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FUZZY ALMOST LINEAR SPACES

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PREFACE

The hard sciences construct exact mathematical models of empirical phenomena and these models are then used to make predictions. However, some aspects of the real world always escape such precise mathematical models, and thus there is an elusive inexactness as part of the original model. The purpose of fuzziness is to introduce a unifying point of view to the notion of inexactness. The FUZZY in the sense used here was first introduced by Zadeh.

Essentially, fuzziness is a type of imprecision that stems from grouping of elements into classes that do not have sharply defined boundaries. Such classes called fuzzy sets arise, whenever we describe ambiguity, vagueness, and ambivalence in mathematical models of empirical phenomena. The theory of fuzzy sets has as one of its aims the development of a methodology for the formulation and solution of problems that are too complex or too ill defined to be susceptible to analysis by conventional techniques. Because of its unorthodoxy, it has been and will continue to be controversial for some time to

come.

The main aim of this dissertation is to define and study the notion of fuzzy almost linear spaces. This dissertation consists of two chapters, each beginning with brief introductions which summarize the material presented in that chapter. Chapter ONE is a survey aimed at clarifying the terminology to be used and recalling basic definitions and facts. this chapter is also devoted to the study of fuzzy vector spaces and almost linear spaces. In chapter TWO we present the notion of fuzzy almost linear spaces, sums and products of fuzzy sets, almost fuzzy subspaces, convex, balanced and absorbing fuzzy sets and results regarding fuzzy basis.

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CHAPTER ONE

DEFINITION AND BASIC CONCEPTS

The aim of this chapter is to present some basic concepts and to explain the terminology used throughout in this dissertation. In section 1.1, we discuss the concept of fuzzy sets and the membership function. In sections 1.2 and 1.3, a brief discussion on fuzzy linear spaces and almost linear spaces is given.

1.1 FUZZY SETS

Russel [10] gave the concept of Vagueness, subsequently Zadeh [13] achieved a remarkable success by introducing the notion of fuzzy set and some related mathematical concepts.

1.1.1 DEFINITION [13];

Let X be a nonempty set. A fuzzy set A in X is characterized by a membership function μ_A which associates to each $x \in X$ with its grade of membership $\mu_A(x) \in [0,1]$; μ_A is called the membership function of A .

For the sake of simplicity, we shall often consider a fuzzy set in X as a function $\mu : X \rightarrow [0,1]$, or

equivalently, as the set

$$\mu = \{(x, \mu(x)) : x \in X\}$$

of ordered pairs, where $\mu(x) \in [0, 1]$ for each $x \in X$.

NOTE;

If A is an ordinary set, then μ_A can take only two values 0 and 1, with

$$\mu_A(x) = \begin{cases} 1 & \text{if } x \in A \\ 0 & \text{if } x \notin A, \end{cases}$$

in this case $\mu_A = \chi_A$, the usual characteristic function of A .

1.1.2 EXAMPLE [7];

Let $X = \mathbb{N}$, the set of all natural numbers, and A be the set of small natural numbers. Then A is a fuzzy set.

Here μ_A may be given subjectively so that

$$\mu_A = \{(1, 1), (2, 0.8), (3, 0.6), (4, 0.4), (5, 0.2), (6, 0), (7, 0), (8, 0), \dots\}.$$

We may write

$$1 \underset{1}{\in} A, 2 \underset{0.8}{\in} A, 3 \underset{0.6}{\in} A, 4 \underset{0.4}{\in} A, 5 \underset{0.2}{\in} A,$$

and $n \underset{0}{\in} A$ for all $n \geq 6$.

1.1.3 EXAMPLE [7];

Let $X = \mathbb{R}$ (The set of all real numbers), and let $A = \{x \in \mathbb{R} : x \gg 0\}$ (the set of real numbers which are much

bigger than zero). Then A is a fuzzy set in X. A membership function μ_A might be subjectively defined by, say

$$\mu_A(x) = \begin{cases} \frac{x^2}{x^2 + 1} & \text{if } x \geq 0 \\ 0 & \text{if } x < 0. \end{cases}$$

Then $\mu_A(0.5) = 0.2, \mu_A(1) = 0.5, \mu_A(2) = 0.8, \dots$

The membership function has some resemblance to a probability function when X is a countable set. But there are essential differences between these concepts.

We next define some notions which are analogous to those in the usual set theory.

1.1.4 DEFINITION [7];

Let μ and λ be two fuzzy sets in X then,

(a) μ is called empty if $\mu = \chi_\phi$ i.e. if $\mu(x) = 0$ for all $x \in X$.

(b) μ is called subset of λ if $\mu \leq \lambda$ i.e. if $\mu(x) \leq \lambda(x)$

for all $x \in X$.

(c) μ is equal to λ if $\mu \equiv \lambda$ i.e. if $\mu(x) = \lambda(x)$ for all

$x \in X$.

(d) The pseudo complement μ^c of μ is defined by

$$\mu^c(x) = 1 - \mu(x) \text{ for all } x \in X.$$

(e) The union (or supremum) $\mu \vee \lambda$ is defined by

$$(\mu \vee \lambda)(x) = \mu(x) \vee \lambda(x) = \max\{\mu(x), \lambda(x)\}, x \in X.$$

(f) The intersection (or infimum) $\mu \wedge \lambda$ is defined by

$$(\mu \wedge \lambda)(x) = \mu(x) \wedge \lambda(x) = \min\{\mu(x), \lambda(x)\}, x \in X.$$

1.1.5 EXAMPLE [7];

Let $X = \{x_1, x_2, x_3, x_4\}$. Consider the fuzzy sets

$$\mu = \{(x_1, 0.4), (x_2, 0.2), (x_3, 0), (x_4, 1)\}.$$

$$\lambda = \{(x_1, 0.3), (x_2, 0.1), (x_3, 0), (x_4, 0.5)\}.$$

Then $\lambda \leq \mu$ since $0.3 < 0.4$, $0.1 < 0.2$, $0 = 0$, $0.5 < 1$. By

writing

$$x_x = \{(x_1, 1), (x_2, 1), (x_3, 1), (x_4, 1)\}, X \text{ is a subset of}$$

itself in the sense of the theory of fuzzy sets and in

this case

$$x_\phi < \lambda < \mu < x_{x'} \text{ where}$$

$$x_\phi = \{(x_1, 0), (x_2, 0), (x_3, 0), (x_4, 0)\}.$$

The next example is an illustration for pseudo-complement, union and intersection, this also shows that in general

$$\mu \vee \mu^c \neq x_{x'}, \mu \wedge \mu^c \neq x_\phi$$

1.1.6 EXAMPLE [7];

Let $X = \{x_1, x_2, x_3, x_4\}$. Consider

$$\mu = \{(x_1, 0.2), (x_2, 0.5), (x_3, 0), (x_4, 1)\}. \text{ Then,}$$

$$\mu^c = \{(x_1, 0.8), (x_2, 0.5), (x_3, 1), (x_4, 0)\},$$

$$\mu \vee \mu^c = \{(x_1, 0.8), (x_2, 0.5), (x_3, 1), (x_4, 1)\} \neq x_{x'},$$

$$\mu \wedge \mu^c = \{(x_1, 0.2), (x_2, 0.5), (x_3, 0), (x_4, 0)\} \neq x_\phi.$$

1.1.7 THEOREM [7, Theorem 2.7];

Let λ, μ, ν be fuzzy sets in X . Then

- (i) $(\lambda \vee \mu)^c = \lambda^c \wedge \mu^c$.
- (ii) $(\lambda \wedge \mu)^c = \lambda^c \vee \mu^c$.
- (iii) $\nu \vee (\lambda \wedge \mu) = (\nu \vee \lambda) \wedge (\nu \vee \mu)$.
- (iv) $\nu \wedge (\lambda \vee \mu) = (\nu \wedge \lambda) \vee (\nu \wedge \mu)$.
- (v) $(\mu^c)^c = \mu$.

