# On Optimization of Sub-differentiable Lower Semi-continuous Pseudo convex Functions on Banach Spaces Using Variational Inequalities

## Ugochukwu Anulobi Osisiogu<sup>1</sup>, Akachukwu Agashi Offia<sup>2\*</sup>, and Theresa Ebele Efor<sup>1</sup>

<sup>1</sup>Department of Mathematics and Computer Science, Ebonyi State University, Abakaliki, Nigeria <sup>2</sup>Department of Mathematics and Statistics, Alex Ekwueme Federal University, Ndufu-Alike, Nigeria <sup>1</sup>Department of Mathematics and Computer Science, Ebonyi State University, Abakaliki, Nigeria

## Abstract

In this work, we extend the relationships between convex functions and corresponding monotone maps to pseudoconvexity and the corresponding pseudo-monotonicity of their sub-differentiable maps. We characterize the lower semi-continuous Clarke-Rockafeller sub-differentiable pseudo-convex functions  $f: C \subseteq X \to \mathbb{R} \cup \{+\infty\}$  on convex subset *C* of infinite-dimensional real Banach spaces *X* with respect to the corresponding monotonicity of their Clarke-Rockafeller sub-differential operators  $\partial f$  and obtain the optimality conditions, using variational inequalities.

**Keywords:** Banach Spaces, Sub-differentials, Lower Semi-continuity, Pseudo-convexity, Pseudo-monotonicity, Optimality Conditions and Variational Inequalities.

## 1. Introduction

Convex optimization, which studies the problem of minimizing convex functions over convex sets, and a subfield of mathematical optimization, plays an important role in many branches of applied mathematics. The foremost reason is that; it is very suitable to extremum problems. For instance, some necessary conditions for the existence of a minimum also become sufficient in the in terms of convexity. And convex optimization can be a smooth or a non-smooth convex optimization. Since the concept of convexity does not satisfy some mathematical models, various generalizations of convexity such as quasi-convexity and pseudo-convexity, which retain some important properties of convexity and equally provide a better representation of reality were introduced in the literature to fill these gaps.

While the quasi-convexity property of a function guarantees the convexity of their sublevel sets, the pseudo-convexity property implies that the critical points are minimizers [25]. One of the features of convexity of functions is the relationship it has with the monotonicity of some maps. For example, a differentiable function is said to be convex if and only if its gradient is a monotone map. In non-smooth analysis, the generalized convexity of functions can be equally characterized in terms of the generalized monotonicity of their related operators, [12].

The concepts of pseudo-convexity, traced to [30], within his research on analytical functions and independently introduced into the field of optimization by [32], has many applications in mathematical programming and economic problems [20, 22, 29]. And pseudo-monotonicity, introduced by [24] as a generalization of monotone operators has been used to describe a property of consumer's demand correspondence [18]. Although the simplest class of pseudo-monotone operators consists of gradients of pseudo-convex functions, there are some monotone operators that are not sub-differentials, [18]. And

generalized monotonicity of maps is frequently used in complementarity problems, equilibrium problems and variational inequalities [17].

Variational inequalities, formulated between late 60s and early 70s by an Italian Mathematician, Stampacchia, solve a wide range of problems in mathematical optimization, operations research, economic equilibrium problems and engineering sciences [11, 15, 37]. Results on relations of variational inequalities with differentiable optimization problems also show that the Stampacchia Variational Inequality (SVI) is a necessary condition for optimality, while the Minty Variational Inequality (MVI) is a sufficient optimality condition and similar result on generalizations of SVI and MVI to multivalued operators for non-smooth optimization problems exists [9].

Consider an optimization problem

*minimize* f(x), subject to  $x \in C$ .

(1)

where  $f: C \subseteq X \to \mathbb{R} \cup \{+\infty\}$  is a lower semi-continuous (l.s.c.) Clarke-Rockafeller sub-differentiable function on as subset *C* of a real Banach space *X*. We characterize (1) by the corresponding monotonicity of their Clarke-Rockafeller sub-differential operator  $\partial f$  and investigate their optimality conditions, using variational inequalities.

## 2. Preliminaries

Let X be a real Banach space with norm  $\|.\|$ ,  $X^*$  be its topological dual and  $\langle x^*, x \rangle$  be the duality pairing between  $x \in X$  and  $x^* \in X^*$ . We denote the closed segment  $[x, y] = \{\lambda x + (1 - \lambda)y: \lambda \in [0,1]\}$  for  $x, y \in X$ , and define (x, y], [x, y) and (x, y) similarly.

**Definition 2.1** [33] Let  $f: X \to \mathbb{R} \cup \{+\infty\}$  be an extended real valued function, the effective domain is defined by

 $\operatorname{dom}(f) = \{ x \in X \colon f(x) < +\infty \}.$ 

**Definition 2.2** [5] A function  $f: X \to \mathbb{R} \cup \{+\infty\}$  is said to be lower semi-continuous at  $x \in X$  if and only if:  $\forall \lambda \in \mathbb{R}$ , such that  $\lambda < f(x), \exists V \subset U(x): \lambda < f(y) \forall y \in V$ .

