

Optimality conditions for set-valued optimization problems via scalarization function

Ha Manh Linh^{1,*}, Nguyen Dinh Huy², Nguyen Thi Thanh Truc¹



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ABSTRACT

One of the most important and popular topics in optimization problems is to find its optimal solutions, especially Pareto optimal points, a well-known solution introduced in multi-objective optimization. This topic is one of the oldest challenges in many issues related to science, engineering and other fields. Many important practical-problems in science and engineering can be expressed in terms of multi-objective/ set-valued optimization problems in order to achieve the proper results/ properties. To find the Pareto solutions, a corresponding scalarization problem has been established and studied. The relationships between the primal problem and its scalarization one should be investigated for finding optimal solutions. It can be shown that, under some suitable conditions, the solutions of the corresponding scalarization problem have uniform spread and have a close relationship to Pareto optimal solutions for the primal one. Scalarization has played an essential role in studying not only numerical methods but also duality theory. It can be usefully applied to get relationships/ important results between other fields, for example optimization, convex analysis and functional analysis. In scalarization, we usually use a kind of scalarized-functions. One of the first and the most popular scalarized-functions used in scalarization method is the Gerstewitz function. In the paper, we mention some problems in set-valued optimization. Then, we propose an application of the Gerstewitz function to these problems. In detail, we establish some optimality conditions for Pareto/ weak solutions of unconstrained/ constrained set-valued optimization problems by using the Gerstewitz function. The study includes the consideration of problems in theoretical approach. Some examples are given to illustrate the obtained results.

Key words: Pareto efficient solution, weak efficient solution, set-valued optimization, Gerstewitz function, optimality condition

INTRODUCTION

Scalarization has an essential role in studying numerical methods and duality theory¹⁻⁴. It can be applied to get relationships between other fields, such as: optimization, convex analysis and functional analysis. Solutions of vector optimization problems can be characterized by those of corresponding scalarized optimization problems.

In scalarization, the Gerstewitz function plays an important role. Its main properties were studied in some papers⁵⁻⁷.

In the paper, we have proposed optimality conditions for set-valued optimization problems using the Gerstewitz function. These results have contributed to applications of the Gerstewitz function in optimization.

RELIMINARIES

In the paper, Let X and Y be normed spaces, K be a pointed, closed, convex cone with nonempty interior in Y . For A be a nonempty subset in Y , we denote

$int(A)$, $cl(A)$ and $cone(A)$ for the interior, the closure of A and the cone generalized by A , respectively.

Let $F := X \rightarrow 2^Y$ be a set-valued map from X to Y , the domain, the image and the graph of F are defined by $dom(F) := \{x \in X | F(x) \neq \emptyset\}$; $im(F) := \{y \in Y | y \in F(dom(F))\}$
 $gr(F) := \{(x, y) \in X \times Y | y \in F(x)\}$.

Definition 2.1. Let $F := X \rightarrow 2^Y$ and $(x_0, y_0) \in gr(F)$.

(i) A point (x_0, y_0) is called a Pareto efficient solution of F on X if $(F(X) - y_0) \cap (-K \setminus \{0\}) = \emptyset$. The set of Pareto efficient solutions of F is denoted by $Min_K F(X)$.

(ii) A point (x_0, y_0) is called a weak efficient solution of F on X if $(F(X) - y_0) \cap (-intK) = \emptyset$. The set of weak efficient solutions of F is denoted by $WMin_K F(X)$.

Note that $Min_K F(X)$ is a subset of $WMin_K F(X)$. In general, the inverse conclusion is not true by the following example.

Example 2.2. Let $X = \mathbb{R}, Y = \mathbb{R}^2, k = \mathbb{R}_+^2, F_1, F_2 : X \rightarrow 2^Y$ be defined by

¹University of Information Technology, VNU-HCM

²University of Technology, VNU-HCM

Correspondence

Ha Manh Linh, University of Information Technology, VNU-HCM

Email: hamanhlinh2002@gmail.com

History

- Received: 09-12-2019
- Accepted: 17-12-2020
- Published: 31-12-2020

DOI : 10.32508/stdjet.v3iSI3.642



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Cite this article : Linh H M, Huy N D, Truc N T T. Optimality conditions for set-valued optimization problems via scalarization function. Sci. Tech. Dev. J. – Engineering and Technology; 3(SI3):SI150-SI153.