PROOF;

- (i) For any $x \in X$,

$$\begin{aligned} (\lambda \vee \mu)^c(x) &= 1 - (\lambda \vee \mu)(x) = 1 - (\lambda(x) \vee \mu(x)) \\ &= (1 - \lambda(x)) \wedge (1 - \mu(x)) = \lambda^c(x) \wedge \mu^c(x). \end{aligned}$$

Thus $(\lambda \vee \mu)^c = \lambda^c \wedge \mu^c$.

- (ii) The proof is similar to the above part, it uses the relation

$$1 - (\lambda(x) \wedge \mu(x)) = (1 - \lambda(x)) \vee (1 - \mu(x)).$$

- (iii) This follows from the relation

$$\nu(x) \vee (\lambda(x) \wedge \mu(x)) = (\nu(x) \vee \lambda(x)) \wedge ((\nu(x) \vee \mu(x))).$$

- (iv) This follows from the relation

$$\nu(x) \wedge (\lambda(x) \vee \mu(x)) = (\nu(x) \wedge \lambda(x)) \vee ((\nu(x) \wedge \mu(x))).$$

- (v) This is trivial.

1.2 FUZZY LINEAR SPACES

Throughout this article E will denote a vector space

over K , where K is the field of real or complex numbers.

1.2.1 DEFINITION [6];

Let A_1, A_2, \dots, A_n be fuzzy sets in E . We define $A_1 \times A_2 \times A_3 \times \dots \times A_n$ to be the fuzzy set A in E , whose membership function is given by

$$\mu_A(x_1, x_2, \dots, x_n) = \min\{\mu_{A_1}(x_1), \mu_{A_2}(x_2), \dots, \mu_{A_n}(x_n)\}.$$

Let $\mu: E^n \rightarrow E$, $\mu(x_1, x_2, \dots, x_n) = x_1 + x_2 + \dots + x_n$.

We define $A_1 + A_2 + \dots + A_n = \mu(A)$. For λ a scalar and B a fuzzy set in E , we define $\lambda(B) = g(B)$, where $g: E \rightarrow E$, defined by $g(x) = \lambda x$.

Let T be a mapping from X to Y . Let B be a fuzzy set in Y with membership function μ_B . The inverse mapping T^{-1} induces a fuzzy set A whose membership function is defined by $\mu_A(x) = \mu_B(y) = \mu_B(Tx)$.

If A is a fuzzy set in X with membership function μ_A , then the membership function for the fuzzy set B in Y which is induced by this mapping T is

$$\mu_B(y) = \max_{x \in T^{-1}(y)} \mu_A(x).$$

$$x \in T^{-1}(y)$$

If T is one-one then $T^{-1}(y)$ is singleton and

$$\mu_B(y) = \mu_A(T^{-1}(y)).$$

For a fuzzy set A of X , the set $\{x \in X : \mu_A(x) > 0\}$ is called support of A and is denoted by $\text{supp } A$ or A_0 .

1.2.2 LEMMA [6, Lemma 2.1];

For $\lambda \neq 0$ $\mu_{\lambda B}(x) = \mu_B((1/\lambda)x)$ for all $x \in E$. For $\lambda = 0$,

$$\mu_{\lambda B}(x) = \begin{cases} 0, & \text{when } x \neq 0 \\ \sup_Y \mu_B(y), & \text{when } x = 0. \end{cases}$$

PROOF;

We know that

$$\mu_{\lambda B}(x) = \mu_B(g^{-1}(x)),$$

where $g(x) = \lambda x$,

$$g^{-1}(x) = (1/\lambda)x \text{ if } \lambda \neq 0, \text{ and } x \neq 0.$$

$$\therefore \mu_{\lambda B}(x) = \mu_B((1/\lambda)x).$$

If $x = 0$, then

$$g(0) = \lambda \cdot 0 = 0,$$

$$g^{-1}(0) = 0 \text{ and}$$

$$\mu_{\lambda B}(0) = \mu_B(0) = 0.$$

If $\lambda = 0$ then

$g(x) = 0$ and

$g(y) = 0$ for all $y \in E$.

$$\begin{aligned}\therefore \mu_{\lambda B}(x) &= \mu_B(g^{-1}(0)) \\ &= \mu_B(Y) \\ &= \sup_y \mu_B(y).\end{aligned}$$

1.2.3 COROLLARY [6, Corollary 2.3];

$\lambda(A + B) = \lambda A + \lambda B$ for all fuzzy sets A, B in E , and all scalars λ .

1.2.4 LEMMA [6 Lemma 2.4];

Let $A_1, \dots, A_n, B_1, \dots, B_m$ be fuzzy sets in E and put,

$$A = A_1 + A_2 + \dots + A_n, \quad B = B_1 + B_2 + \dots + B_m,$$

$$F = A_1 + A_2 + \dots + A_n + B_1 + B_2 + \dots + B_m.$$

Then $F = A + B$.

An ordinary subset A of E can be considered as a fuzzy set with membership function equal to characteristic function of A . In this way we may consider the sums of the form $A+B$ where one (or both) of A, B is an ordinary subset of E . For $x \in X$ and a fuzzy set, We define $x+B = \{x\}+B$. Let $f_x: E \rightarrow E, f_x(y) = x+y$.

1.2.5 LEMMA [6, Lemma 2.5];

Let A be an Ordinary subset of E and B be a fuzzy set. Then

- (i) $x+B = \bigcup_x (B)$;
- (ii) $\mu_{x+B}(z) = \mu_B(z-x)$;
- (iii) $A+B = \bigcup_{x \in B} (x+B)$.

1.2.6 PROPOSITION [6, Proposition 2.6];

Let A_1, A_2, \dots, A_n be fuzzy sets in E and $\lambda_1, \lambda_2, \dots, \lambda_n$ scalar. The following assertions are equivalent:

(i) $\lambda_1 A_1 + \lambda_2 A_2 + \dots + \lambda_n A_n \subset A$.