**Definition 2.3** [3, 29] A lower semi-continuous function  $f: X \to \mathbb{R} \cup \{+\infty\}$  is said to be quasi-convex, if for any  $x, y \in X$  and  $z \in [x, y]$  we have

 $f(z) \le \max\{f(x), f(y)\}.$ 

(2)

**Definition 2.4 (See** [3, 21, 29]) A lower semi-continuous function  $f: X \to \mathbb{R} \cup \{+\infty\}$  is said to be strictly quasiconvex, if the inequality (2) is strict when  $x \neq y$ .

**Definition 2.5 (See** [12]) Let  $T: X \to X^*$  be a multivalued operator with domain  $D(T) = \{x \in X: T(x) \neq \emptyset\}$ . *T* is said to be quasi-monotone if for any  $x, y \in X$ ,  $x^* \in T$  and  $y^* \in T(y)$ , we have

 $\langle x^*, y-x \rangle > 0 \Longrightarrow \langle y^*, y-x \rangle \ge 0.$ 

**Definition 2.6 (See** [12]) Let  $T: X \to X^*$  be a multivalued operator with domain  $D(T) = \{x \in X: T(x) \neq \emptyset\}$ . *T* is said to be pseudo-monotone if for any  $x, y \in X$ ,  $x^* \in T$  and  $y^* \in T(y)$ , we have

 $\langle x^*, y - x \rangle \ge 0 \Longrightarrow \langle y^*, y - x \rangle \ge 0.$ 

(3)

**Definition 2.7 (See** [29]) Let  $T: X \to X^*$  be a multivalued operator with domain  $D(T) = \{x \in X: T(x) \neq \emptyset\}$ . *T* is said to be strictly pseudo-monotone if for any different two points  $x, y \in X, x^* \in T$  and  $y^* \in T(y)$ , we have

 $\langle x^*, y - x \rangle \ge 0 \Longrightarrow \langle y^*, y - x \rangle > 0.$  (4) **Definition 2.8 (See** [2]) An operator  $\partial$  that associates to any lower semi-continuous function  $f: X \to \mathbb{R} \cup \{+\infty\}$  and a point  $x \in X$  a subset  $\partial f(x)$  of  $X^*$  is a sub-differential if it satisfies the following properties:

- (i)  $\partial f(x) = \{x^* \in X^*: \langle x^*, y x \rangle + f(x) \le f(y), \forall y \in X\}$ , whenever f is convex;
- (ii)  $0 \in \partial f(x)$ , whenever  $x \in dom f$  is a local minimum of f;
- (iii)  $\partial (f + g)(x) \subset \partial f(x) + \partial g(x)$ , whenever g is a real a real-valued convex continuous function which is  $\partial$ -differentiable at x,

where g-differentiable at x means that both  $\partial g(x)$  and  $\partial (-g)(x)$  are non-empty. We say that f is  $\partial$ -differentiable at x when  $\partial f(x)$  is non-empty while  $\partial f(x)$  are called the sub-gradients of f at x.

**Definition 2.9 (See** [2]) The Clarke-Rockafellar generalized directional derivative of f at  $x_0 \in \text{dom}(f)$  in the direction  $d \in X$  is given by

$$f^{\uparrow}(x_0, d) = \sup_{\varepsilon > 0} \operatorname{limsup}_{\substack{x \to f^{x_0} \\ \lambda > 0}} \operatorname{inf}_{d' \in B_{\varepsilon}(d)} \frac{f(x + \lambda d') - f(x)}{\lambda},$$

(5)

where  $B_{\varepsilon}(d) = \{d' \in X : ||d' - d|| < \varepsilon\}, \lambda > 0$  indicates the fact that  $\lambda > 0$  and  $\lambda \to 0$ , and  $x \to f^{x_0}$  means that both  $x \to x_0$  and  $f(x) \to f(x_0)$ ; While,

**Definition 2.10 (See** [3]) The Clarke-Rockafellar subdifferential of f at  $x_0$  is defined by

$$\partial f(x_0) = \{x^* \in X^* : (x^*, d) \le f^{\uparrow}(x_0, d), \forall d \in X\};$$

if  $x_0 \in X \setminus \text{dom}(f)$ , then

 $\partial f(x_0) = \emptyset, ([29]).$ 

**Definition 2.11** (See [12, 29]) A lower semi-continuous function  $f: X \to \mathbb{R} \cup \{+\infty\}$  is said to be quasi-convex (with respect to Clarke-RockerfellerSubdifferentials) if for any  $x, y \in X$ ,

 $\exists x^* \in \partial f(x) \colon \langle x^*, y - x \rangle > 0 \Longrightarrow \forall z \in [x, y], \ f(z) \le f(y).$ (7)

