# FUZZY CONVEX FUNCTIONS: A RUDIMENTARY THEORY AND ITS USE IN FUZZY - VALUED INEQUALITIES

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This paper intends to establish a rudimentary convexity theory for the fuzzy-valued functions of a fuzzy variable. Some operations on fuzzy convex functions are investigated and the convexity of fuzzy extension functions of ordinary convex functions is checked. Moreover, two fuzzy-valued inequalities have been proved by using this fuzzy convex function theory.

Keywords: Compact normal convex fuzzy set, Fuzzy convex region, Fuzzy number, Fuzzy convex function, Fuzzy-valued inequality.

### 1. Preliminaries

Throughout this paper, we shall denote by  $\mathbf{R}$  the set of all real numbers, by  $\mathbf{R}^+$  the set of all positive real numbers and by  $\mathbf{R}^-$  the set of all negative real numbers. If  $\tilde{\mathbf{a}}$  is a fuzzy set on  $\mathbf{R}^n$ ,  $\tilde{\mathbf{a}}(\mathbf{x})$  will denote the grade of membership of  $\mathbf{x}$  in  $\tilde{\mathbf{a}}$ ,  $S(\tilde{\mathbf{a}})$  the support of  $\tilde{\mathbf{a}}$  and for each  $\mathbf{r} \in ]0,1]$   $[\tilde{\mathbf{a}}]_r$  the r-level set of  $\tilde{\mathbf{a}}$ :  $[\tilde{\mathbf{a}}]_r = \{\mathbf{x} \in \mathbf{R}^n \colon \tilde{\mathbf{a}}(\mathbf{x}) \geqslant r\}$ . If  $\tilde{\mathbf{a}}$ ,  $\tilde{\mathbf{b}}$  are two fuzzy sets on  $\mathbf{R}^n$  and  $\lambda$  a real number, as special cases of Zadeh's extension principle [14] we define  $\tilde{\mathbf{a}} + \tilde{\mathbf{b}}$  and  $\lambda \tilde{\mathbf{a}}$  as

$$(\tilde{a} + \tilde{b})(x) = \sup_{y+z=x} \tilde{a}(y) \wedge \tilde{b}(z) \text{ and } (\lambda \tilde{a})(x) = \begin{cases} \tilde{a}(\frac{1}{\lambda}x), & \text{if } \lambda \neq 0, \\ \tilde{\theta}_{a}(x), & \text{if } \lambda = 0 \end{cases}$$

respectively, where  $\theta$  is the origin of  $R^n$ ,  $a = \sup \{ \widetilde{a}(x) : x \in R^n \}$  and  $\widetilde{\theta}_a$  is the fuzzy set defined by

$$\widetilde{\theta}_{a}(x) = \begin{cases} a, & \text{if } x = \theta, \\ 0, & \text{if } x \neq \theta. \end{cases}$$

It is well known ( see Nguyen's theory [9] ) that for each re]0,1] we have

$$[\widetilde{a} + \widetilde{b}]_r = [\widetilde{a}]_r + [\widetilde{b}]_r \quad \text{and} \quad [\lambda \widetilde{a}]_r = \lambda [\widetilde{a}]_r \,, \qquad (1.1)$$
 where 
$$[\widetilde{a}]_r + [\widetilde{b}]_r = \{x + y : x \in [\widetilde{a}]_r, y \in [\widetilde{b}]_r \} \quad \text{and} \quad \lambda [\widetilde{a}]_r = \{\lambda x : x \in [\widetilde{a}]_r \} \,.$$

Let  $\tilde{a}$  be a fuzzy set on  $\mathbb{R}^n$ . Following Zadeh [13] and Weiss [12] we shall call  $\tilde{a}$  a compact normal convex fuzzy set if for each  $r \in [0,1]$  [ $\tilde{a}$ ]<sub>r</sub> is a non-empty compact convex subset of  $\mathbb{R}^n$ .

