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# Optimality conditions for interval-valued univex programming

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## Abstract

We introduce concepts of interval-valued univex mappings, consider optimization conditions for interval-valued univex functions for the constrained interval-valued minimization problem, and show examples for the illustration purposes.

**Keywords:** Interval-valued univex mappings; Interval-valued univex programming; Optimality

## 1 Introduction

Convexity and generalized convexity are important in mathematical programming. Invex functions, introduced by Hanson [17], are important generalized convex functions and are successfully used in optimization and equilibrium problems. For example, necessary and sufficient conditions are obtained for  $K$ -invex functions in [14]. The concept of  $G$ -invex functions is introduced by Antczak [3]. Optimality and duality for differentiable  $G$ -multiobjective problems are considered in [4, 5]. Noor [26] considered invex equilibrium problems in the context of invexity. As an extension and refinement of Noor [26], Farajzadeh [15] gave some results for invex Ky Fan inequalities in topological vector spaces.

Another important type of generalized convex functions, called univex functions and preunivex functions, is introduced in [8]. Suppose  $\emptyset \neq X \subseteq R^n$ ,  $\eta : X \times X \rightarrow R^n$ ,  $\Phi : R \rightarrow R$ , and  $b = b(x, y) : X \times X \rightarrow R^+$ . A differentiable function  $F : X \rightarrow R$  is said to be univex at  $y \in X$  with respect to  $\eta$ ,  $\Phi$ ,  $b$  if, for all  $x \in X$ ,

$$b(x, y)\Phi[F(x) - F(y)] \geq \eta^t(x, y)\nabla F(y). \quad (1)$$

Later, some generalized optimality conditions of primal and dual problems were considered by Hanson and Mond [18]. Combing with generalized type I and univex functions, optimality conditions and duality for several mathematical programming problems were considered by many researchers [1, 16, 29], and more and more scholars pay attention to type I and univex functions [24, 25, 34, 35].

The authors of [2, 6, 9, 12, 27, 30, 33, 36–39] have studied generalized convex interval-valued mappings and their connection with interval-valued optimization. For example, Steuer [33] proposed three algorithms, called the F-cone algorithm, E-cone algorithm, and emanating algorithms, to solve the linear programming problems with interval-valued

objective functions. To prove strong duality theorems, Wu [37] derived KKT optimality conditions in the interval-valued problems under convexity hypotheses. Wu [36] also obtained KKT conditions in an optimization problem with an interval-valued objective function using  $H$ -derivatives and the concept of weakly differentiable functions. Since the  $H$ -derivative suffers certain disadvantages, Chalco-Cano et al. [10] gave KKT-type optimality conditions, which were obtained using the  $gH$ -derivatives of interval-valued functions. Also, they studied the relationship between the approach presented with other known approaches given by Wu [36]. However, these methods cannot solve a kind of optimization problems with interval-valued objective functions that are not  $LU$ -convex but univex. Antczak [6] used the classical exact  $l_1$  penalty function method for solving nondifferentiable interval-valued optimization problems under convexity hypotheses. Optimality conditions in invex optimization problems with an interval-valued objective function were discussed by Zhang et al. [39]. Using  $gH$ -differentiability, Li et al. [21] introduced interval-valued invex mappings and gave the optimality conditions for interval-valued objective functions under invexity. By using the weak derivative of fuzzy functions, Li et al. [22] defined fuzzy weakly univex functions and considered optimization conditions for fuzzy minimization problem.

Followed by [21] and [22], in this paper, we introduce the concept of interval-valued univex mappings, consider optimization conditions for interval-valued univex functions for the constrained interval-valued minimization problem, and show examples for illustration purposes. The present paper can be seen as promotion and expansion of [20]. The method presented in this paper is different from that in [6]. Our method cannot solve Example 3.1 of [6] because the objective function is not  $gH$ -differentiable. Example 4.1 shows that the methods given by [6, 33, 36, 37] cannot solve a kind of optimization problems for interval-valued univex mappings. Example 4.2 shows that the methods given by Li et al. [22] cannot solve a kind of fuzzy optimization problems for interval-valued univex mappings. Finally, Example 4.3 shows that the method given in [10] cannot solve a kind of optimization problems for interval-valued univex mappings. In Sect. 3, we introduce the concept of interval-valued univex mappings and discuss some their properties. Section 4 deals with optimality conditions for the constrained interval-valued minimization problem under the assumption of interval-valued univexity.

## 2 Preliminaries

In this paper, a closed interval in  $R$  is denoted by  $A = [a^L, a^U]$ . Every  $a \in R$  is considered as a particular closed interval  $a = [a, a]$ . The set of closed intervals is denoted by  $\mathcal{I}$ .