**Definition 2.12 (See**[12, 29]) A lower semi-continuous function  $f: X \to \mathbb{R} \cup \{+\infty\}$  is said to be pseudo-convex (with respect to Clarke-Rockerfeller Subdifferentials) if for any  $x, y \in X$ :

 $\exists x^* \in \partial f(x) \colon \langle x^*, y - x \rangle \ge 0 \Longrightarrow f(x) \le f(y).$ 

(8)

**Definition 2.13 (See** [3, 21, 29]) A lower semi-continuous function  $f: X \to \mathbb{R} \cup \{+\infty\}$  is said to be strictly pseudo-convex (with respect to Clarke-Rockerfeller Subdifferentials) if for any two different points  $x, y \in X$ :  $\exists x^* \in \partial f(x): \langle x^*, y - x \rangle \ge 0 \Longrightarrow f(x) < f(y)$ , when  $x \neq y$ .

(9)

**Definition 2.14** [29] A lower semi-continuous function  $f: X \to \mathbb{R} \cup \{+\infty\}$  is said to be radially continuous if for all  $x, y \in X$ , f is continuous on [x, y].

**Definition 2.15** [29] A function  $f: X \to \mathbb{R} \cup \{+\infty\}$  is said to be radially non-constant if for all  $x, y \in X$ , with  $x \neq y$ ,  $f \not\equiv$  constant on [x, y].

**Definition 2.16** A sub-differential operator  $\partial f$  is said to satisfy the Stampacchia variational inequality (10) at  $x \in C$  if:

 $\langle x^*, y - x \rangle \ge 0 \forall y \in C, \forall x^* \in \partial f(x).$ 

(10)

**Definition 2.17** While a sub-differential operator  $\partial f$  satisfies the Minty variational inequality (11) at *x*, if:  $\forall y \in C, \langle y^*, y - x \rangle \leq 0, \forall y^* \in \partial f(y).$ 

(11)

**Theorem 2.1** (Approximate mean value inequality). Let  $f: X \to \mathbb{R} \cup \{+\infty\}$  be a Clarke-Rockafeller subdifferentiable lower semi-continuous (l.s.c.) function on a Banach space *X*. Let  $a, b \in X$  with  $a \in \text{dom } f$  and  $a \neq b$ . Let  $\rho \in \mathbb{R}$  be such that  $\rho \leq f(b)$ . Then, there exist  $c \in [a, b)$  and  $x_n \to f^c$  and  $x_n^* \in \partial f(x_n)$  such that

(i)  $\liminf_{n \to +\infty} \langle x_n^*, c - x_n \rangle \ge 0;$ (ii)  $\liminf_{n \to +\infty} \langle x_n^*, b - a \rangle \ge \rho - f(a).$ 

**Proof.** (See [2]).

**Lemma 2.2** Let  $f: X \to \mathbb{R} \cup \{+\infty\}$  be a Clarke-Rockafeller sub-differentiable lower semi-continuous (l.s.c.) function on a Banach space X. Let  $a, b \in X$  with f(a) < f(b). Then, there exist  $c \in [a, b)$ , and two sequences  $c_n \to c$ , and  $c_n^* \in \partial f(c_n)$  with

 $\langle c_n^*, x - c_n \rangle > 0$  for every  $x = c + \lambda(b - a)$  with  $\lambda > 0$ .

**Proof.** By Theorem 2.1, there exists an  $x_0 \in [a, b)$  and a sequence  $x_n \to f^c$  and  $x_n^* \in \partial f(x_n)$  verifying  $\liminf_{n \to +\infty} \langle x_n^*, c - x_n \rangle \ge 0$  and  $\liminf_{n \to +\infty} \langle x_n^*, b - a \rangle > 0$ .

(12)

Putting  $x = c + \lambda(b - a)$  with  $\lambda > 0$  it holds  $\langle x_n^*, x - x_n \rangle = \langle x_n^*, c - x_n \rangle + \lambda \langle x_n^*, b - a \rangle > 0$ 

(13)

for *n* very large.

## 3. Relationship between Generalized Convexity and Generalized Monotonicity

**Theorem 3.1** Let  $f: X \to \mathbb{R} \cup \{+\infty\}$  be a lower semi-continuous (l.s.c.) Clarke-Rockafeller subdifferentiable function on a Banach space *X*. Then, *f* is quasi-convex if and only if  $\partial f$  is quasi-monotone.