We shall denote by  $\widetilde{\mathbf{R}}^n$  the family of all compact normal convex fuzzy sets on  $\mathbf{R}^n$  and, for each non-empty ordinary convex subset  $\mathbf{C} \subset \mathbf{R}^n$ , by  $\widetilde{\mathbf{C}}$  the family of all compact normal convex fuzzy sets with supports contained in  $\mathbf{C} \colon \widetilde{\mathbf{C}} = \{\widetilde{\mathbf{a}} \in \widetilde{\mathbf{R}}^n \colon S(\widetilde{\mathbf{a}}) \subset \mathbf{C}\}$ .

**Definition 1.1.** Let C be a non-empty ordinary convex subset of  $\mathbb{R}^n$ . A non-empty subset  $\underline{D}$  of  $\widetilde{C}$  is called a fuzzy convex region on C if and only if for all  $\widetilde{a},\widetilde{b}\in\underline{D}$  and  $0<\lambda<1$  we have  $\lambda\widetilde{a}+(1-\lambda)\widetilde{b}\in\underline{D}$ .

Note that every singleton of  $\mathbb{C}$  can also be regarded as an element of  $\widetilde{\mathbb{C}}$ , so that each non-empty ordinary convex subset of  $\mathbb{C}$  is a fuzzy convex region on  $\mathbb{C}$ . Further note that, for a fuzzy set  $\widetilde{a} \in \widetilde{\mathbb{R}}^{n}$ ,  $\widetilde{a} \in \widetilde{\mathbb{C}}$  if and only if  $[\widetilde{a}]_{r} \subset \mathbb{C}$  for all  $r \in ]0,1]$ , thus it follows from (1.1) that for any  $\widetilde{a},\widetilde{b} \in \widetilde{\mathbb{C}}$  and  $0<\lambda<1$  we have  $\lambda\widetilde{a}+(1-\lambda)\widetilde{b} \in \widetilde{\mathbb{C}}$ , hence  $\widetilde{\mathbb{C}}$  itself is a fuzzy convex region on  $\mathbb{C}$ .

Now let us recall some knowledge about fuzzy numbers. We shall only consider the class of all compact normal convex fuzzy numbers. Namely, a fuzzy set  $\tilde{a}$  on R is called a fuzzy number if and only if for every  $r \in [0,1]$   $[\tilde{a}]_r$  is a non-empty bounded closed interval. Thus, in our

case,  $\widetilde{R}$  is the set of all fuzzy numbers,  $\widetilde{R}^+$  ( resp.  $\widetilde{R}^-$  ) is the set of all positive ( resp. negative ) fuzzy numbers.

For the arithmetic operations on  $\widetilde{\mathbf{R}}$  we shall follow those of [1,2,7 and 8], namely, for  $\widetilde{\mathbf{a}},\widetilde{\mathbf{b}}\in\widetilde{\mathbf{R}}$  we have

$$(\tilde{a} + \tilde{b})(x) = \sup \{ \tilde{a}(y) \wedge \tilde{b}(z) : y \in S(\tilde{a}), z \in S(\tilde{b}), y + z = x \},$$

$$(\tilde{a} - \tilde{b})(x) = \sup{\tilde{a}(y) \wedge \tilde{b}(z) : y \in S(\tilde{a}), z \in S(\tilde{b}), y - z = x}$$

$$(\tilde{a} \cdot \tilde{b})(x) = \sup \{\tilde{a}(y) \wedge \tilde{b}(z) : y \in S(\tilde{a}), z \in S(\tilde{b}), y \cdot z = x\},$$

and if  $0 \in S(\tilde{b})$ ,

$$(\widetilde{a}/\widetilde{b})(x) = \sup \{\widetilde{a}(y) \wedge \widetilde{b}(z) : y \in S(\widetilde{a}), z \in S(\widetilde{b}), \forall /z = x \},$$
 where we admit  $\sup \phi = 0$  for the empty set  $\phi$ .

