

CONVEXITY AND GLOBAL WELL-POSEDNESS IN SET-OPTIMIZATION

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Abstract. Well-posedness for vector optimization problems has been extensively studied. More recently, some attempts to extend these results to set-valued optimization have been proposed, mainly applying some scalarization. In this paper we propose a new definition of global well-posedness for set-optimization problems.

Using an embedding technique proposed by Kuroiwa and Nuriya (2006), we prove well-posedness property of a class of generalized convex set-valued maps.

1. INTRODUCTION

The notion of well-posedness has been deeply studied in scalar and vector optimization. Especially, for vector optimization, two main classes of definitions have been identified in [18]. Usually a notion of well-posed vector optimization problem is said to be pointwise if it involves a single value in the solution set. Instead global notions consider the solution set as a whole.

In [19] the notion of well-posedness and sensitivity analysis have been studied in the framework of Asplund spaces. By means of coderivatives of set-valued maps, necessary and sufficient conditions for well-posedness properties for set-valued functions are proved. More recently, the notion of well-posedness has been proposed also for optimization problems with set-valued objective map. In [9] a pointwise notion has been proposed, while [21] introduces some global notion. Both papers focus on the so called set-optimization approach as introduced by Kuroiwa and Nuriya in [13], that involves ordering relations among sets.

In this paper we introduce a well-posedness notion which slightly generalizes the one in [21] and we investigate well-posedness properties of convex and generalized convex set-valued maps. We define a new class of quasiconvex set-valued maps that

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guarantees well-posedness of the set-optimization problem. The class is broader than the one of convex functions and strictly included in that of quasiconvex maps proposed by Kuroiwa in [10] and studied also in [3]. The proofs are based on the embedding approach introduced by Kuroiwa in [13], that allows to study set-optimization problems through a suitable vector optimization problem. Therefore this paper generalizes to set-valued maps the well-posedness properties of generalized convex vector functions proved in [4, 5, 6].

The paper is organized as follows. Section 2 presents the basic notations and some results on the embedding technique. Section 3 introduces the class of quasiconvex set-valued maps that is studied in Section 4 in connection with the global well-posedness of the set-optimization problem. Finally Section 5 is devoted to concluding remarks.

2. SETTING

Let $K \subseteq \mathbb{R}^m$ be a pointed closed convex cone with nonempty interior. The usual order relation in \mathbb{R}^m requires that

$$\begin{aligned} x \leq_K y & \text{ if } y - x \in K \\ x <_K y & \text{ if } y - x \in \text{int } K. \end{aligned}$$

To deal with set-optimization problems, six types of order relations among sets have been introduced in [11, 12]. Among them, in this paper, we focus only on the following notion. Let $A, B \in 2^{\mathbb{R}^m}$ be compact and convex sets. Then

$$A \leq_K^l B \text{ if } A + K \supseteq B.$$

and

$$A <_K^l B \text{ if } \exists r > 0 \text{ such that } A + K \supseteq B + r \mathcal{B}_m,$$

where $\mathcal{B}_m := \{y \in \mathbb{R}^m \mid \|y\| \leq 1\}$ is the unit ball in \mathbb{R}^m . If no confusion occurs, the subscript m is omitted and \mathcal{B} denotes the unit ball in the appropriate space. One can easily note that $A <_K^l B$ if and only if $B \subseteq A + \text{int } K$ (see e.g. [15]). We denote by K^+ the positive polar cone of K , that is the set

$$K^+ := \{l \in \mathbb{R}^m \mid \langle l, k \rangle \geq 0, \forall k \in K\}$$

Let $X \subseteq \mathbb{R}^n$ be a closed convex set. In the sequel we deal with the set-optimization problem

$$(\mathbf{P}(F, K)) \quad \min_{x \in X} F : X \subseteq \mathbb{R}^n \rightarrow 2^{\mathbb{R}^m}$$

where F is a set-valued function with $F(x)$ nonempty, compact and convex for all $x \in X$. According to the order given by K we can have different solution concepts (see e.g. [11, 12]).

