

STATISTICAL LINEAR SPACES

Part I. Properties of ε, η -topology

JIŘÍ MICHÁLEK

The definition of the statistical linear space in the Menger sense (*SLM*-space) is given in this paper. The ε, η -topology is introduced and the basic properties of *SLM*-spaces as linear topological spaces are investigated.

0. INTRODUCTION AND PRELIMINARIES

In this paper we shall deal with basic properties of statistical linear spaces in the Menger sense (*SLM*-space) which are a special case of statistical metric spaces in the Menger sense (*SMM*-space). *SMM*-spaces are a generalization of the usual notion of metric spaces in that sense that a metric is replaced by a collection of probability distribution functions. Similarly, *SLM*-spaces are a generalization of linear normed spaces where a norm is substituted by a suitable family of probability distribution functions.

This paper contains in Section 1 the definition of *SLM*-spaces and the main properties of them together with three examples.

The definition of the ε, η -topology and basic properties of *SLM*-spaces as linear topological spaces are in Section 2. Section 3 contains some properties of ε, η -neighbourhoods from a base for the ε, η -topology. In Section 4 properties of the mapping \mathcal{J} , which is defined on an *SLM*-space and takes its values in the Lévy space of probability distribution functions, are studied.

The notation of an *SMM*-space is studied in many details in [1]. A detailing discussion of the original Menger definition of the generalized triangular inequality is made there. Under these conclusions the authors suggested the following definition of an *SMM*-space.

Definition 1. By a statistical metric space in the sense of Menger we shall call a triple (S, \mathcal{K}, T) where S is a nonempty set, \mathcal{K} is a mapping $\mathcal{K} : S \times S \rightarrow \mathcal{F}$,

where \mathcal{F} is the set of all one-dimensional probability distribution functions, satisfying $\mathcal{K}(x, y) = F_{xy}(\cdot)$

1. $(F_{xy}(u) = 1 \text{ for } u > 0) \Leftrightarrow x = y$
2. $F_{xy}(0) = 0$ for every pair $x, y \in S$
3. $F_{xy}(u) = F_{yx}(u)$ for every $u \in \mathbb{R}$ and every pair $x, y \in S$ (\mathbb{R} is the set of reals)
4. $F_{xz}(u + v) \geq T(F_{xy}(u), F_{yz}(v))$ for every $x, y, z \in S$ and every $u, v \in \mathbb{R}$ where T is a t -norm defined on $\langle 0, 1 \rangle \times \langle 0, 1 \rangle$ with values in $\langle 0, 1 \rangle$ and satisfying properties:
 - (a) $T(a, b) = T(b, a); T(a, 1) = a$ for $a > 0$
 - (b) $T(a, b) \leq T(c, d)$ for $a \leq c, b \leq d$
 $T(T(a, b), c) = T(a, T(b, c))$
 - (d) $T(0, 0) = 0$.

Definition 1 yields immediately that every t -norm T satisfies $T(a, b) \leq \min(a, b)$ where \min is a t -norm too. Further important examples of t -norms are $T(a, b) = ab$, $T(a, b) = \max(a + b - 1, 0)$. It is worth quoting [10] where one can see a close relation between t -norms and 2-dimensional copulas.

Further, in [1] the ε, η -topology is introduced by the neighbourhoods of the form

$$N_x(\varepsilon, \eta) = \{y \in S : F_{xy}(\eta) > 1 - \varepsilon\}, \quad x \in S, \quad \eta > 0, \quad 0 < \varepsilon \leq 1$$

and under the continuity of the t -norm T it is proved that these neighbourhoods form a base for a Hausdorff topology in S . This topology is called the ε, η -topology. The paper [2] studies the question under which conditions the ε, η -topology is metrizable. If $\sup_{a < 1} T(a, a) = 1$ then the system $\mathcal{N} = \{U(\varepsilon, \eta)\}$ where $U(\varepsilon, \eta) = \{(x, y) \in S \times S : F_{xy}(\eta) > 1 - \varepsilon\}$ ($\eta > 0, \varepsilon \in (0, 1)$) is a base of a Hausdorff uniformity in $S \times S$.

The mapping $\mathcal{K} : S \times S \rightarrow \mathcal{F}$ where \mathcal{F} is the Lévy space of probability distribution functions is studied in [3]. If $\lim_{v \uparrow 1} T(a, v) = a$ uniformly in $\langle 0, 1 \rangle$, then \mathcal{K} is uniformly continuous with respect to the ε, η -topology in $S \times S$.

The problem of a completion of *SMM*-spaces is solved in [4]. It is proved (under certain conditions on the t -norm T) that every *SMM*-space can be (up to an isomorphism) completed by the maintenance of the t -norm in the unique way.