It is known that both addition and multiplication are associative and commutative. Furthermore, one can easily check that for any  $\tilde{a}, \tilde{b} \in \widetilde{R}$  and  $\lambda \in R$  we have

$$\lambda(\tilde{a} + \tilde{b}) = \lambda \tilde{a} + \lambda \tilde{b}. \tag{1.2}$$

and for any  $\tilde{a} \in \tilde{R}$ , if both  $d, \theta \in R^+$  or both  $d, \theta \in R^-$ , we have

$$(\alpha + \beta) \hat{a} = \alpha \hat{a} + \beta \hat{a} . \qquad (1.3)$$

We shall also use the following proposition which was given by Dubois and Prade [1,2]:

**Proposition 1.1.** Provided that  $\tilde{a}$  is either a positive or a negative fuzzy number and that  $\tilde{b}$  and  $\tilde{c}$  are together positive or negative fuzzy numbers, then

$$\tilde{a} \cdot (\tilde{b} + \tilde{c}) = \tilde{a} \cdot \tilde{b} + \tilde{a} \cdot \tilde{c}$$
.

In order to define fuzzy convex functions, we have to use an inequality relation between fuzzy numbers. The following definition is very popular (see, for examples, [3-7] and [10]):

**Definition 1.2.** Let  $\tilde{a}, \tilde{b} \in \tilde{R}$ . Then  $\tilde{a} \leq \tilde{b}$ , or write  $\tilde{b} \geqslant \tilde{a}$ , if and only if for all  $r \in ]0,1]$  we have  $\inf[\tilde{a}]_r \leq \inf[\tilde{b}]_r$  and  $\sup[\tilde{a}]_r \leq \sup[\tilde{b}]_r$ .

For the properties of  $\leq$ , we give the following proposition which will be used in the sequel.

Proposition 1.2. Let  $\tilde{a}, \tilde{b}, \tilde{c}, \tilde{d} \in \tilde{R}$  and  $\lambda \in R$ .

- (1). If  $\tilde{a} \leq \tilde{b}$  and  $\tilde{b} \leq \tilde{c}$ , then  $\tilde{a} \leq \tilde{c}$ .
- (2). If  $\tilde{a} \leq \tilde{b}$  and  $\tilde{c} \leq \tilde{d}$ , then  $\tilde{a} + \tilde{c} \leq \tilde{b} + \tilde{d}$ .
- (3). If  $\tilde{a} \leq \tilde{b}$ , then for  $\lambda \geq 0$  we have  $\lambda \tilde{a} \leq \lambda \tilde{b}$  and for  $\lambda \leq 0$   $\lambda \tilde{a} \geq \lambda \tilde{b}$ .
- (4). If  $\widetilde{a} \leq \widetilde{b}$  and  $\widetilde{a}, \widetilde{b} \in \widetilde{R}^+ \cup \widetilde{R}^-$ , then for  $\widetilde{c} \in \widetilde{R}^+$  we have  $\widetilde{c} \cdot \widetilde{a} \leq \widetilde{c} \cdot \widetilde{b}$  and for  $\widetilde{c} \in \widetilde{R}^ \widetilde{c} \cdot \widetilde{a} \geq \widetilde{c} \cdot \widetilde{b}$ .

Finally we conclude this preliminary section with a definition:

**Definition 1.3.** Let  $\underline{\mathbb{E}}$  be a subset of  $\widetilde{\mathbf{R}}$ . A mapping f from  $\underline{\mathbb{E}}$  to  $\widetilde{\mathbf{R}}$  is said to be non-decreasing on  $\underline{\mathbb{E}}$  if and only if for all  $\widetilde{\mathbf{a}}, \widetilde{\mathbf{b}} \in \underline{\mathbb{E}}$  such that  $\widetilde{\mathbf{a}} \lesssim \widetilde{\mathbf{b}}$  we have  $f(\widetilde{\mathbf{a}}) \lesssim f(\widetilde{\mathbf{b}})$ .