Given  $A = [a^L, a^U]$  and  $B = [b^L, b^U] \in \mathcal{I}$ , the arithmetic operations and order are defined in [32] as follows:

- (1)  $A + B = [a^L + b^L, a^U + b^U]$  and  $-A = \{-a : a \in A\} = [-a^U, -a^L]$ ;
- (2)  $A \ominus_{gH} B = [\min(a^L - b^L, a^U - b^U), \max(a^L - b^L, a^U - b^U)]$ ;
- (3)  $A \leq B \Leftrightarrow a^L \leq b^L$  and  $a^U \leq b^U$ ;  $A < B \Leftrightarrow A \leq B$  and  $A \neq B$ .

For  $X \subseteq R^n$ , a mapping  $F : X \rightarrow \mathcal{I}$  is called an interval-valued function. Then  $F(x) = [F^L(x), F^U(x)]$ , where  $F^L(x)$  and  $F^U(x)$  are two real-valued functions defined on  $R^n$  and satisfying  $F^L(x) \leq F^U(x)$  for every  $x \in X$ . If  $F^L(x)$  and  $F^U(x)$  are continuous, then  $F(x)$  is said to be continuous.

It is well known that the derivative and subderivative of a function is important in the study of generalized convexity and mathematical programming. For example, a classic

subdifferential is introduced by Azimov and Gasimov [7]. Some theorems connecting operations on the weak subdifferential in the nonsmooth and nonconvex analysis are provided in [13]. The derivative and subderivative of interval-valued functions are extensions of real-valued functions. Due to different arithmetics of intervals, several definitions about derivatives of interval-valued functions are introduced by the authors, such as weakly differentiable functions [36],  $H$ -differentiable functions (based on the Hukuhara difference of two closed intervals [36]),  $gH$ -differentiable functions (based on the operation  $\ominus_{gH}$  of two closed intervals [11, 31]), and subdifferentiable functions (based on the difference  $A - B = [a^L - b^U, a^U - b^L]$  of two closed intervals [6]). In this paper, we always use weakly differentiable and  $gH$ -differentiable functions, which are defined as follows.

Let  $X$  be an open set in  $R^n$ , and let  $F(x) = [F^L(x), F^U(x)]$ . Then  $F(x)$  is called weakly differentiable at  $x_0$  if  $F^L(x)$  and  $F^U(x)$  are differentiable at  $x_0$ .

Let  $x_0 \in (a, b)$  and  $h$  be such that  $x_0 + h \in (a, b)$ . Then

$$F'(x_0) = \lim_{x \rightarrow 0} [F(x_0 + h) \ominus_{gH} F(x_0)]. \tag{2}$$

If  $F'(x_0) \in \mathcal{I}$  exists, then  $F$  is  $gH$ -differentiable at  $x_0$ .

If  $F^L(x)$  and  $F^U(x)$  are differentiable functions at  $x \in (a, b)$ , then  $F(x)$  is  $gH$ -differentiable at  $x$ , and

$$F'(x) = [\min\{(F^L)'(x), (F^U)'(x)\}, \max\{(F^L)'(x), (F^U)'(x)\}]. \tag{3}$$

We say that an interval-valued function  $F$  is  $gH$ -differentiable at  $x = (x_1, \dots, x_n) \in X$  if all the partial  $gH$ -derivatives  $(\frac{\partial F}{\partial x_1})(x), \dots, (\frac{\partial F}{\partial x_n})(x)$  exist on some neighborhood of  $x$  and are continuous at  $x$ . We write

$$\nabla F(x) = \left( \left( \frac{\partial F}{\partial x_1} \right)(x), \left( \frac{\partial F}{\partial x_2} \right)(x), \dots, \left( \frac{\partial F}{\partial x_n} \right)(x) \right)^t,$$

and we call  $\nabla F(x)$  the gradient of a  $gH$ -differentiable interval-valued function  $F$  at  $x$ .

Let  $\mathbb{H}(R^n)$  denote the family of nonempty compact subsets of  $R^n$ . For  $A, B \in \mathbb{H}(R^n)$ , the Hausdorff metric  $h(A, B)$  on  $\mathbb{H}(R^n)$  is defined by

$$h(A, B) = \inf\{\varepsilon \mid A \subseteq N(B, \varepsilon), B \subseteq N(A, \varepsilon)\},$$

where

$$N(A, \varepsilon) = \{x \in R^n \mid d(x, A) < \varepsilon\}, \quad d(x, A) = \inf_{a \in A} \|x - a\|.$$

The following basic result (which can be found in Lemma 3.1. of [19]) of the mathematical analysis is well known:

Suppose that  $\Phi : R^n \rightarrow R^n$  is continuous and let  $X \in \mathbb{H}(R^n)$ . Then the mapping

$$\Psi : \mathbb{H}(X^n) \rightarrow \mathbb{H}(R^n), \quad \Psi(A) = \{\phi(a) \mid a \in A\}$$

is uniformly continuous in  $h$ -metric.

