

ON FUZZY LINEAR SPACES OVER VALUED FIELDS

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Abstract : In this paper, we establish some concepts of fuzzy linear spaces over valued fields. We will also establish convex sets in fuzzy linear spaces over any valued fields.

Definition 1.1:

Let X be a field, and F a fuzzy set in X with membership function μ_F . Suppose the following conditions hold:

- (i) $\mu_F(x + y) \geq \min\{\mu_F(x), \mu_F(y)\}, \quad x, y \in X;$
- (ii) $\mu_F(-x) \geq \mu_F(x), \quad x \in X;$
- (iii) $\mu_F(xy) \geq \min\{\mu_F(x), \mu_F(y)\}, \quad x, y \in X;$
- (iv) $\mu_F(x^{-1}) \geq \mu_F(x), \quad x(\neq 0) \in X.$

Then we call F a fuzzy field in X and denote it by (F, X) . Also, (F, X) is called a fuzzy field of X .

Definition 1.2

Let X be a field, and (F, X) be a fuzzy field of X . Let Y be a linear space over X and V a fuzzy set of Y with membership function μ_V . Suppose the following conditions hold:

- (i) $\mu_V(x + y) \geq \min\{\mu_V(x), \mu_V(y)\}, \quad x, y \in Y;$
- (ii) $\mu_V(-x) \geq \mu_V(x), \quad x \in Y;$
- (iii) $\mu_V(\lambda x) \geq \min\{\mu_F(\lambda), \mu_V(x)\}, \quad \lambda \in X \text{ and } x \in Y;$
- (iv) $\mu_V(1) \geq \mu_V(0).$

Then (V, Y) is called a fuzzy linear space over (F, X) .

Theorem 1.3 : If (V, Y) is a fuzzy linear space over (F, X) , then

- (i) $\mu_F(0) \geq \mu_V(0);$
- (ii) $\mu_V(0) \geq \mu_V(x), \quad x \in Y;$
- (iii) $\mu_F(0) \geq \mu_V(x), \quad x \in Y.$

Theorem 1.4 : Let (F, X) be a fuzzy field of X , and Y a linear space over X . Let V be a fuzzy set of Y . Then (V, Y) is a fuzzy linear space over (F, X) iff

- (i) $\mu_V(\lambda x + \mu y) \geq \min\{\mu_F(\lambda) \wedge \mu_V(x), \mu_F(\mu) \wedge \mu_V(y)\}, \quad \lambda, \mu \in X \text{ and } x, y \in Y;$
- (ii) $\mu_F(1) \geq \mu_V(x), \quad x \in Y.$

Proof : If (V, Y) is a fuzzy linear space over (F, X) , then for $\lambda, \mu \in X$ and $x, y \in Y$,

$$\begin{aligned} \mu_V(\lambda x + \mu y) &\geq \min\{\mu_V(\lambda x), \mu_V(\mu y)\}, \\ &\geq \min\{\mu_F(\lambda) \wedge \mu_V(x), \mu_F(\mu) \wedge \mu_V(y)\} \end{aligned}$$

By Theorem 1.3 (ii) and Definition 1.2 (iv),

$$\mu_F(1) \geq \mu_V(x) \quad \text{for all } x \in Y.$$

On the other hand, if the inequality of the Theorem 1.4 hold for all $x, y \in Y$, then

$$\begin{aligned} \mu_V(x + y) &= \mu_V(1x + 1y) \\ &\geq \min\{\mu_F(1) \wedge \mu_V(x), \mu_F(1) \wedge \mu_V(y)\}, \quad \lambda, \mu \in Y \\ &= \min\{\mu_V(x), \mu_V(y)\} \end{aligned}$$

So we have (i). Further, since (F, X) is a fuzzy field of X ,

$$\mu_F(0) \geq \mu_F(1) \geq \mu_V(x) \quad \text{and} \quad \mu_F(-1) \geq \mu_F(1) \geq \mu_V(x), \quad x \in Y.$$

Hence,

$$\begin{aligned} \mu_V(-x) &= \mu_V(0x + (-1)x) \\ &\geq \min\{\mu_F(0) \wedge \mu_V(x), \mu_F(-1) \wedge \mu_V(x)\} \\ &= \min\{\mu_V(x), \mu_V(x)\} \\ &= \mu_V(x), \end{aligned}$$

i.e. condition (ii) of Definition 1.2 holds.

