Open Mathematics
Research Article
Ting Xie and Zengtai Gong*

# Variational-like inequalities for $n$-dimensional fuzzy-vector-valued functions and fuzzy optimization 

https://doi.org/10.1515/math-2019-0050
Received November 25, 2018; accepted April 25, 2019


#### Abstract

The existing results on the variational inequality problems for fuzzy mappings and their applications were based on Zadeh's decomposition theorem and were formally characterized by the precise sets which are the fuzzy mappings' cut sets directly. That is, the existence of the fuzzy variational inequality problems in essence has not been solved. In this paper, the fuzzy variational-like inequality problems is incorporated into the framework of $n$-dimensional fuzzy number space by means of the new ordering of two $n$-dimensional fuzzy-number-valued functions we proposed [Fuzzy Sets and Systems 295 (2016) 19-36]. As a theoretical basis, the existence and the basic properties of the fuzzy variational inequality problems are discussed. Furthermore, the relationship between the variational-like inequality problems and the fuzzy optimization problems is discussed. Finally, we investigate the optimality conditions for the fuzzy multiobjective optimization problems.


Keywords: $n$-dimensional fuzzy-number-valued functions, generalized convexity, variational-like inequality, fuzzy optimization

## 1 Introduction

Variational inequality theory, where the function is a vector-valued mapping, known either in the form presented by Hartman and Stampacchia [1] or in the form introduced by Minty [2], has become an effective and powerful tool for studying a wide class of linear/nonlinear problems arising in diverse applied fields such as optimization and control, mechanics, economics and engineering sciences. Vector variational inequality, where the function is a matrix-valued mapping, was first introduced and studied by Giannessi [3] in finitedimensional Euclidean spaces. This is a generalization of a scalar variational inequality to the vector case by virtue of multi-criteria considering. In the study of problems related to stochastic impulse control, Bensoussan and Lions [4] proposed quasi-variational inequality [5-7], where the function is a set-valued mapping. However, one frequently observes that there are objects that have an ambiguous status in the real world. The fuzzy set theory, introduced by Zadeh [8] in 1965, offers a wide variety of techniques for analyzing imprecise data and fuzzy numbers [9] have been investigated extensively. In order to deal with the variational inequalities derived from some fuzzy environments, in 1989, Chang and Zhu [10] introduced the concepts of variational inequalities for fuzzy mapping in abstract spaces and investigated the existence of some types

[^0]of variational-like inequalities for fuzzy mappings. Since then, several types of variational inequalities and complementarity problems for fuzzy mappings have been studied by various researchers [11-18].

On the other hand, variational inequalities are efficient tool for the investigation of optimization problems because these inequalities ensure the existence of efficient solutions, under the condition of convexity or generalized convexity. Many works of these type of inequalities have been focused on looking for the relations between the solutions of various type of variational inequalities and optimization problems [19, 20]. While very few investigations have appeared to study the relationships between fuzzy variational inequalities and fuzzy optimization problems. Wu and Xu [21, 22] introduced the generalized monotonicity of fuzzy mappings and discussed the relationship between the fuzzy variational-like inequality and fuzzy optimization problems. Weir [23] and Noor [24, 25] have studied some basic properties of the preinvex functions and their role in optimization and variational-like inequality problems. In [24], Noor has pointed out that the concept of invexity plays exactly the same role in variational-like inequality problems as the classical convexity plays in variational inequality problems, and has shown that the variational-like inequality problems are well defined in the setting of invexity. Recently, Ruiz-Garzon et al. [26] established relationships between vector variational-like inequality and optimization problems, under the assumptions of pseudo-invexity. However, the exiting results on the variational inequalities for fuzzy mappings are focused on two methods. Since the cut set of a 1-dimensional fuzzy number is a close interval on $R$, one method is investigates the $n$-dimensional fuzzy-vector-valued function whose components are the 1-dimensional fuzzy numbers by means of the ordering of two fuzzy numbers proposed by Goetschel and Voxman [27] or by Nanda and Kar [28]; the other method is transformed into the classical set-valued variational inequalities, because the cut set of an $n$-dimensional fuzzy number is a nonempty compact convex subset of $R^{n}$. To the best of our knowledge, very few studies have investigated the variational inequalities for $n$-dimensional fuzzy numbervalued functions directly in $n$-dimensional fuzzy number space. The main reason is that there is almost no related research about the ordering and the difference of $n$-dimensional fuzzy numbers. Until 2016, Gong and Hai [29] introduced the concept of a convex fuzzy-number-valued function based on a new ordering $\preceq c$ of $n$-dimensional fuzzy numbers, and investigated some relations among the convexity and quasiconvex of $n$-dimensional fuzzy-number-valued functions, and also study the local-global minimum properties of the convex fuzzy number-valued functions. The present study is to incorporate the fuzzy variational-like inequality problems into the framework of $n$-dimensional fuzzy number space by the new ordering of two $n$ dimensional fuzzy numbers, which is a further study in theoretical research and more convenient in practical application.

The aim of this paper is to incorporate the fuzzy variational-like inequality problems into the framework of $n$-dimensional fuzzy number space. To make our analysis possible, we present the preliminary terminology used throughout this paper in Section 2. In Section 3, the concept of generalized monotonicity and invexity for $n$-dimensional fuzzy-number-valued functions are presented and some properties are discussed. In Section 4, we introduce the fuzzy variational-like inequality based on the order $\succeq_{c}$ and obtain the existence of a solution of the fuzzy variational-like inequality. The relationship between the variational-like inequality problems and fuzzy optimization problems is given in Section 5. We investigate the optimality conditions for the fuzzy multiobjective optimization problems in Section 6. Section 7 concludes this paper.

## 2 Preliminaries

Throughout this paper, $R^{n}$ denotes the $n$-dimensional Euclidean space, $\mathcal{K}^{n}$ and $\mathcal{K}_{C}^{n}$ denote the spaces of nonempty compact and compact convex sets of $R^{n}$, respectively. Let $\mathcal{F}\left(R^{n}\right)$ be the set of all fuzzy subsets on $R^{n}$. A fuzzy set $u$ on $R^{n}$ is a mapping $u: R^{n} \rightarrow[0,1]$, and $u(x)$ is the degree of membership of the element $x$ in the fuzzy set $u$. For each fuzzy set $u$, we denote its $r$-level set as $[u]^{r}=\left\{x \in R^{n}: u(x) \geq r\right\}$ for any $r \in(0,1]$, and in some references also denoted by $u_{r}$ for short. The support of $u$ we denote by supp $u$ where $\operatorname{supp} u=\left\{x \in R^{n}: u(x)>0\right\}$. The closure of supp $u$ defines the 0-level of $u$, i.e. $[u]^{0}=c l(\operatorname{supp} u)$. Here $c l(M)$ denotes the closure of set $M$. Fuzzy set $u \in \mathcal{F}\left(R^{n}\right)$ is called a fuzzy number if [30, 31]
(i) $u$ is a normal fuzzy set, i.e., there exists an $x_{0} \in R^{n}$ such that $u\left(x_{0}\right)=1$,
(ii) $u$ is a convex fuzzy set, i.e., $u(\lambda x+(1-\lambda) y) \geq \min \{u(x), u(y)\}$ for any $x, y \in R^{n}$ and $\lambda \in[0,1]$,
(iii) $u$ is upper semicontinuous,
(iv) $[u]^{0}=c l(\operatorname{supp} u)=\operatorname{cl}\left(\bigcup_{r \in(0,1]}[u]^{r}\right)$ is compact.

We use $E^{n}$ to denote the fuzzy number space. Note that if $u: R \rightarrow[0,1]$, then $u$ is a 1-dimensional fuzzy number, denoted by $u \in E$, and $[u]^{r}=\left[u_{-}(r), u^{+}(r)\right]$ is a close interval on $R$.

It is clear that each $u \in R^{n}$ can be considered as a fuzzy number $u$ defined by

$$
u(x)=\left\{\begin{array}{l}
1, x=u  \tag{2.1}\\
0, \text { otherwise }
\end{array}\right.
$$

In particular, the fuzzy number 0 is defined as $O(x)=1$ if $x=0$, and $O(x)=0$ otherwise.
Example 2.1. Let $u \in E^{2}$ is defined by

$$
u(x, y)= \begin{cases}\sqrt{1-x^{2}-y^{2}}, & x^{2}+y^{2} \leq 1  \tag{2.2}\\ 0, & \text { otherwise }\end{cases}
$$

then $[u]^{r}=\left\{(x, y): x^{2}+y^{2} \leq 1-r^{2}\right\}, r \in[0,1]$.
Theorem 2.2. [32] If $u \in E^{n}$, then
(i) $[u]^{r}$ is a nonempty compact convex subset of $R^{n}$ for any $r \in(0,1]$,
(ii) $[u]^{r_{1}} \subseteq[u]^{r_{2}}$, whenever $0 \leq r_{2} \leq r_{1} \leq 1$,
(iii) if $r_{n}>0$ and $r_{n}$ converging to $r \in[0,1]$ is nondecreasing, then $\bigcap_{n=1}^{\infty}[u]^{r_{n}}=[u]^{r}$.

Conversely, suppose for any $r \in[0,1]$, there exists an $A^{r} \subseteq R^{n}$ which satisfies the above (i)-(iii), then there exists a unique $u \in E^{n}$ such that $[u]^{r}=A^{r}, r \in(0,1],[u]^{0}=\overline{\bigcup_{r \in(0,1]}[u]^{r}} \subseteq A^{0}$.

Let $u, v \in E^{n}, k \in R$. For any $x \in R^{n}$, the addition and scalar multiplication can be defined, respectively, as:

$$
\begin{align*}
(u+v)(x)= & \sup _{s+t=x} \min \{u(s), v(t)\},  \tag{2.3}\\
& (k u)(x)=u\left(\frac{x}{k}\right), k \neq 0  \tag{2.4}\\
& (0 u)(x)=\left\{\begin{array}{l}
1, x=0 \\
0,
\end{array}, x \neq 0\right. \tag{2.5}
\end{align*}
$$

It is well known that for any $u, v \in E^{n}$ and $k \in R$, the addition $u+v$ and the scalar multiplication $k u$ have the level sets

$$
\begin{array}{r}
{[u+v]^{r}=[u]^{r}+[v]^{r}=\left\{x+y: x \in[u]^{r}, y \in[v]^{r}\right\},} \\
{[k u]^{r}=k[u]^{r}=\left\{k x: x \in[u]^{r}\right\} .} \tag{2.7}
\end{array}
$$

Proposition 2.3. [33] If $u, v \in E^{n}, k, k_{1}, k_{2} \in R$, then
(i) $k(u+v)=k u+k v$,
(ii) $k_{1}\left(k_{2} u\right)=\left(k_{1} k_{2}\right) u$,
(iii) $\left(k_{1}+k_{2}\right) u=k_{1} u+k_{2} u$ when $k_{1} \geq 0$ and $k_{2} \geq 0$.

Give two subsets $A, B \subseteq R^{n}$ and $k \in R$, the Minkowski difference is given by $A-B=A+(-1) B=\{a-b$ : $a \in A, b \in B\}$. However, in general, $A+(-A) \neq 0$, i.e. the opposite of $A$ is not the inverse of $A$ in Minkowski addition (unless $A=\{a\}$ is a singleton). The spaces $\mathcal{K}^{n}$ and $\mathcal{K}_{C}^{n}$ are not linear spaces since they do not contain inverse elements and therefore subtraction is not defined. To partially overcome this situation, Hukuhara [36] introduced the following $H$-difference $A \ominus B=C \Longleftrightarrow A=B+C$ and an important property of $\ominus$ is that $A \ominus A=\{0\}, \forall A \in R^{n}$ and $(A+B) \ominus B=A, \forall A, B \in R^{n}$. The $H$-difference is unique, but a necessary condition
for $A \ominus_{H} B$ to exist is that $A$ contains a translation $\{c\}+B$ of $B$. In order to overcome this situation, Stefanini [37] defined the generalized Hukuhara difference of two sets $A, B \in \mathcal{K}^{n}$ as follows

$$
A \ominus_{g H} B=C \Longleftrightarrow\left\{\begin{array}{r}
\text { (1) } A=B+C  \tag{2.8}\\
\text { or (2) } B=A+(-1) C .
\end{array}\right.
$$

The generalized Hukuhara difference has been extended to the fuzzy case in [38]. For any $u, v \in E^{n}$, the generalized Hukuhara difference ( $g H$-difference for short) is the fuzzy number $w$, if it exists, such that

$$
u \ominus_{g H} v=w \Longleftrightarrow\left\{\begin{align*}
\text { (1) } & u=v+w  \tag{2.9}\\
\text { or (2) } & v=u+(-1) w .
\end{align*}\right.
$$

It is possible that the $g H$-difference of two fuzzy numbers does not exist. To solve this shortcoming, in [39] a new difference between fuzzy numbers was proposed. Using the convex hull (conv) the new difference was defined as follows.