$F_1(x) := \{(y_1, y_2 \in Y | y_2 \geq y_1^2)\};$
 $F_2(x) := \{(y_1, y_2 \in Y | y_2 \geq y_1^2, y_1 \leq 0)\} \cup$
 $\{(y_1, y_2) \in Y | y_2 \geq 0, y_1 \geq 0\}.$
 With F_1 , one has $\text{Min}_K F_1(X) = \text{Wmin}_K F_1(X) = \{(x(y_1, y_2)) \in X \times Y | y_2 = y_1^2, y_1 \leq 0\}.$
 With F_2 , we get
 $\text{Min}_K F_2(X) = \{(x(y_1, y_2)) \in X \times Y | y_2 = y_1^2, y_1 \leq 0\},$
 $\text{Wmin}_K F_2(X) = \{(x(y_1, y_2)) \in X \times Y | y_2 = y_1^2, y_1 \leq 0\} \cup$
 $\{(x(y_1, y_2)) \in X \times Y | y_2 = 0, y_1 \geq 0\}.$ Therefore,
 $\text{Min}_K F_2(X)$ is a subset of $\text{Wmin}_K F_2(X).$

Definition 2.3.⁶ With $e \in \text{int}(K)$, the Gerstewitz function $h_K^e(y) : Y \rightarrow \mathbb{R}$ is defined by $h_K^e(y) := \inf \{t \in \mathbb{R} | y \in te - K\}.$

If K is a pointed, closed, convex cone with nonempty interior then h_K^e is finite and we get $h_K^e(y) = \sup \{h(y) | h \in K^*, h(e) = 1\}, \forall y \in Y,$ where $K^* := \{k^* \in Y^* | \langle k^*, k \rangle \geq 0, \forall k \in K\}.$ It is a convex and continuous function on $Y.$ We recall some properties of the Gerstewitz function as follows.

Proposition 2.4.^{6,7} With $e \in \text{int}(K)$, we have

- (i) $(-\infty, 0]e - K = \{y \in Y | h_K^e(y) \leq 0\}.$
- (ii) $(-\infty, 0)e - K = \{y \in Y | h_K^e(y) < 0\}.$
- (iii) $-K \setminus ((-\infty, 0)e - K) = \{y \in Y | h_K^e(y) = 0\}.$
- (iv) $Y \setminus ((-\infty, 0)e - K) = \{y \in Y | h_K^e(y) \geq 0\}.$
- (v) $Y \setminus ((-\infty, 0]e - K) = \{y \in Y | h_K^e(y) > 0\}.$
- (vi) $\forall \alpha > 0, h_K^e(\alpha y) = \alpha h_K^e(y).$
- (vii) $h_K^e(y_1 + y_2) \leq h_K^e(y_1) + h_K^e(y_2).$

For applications of the Gerstewitz function, the readers are referred to the references⁶⁻⁹.

OPTIMALITY CONDITIONS

Let X, Y, Z be normed spaces, K and D be pointed, closed, convex cones with nonempty interior in Y and $Z,$ respectively, $F : X \rightarrow 2^Y, G : X \rightarrow 2^Z.$ We consider two optimization problems as follows

$$(P_1) = \begin{cases} \min_K F(x), \\ \text{s.t.}, x \in X; \end{cases}$$

$$(P_2) = \begin{cases} \min_K F(x), \\ \text{s.t.}, x \in X, G(x) \cap (-D) \neq \emptyset. \end{cases}$$

Let S_i be feasible sets of $(P_i), i=1,2,$ then

$$S_1 = X \text{ and } S_2 = \{x \in X | G(x) \cap (-D) \neq \emptyset\}$$

A point $(x_0, y_0) \in \text{gr}(F)$ is called a Pareto efficient solution (a weak efficient solution) of $(P_i) (i=1,2)$ if $x_0 \in S_i$ and

$$(F(S_i) - y_0) \cap (-K \setminus \{0\}) = \emptyset.$$

$$(F(S_i) - y_0) \cap (-\text{int}K) = \emptyset.$$

Theorem 3.1. Let $(x_0, y_0) \in \text{gr}(F)$ with $x_0 \in S_1.$ Then, (x_0, y_0) is a weak efficient solution of (P_1) if and only if there exists $\text{int}(K)$ such that $\forall y \in F(S_1),$