(ii) For all x_1, x_2, \dots, x_n in E , and we have

$$\mu_A(\lambda_1 x_1 + \lambda_2 x_2 + \dots + \lambda_n x_n) \geq \min\{\mu_{A_1}(x_1), \dots, \mu_{A_n}(x_n)\}.$$

1.2.7 LEMMA [6, Lemma 2.7];

Let A, B be fuzzy subsets of E . Then,

- (i) $A + OB \subset A$;
- (ii) $A + OB = A$ iff $\sup_{x \in E} \mu_A(x) \leq \sup_{x \in E} \mu_B(x)$.

1.2.8 DEFINITION [6];

A fuzzy set F in E is called a fuzzy subspace if,

- (i) $F + F \subset F$,
- (ii) $\lambda F \subset F$, for every scalar λ .

1.2.9 PROPOSITION [6, Proposition 3.1];

Let F be a fuzzy set in E . Then, the following are equivalent:

- (i) F is a subspace of E .
- (ii) For all scalars k, m , we have $kF + mF \subset F$.
- (iii) For all scalars k, m and all $x, y \in E$, we have

$$\mu_F(kx + my) \geq \min\{\mu_F(x), \mu_F(y)\}.$$

1.2.10 PROPOSITION [6, Proposition 3.3];

If A, B are fuzzy subspaces of E and k is a scalar, then $A + B$ and kA are fuzzy subspaces.

1.2.11 DEFINITION [6];

A fuzzy set A in E is said to be:

- (a) Convex if $kA + (1-k)A \subset A$ for all $k \in [0,1]$.
- (b) Balanced if $kA \subset A$ for all scalars k with $|k| \leq 1$.
- (c) Absorbing if $E = \bigcup_{k>0} kA$.

1.2.12 PROPOSITION [6, Proposition 4.1];

Let A be a fuzzy set in E . Then the following assertions are equivalent:

- (i) A is convex.
- (ii) $\mu_A(kx+(1-k)y) \geq \min\{\mu_A(x), \mu_A(y)\}$ for all $x, y \in X$ and all $k \in [0,1]$.
- (iii) For each $d \in [0,1]$, the set $A_d = \{x \in X: \mu_A(x) \geq d\}$ is convex.

1.2.13 PROPOSITION [6, Proposition 4.1];

Let A be a fuzzy set in E . Then the following assertions are equivalent:

- (i) A is balanced.
- (ii) $\mu_A(kx) \geq \mu_A(x)$ for all k with $|k| \leq 1$.
- (iii) $A_d = \{x \in X: \mu_A(x) \geq d\}$ for each $d \in [0,1]$ is balanced.

1.2.14 PROPOSITION [6, Proposition 4.4];

For a fuzzy set A of E , the following are equivalent:

- (i) A is absorbing.

(ii) For each $x \in X$, $\sup_{k > 0} \mu_A(kx) = 1$.

(iii) For each $d \in [0,1]$, the set $A_d = \{x \in E: \mu_A(x) \geq d\}$ is absorbing.

1.3 ALMOST LINEAR SPACES

An almost linear space is a nonempty set X together with two mappings $s: X \times X \rightarrow X$ and $m: \mathbb{R} \times X \rightarrow X$ satisfying the conditions, given below: For $x, y \in X$ and $\lambda \in \mathbb{R}$ we denote $s(x, y)$ by $x+y$ and $m(\lambda, x)$ by λx . Let $x, y, z \in X$ and $\lambda, \mu \in \mathbb{R}$,

$$L_1) \quad (x+y)+z = x+(y+z);$$

$$L_2) \quad x+y = y+x;$$

$$L_3) \quad \text{there exists an element } 0 \in X \text{ such that } x+0 = x;$$

$$L_4) \quad 1x = x;$$

$$L_5) \quad \lambda(x+y) = \lambda x + \lambda y;$$

$$L_6) \quad 0x = 0;$$

$$L_7) \quad \lambda(\mu x) = (\lambda\mu)x;$$

$$L_8) \quad (\lambda+\mu)x = \lambda x + \mu x \text{ for } \lambda, \mu \geq 0.$$

We denote $-1x$ by $-x$ when this will not lead to misunderstanding, and in the sequel $x-y$ means $x+(-y)$. A non empty set Y of an almost linear space X is called an almost

linear subspace of X , if for each $y_1, y_2 \in Y$ and $\lambda \in \mathbb{R}$, $y_1 + y_2 \in Y$ and $\lambda y_1 \in Y$. For an almost linear space X , the sets V_x and W_x are defined as:

$$V_x = \{x \in X : x - x = 0\},$$

$$W_x = \{x \in X : x = -x\}.$$

NOTE;

Throughout in section (1.3), and chapter 2, X will denote an almost linear space (als).

1.3.1 DEFINITION [2];

A subset B of X is called a basis of X if for each $x \in X \setminus \{0\}$ there exists unique sets $\{b_1, \dots, b_n\} \subset B$, $\{\lambda_1, \dots, \lambda_n\} \in \mathbb{R} \setminus \{0\}$ (n depending on x) such that,

$$x = \sum_{i=1}^n \lambda_i b_i, \text{ where } \lambda_i > 0 \text{ for } b_i \notin V_x.$$

Clearly, if B is a basis of X then $0 \notin B$.

1.3.2 LEMMA [2, Lemma 2.2];

If the als X has a basis B , then the sets $\{\alpha_b : b \in B, \alpha_b > 0 \text{ for } b \notin V_x\}$ are also basis of X .

1.3.3 LEMMA [2, Lemma 2.3];

Let X be an als with a basis and $x_1, x_2 \in X$. If $x_1 + x_2 \in$

V_x then $x_i \in V_x$.

PROOF;

Suppose $x_1 + x_2 \in V_x$ and let $x_3 = -x_1$ and $x_4 = -x_2$. since

X has a basis B , there exist $b_1, \dots, b_n \in B$, $b_i \neq b_j$ for $i \neq j$,

such that $x_i = \sum_{j=1}^n \alpha_{ij} b_j$, where $\alpha_{ij} > 0$ if $b_j \notin V_x$, $1 \leq i \leq 4$. By

hypothesis we get that $\sum_{i=1}^4 x_i = 0$ and so $\sum_{j=1}^n (\sum_{i=1}^4 \alpha_{ij}) b_j = 0$.

Suppose $b_1 \notin V_x$. Then,

$b_1 = (1 + \sum_{i=1}^4 \alpha_{i1}) b_1 + \sum_{j=2}^n (\sum_{i=1}^4 \alpha_{ij}) b_j = 0$. Since $b_1 \in B$, it

follows that $1 + \sum_{i=1}^4 \alpha_{i1} = 1$. But $\alpha_{i1} \geq 0$, $1 \leq i \leq n$, and so

$\alpha_{i1} = 0$, $1 \leq i \leq 4$. Consequently for each $b_j \notin V_x$, $1 \leq j \leq 1$,

we get $\alpha_{ij} = 0$, $1 \leq i \leq 4$, which shows that $x_i \in V_x$, $1 \leq i \leq 4$.

1.3.4 LEMMA [2, Lemma 2.4];

Let X be an als with a basis B . Then $B \cap V_x$ is a basis of V_x .