Definition 2.4. [39, 40] The generalized difference ( $g$-difference for short) of two fuzzy numbers $u, v \in E^{n}$ is given by its level sets as

$$
\begin{equation*}
\left[u \Theta_{g} v\right]^{r}=\operatorname{cl}\left(\operatorname{conv} \bigcup_{\beta \geq r}\left([u]^{\beta} \Theta_{g H}[v]^{\beta}\right)\right), \forall r \in[0,1], \tag{2.10}
\end{equation*}
$$

where the $g H$-difference $\Theta_{g H}$ is with interval operands $[u]^{\beta}$ and $[v]^{\beta}$.
A necessary condition for $u \Theta_{g} v$ to exist is that either [ $\left.u\right]^{r}$ contains a translation of $[v]^{r}$ or $[v]^{r}$ contains a translation of $[u]^{r}$ for any $r \in[0,1]$.

Proposition 2.5. [41] Let $u, v \in E^{n}$. Then
(i) if the $g$-difference exists, it is unique,
(ii) $u \ominus g u=0$,
(iii) $(u+v) \ominus_{g} v=u,(u+v) \ominus_{g} u=v$,
(iv) $u \ominus g v=-(v \ominus g u)$.

Given $u, v \in E^{n}$, the distance $D: E^{n} \times E^{n} \rightarrow[0,+\infty)$ between $u$ and $v$ is defined by the equation

$$
\begin{equation*}
D(u, v)=\sup _{r \in[0,1]} d\left([u]^{r},[v]^{r}\right) \tag{2.11}
\end{equation*}
$$

where $d$ is the Hausdorff metric given by

$$
\begin{aligned}
d\left([u]^{r},[v]^{r}\right) & =\inf \left\{\varepsilon:[u]^{r} \subset N\left([v]^{r}, \varepsilon\right),[v]^{r} \subset N\left([u]^{r}, \varepsilon\right)\right\} \\
& =\max \left\{\sup _{a \in[u]^{r}} \inf _{b \in[v]^{r}}\|a-b\|, \sup _{b \in[v]^{r}} \inf _{a \in[u]^{r}}\|a-b\|\right\}
\end{aligned}
$$

$N\left([u]^{r}, \varepsilon\right)=\left\{x \in R^{n}: d\left(x,[u]^{r}\right)=\inf _{y \in[u]^{r}} d(x, y) \leq \varepsilon\right\}$ is the $\varepsilon$-neighborhood of $[u]^{r}$. Then, $\left(E^{n}, D\right)$ is a complete metric space, and satisfies $D(u+w, v+w)=D(u, v), D(k u, k v)=|k| D(u, v)$ for any $u, v, w \in E^{n}$ and $k \in R$.

Let $S^{n-1}=\left\{x \in R^{n}:\|x\|=1\right\}$ be the unit sphere of $R^{n}$ and $\langle\cdot, \cdot\rangle$ be the inner product in $R^{n}$, i.e. $\langle x, y\rangle=$ $\sum_{i=1}^{n} x_{i} y_{i}$, where $x=\left(x_{1}, x_{2}, \cdots, x_{n}\right) \in R^{n}, y=\left(y_{1}, y_{2}, \cdots, y_{n}\right) \in R^{n}$. Suppose $u \in E^{n}, r \in[0,1]$ and $x \in S^{n-1}$, the support function of $u$ is defined by

$$
\begin{equation*}
u^{\star}(r, x)=\sup _{a \in[u]^{r}}\langle a, x\rangle . \tag{2.12}
\end{equation*}
$$

Theorem 2.6. [42] Suppose $u \in E^{n}, r \in[0,1]$, then

$$
\begin{equation*}
[u]^{r}=\left\{y \in R^{n}:\langle y, x\rangle \leq u^{\star}(r, x), x \in S^{n-1}\right\} . \tag{2.13}
\end{equation*}
$$

For $u \in E^{n}$, we denote the centroid of $[u]^{r}, r \in[0,1]$ as

$$
\left(\frac{\int \cdots \int_{[u]^{r^{2}}} x_{1} d x_{1} d x_{2} \cdots d x_{n}}{\int \cdots \int_{[u]^{r}} 1 d x_{1} d x_{2} \cdots d x_{n}}, \frac{\int \cdots \int_{[u]^{r}} x_{2} d x_{1} d x_{2} \cdots d x_{n}}{\int \cdots \int_{[u]^{r}} 1 d x_{1} d x_{2} \cdots d x_{n}}, \cdots, \frac{\int \cdots \int_{[u]^{r}} x_{n} d x_{1} d x_{2} \cdots d x_{n}}{\int \cdots \int_{[u]^{r}} 1 d x_{1} d x_{2} \cdots d x_{n}}\right),
$$

where $\int \cdots \int_{[u]^{r}} 1 d x_{1} d x_{2} \cdots d x_{n}$ is the solidity of $[u]^{r}, r \in[0,1]$ and $\int \cdots \int_{[u]^{r}} x_{i} d x_{1} d x_{2} \cdots d x_{n}(i=1,2, \cdots, n)$ is the multiple integral of $x_{i}$ on measurable sets $[u]^{r}, r \in[0,1]$. Next we define an order $\preceq_{C}$ for $E^{n}$.

Let $\tau: E^{n} \rightarrow R^{n}$ be a real vector-valued function defined by ([29])

$$
\begin{array}{r}
\tau(u)=\left(2 \int_{0}^{1} r \frac{\int \cdots \int_{[u]^{r}} x_{1} d x_{1} d x_{2} \cdots d x_{n}}{\int \cdots \int_{[u]^{r^{2}}} 1 d x_{1} d x_{2} \cdots d x_{n}} d r, 2 \int_{0}^{1} r \frac{\int \cdots \int_{[u]^{r}} x_{2} d x_{1} d x_{2} \cdots d x_{n}}{\int \cdots \int_{[u]^{r}} 1 d x_{1} d x_{2} \cdots d x_{n}} d r,\right. \\
\left.\cdots, 2 \int_{0}^{1} r \frac{\int \cdots \int_{[u]^{r}} x_{n} d x_{1} d x_{2} \cdots d x_{n}}{\int \cdots \int_{[u]^{r}} 1 d x_{1} d x_{2} \cdots d x_{n}} d r\right), \tag{2.14}
\end{array}
$$

where $\int_{0}^{1} r \frac{\int \cdots \int_{[u]^{r}} x_{i} d x_{1} d x_{2} \cdots d x_{n}}{\int \cdots \int_{[u]^{r}} 1 d x_{1} d x_{2} \cdots d x_{n}} d r(i=1,2, \cdots, n)$ is the Lebesgue integral of $r \frac{\int \cdots \int_{\left[\langle ]^{\prime}\right]^{2}} x_{i} d x_{1} d x_{2} \cdots d x_{n}}{\int \cdots \int_{[u]^{\prime}} 1 d x_{1} d x_{2} \cdots d x_{n}}(i=1,2, \cdots, n)$ on $[0,1]$. The vector-valued function $\tau$ is called a ranking value function defined on $E^{n}$.

Definition 2.7. [29] Let $u, v \in E^{n}, C \subseteq R^{n}$ be a closed convex cone with $0 \in C$ and $C \neq R^{n}$. We say that $u \preceq c v$ (u precedes $v$ ) if

$$
\begin{equation*}
\tau(v) \in \tau(u)+C . \tag{2.15}
\end{equation*}
$$

We say that $u \prec_{c} v$ if $u \preceq_{c} v$ and $\tau(u) \neq \tau(v)$. Sometimes we may write $v \succeq_{c} u$ (resp. $v \succ_{c} u$ ) instead of $u \preceq_{c} v$ (resp. $u \prec_{c} v$ ). In addition, $\widetilde{\varepsilon} \in E^{n}$ is said to be an arbitrary positive fuzzy-number if $\widetilde{\varepsilon} \succ_{c} 0\left(0 \in R^{n}\right)$ and $D(\widetilde{\varepsilon}, 0)<\varepsilon$, where $\varepsilon$ is an arbitrary positive real number.

Example 2.8. If $u, v \in E^{1}$, then $\tau(u)=\int_{0}^{1} r\left(u_{r}^{-}+u_{r}^{+}\right) d r, \tau(v)=\int_{0}^{1} r\left(v_{r}^{-}+v_{r}^{+}\right) d r$. Suppose $C=R^{+}=[0,+\infty) \subseteq R$, $u \preceq c v$ if and only if $\tau(u) \leq \tau(v)$, i.e., $\tau(v) \in \tau(u)+[0,+\infty)$. Therefore, when $u, v \in E^{1}$, Definition 2.7 coincides with the definition of ordering of $u, v$ proposed by Goetschel ([27]).

If $u, v \in E^{2}$, in Definition 2.7, let $C$ be the set of nonnegative orthant of $R^{2}$, i.e., $C=R^{2+}=\left\{\left(x_{1}, x_{2}\right) \in R^{2}\right.$ : $\left.\left.x_{1} \geqslant 0, x_{2} \geqslant 0\right)\right\} \subseteq R^{2}$.

Example 2.9. A special kind of n-dimension fuzzy numbers is the fuzzy n-cell numbers proposed in [43]. Let $u \in$ $L\left(E^{n}\right)$, i.e., $[u]^{r}=\prod_{i=1}^{n}\left[u_{i}^{-}(r), u_{i}^{+}(r)\right]=\left[u_{1}^{-}(r), u_{1}^{+}(r)\right] \times\left[u_{2}^{-}(r), u_{2}^{+}(r)\right] \times \cdots \times\left[u_{n}^{-}(r), u_{n}^{+}(r)\right]$ for any $r \in[0,1]$, where the left endpoint function and the right endpoint function $u_{i}^{-}(r), u_{i}^{+}(r) \in R$ with $u_{i}^{-}(r) \leq u_{i}^{+}(r)(i=1,2, \cdots, n)$, then we have

$$
\begin{equation*}
\tau(u)=\left(\int_{0}^{1} r\left(u_{1}^{-}(r)+u_{1}^{+}(r)\right) d r, \int_{0}^{1} r\left(u_{2}^{-}(r)+u_{2}^{+}(r)\right) d r, \cdots, \int_{0}^{1} r\left(u_{n}^{-}(r)+u_{n}^{+}(r)\right) d r\right) \tag{2.16}
\end{equation*}
$$

For $u, v \in L\left(E^{n}\right)$, suppose $\left.C=R^{n+}=\left\{\left(x_{1}, x_{2}, \cdots, x_{n}\right) \in R^{n}: x_{1} \geqslant 0, x_{2} \geqslant 0, \cdots, x_{n} \geqslant 0\right)\right\} \subseteq R^{n}$, then we have $u \succeq_{c} v \Longleftrightarrow \tau(u) \in \tau(v)+c$. Furthermore, for $k_{1}, k_{2} \in R$, we obtain

$$
\begin{equation*}
\tau\left(k_{1} u+k_{2} v\right)=k_{1} \tau(u)+k_{2} \tau(v) . \tag{2.17}
\end{equation*}
$$

Let $M$ be a convex set of $m$-dimensional Euclidean space $R^{m}$ and $F$ be an $n$-dimensional fuzzy-number-valued function (fuzzy-number-valued function for short) from $M$ into $E^{n}$.