We say that  $\Psi : \mathcal{I} \rightarrow \mathcal{I}$  is increasing if  $A \preceq B$  implies  $\Psi(A) \preceq \Psi(B)$ . From the above result we can prove the following result:

If function  $\Phi : R \rightarrow R$  is increasing, then  $\Psi : \mathcal{I} \rightarrow \mathcal{I}$  is increasing. Moreover,  $\Psi([a^L, a^U]) = [\Phi(a^L), \Phi(a^U)]$ .

### 3 Interval-valued univex functions

In this section, we define interval-valued univex functions as a generalization of univex functions [8] and discuss some their properties.

Let  $X$  be an invex set in  $R^n$  (the concept of an invex set can be found in [8]), and let  $F$  be an interval-valued function. The following definition is a particular case of fuzzy weakly univex functions, which has been introduced in [22].

Suppose  $F$  is a weakly differentiable interval-valued function. Then  $F$  is weakly univex at  $y \in X$  with respect to  $\eta, \Phi, b$  if and only if both  $F^L(x)$  and  $F^U(x)$  are univex at  $y \in X$ , that is, for all  $x \in X$ ,

$$b(x, y)\Phi[F^L(x) - F^L(y)] \geq \eta^t(x, y)\nabla F^L(y), \tag{4}$$

$$b(x, y)\Phi[F^U(x) - F^U(y)] \geq \eta^t(x, y)\nabla F^U(y), \tag{5}$$

where  $\eta = \eta(x, y) : X \times X \rightarrow R^n, \Phi : R \rightarrow R$ , and  $b = b(x, y) : X \times X \times [0, 1] \rightarrow R^+$ .

*Remark 3.1* The concept of  $LU$ -invexity for interval-valued functions is introduced in [39], since it considers the endpoint functions; in this paper, we call them weakly invex. Every interval-valued weakly invex function is interval-valued weakly univex with respect to  $\eta, b, \Phi$ , where

$$\Phi(x) = x, \quad b = 1,$$

but the converse is not true.

*Example 3.1* Consider the function  $F : (-\infty, 0) \rightarrow \mathcal{I}$  defined by

$$F(x) = [1, 2]x^3,$$

$$\eta(x, y) = \begin{cases} x^2 + xy + y^2, & x > y, \\ x - y, & x \leq y, \end{cases}$$

$$b(x, y) = \begin{cases} \frac{y^2}{x-y}, & x > y, \\ 0, & x \leq y. \end{cases}$$

Let  $\Phi : R \rightarrow R$  be defined by  $\Phi(V) = 3V, F^L(x) = 2x^3$ , and  $F^U(x) = x^3$ ; then  $\nabla F^L(x) = 6x^2$  and  $\nabla F^U(x) = 3x^2$ . Then  $F$  is interval-valued weakly univex but not interval-valued weakly invex, since for  $x = -2$  and  $y = -1, F^U(x) - F^U(y) < \eta^t(x, y)\nabla F^U(y)$ .

Let  $X$  be a nonempty open set in  $R^n, \eta : X \times X \rightarrow R^n, \Psi : \mathcal{I} \rightarrow \mathcal{I}$ , and  $b = b(x, y) : X \times X \rightarrow R^+$ .

**Definition 3.1** Suppose  $F$  is a  $gH$ -differentiable interval-valued function. Then  $F$  is univex at  $y \in X$  with respect to  $\eta, \Psi, b$  if for all  $x \in X$ ,

$$b(x, y)\Psi [F(x) \ominus_{gH} F(y)] \succeq \eta^t(x, y)\nabla F(y). \tag{6}$$

The following example shows that an interval-valued univex function may not be an interval-valued weakly univex function.

*Example 3.2* Suppose  $F(x) = [-|x|, |x|], x \in R, b = 1$ , and  $\Phi(a) = a$ . Then  $\Psi[a, b] = [a, b]$  is induced by  $\Phi(a) = a$ , and

$$\eta(x, y) = \begin{cases} x - y, & xy \geq 0, \\ x + y, & xy < 0. \end{cases}$$

Then  $F(x)$  is  $gH$ -differentiable on  $R$ , and  $F'(y) = [-1, 1]$ . We can prove that

$$b\Psi [F(x) \ominus_{gH} F(y)] \succeq \eta^t(x, y)\nabla F(y).$$

Therefore  $F(x)$  is univex with respect to  $\eta, b, \Psi$ , but  $F(x)$  is not weakly univex since  $F^L(x)$  is not univex with respect to  $\eta, b, \Phi$ .