PROOF;

Let $x \in V_x$, then $x \in X$. Since X has a basis B , so for each $x \in X$ there exists unique $\{b_1, \dots, b_n\} \in B$, $\{\lambda_1, \dots, \lambda_n\} \in \mathbb{R} \setminus \{0\}$ such that,

$$x = \sum_{i=1}^n \lambda_i b_i, \text{ where } \lambda_i > 0 \text{ for } b_i \notin V_x.$$

So, $\sum_{i=1}^n \lambda_i b_i \in V_x$, $\Rightarrow \lambda_i b_i \in V_x$, by Lemma 1.3.3. Further

$$1/\lambda_i (\lambda_i b_i) \in V_x \text{ (Because } V_x \text{ is a linear space).}$$

$$\Rightarrow b_i \in V_x, \text{ so } b_i \in B \cap V_x.$$

Hence $B \cap V_x$ is a basis of V_x .

1.3.5 LEMMA [2, Lemma 2.5];

Let X be an almost linear space. The set $B \subset X$ is a basis of X iff $B \cap V_x$ is a basis of V_x and for each $x \in X \setminus V_x$ there exists unique $\{b_1, \dots, b_n\} \in B \setminus V_x$, $v \in V_x$ and $\lambda_1, \dots, \lambda_n > 0$, such that

$$x = \sum_{i=1}^n \lambda_i b_i + v.$$

PROOF;

Suppose that $B \subset X$ is a basis of X . By Lemma 1.3.4 $B \cap V_x$ is a basis of V_x . Now suppose that $x \in X \setminus V_x$ and by definition of basis for each $x \in X$ there exists unique $\{b_1,$

$\dots, b_n \} \in B \setminus V_x$, $\{\lambda_1, \dots, \lambda_n\} \in \mathbb{R} \setminus \{0\}$ such that

$$x = \sum_{i=1}^n \lambda_i b_i, \text{ where } \lambda_i > 0 \text{ for } b_i \notin V_x.$$

Now suppose for $i = 1, \dots, m$, $b_i \notin V_x$, and $b_i \in V_x$ for $i =$

$m+1, \dots, n$. For $b_i \in V_x$, we may have $\lambda_i < 0$.

$$x = \sum_{i=1}^m \lambda_i b_i + \sum_{i=m+1}^n \lambda_i b_i.$$

As $b_i \in V_x$, then $\lambda_i b_i$ and $\sum_{i=m+1}^n \lambda_i b_i \in V_x$ ($\because V_x$ is a linear

spaces), put $v = \sum_{i=m+1}^n \lambda_i b_i$, then $x = \sum_{i=1}^m \lambda_i b_i + v$.

Conversely;

Suppose that $x \in X$, then either $x \in V_x$ or $x \notin V_x$.

If $x \in V_x$, then $B \cap V_x$ is a basis of V_x , by given condition.

If $x \notin V_x$, then $x \in X \setminus V_x$ and again by given condition for

each $x \in X \setminus V_x$ there exists unique $\{b_1, \dots, b_n\} \in B \setminus V_x$ $\{\lambda_1,$

$\dots, \lambda_n] \in R \setminus \{0\}$ such that

$$x = \sum_{i=1}^n \lambda_i b_i + v$$

$$\text{As } v \in V_x \text{ so } v = \sum_{i=1}^m \lambda_i b_i$$

Hence B is a basis of X .

1.3.6 LEMMA [2, Lemma 2.6];

Let B be a basis of the als X . Then for each $b \in B \setminus V_x$ there exist unique $\psi(b) \in B \setminus V_x$, $V(b) \in V_x$ and $\lambda(b) > 0$ such that $-b = \lambda(b) \psi(b) + V(b)$.

PROOF;

Let $b \in B \setminus V_x$. Then $-b \notin V_x$ and by Lemma 1.3.5 we get

$$-b = \sum_{i=1}^k \lambda_i b_i + v \quad (i)$$

where $b_1, \dots, b_k \in B \setminus V_x$, $k \geq 1$, $b_i \neq b_j$ for $i \neq j$, $v \in V_x$ and $\lambda_i > 0$, $1 \leq i \leq k$, are uniquely determined. Clearly the Lemma is proved if we show that $k = 1$. Let $e_1, \dots, e_m \in B \setminus V_x$, $e_i \neq e_j$ for $i \neq j$, $v_i \in V_x$, $U_{ij} \geq 0$ $1 \leq i \leq k$, $1 \leq j \leq m$, s.t

$$-b_i = \sum_{j=1}^m U_{ij} e_j + v_i \quad (1 \leq i \leq k) \quad (ii)$$

Multiplying (i) by -1 and using (ii) we get

$$b = \sum_{j=1}^m \left(\sum_{i=1}^k \lambda_i u_{ij} \right) e_j + \sum_{i=1}^k \lambda_i v_i + v \quad (\text{iii})$$

Since $b \in B \setminus V_x$, there exists an index $j_0 \in \{1, \dots, m\}$ -

say $j_0 = 1$ such that $b = e_1$ and we must have

$$\sum_{i=1}^k \lambda_i u_{ij} = 0, \quad 2 \leq j \leq m.$$

Since $\lambda_i > 0$ and $u_{ij} \geq 0$ it follows that $u_{ij} = 0$ for each $1 \leq i$

$\leq k$ and each $2 \leq j \leq m$. Consequently, we get by (ii)

$$-b_i = u_{i1} e_1 + v_i \quad (1 \leq i \leq k) \quad (\text{iv})$$

and $u_{ij} \geq 0$, since $-b_i \notin V_x$, $1 \leq i \leq k$. Suppose $k > 1$. By (iv)

for $i = 1, 2$ we get that $e = (-b_1 - v_1)/u_{11} = (-b_2 - v_2)/u_{21}$

and so $b_1 = (u_{11}/u_{21})b_2 + ((v_2/u_{21}) - (v_1/u_{11}))$, contradicting

Lemma 1.3.5.

Let $\psi : B \setminus V_x \longrightarrow B \setminus V_x$ be defined as in Lemma 1.3.6. Then ψ is well-defined and we have :

1.3.7 LEMMA [2, Lemma 2.7];

The mapping $\psi : B \setminus V_x \longrightarrow B \setminus V_x$ defined as above is injective and $\psi(\psi(b)) = b$ for each $b \in B \setminus V_x$. In particular ψ is surjective.

PROOF;

Let $b_1, b_2 \in V_x$ such that $\psi(b_1) = \psi(b_2) = b \in B \setminus V_x$. Then $-b_i = \lambda_i b + v_i$, $\lambda_i > 0$, $v_i \in V_x$, $i = 1, 2$, and similarly with the proof given at the end of Lemma 1.3.6 this contradicts Lemma 1.3.5,

Let now $b \in B \setminus V_x$. Then $-b = \lambda \psi(b) + v$, where $\lambda > 0$, $v \in V_x$ and $\psi(b) \in B \setminus V_x$ are given by Lemma 1.3.6. Then $-\psi(b) = (b/\lambda) + (v/\lambda)$, and so, again by Lemma 1.3.6 we get $\psi(\psi(b)) = b$.