Definition 1. A vector $x^0 \in X$ is a weak minimizer of $P(F, K)$ when

$$(1) \quad F(x) \not\leq_K^l F(x^0), \quad \forall x \in X.$$

The set of all weak minimizers of $P(F, K)$ is denoted by $\text{WEff}(F, X)$.
 A vector $x^0 \in X$ is a minimizer of $P(F, K)$ when

$$(2) \quad F(x) \leq_K^l F(x^0) \implies F(x^0) \leq_K^l F(x), \quad \forall x \in X$$

For the vector-valued optimization problem

$$(\text{VP}(f, K)) \quad \min_{x \in X} f : X \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^m$$

Letting $F(x) = \{f(x)\}$ for all $x \in X$, Definition 1 reduces to the classical notions of weak efficient solution and efficient solution. Therefore, if no confusion occurs on the valuedness of the objective function, we may refer to $\text{WEff}(f, X)$ as the set of all weak efficient solutions of Problem $\text{VP}(f, K)$. Throughout the paper, lower case letters are devoted to vector-valued functions and capital letters to set-valued ones. Moreover we assume that $\text{WEff}(F, X)$ is nonempty.

Let χ be the family of all compact convex subsets in \mathbb{R}^m . Following the approach in [20, 13] any two couples $(A, B); (C, D) \in \chi^2$ are equivalent if $A + D + K = B + C + K$. When this occurs we write $(A, B) \equiv (C, D)$. The equivalence family of the couple (A, B) is defined by the set

$$[A, B] := \{(C, D) \in \chi^2 \mid (A, B) \equiv (C, D)\}.$$

In [13] it has been introduced the vector space $(\chi^2 / \equiv, +, \cdot)$, where

- $[A, B] + [C, D] = [A + C, B + D];$
- $\lambda \cdot [A, B] = \begin{cases} [\lambda A, \lambda B], & \lambda \geq 0 \\ [-\lambda A, -\lambda B], & \lambda < 0 \end{cases}.$

Given a compact base W of K , the embedding space $(\chi^2 / \equiv, +, \cdot)$ is normed (see e.g. [13, 14]), introducing

$$(3) \quad \|[A, B]\| := \sup_{w \in W} |\inf \langle w, A \rangle - \inf \langle w, B \rangle|.$$

A partial order in χ^2 / \equiv can be introduced through the pointed, closed and convex cone

$$\mu(K) := \{[A, B] \in \chi^2 / \equiv \mid B \leq_K^l A\}$$

depending on the ordering cone K on \mathbb{R}^m . The interior of $\mu(K)$ is defined as

$$\text{int } \mu(K) := \{[A, B] \in \chi^2 / \equiv \mid B <_K^l A\}.$$

Therefore we can define order relations in the vector space χ^2/ \equiv by

$$[A, B] \leq_{\mu(K)} [C, D] \text{ if } [C, D] - [A, B] \in \mu(K)$$

and

$$[A, B] <_{\mu(K)} [C, D] \text{ if } [C, D] - [A, B] \in \text{int } \mu(K)$$

Problem $P(F, K)$ can be embedded into the vector optimization problem on $(\chi^2/ \equiv, +, \cdot)$

$$(\text{VP}(f, \mu(K))) \quad \min_{x \in X} f : X \rightarrow \chi^2/ \equiv$$

where, for all $x \in X$,

$$(4) \quad f(x) = (\varphi \circ F)(x) = [F(x), \{0\}].$$

In [11, 15] we find the following result.

Theorem 1. *Let $F : X \subseteq \mathbb{R}^n \rightarrow 2^{\mathbb{R}^m}$ and let $f : X \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^m$ be defined by (4). Then*

- (i) $x^0 \in X$ is a minimizer of $P(F, K)$ if and only if it is an efficient solution of $\text{VP}(f, \mu(K))$;
- (ii) $x^0 \in X$ is a weak minimizer of $P(F, K)$ if and only if it is a weak efficient solution of $\text{VP}(f, \mu(K))$.