**Proof.** We show that if *f* is not quasi-convex, then  $\partial f$  is not quasi-monotone.

Suppose that there exist some x, y, z in X with  $z \in [x, y]$  and  $f(z) > \max\{f(x), f(y)\}$ . According to Lemma 2.2 applied with a = x and b = z, there exists a sequence  $y_n \in \text{dom}\partial f$  and  $y_n^* \in \partial f(y_n)$  such that

 $y_n \to \bar{y} \in [x, z], \ \bar{y} \neq z \text{ and } \langle y_n^*, y - y_n \rangle > 0.$  (14) Let  $0 < \lambda \le 1$  be such that  $z = \bar{y} + \lambda(y - \bar{y})$  and set  $z_n = y_n + \lambda(y - y_n)$ , so that  $z_n \to z$ . Since f is lower semi-continuous, we may pick  $n \in \mathbb{N}$  very large with  $f(z_n) > f(y)$ . Apply Lemma 2.2 again with a = y and  $b = z_n$  to find sequences  $x_k \in \text{dom}\partial f, x_k^* \in \partial f(x_k)$  such that

 $x_k \to \bar{x} \in [y, z_n], \ \bar{x} \neq z_n \text{ and } \langle x_k^*, y_n - x_k \rangle > 0.$ In particular,  $\bar{x} \neq y_n$  and (15)

$$\langle y_n^*, \bar{x} - y_n \rangle = \frac{\|\bar{x} - y_n\|}{\|y - y_n\|} \langle y_n^*, y - y_n \rangle > 0;$$
(16)

hence,  $\langle y_n^*, x_k - y_n \rangle > 0$  for k sufficiently large. But  $\langle y_n^*, y_n - x_k \rangle > 0$ , showing that  $\partial f$  is not quasi-monotone.

Conversely, we suppose that f is quasi-convex and show that  $\partial f$  is quasi-monotone. Let  $x^* \in \partial f(x)$ and  $y^* \in \partial f(y)$  with  $\langle x^*, y - x \rangle > 0$ . We need to verify that  $f^{\uparrow}(y, x - y) \leq 0$ . We fix  $\varepsilon > 0$  and  $\omega \in (0, \varepsilon)$ such that

 $\langle x^*, v - x \rangle > 0$  for all  $v \in B_{\omega}(y)$ . We fix  $v \in B_{\omega}(y)$ . Since  $f^{\uparrow}(y, x - y) > 0$  we can find  $\varepsilon' \in (0, \varepsilon - \omega)$ ,  $u \in B_{\varepsilon'}(x)$  and  $t \in (0,1)$  such that f(u + t(v - u) > f(u). From the quasi-convexity of f we deduce that f(u) < f(v), whence,

 $f(v + \lambda(u - v)) \le f(v)$  for all  $\lambda \in (0,1)$ ,

so that

$$\inf_{\mu\in B_{\mathcal{E}}(x-y)}\frac{f(v+\lambda\mu)-f(v)}{\lambda}\leq \frac{f(v+\lambda(u-v))-f(v)}{\lambda}\leq 0 \text{ for all } \lambda\in(0,1).$$

Combining the inequalities and for any  $\varepsilon > 0$  there exists  $\omega > 0$  such that

$$\sup_{\substack{\nu \in B_{\omega}(y)\\\lambda \in (0,1)}} \left[ \inf_{\mu \in B_{\varepsilon}(x-y)} \frac{f(\nu+\lambda\mu) - f(\nu)}{\lambda} \right] \le 0,$$

which shows that  $f^{\uparrow}(y, x - y) \leq 0$ .

We consider the relationship between pseudo-convexity and quasi-convexity.

**Theorem 3.2** Let  $f: X \to \mathbb{R} \cup \{+\infty\}$  be a lower semi-continuous (l.s.c.) function on a Banach space *X* such that *f* Clarke-Rockafeller suddifferentiable. Consider the following assertions:

- (i) f is pseudoconvex.
- (ii) f is quasiconvex and  $(0 \in \partial f(x) \Rightarrow x$  is a global minimum of f).

Then, (i) implies (ii). And (ii) implies (i) if f is radially continuous.

**Proof.** (i)  $\Rightarrow$ (ii). We want to prove that f is quasiconvex. Suppose to the contrary that for some  $x, y \in X$ ,  $z \in (x, y)$  we have  $f(z) > \max\{f(x), f(y)\}$ . Since f is lower semicontinuous, we can find some  $\varepsilon > 0$  such that  $f(z') > \max\{f(x), f(y)\}$ , for all  $z' \in B_{\varepsilon}(z)$ . Since z cannot be a local nor global minimizers, there exist some  $v \in B_{\varepsilon}(z)$  such that f(v) < f(z). From **Lemma 2.2**, there exist  $u_n \to u \in [v, z)$  and  $u_n^* \in \partial f(u_n^*)$  such that

 $\langle u_n^*, z - u_n \rangle > 0.$ 

But since  $z \in (x, y)$ , either of the following must hold

 $\langle u_n^*, x - u_n \rangle > 0 \text{ or } \langle u_n^*, y - u_n \rangle > 0.$ 

Therefore,

 $f(u_n) \le \max\{f(x), f(y)\}.$ 

Which is a contradiction.

