Overview on Generalized Convexity and Vector Optimization

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Overview on Generalized convexity and VO



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 $E \neq \emptyset$ with partial order (reflexive and transitive) \preccurlyeq ; $A \subseteq E$. $\bar{a} \in A$ is efficient of A if

$$a \in A, a \preccurlyeq \bar{a} \Longrightarrow \bar{a} \preccurlyeq a.$$

The set of \bar{a} is denoted $Min(A, \preccurlyeq)$. Given $x \in E$, lower and upper section at x,

$$L_x \doteq \{y \in E : y \preccurlyeq x\}, \ S_x \doteq \{y \in E : x \preccurlyeq y\},$$

Set

$$S_A \doteq \bigcup_{x \in A} S_x.$$

When $\preccurlyeq = \leq_P, P$ being a convex cone, then

$$(x \preccurlyeq y \Longleftrightarrow y - x \in P) \ L_x = x - P, \ S_x = x + P, \ S_A = A + P.$$



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• Property (*Z*): each totally ordered (chain) subset of *A* has a lower bound in *A*

• A is order-totally-complete (it has no covering of form $\{(L_x)^c : x \in D\}$ with $D \subseteq A$ being totally ordered)

 each maximal totally ordered subset of A has a lower bound in A.

$$A \not\subset \bigcup_{x \in D} L_x^c \Leftrightarrow \emptyset \neq A \cap \left(X \setminus \bigcup_{x \in D} L_x^c \right) \Leftrightarrow \emptyset \neq A \bigcap \bigcap_{x \in D} L_x \Leftrightarrow \exists \, \text{LB}.$$

Sonntag-Zalinescu, 2000; Ng-Zheng, 2002; Corley, 1987; Luc, 1989; Ferro, 1996, 1997, among others.



Introduction

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Basic Definitions:

(a) [Ng-Zheng, 2002] A is order-semicompact (resp. order-s-semicompact) if every covering of A of form {L^c_x : x ∈ D}, D ⊆ A (resp. D ⊆ E), has a finite subcover.
(b) [Luc, 1989; FB-Hernández-Novo, 2008] A es order-complete if *A* covering of form {L^c_{xα} : α ∈ I} where {x_α : α ∈ I} is a decreasing net in A.

A directed set (I, >) is a set $I \neq \emptyset$ together with a reflexive and transitive relation >: for any two elements α , $\beta \in I$ there exists $\gamma \in I$ with $\gamma > \alpha$ and $\gamma > \beta$. A net in *E* is a map from a directed set (I, >) to *E*. A net $\{y_{\alpha} : \alpha \in I\}$ is *decreasing* if $y_{\beta} \preccurlyeq y_{\alpha}$ for each $\alpha, \beta \in I, \beta > \alpha$.



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Theorem

If *A* is order-totally-complete then $Min A \neq \emptyset$.

Proof. Let \mathcal{P} = set of totally ordered sets in A. Since $A \neq \emptyset$, $\mathcal{P} \neq \emptyset$. Moreover, \mathcal{P} equipped with the partial order - inclusion, becomes a partially ordered set. By standard arguments we can prove that any chain in \mathcal{P} has an upper bound and, by Zorn's lemma, we get a maximal set $D \in \mathcal{P}$. Applying a previous equivalence, there exists a lower bound $a \in A$ of D. We claim that $a \in Min A$. Indeed, if $a' \in A$ satisfies that $a' \prec a$ then a' is also a lower bounded of D. Thus, $a' \in D$ by the maximality of *D* in \mathcal{P} . Hence, $a \preccurlyeq a'$ and therefore $a \in Min A$. In particular, if $A \subseteq E$ is order-s-semicompact, order-semicompact or order-complete, then Min $A \neq \emptyset$.



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Teorema [Ng-Zheng, 2002; FB-Hernández-Novo, 2008]

The following are equivalent:

- (a) $Min(A, \preccurlyeq) \neq \emptyset;$
- (b) A has a maximal totally ordered subset minorized by an order-s-semicompact subset H of S_A;
- (c) A has a nonempty section which is order-complete;
- (*d*) *A* has a nonempty section which is order-totally-complete (equiv. satisfies property (*Z*)).

$$S_A \doteq \bigcup_{x \in A} \{ y \in E : x \preccurlyeq y \}.$$

 $(\preccurlyeq = \leq_{P}, \ I(P) = \{0\}); \bar{a} \in \operatorname{Min} A \Longleftrightarrow A \cap (\bar{a} - P) = \{\bar{a}\}.$



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Sketch - proof

 $(a) \Longrightarrow (b)$: Take $a \in Min A$, and consider

 $\mathcal{P} \doteq \{ D \subseteq E \colon L_a \cap A \subseteq D \subseteq S_a \cap A \text{ and } D \text{ is totally ordered } \}.$

It is clear $L_a \cap A$ is totally ordered, $L_a \cap A \in \mathcal{P}$. By equipping \mathcal{P} with the partial order - inclusion- we can prove by standard arguments that any chain in \mathcal{P} has an upper bound. Therefore, there exists a maximal totally ordered element $D_0 \in \mathcal{P}$, i.e.,

$$L_a \cap A \subseteq D_0 \subseteq S_a \cap A \subseteq S_a$$
.

Set $H = \{a\}$. Then D_0 is minorized by H which is an order-*s*-semicompact subset of S_A . It generalizes and unifies results by Luc 1989, Ng-Zheng 2002 among others.



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Optimization problem

X Hausdorff top. s.p; $f : X \to (E, \preccurlyeq)$. Consider

$$\min\{f(x): x \in X\} \qquad (P)$$

 $f(X) \doteq \{f(x) : x \in X\}$. A sol $\bar{x} \in X$ to (*P*) is such that $f(\bar{x}) \in Min(f(X), \preccurlyeq)$.

Theorem [FB-Hernández-Novo, 2008]

Let X compact. If $f^{-1}(L_y)$ closed $\forall y \in f(X)$ (resp. $\forall y \in E$), then f(X)

- (a) is order-semicomp. (resp. f(X) is order-s-semicomp.);
- (b) has the domination property, i.e., every lower section of f(X) has an efficient point.

As a consequence, $Min(f(X), \preccurlyeq) \neq \emptyset$.



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Proof.

We only prove (a) when $f^{-1}(L_y)$ is closed for all $y \in f(X)$. Suppose $\bigcup_{d \in D} L_d^c$ is a covering of f(X) with $D \subseteq f(X)$. Put

$$U_d \doteq \{x \in X \colon f(x) \notin L_d\}.$$

Then, $X = \bigcup_{d \in D} U_d$. Since $f^{-1}(L_d)$ is closed, $U_d = (f^{-1}(L_d))^c$ is open $\forall d \in D$. Moreover, as X is compact, \exists finite set $\{d_1, \ldots, d_r\} \subseteq D$ such that

$$X = U_{d_1} \cup \cdots \cup U_{d_r}.$$

Hence, $L_{d_1}^c \cup \cdots \cup L_{d_r}^c$ covers f(X) and therefore f(X) is order-semicompact.



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We introduce the following new

Definition [FB-Hernández-Novo, 2008]: Let $x_0 \in X$.

We say *f* is decreasingly lower bounded at x_0 if for each net $\{x_{\alpha} : \alpha \in I\}$ convergent to x_0 such that $\{f(x_{\alpha}) : \alpha \in I\}$ is decreasing, the following holds

$$\forall \alpha \in I : f(x_0) \in L_{f(x_\alpha)}.$$

We say that *f* is decreasingly lower bounded (in *X*) if it is for each $x_0 \in X$.



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Proposition [FB-Hernández-Novo, 2008]

If $f^{-1}(L_y)$ is closed $\forall y \in f(X)$, then *f* is decreasingly lower bounded.

Theorem [FB-Hernández-Novo, 2008]

Let X compact. If f is decreasingly lower bounded, then

- (a) f(X) is order-complete;
- (b) f(X) has the domination property;
- (c) $\operatorname{Min} f(X) \neq \emptyset$.



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Special situation

Y top. vec. space ordered by a closed convex cone $P \subseteq Y$. Define \preccurlyeq^{I} in 2^{*Y*}. If $A, B \in 2^{Y}$ then

$$A \preccurlyeq' B \iff B \subseteq A + P.$$

This is partial order: reflexive and transitive [Jahn, 2003; Kuroiwa, 1998, 2003].