Moreover we have also that continuity and convexity are preserved by embedding. We recall that $F : X \subseteq \mathbb{R}^n \rightarrow 2^{\mathbb{R}^m}$ is K -convex when, for all $x_1, x_2 \in X$ and $\lambda \in [0, 1]$ we have

$$(5) \quad \lambda F(x_1) + (1 - t) F(x_2) \subseteq F(\lambda x_1 + (1 - t)x_2) + K$$

In [14] it has been remarked that clearly (5) holds if and only if

$$\varphi \circ F(t x_1 + (1 - t)x_2) \leq_{\mu(K)} t(\varphi \circ F(x_1)) + (1 - t)(\varphi \circ F(x_2))$$

that is if and only if the vector valued function $f(x) = [F(x), \{0\}]$ is $\mu(K)$ -convex.

Continuity for set-valued functions has been defined e.g. in [1]. A set-valued function F is upper semicontinuous (usc) at $x^0 \in X$ if $\forall \varepsilon > 0, \exists \delta > 0$ s.t.

$$F(x) \subseteq F(x^0) + \varepsilon \mathcal{B}, \quad \forall x \in (x^0 + \delta \mathcal{B}) \cap X$$

Analogously we can define lower semicontinuity (lsc) at $x^0 \in X$ if $\forall \varepsilon > 0, \exists \delta > 0$ s.t.

$$F(x^0) \subseteq F(x) + \varepsilon \mathcal{B}, \quad \forall x \in (x^0 + \delta \mathcal{B}) \cap X$$

A set-valued function is Hausdorff continuous at $x^0 \in X$ if and only if it is both upper and lower semicontinuous.

Proposition 1. *The set-valued function $F : X \subseteq \mathbb{R}^n \rightarrow 2^{\mathbb{R}^m}$ is such that $F + K$ is Hausdorff continuous if and only if the vector valued embedded function $f = (\varphi \circ F) : X \rightarrow \chi^2 / \equiv$ is continuous.*

Proof. Assume $f = \varphi \circ F$ is continuous at $x^0 \in X$. Given a compact base W of K^+ , continuity of f at x^0 means that $\forall \varepsilon > 0, \exists \delta > 0$ such that $\|[F(x), F(x^0)]\| < \varepsilon, \forall x \in (x^0 + \delta\mathcal{B}) \cap X$. This is equivalent to

$$(6) \quad -\varepsilon < \inf \langle w, F(x) \rangle - \inf \langle w, F(x^0) \rangle < \varepsilon, \quad \forall w \in W, \quad \forall x \in (x^0 + \delta\mathcal{B}) \cap X.$$

Let $p \in \text{int } K$ and consider the compact base $W = \{l \in K^+ : \langle l, p \rangle = 1\}$. We first prove lower semicontinuity of $F + K$. Indeed (6) implies

$$\begin{aligned} & \inf \langle w, F(x) \rangle < \inf \langle w, F(x^0) \rangle + \varepsilon \\ & = \inf \langle w, F(x^0) + \varepsilon p \rangle, \forall w \in W, \forall x \in (x^0 + \delta\mathcal{B}) \cap X \end{aligned}$$

Hence, see e.g. [13]

$$\begin{aligned} & F(x^0) + \varepsilon p \subseteq F(x) + K \\ \text{i.e. } & F(x^0) \subseteq F(x) + K - \varepsilon p \subseteq F(x) + K + \varepsilon\mathcal{B} \end{aligned}$$

Therefore $F + K$ is lsc at x^0 . By a similar argument we prove upper semicontinuity and then Hausdorff continuity of $F + K$. To prove the reverse implication, let a norm in \mathbb{R}^m be defined as the Minkowski functional of the set $\tilde{\mathcal{B}} = \text{conv}((-W) \cup W)$ (every norm in \mathbb{R}^m is topologically equivalent). Hence $\forall \varepsilon > 0$, the set $\varepsilon\tilde{\mathcal{B}}$ is a neighborhood of 0. Hausdorff continuity of $F + K$ implies that $F(x^0) + K \subseteq F(x) + K + \varepsilon\tilde{\mathcal{B}}$. Then $\forall w \in W$, we have

$$\begin{aligned} \inf \langle w, F(x^0) + K \rangle & \geq \inf \langle w, F(x) + K + \varepsilon\tilde{\mathcal{B}} \rangle \\ & \geq \inf \langle w, F(x) + K \rangle + \inf \langle w, \varepsilon\tilde{\mathcal{B}} \rangle \\ & \geq \inf \langle w, F(x) + K \rangle - \varepsilon. \end{aligned}$$