**Theorem 3.1** *Suppose  $F(x)$  is  $gH$ -differentiable. If  $F(x)$  is an interval-valued weakly univex function with respect to  $\eta, b, \Phi$  and  $\Phi$  is increasing, then  $F(x)$  is an interval-valued univex function with respect to the same  $\eta, b$ , and  $\Psi$ , where  $\Psi$  is an extension of  $\Phi$ .*

*Proof* Since  $F(x)$  is weakly univex at  $y$ , then real-valued functions  $F^L$  and  $F^U$  are univex at  $y$ , that is,

$$\begin{aligned} b(x, y)\Phi [F^L(x) - F^L(y)] &\geq \eta^t(x, y)\nabla F^L(y) \quad \text{and} \\ b(x, y)\Phi [F^U(x) - F^U(y)] &\geq \eta^t(x, y)\nabla F^U(y) \end{aligned}$$

for all  $x \in X$ .

(i) Under the condition  $\eta^t(x, y)\nabla F^L(y) \leq \eta^t(x, y)\nabla F^U(y)$ , we have

$$\eta^t(x, y)\nabla F(y) = [\eta^t(x, y)\nabla F^L(y), \eta^t(x, y)\nabla F^U(y)].$$

If  $F(x) \ominus_{gH} F(y) = [F^L(x) - F^L(y), F^U(x) - F^U(y)]$ , then since  $\Phi$  is increasing, we have

$$\begin{aligned} &b(x, y)\Psi [F(x) \ominus_{gH} F(y)] \\ &= [b(x, y)\Phi (F^L(x) - F^L(y)), b(x, y)\Phi (F^U(x) - F^U(y))] \\ &\geq [\eta^t(x, y)\nabla F^L(y), \eta^t(x, y)\nabla F^U(y)] \\ &= \eta^t(x, y)\nabla F(y). \end{aligned}$$

If  $F(x) \ominus_{gH} F(y) = [F^U(x) - F^U(y), F^L(x) - F^L(y)]$ , then

$$\begin{aligned} & b(x, y)\Phi(F^L(x) - F^L(y)) \\ & \geq b(x, y)\Phi(F^U(x) - F^U(y)) \\ & \geq \eta^t(x, y)\nabla F^U(y) \\ & \geq \eta^t(x, y)\nabla F^L(y), \end{aligned}$$

and since  $\Phi$  is increasing, we have

$$\begin{aligned} & b(x, y)\Psi[F(x) \ominus_{gH} F(y)] \\ & = [b(x, y)\Psi[F^U(x) - F^U(y), F^L(x) - F^L(y)]] \\ & = [b(x, y)\Phi(F^U(x) - F^U(y)), b(x, y)\Phi(F^L(x) - F^L(y))] \\ & \geq [\eta^t(x, y)\nabla F^L(y), \eta^t(x, y)\nabla F^U(y)] \\ & = \eta^t(x, y)\nabla F(y). \end{aligned}$$

(ii) Under the condition  $\eta^t(x, y)\nabla F^L(y) > \eta^t(x, y)\nabla F^U(y)$ , we have

$$\eta^t(x, y)\nabla F(y) = [\eta^t(x, y)\nabla F^U(y), \eta^t(x, y)\nabla F^L(y)].$$

If  $F(x) \ominus_{gH} F(y) = [F^U(x) - F^U(y), F^L(x) - F^L(y)]$ , then since  $\Phi$  is increasing, we have

$$\begin{aligned} & b(x, y)\Psi[F(x) \ominus_{gH} F(y)] \\ & = [b(x, y)\Phi(F^U(x) - F^U(y)), b(x, y)\Phi(F^L(x) - F^L(y))] \\ & \geq [\eta^t(x, y)\nabla F^U(y), \eta^t(x, y)\nabla F^L(y)] \\ & = \eta^t(x, y)\nabla F(y). \end{aligned}$$

If  $F(x) \ominus_{gH} F(y) = [F^L(x) - F^L(y), F^U(x) - F^U(y)]$ , then

$$\begin{aligned} & b(x, y)\Phi(F^U(x) - F^U(y)) \\ & \geq b(x, y)\Phi(F^L(x) - F^L(y)) \\ & \geq \eta^t(x, y)\nabla F^L(y) \\ & \geq \eta^t(x, y)\nabla F^U(y). \end{aligned}$$

Since  $\Phi$  is increasing, we have

$$\begin{aligned} & b(x, y)\Psi[F(x) \ominus_{gH} F(y)] \\ & = b(x, y)\Psi[F^L(x) - F^L(y), F^U(x) - F^U(y)] \\ & = [b(x, y)\Phi(F^L(x) - F^L(y)), b(x, y)\Phi(F^U(x) - F^U(y))] \\ & \geq [\eta^t(x, y)\nabla F^U(y), \eta^t(x, y)\nabla F^L(y)] \\ & = \eta^t(x, y)\nabla F(y). \end{aligned}$$

□

*Remark 3.2* If  $\Phi$  is nonincreasing, then Theorem 3.1 may not be true (as shown in the following Example 3.3).