1.3.8 THEOREM [2, Lemma 2.8];

Let B be a basis of the als X . Then there exists a basis B' of X with the property that for each $b' \in B' \setminus V_x$, we have $-b' \in B' \setminus V_x$.

CHAPTER TWO

FUZZY ALMOST LINEAR SPACES

Godini [2] gave the notion of almost linear spaces. The aim of this chapter is to define and study the fuzzy almost linear spaces. Section 2.1 contains sums and products of fuzzy sets in almost linear spaces. Section 2.2 deals with the almost fuzzy subspaces. Section 2.3 contains convex, balanced, absorbing fuzzy sets, in almost linear spaces. Section 2.4 includes the definition of fuzzy basis and its relevant results. Throughout this chapter F will denote a fuzzy subset of X .

2.1 SUMS AND PRODUCTS OF FUZZY SETS IN ALMOST LINEAR SPACES

2.1.1 DEFINITION;

Let A_1, \dots, A_n be fuzzy sets in X^n . We define $A_1 \times A_2 \times \dots \times A_n$ to be the fuzzy set A in X whose membership function is given by

$$\mu_A(x_1, \dots, x_n) = \min\{\mu_{A_1}(x_1), \dots, \mu_{A_n}(x_n)\}.$$

Let $f: X^n \rightarrow X$, be defined by $f(x_1, \dots, x_n) =$

$$x_1 + \dots + x_n.$$

Define $A_1 + \dots + A_n = f(A)$. If A and B are fuzzy sets then the membership function of A+B is given by

$$\mu_{A+B}(x) = \sup_{x=a+b} \{\min\{\mu_A(a), \mu_B(b)\}\}.$$

For λ a scalar and B a fuzzy set in X, define $\lambda B = g(B)$ where $g: X \rightarrow X$, $g(x) = \lambda x$ and

$$\mu_{\lambda B}(x) = \begin{cases} \mu_B(g^{-1}(x)) = \mu_B((1/\lambda)x) & \text{if } \lambda \neq 0 \text{ and all } x \in E \\ 0 & \text{if } \lambda = 0 \text{ and } x \neq 0, \\ \sup_y \mu_B(y) & \text{if } \lambda = 0 \text{ and } x = 0. \end{cases}$$

Let $f_x: X \rightarrow X$ be defined by $f_x(y) = x + y$.

2.1.2 LEMMA;

Let A be any set and B a fuzzy set in X.

Then for $x \in X$,

$$(i) \quad x + B = f_x(B),$$

$$(ii) \quad \mu_{x+B}(a) = \begin{cases} \sup_{a=x+y} \mu_B(y) , \\ 0 & \text{if } a \neq x + y, \end{cases}$$

$$(iii) \quad A + B = \bigcup_{x \in A} x + B .$$

PROOF;

(i). As $x + B = \{x\} + B$, therefore by definition of sum

of two fuzzy sets

$$\begin{aligned} \mu_{x+B}(a) &= \sup_{a=z+y} \{\min\{\mu_{\{x\}}(z), \mu_B(y)\}\} \\ &= \begin{cases} \sup_{a=z+y} \{\mu_B(y)\} & \text{if } \mu_{\{x\}}(z)=1, \\ 0 & \text{if } \mu_{\{x\}}(z)=0, \text{ and } a \neq \\ & x+y \text{ for } y \in X \end{cases} \end{aligned}$$

On the other hand

$$\begin{aligned} \mu_{f_x^{-1}(B)}(a) &= \sup_{y \in f_x^{-1}(B)} \mu_B(y) \\ &= \begin{cases} \sup_{a=x+y} \mu_B(y) & \text{if } a = x+y \text{ for some } y \in X, \\ 0 & \text{if } a \neq x+y \text{ for any } y \in X. \end{cases} \end{aligned}$$

Thus $\{x\} + B = f_x^{-1}(B)$.

(ii) It follows from the proof of (i).

(iii) Since $\mu_{A+B}(x) = \sup_{x=a+b} \{\min\{\mu_A(a), \mu_B(b)\}\}$

$$= \begin{cases} \sup_{x=a+b} \mu_B(b) \\ 0 \end{cases}$$

Hence $A + B = \bigcup_{x \in A} x + B$.

2.1.3 PROPOSITION;

Let A, A_1, \dots, A_n be fuzzy sets in X and $\lambda_1, \lambda_2, \dots$

..., $\lambda_n \in \mathbb{R}$.

Then the following are equivalent:

(i) $\lambda_1 A_1 + \dots + \lambda_n A_n \subset A$.

(ii) For all $x_1, \dots, x_n \in X$ we have

$$\mu_A(\lambda_1 x_1 + \dots + \lambda_n x_n) \geq \min\{\mu_{A_1}(x_1), \dots, \mu_{A_n}(x_n)\}.$$

PROOF;

(i) \Rightarrow (ii):

$$\begin{aligned} \mu_A(\lambda_1 A_1 + \lambda_2 A_2 + \dots + \lambda_n A_n) &\geq \mu_{\lambda_1 A_1 + \dots + \lambda_n A_n}(\lambda_1 x_1 + \dots + \lambda_n x_n) \\ &\geq \min\{\mu_{\lambda_1 A_1}(\lambda_1 x_1), \dots, \mu_{\lambda_n A_n}(\lambda_n x_n)\} \\ &\geq \min\{\mu_{A_1}(x_1), \dots, \mu_{A_n}(x_n)\}. \end{aligned}$$

Now (ii) \Rightarrow (i):

By rearranging the order if necessary, we may assume

that $\lambda_i \neq 0$ for $i = 1, 2, \dots, k$, and $\lambda_i = 0$ for

$k < i \leq n$. Let x_1, x_2, \dots, x_n be elements of E . For

all y_1, y_2, \dots, y_{n-k} in E we have

$$\begin{aligned} \mu_A(\lambda_1 x_1 + \dots + \lambda_n x_n) &\geq \{\mu_{A_1}(x_1), \dots, \mu_{A_k}(x_k), \mu_{A_1}(y_1), \dots \\ &\quad, \mu_{A_{n-k}}(y_{n-k})\}. \end{aligned}$$

n

Since $\mu_{O\Lambda_i}(0) = \sup_{y \in E} \mu_{\Lambda_i}(y)$, we get

$$\mu_{\Lambda}(\lambda_1 x_1 + \dots + \lambda_n x_n) \geq \min\{\mu_{\Lambda_1}(x_1), \dots, \mu_{\Lambda_k}(x_k), \dots, \mu_{O\Lambda_{k+1}}(0), \dots, \mu_{O\Lambda_n}(0)\}.$$