## 2. Fuzzy convex functions

Throughout this section,  $\underline{\mathbb{D}}$  will always denote a fuzzy convex region on  $\mathbf{R}^n$ .

**Definition 2.1.** A mapping f from  $\mathbb{D}$  to  $\widetilde{\mathbf{R}}$  is called a fuzzy convex function, or say that f is F-convex on  $\mathbb{D}$ , if and only if for all  $\widetilde{\mathbf{a}}$ ,  $\widetilde{\mathbf{b}}$   $\in \mathbb{D}$  and  $0 < \lambda < 1$  we have

$$f(\lambda \tilde{a} + (1-\lambda)\tilde{b}) \leq \lambda f(\tilde{a}) + (1-\lambda)f(\tilde{b}).$$
 (2.1)

**Definition 2.2.** A mapping f from  $\mathbb{D}$  to  $\widetilde{\mathbf{R}}$  is called a fuzzy concave function,or say that f is F-concave on  $\mathbb{D}$ , if and only if for all  $\widetilde{\mathbf{a}}$ ,  $\widetilde{\mathbf{b}}$   $\in \mathbb{D}$  and  $0<\lambda<1$  the inverse inequality of (2.1) holds.

It is easy to see from (1.2) and Proposition 1.2 (3) that a mapping f from  $\mathbb D$  to  $\widetilde{\mathbf R}$  is F-concave iff -f is F-convex. Thus in the following we shall mainly discuss fuzzy convex functions.

Let C be a non-empty ordinary convex subset of  $R^n$  and f be an ordinary convex function on C. Since C is a special fuzzy convex region and R is a set of special fuzzy numbers, of course  $f:C \rightarrow R$  is a special fuzzy convex function. Furthermore, Zadeh's extension principle [14] tells us that, for each fuzzy convex region D on C, f can be extended as a mapping from D to the family of all fuzzy sets on R such that for  $\widetilde{a} \in D$ 

$$f(\tilde{a})(t) = \sup_{x \in f^{-1}(t)} \tilde{a}(x) \quad \text{for } t \in \mathbb{R},$$
 (2.2)

and Nguyen [9] tells us that in the case f is continuous on C we have

$$[f(\tilde{a})]_r = f([\tilde{a}]_r)$$
 for  $r \in [0,1]$ ,

and thus, by the compactness and convexity of  $[\tilde{a}]_r$ , for each  $r \in [0,1]$   $[f(\tilde{a})]_r$  is a bounded closed interval, namely the fuzzy extension function defined by (2.2) is in fact a mapping from D to  $\tilde{R}$ .

Theorem 2.1. Let C be a non-empty ordinary convex subset of  $R^n$  and f be an ordinary convex function on C. If f is continuous on C, then for any fuzzy convex region  $D \subset \widetilde{C}$  the fuzzy extension function  $f:D \to \widetilde{R}$  defined by (2.2) is a fuzzy convex function.

**Proof.** Let  $\tilde{a}, \tilde{b} \in \mathbb{D}$  and  $0 < \lambda < 1$ . Since for each  $r \in [0,1]$   $[\tilde{a}]_r, [\tilde{b}]_r$  and  $[\lambda \tilde{a} + (1-\lambda)\tilde{b}]_r = \lambda [\tilde{a}]_r + (1-\lambda)[\tilde{b}]_r$  are all non-empty compact subset of  $\mathbf{C}$ , and since f is continuous, there should exist  $x_{\min} \in [\tilde{a}]_r, y_{\min} \in [\tilde{b}]_r$  and  $\lambda x^* + (1-\lambda)y^* \in [\lambda \tilde{a} + (1-\lambda)\tilde{b}]_r$  ( $x^* \in [\tilde{a}]_r, y^* \in [\tilde{b}]_r$ ) such that

$$f(x_{\min}) = \inf f(\tilde{l}al_r), \quad f(y_{\min}) = \inf f(\tilde{l}bl_r)$$
 (2.3)

and

$$f(\lambda x^* + (1 - \lambda)y^*) = \sup f([\lambda \tilde{a} + (1 - \lambda)\tilde{b}]_r). \qquad (2.4)$$