Therefore $\inf \langle w, F(x) + K \rangle - \inf \langle w, F(x^0) + K \rangle \leq \varepsilon$. Analogously we can prove the inequality $\inf \langle w, F(x^0) + K \rangle - \inf \langle w, F(x) + K \rangle \geq -\varepsilon$ and therefore the continuity of f at x^0 . ■

In general, continuity of $\varphi \circ F$ does not guarantee Hausdorff continuity of F , as the following example shows.

Example 1. Let $K := [0, +\infty) \subseteq \mathbb{R}$ and $F : \mathbb{R} \rightarrow 2^{\mathbb{R}}$ be defined as

$$F(x) = \begin{cases} [-x^2, x^2] & \text{if } x \in (0, 1] \\ [0, 1] & \text{if } x = 0 \end{cases}$$

Then $\|F(x_n) - F(0)\| \rightarrow 0$ for all $x_n \downarrow 0$. Indeed, it is enough to take $W = \{1\}$ as a basis for \mathbb{R}_+ and for all $x_n \downarrow 0$ $\inf w F(x_n) = \inf F(x_n) = -(x_n)^2$, where $w = 1$. Moreover $\inf w F(0) = 0$ and hence $\|[F(x_n), F(0)]\| \rightarrow 0$, as $n \rightarrow +\infty$. However function F is not Hausdorff continuous at zero, while $F + K$ is Hausdorff continuous.

3. QUASICONVEXITY

Both for scalar and vector optimization, well-posedness is strictly related to convexity and quasi-convexity. Some notion of quasiconvexity for set-valued functions F have been provided in the literature. The following we quote from [3].

Definition 2. A set-valued function $F : X \subseteq \mathbb{R}^n \rightarrow 2^{\mathbb{R}^m}$ is said K -quasiconvex on X when the level sets $F^{-1}(y - K) := \{x \in X : y \in F(x) + K\}$ are convex for all $y \in \mathbb{R}^m$.

Quasiconvexity of a vector valued function $f : X \rightarrow \mathbb{R}^m$ is a special case of Definition 2. In this paper we propose the following alternative definition.

Definition 3. A set-valued function $F : X \subseteq \mathbb{R}^n \rightarrow 2^{\mathbb{R}^m}$ is said K -quasiconvex in the embedding sense when the vector-valued function $f : X \rightarrow \chi^2/ \equiv$ defined by (4) is $\mu(K)$ -quasiconvex.

Definition 3 can be stated also through convexity of appropriate level sets. According to the embedding technique used, a level set can be defined as the set

$$(7) \quad \{x \in X : F(x) + B + K \supseteq A\} = \left\{x \in X : F(x) + B \leq_K^l A\right\}$$

for some $A, B \subseteq \mathbb{R}^m$ compact and convex.

Proposition 2. A set-valued function $F : X \subseteq \mathbb{R}^n \rightarrow 2^{\mathbb{R}^m}$ is K -quasiconvex in the embedding sense if and only if $\forall A, B \subseteq \mathbb{R}^m$, with A, B compact and convex subsets, the level set (7) is convex.

Proof. Level sets as in (7) are equivalent to the following:

$$(8) \quad \{x \in X : [F(x), \{0\}] \leq_{\mu(K)} [A, B]\} = \{x \in X : f(x) \leq_{\mu(K)} [A, B]\}$$

where $f(x)$ is defined by (4). For any compact convex sets $A, B \in 2^{\mathbb{R}^m}$, $[A, B] \in \chi^2/ \equiv$ is an element of the image space of f . In [17], a vector-valued function is said to be quasiconvex when its level sets are convex. Since (8) are level sets for the vector-valued function $\varphi \circ F$, with respect to the ordering cone $\mu(K)$, the proof follows straightforward. ■

Remark 1. Since Definition 3 provide a characterization of quasiconvexity for set-valued maps through quasiconvexity of vector-valued ones, it is easy to see that any K -convex set-valued map is also K -quasiconvex in the embedding sense. Indeed K -convexity implies $\mu(K)$ -convexity of f , defined in (4), which, in turns, implies $\mu(K)$ -quasiconvexity of f .