For all $\lambda \in X$ and $x \in Y$,

$$\begin{aligned} \mu_V(\lambda x) &= \mu_V(0x + \lambda x) \\ &\geq \min\{\mu_F(0) \wedge \mu_V(x), \mu_F(\lambda) \wedge \mu_V(x)\} \\ &= \min\{\mu_V(x), \mu_F(\lambda) \wedge \mu_V(x)\} \\ &= \min\{\mu_F(\lambda), \mu_V(x)\} \end{aligned}$$

Hence condition (iii) of the definition holds.

That condition (iv) of Definition 1.2 is obvious.

Hence, (V, Y) is a fuzzy linear space over (F, X) .

This completes the proof.

We state two propositions without proof which are useful in over further study.

Theorem 1.5 : Let Y and Z be linear spaces over the field X , and f a linear transformation of Y into Z . Let (F, X) be a fuzzy field of X , and (W, Z) be a fuzzy linear space over (F, X) . Then $(f^{-1}(W), Y)$ is a fuzzy linear space over (F, X) .

Theorem 1.6 : Let Y and Z be linear spaces over the field X , and f a linear transformation of Y into Z . Let (F, X) be a fuzzy field of X and (V, Y) be a fuzzy linear space over (F, X) . Then $(f(V), Z)$ is a fuzzy linear space over (F, X) .

Proof : For all $\lambda, \mu \in X$ and $u, v \in Z$ if either $f^{-1}(u)$ or $f^{-1}(v)$ is empty, then the inequality (i) of

Theorem 1.4 is satisfied. Suppose neither $f^{-1}(u)$ nor $f^{-1}(v)$ is empty, then $f^{-1}(\lambda u + \mu v) \neq \emptyset$

. Let $r \in f^{-1}(u), s \in f^{-1}(v)$. Then $f(\lambda r + \mu s) = \lambda f(r) + \mu f(s) = \lambda u + \mu v$.

So

$$\begin{aligned} \mu_{f(v)}(\lambda u + \mu v) &= \sup_{w \in f^{-1}(\lambda u + \mu v)} \mu_v(w) \\ &\geq \sup_{r \in f^{-1}(u), s \in f^{-1}(v)} \mu_v(\lambda r + \mu s) \\ &\geq \sup_{r \in f^{-1}(u), s \in f^{-1}(v)} \{ \min \mu_F(\lambda) \wedge \mu_v(r), \mu_F(\mu) \wedge \mu_v(s) \} \\ &= \min \left\{ \min \left\{ \mu_F(\lambda), \sup_{r \in f^{-1}(u)} \mu_v(r) \right\}, \min \left\{ \mu_F(\mu), \sup_{s \in f^{-1}(v)} \mu_v(s) \right\} \right\} \\ &= \min \{ \min \{ \mu_F(\lambda), \mu_{f(v)}(u) \}, \min \{ \mu_F(\mu), \mu_{f(v)}(v) \} \} \\ &= \min \{ \mu_F(\lambda) \wedge \mu_{f(v)}(u), \mu_F(\mu) \wedge \mu_{f(v)}(v) \} \end{aligned}$$

Obviously, for any $x \in Z, \mu_F(1) \geq \mu_{f(v)}(x)$.

Thus $(f(V), Z)$ is a fuzzy linear space over (F, X) , which ends the proof.

Let X be any set and L a lattice (in particular L could be $[0, 1]$). Then a fuzzy set A in X is characterised by a mapping $A : X \rightarrow L$. We now quote the following definitions which will be needed in the sequel.

Definition 1.7

Let E be a field and Y a linear space over E . Let X be a fuzzy subset of Y . Then X is a fuzzy linear space (fls in short) in Y if for all $x, y \in Y$ and $\lambda \in E$,

- (i) $X(x+y) \geq \min \{ X(x), X(y) \}$.
- (ii) $X(\lambda x) \geq X(x)$.