Now,

$$\begin{aligned} & \mu_{\lambda_1 \Lambda_1 + \dots + \lambda_n \Lambda_n}(z) \\ &= \sup_{x_1 + x_2 + \dots + x_k = z} [\min\{\mu_{\lambda_1 \Lambda_1}(x_1), \dots, \mu_{\lambda_n \Lambda_n}(x_n)\}] \\ &= \sup_{x_1 + \dots + x_k = z} [\min\{\mu_{\lambda_1 \Lambda_1}(x_1), \dots, \mu_{\lambda_k \Lambda_k}(x_k), \mu_{O\Lambda_{k+1}}(0), \dots, \mu_{O\Lambda_n}(0)\}] \\ &= \sup_{x_1 + \dots + x_k = z} [\min\{\mu_{\Lambda_1}((1/\lambda_1)x_1), \dots, \mu_{\Lambda_k}((1/\lambda_k)x_k), \mu_{O\Lambda_{k+1}}(0), \dots, \mu_{O\Lambda_n}(0)\}] \\ &\leq \sup_{x_1 + \dots + x_k = z} \mu_{\Lambda_1}(\lambda_1(1/\lambda_1)x_1 + \dots + \lambda_k(1/\lambda_k)x_k) = \mu_{\Lambda}(z) \end{aligned}$$

2.1.4 LEMMA;

Let A and B be two fuzzy sets in X. Then

$$(i) \quad A + 0B \subset A;$$

$$(ii) \quad A + 0B = A \quad \text{iff} \quad \sup_{x \in X} \mu_A(x) \leq \sup_{x \in X} \mu_B(x).$$

PROOF;

$$(i) \quad \mu_{A+0B}(x) = \mu_A(x) \geq \min\{\mu_A(x), \mu_B(y)\}. \text{ Hence (i)}$$

follows from proposition 2.1.3.

$$(ii) \quad \text{Suppose that } \sup \mu_A(x) \leq \sup \mu_B(x) = \mu_{0B}(0).$$

$$\begin{aligned} \text{Then } \mu_{A+0B}(z) &= \sup_{x+y=z} [\min\{\mu_A(x), \mu_{0B}(y)\}] = \\ & \min\{\mu_A(z), \mu_{0B}(0)\} = \mu_A(z). \end{aligned}$$

On the other hand, if $\mu_A(z) > \sup \mu_B(x) =$

$\mu_{0B}(0)$ for some z, then

$$\mu_{A+0B}(z) = \min\{\mu_A(z), \mu_{0B}(0)\} < \mu_A(z),$$

and hence $A + 0B \neq A$.

2.2 ALMOST FUZZY SUBSPACES

2.2.1 DEFINITION;

A nonempty fuzzy set F in X is called a fuzzy almost linear subspace of X if

- (i) $F + F \subset F$,
- (ii) $\lambda F \subset F$ for all $\lambda \in \mathbb{R}$.

2.2.2 PROPOSITION;

Let F be a fuzzy set in X . Then the following are equivalent:

- (i) F is a fuzzy almost linear subspace of X .
- (ii) For all $k, m \in \mathbb{R}$, $kF + mF \subset F$.
- (iii) For all $k, m \in \mathbb{R}$ and $x, y \in X$,

$$\mu_F(kx + my) \geq \min\{\mu_F(x), \mu_F(y)\}.$$

PROOF;

$$\begin{aligned} \text{(i)} \quad \Rightarrow \quad \text{(ii):} \quad \mu_{kF+mF}(x) &= \sup_{x=a+b} \{\min\{\mu_{kF}(a), \\ &\quad \mu_{mF}(b)\}\} \\ &\leq \sup_{x=a+b} \{\min\{\mu_F(a), \mu_F(b)\}\} \\ &= \mu_{F+F}(x) \leq \mu_F(x). \end{aligned}$$

Therefore $kF + mF \subseteq F$.

- (ii) \Rightarrow (i): Take $k = 1 = m$, then $F + F \subset F$. Also if $m = 0$, then by Lemma 2.1.4, $kF = kF + oF \subset F$.

(ii) \Leftrightarrow (iii): Follows from Proposition 2.1.3.

2.2.3 LEMMA;

If A is an almost linear subspace of X then μ_A (characteristic function of A) is a fuzzy almost linear subspace of X .

PROOF;

Let $x, y \in X$ and $k, m \in \mathbb{R}$. If $kx + my \in A$, then

$$\mu_A(kx + my) = 1 \geq \min\{\mu_A(x), \mu_A(y)\} .$$

If $kx + my \notin A$ then either x or y does not belong to A and so $\min\{\mu_A(x), \mu_A(y)\} = 0$. Therefore

$$\mu_A(kx + my) \geq \min\{\mu_A(x), \mu_A(y)\} .$$

Thus by Proposition 2.2.2, A is a fuzzy almost linear subspace of X .

2.2.4 LEMMA;

If F is a fuzzy almost linear subspace of X , then the following sets are almost linear subspaces of X :

$$(i) \quad F_1 = \{x \in X: \mu_F(x) = \mu_F(o)\}.$$

$$(ii) \quad F_o = \{x \in X: \mu_F(x) > 0\}.$$

PROOF;

(i). Obviously, F_1 is nonempty. If $x, y \in F_1$ then

$$\begin{aligned} \mu_F(x) &= \mu_F(y) = \mu_F(o) \text{ and } \mu_F(x+y) \geq \min\{\mu_F(x), \mu_F(y)\} \\ &= \mu_F(o). \end{aligned}$$

But $\mu_F(o) \geq \mu_F(x)$ for all $x \in X$. Therefore $\mu_F(x+y) = \mu_F(o)$.

Thus $x+y \in F_1$. Also $\mu_F(\lambda x) \geq \mu_F(x) = \mu_F(o)$. It implies

that $\mu_F(\lambda x) = \mu_F(o)$. Thus $\lambda x \in F_1$. Hence F_1 is an almost

linear subspace of X .

(ii). Obviously F_o is nonempty. Let $x, y \in F_o$ then

$$\begin{aligned} \mu_F(x) > 0 \text{ and } \mu_F(y) > 0. \text{ Thus } \mu_F(x+y) &\geq \min\{\mu_F(x), \mu_F(y)\} \\ &> 0. \end{aligned}$$

Therefore $x+y \in F_o$. Also $\mu_F(\lambda x) \geq \mu_F(x) > 0$. Thus $\lambda x \in F_o$.

Hence F_\circ is an almost linear subspace of X .

2.2.5 PROPOSITION;

If A and B are fuzzy almost linear subspaces of X and k is a non zero real number, then $A+B$ and kA are fuzzy almost linear subspaces.