*Example 3.3* Suppose  $F(x) = [-2, 1]x^2, x < 0$ . Then  $F(x)$  is  $gH$ -differentiable and weakly differentiable. It is easy to check that  $F(x)$  is weakly univex with respect to  $\eta(x, y) = x - y$ ,

$$b(x, y) = \begin{cases} 1, & x \leq y < 0, \\ \frac{-2y(x-y)}{-x^2+y^2}, & y < x < 0, \end{cases}$$

and  $\Phi(a) = |a|$ . However,  $F(x)$  is not univex with respect to the same  $\eta(x, y), b$ , and  $\Psi$ , where  $\Psi$  is defined by the extension of  $\Phi(a) = |a|$ .

#### 4 Optimality criteria for interval-valued univex mappings

In this section, for  $gH$ -differentiable interval-valued univex functions, we establish sufficient optimality conditions for a feasible solution  $x^*$  to be an optimal solution or a non-dominated solution for (P).

Suppose  $F(x), g_1(x), \dots, g_m(x)$  are  $gH$ -differentiable interval-valued mappings defined on a nonempty open set  $X \subseteq R^n$ . Then, we consider the primal problem:

$$(P) \quad \min F(x) \\ \text{s.t. } g(x) \leq 0.$$

Let  $P := \{x \in X : g(x) \leq 0\}$  denote the feasible set of (P).

Since  $\leq$  is a partial order, the optimal solution may not exist for some interval-valued optimization problems. Therefore, authors always consider the concept of a nondominated solution in this situation. We reconsider an optimal solution and nondominated solution as follows.

##### Definition 4.1

- (i)  $x^* \in P$  is an optimal solution of (P)  $\Leftrightarrow F(x^*) \leq F(x)$  for all  $x \in P$ . In this case,  $F(x^*)$  is called the optimal objective value of  $F$ .
- (ii)  $x^* \in P$  is a nondominated solution of (P)  $\Leftrightarrow$  there exists no  $x_0 \in P$  such that  $F(x_0) < F(x^*)$ . In this case,  $F(x^*)$  is called the nondominated objective value of  $F$ .

**Theorem 4.1** *Let  $x^*$  be P-feasible. Suppose that:*

- (i) *there exist  $\eta, \Psi_0, b_0, \Psi_i, b_i, i = 1, 2, \dots, m$ , such that*

$$b_0(x, y)\Psi_0[F(x) \ominus_{gH} F(x^*)] \geq \eta^t(x, x^*)\nabla F(x^*) \tag{7}$$

and

$$-b_i(x, x^*)\Psi_i[g_i(x^*)] \geq \eta^t(x, x^*)\nabla g_i(x^*) \tag{8}$$

for all feasible  $x$ ;

(ii) there exists  $y^* \in R^m$  such that

$$\nabla F(x^*) = -y^{*t} \nabla g(x^*), \tag{9}$$

$$y^* \geq 0. \tag{10}$$

Further suppose that

$$\Psi_0(\mu) \geq 0 \Rightarrow \mu \geq 0, \tag{11}$$

$$\mu \leq 0 \Rightarrow \Psi_i(\mu) \geq 0, \tag{12}$$

and

$$b_0(x, x^*) > 0, \quad b_i(x, x^*) \geq 0, \tag{13}$$

for all feasible  $x$ . Then  $x^*$  is an optimal solution of (P).

*Proof* Let  $x$  be  $P$ -feasible. Then

$$g(x) \leq 0.$$

This, along with (12), yields

$$\Psi_i[g_i(x)] \geq 0.$$

From (7)–(13) it follows that

$$\begin{aligned} b_0(x, x^*) \Psi_0[F(x) \ominus_{gH} F(x^*)] &\geq \eta^t(x, x^*) \nabla F(x^*) \\ &= -\eta^t(x, x^*) \sum_{i=1}^m y_i \nabla g_i(x^*) \\ &\geq \sum_{i=1}^m b_i(x, x^*) y_i \Psi_i[g_i(x^*)] \\ &\geq 0. \end{aligned}$$

From (13) it follows that

$$\Psi_0[F(x) \ominus_{gH} F(x^*)] \geq 0.$$

By (11) we have

$$F(x) \ominus_{gH} F(x^*) \geq 0.$$

Thus

$$F(x) \geq F(x^*).$$

Therefore  $x^*$  is an optimal solution of (P). □



*Remark 4.1* If we change the condition

$$\Psi_0(\mu) \geq 0 \Rightarrow \mu \geq 0$$

of Theorem 4.1 by

$$\Psi_0(\mu) \not\geq 0 \Rightarrow \mu \not\geq 0, \tag{14}$$

then  $x^*$  is a nondominated solution of (P).