Kuroiwa introduces the notion of efficient set for a family of $\mathcal{F} \subseteq$ of nonempty subsets of *Y*. We say $A \in \mathcal{F}$ is a *I*-minimal set $(A \in 1 \operatorname{Min} \mathcal{F})$ if

$$B \in \mathcal{F}, \ B \preccurlyeq' A \implies A \preccurlyeq' B.$$



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- X real Hausd. top. vect. spac.; Y real normed vect. spac.;
- $P \subseteq Y$ a convex cone, int $P \neq \emptyset$, $I(P) \doteq P \cap (-P)$;
- $K \subseteq X$ a closed set; $F : K \to Y$ a vector function.
- E = the set of \bar{x} such that

$$\bar{x} \in K : F(x) - F(\bar{x}) \notin -P \setminus I(P) \ \forall x \in K.$$

Its elements are called efficient points; E_W = the set of \bar{x} such that

$$\bar{x} \in K$$
: $F(x) - F(\bar{x}) \notin -int P \quad \forall x \in K.$

Its elements are called weakly efficient points.

$$E \subseteq E_W = \bigcap_{x \in K} \left\{ \overline{x} \in K : F(\overline{x}) - F(x) \notin \text{ int } P \right\}.$$



Setting of the problem

How we can compute the efficient points?

Theorem: Consider $P = \mathbb{R}^n_+$, F(x) = Cx (linear), K polyhedra.

 \bar{x} is efficient $\iff \exists p^* > 0$ such that \bar{x} solves

 $\min\{\langle p^*, F(x) \rangle : Ax > b, x > 0\}.$

In a standar notation $\bar{x} \in \operatorname{argmin}_{\kappa} \langle p^*, F(\cdot) \rangle$, $K \doteq \{x \in \mathbb{R}^n : Ax \ge b, x \ge 0\}.$

Does the previous theorem remains valid for non linear F?

 $\bar{x} \in E \iff \bar{x} \in [] \operatorname{argmin}_{K} \langle p^{*}, F(\cdot) \rangle (\iff \text{always!!});$ $D^* \in \mathbb{R}^m_{++}$

 \implies weighting method

How to choice $p^* \in \mathbb{R}^m_{++}$?



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Example 1.1.

Let
$$F(x_1, x_2) = (x_1, x_2), x \in K \doteq \{(x_1, x_2) \in \mathbb{R}^2 : x_1 + x_2 \ge 1\}.$$

Here, $E = E_W = \{(x_1, x_2) \in \mathbb{R}^2 : x_1 + x_2 = 1\}.$ However,

$$\inf_{x\in\mathbb{R}}\langle \boldsymbol{\rho}^*,\boldsymbol{F}(x)\rangle=-\infty,\ \boldsymbol{\rho}^*=(\boldsymbol{\rho}_1^*,\boldsymbol{\rho}_2^*),\ \boldsymbol{\rho}_1^*\neq\boldsymbol{\rho}_2^*.$$

Example 1.2.

Let
$$F(x) = (\sqrt{1 + x^2}, x), x \in K = \mathbb{R}$$
. Here,
 $E = E_W =] -\infty, 0]$. However, if $p_2^* > p_1^* > 0$, and
 $\inf_{x \in \mathbb{R}} \langle p^*, F(x) \rangle = -\infty, \ p^* = (p_1^*, p_2^*)$.

A lot of work to do !!!



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Let $h: X \to \mathbb{R} \cup \{+\infty\}$. It is

quasiconvex if

$$h(x) \leq h(y) \Longrightarrow h(\xi) \leq h(y) \quad \forall \xi \in (x, y);$$

or equivalently, $\{x : h(x) \le t\}$ is convex for all $t \in \mathbb{R}$. • semistrictly guasiconvex if

$$h(x) < h(y) \Longrightarrow h(\xi) < h(y) \quad \forall \ \xi \in (x, y).$$

Proposition

If $h: X \to \mathbb{R} \cup \{+\infty\}$ is semistrictly quasiconvex and lower semicontinuous, then it is quasiconvex.



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Theorem [Malivert-Boissard, 1994] $K \subseteq \mathbb{R}^n$ convex

each f_i (i = 1, ..., m) is quasiconvex, semistrictly quasiconvex, and lsc along lines in K. Then

$$E_W = \bigcup \{E(J): J \subseteq \{1,\ldots,m\}, J \neq \emptyset\}.$$

Example 2.

Consider
$$F = (f_1, f_2), K = [0, +\infty[,$$

$$f_1(x) = \begin{cases} 2, & \text{if } x \notin [1,2] \\ 1, & \text{if } x \in [1,2] \end{cases}$$
 $f_2(x) = |x-5|.$

Here, $E = \{2, 5\}, E_W = [1, 8].$



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Some notations

$$G^{\lambda} \doteq \{ x \in K : F(x) - \lambda \in -P \}; G_{\lambda} \doteq \{ x \in K : F(x) - \lambda \notin \text{ int } P \};$$
$$G(y) \doteq \{ x \in K : F(x) - F(y) \notin \text{ int } P \};$$
$$\text{epi } F \doteq \{ (x, y) \in K \times Y : y \in F(x) + P \}.$$

There is no relationship between the closedness of G^{λ} for all $\lambda \in Y$ and the closedness of G(y) for all $y \in K$ even when P is additionally closed.

- *F* : *K* → *Y* is [Penot-Therá, 1979] *P*-lower semicontinuous (*P*-lsc) at *x*₀ ∈ *K* if ∀ open set *V* ⊆ *Y* st *F*(*x*₀) ∈ *V* ∃ an open neighborhood *U* ⊆ *X* of *x*₀ st *F*(*U* ∩ *K*) ⊆ *V* + *P*. We shall say that *F* is *P*-lsc (on *K*) if it is at every *x*₀ ∈ *K*.
- F is \mathbb{R}^m_+ -lsc if and only if each f_i is lsc.



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Proposition [FB, 2003; Bianchi-Hadjisavvas-Schaible, 1997; The Luc, 1989]

 $P \subseteq Y$ is convex cone, $K \subseteq X$ and $S \subseteq Y$ be closed sets such that $S + P \subseteq S$ and $S \neq Y$; $F : K \rightarrow Y$. The following hold.

- (a) If *F* is a *P*-lsc function, then $\{x \in K : F(x) \in \lambda S\}$ is closed for all $\lambda \in Y$;
- (b) Assume int $P \neq \emptyset$ and P closed: F is P-lsc if and only if $\{x \in K : F(x) \lambda \notin int P\}$ is closed for all $\lambda \in Y$;
- (c) Assume int $P \neq \emptyset$ and P closed: epi F is closed if and only if $\{x \in K : F(x) \lambda \in -P\}$ is closed for all $\lambda \in Y$;
- (d) Assume int $P \neq \emptyset$ and P closed: if F is P-lsc then epi F is closed.



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Theorem [Ferro, 1982; (set-valued) Ng-Zheng, 2002]

P convex cone; *K* compact; G^{λ} closed for all $\lambda \in Y$ (\iff epi *F* is closed if int $P \neq \emptyset$). Then $E \neq \emptyset$.

Proof. We know Min $F(X) \neq \emptyset$, thus $E \neq \emptyset$.

$$G^{\lambda} \doteq \{ x \in K : F(x) - \lambda \in -P \}.$$



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Theorem: *P* convex cone, int $P \neq \emptyset$; *K* compact;

 $G(y) \doteq \{x \in K : F(x) - F(y) \notin \text{int } P\} \text{ closed } \forall y \in K. \text{ Then } E_W \neq \emptyset.$

Proof. Notice that $E_W = E(F(K)|C)$ for $C = (\text{int } P) \cup \{0\}$. The closedness of E_W is obvious; it suffices to show that $E_W \neq \emptyset$. If it is not order-complete for C, let $\{F(x_\alpha)\}$ be a decreasing net with $\{(F(x_\alpha) - C)^c\}_\alpha$ forming a covering of F(K). By compactness, (assume) $x_\alpha \to x_0$ for some $x_0 \in K$. If $E_W = \emptyset$, $\exists y \in K$ such that $F(y) - F(x_0) \in -\text{int } P$. For F(y), $\exists \alpha_0$ such that $F(y) - F(x_{\alpha_0}) \notin -C$. This implies

$$\begin{array}{l} F(y) - F(x_{\alpha}) = F(y) - F(x_{\alpha_0}) + F(x_{\alpha_0}) - F(x_{\alpha_0}) \\ \in (Y \setminus -C) + C \subseteq Y \setminus -C \subseteq Y \setminus -\operatorname{int} P \ \forall \ \alpha > \alpha_0. \end{array}$$
$$G(y) \text{ closed implies } F(x_0) - F(y) \notin \operatorname{int} P, \text{ a contradiction.} \end{array}$$



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Since $\operatorname{int} P \neq \emptyset$, take $\overline{y} \in \operatorname{int} P$. Then the set

$$\textit{B} = \{\textit{y}^* \in \textit{P}^* : \langle\textit{y}^*, \bar{\textit{y}}\rangle = 1\}$$

is a *w*^{*}-compact convex base for *P*^{*}, i.e., $0 \notin B$ and $P^* = \bigcup_{t>0} tB$. In this case,

$$oldsymbol{
ho}\inoldsymbol{P}\Longleftrightarrow\langleoldsymbol{
ho}^*,oldsymbol{
ho}
angle\geq 0 \ orall \ oldsymbol{
ho}^*\inoldsymbol{B};$$

$$\boldsymbol{\rho} \in \operatorname{int} \boldsymbol{P} \iff \langle \boldsymbol{\rho}^*, \boldsymbol{\rho} \rangle > 0 \ \forall \ \boldsymbol{\rho}^* \in \boldsymbol{B}.$$

The set E^* of the extreme points of *B* is nonempty by the Krein-Milman theorem.