(ii)  $\Rightarrow$ (i). Let  $x \in \text{dom}\partial f$ ,  $y \in X$ , and  $x^* \in \partial f(x)$  such that  $\langle x^*, y - x \rangle \ge 0$ . If  $0 \in \partial f(x)$ , then x is a global minimum of f and  $f(x) \le f(y)$  in particular. Otherwise,  $[0 \notin \partial f(x)]$ , there exist  $d \in X$  such that  $\langle x^*, d \rangle > 0$ . We define a sequence  $\{y_n\}$  by

$$y_n = y + \left(\frac{1}{2n\|d\|}\right) d.$$

For every  $n \in \mathbb{N}$ , the point  $y_n$  satisfies

$$\begin{split} y_n &\in B_{1/n}(y), \\ &\langle x^*, y_n - x \rangle = \langle x^*, y_n - y \rangle + \langle x^*, y - x \rangle \geq \left(\frac{1}{2n \|d\|}\right) \langle x^*, d \rangle > 0. \end{split}$$

Using (8), we obtain that, for every n,

 $f(y_n) \ge f(x)$  and by radial continuity of  $f, f(y) \ge f(x)$ .

**Theorem 3.3** Let  $f: X \to \mathbb{R} \cup \{+\infty\}$  be a lower semi-continuous (l.s.c.) Clarke-Rockafeller subdifferentiable function. Consider the following assertions:

(i) f is pseudoconvex.

(ii)  $\partial f$  is pseudomonotone

Then, (i) implies (ii). And (ii) implies (i) if f is radially continuous.

**Proof.** (i)  $\Rightarrow$ (ii). Suppose  $x^* \in \partial f(x)$  such that  $\langle x^*, y - x \rangle \ge 0$ . By **Theorem 3.2**, f is quasiconvex. By

**Theorem 3.1,** we conclude that  $\partial f$  is quasimonotone. Hence  $\langle y^*, y - x \rangle \ge 0$ , for all  $y^* \in \partial f(y)$ . Suppose to the contrary that for some  $y^* \in \partial f(y)$ , we have  $\langle y^*, y - x \rangle = 0$ . From (8), we obtain  $f(x) \ge f(y)$ .

However, since  $f^{\uparrow}(x, y - x) > 0$ , there exist  $\varepsilon > 0$ , such that for some  $x_n \to x$ ,  $\lambda_n \searrow 0$  and for all  $y' \in B_{\varepsilon}(y)$ , we have  $f(x_n + t_n(y' - x_n)) > f(x_n)$ . By the quasiconvexity of f, it implies that  $f(y') > f(x_n)$  for every  $y' \in B_{\varepsilon}(y)$ . In particular, f(y') > f(x) because f is lower semicontinuous. Thus,  $f(y') \ge f(y)$ . This shows that y is a local minimum and also a global minimum, which is a contradiction since we can have that  $f(y) > f(x_n)$ .

(ii)  $\Rightarrow$ (i). Using Theorem , we prove that f is pseudoconvex. Since  $\partial f$  is pseudomonotone,  $\partial f$  is quasimonotone. By Theorem 3.2, f is quasi-convex. On the other hand, if x is not a minimizer of f, there exists  $y \in X$  such that f(y) < f(x). Using **Lemma 2.2**, we find  $u \in \text{dom}\partial f$  and  $u^* \in \partial f(u)$  such that  $\langle u^*, x - u \rangle > 0$  and by the pseudo-monotonicity of  $\partial f$ ,  $\langle x^*, x - u \rangle > 0$  for every  $x^* \in \partial f(x)$ . Hence, 0 does not  $\partial f(x)$ . Consequently, f satisfies condition  $0 \in \partial f(x)$ , which implies that x is a global minimum of f, which completes the proof.

**Theorem 3.4** Let  $f: X \to \mathbb{R} \cup \{+\infty\}$  be a lower semi-continuous (l.s.c.) Clarke-Rockafeller subdifferentiable function on a Banach space *X*. Consider the following assertions:

- (i) f is strictly pseudoconvex.
- (ii) f is strictly quasiconvex and  $(0 \in \partial f(x) \Rightarrow x$  is a global minimum of f),

Then, (i) implies (ii). And (ii) implies (i) if f is radially continuous.

**Proof.** (i)  $\Rightarrow$ (ii). We want to prove that *f* is strictly quasiconvex. Let *f* be a strictly pseudo-convex function, then by **Theorem 3.2**, the function *f* is quasiconvex and satisfies the optimality condition

 $0 \in \partial f(x) \Longrightarrow (x \text{ is a global minimum of } f).$ 

Since *f* is quasiconvex, then according to [14], it suffices to prove that *f* is radially non-constant. Assume by contradiction that there exists a closed segment [x, y] with with  $x \neq y$  where with *f* is constant. Let  $z \in (x, y)$  and apply the strict pseudo-convexity property to *x* and *z*, then

 $f(z) = f(x) \Longrightarrow (\forall z^* \in \partial f(z): \langle z^*, x - z \rangle < 0).$ 

Using the same argument for zand ywe obtain

 $f(z) = f(y) \Longrightarrow (\forall z^* \in \partial f(z): \langle z^*, y - z \rangle < 0).$ 

Since  $\partial f(z)$  is nonempty, it follows that

for all  $z^* \in \partial f(z), \langle z^*, x - y \rangle < 0$  and  $\langle z^*, x - y \rangle > 0$ ), which is a contradiction.