In Theorem 18 of [20], the authors also gave a sufficient optimality condition for a feasible solution  $x^*$  to be an optimal solution. In this theorem, the equation

$$\nabla F(x^*) + y^{*t} \nabla g(x^*) = 0$$

was used, substituted for (9) of Theorem 4.1. We can prove that the previous equation is very restrictive. In fact, in case  $F(x)$  is a unary function, suppose  $\nabla F(x^*) = [a, b]$  and  $y^{*t} \nabla g(x^*) = [yc, yd]$ . Then we have  $[a, b] + [yc, yd] = [a + yc, b + yd] = 0$ , where  $a \leq b$  and  $yc \leq yd$ . Therefore we have  $a = b$  and  $c = d$  since  $y \geq 0$ . That is to say,  $\nabla F(x^*)$  is a real number instead of an interval. In the following example, we can observe that  $x^*$  is an optimal solution of (P), but  $x^*$  do not satisfies the previous equation. The following example also shows the advantages of our method over [6, 33, 36, 37].

*Example 4.1*

$$\begin{aligned} \min F(x) &= \left[ \frac{1}{2}, \frac{3}{2} \right] \sin^2 x_1 + \left[ \frac{1}{2}, \frac{3}{2} \right] \sin^2 x_2 \\ \text{s.t. } g(x) &= \left[ \frac{1}{2}, \frac{3}{2} \right] (\sin x_1 - 1)^2 + \left[ \frac{1}{2}, \frac{3}{2} \right] (\sin x_2 - 1)^2 \leq \frac{1}{4} \left[ \frac{1}{2}, \frac{3}{2} \right], \\ x_1, x_2 &\in \left( 0, \frac{\pi}{2} \right). \end{aligned}$$

We can observe that  $F(x)$  is weakly differentiable,  $H$ -differentiable, and  $gH$ -differentiable. Since the interval-valued function  $F(x)$  is not convex, the method in [6, 33, 36, 37] cannot be used.

The function  $F(x)$  is interval-valued univex with respect to

$$\begin{aligned} \eta(x, y) &= \begin{cases} \left( \frac{\sin x_1 - \sin y_1}{\cos y_1}, \frac{\sin x_2 - \sin y_2}{\cos y_2} \right)^t, & (x_1, x_2) \geq (y_1, y_2), \\ 0 & \text{otherwise,} \end{cases} \\ b_0(x, y) &= \begin{cases} 1, & (x_1, x_2) \geq (y_1, y_2), \\ 0 & \text{otherwise,} \end{cases} \end{aligned}$$

and  $\Psi$  is induced by  $\Phi(a) = 2a$ ,  $b_1(x, y) = b_0(x, y)$ , and  $\Psi_1$  is induced by  $\Phi_1(a) = |a|$ , where  $x = (x_1, x_2)^t$  and  $y = (y_1, y_2)^t$ . The point  $x^* = (\sin^{-1}(1 - \frac{1}{2\sqrt{2}}), \sin^{-1}(1 - \frac{1}{2\sqrt{2}}))^t$  is a feasible solution. We can also see that  $(F, g)$  satisfies the hypotheses of Theorem 4.1. Therefore  $x^* = (\sin^{-1}(1 - \frac{1}{2\sqrt{2}}), \sin^{-1}(1 - \frac{1}{2\sqrt{2}}))^t$  is an optimal solution.

**Theorem 4.2** *Let  $x^*$  be  $P$ -feasible. Suppose that:*

(i) *there exist  $\eta, \Psi_0, b_0, \Psi_i, b_i, i = 1, 2, \dots, m$ , such that*

$$b_0(x, y)\Psi_0[F(x) \ominus_{gH} F(x^*)] \geq \eta^t(x, x^*)\nabla F(x^*) \tag{15}$$

and

$$-b_i(x, x^*)\Psi_i[g_i(x^*)] \geq \eta^t(x, x^*)\nabla g_i(x^*) \tag{16}$$

for all feasible  $x$ ;

(ii) *there exists  $y^* \in R^m$  such that*

$$\{\nabla F(x^*)\}^L = \{-y^{*t}\nabla g(x^*)\}^L, \tag{17}$$

$$y^* \geq 0. \tag{18}$$

Further, suppose that

$$\Psi_0(\mu) \not\leq 0 \Rightarrow \mu \not\leq 0, \tag{19}$$

$$\mu \leq 0 \Rightarrow \Psi_i(\mu) \geq 0, \tag{20}$$

and

$$b_0(x, x^*) > 0, \quad b_i(x, x^*) \geq 0 \tag{21}$$

for all feasible  $x$ . Then  $x^*$  is a nondominated solution of  $(P)$ .