Example 2.10. The following function is a 2-dimensional fuzzy-number-valued function. For constants $s, t \in R$, $F:\left[-\sqrt{-\ln \frac{1}{5}}, \sqrt{-\ln \frac{1}{5}}\right]^{2} \rightarrow E^{2}$ is defined as

$$
F(s, t)(x, y)= \begin{cases}\frac{5}{4} e^{-\left[(x-s)^{2}+(y-t)^{2}\right]}-\frac{1}{4}, & -\sqrt{-\ln \frac{1}{5}} \leq x, y \leq \sqrt{-\ln \frac{1}{5}}  \tag{2.18}\\ 0, & \text { otherwise } .\end{cases}
$$

Example 2.11. The following function is a fuzzy 1-cell number function. Furthermore, for constants $s \in R$, $F(s)(x)=f(s) u(x)$.

$$
F(s)(x)= \begin{cases}\frac{x+e^{s}}{2 e^{s}}, & -e^{s} \leq x \leq e^{s}  \tag{2.19}\\ \sqrt{\frac{2 e^{s}-x}{e^{s}}}, & e^{s} \leq x \leq 2 e^{s} \\ 0, & \text { otherwise }\end{cases}
$$

where $f(s)=e^{s}$, and

$$
u(x)= \begin{cases}\frac{x+1}{2}, & -1 \leq x \leq 1 \\ \sqrt{2-x}, & 1<x<2 \\ 0, & \text { otherwise }\end{cases}
$$

The epigraph of $F$, denoted by epi $(F)$, is defined as

$$
\begin{equation*}
\operatorname{epi}(F)=\left\{(t, u): t \in M, u \in E^{n}, F(t) \preceq_{c} u\right\} . \tag{2.20}
\end{equation*}
$$

For $u, v \in E^{n}$, we say that $u$ and $v$ are comparable, if either $u \preceq_{c} v$ or $v \preceq_{c} u$,; otherwise, they are non-comparable. $F$ is said to be a comparable fuzzy-number-valued function if for each pair $t_{1}, t_{2} \in M$ and $t_{1} \neq t_{2}, F\left(t_{1}\right)$ and $F\left(t_{2}\right)$ are comparable; otherwise, $F$ is said to be a non-comparable fuzzy-number-valued function.
$F$ is said to be lower semicontinuous (l.c.) at $t_{0} \in M$, if for any $\widetilde{\varepsilon} \succ_{c} 0$, there exists a neighborhood $U$ of $t_{0}$, when $t \in U$, we have $F\left(t_{0}\right) \prec{ }_{c} F(t)+\widetilde{\varepsilon} ; F$ is said to be upper semicontinuous (u.c.) at $t_{0} \in M$, if for any $\widetilde{\varepsilon} \succ_{c} 0$, there exists a neighborhood $U$ of $t_{0}$, when $t \in U$, we have $F(t) \prec_{c} F\left(t_{0}\right)+\widetilde{\varepsilon}$. $F$ is continuous at $t_{0} \in M$, if it is both l.c. and u.c. at $t_{0}$, and that it is continuous if and only if it is continuous at every point of $M$ ([29]).

Definition 2.12. ([29]) Let $F: M \rightarrow E^{n}$ be a fuzzy-number-valued function.
(1) An element $t_{0} \in M$ is called a local minimum point of $F$ if there exists a neighborhood $U$ of $t_{0}, F\left(t_{0}\right) \preceq_{c} F(t)$ for any $t \in U$.
(2) An element $t_{0} \in M$ is called a global minimum point of $F$ if $F\left(t_{0}\right) \preceq_{c} F(t)$ for any $t \in M$.
(3) An element $t_{0} \in M$ is called a strictly local minimum point of $F$ if there exists a neighborhood $U$ of $t_{0}$, $F\left(t_{0}\right) \prec_{c} F(t)$ for any $t \in U$ and $t \neq t_{0}$.
(4) An element $t_{0} \in M$ is called a strictly global minimum point of $F$ if $F\left(t_{0}\right) \prec_{c} F(t)$ for any $t \in M$ and $t \neq t_{0}$.

Definition 2.13. Let $\mathbf{A}=\left(u_{1}, u_{2}, \cdots, u_{n}\right) \in\left(E^{n}\right)^{n}, u_{i} \in E^{n}, i=1,2, \cdots, n$, and $T=\left(t_{1}, t_{2}, \cdots, t_{n}\right) \in R^{n}$ be an n-dimensional fuzzy vector and an n-dimensional real vector, respectively. We define the product of a fuzzy vector with a real vector as $T \mathbf{A}=\sum_{i=1}^{n} t_{i} u_{i}$, which is an $n$-dimensional fuzzy number. In addition, if $T \mathbf{A}=0$, then we say $\mathbf{A}$ is fuzzy orthogonal to $T$.

We denote the fuzzy vector $\mathbf{0}$ by $\mathbf{0}=\{\underbrace{0,0, \cdots, 0}_{n}\}$, where $0 \in E^{n}$. If $\mathbf{A}=\left(u_{1}, u_{2}, \cdots, u_{n}\right) \in(E)^{n}, u_{i} \in E^{1}, i=$ $1,2, \cdots, n$, then Definition 2.13 coincides with Definition 2.4 proposed in [34]. It is not difficult to obtain

$$
\begin{equation*}
[T \mathbf{A}]^{r}=\bigcup_{i=1}^{n} t_{i}\left[u_{i}\right]^{r}=\bigcup_{i=1}^{n}\left\{t_{i} x_{i}: x \in\left[u_{i}\right]^{r}\right\} \tag{2.21}
\end{equation*}
$$

For any $n$-dimensional fuzzy vectors $\mathbf{X}$ and $\mathbf{Y}$, let $\mathbf{X}=\left\{X_{1}, X_{2}, \cdots, X_{n}\right\}, \mathbf{Y}=\left\{Y_{1}, Y_{2}, \cdots, Y_{n}\right\}$, we use the following convention for equalities and inequalities throughout the paper:
(a) $\mathbf{X} \leq \mathbf{Y} \Longleftrightarrow x_{i} \preceq c y_{i}, i=1,2, \cdots n$, with strict inequality holding for at least one $i$;
(b) $\mathbf{X} \leqq \mathbf{Y} \Longleftrightarrow x_{i} \preceq_{c} y_{i}, i=1,2, \cdots n$;
(c) $\mathbf{X}=\mathbf{Y} \Longleftrightarrow x_{i}=c y_{i}, i=1,2, \cdots n$;
(d) $\mathbf{X}<\mathbf{Y} \Longleftrightarrow x_{i} \prec_{c} y_{i}, i=1,2, \cdots n$.

In the following, we assume that the fuzzy-number-valued function $F: M \rightarrow E^{n}$ and fuzzy-vector-valued function $\mathbf{F}: M \rightarrow\left(E^{n}\right)^{n}$ are comparable, respectively.

Definition 2.14. Let $\mathbf{F}: M \rightarrow\left(E^{n}\right)^{n}$ be an n-dimensional fuzzy-vector-valued function, denoted by $\mathbf{F}(t)=$ $\left(u_{1}(t), u_{2}(t), \cdots, u_{n}(t)\right)$, where $u_{i}(t)(i=1,2, \cdots, n)$ is a fuzzy-number-valued function on $M$. For the sake of brevity, $F$ is called a fuzzy-vector-valued function.
(1) $\mathbf{F}$ is said to be a comparable fuzzy-vector-valued function if any $u_{i}(t)(i=1,2, \cdots, n)$ is a comparable fuzzy-number-valued function.
(2) For $s, t \in M$, we define $g$-difference of fuzzy-vector-valued functions as

$$
\begin{equation*}
\mathbf{F}(s) \Theta_{g} \mathbf{F}(t)=\left(u_{1}(s) \Theta_{g} u_{1}(t), u_{2}(s) \Theta_{g} u_{2}(t) \cdots, u_{n}(s) \Theta_{g} u_{n}(t)\right) . \tag{2.22}
\end{equation*}
$$

Example 2.15. Let $f(t)=\left(t_{1}, t_{2}, t_{3}\right)=\left(e^{t}, \frac{e^{t}\left(e^{t}-1\right)}{3}, \frac{e^{t}}{3}\right) \in R^{3}, t \in R$, be a 3-dimensional real-vector-valued function, and $\mathbf{u}=\left(u_{1}, u_{2}, u_{3}\right) \in(E)^{3}\left(u_{i} \in E, i=1,2,3\right)$ be a 3-dimensional fuzzy-vector-valued function, where

$$
\begin{gathered}
u_{1}= \begin{cases}x+2, & -2 \leq x \leq-1, \\
1, & -1 \leq x \leq 0 \\
0, & \text { otherwise },\end{cases} \\
u_{2}= \begin{cases}1, & 0 \leq x \leq 1, \\
0, & \text { otherwise },\end{cases}
\end{gathered}
$$

and

$$
u_{3}= \begin{cases}1-x, & 0 \leq x \leq 1 \\ 0, & \text { otherwise }\end{cases}
$$

Then according to Definition 2.13, we have the following 1-dimensional fuzzy-number-valued function $F$ : $(0, \infty)^{2} \rightarrow E$ and

$$
F(t)(x)=f \mathbf{u}=f_{1} u_{1}+f_{2} u_{2}+f_{3} u_{3} \begin{cases}\frac{2 e^{t}+x}{e^{t}}, & -2 e^{t} \leq x \leq-e^{t} \\ 1, & -e^{t} \leq x \leq \frac{e^{2 t}-e^{t}}{3} \\ \frac{2 e^{t}-3 x}{e^{t}}, & \frac{e^{2 t}-e^{t}}{3} \leq x \leq \frac{e^{2 t}}{3} \\ 0, & \text { otherwise }\end{cases}
$$

and

$$
\begin{aligned}
F_{r}(t) & =\left[e^{t}(r-2), \frac{e^{t}\left(e^{t}-r\right)}{3}\right] \\
& =e^{t}[r-2,0]+\frac{e^{t}\left(e^{t}-1\right)}{3}[0,1]+\frac{e^{t}}{3}[0,1-r], \forall r \in[0,1]
\end{aligned}
$$

Theorem 2.16. Let $\mathbf{F}, \mathbf{G} \in\left(E^{n}\right)^{n}$. Then
(i) if the $g$-difference exists, it is unique,
(ii) $\mathbf{F} \ominus_{g} \mathbf{F}=\mathbf{0}$,
(iii) $(\mathbf{F}+\mathbf{G}) \ominus_{g} \mathbf{G}=\mathbf{F},(\mathbf{F}+\mathbf{G}) \ominus_{\mathrm{g}} \mathbf{F}=\mathbf{G}$,
(iv) $\mathbf{F} \ominus_{g} \mathbf{G}=-\left(\mathbf{G} \ominus_{g} \mathbf{F}\right)$.

Proof. It is not difficult to obtain from Proposition 2.5 and Definition 2.14.
Definition 2.17. Let $\mathbf{F}: M \rightarrow\left(E^{n}\right)^{n}$ be a fuzzy-vector-valued function, denoted by $\mathbf{F}(t)=$ $\left(u_{1}(t), u_{2}(t), \cdots, u_{n}(t)\right)$, where $u_{i}(t)(i=1,2, \cdots, n)$ is a fuzzy-number-valued function on $M$.
(1) $\mathbf{F}$ is said to be lower semicontinuous (l.c.) at $t_{0} \in M$ if there exists a neighborhood $U$ of $t_{0}$, any $u_{i}(t)(i=$ $1,2, \cdots, n$ ) is l.c. at $t_{0}$.
(2) $\mathbf{F}$ is said to be upper semicontinuous (u.c.) at $t_{0} \in M$ if there exists a neighborhood $U$ of $t_{0}$, any $u_{i}(t)(i=$ $1,2, \cdots, n)$ is u.c. at $t_{0}$.
A fuzzy-vector-valued function $\mathbf{F}: M \rightarrow\left(E^{n}\right)^{n}$ is continuous at $t_{0} \in M$, if it is both l.c. and u.c. at $t_{0}$, and that it is continuous if and only if it is continuous at every point of $M$.

Definition 2.18. Let $F: M \rightarrow E^{n}$ be a fuzzy-number-valued function, $t_{0}=\left(t_{1}^{0}, t_{2}^{0}, \cdots, t_{m}^{0}\right) \in \operatorname{int} M$. If $g$ difference $F\left(t_{1}^{0}, \cdots, t_{j}^{0}+h, \cdots, t_{m}^{0}\right) \ominus_{g} F\left(t_{1}^{0}, \cdots, t_{j}^{0}, \cdots, t_{m}^{0}\right)$ exists and there exists $u_{j} \in E^{n}(j=1,2, \cdots, m)$, such that

$$
\lim _{h \rightarrow 0} \frac{F\left(t_{1}^{0}, \cdots, t_{j}^{0}+h, \cdots, t_{m}^{0}\right) \ominus_{g} F\left(t_{1}^{0}, \cdots, t_{j}^{0}, \cdots, t_{m}^{0}\right)}{h}=u_{j}
$$

then we say that $F$ has the jth partial generalized derivative ( $g$-derivative for short) at $t_{0}$, denoted by $u_{j}=\partial F / \partial t_{j}^{0}$. Here the limit is taken in the metric space $\left(E^{n}, D\right)$. If all the partial g-derivatives at $t_{0}$ exist, then we say $F$ is said to be generalized differentiable ( $g$-differentiable for short) on $t_{0}$. If $F$ is $g$-differentiable at any interior point of $M$, then $F$ is said to be $g$-differentiable on $M$. The fuzzy vector $\left(u_{1}, u_{2}, \cdots, u_{m}\right) \in\left(E^{n}\right)^{m}$ is said to be the gradient of $F$ at $t_{0}$, denoted by $\nabla F\left(t_{0}\right)$, that is,

$$
\nabla F\left(t_{0}\right)=\left(u_{1}, u_{2}, \cdots, u_{m}\right)=\left(\partial F / \partial t_{1}^{0}, \partial F / \partial t_{2}^{0}, \cdots, \partial F / \partial t_{m}^{0}\right)
$$

In addition, $t_{0} \in M$ is said to be a stationary point of $F$ if $\nabla F\left(t_{0}\right)=\mathbf{0}$.
Note that if $M=[a, b]$, then Definition 2.18 coincides with the definition of $F$ is $g$-differentiable on $[a, b]$ proposed by Gong and Hai ([41]).