PROOF;

Let $x, y \in X$, then

$$\mu_{A+B}(x) = \sup_{x=a+b} \{\mu_A(a) \wedge \mu_B(b)\},$$

and

$$\mu_{A+B}(y) = \sup_{y=c+d} \{\mu_A(c) \wedge \mu_B(d)\}.$$

It further implies that,

$$\begin{aligned} \mu_{A+B}(x) \wedge \mu_{A+B}(y) &= \inf \{ \sup_{x=a+b} \{\mu_A(a) \wedge \mu_B(b)\}, \\ &\quad \sup_{y=c+d} \{\mu_A(c) \wedge \mu_B(d)\} \}. \\ &= \sup_{x=a+b} \sup_{y=c+d} \{\mu_A(a) \wedge \mu_B(b)\} \wedge \\ &\quad \{\mu_A(c) \wedge \mu_B(d)\}. \\ &\leq \sup_{x+y=e+f} \{\mu_A(e) \wedge \mu_B(f)\}. \end{aligned}$$

$$= \mu_{A+B}(x+y) . \quad (1)$$

Also,

$$\begin{aligned} \mu_{A+B}(x) &= \sup_{x=a+b} \{ \mu_A(a) \wedge \mu_B(b) \} \\ &\leq \sup_{x=a+b} \{ \mu_A(ka) \wedge \mu_B(kb) \} \\ &\leq \sup_{kx=ka+kb} \{ \mu_A(ka) \wedge \mu_B(kb) \} \\ &= \mu_{A+B}(kx) . \end{aligned} \quad (2)$$

From (1) and (2)

$$\mu_A(kx+my) \geq \min\{ \mu_{A+B}(x), \mu_{A+B}(y) \} .$$

Thus $A+B$ is a fuzzy almost linear subspace of X .

Now we show that kA is a fuzzy almost linear subspace of X . Let $x, y \in X$ and $m, n \in \mathbb{R}$, then

$$\begin{aligned} \mu_{kA}(mx+ny) &= \mu_A\left(\frac{1}{k}(mx+ny)\right) \\ &= \mu_A\left(\frac{m}{k}x + \frac{n}{k}y\right) \\ &\geq \min\left\{ \mu_A\left(\frac{x}{k}\right), \mu_A\left(\frac{y}{k}\right) \right\} \end{aligned}$$

$$= \min\{\mu_{kA}(x), \mu_{kA}(y)\}.$$

Thus by Proposition 2.2.2 kA is a fuzzy almost linear subspace of X .

2.3 CONVEX, BALANCED AND ABSORBING FUZZY SETS IN ALMOST LINEAR SUBSPACES

2.3.1 DEFINITION;

A fuzzy set A in X is said to be:

- (a) Convex if $kA + (1-k)A \subset A$ for all $k \in [0,1]$.
- (b) Balanced if $kA \subset A$ for all scalars k with $|k| \leq 1$
- (c) Absorbing if $X = \bigcup_{k>0} kA$.

2.3.2 PROPOSITION;

Let A be a fuzzy set in X . Then the following assertions are equivalent:

- (i) A is convex.
- (ii) $\mu_A(kx+(1-k)y) \geq \min\{\mu_A(x), \mu_A(y)\}$ for all $x, y \in X$ and all $k \in [0,1]$.
- (iii). For each $d \in [0,1]$, the set $A_d = \{x \in X: \mu_A(x) \geq d\}$

is convex.

PROOF;

(i) \Rightarrow (ii): From the definition of convex fuzzy set, $kA+(1-k)A \subset A$ for all $k \in [0,1]$. Proposition 2.1.3 further implies

$$\mu_A(kx+(1-k)y) \geq \min\{\mu_A(x), \mu_A(y)\}.$$

(ii) \Rightarrow (i): Follows from Proposition 2.1.3.

(ii) \Rightarrow (iii): Let $x, y \in A_d$ then $\mu_A(x)$ and $\mu_A(y) \geq d$.

By hypothesis $\mu_A(kx+(1-k)y) \geq \min\{\mu_A(x), \mu_A(y)\} \geq d$.

Therefore

$$kx + (1-k)y \in A_d.$$

Hence A_d is a convex set.

(iii) \Rightarrow (ii): Let $x, y \in X$ and $\mu_A(x) \leq \mu_A(y)$. Take $\mu_A(x) = d$ then $x, y \in A_d$ and so by (iii) $kx+(1-k)y \in A_d$. Thus $\mu_A(kx+(1-k)y) \geq d$.

Hence $\mu_A(kx+(1-k)y) \geq \min\{\mu_A(x), \mu_A(y)\}$.

2.3.3 PROPOSITION;

Let A be a fuzzy set in X . Then the following assertions are equivalent:

(i). A is balanced.

(ii). $\mu_A(kx) \geq \mu_A(x)$ for all x with $|k| \leq 1$.

(iii). $A_d = \{x \in X: \mu_A(x) \geq d\}$ for each $d \in [0,1]$ is balanced.

PROOF;

(i) \Rightarrow (ii): By the definition of balanced fuzzy set $kA \subset A$. Therefore $\mu_A(kx) \geq \mu_{kA}(kx)$ for all $x \in X$ and scalars k with $|k| \leq 1$. Hence $\mu_A(kx) \geq \mu_A\left(\frac{1}{k}(kx)\right) = \mu_A(x)$.

(ii) \Rightarrow (i): By hypothesis, $\mu_A(kx) \geq \mu_A(x)$ for all k with $|k| \leq 1$. Therefore $\mu_A(x) \geq \mu_A\left(\frac{1}{k}x\right) = \mu_{kA}(x)$. Hence $kA \subset A$ for all scalars k with $|k| \leq 1$.

(ii) \Rightarrow (iii): Let $x \in A_d$ then $\mu_A(x) \geq d$. By hypothesis, $\mu_A(kx) \geq \mu_A(x)$. Therefore $\mu_A(kx) \geq d$ for $|k| \leq 1$. Thus $kx \in A_d$ for $|k| \leq 1$. Hence A_d is balanced.

(iii) \Rightarrow (ii): Let $x \in X$ and take $d = \mu_A(x)$ then A_d is balanced and $x \in A_d$. Therefore $kx \in A_d$ for all scalar k with $|k| \leq 1$. Thus $\mu_A(kx) \geq d$. Hence $\mu_A(kx) \geq \mu_A(x)$.

2.3.4 PROPOSITION;

For a fuzzy set A of X , the following are equivalent:

- (i) A is absorbing.
- (ii) For each $x \in X$, $\sup_{k>0} \mu_A(kx) = 1$.
- (iii) For each $d \in [0,1]$, the set $A_d = \{x \in X: \mu_A(x) \geq d\}$ is absorbing.

PROOF;

(i) \Rightarrow (ii): Let A be an absorbing fuzzy set, then by definition $\bigcup_{k>0} kA = X$. Therefore for each $x \in X$

$$\mu_{\bigcup_{k>0} kA}(x) = \mu_X(x).$$

Thus $\sup_{k>0} \mu_{kA}(x) = \mu_X(x) = 1$. In particular $\sup_{k>0} \mu_{kA}(k^2x) = 1$.

Hence $\sup_{k>0} \mu_A(kx) = 1$.

(ii) \Rightarrow (i): By hypothesis for each $x \in X$, $\sup_{k>0} \mu_A(kx) = 1$

Therefore $\sup_{k>0} \mu_A((1/k)x) = 1$. Thus $\sup_{k>0} \mu_{kA}(x) = 1$.

Hence $\bigcup_{k>0} kA = X$.