Note that we always have

$$f(\lambda x_{\min} + (1-\lambda)y_{\min}) \ge \inf f([\lambda \tilde{a} + (1-\lambda)\tilde{b}]_r),$$
 (2.5)

$$f(x^*) \le \sup f(\tilde{a}_r)$$
 and  $f(y^*) \le \sup f(\tilde{b}_r)$ , (2.6)

and note that also by applying Nguyen's theory [9] we have

$$\inf f([\widehat{a}]_r) = \inf [f(\widehat{a})]_r, \inf f([\widehat{b}]_r) = \inf [f(\widehat{b})]_r, \qquad (2.7)$$

$$\sup f(\tilde{a}_r) = \sup [f(\tilde{b})]_r, \sup f(\tilde{b}_r) = \sup [f(\tilde{b})]_r, \qquad (2.8)$$

$$\sup f([\lambda \tilde{a} + (1-\lambda)\tilde{b}]_r) = \sup [f(\lambda \tilde{a} + (1-\lambda)\tilde{b})]_r, \qquad (2.9)$$

and

$$\inf f([\lambda \tilde{a} + (1-\lambda)\tilde{b}]_r) = \inf [f(\lambda \tilde{a} + (1-\lambda)\tilde{b})]_r.$$
 (2.10)

Moreover, by the convexity of f we have

$$f(\lambda x_{\min} + (1-\lambda)y_{\min}) \leq \lambda f(x_{\min}) + (1-\lambda)f(y_{\min}), \qquad (2.11)$$

and

$$f(\lambda x^* + (1-\lambda)y^*) \le \lambda f(x^*) + (1-\lambda)f(y^*).$$
 (2.12)

Thus we have by (2.10),(2.5),(2.11),(2.3) and (2.7)

$$\inf [f(\lambda \tilde{a} + (1-\lambda)\tilde{b})]_r \leq \lambda (\inf [f(\tilde{a})]_r) + (1-\lambda) (\inf [f(\tilde{b})]_r), \qquad (2.13)$$

and by (2.9),(2.4),(2.12),(2.6) and (2.8)

$$\sup [f(\lambda \tilde{a} + (1-\lambda)\tilde{b})]_r \leq \lambda (\sup [f(\tilde{a})]_r) + (1-\lambda)(\sup [f(\tilde{b})]_r). \tag{2.14}$$

Furthermore, one can easily check that we have

$$\lambda(\inf[f(\widetilde{a})]_r) + (1-\lambda)(\inf[f(\widetilde{b})]_r) = \inf(\lambda[f(\widetilde{a})]_r + (1-\lambda)[f(\widetilde{b})]_r)$$
$$= \inf[\lambda f(\widetilde{a}) + (1-\lambda)f(\widetilde{b})]_r, \quad (2.15)$$

and

$$\lambda(\sup[f(\widetilde{a})]_r) + (1-\lambda)(\sup[f(\widetilde{b})]_r) = \sup[\lambda[f(\widetilde{a})]_r + (1-\lambda)[f(\widetilde{b})]_r)$$

$$= \sup[\lambda f(\widetilde{a}) + (1-\lambda)f(\widetilde{b})]_r. \quad (2.16)$$

Consequently we have by (2.13), (2.14), (2.15) and (2.16)

$$\inf[f(\lambda \tilde{a} + (1-\lambda)\tilde{b})]_r \leq \inf[\lambda f(\tilde{a}) + (1-\lambda)f(\tilde{b})]_r$$

and

$$\sup[f(\lambda \tilde{a} + (1-\lambda)\tilde{b})]_{r} \leq \sup[\lambda f(\tilde{a}) + (1-\lambda)f(\tilde{b})]_{r},$$

namely we have

$$f(\lambda \tilde{a} + (1-\lambda)\tilde{b}) \leq \lambda f(\tilde{a}) + (1-\lambda)f(\tilde{b}).$$

This completes the proof.