We call $\mathbf{F}:[a, b] \rightarrow\left(L\left(E^{n}\right)\right)^{n}$, denoted by $\mathbf{F}=\left(F_{1}, F_{2}, \cdots, F_{n}\right)$, is an $n$-dimensional fuzzy $n$-cell vectorvalued function (fuzzy $n$-cell vector-valued function for short). If $\mathbf{F}=\left(f_{1}(t) u_{1}, f_{2}(t) u_{2}, \cdots, f_{n}(t) u_{n}\right)$, where $f_{i}:[a, b] \rightarrow R, u_{i} \in L\left(E^{n}\right), i=1,2, \cdots n$, then the gradient of $\mathbf{F}$ at $t_{0}$ is defined as

$$
\nabla \mathbf{F}\left(t_{0}\right)=\left(\nabla F_{1}, \nabla F_{2}, \cdots, \nabla F_{n}\right)
$$

and it is not difficult to obtain $\nabla F_{i}=\left(u_{i} \partial f_{i} / \partial t_{1}^{0}, u_{i} \partial f_{i} / \partial t_{2}^{0}, \cdots, u_{i} \partial f_{i} / \partial t_{m}^{0}\right), i=1,2, \cdots n$.
Definition 2.19. The function $\eta: M \times M \rightarrow R^{n}$ is said to be a skew function if

$$
\begin{equation*}
\eta(x, y)=-\eta(y, x), \quad \forall x, y \in M \tag{2.23}
\end{equation*}
$$

Definition 2.20. [35] An n-dimensional fuzzy set $u$ is a fuzzy cone if $u(\gamma x)=u(x)$ for all $\gamma>0$ and $x \in R^{n}$.
Definition 2.21. Let $\mathbf{A}=\left(u_{1}, u_{2}, \cdots, u_{n}\right) \in\left(E^{n}\right)^{n}\left(u_{i} \in E^{n}, i=1,2, \cdots, n\right)$ be an $n$-dimensional fuzzy vector. A fuzzy dual cone of $\mathbf{A}$ is the n-dimensional fuzzy vector $\mathbf{A}^{\star}$ given by

$$
\begin{equation*}
\mathbf{A}^{\star}(y)=\left(\inf _{x \in R^{n}: x y<0}\left(1-u_{1}(x)\right), \inf _{x \in R^{n}: x y<0}\left(1-u_{2}(x)\right), \cdots, \inf _{x \in R^{n}: x y<0}\left(1-u_{n}(x)\right)\right) \tag{2.24}
\end{equation*}
$$

for nonzero $y \in R^{n}$, and $\mathbf{A}^{\star}(0)=(\underbrace{1,1, \cdots, 1}_{n})$.
Notice that if $\mathbf{A}=(u) \in E^{n}$ is a 1-dimensional fuzzy vector, i.e., an $n$-dimensional fuzzy number, then Definition 2.21 reduces to Definition 8 proposed in [35].

## 3 Generalized convex fuzzy-number-valued functions

It is well known that the role of generalized monotonicity of the operator in vector variational inequality problems corresponds to the role of generalized convexity of the objective function in the optimization problem. In this section, we generalize convexity from vector-valued maps to fuzzy number-valued functions. The concepts of invexity and generalized monotonicity for $n$-dimensional fuzzy-number-valued functions are presented and some relative properties are discussed. In the following, suppose $M \subseteq R^{n}$ be a convex set.

Definition 3.1. The mapping $\mathbf{F}: M \rightarrow\left(E^{n}\right)^{n}$ is said to be
(1) fuzzy monotone over $M$ if

$$
\begin{equation*}
(y-x)\left(\mathbf{F}(y) \Theta_{g} \mathbf{F}(x)\right) \succeq_{c} 0, \forall x, y \in M \tag{3.1}
\end{equation*}
$$

(2) fuzzy invex monotone over $M$ if there exists a continuous map $\eta: M \times M \rightarrow R^{n}$ such that

$$
\begin{equation*}
\eta(y, x)\left(\mathbf{F}(y) \Theta_{g} \mathbf{F}(x)\right) \succeq_{c} 0, \forall x, y \in M . \tag{3.2}
\end{equation*}
$$

Note that this definition reduces to the definition of monotone functions if $\eta(y, x)=y-x$.
(3) fuzzy strictly invex monotone over $M$ if there exists a continuous map $\eta: M \times M \rightarrow R^{n}$ such that

$$
\begin{equation*}
\eta(y, x)\left(\mathbf{F}(y) \Theta_{g} \mathbf{F}(x)\right) \succeq_{c} 0, \forall x, y \in M, x \neq y \tag{3.3}
\end{equation*}
$$

Definition 3.2. A g-differentiable fuzzy mapping $F: M \rightarrow E^{n}$ is called
(1) fuzzy invex (FIX) with respect to a function $\eta: M \times M \rightarrow R^{n}$, if for all $x, y \in M$

$$
\begin{equation*}
F(x) \Theta_{g} F(y) \succeq_{c} \eta(x, y) \nabla F(y) \tag{3.4}
\end{equation*}
$$

(2) fuzzy strictly invex (FSIX) with respect to a function $\eta: M \times M \rightarrow R^{n}$, if for all $x, y \in M$

$$
\begin{equation*}
F(x) \Theta_{g} F(y) \succ_{c} \eta(x, y) \nabla F(y), \quad \forall x \neq y \tag{3.5}
\end{equation*}
$$

(3) fuzzy incave (FIC) with respect to a function $\eta: M \times M \rightarrow R^{n}$, if for all $x, y \in M$

$$
\begin{equation*}
F(x) \Theta_{g} F(y) \preceq_{c} \eta(x, y) \nabla F(y) \tag{3.6}
\end{equation*}
$$

(4) fuzzy strictly incave (FSIC) with respect to a function $\eta: M \times M \rightarrow R^{n}$, if for all $x, y \in M$

$$
\begin{equation*}
F(x) \ominus_{g} F(y) \prec_{c} \eta(x, y) \nabla F(y), \quad \forall x \neq y \tag{3.7}
\end{equation*}
$$

Theorem 3.3. The function $F: M \rightarrow E^{n}$ will be a fuzzy invex fuzzy-number-valued function with respect to some function $\eta$ if and only if each stationary point of $F$ is a global minimum point.

Proof. Necessity. Let $F$ be a fuzzy invex fuzzy-number-valued function with respect to some function $\eta$. If $x_{0}$ is a stationary point of $F$, then $\nabla F\left(x_{0}\right)=\mathbf{0}$. Since $F$ is fuzzy invex, using (3.4), we have $F(x) \Theta_{g} F\left(x_{0}\right) \succeq_{c}$ $\eta(x, y) \nabla F\left(x_{0}\right)=0, \forall x \in M$. Thus, we obtain $F\left(x_{0}\right) \preceq_{c} F(x), \forall x \in M$. Therefore, $x_{0}$ is a global minimum point.

Sufficiency. If $y$ is a stationary point of $F$, i.e., $\nabla F=\mathbf{0}$, and also a global minimum point of $F$, then for a function $\eta: M \times M \rightarrow R^{n}$, we have

$$
F(x) \ominus_{g} F(y) \succeq_{c} 0=\eta(x, y) \nabla F, \quad \forall x \in M .
$$

Therefore, $F$ is a fuzzy invex fuzzy-number-valued function.
Theorem 3.4. If a g-differentiable fuzzy mapping $F: M \rightarrow E^{n}$ is fuzzy invex on $M$ with respect to $\eta: M \times M \rightarrow$ $R^{n}$ and $\eta$ is a skew function. Then, $\nabla F: M \rightarrow\left(E^{n}\right)^{n}$ is fuzzy invex monotone with respect to the same $\eta$.

Proof. By the fuzzy invexity of $F$, there exists $\eta(x, y) \in R^{n}$, such that

$$
F(x) \Theta_{g} F(y) \succeq_{c} \eta(x, y) \nabla F(y), \quad \forall x, y \in M
$$

By changing $x$ for $y$,

$$
F(y) \Theta_{g} F(x) \succeq_{c} \eta(y, x) \nabla F(x)
$$

Adding the above two formulas, we obtain

$$
0 \succeq_{c} \eta(x, y) \nabla F(y)+\eta(y, x) \nabla F(x)
$$

Since $\eta$ is a skew function, $\eta(y, x)=-\eta(x, y)$, thus, we have

$$
\eta(y, x) \nabla F(y) \Theta_{g} \eta(y, x) \nabla F(x) \succeq_{c} 0
$$

that is,

$$
\eta(y, x)\left(\nabla F(y) \Theta_{g} \nabla F(x)\right) \succeq_{c} 0
$$

Therefore, $\nabla F$ is fuzzy invex monotone.

Theorem 3.5. If a g-differentiable fuzzy mapping $F: M \rightarrow E^{n}$ is fuzzy strictly invex on $M$ with respect to $\eta: M \times M \rightarrow R^{n}$ and $\eta$ is a skew function. Then, $\nabla F: M \rightarrow\left(E^{n}\right)^{n}$ is fuzzy strictly invex monotone with respect to the same $\eta$.

Proof. By the fuzzy invexity of $F$, there exists $\eta(x, y) \in R^{n}$, such that

$$
F(x) \Theta_{g} F(y) \succ_{c} \eta(x, y) \nabla F(y), \quad \forall x, y \in M .
$$

By changing $x$ for $y$,

$$
F(y) \Theta_{g} F(x) \succ_{c} \eta(y, x) \nabla F(x) .
$$

Adding the above two formulas, we obtain

$$
\eta(x, y) \nabla F(y)+\eta(y, x) \nabla F(x) \prec_{c} 0 .
$$

Since $\eta$ is a skew function, $\eta(y, x)=-\eta(x, y)$, thus, we have

$$
\eta(y, x) \nabla F(y) \Theta_{g} \eta(y, x) \nabla F(x) \succ_{c} 0
$$

that is,

$$
\eta(y, x)\left(\nabla F(y) \Theta_{g} \nabla F(x)\right) \succ_{c} 0
$$

Therefore, $\nabla F$ is fuzzy strictly invex monotone.

## 4 Variational-like inequalities for fuzzy-vector-valued functions

The existing results on the variational inequality problems for fuzzy mappings and their applications were based on Zadeh's decomposition theorem and were formally characterized by the precise sets which are the fuzzy mappings' cut sets directly. In this section, the fuzzy variational-like inequality problems is incorporated into the framework of $n$-dimensional fuzzy number space and proposed by means of the new ordering of two $n$-dimensional fuzzy-number-valued functions we proposed in [29]. In addition, we give the extension principle of the fuzzy variational inequality problems.

Theorem 4.1. (Decomposition theorem)[39] If $u \in E^{n}$, then

$$
\begin{equation*}
u=\bigcup_{\lambda \in[0,1]}\left(\lambda \cdot[u]^{\lambda}\right) \tag{4.1}
\end{equation*}
$$

Let $f: M \rightarrow E^{n}$ be a fuzzy-number-valued function, then $\forall x \in M,[f(x)]^{\alpha}=f_{\alpha}(x)=f(x)(\alpha)=f(x, \alpha)=\{x \in$ $\left.R^{n}: f(x) \geqslant \alpha\right\}, \alpha \in[0,1]$, denotes the $\alpha$-cut set of $f$. According to Theorem 2.2, $\forall x \in M, f_{\alpha}(x) \subseteq \mathscr{K}_{C}^{n} \subseteq 2^{R^{n}}$, where $2^{R^{n}}$ is the family of all nonempty subsets of $R^{n}$.