*Proof* Let  $x$  be  $P$ -feasible. Then

$$\tilde{g}(x) \leq 0.$$

From (20) we conclude

$$\Psi_i[g_i(x)] \geq 0.$$

From (15), (16) it follows that

$$b_0(x, y)\{\Psi_0[F(x) \ominus_{gH} F(x^*)]\}^L \geq \{\eta^t(x, x^*)\nabla F(x^*)\}^L,$$

$$b_0(x, y)\{\Psi_0[F(x) \ominus_g F(x^*)]\}^U \geq \{\eta^t(x, x^*)\nabla F(x^*)\}^U,$$

and

$$b_i(x, x^*)\{\Psi_i[g_i(x^*)]\}^L \leq \{-\eta^t(x, x^*)\nabla g_i(x^*)\}^L,$$

$$b_i(x, x^*)\{\Psi_i[g_i(x^*)]\}^U \leq \{-\eta^t(x, x^*)\nabla g_i(x^*)\}^U.$$

Since

$$\begin{aligned} \eta^t(x, x^*) \nabla F(x^*) &= \eta^t(x, x^*) [\{\nabla F(x^*)\}^L, \{\nabla F(x^*)\}^U] \\ &= \begin{cases} [\eta^t(x, x^*) \{\nabla F(x^*)\}^L, \eta^t(x, x^*) \{\nabla F(x^*)\}^U], & \eta^t(x, x^*) \geq 0, \\ [\eta^t(x, x^*) \{\nabla F(x^*)\}^U, \eta^t(x, x^*) \{\nabla F(x^*)\}^L], & \eta^t(x, x^*) < 0, \end{cases} \end{aligned}$$

and

$$\begin{aligned} -\eta^t(x, x^*) \nabla g_i(x^*) &= -\eta^t(x, x^*) [\{\nabla g_i(x^*)\}^L, \{\nabla g_i(x^*)\}^U] \\ &= \begin{cases} [\eta^t(x, x^*) \{-\nabla g_i(x^*)\}^U, \eta^t(x, x^*) \{-\nabla g_i(x^*)\}^L], & \eta^t(x, x^*) \geq 0, \\ [\eta^t(x, x^*) \{-\nabla g_i(x^*)\}^L, \eta^t(x, x^*) \{-\nabla g_i(x^*)\}^U], & \eta^t(x, x^*) < 0, \end{cases} \end{aligned}$$

we consider the following two cases.

Case (i)

$$\{\eta^t(x, x^*) \nabla F(x^*)\}^L = \eta^t(x, x^*) \{\nabla F(x^*)\}^L$$

and

$$\{-\eta^t(x, x^*) \nabla g_i(x^*)\}^L = \eta^t(x, x^*) \{-\nabla g_i(x^*)\}^U$$

yield

$$\{\eta^t(x, x^*) \nabla F(x^*)\}^U = \eta^t(x, x^*) \{\nabla F(x^*)\}^U$$

and

$$\{-\eta^t(x, x^*) \nabla g_i(x^*)\}^U = \eta^t(x, x^*) \{-\nabla g_i(x^*)\}^L.$$

Thus

$$\begin{aligned} b_0(x, y) \{\Psi_0[F(x) \ominus_{gH} F(x^*)]\}^L &\geq \{\eta^t(x, x^*) \nabla F(x^*)\}^L \\ &= \eta^t(x, x^*) \{\nabla F(x^*)\}^L \\ &= \eta^t(x, x^*) \{-y^{*t} \nabla g(x^*)\}^L \\ &\geq \sum_{i=1}^m b_i(x, x^*) y_i \{\Psi_i[g_i(x^*)]\}^L \\ &\geq 0. \end{aligned}$$

From (21) it follows that

$$\Psi_0[F(x) \ominus_{gH} F(x^*)] \geq 0.$$

Then

$$F(x) \ominus_{gH} F(x^*) \not\leq 0,$$

and thus

$$F(x) \not\leq F(x^*).$$

Therefore  $x^*$  is a nondominated solution of (P).

Case (ii)

$$\{\eta^t(x, x^*) \nabla F(x^*)\}^L = \eta^t(x, x^*) \{\nabla F(x^*)\}^U$$

and

$$\{-\eta^t(x, x^*) \nabla g_i(x^*)\}^L = \eta^t(x, x^*) \{-\nabla g_i(x^*)\}^L$$

yield

$$\{\eta^t(x, x^*) \nabla F(x^*)\}^U = \eta^t(x, x^*) \{\nabla F(x^*)\}^L$$

and

$$\{-\eta^t(x, x^*) \nabla g_i(x^*)\}^U = \eta^t(x, x^*) \{-\nabla g_i(x^*)\}^U.$$

Thus

$$\begin{aligned} b_0(x, y) \{\Psi_0[F(x) \ominus_{gH} F(x^*)]\}^U &\geq \{\eta^t(x, x^*) \nabla F(x^*)\}^U \\ &= \eta^t(x, x^*) \{\nabla F(x^*)\}^L \\ &= \eta^t(x, x^*) \{-\gamma^{*t} \nabla g(x^*)\}^L \\ &\geq \sum_{i=1}^m b_i(x, x^*) y_i \{\Psi_i[g_i(x^*)]\}^L \\ &\geq 0, \end{aligned}$$