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Definitions: Let $\emptyset \neq K \subseteq X$, $F : K \rightarrow Y$ is said to be:

1. *P*-convex if, $x, y \in K$,

 $tF(x) + (1-t)F(y) \in F(tx + (1-t)y) + P, \ \forall t \in]0,1[;$

F is \mathbb{R}^{m}_{+} -convex if and only if each f_{i} is convex.

2. properly *P*-quasiconvex [Ferro, 1982] if, $x, y \in K, t \in [0, 1[,$

$$F(tx+(1-t)y) \in F(x)-P$$
 or $F(tx+(1-t)y) \in F(x)+P$,

or equivalently, $\{\xi \in K : F(\xi) \notin \lambda + P\}$ is convex $\forall \lambda \in Y$. $F(x) = (x, -x^2), K =] - \infty, 0]$, satisfies 2 but not 1; $F(x) = (x^2, -x), K = \mathbb{R}$, satisfies 1 but not 2;



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More Definitions

- 3. naturally *P*-quasiconvex [Tanaka, 1994] if, $x, y \in K$, $t \in [0, 1[$
 - $F(tx+(1-t)y) \in \mu F(x)+(1-\mu)F(y)-P$, for some $\mu \in [0,1]$,

or equivalently, $F([x, y]) \in co\{F(x), F(y)\} - P$. $F(x) = (x^2, 1 - x^2), K = [0, 1]$, satisfies 3 but not 2 or 1.

4. scalarly *P*-quasiconvex [Jeyakumar-Oettli-Natividad, 1993] if, for $p^* \in P^* \setminus \{0\}$, $x \in K \mapsto \langle p^*, F(x) \rangle$ is quasiconvex.

Both are equivalent [FB-Hadjisavvas-Vera, 2007] if int $P \neq \emptyset$.

 \implies F(K) + P is convex.



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More Definitions

5. P-quasiconvex [Ferro, 1982] if,

$$\{\xi \in K : F(\xi) - \lambda \in -P\}$$
 is convex $\forall \lambda \in Y$.

F is \mathbb{R}^m_+ -quasiconvex if and only if each f_i is quasiconvex. [Benoist-Borwein-Popovici, 2003] This is equivalent to: given any $p^* \in E^*$, $x \in K \mapsto \langle p^*, F(x) \rangle$ is quasiconvex.

6. semistrictly-P-quasiconvex at y [Jahn-Sachs, 1986] if,

$$x \in K, F(x)-F(y) \in -P \Longrightarrow F(\xi)-F(y) \in -P \ \forall \xi \in]x,y[.$$

[R. Cambini, 1998] When $X = \mathbb{R}^n$, $Y = \mathbb{R}^2$, $P \subseteq \mathbb{R}^2$ polyhedral, int $P \neq \emptyset$, $F : K \to \mathbb{R}^2$ continuous, both are equivalent.



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One more Definition

7. semistrictly-($Y \setminus -int P$)-quasiconvex at y [FB, 2004] if, $x \in K$,

$$F(x) - F(y) \notin \operatorname{int} P \Longrightarrow F(\xi) - F(y) \notin \operatorname{int} P \ \forall \xi \in]x, y[.$$

Teorema [FB, 2004]. Sean X, Y, K, P, F as above. We have:





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P-quasiconv. implies semistrictly $(Y \setminus -int P)$ -quasiconv. **Proof.** Take any $x, y \in K$ such that $F(x) - F(y) \notin int P$, and suppose $\exists \xi \in]x, y[$ satisfying $F(\xi) - F(y) \in int P$. If $F(x) - F(\xi) \in P$, the latter inclusion implies $F(x) - F(y) \in int P$ which cannot happen by the choice of x, y. Hence $F(x) - F(\xi) \notin P$. By a Lemma due to Bianchi-Hadjisavvas-Schaible (1997) ($a \ge 0 \ b < 0 \Rightarrow \exists c \ge 0$, $a \le c, b \le c$) there exists $c \notin P$ such that

$$F(x) - F(\xi) - c \in -P$$
 and $F(y) - F(\xi) - c \in -P$.

By the *P*-quasiconvexity of *F*, we conclude in particular $F(\xi) - F(\xi) - c = -c \in -P$ giving a contradiction. Consequently $F(\xi) - F(y) \notin \text{int } P$ for all $\xi \in]x, y[$, proving the desired result.



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Definition

Given $S \subseteq Y$, $K \subseteq X$ convex. The function $F : K \to Y$ is semistrictly (S)-quasiconvex at $y \in K$, if for every $x \in K$, $x \neq y$, $F(x) - F(y) \in -S \implies F(\xi) - F(y) \in -S \quad \forall \xi \in]x, y[$. We say that F is semistrictly (S)-quasiconvex (on K) if it is at every $y \in K$.

$$F_1(x) = (e^{-x^2}, x^2), \ x \in \mathbb{R}; \ F_2(x) = (rac{1}{1+|x|^2}, |x|), \ x \in \mathbb{R};$$

$$F_3(x_1, x_2) = \left(\frac{x_1^2}{1+x_1^2}, x_2^3\right), \ (x_1, x_2) \in \mathbb{R}^2,$$

are semistrictly ($\mathbb{R}^2\setminus -int\ \mathbb{R}^2_+)\text{-quasiconvex}.$



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Particular cases

 $Y = \mathbb{R}, \mathbb{R}_{+} \doteq [0, +\infty[, \mathbb{R}_{++} \doteq]0, +\infty[:$ semistrict (\mathbb{R}_{+})-quasiconvexity = quasiconvexity; semistrict (\mathbb{R}_{++})-quasiconvexity = semistrict quasiconvexity.

The previous definition is related to the problem of finding $\bar{x} \in X$ satisfying

$$\bar{x} \in K$$
 such that $F(x) - F(\bar{x}) \in S \ \forall x \in K$.

The set of such \bar{x} is denoted by E_S .



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Set
$$\mathcal{L}_y \doteq \{x \in \mathcal{K} : F(x) - F(y) \in -S\}.$$

Proposition

Assume $0 \in S$ (for instance $S = Y \setminus -int P$, $S = Y \setminus -P \setminus I(P)$); *K* convex; $F : K \to Y$, $y \in K$. The FAE:

- (a) F is semistrictly (S)-quasiconvex at y;
- (b) \mathcal{L}_y is starshaped at y.

If $X = \mathbb{R}$, (b) may be substituted by the convexity of \mathcal{L}_{y} .

Proposition

Let *S*, *K* as above, and $\bar{x} \in K$ be a local *S*-minimal for *F* on *K*. Then, $\bar{x} \in E_S \iff F$ is semistrictly $(Y \setminus -S)$ -quasiconvex at \bar{x} .