$$h_K^e(y - y_0) \geq 0. \tag{1}$$

Proof. "If:" Suppose that there exists $e \in \text{int}(K)$ such that $h_K^e(y - y_0) \geq 0, \forall y \in F(S_1).$ Let $v \in F(S_1) - y_0,$ then $y \in F(S_1)$ there is with $v = y - y_0.$ Because (1) is fulfilled, i.e., $h_K^e(v) \geq 0,$ by Proposition 2.4(iv) we obtain $v \in Y \setminus ((-\infty, 0)e - K),$ thus $v \in Y \setminus (-\text{int}K).$ Consequently, one gets

$$(F(S_1) - y_0) \cap (-\text{int}K) = \emptyset.$$

Therefore, (x_0, y_0) a weak efficient solution of $(P_1).$

"Only if:" Let (x_0, y_0) be a weak efficient solution of $(P_1),$ suppose that (1) does not hold, i.e., there exist $x \in S_1, y \in F(S_1)$ with $h_K^e(y - y_0) < 0$ By Proposition 2.4(ii), we get $y - y_0 \in (-\infty, 0)e - K = -\text{int}K,$ which contradicts to the weak efficiency of $(x_0, y_0).$ ■

Theorem 3.2. Let $(x_0, y_0) \in \text{gr}(F)$ with $x_0 \in S_1.$

(i) If (x_0, y_0) is a Pareto efficient solution of (P_1) then there exists $e \in \text{int}(K)$ such that (1) holds $\forall y \in F(S_1).$

(ii) If there exists $e \in \text{int}(K)$ such that (1) holds with strict inequality $\forall y \in F(S_1) \setminus \{y_0\}$ then (x_0, y_0) is a Pareto efficient solution of $(P_1).$

Proof. (i) Since (x_0, y_0) is a Pareto efficient solution of $(P_1), (x_0, y_0)$ is a weak efficient solution of $(P_1).$ It follows from Theorem 3.1 that we are done.

(ii) Suppose that there is $e \in \text{int}(K)$ with $h_K^e(y - y_0) > 0, \forall y \in F(S_1) \setminus \{y_0\}.$ Let $v \in F(S_1) - y_0,$ then there exists $y \in F(S_1)$ such that $v = y - y_0.$ Since (1) holds with strict inequality, i.e., $h_K^e(v) > 0$ by Proposition 2.4(v), we get $v \in Y \setminus ((-\infty, 0)e - K),$ so $v \in Y \setminus (-K).$

Therefore, we get

$$(F(S_1) - y_0) \cap (-K \setminus \{0\}) = \emptyset.$$

Hence, (x_0, y_0) is a Pareto efficient solution of $(P_1).$ ■

With the problem $(P_2),$ we can obtain optimality conditions similar to those in Theorem 3.1 and 3.2 replacing S_1 by $S_2.$ However, since (P_2) is a constrained optimization problem, we propose other results for this problem.

Let $O(Z, Y)$ be a set of continuous maps from Z to $Y,$ and

$$\Pi := \{T \in O(Z, Y) | T(D) \subseteq K\}$$

We define the map $L : X \times \Pi \rightarrow 2^Y$ by $L(x, T) := F(x) + T(G(x)).$

Theorem 3.3. Let $(x_0, y_0) \in \text{gr}(F)$ with $x_0 \in S_2$ and $z_0 \in G(x) \cap (-D).$ Then, (x_0, y_0) is a weak efficient solution of (P_2) if and only if there exist $e \in \text{int}(K), T \in \Pi$ such that $\forall x \in X, y \in L(x, T),$

$$h_K^e(y - y_0) \geq 0. \tag{2}$$

Proof. "If:" Since (2) holds, by Proposition 2.4(iv), one gets $L(x, T) - y_0 \subseteq Y \setminus ((-\infty, 0)e - K) = Y \setminus (-\text{int}K),$ or

$$F(x) + T(G(x)) - y_0 \subseteq Y \setminus (-\text{int}K) \tag{3}$$

Suppose that (x_0, y_0) is not a weak efficient solution of (P_2) , there exist $x \in \tilde{X}, z \in \tilde{G}(x) \cap (-\tilde{D}), y \in \tilde{F}(x)$ such that $y - y_0 \in -\text{int}K$. Therefore, $y + \tilde{T}(z) - y_0 \in -\text{int}K - K = -\text{int}K$, which contradicts to (3). Hence, (x_0, y_0) is a weak efficient solution of (P_2) .