From (21) it follows that

$$\Psi_0[F(x) \ominus_{gH} F(x^*)] \not\leq 0.$$

Then

$$F(x) \ominus_{gH} F(x^*) \not\leq 0,$$

and thus

$$F(x) \not\leq F(x^*).$$

Therefore  $x^*$  is a nondominated solution of (P). □

**Theorem 4.3** *Let  $x^*$  be  $P$ -feasible. Suppose that:*

(i) *there exist  $\eta, \Psi_0, b_0, \Psi_i, b_i, i = 1, 2, \dots, m$ , such that*

$$b_0(x, y)\Psi_0[F(x) \ominus_{gH} F(x^*)] \geq \eta^t(x, x^*)\nabla F(x^*) \tag{22}$$

and

$$-b_i(x, x^*)\Psi_i[g_i(x^*)] \geq \eta^t(x, x^*)\nabla g_i(x^*) \tag{23}$$

for all feasible  $x$ ;

(ii) *there exists  $y^* \in R^m$  such that*

$$\{\nabla F(x^*)\}^U = \{-y^{*t}\nabla g(x^*)\}^U, \tag{24}$$

$$y^* \geq 0. \tag{25}$$

Further, suppose that

$$\Psi_0(\mu) \not\leq 0 \Rightarrow \mu \not\leq 0, \tag{26}$$

$$\mu \leq 0 \Rightarrow \Psi_i(\mu) \geq 0, \tag{27}$$

and

$$b_0(x, x^*) > 0, \quad b_i(x, x^*) \geq 0 \tag{28}$$

for all feasible  $x$ . Then  $x^*$  is a nondominated solution of  $(P)$ .

The following example shows the advantages of our method over [22].

*Example 4.2*

$$\begin{aligned} \min F(x) &= [-1, 1]|x| \\ \text{s.t. } g(x) &= x - 1 \leq 0. \end{aligned}$$

Since  $F^L(x) = -|x|$  and  $F^U(x) = |x|$  is not differentiable at  $x = 0$ ,  $F(x)$  is not weakly differentiable at  $x = 0$ . Therefore the method in [22] cannot be used.

Note that the objective function  $F(x)$  is  $gH$ -differentiable on  $R$  and that  $F'(y) = [-1, 1]$ . Let

$$b_0(x, y) = \begin{cases} 1, & x < y < 0 \text{ or } 0 < x < y, \\ 0 & \text{otherwise.} \end{cases}$$

the function  $\Psi_0[a, b] = [a, b]$  is induced by  $\Phi_0(a) = a$ , and

$$\eta(x, y) = \begin{cases} x - y, & x < y < 0 \text{ or } 0 < x < y, \\ 0 & \text{otherwise.} \end{cases}$$

Let  $b_1 = 1$ , and let  $\Psi_1$  be induced by  $\Phi_1(a) = |a|$ . The point  $x^* = 1$  is a feasible solution. We can see that  $(F, g)$  satisfies the hypotheses of Theorem 4.2. Therefore  $x^* = 1$  is a nondominated solution.

The following example also shows the advantages of our method over [10] and [23, 28].

*Example 4.3*

$$\begin{aligned} \min F(x) &= [-2, 1]x^2, \quad x < 0, \\ \text{s.t. } g(x) &= x + 1 \leq 0. \end{aligned}$$

Then  $F(x)$  is  $gH$ -differentiable and weakly differentiable. Since  $F(x)$  is not  $LU$ -convex, the methods of [10] cannot be used, and since  $F^L(x) + F^U(x) = -x^2$  is not convex, the methods of [23, 28] cannot be used.

Let

$$b_0(x, y) = \begin{cases} 1, & x \leq y < 0, \\ \frac{-2y(x-y)}{-x^2+y^2}, & y < x < 0, \end{cases} \quad \text{and} \quad \Psi_0[a, b] = \begin{cases} [a, b], & [a, b] \leq 0, \\ \Psi([a, b]), & [a, b] \not\leq 0, \end{cases}$$

where  $\Psi([a, b])$  induced by  $\Phi(a) = |a|$ , and

$$\eta(x, y) = \begin{cases} x - y, & x < y < 0 \text{ or } 0 < x < y, \\ 0 & \text{otherwise.} \end{cases}$$

Let  $b_1(x, y) = 1$  and  $\Phi_1(a) = \Phi(a) = |a|$ . The point  $x^* = -1$  is a feasible solution. We can see that  $(F, g)$  satisfies the hypotheses of Theorem 4.3, and therefore  $x^* = -1$  is a nondominated solution.

**5 Conclusion**

The objective of this paper is to introduce the concept of  $gH$ -differentiable interval-valued univex mappings and discuss the relationship between interval-valued univex mappings and interval-valued weakly univex mappings. We derive sufficient optimality conditions for constrained interval-valued minimization problem under interval-valued univex mappings. In future work, we hope to give sufficient optimality conditions for a nondifferentiable interval-valued optimization problem under univexity hypotheses.

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**Competing interests**

The authors declare that they have no competing interests.

**Authors' contributions**

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