The next example shows that this new class of generalized convex set-valued functions is not empty and broader than the class of K -convex set-valued functions.

Example 2. Let $X = [0, 1]$, $K = \mathbb{R}_+$ and $F : X \subseteq \mathbb{R}^n \rightarrow 2^{\mathbb{R}}$ be defined as $F(x) = [-x^2; x^2]$. The function is not K -convex. Indeed let $x_1 = 0$ and $x_2 = 1$. Therefore since $F(0) = \{0\}$ and $F(1) = [-1; 1]$, we have

$$\frac{1}{2}F(0) + \frac{1}{2}F(1) = \left[-\frac{1}{2}; \frac{1}{2}\right]$$

but $F\left(\frac{1}{2}\right) = \left[-\frac{1}{4}; \frac{1}{4}\right]$, proving that $\frac{1}{2}F(0) + \frac{1}{2}F(1) \not\subseteq F\left(\frac{1}{2}\right) + K$.

However, F is K -quasiconvex in the embedding sense. Indeed $\forall \bar{x}, \hat{x} \in X$ with $\bar{x} < \hat{x}$ it holds that $F(\bar{x}) + K \subseteq F(\hat{x}) + K$. Therefore, $\forall A, B \subseteq \mathbb{R}$, compact and convex such that $A \subseteq F(x_i) + B + K$, $i = 1, 2$, $x_1 < x_2$, it holds that $F(x_1) + K \subseteq F(\lambda x_1 + (1 - \lambda)x_2) + K$, for all $\lambda \in [0, 1]$, since $x_1 \leq \lambda x_1 + (1 - \lambda)x_2$. Hence

$$A \subseteq F(x_1) + K + B \subseteq F(\lambda x_1 + (1 - \lambda)x_2) + K + B \quad \forall \lambda \in [0, 1]$$

proving the convexity of sets (7). Finally the same function also fulfills Definition 2. Indeed, this is easily seen since $\forall x_1 < x_2 \in X$ it holds that $F(x_1) + K \subseteq F(x_2) + K$.

Proposition 3. *If $F : X \subseteq \mathbb{R}^n \rightarrow 2^{\mathbb{R}^m}$ is K -quasiconvex in the embedding sense, then F is K -quasiconvex.*

Proof. The implication easily follows assuming $A = \{y\}$ and $B = \{0\}$ in the definition of K -quasiconvexity in the embedding sense. ■

However the converse is not necessarily true, even assuming continuity, as the next example shows.

Example 3. Let $K = \mathbb{R}_+^2$ and $F : [0, 1] \subseteq \mathbb{R} \rightarrow 2^{\mathbb{R}^2}$ be defined as follows.

$$F(x) = \begin{cases} \{(x - 1/2, 1/2)\} & \text{if } x \in [0, 1/2] \\ \{(0, 1 - x)\} & \text{if } x \in (1/2, 1] \end{cases}.$$

Function F is Hausdorff continuous and K -quasiconvex, but not K -quasiconvex in the embedding sense. Indeed, if B is the segment joining points $(1/2, -1/2)$ and $(-1/2, 1/2)$, and $A = \{(-1/4, 3/4)\}$, we have $A \subseteq F(0) + B + K$ and $A \subseteq F(1) + B + K$. However $A \not\subseteq F(1/2) + B + K$.

4. GLOBAL WELL-POSEDNESS AND CONVEXITY

We recall the notion of global well-posedness for vector optimization, starting from the definition of minimizing sequence.

Definition 4. Let $p \in \text{int} K$ be given. A minimizing sequence for (w.r.t. p) $f : X \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a sequence $\{x^n\} \subseteq X$ for which $\exists \varepsilon_n \downarrow 0$ s.t. $(f(X) - f(x^n)) \cap (-\text{int} K - \varepsilon_n p) = \emptyset$.