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To fix ideas let $X = \mathbb{R}^n$, the asymptotic cone of *C* is

$$\boldsymbol{C}^{\infty} \doteq \{\boldsymbol{v} \in \boldsymbol{X}: \exists t_n \downarrow \boldsymbol{0}, \exists x_n \in \boldsymbol{C}, t_n x_n \rightarrow \boldsymbol{v}\},\$$

When *C* is closed and starshaped at $x_0 \in C$, one has

$$C^{\infty}=\bigcap_{t>0}t(C-x_0).$$

If *C* is convex the above expression is independent of $x_0 \in C$.

$$E_{S} \doteq \bigcap_{y \in K} \left\{ x \in \mathcal{K} : F(x) - F(y) \in -S \right\},$$
$$(E_{S})^{\infty} \subseteq \bigcap_{y \in \mathcal{K}} \left\{ x \in \mathcal{K} : F(x) - F(y) \in -S \right\}^{\infty}$$
$$(E_{S})^{\infty} \subseteq \bigcap_{y \in \mathcal{K}} \left\{ v \in \mathcal{K}^{\infty} : F(y + \lambda v) - F(y) \in -S \ \forall \ \lambda > 0 \right\} \doteq R_{S}.$$

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We introduce the following cones in order to deal with the case when *K* unbounded. Here $S \subseteq Y$,

$${m R}_{{m P}}\doteqigcap_{y\in K}\Big\{m v\in K^\infty:\; {m F}(y+\lambdam v)-{m F}(y)\in -{m P}\;\;orall\;\lambda>0\Big\},$$

$$\mathbf{R}_{\mathbf{S}} \doteq \bigcap_{\mathbf{y}\in\mathbf{K}} \Big\{ \mathbf{v}\in\mathbf{K}^{\infty}: \ \mathbf{F}(\mathbf{y}+\lambda\mathbf{v})-\mathbf{F}(\mathbf{y})\in-\mathbf{S} \ \forall \ \lambda>\mathbf{0} \Big\}.$$

We recall that E_S denotes the set of $\bar{x} \in X$ satisfying

$$\bar{x} \in K$$
 such that $F(x) - F(\bar{x}) \in S \ \forall x \in K$.

$$E_{\mathcal{S}} \neq \emptyset \implies 0 \in \mathcal{S}.$$

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$$\mathcal{L}_{y} \doteq \{x \in K; F(x) - F(y) \in -S\}.$$

Theorem

K closed convex; *P* convex cone; $S \subseteq Y$ such that $S + P \subseteq S$; *F* : *K* \rightarrow *Y* semistrictly (*S*)-quasiconvex and \mathcal{L}_y is closed $\forall y \in K$. The following hold:

•
$$E_S + R_P = E_S, R_P \subseteq (E_S)^\infty \subseteq R_S;$$

• if $E_S \neq \emptyset$ and either $X = \mathbb{R}$ or $Y = \mathbb{R}$ (with $P = [0, +\infty[), \implies E_S$ is convex and $(E_S)^{\infty} = R_S$;

•
$$E_P \neq \emptyset \Longrightarrow (E_S)^{\infty} = R_S, (E_P)^{\infty} = R_P.$$

Models: S = P, $S = Y \setminus -(P \setminus I(P))$, $S = Y \setminus -intP$.



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Proposition [FB-Vera, 2006]

 $K \subseteq \mathbb{R}^n$ closed convex; $S \subseteq Y$; $F : K \to Y$ semistrictly (S)-quasiconvex and \mathcal{L}_y closed for all $y \in K$.

• If
$$R_{\mathcal{S}} = \{0\} \Longrightarrow (\mathcal{K}_r \doteq \mathcal{K} \cap \bar{\mathcal{B}}(0,r))$$

$$\exists r > 0, \forall x \in K \setminus K_r, \exists y \in K_r : F(y) - F(x)
ot\in S; \quad (*)$$

• if $X = \mathbb{R}$ then (without the closedness of \mathcal{L}_{y}),

$$R_S = \{0\} \iff (*)$$
 holds.

when $S = Y \setminus -int P$, we denote $E_S = E_W$, $R_S = \tilde{R}_W$.


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Theorem

$$\begin{split} & \mathcal{K} \subseteq \mathbb{R}^n \text{ closed convex; } P \subseteq Y \text{ closed cone; } F : \mathcal{K} \to Y \\ & \text{semistrictly } (Y \setminus -\text{int } P) \text{-quasiconvex with} \\ & G(y) = \{x \in \mathcal{K} : F(x) - F(y) \not\in \text{ int } P\} \text{ closed } \forall y \in \mathcal{K}. \text{ Then} \\ & \tilde{\mathcal{R}}_W = \{0\} \Longrightarrow E_W \neq \emptyset \text{ and compact.} \end{split}$$

Remarks

- Unfortunately, we do not know whether the condition $\tilde{R}_W = \{0\}$ is also necessary for the nonemptines and compactness of E_W in this general setting.
- convex case If $P = \mathbb{R}^m_+$ and each component of F is convex and lsc, the equivalence holds [Deng, 1998]. It will be extended for general cones latter on.
- a nonconvex case If n = 1 or $Y = \mathbb{R}$...



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We set
$$X = \mathbb{R}^n$$
, $Y = \mathbb{R}^m$,

Hypothesis on the cone P

 $P \subseteq \mathbb{R}^m$ is a closed convex cone, int $P \neq \emptyset$ (thus $P^* = \bigcup_{t>0} tB$ for some compact convex set *B*). We require that the set B_0 of extreme points of *B* is closed.

Obviously the polhyedral and the ice-cream cones satisfy the previous hypothesis.



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Set
$$(S = \mathbb{R}^m \setminus -int P)$$

$$E_{W} \doteq \bigcap_{y \in K} \bigcup_{q \in B_{0}} \Big\{ x \in K : \langle q, F(x) - F(y) \rangle \leq 0 \Big\},$$

$$R_{P} = \bigcap_{y \in K} \bigcap_{\lambda > 0} \bigcap_{q \in B_{0}} \Big\{ v \in K^{\infty} : \langle q, F(y + \lambda v) - F(y) \rangle \leq 0 \Big\},$$

Additionally, we also consider the cone

$$ilde{\pmb{\mathcal{R}}}_{\pmb{W}} = igcap_{\pmb{y}\in \pmb{\mathcal{K}}} igcup_{\pmb{q}\in \pmb{B_0}} \Big\{ \pmb{v}\in \pmb{\mathcal{K}}^\infty: \ \langle \pmb{q},\pmb{F}(\pmb{y}+\lambda\pmb{v})-\pmb{F}(\pmb{y})
angle \leq \pmb{0} \ \ \forall \ \lambda > \pmb{0} \Big\}.$$



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Corollary: $h_q(x) = \langle q, F(x) \rangle, \ q \in P^*, x \in K.$

Assume h_q is convex for all $q \in B_0$; $F : K \to \mathbb{R}^m$ *P*-lsc. Then, • if $E_W \neq \emptyset$,

$$igcap_{q\in B_0}\left\{oldsymbol{v}\in K^\infty:\ h^\infty_q(oldsymbol{v})\leq 0
ight\}\subseteq (E_W)^\infty\subseteq$$

$$\bigcup_{q\in B_0}\Big\{v\in \mathcal{K}^\infty:\ h^\infty_q(v)\leq 0\Big\};$$

• if $\operatorname{argmin}_{\mathcal{K}} h_q \neq \emptyset$ for all $q \in B_0$,

$$(E_W)^\infty = igcup_{q\in B_0} \left\{ v\in \mathcal{K}^\infty: \ h^\infty_q(v)\leq 0
ight\} = ilde{\mathcal{R}}_W.$$



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Examples showing optimality of the assumptions

Example 3.1.

Take
$$P = \mathbb{R}^2_+$$
, $K = \mathbb{R}^2$, $f_1(x_1, x_2) = x_1^2$, $f_2(x_1, x_2) = e^{x_2}$. Then $f_1^{\infty}(v_1, v_2) = 0$ if $v_1 = 0$, $f_1^{\infty}(v_1, v_2) = +\infty$ elsewhere; $f_2^{\infty}(v_1, v_2) = 0$ if $v_2 \le 0$, $f_2^{\infty}(v_1, v_2) = +\infty$ elsewhere. Thus,

$$\textit{\textbf{R}}_{\textit{P}} = \{0\} \times \] - \infty, 0], \ \tilde{\textit{\textbf{R}}}_{\textit{W}} = \Bigl(\{0\} \times \mathbb{R}\Bigr) \cup \Bigl(\mathbb{R} \times \] - \infty, 0] \Bigr),$$

while $E_W = \{0\} \times \mathbb{R} = (E_W)^{\infty}$. Notice that $\operatorname{argmin}_{\mathcal{K}} f_2 = \emptyset$.