(ii)  $\Rightarrow$ (i). Assume that *f* satisfies condition ii) and *f* is radially continuous. Then by Theorem 3.2, *f* is pseudoconvex. We prove that *f* is pseudo-convex. Suppose by contradiction that there exist  $x \neq y$  in X and  $x^* \in \partial f(x)$  such that

$$\langle x^*, y - x \rangle \ge 0$$
 and  $f(x) \ge f(y)$ .

Then, it follows by pseudo-convexity property that

 $\forall z \in [x, y], f(z) = f(x).$ 

Since *f* is quasi-convex, then we have

 $\forall z \in [x, y], f(z) \ge f(x) \ge f(y).$ 

So *f* is not radially non-constant on *X* (since *f* is constant on [x, y]) which contradicts the fact *f* is strictly quasiconvex.

**Theorem 3.5** Let  $f: X \to \mathbb{R} \cup \{+\infty\}$  be a lower semi-continuous (l.s.c.) function such that f is radially Clarke-Rockafeller Sub-differentiable. Consider the following assertions:

- (i) f is strictly pseudo-convex.
- (ii)  $\partial f$  is strictly pseudomonotone

Then, (i) implies (ii). And (ii) implies (i) if f is radially continuous.

**Proof.** (i)  $\Rightarrow$ (ii). Suppose that *f* is strictly pseudoconvex. We want to prove that  $\partial f$  is strictly pseudomonotone. Suppose to the contrary that there exist two distinct points  $x, y \in X$ ,  $x^* \in \partial f(x)$  and  $y^* \in \partial f(y)$  such that  $\langle x^*, y - x \rangle \ge 0$  and  $\langle y^*, y - x \rangle \le 0$ .

Since *f* is strictly pseudoconvex, we have that

f(x) < f(y) and f(y) < f(x).

Which is a contradiction. Therefore,  $\partial f$  is strictly pseudomonotone.

(ii)  $\Rightarrow$ (i). Suppose that *f* satisfies condition (ii) and *f* is radially continuous. We want to prove that *f* is strictly pseudoconvex. Suppose to the contrary that there exist two distinct points  $x, y \in X$ , and  $x^* \in \partial f(x)$  such that

$$\langle x^*, y - x \rangle \ge 0$$
 and  $f(x) \ge f(y)$ 

Then,

 $\langle x^*, z - x \rangle \ge 0$  for all  $z \in [x, y]$ .

(17)

By theorem 3.3, f is quasiconvex. Consequently, f must be constant on [x, y]. Contrarily, from (16) and the strict monotonicity of  $\partial f(x)$ , we have

 $\langle x^*, z - x \rangle > 0, \forall z \in (x, y) \text{ and } \forall z^* \in \partial f(z).$  (18)

Pick  $z_0 \in (x, y)$  such that  $\partial f(z_0) \neq \emptyset$  (such a  $z_0$  exists since f is a radially Clarke-Rockafeller sub differentiable function). Choose any  $z_0^* \in \partial f(z_0)$ . Then,  $\langle z_0^*, z_0 - x \rangle > 0$ . Therefore,  $\langle z_0^*, y - z_0 \rangle > 0$ . Consequently, there exist  $\varepsilon > 0$  such that

 $\langle \mathbf{z}_0^*, \mathbf{y}' - \mathbf{z}_0 \rangle > 0$  for all  $\mathbf{y}' \in B_{\varepsilon}(\mathbf{y})$ .

By the pseudo-convexity of f, it follows that y is a global minimum of f. Hence,  $z_0$  is also a global minimum of f. Thus,  $0 \in \partial f(z_0)$  and this is a contradiction with (19).

## 4. Optimality Conditions and Variational Inequalities

We study the necessary and sufficient conditions for a point x to be a global minima of a pseudoconvex, lower semi-continuous and radially continuous function f over a convex set C.

**Theorem 4.1** Let  $f: C \subseteq X \to \mathbb{R} \cup \{+\infty\}$  be a lower semi-continuous Clarke-Rockafeller differentiable pseudoconvex function, and  $\bar{x} \in C$ . Then the following assertions are equivalent

- (i)  $\bar{x}$  is an optimal solution of (1).
- (ii)  $\partial f$  satisfies (11) at  $\bar{x} \in C$ .

Proof. (i)  $\Rightarrow$ (ii). Suppose that  $\bar{x}$  is a solution of (1), then by Definition 2.12, if  $f(\bar{x}) \le f(x)$ , then we have  $\forall x^* \in \partial f(x), \qquad \langle x^*, \bar{x} - x \rangle \le 0.$ 

So, the variational inequality (11) holds.