Definition 4 does not depend on the choice of $p \in \text{int} K$. To show this we first need the following Lemma we quote from [4] without proof.

Lemma 1. Let p^1 and p^2 be vectors in $\text{int} K$. Then, there exist positive constants α and β , such that for every $\varepsilon > 0$ it holds:

$$-K - \alpha \varepsilon p^2 \subseteq -K - \varepsilon p^1 \subseteq -K - \beta \varepsilon p^2.$$

Lemma 2. Let $p^1, p^2 \in \text{int} K$. A sequence $\{x^n\} \subseteq X$ is minimizing for problem $P(F, K)$ w.r.t. p^1 if and only if it is a minimizing sequence w.r.t. p^2 .

Proof. By assumption there exists $\varepsilon_n \downarrow 0$ s.t.

$$(f(X) - f(x^n)) \cap (-\text{int} K - \varepsilon_n p^1) = \emptyset.$$

Hence, by Lemma 1 there exists $\alpha > 0$ s.t.

$$(f(X) - f(x^n)) \cap (-\text{int} K - \alpha \varepsilon_n p^2) = \emptyset$$

proving the sequence is minimizing also w.r.t. p^2 . ■

According to this notion, in [5] the following definition of global well-posedness has been introduced.

Definition 5. Problem $\text{VP}(f, K)$ is (globally) well-posed when for every minimizing sequence there exists a subsequence $\{x^{n_k}\}$ such that $\text{dist}(x^{n_k}, \text{WEff}(f, X)) \rightarrow 0$.

Motivated by the previous notions, we can introduce the following definition of (globally) minimizing sequence for problem $P(F, K)$.

Definition 6. Let $p \in \text{int} K$. A sequence $\{x^n\} \in X$ is a *minimizing sequence* for $P(F, K)$ when $\exists \varepsilon_n \downarrow 0$ s.t.

$$(9) \quad F(x^n) \not\subseteq F(x) + \text{int} K + \varepsilon_n p, \quad \forall x \in X$$

i.e.

$$F(x^n) - \varepsilon_n p \not\prec_K^l F(x), \quad \forall x \in X$$

Therefore the following definition of global well-posedness is straightforward.

Definition 7. Problem $P(F, K)$ is (globally) well-posed when every minimizing sequence $\{x^n\}$ admits a subsequence $\{x^{n_k}\}$ s.t. $\text{dist}(x^{n_k}, \text{WEff}(F, X)) \rightarrow 0$.

In [21] a first approach to extend global well-posedness to set-optimization has been proposed. Definition 7 is slightly more general since it does not require that minimizing sequences converge to some specific weak efficient solution but just that the distance between the minimizing sequence and the set $\text{WEff}(F, K)$ converges to 0.

Proposition 4. *Problem $P(F, K)$ is globally well-posed if and only if problem $VP(f, \mu(K))$ with objective function $f : X \subseteq \mathbb{R}^n \rightarrow \chi^2 / \equiv$ defined by (4), is well-posed according to Definition 5.*

Proof. In view of Theorem 1, it is enough to prove that $\{x^n\}$ is a minimizing sequence for $P(F, K)$ if and only if it is so for $VP(f, \mu(K))$.
Indeed from (9) we obtain

$$F(x^n) - \varepsilon_n p \not\subseteq F(x) + \text{int } K, \quad \forall x \in X.$$

Hence

$$[F(x), \{0\}] \not\prec_{\mu(K)} [F(x^n) - \varepsilon_n p, \{0\}], \quad \forall x \in X$$

or, equivalently, $\forall x \in X$

$$[F(x), \{0\}] \not\prec_{\mu(K)} [F(x^n), \{0\}] + [\{-\varepsilon_n p\}, \{0\}], \quad i.e.$$

$$[F(x), \{0\}] \not\subseteq [F(x^n), \{0\}] - \text{int } \mu(K) - \varepsilon_n [\{p\}, \{0\}].$$

The proof is complete, observing that $[\{p\}, \{0\}] \in \text{int } \mu(K)$.