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The convex case

Theorem [Deng, 1998; FB-Vera, 2006]

 $K \subseteq \mathbb{R}^n$ closed convex; *P* closed convex cone as above. Assume $F : K \to \mathbb{R}^m$ is *P*-lsc such that $\langle q, F(\cdot) \rangle : K \to \mathbb{R}$ is convex $\forall q \in B_0$. The FAE:

(a) E_W is nonempty and compact;

(b) $\operatorname{argmin}_{\mathcal{K}}\langle q, F(\cdot) \rangle$ is nonempty and compact for all $q \in B_0$; (c) $\tilde{R}_w = \{0\}$;



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The nonconvex case: non quasiconvexity

Theorem [FB-Vera, 2006]

 $\begin{array}{l} \mathcal{K} \subseteq \mathbb{R} \text{ closed convex}; P \subseteq Y \text{ convex cone, int } P \neq \emptyset; \\ F : \mathcal{K} \to Y \text{ is semistrictly } (Y \setminus -\text{int } P) \text{-quasiconvex such that} \\ \mathcal{L}_y \text{ is closed } \forall y \in \mathcal{K}. \text{ Then, } E_W \text{ is closed convex, and the FAE:} \\ \textbf{(a)} \quad \tilde{R}_W = \{0\}; \\ \textbf{(b)} \quad \exists r > 0, \forall x \in \mathcal{K} \setminus \mathcal{K}_r, \exists y \in \mathcal{K}_r : F(y) - F(x) \in -\text{int } P, \\ \text{ where } \mathcal{K}_r = [-r, r] \cap \mathcal{K}; \end{array}$

(c) $E_W \neq \emptyset$ and bounded (it is already closed and convex).

When $P = \mathbb{R}^m_+$ some of the components of F may be not quasiconvex.



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The nonconvex case: quasiconvexity

Theorem [FB, 2004; FB-Vera, 2006]

 $Y = \mathbb{R}^n, K \subseteq \mathbb{R}$ is closed convex; $P \subseteq \mathbb{R}^m$ closed convex cone as above. Assume $\langle q, F(\cdot) \rangle : K \to \mathbb{R}$ is lsc and semistrictly quasiconvex $\forall q \in B_0$. The FAE:

(a) E_W is a nonempty compact convex set;

- (b) $\operatorname{argmin}_{\mathcal{K}}\langle q, F(\cdot) \rangle$ is a nonempty compact convex set for all $q \in B_0$;
- (c) $\exists r > 0, \forall x \in K \setminus K_r, \exists y \in K_r (K_r = [-r, r] \cap K)$:

$$\langle q, F(y) - F(x) \rangle < 0 \ \forall \ q \in B_0.$$



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Examples showing optimality of the assumptions

Example 4.1.

Consider
$$P = \mathbb{R}^2_+$$
, $K = \mathbb{R}$, $F(x) = (\sqrt{|x|}, \frac{x}{1+|x|})$, $x \in \mathbb{R}$. Here, $E_W =] - \infty, 0]$.

Example 4.2.

Consider
$$P = \mathbb{R}^2_+$$
, $F = (f_1, f_2)$, $K = [0, +\infty[$ where,

$$f_1(x) = \begin{cases} 2, & \text{if } x \notin [1,2] \\ 1, & \text{if } x \in [1,2] \end{cases} \quad f_2(x) = \begin{cases} -e^{-x+5}, & \text{if } x \ge 5 \\ 4-x, & \text{if } x < 5 \end{cases}$$

Here, $E_W = [1, +\infty[.$



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Conjecture:

Assume that each $f_i : K \subseteq \mathbb{R}^n \to \mathbb{R}$ is semistrictly quasiconvex and lsc, i = 1, ..., m. The FAE:

- *E_W* is nonempty and compact;
- each $\operatorname{argmin}_{K} f_{i}$ is nonempty and compact.



Althernative theorems Characterization through linear scalarization

The starting point: linear case

Theorem [Gordan Paul, 1873] Let A matrix.

Then, exactly one of the following sistems has solution:

(1)
$$Ax < 0;$$

(11) $A^{\top}p = 0, p \ge 0, p \ne 0.$



Althernative theorems Characterization through linear scalarization

The convex case

Theorem [Fan-Glicksberg-Hoffman, 1957] Let $K \subseteq \mathbb{R}^n$ convex,

 $f_i : K \to \mathbb{R}, i = 1, ..., m$, convex. Then, exactly one of the following two sistems has solution:

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

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

Sketch of Proof. Set $F = (f_1, \ldots, f_m)$.

Not $(I) \iff F(K) \cap (-\operatorname{int} \mathbb{R}^m_+) = \emptyset \iff (F(K) + \mathbb{R}^m_+) \cap (-\operatorname{int} \mathbb{R}^m_+) = \emptyset$ $(F(K) + \mathbb{R}^m_+ \text{ is convex } \longrightarrow (II))$ $\overline{\operatorname{cone}}(F(K) + \mathbb{R}^m_+) \cap (-\operatorname{int} \mathbb{R}^m_+) = \emptyset$



Althernative theorems Characterization through linear scalarization

Let *P* closed convex cone with int $P \neq \emptyset$

$$F(K) \approx A \subseteq Y, \ \mathbb{R}^n_+ \approx P$$

(*I*)
$$A \cap (-int P) \neq \emptyset$$
,
(*II*) $co(A) \cap (-int P) = \emptyset$.
Trivial part (*I*) y (*II*) \implies absurd.
Non trivial part: Hipothesis (¿?)

$$A \cap (-\operatorname{int} P) = \emptyset \Longrightarrow \operatorname{co}(A) \cap (-\operatorname{int} P) = \emptyset.$$

 $A \cap (-int P) = \emptyset \iff \overline{cone}(A + P) \cap (-int P) = \emptyset.$

It suffices the convexity of $\overline{\text{cone}}(A + P)!!$



Definition: Let $P \subseteq Y$ closed convex cone, int $P \neq \emptyset$.

The set $A \subseteq Y$ is:

(a) generalized subconvexlike [Yang-Yang-Chen, 2000] if $\exists u \in int P, \forall x_1, x_2 \in A, \forall \alpha \in]0, 1[, \forall \varepsilon > 0, \exists \rho > 0$ such that

$$\varepsilon \boldsymbol{u} + \alpha \boldsymbol{x}_1 + (1 - \alpha) \boldsymbol{x}_2 \in \rho \boldsymbol{A} + \boldsymbol{P}; \tag{1}$$

- (b) presubconvexlike [Zeng, 2002] if $\exists u \in Y, \forall x_1, x_2 \in A, \forall \alpha \in]0, 1[, \forall \varepsilon > 0, \exists \rho > 0$ such that (1) holdse;
- (c) nearly subconvexlike [Sach, 2003; Yang-Li-Wang, 2001] if $\overline{\text{cone}}(A + P)$ is convex.



$$\operatorname{cone}_+(A + \operatorname{int} P)$$
 is $\operatorname{convex} \iff \operatorname{cone}(A + \operatorname{int} P)$ is convex .
 $\implies \overline{\operatorname{cone}}(A + \operatorname{int} P) = \overline{\operatorname{cone}}(\overline{A + \operatorname{int} P}) = \overline{\operatorname{cone}}(\overline{A + \operatorname{int} P}) = \overline{\operatorname{cone}}(A + P) = \overline{\operatorname{cone}}(A + P)$ is convex .

Also,

 $\operatorname{int}(\overline{\operatorname{cone}_+(A+P)}) = \operatorname{int}(\overline{\operatorname{cone}_+(A)+P}) = \operatorname{cone}_+(A) + \operatorname{int} P = \operatorname{cone}_+(A+\operatorname{int} P) \text{ is convex. Consequently,}$ $\operatorname{cone}(A+P) \text{ is convex} \iff \operatorname{cone}(A+\operatorname{int} P) \text{ is convex.}$ Here, $\overline{\operatorname{cone}(M)} = \overline{\operatorname{cone}_+(M)}$.

$$\operatorname{cone}(M) = \bigcup_{t \ge 0} tM, \ \operatorname{cone}_+(M) \doteq \bigcup_{t > 0} tM, \ \overline{\operatorname{cone}}(M) = \overline{\operatorname{cone}(M)}.$$

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Theorem [Yang-Yang-Chen, 2000; Yang-Li-Wang, 2001]

 $P \subseteq Y$ as above, $A \subseteq Y$. Assume $\overline{\text{cone}}(A + P)$ is convex. Then

$$A \cap (-\operatorname{int} P) = \emptyset \Longrightarrow \operatorname{co}(A) \cap (-\operatorname{int} P) = \emptyset.$$

Example: [FB-Hadjisavvas-Vera, 2007]



Althernative theorems Characterization through linear scalarization

Def: A cone $K \subseteq Y$ is called "pointed" if

 $x_1 + \cdots + x_k = 0$ is impossible for x_1, x_2, \dots, x_k in K unless $x_1 = x_2 = \cdots = x_k = 0$. (\iff co $K \cap (-co K) = \{0\}$).