Definition 1.8

Let E be a field. The map $|\cdot| : E \rightarrow R$ (where R denotes the real numbers) is called a valuation on E if or all $\alpha, \beta \in E$,

- (i) $|\alpha| \geq 0$ and $= 0$ iff $\alpha = 0$,
- (ii) $|\alpha\beta| = |\alpha| |\beta|$,
- (iii) $|\alpha + \beta| = |\alpha| + |\beta|$.

A field E along with a valuation is called a valued field. The valuation is said to be non-Archimedean if (iii) is replaced by

$$(iii)' \quad |\alpha + \beta| \leq \max (|\alpha|, |\beta|),$$

Otherwise it is called Archimedean. The set $V = \{ \alpha \in E : |\alpha| \leq 1 \}$ is a subring of E and is called the valuation ring of E or the ring of integers of E if the valuation is non-Archimedean.

Definition 1.9

Let Y be a linear space over a non-Archimedean

valued field E and let X be a fuzzy linear space in Y . Let S be a fuzzy set in X . Then S is said to be balanced if

$$S(kx) = S(x) \quad \text{for all } k \in V \text{ and } x \in X.$$

S is said to be convex if

$$S(k_1x + k_2y) \geq \min \{ S(x), S(y) \} \quad \text{for all } k_1, k_2 \in V \text{ and } x, y \in X.$$

Theorem 1.10 : Let G_1, G_2 be linear spaces over E and $f : G_1 \rightarrow G_2$ a linear map. Let F_1 and F_2 be fuzzy linear spaces in G_1 and G_2 , respectively. If A is a convex (balanced) fuzzy set in F_1 , then $f(A)$ is a convex (balanced) fuzzy set in F_2 . Similarly if B is a convex (balanced) fuzzy set in F_2 , then $f^{-1}(B)$ is a convex (balanced) fuzzy set in F_1 .

Proof : We shall prove the result only for the convex case. Let $k_1, k_2 \in V$, the valuation ring of E . Let A be a convex fuzzy set in F_1 . Then $k_1f(A) + k_2f(A) = f(k_1A + k_2A) \subset f(A)$ which implies that $f(A)$ is convex.

Let B be a convex fuzzy set in F_2 and let $k_1, k_2 \in V$. Set

$$M = k_1f^{-1}(B) + k_2f^{-1}(B).$$

Then

$$f(M) = k_1f(f^{-1}(B)) + k_2f(f^{-1}(B)) \subset k_1B + k_2B \subset B.$$

Hence $M \subset f^{-1}(B)$ and this completes the proof.

Theorem (Lowen) 1.11 : Let L be a complete lattice and let $\{A_i\}$ be a family of convex (balanced) fuzzy sets in X . Then $A = \bigcap_{i \in I} A_i$ is a convex (balanced) fuzzy set in X .

Proof : For $k_1, k_2 \in V$,

$$\begin{aligned} A(k_1x + k_2y) &= \inf_{i \in I} A_i(k_1x + k_2y) \\ &\geq \inf_{i \in I} [\min \{ A_i(x), A_i(y) \}] \quad \dots(\text{from Definition 1.3}) \\ &= \min [\inf_{i \in I} A_i(x), \inf_{i \in I} A_i(y)] \\ &= \min [A(x), A(y)]. \end{aligned}$$

For $k \in V$,

$$\{x \in X : kA(x) \geq d\} \subset \{x \in X : A(x) \geq d\}.$$

This completes the proof.

Definition 1.12

Let A be a fuzzy set in X . The convex (balanced) hull of A is the intersection of all convex (balanced) fuzzy sets in X which contain A .

It follows from Theorem 1.11 that the convex (balanced) hull of A is the smallest convex (balanced) fuzzy set in X which contains A .

Theorem 1.13 : Let A be a fuzzy set in X . Then the balanced hull of A is the fuzzy set $\bigcup_{\lambda \in V} \lambda A$.

Proof : It is easy to see that the fuzzy set

$B = \bigcup_{\lambda \in V} \lambda A$ is included in any balanced fuzzy set which contains A. Since $B \supset A$, it suffices to show that B is balanced. Let $a \in V$ and $x \in X$. Then

$$B(x) = \sup_{\lambda \in V} \lambda A(x) \leq \sup_{\lambda \in V} a \lambda A(x)$$

$$= \sup_{\lambda \in V} \lambda A(ax) = B(ax).$$

Hence, a $B \subset B$ and this completes the proof.

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