(ii) \Rightarrow (iii): Clearly $\bigcup_{k>0} kA_d \subseteq X$. Let $x \in X$ then $\sup_{k>0}$

$\mu_A(kx) = 1$. Therefore there exists $k_0 > 0$ such that $\mu_A((1/k_0)x) = 1$. It further implies that $\frac{1}{k_0}x \in A_d$. Thus $x \in k_0 A_d$. Hence $X = \bigcup_{k>0} kA_d$.

(iii) \Rightarrow (ii): Since A_d is absorbing therefore $\bigcup_{k>0} kA_d = X$. If $d = 1$, then $\bigcup_{k>0} kA_1 = X$. Let $x \in X$ then $x \in \bigcup_{k>0} kA_1$. Thus there exists $k_0 > 0$ such that $x \in k_0 A_1$. It follows that $x = k_0 y$ where $y \in A_1$. Therefore $\mu_A(\frac{1}{k_0}x) = \mu_A(y) = 1$.

Hence $\sup_{k>0} \mu_A(kx) = 1$.

2.4 FUZZY BASIS

2.4.1 DEFINITION;

Let F be a fuzzy almost linear subspace of X . A fuzzy subset B of X is said to be a fuzzy basis of F if for each $0 \neq x \in X$, with $\mu_F(x) > 0$ there exist unique sets $\{b_1, b_2, \dots, b_n\} \subset X$, $\{\lambda_1, \dots, \lambda_n\} \subset \mathbb{R} \setminus \{0\}$ (b_i depending on x) with $\mu_B(b_i) > 0$ such that

$$x = \sum_{i=1}^n \lambda_i b_i \quad \text{where } \lambda_j > 0 \text{ for } b_j \notin V_x.$$

Clearly if B is a fuzzy basis of F then B_0 is a basis for F_0 . Conversely, if for some fuzzy set B , B_0 is a basis for

F_0 , then B is a fuzzy basis for F .

2.4.2 LEMMA;

If a fuzzy set B is a fuzzy basis for a fuzzy almost linear subspace F of X then aB is also a fuzzy basis for F , where $a > 0$.

PROOF;

Let $0 \neq x \in X$ such that $\mu_F(x) > 0$ then there exists

$\{\lambda_1, \lambda_2, \dots, \lambda_n\} \subseteq \mathbb{R} \setminus \{0\}$ and $\{b_1, b_2, \dots, b_n\} \subset X$ with $\mu_B(b_i) > 0$ such that $x = \sum_{i=1}^n \lambda_i b_i$ and $\lambda_j > 0$ for $b_j \notin V_X$.

Take $C_i = ab_i$. If $b_j \notin V_X$ then $ab_j = C_j \notin V_X$. Also $\mu_{aB}(C_i) = \mu_B(\frac{1}{a} C_i) = \mu_B(b_i) > 0$. Thus $x = \sum_{i=1}^n \lambda_i b_i = \sum_{i=1}^n \lambda_i (\frac{1}{a} C_i) = \sum_{i=1}^n \frac{\lambda_i}{a} C_i = \sum_{i=1}^n \eta_i C_i$ where $\eta_i = \lambda_i/a$. Hence, there

exists $\{\eta_1, \dots, \eta_n\} \subseteq \mathbb{R} \setminus \{0\}$ and $\{C_1, C_2, \dots, C_n\} \subset X$ with

$\mu_{aB}(C_i) > 0$ and

$x = \sum_{i=1}^n \eta_i C_i$ and $\eta_j > 0$ for $C_j \notin V_X$.

2.4.3 PROPOSITION;

Let B be a fuzzy set and F be a fuzzy almost linear subspace of X . If B is a fuzzy basis of F and $\mu_F(x) \leq \min\{\mu_B(b_i)\}_{i=1}^n$ where b_i are elements which satisfy $x = \sum_{i=1}^n \lambda_i b_i$. Then B_d is a basis for F_d where $0 < d \leq \mu_F(0)$.

PROOF;

If $0 < d \leq \mu_F(0)$ then F_d is an almost linear subspace of X . Suppose $x \in F_d$ then $\mu_F(x) \geq d$. Then there exist $\{\lambda_1, \dots, \lambda_n\}$ and $\{b_1, \dots, b_n\}$ with $\mu_B(b_i) > 0$ such that $x = \sum_{i=1}^n \lambda_i b_i$ where $\lambda_j > 0$ for $b_j \notin V_x$. As $\mu_F(x) \leq \min\{\mu_B(b_i)\}_{i=1}^n$, therefore $\mu_B(b_i) \geq d$. Hence $b_i \in B_d$. Thus B_d is a basis for F_d .

Let F be a fuzzy almost linear subspace of X . Define $V_{F_0} = V_x \wedge F_0 = \{x \in F_0 : x-x = 0\}$. Then V_{F_0} is an almost linear subspace of F_0 .

2.4.4 LEMMA;

If a fuzzy set B of X is a fuzzy basis for a fuzzy almost linear subspace F of X then $B \cap V_{F_0}$ is a fuzzy basis for V_{F_0} .

PROOF;

If B is a fuzzy basis for F then B_o is a basis for F_o . Hence by Lemma 1.3.4 $B_o \cap V_{F_o}$ is a basis for V_{F_o} . Thus $B \cap V_{F_o}$ is a fuzzy basis for V_{F_o} (considered as a fuzzy almost linear subspace).

2.4.5 LEMMA;

Let B be a fuzzy set of X and F be a fuzzy almost linear subspace of X . Then B is a fuzzy basis for F if and only if $B_o \cap V_{F_o}$ is a basis for V_{F_o} and for each $x \notin V_{F_o}$ and $\mu_F(x) > 0$, there exist $b_1, \dots, b_n \in B_o \setminus V_{F_o}$, $\lambda_i > 0$, $v \in V_{F_o}$ and $\lambda_1, \dots, \lambda_n$ such that $x = \sum_{i=1}^n \lambda_i b_i + v$.

PROOF;

The fuzzy set B is a fuzzy basis for F if and only if B_o is a basis for F_o . Thus by Lemma 1.3.4 the Lemma follows.

2.4.6 LEMMA;

Let B be a fuzzy basis of F . Then for each $b \notin V_{F_o}$, there exists unique $\psi(b) \notin V_{F_o}$, $v(b) \in V_{F_o}$ and $\lambda(b) > 0$

such that $-b = \lambda(b) \psi(b) + v(b)$.

PROOF;

This is a consequence of Lemma 2.4.5.

2.4.7 THEOREM;

Let B be a fuzzy basis of F , then there exists a fuzzy basis B' of F such that for each $b' \in B'_0 \setminus V_{F_0}$, $-b' \in B'_0 \setminus V_{F_0}$.

PROOF;

If B is a fuzzy basis of F then B_0 is a basis of F_0 . By Theorem 1.3.8 there exists a basis B'_0 of X with the property that for each $b' \in B'_0 \setminus V_{F_0}$, we have $-b' \in B'_0 \setminus V_{F_0}$. Hence the fuzzy set B' is a basis of F such that for each $b' \in B'_0 \setminus V_{F_0}$ we have $-b' \in b'_0 \setminus V_{F_0}$.