In [5] it is suggested one of the possible generalizations of the triangular inequality. The demand 4 in Definition 1 is replaced by 4': $(F_{xy}(u) = 1 \text{ and } F_{yz}(v) = 1) \Rightarrow F_{xz}(u + v) = 1$, which is of course weaker than 4 in Definition 1. Further, in this paper a relation between the mapping \mathcal{K} (mentioned above) and a certain class of semimetrics on S is studied and it is proved, in the case of the t -norm $T = \min(a, b)$ the existence of a probability space (D, \mathcal{B}, μ) where D contains some semimetrics on S , all sets of the form $\{d \in D : d(x, y) > u\}$, $x, y \in S, u \in \mathbb{R}$ belong to \mathcal{B} and

$$\mu\{d \in D : d(x, y) > u\} = F_{xy}(u).$$

At the beginning the theory of *SMM*-spaces belonged rather to the functional analysis than to the probability theory; e.g. many articles are devoted to problems of fixed points of mappings defined on *SMM*-spaces. Recently, some papers occurred where the connection with the probability theory is quite evident, see, e.g. [7], [8], [9].

1. DEFINITION OF *SML*-SPACE, BASIC PROPERTIES, EXAMPLES

In this paper a special case of statistical metric spaces is considered. The definition of *SMM*-spaces is based on that fact that although the distance of two points is a fixed nonnegative number, an observer can measure this distance with certain errors. His measurements are affected by errors and from this point of view a distance is a random variable with its distribution function. Similarly, we can consider the case of a normed linear space, where a norm is the distance measured from the zero element. Properties of a norm and Definition 1 of the *SMM*-space lead us to the following definition of the linear statistical space.

Definition 2. Let S be a real linear space, let \mathcal{F} be the set of all probability distribution functions defined on the real line \mathbb{R} . Let $\mathcal{J} : S \rightarrow \mathcal{F}$ be a given mapping. For every $x \in S$ let us denote $\mathcal{J}(x) = F_x \in \mathcal{F}$ and we demand that \mathcal{J} satisfies:

1. $x = 0 \Leftrightarrow F_x = H$ where $H(u) = 0 \ u \leq 0; H(u) = 1 \ u > 0$
2. $F_{\lambda x}(u) = F_x(u/|\lambda|)$ for every $x \in S$ and every $\lambda \neq 0$.
3. $F_x(u) = 0$ for every $u \leq 0$ and every $x \in S$.
4. $T(F_x(u), F_y(v)) \leq F_{x+y}(u+v)$ for every $u, v \in \mathbb{R}$ and every pair $x, y \in S$ where T is a t -norm satisfying (a), (b), (c), (d) in Definition 1.

Under these conditions the triple (S, \mathcal{J}, T) is called a linear statistical space in the Menger sense (*SLM*-space).

Example 1. Let $S = \mathbb{R}$, let G be a distribution function with $G(0) = 0$ and $G \neq H$. If $x \in S$ let us define

$$\mathcal{J}(x) = F_x(\cdot) = G\left(\frac{\cdot}{|x|}\right) \quad \text{for } x \neq 0$$

$$\mathcal{J}(0) = H(\cdot) \quad \text{and} \quad T(a, b) = \min(a, b);$$

then $(\mathbb{R}, \mathcal{J}, \min)$ is an *SLM*-space. As we assume $G \neq H$ then $x = 0$ if and only if $F_x = H$. Further, $F_x(0) = 0$ for every $x \in \mathbb{R}$ thanks to the assumption $G(0) = 0$. Thus, we have

$$F_{\lambda x}(u) = G\left(\frac{u}{|\lambda x|}\right) = G\left(\frac{u}{|\lambda| |x|}\right) = G\left(\frac{u}{|\lambda|} \frac{1}{|x|}\right) = F_x\left(\frac{u}{|\lambda|}\right)$$

for every $\lambda \in \mathbb{R}, \lambda \neq 0$ and every $x \in \mathbb{R}$. The main problem is to prove the triangular

inequality in the form

$$F_{x+y}(u+v) \geq \min(F_x(u), F_y(v)), \quad \text{i.e.}$$

$$(*) \quad G\left(\frac{u+v}{|x+y|}\right) \geq \min\left(G\left(\frac{u}{|x|}\right), G\left(\frac{v}{|y|}\right)\right).$$