Now let us discuss some fundamental operations on fuzzy convex functions.

Theorem 2.2. If both f and g are F-convex on  $\mathbb{D}$ , then so are f+g and  $\lambda f$  ( $\lambda \ge 0$ ). Namely the family of all fuzzy convex functions on  $\mathbb{D}$  is closed under positive linear combinations.

Proof. Immediate from (1.2) and Proposition 1.2 (2) and (3).

Theorem 2.3. If f is F-convex on  $\mathbb D$  and if for all  $\widetilde{a} \in \mathbb D$  we have  $f(\widetilde{a}) \in \widetilde{R}^+$  or for all  $\widetilde{a} \in \mathbb D$  we have  $f(\widetilde{a}) \in \widetilde{R}^-$ , then for any  $\widetilde{\lambda} \in \widetilde{R}^+$ ,  $\widetilde{\lambda} \cdot f$  is a fuzzy convex function on  $\mathbb D$ .

Proof. Immediate from Proposition 1.1 and Proposition 1.2 (4).

Theorem 2.4. If f is a fuzzy convex function from  $\mathbb{D}$  to a fuzzy convex region  $\mathbb{E} \subset \widetilde{\mathbb{R}}$ , then for any a non-decreasing fuzzy convex function g on  $\mathbb{E}$ , the compound function  $g(f(\widetilde{x}))$  is F-convex on  $\mathbb{D}$ .

**Proof.** Let  $\tilde{a}, \tilde{b} \in D$  and  $0 < \lambda < 1$ . Since it is known that we have  $f(\lambda \tilde{a} + (1-\lambda)\tilde{b}) \leq \lambda f(\tilde{a}) + (1-\lambda)f(\tilde{b}),$ 

and since g is non-decreasing, of course we have

$$g(f(\lambda \tilde{a} + (1-\lambda)\tilde{b})) \leq g(\lambda f(\tilde{a}) + (1-\lambda)f(\tilde{b})),$$

and hence by the F-convexity of g we can get

$$g(f(\lambda \tilde{a} + (1-\lambda)\tilde{b}) \leq \lambda g(f(\tilde{a})) + (1-\lambda)g(f(\tilde{b})).$$

This completes the proof.

In order to obtain some more operations on fuzzy convex functions we need the following lemma:

**Lemma 2.5.** Let I be a non-empty interval of R and f be an ordinary non-decreasing function on I. If f is continuous on I, then for any

subset  $\mathbb{E}$  of  $\widetilde{\mathbf{I}}$ , the fuzzy extension function  $f: \mathbb{E} \to \widetilde{\mathbf{R}}$  defined by (2.2) is also non-decreasing.

Proof. Straightforward.

**Theorem 2.6.** Let I be a non-empty interval of R, and let f be a fuzzy convex function from D to  $\widetilde{I}$ . Then for any a non-decreasing ordinary convex function g on I, the compound function  $g(f(\widetilde{x}))$  is also a fuzzy convex function on D.

Proof. Immediate from Lemma 2.5, Theorem 2.1 and Theorem 2.4.

Corollary 2.7. If f is F-convex on  $\mathbb{D}$ , then so is  $e^{f(\widetilde{X})}$ .

**Corollary 2.8.** If f is F-convex and positive on  $\mathbb{D}$ , then so is  $[f(\widetilde{x})]^p$  for p > 1.

Corollary 2.9. If f is F-concave and positive on  $\mathbb{D}$ , then 1/f(x) is F-convex on  $\mathbb{D}$ .

**Proof.** It is known that -f is F-convex on  $\mathbb D$  and for all  $\widetilde a \in \mathbb D$  we have  $f(\widetilde a) \in \widetilde R^-$ . Further we know that g(t) = -1/t is an ordinary increasing convex function on  $R^-$ . Hence by virtue of Theorem 2.6 the compound function g(f(x)) = -1/(-f(x)) = 1/f(x) is F-convex on  $\mathbb D$ .

