containment convex-reverse convex:**  $A \subset \{x \in X \mid h_w(x) \geq 0, w \in W\}$  ( $h_w$  as  $f_t$ ) iff

$$0 \in \bigcap_{w \in W} \text{cl} \{ \text{epi} h_w^* + K \}$$

Precedents: Jeyakumar (2003) and Bot & Wanka (2005), in CSISs.

The following *closedness condition* was introduced by Burachik & Jeyakumar (2005):

**(CC)**

$\text{epi} f^* + \text{cl} K$  is weak\*-closed.

Each of the following conditions implies **(CC)**:

(i)  $\text{epi} f^* + K$  is weak\*-closed.

(ii)  $\sigma$  is FM and  $f$  is linear.

(ii)  $\sigma$  is FM and  $f$  is continuous at some point of  $A$ .

**Non-asymptotic Farkas Lemma for reverse-convex inequalities:** if  $\sigma$  is FM, **(CC)** holds, and  $\alpha \in \mathbb{R}$ , then

(i)  $f(x) \geq \alpha$  is consequence of  $\sigma$

$\Leftrightarrow$

(ii)  $(0, -\alpha) \in \text{epi} f^* + K$

$\Leftrightarrow$

(iii)  $\exists \bar{\lambda} \in \mathbb{R}_+^{(T)}$  such that

$$f(x) + \sum_{t \in T} \bar{\lambda}_t f_t(x) \geq \alpha, \quad \forall x \in C$$

Precedents: Gwinner (1987) and Dinh, Jeyakumar & Lee (2005) under strong assumptions.

## FM SYSTEMS IN CONVEX OPTIMIZATION

From now on we consider the CP problem

$$\begin{aligned} \text{(P)} \quad & \text{Minimize } f(x) \\ & \text{s.t. } f_t(x) \leq 0, \quad t \in T, \\ & \quad x \in C. \end{aligned}$$

**Solvability theorem:** if  $X = \mathbb{R}^n$  and  $\sigma$  satisfies **(RC)**, then (P) is solvable.

This is not true for reflexive Banach spaces, unless  $f + \delta_A$  is coercive (Zalinescu, 2002).



**Example 4:** let  $X = \ell^2$  (Hilbert space),

$$C := \left\{ x = \{\xi_n\} \in \ell^2 \mid |\xi_n| \leq n \ \forall n \in \mathbb{N} \right\},$$

and  $f(x) := \sum_{n=1}^{\infty} \frac{\xi_n}{n}$ , with  $f \in X'$ .

$C$  is a closed convex set which is not bounded (because  $ne_n \in C$ , for every  $n \in \mathbb{N}$ ) and such that  $C^\infty = \{0\}$ . Thus **(RC)** holds.

Consider  $c^k := (\gamma_n^k)_{n \geq 1}$ ,  $k = 1, 2, \dots$ ,

$$\gamma_n^k := \begin{cases} -n, & \text{if } n \leq k, \\ 0, & \text{if } n > k. \end{cases}$$

We have  $\{c^k\} \in C$  and  $f(c^k) = -k$ ,  $k \in \mathbb{N}$ , so that  $f$  is not bounded from below on  $C$  and no minimizer exists.

**KKT optimality theorem:** assume that  $\sigma$  is FM, that **(CC)** holds, and let  $a \in A \cap \text{dom } f$ . Then  $a$  is a minimizer of (P) iff

$\exists \lambda \in \mathbb{R}_+^{(T)}$  such that

(i)  $\partial f_t(a) \neq \emptyset \quad \forall t \in \text{supp } \lambda$

(ii)  $\lambda_t f_t(a) = 0, \quad \forall t \in T$ , and

(iii)  $0 \in \partial f(a) + \sum_{t \in T} \lambda_t \partial f_t(a) + N_C(a)$

Precedent: without the FM property, the optimality condition is  $0 \in \partial f(a) + N_A(a)$  (Burachik & Jeyakumar, 2005).

Now we consider the *parametric problem*  $(P_u)$ , for  $u \in \mathbb{R}^T$ ,

$$(P_u) \quad \begin{array}{ll} \text{Minimize} & f(x) \\ \text{subject to} & f_t(x) \leq u_t, \quad t \in T, \\ & x \in C, \end{array}$$

with feasible set  $A_u$ .