Definition 4.2. Let $M$ be a closed and convex set in $R^{m}$. Given a continuous mapping $\eta: M \times M \rightarrow R^{n}$.
(1) The variational-like inequality problem for $n$-dimensional fuzzy mappings (fuzzy variational-like inequality problem for short), denoted by $\operatorname{FVLIP}(M, \mathbf{F}, \eta)$, is to find $x \in M$ such that

$$
\begin{equation*}
\eta(x, y) \mathbf{F}(x) \succeq_{c} 0, \quad \forall y \in M \tag{4.2}
\end{equation*}
$$

where $\mathbf{F}: M \rightarrow\left(E^{n}\right)^{n}$ is a continuous fuzzy-vector-valued mapping.
(2) The variational inequality problem for n-dimensional fuzzy mappings (fuzzy variational inequality problem for short), denoted by $\operatorname{FVIP}(M, \mathbf{F})$, is to find $x \in M$ such that

$$
\begin{equation*}
(y-x) \mathbf{F}(x) \succeq_{c} 0, \quad \forall y \in M \tag{4.3}
\end{equation*}
$$

where $\mathbf{F}: M \rightarrow\left(E^{n}\right)^{n}$ is a continuous fuzzy-vector-valued mapping.
(3) The generalized variational-like inequality problem for n-dimensional fuzzy mappings (generalized fuzzy variational-like inequality problem for short), denoted by $\operatorname{GFVLIP}(M, \mathbf{F}, \eta)$, is to find $x \in M$ with $\mathbf{x}^{*} \in \mathbf{F}(x)$ such that

$$
\begin{equation*}
\eta(x, y) \mathbf{x}^{\star} \succeq_{c} 0, \quad \forall y \in M \tag{4.4}
\end{equation*}
$$

where $\mathbf{F}: M \rightarrow 2^{\left(E^{n}\right)^{n}}$ is a continuous set-valued fuzzy vector mapping, and $2^{\left(E^{n}\right)^{n}}$ is the family of all nonempty subsets of $\left(E^{n}\right)^{n}$.
(4) The generalized variational inequality problem for n-dimensional fuzzy mappings (generalized fuzzy variational inequality problem for short), denoted by $\operatorname{GFVIP}(M, \mathbf{F})$, is to find $x \in M$ with $\mathbf{x}^{\star} \in \mathbf{F}(x)$ such that

$$
\begin{equation*}
(y-x) \mathbf{x}^{\star} \succeq_{c} 0, \quad \forall y \in M \tag{4.5}
\end{equation*}
$$

where $\mathbf{F}: M \rightarrow 2^{\left(E^{n}\right)^{n}}$ is a continuous set-valued fuzzy vector mapping, and $2^{\left(E^{n}\right)^{n}}$ is the family of all nonempty subsets of $\left(E^{n}\right)^{n}$.

Here we would like to point out that (FVLIP) and (FVIP) include many kinds of variational inequality problems as their special cases. For example,
(i) If $\mathbf{F}: M \rightarrow R^{n}$ is a continuous real-vector-valued mapping, and $C=R^{+}=[0, \infty)$, then (4.3) reduces to the classical variational inequality problem: to finding $x \in K$ such that

$$
\begin{equation*}
(y-x) F(x) \geq 0, \quad \forall y \in M \tag{4.6}
\end{equation*}
$$

which was considered by Stampacchia [1].
(ii) If $\mathbf{F}: M \rightarrow 2^{R^{n}}$ is a continuous set-valued real-vector mapping, then (4.5) reduces to the classical generalized variational inequality problem: to finding $x \in M$ with $x^{\star} \in \mathbf{F}(x)$ such that

$$
\begin{equation*}
(y-x) x^{\star} \geq 0, \quad \forall y \in M \tag{4.7}
\end{equation*}
$$

This problem was considered and studied by Noor [24].
Suppose for any $r \in[0,1]$, there exists an $A^{r} \subseteq R^{n}$ which satisfies the conditions (i)-(iii) in Theorem 2.2, then there exists a unique $F \in E^{n}$ such that $[F]^{r}=A^{r}, r \in(0,1],[F]^{0}=\overline{\bigcup_{r \in(0,1]}[F]^{r}} \subseteq A^{0}$. We denote $A=\left\{A^{r}: r \in[0,1]\right\}$, then $A \subseteq \mathcal{K}_{C}^{n} \subseteq 2^{R^{n}}$.
(iii) Let $G: M \rightarrow A$ be a set-valued mapping. Now we define a 1-dimensional fuzzy-vector-valued mapping F by

$$
\mathbf{F}: M \rightarrow \mathcal{F}(A), \quad x \mapsto r \cdot \chi_{A}(x) \subseteq\left(E^{n}\right)^{1}
$$

where $\chi_{A}(x)=\left\{\begin{array}{l}1, x \in A, \\ 0, x \notin A,\end{array}\right.$ is the characteristic function of the set $A$. Then (4.3) is equivalent to the variational inequality for fuzzy mapping, which was considered and studied by Noor [17], i.e., is to find $x \in M$ with $x^{\star} \in G(x)$ such that

$$
\begin{equation*}
(y-x) x^{\star} \geq 0, \quad \forall y \in M \tag{4.8}
\end{equation*}
$$

(iv) Let $G: M \rightarrow A$ be a set-valued mapping. Now we define a 1-dimensional fuzzy-vector-valued mapping F by

$$
\mathbf{F}: M \rightarrow \mathcal{F}(A), \quad x \mapsto r \cdot \chi_{A}(x) \subseteq\left(E^{n}\right)^{1}
$$

where $\chi_{A}(x)=\left\{\begin{array}{l}1, x \in A, \\ 0, x \notin A,\end{array}\right.$ is the characteristic function of the set $A$. Then (4.2) is equivalent to the variational-like inequality for fuzzy mapping,which was considered and studied by Rufián-Lizana [44], i.e., is to find $x \in M$ with $x^{\star} \in G(x)$ such that

$$
\begin{equation*}
\eta(x, y) x^{\star} \geq 0, \quad \forall y \in M \tag{4.9}
\end{equation*}
$$

Let $a: R^{n} \rightarrow[0,1]$ be a function, we have $f_{a(x)}=\left\{x \in R^{n}: f(x) \geqslant a(x)\right\} . \forall x \in R^{n}$, suppose for any $a(x) \in[0,1]$, there exists an $A^{a(x)} \subseteq R^{n}$ which satisfies the conditions (i)-(iii) in Theorem 2.2, then there exists a unique $F \in E^{n}$ such that $[F]^{a(x)}=A^{a(x)}, a(x) \in(0,1]$, and $[F]^{0}=\overline{\bigcup_{a(x) \in(0,1]}[F]^{a(x)}} \subseteq A^{0}$. We denote $A=\left\{A^{a(x)}: a(x) \in[0,1]\right\}$, then $A \subseteq \mathcal{K}_{C}^{n} \subseteq 2^{R^{n}}$.
(v) Let $G: M \rightarrow A$ be a set-valued mapping. Now we define a 1-dimensional fuzzy-vector-valued mapping F by

$$
\mathbf{F}: M \rightarrow \mathcal{F}(A), \quad x \mapsto a(x) \cdot \chi_{A}(x) \subseteq\left(E^{n}\right)^{1}
$$

where $\chi_{A}(x)=\left\{\begin{array}{l}1, x \in A, \\ 0, x \notin A,\end{array}\right.$ is the characteristic function of the set $A$. Then (4.2) is equivalent to the variational inequality for fuzzy mapping: is to find $x \in M$ with $\mathbf{x}^{\star} \in \mathbf{F}_{a(x)}(x)$ such that

$$
\begin{equation*}
(y-x) \mathbf{x}^{\star} \succeq_{c} 0, \quad \forall y \in M \tag{4.10}
\end{equation*}
$$

This problem was studied by Huang [16], where the cut $a(x)$ depends on $x$. It is slightly different from that of Noor [17], where the cut is a constant. The advantages are that the cuts have more freedom than those of Noor, and the model includes that of Noor as a special case in the viewpoint of mathematics.

Remark 4.3. Let $M$ be a convex cone in $R^{m}$ and $\mathbf{F}: M \rightarrow\left(E^{n}\right)^{n}$. The fuzzy variational-like inequality problem is called complementarity-like problem, denoted by $\operatorname{NCLP}(\mathbf{F})$. The $\operatorname{NCLP}(\mathbf{F})$ is an important special case of $\operatorname{FVILP}(M, \mathbf{F}, \eta)$. That is, the $\operatorname{NCLP}(\mathbf{F})$ is to find $x^{\star} \in M$ such that

$$
\begin{equation*}
\mathbf{F}\left(x^{\star}\right) \in \mathbf{F}^{\star}, \eta\left(x^{\star}\right) \mathbf{F}\left(x^{\star}\right)=0 \tag{4.11}
\end{equation*}
$$

where $\mathbf{F}^{\star}$ denotes the fuzzy dual cone of $\mathbf{F}$, i.e.,

$$
\mathbf{F}^{\star}(y)=\left(\inf _{x \in D: x y<0}\left(1-u_{1}(x)\right), \inf _{x \in M: x y<0}\left(1-u_{2}(x)\right), \cdots, \inf _{x \in M: x y<0}\left(1-u_{n}(x)\right)\right), y \in M
$$

Remark 4.4. If $\mathbf{F}: M \rightarrow\left(L\left(E^{n}\right)\right)^{n}$, the fuzzy variational-like inequality problem is called the fuzzy box constrained variational-like inequality problem, denoted by $\operatorname{FBVLIP}(M, \mathbf{F}, \eta)$.

Example 4.5. If $\mathbf{F}: M \rightarrow\left(L\left(E^{n}\right)\right)^{n}$ be fuzzy $n$-cell vector-valued function, then (FBVLIP) is to find $x \in M$ such that

$$
\begin{equation*}
\eta(x, y) F(x) \succeq_{c} 0, \quad \forall y \in M \tag{4.12}
\end{equation*}
$$

which is equivalent to

$$
\begin{equation*}
\tau(\eta(x, y) F(x)) \in \tau(0)+C, \quad \forall y \in M \tag{4.13}
\end{equation*}
$$

where $C=R^{n+}=\left\{\left(x_{1}, x_{2}, \cdots, x_{n}\right) \in R^{n}: x_{1} \geqslant 0, x_{2} \geqslant 0, \cdots, x_{n} \geqslant 0\right\} \subseteq R^{n}$. Suppose that $\mathbf{F}=\left(F_{1}, F_{2}, \cdots, F_{n}\right), \eta=\left(\eta_{1}, \eta_{2}, \cdots, \eta_{n}\right)$, then (FVIP) is to find $x \in M$ such that

$$
\begin{equation*}
\tau(\eta(x, y) \mathbf{F}(x))=\tau\left(\sum_{i=1}^{n} \eta_{i} F_{i}(x)\right)=\sum_{i=1}^{n} \eta_{i} \tau\left(F_{i}(x)\right) \geq 0 \tag{4.14}
\end{equation*}
$$

where $\tau\left(F_{i}(x)\right)=\left(\int_{0}^{1} r\left(F_{i 1}^{-}(x)(r)+F_{i 1}^{+}(x)(r)\right) d r, \int_{0}^{1} r\left(F_{i 2}^{-}(x)(r)+F_{i 2}^{+}(x)(r)\right) d r, \cdots, \int_{0}^{1} r\left(F_{i n}^{-}(x)(r)+F_{i n}^{+}(x)(r)\right) d r\right)$, $i=1,2, \cdots n$.

Theorem 4.6. Let $M$ be a nonempty, compact and convex subset of $R^{m}$ and let $\mathbf{F}$ be a continuous mapping from $X$ into $\left(E^{n}\right)^{n}$. Then there exists a solution to the problem $\operatorname{FVLIP}(M, \mathbf{F}, \eta)$, that is, there exists $x_{0} \in M$ such that

$$
\begin{equation*}
\eta\left(y, x_{0}\right) \mathbf{F}\left(x_{0}\right) \succeq_{c} 0, \quad \forall y \in M \tag{4.15}
\end{equation*}
$$

Proof. If $M$ is a point, the theorem is trivial. If $M$ is not a point, then it can be supposed that $M$ has interior points for otherwise, without loss of generality, $R^{n}$ is replaced by a suitable subspace of $R^{n}$ containing $M$. Since a translation of the space $R^{n}$ dose not affect the assumption or assertion, it can be supposed that $x=0$ is an interior point of $M$. We denote a half-space by $\partial M=\left\{x \in R^{m}:(x-p) n \leq 0\right\}$, where $p$ is a point in $R^{m}$ and $n$ is an nonzero vector in $R^{m}$.