Finally we conclude our this fuzzy convex function theory with the following fuzzy extension of Jenson's inequality which will be useful in the theory of fuzzy-valued inequalities.

**Theorem 2.10.** Let f be a mapping from  $\mathbb{D}$  to  $\widetilde{\mathbf{R}}$ . Then f is F-convex on  $\mathbb{D}$  if and only if for any  $\widetilde{\mathbf{a}}_1, \dots, \widetilde{\mathbf{a}}_m \in \mathbb{D}$ 

$$\texttt{f}(\lambda_1 \widetilde{\texttt{a}}_1 + \cdots + \lambda_m \widetilde{\texttt{a}}_m) \lesssim \lambda_1 \texttt{f}(\widetilde{\texttt{a}}_1) + \cdots + \lambda_m \texttt{f}(\widetilde{\texttt{a}}_m)$$

whenever  $\lambda_1 \ge 0, \dots, \lambda_m \ge 0, \lambda_1 + \dots + \lambda_m = 1$ .

**Proof.** Analogous to the proof of Jenson's inequality in the non-fuzzy case.

## 3. Applications to fuzzy-valued inequalities

In paper [15], we have established a rudimentary theory of fuzzy-valued inequalities and have extended four classical inequalities to fuzzy-valued case. In this section we shall now use our fuzzy convex function theory to prove two more fuzzy-valued inequalities.

Inequality 1. Let  $\lambda_{i,j}$  (  $i,j=1,\dots,n$  ) be  $n^2$  positive real numbers satisfying  $\lambda_{i,1}+\dots+\lambda_{i,n}=1$  and  $\lambda_{1,j}+\dots+\lambda_{i,n}=1$  (  $i,j=1,\dots,n$  ) and let  $\widetilde{a}_1,\dots,\widetilde{a}_n\in\widetilde{R}^+$ . If  $\widetilde{b}_i=\lambda_{i,1}\widetilde{a}_1+\dots+\lambda_{i,n}\widetilde{a}_n$  (  $i=1,\dots,n$  ), then we have  $\widetilde{b}_1,\dots,\widetilde{b}_n\geq\widetilde{a}_1,\dots,\widetilde{a}_n$  (3.1)

**Proof.** Theorem 2.1 tells us that  $f(\tilde{x}) = -\ln \tilde{x}$  is F-convex on  $\tilde{R}^+$ . Hence for  $i = 1, \dots, n$  we have by Theorem 2.10

 $\ln \widetilde{b}_{i} = \ln(\lambda_{i} + \widetilde{a}_{1} + \cdots + \lambda_{i} + \widetilde{a}_{n}) \geq \lambda_{i} + \ln \widetilde{a}_{1} + \cdots + \lambda_{i} + \ln \widetilde{a}_{n}$ 

So that by (1.3) we have

$$\ln \widetilde{b}_1 + \cdots + \ln \widetilde{b}_n \geqslant \ln \widetilde{a}_1 + \cdots + \ln \widetilde{a}_n$$
 (3.2)

Since lnt is strictly increasing on  $R^+$ , so one can easily check that for any  $\widetilde{a},\widetilde{b}\in\widetilde{R^+}$  we have

$$\ln \tilde{a} + \ln \tilde{b} = \ln (\tilde{a} \cdot \tilde{b}) \tag{3.3}$$

and that  $\ln \tilde{a} \leq \ln \tilde{b}$  iff  $\tilde{a} \leq \tilde{b}$ . Therefore (3.2) is equivalent to (3.1).