Defining  $\psi(x, u) := f(x) + \delta_{A_u}(x)$ ,  $\psi : X \times \mathbb{R}^T \rightarrow \mathbb{R} \cup \{+\infty\}$ , we can write

$$(P_u) \quad \text{Minimize } \psi(x, u), x \in X.$$

Then

$$(P) \equiv (P_0) \quad \text{Minimize} \quad \psi(x, 0), x \in X.$$

The *dual problem* of (P) is

$$(D) \quad \text{Maximize} \quad -\psi^*(0, \lambda), \lambda \in \mathbb{R}_+^{(T)}.$$

**Duality theorem:** if (P) is bounded,  $\sigma$  is FM, and **(CC)** holds, then  $v(D) = v(P)$  and (D) is solvable.

Precedents: Rockafellar (1974) and Bonnans & Shapiro (2000).

Consider the **Lagrange function**  $L : X \times \mathbb{R}^{(T)} \rightarrow \mathbb{R} \cup \{+\infty\}$ , where  $L(x, \lambda)$  is

$$\begin{cases} f(x) + \sum_{t \in T} \lambda_t f_t(x), & \text{if } x \in C, \lambda \in \mathbb{R}_+^{(T)}, \\ +\infty, & \text{otherwise.} \end{cases}$$

**Lagrange optimality theorem:** suppose that  $\sigma$  is FM and that **(CC)** holds. Then a point  $a \in A$  is minimizer of (P) iff

$\exists \lambda_0 \in \mathbb{R}_+^{(T)}$  such that  $(a, \lambda_0)$  is a **saddle point** of the Lagrangian function  $L$ , i.e.,

$$L(a, \lambda) \leq L(a, \lambda_0) \leq L(x, \lambda_0), \quad \forall \lambda \in \mathbb{R}_+^{(T)} \quad \forall x \in C.$$

Then  $\lambda_0$  is a maximizer of (D).

Denote by  $v(u)$  the value of  $(P_u)$ , so that  $v(P) = v(0)$ . The following stability concepts (Laurent, 1972) involve the *value function*  $v : \mathbb{R}^T \rightarrow \overline{\mathbb{R}}$ , whose directional derivative at 0 in the direction  $u$  is denoted by  $v'(0, u)$ .

$(P)$  is called:

◆ *inf-stable* if  $v(0) \in \mathbb{R}$  and  $v$  is l.s.c. at 0.

◆ *inf-dif-stable* if  $v(0) \in \mathbb{R}$  and  $\exists \lambda_0 \in \mathbb{R}^{(T)}$  such that

$$v'(0, u) \geq \lambda_0(u), \quad \forall u \in \mathbb{R}^T.$$

These concepts are related as follows:

(P) inf-stable



$v(D) = v(P) \in \mathbb{R}$  (called *normality* in Zalinescu, 2002).

(P) inf-dif-stable



$\partial v(0) \neq \emptyset$  (called *calmness* in Clarke, 1976)



$v(D) = v(P)$  and (D) is solvable.

Thus,  $(P)$  inf-dif-stable  $\implies (P)$  inf-stable

**Stability Theorem:** if  $(P)$  is bounded,  $\sigma$  is FM, and **(CC)** holds, then  $(P)$  is inf-dif-stable.



## LOCALLY FARKAS-MINKOWSKI SYSTEMS

The following local c.q. was introduced by Puente & Vera de Serio (1999), in LSIP. It is also equivalent to the so-called basic c.q. in CP (Hiriart Urruty & Lemarechal, 1993) if  $x \in \text{int } A$  and  $\sup_{t \in T} f_t$  is continuous at  $x$ :

$\sigma$  is **LFM** at  $x \in A$  if

$$N_A(x) \subseteq N_C(x) + \text{cone} \left( \{ \partial f_t(x) \}_{t \in T(x)} \right),$$

where  $T(x) := \{t \in T \mid f_t(x) = 0\}$ .

$\sigma$  is said to be **LFM** if it is LFM at every feasible point  $x \in A$ .

As a consequence of the optimality theorem,

$$\sigma \text{ FM} \Rightarrow \sigma \text{ LFM}$$

If  $\sigma$  is LFM at  $x \in A$ , then every continuous linear consequence of  $\sigma$  which is binding at  $x$  is also consequence of a finite subsystem of  $\sigma$ . The converse statement holds if  $\sigma$  is linear (but not if  $\sigma$  is convex).

$\sigma$  is LFM at  $a$  iff the KKT optimality theorem holds for any l.s.c. convex function  $f$  such that  $a \in \text{dom } f$  and  $f$  is continuous at some point of  $A$ .

Precedent: Li & Ng (2005), with real-valued functions and basic c.q.

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