Our first main result is the following:

Theorem [FB-Hadjisavvas-Vera, 2007]

: $\emptyset \neq A \subseteq Y$, $P \subseteq Y$ convex closed cone, int $P \neq \emptyset$. The FAE: (a) cone(A + int P) is pointed;

(b) $\operatorname{co}(A) \cap (-\operatorname{int} P) = \emptyset$.



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Sketch of proof

We first prove cone(A + int P) is pointed $\implies A \cap (-int P) = \emptyset$. If $\exists x \in A \cap (-int P)$, then $x = 2(x - \frac{x}{2}) \in \operatorname{cone}(A + int P)$ and $-x = x + (-2x) \in A + \text{int } P \subseteq \text{cone}(A + \text{int } P)$. By pointedness, 0 = x + (-x) implies $x = 0 \in int P$, a contradiction. Now assume that (a) holds. If (b) does not hold, $\exists x \in -int P$ such that $x = \sum_{i=1}^{m} \lambda_i a_i$ with $\sum_{i=1}^{m} \lambda_i = 1, \lambda_i > 0, a_i \in A$. Thus, $0 = \sum_{i=1}^{m} \lambda_i(a_i - x)$. Using (a), $\lambda_i(a_i - x) = 0 \ \forall i = 1, ..., m$, a contradiction. Conversely, assume (b) holds. If cone(A + int P)is not pointed, then $\exists x_i \in \text{cone}(A + \text{int } P) \setminus \{0\}, i = 1, 2, ..., n$ $\sum_{i=1}^{n} x_i = 0$. So, $x_i = \lambda_i (y_i + u_i)$ with $\lambda_i > 0$, $y_i \in A$ and $u_i \in \text{int } P$. Hence $\sum_{i=1}^n \lambda_i y_i = -\sum_{i=1}^n \lambda_i u_i$. Setting $\mu_i = \lambda_i / \sum_{i=1}^n \lambda_i$ we get $\sum_{i=1}^{n} \mu_i y_i = -\sum_{i=1}^{n} \mu_i u_i \in co(A) \cap (-int P)$, a contradiction.

Althernative theorems Characterization through linear scalarization

The optimal 2D alternative theorem

Theorem [FB-Hadjisavvas-Vera, 2007]

Let $P \subseteq \mathbb{R}^2$ be a cone as before with int $P \neq \emptyset$, and $A \subseteq \mathbb{R}^2$ be satisfying $A \cap (-int P) = \emptyset$. The following hold:

$$co(A) \cap (-int P) = \emptyset \iff cone(A + P)$$
 is convex \iff

 $\operatorname{cone}(A + \operatorname{int} P)$ is $\operatorname{convex} \iff \operatorname{cone}(A) + P$ is $\operatorname{convex} \iff$

 $\overline{\text{cone}}(A+P)$ is convex.

We are in \mathbb{R}^2 , int $(\operatorname{cone}_+(A + P)) \cup \{0\} = \operatorname{cone}(A + \operatorname{int} P) \subseteq$

 $\operatorname{cone}(A + \operatorname{int} P) \subseteq \operatorname{cone}(A + P) \subseteq \operatorname{cone}(A) + P \subseteq \overline{\operatorname{cone}}(A + P).$



Althernative theorems Characterization through linear scalarization

Remark

$$A \cap (-\operatorname{int} P) = \emptyset \& \operatorname{cone}(A + P) \text{ is convex} \iff \operatorname{co}(A) \cap (-\operatorname{int} P) = \emptyset.$$



Theorem [FB-Hadjisavvas-Vera, 2007] Y LCTVS.

 $P \subseteq Y$ closed convex cone, int $P \neq \emptyset$ and int $P^* \neq \emptyset$. The FAE: (*a*) for every $A \subseteq Y$ one has

 $co(A) \cap (-int P) = \emptyset \Rightarrow \overline{cone}(A + P)$ is convex;

(b) for every $A \subseteq Y$ one has

 $co(A) \cap (-int P) = \emptyset \Rightarrow cone (A) + P$ is convex;

(c) for every $A \subseteq Y$ one has

 $co(A) \cap (-int P) = \emptyset \Rightarrow cone (A + int P)$ is convex;

(d) Y is at most two-dimensional.

The assumption int $P^* \neq \emptyset$ (which corresponds to pointedness of *P* when *Y* is finite-dimensional) cannot be removed. Indeed, let $P = \{y \in Y : \langle p^*, y \rangle \ge 0\}$ where $p^* \in Y^* \setminus \{0\}$. Then $P^* = \operatorname{cone}(\{p^*\})$, int $P^* = \emptyset$. For any nonempty $A \subseteq Y$, the set $\operatorname{cone}(A + \operatorname{int} P)$ is convex if $A \cap (-\operatorname{int} P) = \emptyset$ $(\iff A \subseteq P \iff \operatorname{co}(A) \cap (-\operatorname{int} P) = \emptyset)$. Thus, the previous implication holds independently of the dimension of the space *Y*.



Althernative theorems Characterization through linear scalarization

Characterization of weakly efficient solutions via linear scalarization

 $K \subseteq \mathbb{R}^n$ convex and *P* as above. Given $F : K \to \mathbb{R}^m$, we consider

$$\bar{x} \in K$$
: $F(x) - F(\bar{x}) \notin -int P$, $\forall x \in K$,

Clearly, $\bar{x} \in E_W \iff (F(K) - F(\bar{x})) \cap -int P = \emptyset$.

Teorema[FB-Hadjisavvas-Vera, 2007]: The FAE

(a)

$$\bar{x} \in \bigcup_{p^* \in P^*, p^* \neq 0} \operatorname{argmin}_{\mathcal{K}} \langle p^*, F(\cdot) \rangle;$$

(b) $\operatorname{cone}(F(K) - F(\bar{x}) + \operatorname{int} P)$ is pointed.



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Theorem [FB-Hadjisavvas-Vera, 2007]

Set m = 2. The FAE:

(a)

$$\bar{x} \in \bigcup_{p^* \in P^*, p^* \neq 0} \operatorname{argmin}_{\mathcal{K}} \langle p^*, F(\cdot) \rangle;$$

- (b) $\bar{x} \in E_W$ and $\operatorname{cone}(F(K) F(\bar{x}) + \operatorname{int} P)$ is convex.
- (c) $\bar{x} \in E_W$ and $\operatorname{cone}(F(K) F(\bar{x}) + P)$ is convex.
- (d) $\bar{x} \in E_W$ and $\operatorname{cone}(F(K) F(\bar{x})) + P$ is convex.

$$\operatorname{cone}(A) = \bigcup_{t \ge 0} tA.$$



Althernative theorems Characterization through linear scalarization

Example

Consider $F = (f_1, f_2), K = [0, +\infty[$ where,

$$f_1(x) = \begin{cases} 2, & \text{if } x \notin [1,2] \\ 1, & \text{if } x \in [1,2] \end{cases}$$
 $f_2(x) = |x-5|.$

Here, $E_W = [1, 8]$, whereas

$$\bigcup_{\boldsymbol{\rho}^* \in \mathbb{R}^2_+, \boldsymbol{\rho}^* \neq 0} \operatorname{argmin}_{\boldsymbol{K}} \langle \boldsymbol{\rho}^*, \boldsymbol{F}(\cdot) \rangle = [1, 5].$$



Althernative theorems Characterization through linear scalarization





Althernative theorems Characterization through linear scalarization

Open problem

to find an assumption convexity of ?? (*) such that

(*) &
$$A \cap (-\operatorname{int} P) = \emptyset \implies \operatorname{co}(A) \cap (-\operatorname{int} P) = \emptyset$$
.

At least for $A \approx G(K)$ some class of vector functions $G: K \rightarrow Y$.



Althernative theorems Characterization through linear scalarization

Characterizing the Fritz-John type optimality conditions in VO

Take *X* normed space. It is known that if \bar{x} is a local minimum point for (differentiable) $F : K \to \mathbb{R}$ on *K*, then

 $\nabla F(\bar{x}) \in (T(K;\bar{x}))^*.$

Here, $T(C; \bar{x})$ denotes the contingent cone of C at $\bar{x} \in C$,

$$T(C;\bar{x}) = \Big\{ v \in X : \exists t_k \downarrow 0, v_k \in X, v_k \to v, \bar{x} + t_k v_k \in C \ \forall k \Big\}.$$

How to extend to the vector case ?