Inequality 2. If  $\tilde{a}_1, \dots, \tilde{a}_n \in \widetilde{\mathbf{I}}$  (  $\mathbf{I} = \mathbb{J}0, 1\mathbb{I}$  ), then for any  $\lambda_1, \dots, \lambda_n \in \mathbb{R}^+$  such that  $\lambda_1 + \dots + \lambda_n = 1$  we have

$$\frac{\lambda_1}{1+\widetilde{a}_1}+\cdots+\frac{\lambda_n}{1+\widetilde{a}_n} \lesssim \frac{1}{1+\widetilde{a}_1^{\lambda_1}\cdots\widetilde{a}_n^{\lambda_n}}$$
 (3.4)

**Proof.** Let  $\widetilde{b}_1 = \ln \widetilde{a}_1$ . Then we have  $\widetilde{b}_1 \in \widetilde{R}^-$ . Further it is easy to check

that  $\tilde{a}_i = e^{\tilde{b}_i}$  (  $i=1,\cdots,n$  ). Since  $f(t) = -1/(1+e^t)$  is convex on  $R^-$ , it follows at once from Theorem 2.1 and Theorem 2.10 that we have

$$\frac{\lambda_1}{1+e^{\widetilde{b}_1}} + \cdots + \frac{\lambda_n}{1+e^{\widetilde{b}_n}} \lesssim \frac{1}{1+e^{\lambda_1 \widetilde{b}_1 + \cdots + \lambda_n \widetilde{b}_n}}.$$
 (3.5)

Note that for  $\tilde{a}, \tilde{b} \in \tilde{R}$  and  $\lambda \in R$  we also have  $e^{\tilde{a}+\tilde{b}} = e^{\tilde{a}} \cdot e^{\tilde{b}}$  and  $e^{\lambda \tilde{a}} = (e^{\tilde{a}})^{\lambda}$  (see [2]). Hence (3.5) is equivalent to (3.4).

### References

- [1] D.Dubois and H.Prade, Fuzzy real algebra: Some results, Fuzzy sets and Systems, 4(1979) 327-348.
- [2] D.Dubois and H.Prade, Fuzzy Sets and Systems: Theory and Applications, (Academic Press, New York, 1980).
- [3] D.Dubois and H.Prade, Comment on tolerance analysis using fuzzy sets "and" a procedure for multiple aspect decision making, Int. J. Systems Sci., 9(1978) 357-348.
- [4] A.N.S.Freeling, Fuzzy sets and decision analysis, IEEE Trans. Systems, Man Cybernet. 10(1980) 341-354.
- [5] R.Goetschel and W.Voxman, Eigen fuzzy number sets, Fuzzy Sets and Systems, 16(1985) 75-85.
- [6] O.Kaleva and S.Seikkala, On fuzzy metric spaces, Fuzzy Sets and Systems, 14(1984) 215 229.
- [7] M.Mizumoto and K.Tanaka, Some properties of fuzzy numbers, in: M.M. Gupta, R.K.Ragade and R.R.Yager, Eds., Advances in Fuzzy Set Theory and Applications (North-Holland, Amsterdam, 1979) 153-164.
- [8] M.Mizumoto and K.Tanaka, The four operations of arithmetic on fuzzy numbers, Systems Comput. Controls, 5(1976) 73 81.
- [9] H.T.Nguyen, A note on the extension principle for fuzzy sets, J. Math. Anal. Appl. 64(1978) 369-380.
- [10] J.Ramík and J.Římánek, Inequality relation between fuzzy numbers and its use in fuzzy optimization, Fuzzy Sets and Systems, 16(1985) 123 138.
- [11] R.T.Rockafellar, Convex Analysis (Princeton University Press, second printing, 1972).
- [12] M.D.Weiss, Fixed points, separation and induced topologies for fuzzy sets, J. Math. Anal. Appl. 50(1975) 142-150.
- [13] L.A. Zadeh, Fuzzy sets, Information and Control, 8(1965) 338 353.
- [14] L.A.Zadeh, The concept of a linguistic variable and its application to approximate reasoning, Part 1,2 and 3, Inform. Sci. 8, 199 249; 8, 301 357; 9, 43 80(1975).
- [15] Zhou Feiyue, On fuzzy-valued inequalities, Fuzzy Sets and Systems, To appear.