(ii)  $\Rightarrow$ (i). let  $x \in C$  such that  $x \neq \overline{x}$  then for some  $y \in (\overline{x}, x)$ . Then,

 $\forall y^* \in \partial f(y), \qquad \langle y^*, \bar{x} - y \rangle \le 0.$ 

Which follows that

 $\forall y^* \in \partial f(y), \qquad \langle y^*, x - y \rangle \le 0.$ 

Since  $\partial f(y)$  is nonempty and from the pseudoconvexity of *f* we have

 $\forall y \in (\bar{x}, x), \qquad \qquad f(y) \le f(x).$ 

But since *f* is lsc, then  $f(\bar{x}) \le f(x)$ .

We proceed to maximization problem.

Let  $C \subseteq X$  be convex and nonempty. Consider the maximization problem:

maximise f(x), subject to  $x \in C$ ,

(19)

where *f* a strictly pseudoconvex lower semi-continuous (l.s.c.) and radially Clarke-Rockafeller differentiable function.

**Theorem 4.3** Let  $f: C \subseteq X \to \mathbb{R} \cup \{+\infty\}$  be a lower semi-continuous Clarke-Rockafeller Sub-differentiable pseudo-convex function. let  $\bar{x} \in C$  such that

$$-\infty \le \inf_{c} f < f(\bar{x}). \tag{20}$$

Then,  $\bar{x}$  is a maximum of f on C if and only if for all  $x \in C$  such that  $f(x) = f(\bar{x})$  and for all  $x^* \in \partial f(x)$  we have:

 $\langle x^*, y-x \rangle \leq 0 \quad \forall y \in C \setminus \{x\}.$ 

**Proof.**  $\Rightarrow$  Suppose that x is a solution of (14). Let  $x \in X$  such that  $f(x) = f(\bar{x})$  and let  $x^* \in \partial f(x)$ . Then,  $f(y) = f(x), \forall y \in C$ .

Since f is strictly pseudoconvex, then

 $\langle x^*, y-x \rangle < 0, \forall y \in C \setminus \{x\}.$ 

 $\Leftarrow$  Suppose that there exist  $z \in C$  such that  $f(z) > f(\bar{x})$ . By the hypothesis, there exist  $z_0 \in C$  such that  $f(z_0) < f(\bar{x})$ . Since f is strongly radially Clarke-Rockafeller sub-differentiable, then there exists some  $x_0 \in (z_0, z)$  such that  $f(x_0) = f(\bar{x})$  and  $\partial f(x_0) \neq \emptyset$ . Pick any  $x_0^* \in \partial f(x_0)$ . Then,

 $\langle x_0^*, z - x_0 \rangle < 0$  and  $\langle x_0^*, x_0 - x_0 \rangle < 0$ , which is a contradiction.

Thus,  $\bar{x}$  is a maximum of f on C.

#### 5. Conclusion

We extended the relationships between convex functions and corresponding monotone maps to pseudo-convexity and the corresponding pseudo-monotonicity of their sub-differentiable maps. We characterized the lower semi-continuous Clarke-Rockafeller sub-differentiable pseudo-convex functions (1) by the corresponding monotonicity of their Clarke-Rockafeller sub-differential operator  $\partial f$  and presented their optimality conditions, using variational inequalities.

## References

D. Aussel, J. Corvellec, and M. Lassonde, Subdifferential Characterization of Quasiconvexity and Convexity, Journal of Convex Analysis, 1 (1994) pp. 195-201.

D. Aussel, J.Corvellec, and M. Lassonde, Mean Value Property and Subdifferential

Criteria for Lower Semi-Continuous Functions, Transactions of the American Mathematical Society, 347(1995)pp 1-15.

D. Aussel, Subdifferential properties of quasiconvex and pseudoconvex functions: unified approach, J. Optim. Theory Appl, 97 (1998) pp. 29–45.

H.Brezis, Equations et inequationsnonlin¶eairesendualite. Annales de l'InstitutFourier, (Grenoble) 18, fasc. 1, (1968) pp. 115-175.

C. E. Chidume, Applicable Functional Analysis, TETFUND Book project, Ibadanuniversity press publishing house, ISBN:978-978-8456-31-5, 2014.

F. H.Clarke, Optimization and Non-smooth Analysis, Wiley-Interscience, New-York, (1983) Pp. 308.

R. Correa, A.Joffre, and T.Thibault, Characterization of lower semi-continuous convex functions, Proc. Amer. Math. Soc. 116, (1992) pp. 67-72.

*R.* Correa, A.Joffre, and T.Thibault, Sub-differential monotonicity as characterization of convex functions, Numer. Funct. Anal. Optim. 15, (1994) pp.531-535.

G. P. Crepsi, and M. Rocca, Minty Variational Inequality and Monotone Trajectories of Differential Inclusions, Journal of Inequalities in Pure and Applied Mathematics, Art. 48. Vol. 5, No. 2, 2004.