Let $x_{0} \in \partial M$. Then (4.15) holds if and only if there is a hyperplane $\pi$ through $x_{0}$, that is, $\pi=\left\{x \in R^{m}\right.$ : $(x-p) n=0\}$, supporting $M$ such that if $\mathbf{N} \neq 0$ is a fuzzy vector which is fuzzy orthogonal to $\pi$ and pointing into the half-space not containing $M$, then $\mathbf{F}\left(x_{0}\right)=-t \mathbf{N}$ for some $t \geq 0$.

Case1. $\partial M$ is of class $C^{1}$. Assume that (4.15) fails to hold for all $x_{0} \in M$. We shall show that

$$
\begin{equation*}
\mathbf{F}(x)=\mathbf{0} \tag{4.16}
\end{equation*}
$$

has a solution $x_{0} \in M$, which satisfies (4.15) trivially.
Let $\mathbf{N}\left(x_{0}\right)$ be the outward, unit normal vector at $x_{0} \in \partial M$. Then

$$
\mathbf{F}\left(x_{0}, t\right)=(1-t) \mathbf{F}\left(x_{0}\right)+t \mathbf{N}\left(x_{0}\right), 0 \leq t \leq 1,
$$

is a deformation of the vector field $\mathbf{F}\left(x_{0}\right), x_{0} \in \partial M$, into the vector field $\mathbf{N}\left(x_{0}\right)$. The assumption that (4.15) dose not hold for $x_{0} \in \partial M$ implies that $\mathbf{F}\left(x_{0}\right) \neq \mathbf{0}$ for $x_{0} \in \partial M, 0 \leq t \leq 1$. Hence the indices of the vector fields $\mathbf{F}\left(x_{0}\right), \mathbf{N}\left(x_{0}\right)$ with respect to $x=0$ are identical.

There is a deformation $\mathbf{D}\left(x_{0}, s\right)=(1-s) \mathbf{N}\left(x_{0}\right)+s x_{0}, 0 \leq s \leq 1$, of $\mathbf{N}\left(x_{0}\right)$ into $x_{0}$ and $\mathbf{D}\left(x_{0}, s\right) \neq \mathbf{0}$ since $x=0$ is an interior point of $M$. Since the vector field $x_{0}, x_{0} \in \partial M$, has index 1 with respect to $x=0$, the index of $\mathbf{N}\left(x_{0}\right)$ and, hence, of $\mathbf{F}\left(x_{0}\right)$ is 1 . This proves that (4.16) has a solution in $M$.

Case2. $\partial M$ is not of class $C^{1}$. By a theorem of Minkowski (see [45], pp. 36-37), there exists a sequence of compact convex sets $M_{1} \subseteq M_{2} \subseteq \cdots$ such that $M$ is the closure of the union $M_{1} \cup M_{2} \cup \cdots$ and $\partial M_{m}, m=$ $1,2, \cdots$, is of class $C^{1}$. By case 1 , there exists $x_{m} \in M$ satisfying

$$
\eta\left(y, x_{m}\right) \mathbf{F}\left(x_{m}\right) \succeq_{c} 0, \quad \forall y \in M_{m} .
$$

After a selection of a subsequence, it can be supposed that $x_{0}=\lim x_{m}$ exists. Then, by continuity of $\mathbf{F}$, it follows that

$$
\eta\left(y, x_{0}\right) \mathbf{F}\left(x_{0}\right) \succeq_{c} 0, \quad \forall y \in M_{m} .
$$

This implies (4.15) and completes the proof.
Corollary 4.7. Let $M$ be a nonempty, closed and fuzzy invex subset of $R^{m}$ and let $\mathbf{F}: R^{n} \rightarrow\left(E^{n}\right)^{n}$ be continuous. If there exists a nonempty bounded subset $B$ of $M$ such that for every $x \in M \backslash B$ there is $a y \in B$ with

$$
\eta(x, y) \mathbf{F}(x) \geq 0,
$$

then the problem $\operatorname{FVLIP}(M, \mathbf{F}, \eta)$ has a solution.

## 5 Relationship between fuzzy variational-like inequality problems and fuzzy optimization problems

In this section, we investigate the relationships between fuzzy variational-like inequality problems and fuzzy optimization problems.

The Fuzzy Optimization Problem (FOP) is defined as
$\min \quad f(t)$
subject to $t \in M$,
where $M$ is closed and convex set and in $R^{n}$ and $f: M \rightarrow E^{n}$ is continuously $g$-differentiable.

Theorem 5.1. Suppose that $f: M \rightarrow E^{n}$ is fuzzy invex with respect to some continuous map $\eta: M \times M \rightarrow R^{n}$. If $t^{\star} \in M$ is a solution to $\operatorname{FVLIP}(M, F, \eta)$, where $\mathbf{F}(t)=\nabla f$, then $t^{\star}$ is a solution to the (FOP).

Proof. By the fuzzy invexity of $f$, we have

$$
f(t) \Theta_{g} f\left(t^{\star}\right) \succeq_{c} \eta\left(t, t^{\star}\right) \nabla f\left(t^{\star}\right), \forall t \in M
$$

Since $t^{\star} \in M$ is a solution to $\operatorname{FVLIP}(M, F, \eta)$, we have

$$
\eta\left(t, t^{\star}\right) \mathbf{F}\left(t^{\star}\right) \succeq_{c} 0, \forall t \in M
$$

Now, setting $\mathbf{F}\left(t^{\star}\right)=\nabla f\left(t^{\star}\right)$, we obtain

$$
f(t) \Theta_{g} f\left(t^{\star}\right) \succeq_{c} 0, \forall t \in M
$$

that is,

$$
f(t) \succeq_{c} f\left(t^{\star}\right), \forall t \in M
$$

Thus, we have

$$
f\left(t^{\star}\right)=\min _{t \in M} f(t)
$$

Therefore, $t^{\star}$ is a solution to the (FOP).
Theorem 5.2. Let $K \subseteq R^{n}$ be an invex set with respect to $\eta, x^{\star} \in K$, and $F: K \rightarrow E^{n}$ be a g-differentiable incave fuzzy mapping (FIC) with respect to $\eta$. If $x^{\star}$ is a strictly local optimal solution to (FOP), then ( $x^{\star}, \nabla F\left(x^{*}\right)$ ) is a solution to (FVLIP).

Proof. Let $x^{*}$ be a strictly local optimal solution to (FOP). By contradiction, suppose that there exists an $\bar{x} \in K$ such that

$$
\eta\left(\bar{x}, x^{\star}\right) \nabla F\left(x^{\star}\right) \preceq_{c} 0 .
$$

Since $F$ is a $g$-differentiable incave fuzzy mapping,

$$
F(\bar{x}) \ominus_{g} F\left(x^{\star}\right) \preceq_{c} \nabla \eta\left(\bar{x}, x^{\star}\right) F\left(x^{\star}\right) .
$$

Thus, we have

$$
F(\bar{x}) \preceq_{c} F\left(x^{\star}\right) .
$$

This contradicts the fact that $x^{\star}$ is a strictly local optimal solution of (FOP).
Theorem 5.3. Let $K \subseteq R^{n}$ be an open invex set with respect to $\eta, x^{\star} \in K$, and $F: K \rightarrow E^{n}$ be a g-differentiable strictly incave fuzzy mapping (FSIC) with respect to $\eta$. If $x^{\star}$ is an optimal solution of (FOP), then ( $x^{\star}, \nabla F\left(x^{\star}\right)$ is a solution to (FVLIP).

Proof. Let $x^{\star}$ be an optimal solution to (FOP). By contradiction, suppose that there exists an $\bar{x} \in K$ such that

$$
\eta\left(\bar{x}, x^{\star}\right) \nabla F\left(x^{\star}\right) \preceq_{c} 0 .
$$

Since $F$ is a $g$-differentiable strictly incave fuzzy mapping,

$$
F(\bar{x}) \ominus_{g} F\left(x^{\star}\right) \prec_{c} \nabla \eta\left(\bar{x}, x^{\star}\right) F\left(x^{\star}\right) .
$$

Therefore,

$$
F(\bar{x}) \prec c F\left(x^{\star}\right)
$$

This contradicts the fact that $x^{\star}$ is an optimal solution to (FOP).

## 6 Fuzzy multiobjective optimization

In this section, we investigate the optimality conditions for the multiobjective optimization problems.
The Fuzzy Multiobjective Optimization Problem (FMOP1) is defined as

$$
\begin{array}{ll}
\min & \mathbf{F}(x)=\left(f_{1}(x), f_{2}(x), \cdots, f_{p}(x)\right) \\
\text { subject to } & \mathbf{G}(x) \leq \mathbf{0},  \tag{6.1}\\
& \mathbf{H}(x)=\mathbf{0}, \\
& x \in M,
\end{array}
$$

where $M \subseteq R^{m}$ is closed and convex set, the objective function $\mathbf{F}(x): M \rightarrow\left(L\left(E^{n}\right)\right)^{p}$ is a fuzzy-vectorvalued function, $\mathbf{G}(x): M \rightarrow\left(L\left(E^{n}\right)\right)^{l}$ and $\mathbf{H}(x): M \rightarrow\left(L\left(E^{n}\right)\right)^{t}$ in constraint conditions are fuzzy-vector-valued functions, denoted by $\mathbf{G}(x)=\left(g_{1}(x), g_{2}(x), \cdots, g_{l}(x)\right), \mathbf{H}(x)=\left(h_{1}(x), h_{2}(x), \cdots, h_{t}(x)\right)$, where $f_{i}(x), g_{k}(x), h_{s}(x): M \rightarrow E^{n}, i=1,2, \cdots, p, k=1,2, \cdots, l, s=1,2, \cdots, t$.
$X=\{x \in M: \mathbf{G}(x) \leq \mathbf{0}, \mathbf{H}(x)=\mathbf{0}\}$ is said to be the feasible set of (FMOP1). Let $x_{0} \in X$, if there does not exist $x \in M$ such that $\mathbf{F}(x) \leq \mathbf{F}\left(x_{0}\right)$, then $x_{0}$ is said to be an optimal solution to (FMOP1).

Let $C=R^{n+}=\left\{\left(x_{1}, x_{2}, \cdots, x_{n}\right) \in R^{n}: x_{1} \geqslant 0, x_{2} \geqslant 0, \cdots, x_{n} \geqslant 0\right\} \subseteq R^{n}$. Then we have

$$
\begin{equation*}
\mathbf{G}(x) \leq \mathbf{0} \Longleftrightarrow \tau\left(g_{k}(x)\right) \leq 0\left(0 \in R^{n}\right), k=1,2, \cdots, l . \tag{6.2}
\end{equation*}
$$

where $\tau\left(g_{k}(x)\right)=\left(\int_{0}^{1} r\left(g_{k 1}^{-}(x)(r)+g_{k 1}^{+}(x)(r)\right) d r, \int_{0}^{1} r\left(g_{k 2}^{-}(x)(r)+g_{k 2}^{+}(x)(r)\right) d r, \cdots, \int_{0}^{1} r\left(g_{k n}^{-}(x)(r)+g_{k n}^{+}(x)(r)\right) d r\right)$, $k=1,2, \cdots l$, thus, we obtain

$$
\begin{equation*}
\mathbf{G}(x) \leq \mathbf{0} \Longleftrightarrow \int_{0}^{1} r\left(g_{k j}^{-}(x)(r)+g_{k j}^{+}(x)(r)\right) d r \leq 0(0 \in R), k=1,2, \cdots, l, j=1,2, \cdots, n . \tag{6.3}
\end{equation*}
$$

Similarly, we have

$$
\begin{equation*}
\mathbf{H}(x)=\mathbf{0} \Longleftrightarrow \int_{0}^{1} r\left(h_{s j}^{-}(x)(r)+h_{s j}^{+}(x)(r)\right) d r=0(0 \in R), s=1,2, \cdots, t, j=1,2, \cdots, n \tag{6.4}
\end{equation*}
$$

We denote $G_{k^{\prime}}(x)=\int_{0}^{1} r\left(g_{k j}^{-}(x)(r)+g_{k j}^{+}(x)(r)\right) d r, H_{s^{\prime}}(x)=\int_{0}^{1} r\left(h_{s j}^{-}(x)(r)+h_{s j}^{+}(x)(r)\right) d r, k^{\prime}=1,2, \cdots, l \times$ $n, s^{\prime}=1,2, \cdots, t \times n$, then the fuzzy multiobjective optimization problem (FMOP1) can be transformed into the following fuzzy multiobjective optimization problem (FMOP2)

$$
\begin{array}{ll}
\min & \mathbf{F}(x)=\left(f_{1}(x), f_{2}(x), \cdots, f_{p}(x)\right) \\
\text { subject to } & G_{k^{\prime}}(x) \leq 0, \\
& H_{s^{\prime}}(x)=0,  \tag{6.5}\\
& x \in M,
\end{array}
$$

where $G_{k^{\prime}}, H_{s^{\prime}}: M \rightarrow R$.
Obviously, the feasible set of (FMOP2) is equivalent to the feasible set of (FMOP1).
In the following, suppose that the feasible set of (FMOP2) $X=\left\{x \in \operatorname{int} M: G_{k^{\prime}}(x) \leq 0, H_{s^{\prime}}(x)=0, k^{\prime}=\right.$ $\left.1,2, \cdots, l \times n, s^{\prime}=1,2, \cdots, t \times n\right\} \subseteq \mathcal{K}_{C}^{n}$, the real-valued functions $G_{k^{\prime}}(x), k^{\prime}=1,2, \cdots, l \times n$, are convex on $M$, continuous and differentiable at $x_{0} \in X$.