Conversely, assume that $\{x^n\}$ is a minimizing sequence for $VP(f, \mu(K))$. Then, for some $[P, Q] \in \text{int } \mu(K)$ we have

$$(\varphi \circ F)(x) \not\prec_{\mu(K)} (\varphi \circ F)(x^n) - \varepsilon_n [P, Q], \quad \forall x \in X$$

By Lemma 2, we can choose $[P, Q] = [\{p\}, \{0\}] \in \text{int } \mu(K)$, with $p \in \text{int } K$. Hence, for all $x \in X$ we have

$$[F(x), \{0\}] \not\subseteq [F(x^n), \{0\}] - \text{int } \mu(K) - \varepsilon_n [\{p\}, \{0\}]$$

from which the conclusion easily follows. ■

The embedding technique allows us to prove Theorem 2 below, that relates well-posedness to quasiconvexity. In order to prove it, we need the following result from [5].

Theorem 1. *Let $f : X \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a continuous and K -quasiconvex function. Moreover assume that, for every $y \in \mathbb{R}^m$, the level set*

$$f^{-1}(y - K) := \{x \in X \mid y \in f(x) + K\}$$

is bounded and $\text{WEff}(f, X)$ is bounded. Then problem $\text{VP}(f, K)$ is globally well-posed.

Theorem 2. *Let $F : X \subseteq \mathbb{R}^n \rightarrow 2^{\mathbb{R}^m}$ be K -quasiconvex in the embedding sense and such that $F + K$ is Hausdorff continuous. Moreover, assume that, for every $A, B \subseteq \mathbb{R}^m$ compact and convex, the set $\{x \in X \mid A \subseteq F(x) + B + K\}$ is bounded and $\text{WEff}(F, X)$ is nonempty and bounded. Then problem $P(F, K)$ is globally well-posed.*

Proof. Embedding problem $P(F, K)$ in the vector valued problem $\text{VP}(f, \mu(K))$ we obtain from (4) the objective function $f(x) = [F(x), \{0\}]$. According to Propositions 1 and 2, f is continuous and $\mu(K)$ -quasiconvex. Moreover, level sets of f , w.r.t. $\mu(K)$, that is the sets $\{x \in X : y \in f(x) + \mu(K)\}$, coincide with sets $\{x \in X \mid A \subseteq F(x) + B + K\}$, for all $y = [A, B] \in \chi^2 / \equiv$ and, therefore are bounded. Finally, Theorem 1 guarantees that $\text{WEff}(f, X) = \text{WEff}(F, X)$ and hence it is bounded. We can finally apply Theorem 1 to problem $\text{VP}(f, \mu(K))$ to get the thesis. ■

In [6] it has been proved that Theorem 1 holds with weaker assumptions for K -convex vector-valued functions.

Theorem 3. *Let $f : X \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a K -convex function and assume $\text{WEff}(f, X)$ is nonempty and bounded. Then problem $\text{VP}(f, \mu(K))$ is well-posed.*

Corollary 1. *Let $F : X \subseteq \mathbb{R}^n \rightarrow 2^{\mathbb{R}^m}$ be K -convex. Moreover assume that $\text{WEff}(F, X)$ is nonempty and bounded. Then problem $P(F, K)$ is globally well-posed.*

Proof. Function f defined by (4) is $\mu(K)$ -convex, according to Proposition 1. The set $\text{WEff}(\varphi \circ F, X) = \text{WEff}(F, X)$ is bounded. Hence we can apply Theorem 3 to problem $\text{VP}(f, \mu(K))$ to prove the thesis. ■

5. CONCLUDING REMARKS

In this paper we introduced a notion of global well-posedness for set-optimization that most intuitively extend the notion introduced in [4] for vector optimization. The definition is motivated by the embedding technique popularized by [13], that allows to study problem $P(F, K)$ through an equivalent vector-valued problem in the embedding space.

We proved that a certain class of generalized convex set-valued functions implies well-posedness of the set-optimization problem. This class, broader than the class of K -convex functions, is not equivalent to the class of K -quasiconvex set-valued function studied in [3]. We leave as an open question whether Theorem 2 can be extended to this larger class.

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