J. P. Crouzeix, A. Eberhard and D. Ralph, A geometrical insight on Pseudo-convexityandPseudo-

monotonicity, Mathematical Programming, Springer Verlag, 123No. 1, (2009), pp. 61-83.

P. Cubiotti, Existence of Solutions for Lower Semi-Continuous Quasi-Equilibrium Problems, 30 No. 12, (1995), pp. 11-22.

A. Daniilidis, and N. Hadjisavvas, On the Sub-differentials of Quasi-Convex and Pseudo- Convex Functions and Cyclic Monotonicity, Journal of Mathematical Analysis and Applications, 237 (1999), pp. 30-42.

A. Daniilidis, and N. Hadjisawas, Characterization of Non-Smooth Semi-Strictly Quasi-Convex and Strictly Quasi-Convex Functions, Journal of Optimization Theory and Applications: Vol. 102, No. 3. (1999), pp. 525-536.

W. E. Diewert, Alternative characterizations of six kinds quasi-convexity in the non-differentiable case with applications non-smooth programming, Academic Press, (1981).

*R. Ferrentino, Variational Inequalities and Optimization Problems, Department of Economics and Sciences, university of Salemo, (2007), pp. 1-20.* 

N. Hadjisavvas, S. Komlosi, and S. Schaible, Nonconvex Optimizations and Applications: Handbook of Generalized Convexity and Generalized Monotonicity. Springer Science + Business Media, Inc., Boston, 2005.

N. Hadjisavvas, S. Schaible, and N. C. Wong, Pseudo-monotone Operators: A Survey of the Theory and Its Applications, J Optim Theory Appl, 152 (2012), pp. 1-20.

A. Hassouni, Quasi-monotone Multi-functions: Applications to Optimality Conditions in Quasi-Convex programming, Numer. Funct. Anal. Optim., 13 (3-4) (1992), pp. 267–275.

A. Hassouni, and Jaddar, On Generalized Monotone Multi-functions with Applications to Optimality Conditions in Generalized Convex Programming, Journal of Inequalities in Pure Mathematics, Volume 4, Issue 4, Article 67, (2003).

A. Hassouni, and A. Jaddar, A. On Pseudo-convex Functions and Applications to GlobalOptimization, ESIAM Proceedings, **20** (2007), pp. 138-148.

V. I. Ivanov, Characterization of Radially Lower Semi-continuous Pseudo-convexFunctions, Journal of Optimization Theory and Applications, Springer Science+BusinessMedia, LLC, part of Springer Nature, (2019), pp. 1-19.

A.Jaddar, and Y.Jabri, Characterization of continuous pseudoconvexfunctions' Extrema, Annals of University of Craiova, Math. Comp. Sci. Ser. **31**, (2004) pp. 45–50.

S.Karamaradian, and S.Schaible, Some kinds of monotone maps, J. Optim. Theory Appl. 66, (1990) pp. 37-46.

S.Karamaradian, Complementarity over cones with monotone and pseudo-monotone maps, Journal of Optimization Theory and Applications. **18**,(1976), pp. 445-454.

P. D. Khanh, and V. T. Phat, Second-order characterizations of quasiconvexity and pseudo-convexity for differentiable functions with Lipschitzian derivatives, Optimization Letters, Springer-Verlag GmbH Germany, part of Springer Nature, (2020), pp. 1-15.

B. T.Kien, J. C.Yao, and N. D.Yen, On the solution existence of pseudomonotonevariational Inequalities, J Glob Optim. 41, (2008), pp. 135–145.

*I.V.Konnov, On the convergence of a regularization method for non-monotone variational inequalities, Comput. Math. Math. Phys.* 46, (2006), 541–547.

*I.V. Konnov, M.S. Ali, and E.O.Mazurkevich, Regularization of non-monotone variational inequalities, Appl. Math. Optim.* 53, (2006),311–330.

S. Lahrech, A. Jaddar, A. Ouahab, and A. Mbarki, Some Remarks About StrictlyPseudoconvex Functions with Respect to the Clarke-RockerfellaSubdifferential,Lobachevskii Journal of Mathematics, 20 (2006), 55-62.

E. E. Levi, Studi suit puntisingolariessenzialidellefunzionianalitiche di due opiu variabilicomplesse. Ann. Mat. Pura Appl. 17 (1910), pp. 61-68.

N. B. Okelo, On Certain Conditions for Convex Optimization in Hilbert Spaces. Khayyam Journal of Mathematics, Vol. 5, No. 2, (2019), pp. 108-112.

B. B. Upadhyay, and P. Mishra, On Generalized Minty and StampacchiaVectorVariational-Like Inequalities and

Nonsmooth Vector Optimization Problem Involving Higher Order Strong Invexity, Journal of Scientific Research, 64 (1), (2020), pp. 282-291.

*A. Wald, Über die Produktionsgleichungen der ökonomischenWertlehre, ErgebnisseeinesmathematischenKolloquiums, 7: (1936), pp. 1-6.*