Definition 6.1. Let $\mathbf{F}: M \rightarrow\left(L\left(E^{n}\right)\right)^{p}$, denoted by $\mathbf{F}(x)=\left(f_{1}(x), f_{2}(x), \cdots, f_{p}(x)\right)$. If for any $x_{1}, x_{2} \in \operatorname{int} M$ and $\lambda \in[0,1]$, the inequalities

$$
\begin{equation*}
f_{i j}^{-}\left(\lambda x_{1}+(1-\lambda) x_{2}\right) \leq \lambda f_{i j}^{-}\left(x_{1}, r\right)+(1-\lambda) f_{i j}^{-}\left(x_{2}, r\right), i=1,2, \cdots, p, j=1,2, \cdots, n, \tag{6.6}
\end{equation*}
$$

and

$$
\begin{equation*}
f_{i j}^{+}\left(\lambda x_{1}+(1-\lambda) x_{2}\right) \leq \lambda f_{i j}^{+}\left(x_{1}, r\right)+(1-\lambda) f_{i j}^{+}\left(x_{2}, r\right), i=1,2, \cdots, p, j=1,2, \cdots, n, \tag{6.7}
\end{equation*}
$$

uniformly hold for all $r \in[0,1]$, then $\mathbf{F}(x)$ is said to be endpoint-wise fuzzy convex.

Definition 6.2. Let $\mathbf{F}: M \rightarrow\left(L\left(E^{n}\right)\right)^{p}$, denoted by $\mathbf{F}(x)=\left(f_{1}(x), f_{2}(x), \cdots, f_{p}(x)\right)$. Then we say $F$ is endpointwise differentiable at $x_{0}$, that is, if there exists $u_{k i j}^{-}, u_{k i j}^{+} \in R, k=1,2, \cdots, p, i=1,2, \cdots, n, j=1,2, \cdots, m$, such that

$$
\lim _{h \rightarrow 0} \frac{f_{k i}^{-}\left(x_{1}^{0}, \cdots, x_{j}^{0}+h, \cdots, x_{m}^{0}, r\right)-f_{k i}^{-}\left(x_{1}^{0}, \cdots, x_{j}^{0}, \cdots, x_{m}^{0}, r\right)}{h}=u_{k i j}^{-}, k=1,2, \cdots, p
$$

and

$$
\lim _{h \rightarrow 0} \frac{f_{k i}^{+}\left(x_{1}^{0}, \cdots, x_{j}^{0}+h, \cdots, x_{m}^{0}, r\right)-f_{k i}^{+}\left(x_{1}^{0}, \cdots, x_{j}^{0}, \cdots, x_{m}^{0}, r\right)}{h}=u_{k i j}^{+}, k=1,2, \cdots, p,
$$

uniformly for $r \underset{\partial F^{-}}{\in}[0,1]$, then we say $\mathbf{F}$ has jth partial endpoint-wise differentiable at $x_{0}$, and we denote $\frac{\partial F_{i}^{-}\left(x_{0}, r\right)}{\partial x_{j}^{0}}=u_{i j}^{-}, \frac{\partial F_{i}^{+}\left(x_{0}, r\right)}{\partial x_{j}^{0}}=u_{i j}^{+}$. If all the partial endpoint-wise derivatives at $x_{0}$ exist, then we say $\mathbf{F}$ is endpointwise differentiable at $x_{0}$.

Theorem 6.3. Let the objective function $\mathbf{F}: M \rightarrow\left(L\left(E^{n}\right)\right)^{p}$, denoted by $\mathbf{F}(x)=\left(f_{1}(x), f_{2}(x), \cdots, f_{p}(x)\right)$, be endpoint-wise fuzzy convex, let $\mathbf{F}$ be continuous and endpoint-wise differentiable at $x_{0}=\left\{x_{1}^{0}, x_{2}^{0}, \cdots, x_{m}^{0}\right\} \in$ $\operatorname{intM}$. If for any $r \in[0,1]$, there exist $\omega(r)=\left\{\omega_{1}(r), \omega_{2}(r), \cdots, \omega_{p}(r)\right\} \in R^{p+}, \alpha(r)=$ $\left\{\alpha_{1}(r), \alpha_{2}(r), \cdots, \alpha_{l \times n}(r)\right\} \in R^{(l \times n)+}$ and $\beta(r)=\left\{\beta_{1}(r), \beta_{2}(r), \cdots, \beta_{t \times n}(r)\right\} \in R^{t \times n}$ such that
(1) $\left.\sum_{i=1}^{p} \omega_{i}(r) \frac{\partial f_{i j}^{-}(x, r)}{\partial x_{j^{\prime}}}\right|_{x_{0}}+\left.\sum_{i=1}^{p} \omega_{i}(r) \frac{\partial f_{i j}^{+}(x, r)}{\partial x_{j^{\prime}}}\right|_{x_{0}}+\left.\sum_{k^{\prime}=1}^{l \times n} \alpha_{k^{\prime}}(r) \frac{\partial G_{k^{\prime}}(x)}{\partial x_{j^{\prime}}}\right|_{x_{0}}+\left.\sum_{s^{\prime}=1}^{t \times n} \beta_{s^{\prime}}(r) \frac{\partial H_{s^{\prime}}(x)}{\partial x_{j^{\prime}}}\right|_{x_{0}}=0, j^{\prime}=1,2, \cdots, m$,
(2) $\alpha_{k^{\prime}}(r) G_{k^{\prime}}\left(x_{0}\right)=0, k^{\prime}=1,2, \cdots l \times n$,
then $x_{0}$ is an optimal solution to (FMOP2).
Note that $\omega(r), \alpha(r), \beta(r)$ are called Lagrange multiplier vectors containing parameters, the condition (1) and (2) are called the Karush-Kuhn-Tucker (KKT for short) conditions for (FMOP2).

Proof. $\forall r \in[0,1]$, we denote $\bar{f}(x, r)=\left\{\bar{f}_{1}(x, r), \bar{f}_{2}(x, r), \cdots, \bar{f}_{p}(x, r)\right\}$, and

$$
\bar{f}_{i j}(x, r)=f_{i}^{-}(x, r)+f_{i j}^{+}(x, r), i=1,2, \cdots, p, j=1,2, \cdots, n .
$$

Since $\mathbf{F}$ is endpoint-wise fuzzy convex on $M$, and continuous and endpoint-wise differentiable at $x_{0}$, then the real-valued function $f_{i j}^{-}(x, r)$ and $f_{i j}^{+}(x, r), i=1,2, \cdots, p, j=1,2, \cdots, n$, is convex on $M$, and continuous and differentiable at $x_{0}$. Therefore, for all $r \in[0,1], \bar{f}_{i j}(x, r)$ is convex, and continuous and differentiable at $x_{0}$, furthermore, we have

$$
\left.\frac{\partial \bar{f}_{i j}(x, r)}{\partial x_{j^{\prime}}}\right|_{x_{0}}=\left.\frac{\partial f_{i j}^{-}(x, r)}{\partial x_{j^{\prime}}}\right|_{x_{0}}+\left.\frac{\partial f_{i j}^{+}(x, r)}{\partial x_{j^{\prime}}}\right|_{x_{0}}, i=1,2, \cdots, p, j=1,2, \cdots, n, j^{\prime}=1,2, \cdots, m
$$

Since $\forall r \in[0,1]$, the KKT conditions are equivalent to
(1) $\left.\sum_{i=1}^{p} \omega_{i}(r) \frac{\partial \bar{f}_{i j}(x, r)}{\partial x_{j^{\prime}}}\right|_{x_{0}}+\left.\sum_{k^{\prime}=1}^{l \times n} \alpha_{k^{\prime}}(r) \frac{\partial G_{k^{\prime}}(x)}{\partial x_{j^{\prime}}}\right|_{x_{0}}+\left.\sum_{s^{\prime}=1}^{t \times n} \beta_{s^{\prime}}(r) \frac{\partial H_{s^{\prime}}(x)}{\partial x_{j^{\prime}}}\right|_{x_{0}}=0$,
(2) $\alpha_{k^{\prime}}(r) G_{k^{\prime}}\left(x_{0}\right)=0, k^{\prime}=1,2, \cdots l \times n$,
thus, $x_{0}$ is an optimal solution to the multiobjective optimization problem under this constraint conditions (1) and (2), where the mutiobjective function $\bar{f}(x, r)=\left\{\bar{f}_{1}(x, r), \bar{f}_{2}(x, r), \cdots, \bar{f}_{p}(x, r)\right\}$, that is $\forall x \in \operatorname{int} M$, we have $\bar{f}\left(x_{0}, r\right) \leq \bar{f}(x, r)$, that is,

$$
\begin{equation*}
\bar{f}_{i j}\left(x_{0}, r\right) \leq \bar{f}_{i j}(x, r), i=1,2, \cdots, p, j=1,2, \cdots, n, \tag{6.8}
\end{equation*}
$$

By reductio ad absurdum, suppose that $x_{0}$ is not an optimal solution of (FMOP2), then there exists $x^{\prime} \in \operatorname{int} M$ such that $\mathbf{F}\left(x^{\prime}\right)<\mathbf{F}\left(x_{0}\right)$.

Let $C=R^{n+} \subseteq R^{n}$, according to Definition 2.7, we have

$$
\int_{0}^{1} r\left(f_{i j}^{-}\left(x^{\prime}\right)(r)+f_{i 1}^{+}\left(x^{\prime}\right)(r)\right) d r<\int_{0}^{1} r\left(f_{i j}^{-}\left(x_{0}\right)(r)+f_{i 1}^{+}\left(x_{0}\right)(r)\right) d r, i=1,2, \cdots, p, j=1,2, \cdots, n
$$

that is,

$$
\int_{0}^{1} r \bar{f}_{i j}\left(x^{\prime}, r\right) d r<\int_{0}^{1} r \bar{f}_{i j}\left(x_{0}, r\right) d r, i=1,2, \cdots, p, j=1,2, \cdots, n
$$

which is in contradiction to (6.8). Therefore, $x_{0}$ is an optimal solution to (FMOP2).
Theorem 6.4 Let the objective function $\mathbf{F} \quad: \quad M \quad \rightarrow \quad\left(L\left(E^{n}\right)\right)^{p}$ be denoted by $\mathbf{F}(x)=\left(f_{1}(x) u_{1}, f_{2}(x) u_{2}, \cdots, f_{p}(x) u_{p}\right)$, where $f_{i}:[a, b] \rightarrow R, u_{i} \in L\left(E^{n}\right), i=1,2, \cdots p$, and $u_{i} \succeq_{c} 0$. Let $\mathbf{F}$ be endpoint-wise fuzzy convex, continuous and endpoint-wise differentiable at $x_{0}=\left\{x_{1}^{0}, x_{2}^{0}, \cdots, x_{m}^{0}\right\} \in \operatorname{int} M$. If there exist $\omega=\left\{\omega_{1}, \omega_{2}, \cdots, \omega_{p}\right\} \in R^{p+}, \alpha(r)=\left\{\alpha_{1}(r), \alpha_{2}(r), \cdots, \alpha_{l \times n}(r)\right\} \in R^{(l \times n)+}$ and $\beta(r)=$ $\left\{\beta_{1}(r), \beta_{2}(r), \cdots, \beta_{t \times n}(r)\right\} \in R^{t \times n}$ such that
(1) $\sum_{\omega_{i}=1}^{p} \omega_{i} \nabla f_{i}\left(x_{0}\right)+\left.\sum_{k^{\prime}=1}^{l \times n} \alpha_{k^{\prime}}(r) \frac{\partial G_{k^{\prime}}(x)}{\partial x_{j^{\prime}}}\right|_{x_{0}}+\left.\sum_{s^{\prime}=1}^{t \times n} \beta_{s^{\prime}}(r) \frac{\partial H_{s^{\prime}}(x)}{\partial x_{j^{\prime}}}\right|_{x_{0}}=0, j^{\prime}=1,2, \cdots, m$,
(2) $\alpha_{k^{\prime}}(r) G_{k^{\prime}}\left(x_{0}\right)=0, k^{\prime}=1,2, \cdots l \times n$,
then $x_{0}$ is an optimal solution to (FMOP2).
Note that $\omega, \alpha, \beta$ are called Lagrange multiplier vectors.
Proof. By Definition 2.18, $\forall x_{0} \in M$, we have $\nabla \mathbf{F}\left(x_{0}\right)=\left(\nabla\left(f_{1}\left(x_{0}\right) u_{1}\right), \nabla\left(f_{2}\left(x_{0}\right) u_{2}\right), \cdots, \nabla\left(f_{p}\left(x_{0}\right) u_{p}\right)\right)$, and

$$
\begin{equation*}
\nabla\left(f_{i}\left(x_{0}\right) u_{i}\right)=\left(u_{i} \frac{\partial f_{i}}{\partial x_{1}^{0}}, u_{i} \frac{\partial f_{i}}{\partial x_{2}^{0}}, \cdots, u_{i} \frac{\partial f_{i}}{\partial x_{m}^{0}}\right), i=1,2, \cdots, p \tag{6.9}
\end{equation*}
$$