 $K \subseteq X$ closed; $F : K \to \mathbb{R}^m$; $P \subseteq \mathbb{R}^m$, int $P \neq \emptyset$, a vector $\bar{x} \in K$ is a local weakly efficient solution for F on K ($\bar{x} \in E_W^{loc}$), if there exists an open neighborhood V of \bar{x} such that

 $(F(K \cap V) - F(\bar{x})) \cap (-\operatorname{int} P) = \emptyset.$

We say that a function $h: X \to \mathbb{R}$ admits a Hadamard directional derivative at $\bar{x} \in X$ in the direction v if

$$\lim_{\substack{(t,u)\to(0^+,v)}}\frac{h(\bar{x}+tu)-h(\bar{x})}{t}\in\mathbb{R}.$$

In this case, we denote such a limit by $dh(\bar{x}; v)$.



Althernative theorems Characterization through linear scalarization

If $F = (f_1, \ldots, f_m)$, we set

$$\mathcal{F}(\mathbf{v}) \doteq ((df_1(\bar{\mathbf{x}}; \mathbf{v}), \ldots, df_m(\bar{\mathbf{x}}; \mathbf{v})),$$

$$\mathcal{F}(T(K;\bar{x})) = \{\mathcal{F}(v) \in \mathbb{R}^m : v \in T(K;\bar{x})\}.$$

It is known that if $df_i(\bar{x}; \cdot)$, i = 1, ..., m do exist in $T(K; \bar{x})$, and $\bar{x} \in E_W^{loc}$, then

 $(df_1(\bar{x}; v), \ldots, df_m(\bar{x}; v)) \in \mathbb{R}^m \setminus -int P, \forall v \in T(K; \bar{x}),$

or equivalently, $\mathcal{F}(T(K; \bar{x})) \cap (-int P) = \emptyset$.



Theorem [FB-Hadjisavvas-Vera, 2007][$Y = \mathbb{R}^{m}$]

Under the assumptions above, the FAE:

- $\exists (\alpha_1^*, \ldots, \alpha_m^*) \in \mathcal{P}^* \setminus \{0\}, \ \alpha_1^* df_1(\bar{x}, v) + \cdots + \alpha_m^* df_m(\bar{x}, v) \geq 0 \ \forall \ v \in T(K; \bar{x});$
- $\operatorname{cone}(\mathcal{F}(T(K;\bar{x})) + \operatorname{int} P)$ is pointed.

A more precise formulation may be obtained when m = 2.

Theorem [FB-Hadjisavvas-Vera, 2007][m = 2]

The FAE:

- $\exists (\alpha_1^*, \alpha_2^*) \in P^* \setminus \{0\}, \ \alpha_1^* df_1(\bar{x}, v) + \alpha_2^* df_2(\bar{x}, v) \ge 0 \ \forall v \in T(K; \bar{x});$
- $\mathcal{F}(T(K;\bar{x})) \cap (-int P) = \emptyset \& \operatorname{cone}(\mathcal{F}(T(K;\bar{x})) + int P) \text{ is convex.}$



Althernative theorems Characterization through linear scalarization

 $P = \mathbb{R}^m_+, f_i : \mathbb{R}^n \to \mathbb{R}$ is diff. for i = 1, ..., m. Then $df_i(\bar{x}, v) = \langle \nabla f_i(\bar{x}), v \rangle$,

$$\mathcal{F}(\mathbf{v}) = (\langle \nabla f_1(\bar{\mathbf{x}}), \mathbf{v} \rangle, \dots, \langle \nabla f_m(\bar{\mathbf{x}}), \mathbf{v} \rangle).$$

Moreover,



Althernative theorems Characterization through linear scalarization

Example [FB-Hadjisavvas-Vera, 2007]

$$K = \{(x_1, x_2) : (x_1 + 2x_2)(2x_1 + x_2) \le 0\}.$$
 Take $f_i(x_1, x_2) = x_i$,
 $\bar{x} = (0, 0) \in E_W$: $T(K; \bar{x}) = K$ is nonconvex; $\mathcal{F}(v) = v$;
 $(T(K; \bar{x}))^* = \{(0, 0)\}$, and

 $\operatorname{co}(\{\nabla f_i(\bar{x}): i=1,2\}) \cap (T(K;\bar{x}))^* = \emptyset.$





Althernative theorems Characterization through linear scalarization

In the example,

$$\operatorname{cone}(\mathcal{F}(T(\mathcal{K};\bar{x})) + \mathbb{R}^2_+) = \bigcup_{t \ge 0} t(T(\mathcal{K};\bar{x}) + \mathbb{R}^2_+) \text{ is nonconvex.}$$

On the other hand, due to the linearity of \mathcal{F} (when each f_i is differentiable), if $\mathcal{T}(K; \bar{x})$ is convex then

$$\operatorname{cone}(\mathcal{F}(T(K;\bar{x})) + \mathbb{R}^m_+) = \bigcup_{t \ge 0} t(\mathcal{F}(T(K;\bar{x})) + \mathbb{R}^m_+) \text{ is also convex.}$$

This fact was point out earlier in [Wang, 1988], i.e., if $T(K; \bar{x})$ is convex the condition above is a necessary optimality condition. The convexity of $T(K; \bar{x})$ is the only case ??



Althernative theorems Characterization through linear scalarization

Thus,

$$\operatorname{co}(\{\nabla f_i(\bar{x}): i = 1, 2\}) \cap (T(K; \bar{x}))^* \neq \emptyset. \text{ And}$$
$$\bigcup_{t \ge 0} t(\mathcal{F}(T(K; \bar{x})) + \mathbb{R}^2_+) = \bigcup_{t \ge 0} t(T(K; \bar{x}) + \mathbb{R}^2_+) \text{ is convex.}$$

Althernative theorems Characterization through linear scalarization

A non linear scalarization procedure

Def.: Let $a \in Y$, $e \in int P$.

Define $\xi_{e,a} \colon Y \longrightarrow \mathbb{R} \cup \{-\infty\}$, by

$$\xi_{e,a}(y) = \operatorname{sef}\{t \in \mathbb{R} \colon y \in te + a - P\}.$$


Althernative theorems Characterization through linear scalarization

Def.
$$A \subseteq Y$$
, $\xi_{e,A} : Y \to \mathbb{R} \cup \{-\infty\}$:

$$\xi_{e,A}(y) = \operatorname{suf}\{t \in \mathbb{R} \colon y \in te + A - P\}.$$

$$\xi_{e,A}(y) = \inf_{a \in A} \xi_{e,a}(y). \tag{1}$$





Althernative theorems Characterization through linear scalarization

Lemma [Hernández-Rodriguez, 2007]: Let $\emptyset \neq A \subseteq Y$ and *P* as above.

Then,
$$A - P \neq Y \iff \xi_{e,A}(y) > -\infty \quad \forall \ y \in Y.$$

By taking into account that

$$\operatorname{int}(\overline{A-P}) = \operatorname{int}(A-P) = A - \operatorname{int}P, \ \overline{A-P} = \overline{A - \operatorname{int}P},$$

one can prove,

Lemma: Let $A \subseteq Y$, $r \in \mathbb{R}$, $y \in Y$.

Then

(a)
$$\xi_{e,A}(y) < r \Leftrightarrow y \in re + A - int(P);$$

(b) $\xi_{e,A}(y) \leq r \Leftrightarrow y \in re + \overline{A - P};$
(c) $\xi_{e,A}(y) = r \Leftrightarrow y \in re + \partial(A - P).$



Althernative theorems Characterization through linear scalarization

Corollary: Let $\emptyset \neq P \subseteq Y$ closed convex proper cone.

(a) If int $P \neq \emptyset$ and $E_W \neq \emptyset$, then

$$E_W = E(\xi_{e,f(E_W)} \circ f, K) = \bigcup_{x \in E_W} E(\xi_{e,f(x)} \circ f, K).$$

If in addition $E(\xi_{e,f(x)} \circ f, K) \neq \emptyset$ for some $x \in K$, then

$$E_W = \bigcup_{x \in K} E(\xi_{e,f(x)} \circ f, K);$$

(b) if $E \neq \emptyset$, then

$$E = \bigcup_{x \in E} E(\xi_{e,f(x)} \circ f, K) \subseteq E(\xi_{e,f(E)} \circ f, K);$$



The positive orthant

Example 1.

Consider $F(x) = (x, \sqrt{1 + x^2}), x \in K = \mathbb{R}$. Here, $E_W =] - \infty, 0]$. However, if $p_1^* > p_2^* > 0$, then

$$\inf_{x\in\mathbb{R}}\langle p^*,F(x)\rangle=-\infty,\ p^*=(p_1^*,p_2^*).$$

Example 2.