“Only if:” Suppose that (2) does not hold, i.e., for all $e \in \text{int}(K), T \in \Pi$ one has $x \in X$ such that $h_K^e(L(x, T) - y_0) < 0$. By Proposition 2.4(ii), we obtain

$$L(x, T) - y_0 \subseteq (-\infty, 0)e - K \subseteq -\text{int}K$$

It means that $F(x) + T(G(x)) - y_0 \subseteq -\text{int}K$. For x with $\tilde{G}(x) \cap (-D) \neq \emptyset$, there is $z \in \tilde{G}(x) \cap (-\tilde{D}), y \in \tilde{F}(x)$ such that $y - y_0 \in -\text{int}K$, which is a contradiction. Hence, (2) is fulfilled. ■

Theorem 3.4. Let $(x_0, y_0) \in \text{gr}(F)$ with $x_0 \in S_2$ and $z_0 \in G(x_0) \cap (-D)$.

(i) If (x_0, y_0) is a Pareto efficient solution of (P_2) then there exist $e \in \text{int}(K)$, such that (2) holds $\forall x \in X, y \in L(x, T)$.

(ii) If there exist $e \in \text{int}(K), T \in \Pi$ such that (2) holds with strict inequality $\forall x \in X, y \in L(x, T) \setminus \{y_0\}$ then (x_0, y_0) is a Pareto efficient solution of (P_2) .

Proof. (i) Since (x_0, y_0) is a Pareto efficient solution of (P_2) , (x_0, y_0) is a weak efficient solution of (P_2) . By Theorem 3.3, we are done.

(ii) Since (2) holds with strict inequality, by Proposition 2.4(v), one gets $L(x, T) - y_0 \subseteq ((-\infty, 0]e - K) = Y \setminus (-K)$, or (4).

$$F(x) + T(G(x)) - y_0 \subseteq Y \setminus (-K). \tag{4}$$

Suppose that (x_0, y_0) is not a Pareto efficient solution of (P_2) , there exist $x \in \tilde{X}, z \in \tilde{G}(x) \cap (-\tilde{D}), y \in \tilde{F}(x)$ such that $y - y_0 \in -K \setminus \{0\}$. Therefore, $y + \tilde{T}(z) - y_0 \in -K \setminus \{0\} - K \subseteq -K$, which contradicts to (4). Hence, (x_0, y_0) is a Pareto efficient solution of (P_2) . ■

To illustrate Theorem 3.3, we consider the following example.

Example 3.5. Let $X = Y = Z = \mathbb{R}, K = D = \mathbb{R}_+, F : X \rightarrow 2^Y, G : X \rightarrow 2^Z$ be defined by $F(x) := \{y \in Y | 0 \leq y \leq x^2\}; G(x) := \{z \in Z | 0 \leq z \leq |x|\}$.

With (P_2) , one has $\{x \in X | G(x) \cap (-D) \neq \emptyset\} = \mathbb{R}$. It is easy to see that $(x_0, y_0) = (0, 0)$ is a weak efficient solution of (P_2) . We now check that condition (2) holds. In fact, with $e=1, T(x) = x$ is a linear operator satisfying $T(D) \subseteq K$, then

$$L(x, T) := F(x) + T(G(x)) = [0, x^2] + [0, |x|] = [0, x^2 + |x|].$$

For all $y \in L(x, T), x \in X$, one gets

$$\begin{aligned} h_K^e(y - y_0) &:= \inf \{t \in \mathbb{R} | y \in te - K\} \\ &= \inf \{t \in \mathbb{R} | [0, x^2 + |x|] \in t - K\} \\ &= x^2 + |x| \geq 0. \end{aligned}$$

CONCLUSIONS

In the paper, we first recall the Gerstewitz scalar function and its basic properties. Via this function, optimality conditions for some kinds of optimization problems are established concerning Pareto efficient/weak efficient solutions.

In set-valued optimization, we have several kinds of solutions, such as: Geoffrion efficient solution¹⁰, Borwein efficient solution¹¹, Benson efficient solution¹² ... They have been studied in some recent results^{13,14}. Thus, for the possible development, we think that giving optimality conditions for the above-mentioned solutions may be a promising approach.

AUTHOR S' CONTRIBUTION

All authors contributed equally to this work. All authors have read and agreed to the published version of the manuscript.

CONFLICT OF INTEREST

We declare that there is no conflict of whatsoever involved in publishing this research.

ACKNOWLEDGMENT

This research is funded by Vietnam National University of Hochiminh City (VNU HCMC) under grant number C2019-26-03.

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