Since $\mathbf{F}$ is endpoint-wise fuzzy convex $M$, and continuous and endpoint-wise differentiable at $x_{0}$, then $f_{i}(x), i=1,2, \cdots, p$, is convex on $M$, continuous and differential at $x_{0} \in \operatorname{int} M$, that is, the real-vector-valued function $f=\left(f_{1}(x), f_{2}(x), \cdots, f_{p}(x)\right)$ is convex on $M$, continuous and differential at $x_{0} \in \operatorname{int} M$. Consider the following multiobjective optimization problem

$$
\begin{array}{ll}
\min & f(x)=\left(f_{1}(x), f_{2}(x), \cdots, f_{p}(x)\right) \\
\text { subject to } G_{k^{\prime}}(x) \leq 0,  \tag{6.10}\\
& H_{s^{\prime}}(x)=0, \\
& x \in M .
\end{array}
$$

Obviously, the conditions (1) and (2) are the KKT conditions for this problem. Therefore, $x_{0}$ is an optimal solution to this problem, that is, $\forall x \in \operatorname{int} M$, we have $f\left(x_{0}\right) \leq f(x)$, that is,

$$
\begin{equation*}
f_{i}\left(x_{0}\right) \leq f_{i}(x), i=1,2, \cdots, p \tag{6.11}
\end{equation*}
$$

By reductio ad absurdum, suppose that $x_{0}$ is not an optimal solution to (FMOP2), then there exists $x^{\prime} \in \operatorname{int} M$ such that $\mathbf{F}\left(x^{\prime}\right)<\mathbf{F}\left(x_{0}\right)$, that is, $f_{i}\left(x^{\prime}\right) u \prec c f_{i}\left(x_{0}\right) u, i=1,, 2, \cdots, p$.

Let $C=R^{n+} \subseteq R^{n}$, according to Definition 2.7, we have

$$
\tau\left(f_{i}\left(x^{\prime}\right) u\right) \in \tau\left(f_{i}\left(x_{0}\right) u\right)+C
$$

thus, we obtain $f_{i}\left(x^{\prime}\right) \tau(u) \in f_{i}\left(x_{0}\right) \tau(u)+C$ and $f_{i}\left(x^{\prime}\right) \tau(u) \neq f_{i}\left(x_{0}\right) \tau(u)$. Since $u_{i} \succeq_{c} 0$, we have $f_{i}\left(x^{\prime}\right)<f_{i}\left(x_{0}\right)$, which is in contradiction to (6.11). Therefore, $x_{0}$ is an optimal solution to (FMOP2).

## 7 Conclusions

We define the fuzzy variational-like inequality problems by using the new ordering of two $n$-dimensional fuzzy-number-valued functions, and the existence and the basic properties of the fuzzy variational inequality problems are also investigated. We examine the relationship between the variational-like inequality problems and fuzzy optimization problems. Additionally, we discuss the optimality conditions for fuzzy multiobjective optimization. The next step for the continuation of the research direction proposed here is to investigate illposedness and regularization methods of the fuzzy variational-like inequality problems, and the duality for
the fuzzy multiobjective optimization problems.

Acknowledgments: We would like to thank the anonymous reviewers for their careful work and helpful comments. This research is supported by the National Natural Science Foundation of China (61763044) and the Strategic Priority Research Program of Chinese Academy of Sciences (XDA21010202).

## References

[1] Hartman P., Stampacchia G., On some nonlinear elliptic differential functional equations, Acta Math., 1966, 115, 153-188
[2] Minty G.J., On the generalization of a direct method of the calculus of variations, Bull. Amer. Math. Soc., 1967, 73, 314-321
[3] Giannessi F., Theorem of alternative, quadratic programs and complementarity problems, in: R.W. Cottle, F. Giannessi, J.L. Lions (Eds.), Variational Inequalities and Complementarity Problems, JohnWiley and Sons, New York, 1980, pp. 151-186
[4] Bensoussan A., Lions J.L., Nouvelle formulation de problems de control impulsionel et applications, C. R. Ac. Sci., 1973, 1189-1192
[5] Cubiotti P., Generalized quasi-variational inequalities without continuities, Journal of Optimization Theory and Applications, 1997, 92, 477-495
[6] Lunsford M.L., Existence of results for generalized variational inequalities, Ph.D, The University of Alabama in Huntsville, 1995, AAI9625420
[7] Noor M.A., Generalized multivalued quasi-variational inequalites (II), Computers and Mathematics with Applications, 1998, 35, 63-78
[8] Zadeh L.A., Fuzzy sets, Inform Control., 1965, 8, 338-353
[9] Zadeh L.A., The concept of a linguistic variable and its application to approximate reasoning-I, Inform Sci., 1975, 8, 199-249
[10] Chang S.S., Zhu Y.G., On variational inequalities for fuzzy mappings, Fuzzy Sets Syst., 1989, 32, 359-367
[11] Ahmad M.K., Salahuddin, A fuzzy extension of generalized implicit vector variational-like inequalities, Positivity, 2007, 11, 477-484
[12] Chang S.S., Huang N.J., Generalized complementarity problems for fuzzy mappings, Fuzzy Sets Syst., 1993, 55, 227-234
[13] Chang S.S., Salahuddin, Existence of vector quasi-variational-like inequalities for fuzzy mappings, Fuzzy Sets Syst., 2013, 233, 89-95
[14] Chang S.S., Salahuddin, Ahmad M.K., Wang X.R., Generalized vector variational-like inequalities in fuzzy environment, Fuzzy Sets Syst., 2015, 265, 110-120
[15] Dai H.X., Generalized mixed variational-like inequality for random fuzzy mappings, J. Comput. Appl. Math., 2009, 224, 20-28
[16] Huang N.J., A new method for a class of nonlinear variational inequalities with fuzzy mappings, Appl. Math. Lett., 1997, 10, 129-133
[17] Noor M.A., Variational inequalities for fuzzy mappings (I), Fuzzy Sets Syst., 1993, 55, 309-312
[18] Tang G.J., Zhao T., Wan Z.P., He D.X., Existence results of a perturbed variational inequality with a fuzzy mapping, Fuzzy Sets and Syst., 2018, 331, 68-77
[19] Ward D.E., Lee G.M., On relations between vector variational inequality and vector optimization problem, Journal of Optimization Theory and Application, 2002, 113, 583-596
[20] Yang X.M., Yang X.Q., Vector variational-like inequality with pseudoinvexity, Optimization, 2006, 55 (2006) 157-170
[21] Wu Z.Z., Xu J.P., Generalized convex fuzzy mappings and fuzzy variational-like inequality, Fuzzy Sets Syst., 2009, 160, 15901619
[22] Wu Z.Z., Xu J.P., A class of fuzzy variational inequality based on monotonicity of fuzzy mappings, Abstract and Applied Analysis, 2013, 2013, Article ID 854751, 17 pages
[23] Weir T., Mond B., Preinvex functions in multiobjective optimization, J. Math. Anal.Appl., 1988, 136, 29-38
[24] Noor M.A., Variational-like inequalities, Optimization, 1994, 30, 323-330
[25] Noor M.A., On generalized preinvex functions and monotonicities, J. Inequal. Pure Appl. Math., 2004, 5, 1-9
[26] Ruiz-Garzon G., Osuna-Gomez R., Rufian-Lizan A., Relationships between vector variational-like inequality and optimization problems, European J. Oper. Res., 2004, 157, 113-119
[27] Goetschel R., Voxman W., Elementary fuzzy calculus, Fuzzy Sets Syst., 1986, 18, 31-43
[28] Nanda S., Kar K., Convex fuzzy mappings, Fuzzy Sets Syst., 1992, 48, 129-132
[29] Gong Z.T., S.X. Hai, The convexity of $n$-dimensional fuzzy-number-valued functions and its applications, Fuzzy Sets Syst., 2016, 296, 19-36
[30] Diamond P., Kloeden P., Characterization of compact subsets of fuzzy sets, Fuzzy Sets Syst., 1989, 29, 341-348
[31] Ma M., On embedding problems of fuzzy number spaces: Part 5, Fuzzy Sets Syst., 1993, 55, 313-318
[32] Kaleva O., Fuzzy differential equations, Fuzzy Sets Syst., 24 (1987) 301-317
[33] Wu C.X., Ma M., Fang J.X., Structure Theory of fuzzy Analysis, 1994, Guizhou Scientific Publication (in Chinese).
[34] Li L.F., Liu S.Y., Zhang J.K., On fuzzy generalized convex mappings and optimality conditions for fuzzy weakly univex mappings, Fuzzy Sets Syst., 2015, 280, 107-132
[35] Baskov O.V., Some properties of fuzzy dual cones, Fuzzy Sets Syst., 2018, 331, 78-84
[36] Hukuhara M., Integration des applications mesurables dont la valeur est un compact convex, Funkcial. Ekvac., 1967, 10, 205-229
[37] Stefanini L., A generalization of Hukuhara difference, in: D. Dubois, M.A. Lubiano, H. Prade, M.A. Gil, P. Grzegorzewski, 0. Hryniewicz (Eds.), Soft Methods for Handling Variability and Imprecision, in: Series on Advances in Soft Computing, 2008, Springer
[38] Stefanini L., A generalization of Hukuhara difference and division for interval and fuzzy arithmetic, Fuzzy Sets Syst., 2010, 161, 1564-1584
[39] Bede B., Stefanini L., Generalized differentiability of fuzzy-valued functions, Fuzzy Sets Syst., 2013, 230, 119-141
[40] Gomes L.T., Barros L.C., A note on the generalized difference and the generalized differentiability, Fuzzy Sets Syst., 2015, 280, 142-145
[41] Hai S.X., Gong Z.T., Li H.X., Generalized differentiability for $n$-dimensional fuzzy-number-valued functions and fuzzy optimization, Inform Sci., 2016, 374, 151-163
[42] Zhang B.K., On measurability of fuzzy-number-valued functions, Fuzzy Sets Syst., 2001, 120, 505-509
[43] Wang G.X., Wu C.X., Fuzzy n-cell numbers and the differential of fuzzy n-cell number value mappings, Fuzzy Sets and Syst., 2002, 130, 367-381
[44] Rufián-Lizana A., Chalco-Cano Y., Osuna-Gómez R., Ruiz-Garzón G., On invex fuzzy mappings and fuzzy variational-like inequalities, Fuzzy Sets Syst., 2012, 200, 84-98
[45] Bonnesen T., Fenchel W., Theorie der konvexen Körper, Ergeb. Math., 1934, Berlin


[^0]:    *Corresponding Author: Zengtai Gong: College of Mathematics and Statistics, Northwest Normal University, Lanzhou 730070, China, E-mail: zt-gong@163.com, gongzt@nwnu.edu.cn
    Ting Xie: College of Mathematics and Statistics, Northwest Normal University, Lanzhou 730070, China and Institute of Modern Physics, Chinese Academy of Sciences, Lanzhou 730000, China, E-mail: xietingr@163.com