Consider
$$F = (f_1, f_2), K = [0, +\infty[$$
 where,

$$f_1(x) = \left\{ egin{array}{ccc} 2, & ext{if } x
ot\in [1,2] \ 1, & ext{if } x \in [1,2] \end{array}
ight. f_2(x) = |x-5|. ext{ Here } E_W = [1,8]. \end{array}
ight.$$



Theorem [FB-Vera, 2008]

 $K \subseteq \mathbb{R}$ is closed convex; $f_i : K \to \mathbb{R}$ is lsc and quasiconvex for all i = 1, ..., m. The following assertions hold: (a) if $\emptyset \neq E_W \neq \mathbb{R}$, then there exists *j* such that $\operatorname{argmin}_{K} f_i \neq \emptyset$;

(b) if $K \neq \mathbb{R}$: then $E_W \neq \emptyset \iff \exists j$, $\operatorname{argmin}_K f_j \neq \emptyset$.

Theorem [FB-Vera, 2008]

 $K \subseteq \mathbb{R}$ is closed convex; $f_i : K \to \mathbb{R}$ is lsc and semistrictly quasiconvex for all i = 1, ..., m. Assume $E_W \neq \emptyset$. Then, either

$$E_W = \mathbb{R}$$
 or $E_W = \operatorname{co}\left(\bigcup_{j \in J} \operatorname{argmin}_{\mathcal{K}} f_j\right) + R_W.$



The bicriteria case

We consider $F: K \subseteq \mathbb{R} \to \mathbb{R}^2$ such that

$$[\alpha_1,\beta_1] \doteq \operatorname{argmin}_{K} f_1, \ [\alpha_2,\beta_2] \doteq \operatorname{argmin}_{K} f_2,$$

$$-\infty < \alpha_1 \le \beta_1 < \alpha_2 \le \beta_2 < +\infty.$$

Set

$$A_{+} \doteq \{ x \in [\beta_{1}, \alpha_{2}] : f_{1}(x) = f_{1}(\alpha_{2}) \}, \\ A_{-} \doteq \{ x \in [\beta_{1}, \alpha_{2}] : f_{2}(x) = f_{2}(\beta_{1}) \}.$$

$$\gamma_{+} = \begin{cases} f_{2}(\alpha_{0}^{+}), & \mathbf{A}_{+} =]\alpha_{0}^{+}, \alpha_{2}] \\ \lambda_{+}, & \mathbf{A}_{+} = [\alpha_{0}^{+}, \alpha_{2}] \end{cases} \qquad \lambda_{+} \doteq \lim_{t \downarrow 0} f_{2}(\alpha_{0}^{+} - t).$$





$$M_{1}^{+} \doteq \{ x \in K : x > \beta_{2}, f_{1}(x) = f_{1}(\alpha_{2}) \},$$
$$M_{2}^{+} \doteq \{ x \in K : x > \beta_{2}, f_{2}(x) = \gamma_{+} \}.$$



Theorem [FB-Vera, 2008]

 $K \subseteq \mathbb{R}$ is closed convex; $f_i : K \to \mathbb{R}$ is lsc and quasiconvex for all i = 1, 2. Then A_+ and A_- are convex and nonempty. Moreover, we also have:

(a)
$$\bar{x} > \beta_2$$
: if $A_+ =]\alpha_0^+, \alpha_2], \alpha_0^+ \ge \beta_1$, then
 $\bar{x} \in E_W \iff f_2(\bar{x}) \le f_2(\alpha_0^+), f_1(\bar{x}) = f_1(\alpha_2);$
(b) $\bar{x} > \beta_2$: if $A_+ = [\alpha_0^+, \alpha_2], \alpha_0^+ > \beta_1$, then
 $\bar{x} \in E_W \iff f_2(\bar{x}) \le \lambda_+, f_1(\bar{x}) = f_1(\alpha_2);$
where $\lambda_+ \doteq \lim_{t \downarrow 0} f_2(\alpha_0^+ - t) = \inf_{y < \alpha_0^+} f_2(y).$



Theorem [continued...]

(c)
$$\bar{x} < \alpha_1$$
: if $A_- = [\beta_1, \alpha_0^-[, \alpha_0^- \le \alpha_2, \text{then}]$
 $\bar{x} \in E_W \iff f_1(\bar{x}) \le f_1(\alpha_0^-), f_2(\bar{x}) = f_2(\beta_1)$ }
(d) $\bar{x} < \alpha_1$: if $A_- = [\beta_1, \alpha_0^-], \alpha_0^- < \alpha_2, \text{ then}]$
 $\bar{x} \in E_W \iff f_1(\bar{x}) \le f_1(\lambda_-), f_2(\bar{x}) = f_2(\beta_1)$ }
where $\lambda_- \doteq \lim_{t \downarrow 0} f_2(\alpha_0^- + t)$.



Theorem [FB-Vera, 2008]

 $K \subseteq \mathbb{R}$ convex closed; $f_i : K \to \mathbb{R}$ be lsc and quasiconvex for i = 1, 2.

(a) If f_2 is semistrictly quasiconvex and $M_1^+ \cap M_2^+ \neq \emptyset$, then $E_w = [\alpha_1, \bar{x}]$, where $\bar{x} \in K$ solves the system

$$\bar{x} > \beta_2 \ f_1(\bar{x}) = f_1(\alpha_2), \ f_2(\bar{x}) = \gamma_+.$$

(b) If f_1 is semistrictly quasiconvex and $M_1^- \cap M_2^- \neq \emptyset$, then $E_w = [\bar{x}, \beta_2]$, here $\bar{x} \in K$ solves the system

$$\bar{x} < \alpha_1 \ f_2(\bar{x}) = f_2(\beta_1), \ f_1(\bar{x}) = \gamma_-.$$



Example.

Consider
$$F = (f_1, f_2), K = [0, +\infty[,$$

$$f_1(x) = \begin{cases} 2, & \text{if } x \notin [1,2] \\ 1, & \text{if } x \in [1,2] \end{cases} \quad f_2(x) = |x-5|.$$

Here,
$$E = \{2, 5\}, E_W = [1, 8].$$





TABLE 1

error	total cpu time	γ_+ /iterations	max <i>E_w</i> /iterations
10 ⁻³	0.0150000000	2.9992675781/12	7.9993314775/40
10 ⁻⁴	0.0160000000	2.9999084430/15	7.9999226491/42
10 ⁻⁵	0.0160000000	2.9999942780/19	7.9999965455/45
10 ⁻⁶	0.0160000000	2.9999992847/22	7.9999993877/49



Example.

Let
$$K = [0, +\infty[$$
,

$$f_1(x) = \begin{cases} 2 & \text{si } x < 1, \\ 1 & \text{if } x \in [1, 2], \\ 2 & \text{if } x \in]2, 7[, \\ \sqrt{x - 7} + 2 & \text{if } x > 7, \end{cases}$$
$$f_2(x) = \begin{cases} 6 - x & \text{if } x < 4, \\ e^{-(x - 4)^2} + 3 & \text{if } x \ge 4, \end{cases}$$

Here $E_w = [0, 7]$.





TABLE2

error	total cpu time	γ_+ /iterations	max <i>E_w</i> /iterations
10 ⁻³	0.0140000000	3.9990234375/11	6.9999681538/40
10 ⁻⁴	0.0160000000	3.9999389648/15	6.9999908912/42
10 ⁻⁵	0.0160000000	3.9999923706/18	6.9999997729/48
10 ⁻⁶	0.0160000000	3.9999990463/21	6.9999997729/48



Multicriteria case

We describe E_W in the multicriteria case, that is when m > 2, since

$$E_{W} = \bigcup \{ E_{W}(I) : I \subseteq \{1, 2..., m\}, |I| \le 2 \},$$
(1)

where $E_W(I)$ is the set of \bar{x} solutions to the subproblem

$$\bar{x} \in \mathcal{K}: F_{l}(x) - F_{l}(\bar{x}) \notin -\mathrm{int} \mathbb{R}_{+}^{|l|} \quad \forall x \in \mathcal{K}.$$

Here, $F_I = (f_i)_{i \in I}$ and $\mathbb{R}^{|I|}_+$ is the positive orthant in $\mathbb{R}^{|I|}$. One inclusion in (1) trivially holds since $E_W(I) \subseteq E_W(I')$ if $I \subseteq I'$; the other is a consequence of the following Helly's theorem since each f_i is quasiconvex.



Helly's theorem

Let C_i , i = 1, ..., m, be a collection of convex sets in \mathbb{R}^n . If every subcollection of n + 1 or fewer of these C_i has a nonempty intersection, then the entire collection of the *m* sets has a nonempty intersection.





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