If $u \leq 0$ or $v \leq 0$ then the inequality (*) is true because $G(0) = 0$. In the case $u > 0$ and $v > 0$, $x = 0$ or $y = 0$ or $x + y = 0$ the generalized triangular inequality is trivial. As the function G is nondecreasing, the inequality (*) for $u > 0$, $v > 0$, $|x + y| > 0$, $|x| > 0$, $|y| > 0$ follows from the inequality

$$\frac{u+v}{|x+y|} \geq \min\left(\frac{u}{|x|}, \frac{v}{|y|}\right).$$

Indeed, let us assume $u > 0$, $v > 0$, $|x + y| > 0$ and $(u+v)/|x+y| > \min(u/|x|, v/|y|)$. It implies that simultaneously $(u+v)/|x+y| > u/|x|$ and $(u+v)/|x+y| > v/|y|$, thus $(u+v)|x| > u|x+y|$ and $(u+v)|y| > v|x+y|$, what gives $|x| + |y| > |x+y|$ and that is a contradiction. This completes the proof of that fact that $(\mathbb{R}, \mathcal{F}, \min)$ is an *SLM*-space.

Example 2. Let S be the set of all real sequences, i.e. $S = \{x : x = (x_1, x_2, x_3, \dots, x_n, \dots)\}$, where the operations of addition and scalar multiplication are defined coordinatewisely. Let $\{a_n\}_{n=1}^{\infty}$ be a sequence of positive numbers such that $\sum_{n=1}^{\infty} a_n = 1$. Let us define the mapping $\mathcal{F} : S \rightarrow \mathcal{F}$ in the following way:

if $x = (x_1, x_2, x_3, \dots, x_n, \dots)$ then we put

$$\begin{aligned} F_x(u) &= 0 & \text{for } u &\leq |x_1| \\ F_x(u) &= a_1 & \text{for } |x_1| < u \leq |x_1| + |x_2| \\ F_x(u) &= a_1 + a_2 & \text{for } |x_1| + |x_2| < u \leq |x_1| + |x_2| + |x_3| \\ &\vdots & \vdots \\ F_x(u) &= \sum_{i=1}^n a_i & \text{for } \sum_{i=1}^n |x_i| < u \leq \sum_{i=1}^{n+1} |x_i| \\ &\vdots & \vdots \end{aligned}$$

In the case if $\sum_{i=1}^{\infty} |x_i| < \infty$ we must consider two possibilities:

- $\sum_{i=1}^{\infty} |x_i|$ contains infinitely many non-zero elements, then $F_x(u) = 1$ for $u \geq \sum_{i=1}^{\infty} |x_i|$
- $\sum_{i=1}^{\infty} |x_i|$ contains finitely many non-zero elements only, then $F_x(u) = 1$ for $u > \sum_{i=1}^{\infty} |x_i|$.

We do not eliminate the case of an empty interval.

As a t -norm we choose again the function $\min(a, b)$. Then the triple (S, \mathcal{F}, \min) is an *SLM*-space. Surely, $F_x = H$ if and only if $x = 0$ because for every $x \neq 0$ at

least one coordinate x_i differs from zero. Further, $F_{\lambda x}(u) = F_x(u/|\lambda|)$ for every $x \in S$, $\lambda \neq 0$, $u \in \mathbb{R}$ because if $\lambda \neq 0$, $u > 0$, $x = 0$ then $\lambda x = 0$ and $F_{\lambda x}(u) = 1$. If $u \leq 0$ then for every $x \in S$ it is $F_x(u) = 0$ hence $F_{\lambda x}(u) = 0$ also for every $\lambda \in \mathbb{R}$. Now, in the last case $\lambda \neq 0$, $u > 0$, $x \neq 0$ we have

$$F_x\left(\frac{u}{|\lambda|}\right) = \sum_{i=1}^n a_i \quad \text{if and only if} \quad \sum_{i=1}^n |x_i| < \frac{u}{|\lambda|} \leq \sum_{i=1}^{n+1} |x_i|,$$

what is

$$\sum_{i=1}^n |\lambda x_i| < u \leq \sum_{i=1}^{n+1} |\lambda x_i|.$$

The previous inequality expresses the value of $F_{\lambda x}$ at the point u , i.e.

$$F_{\lambda x}(u) = \sum_{i=1}^n a_i \quad \text{if and only if} \quad \sum_{i=1}^n |\lambda x_i| < u \leq \sum_{i=1}^{n+1} |\lambda x_i|.$$

At the end we must verify the generalized triangular inequality with the t -norm min.

If $u + v \in (\sum_{i=1}^n |x_i + y_i|, \sum_{i=1}^{n+1} |x_i + y_i|)$ then either $u \leq \sum_{i=1}^{n+1} |x_i|$ or $v \leq \sum_{i=1}^{n+1} |y_i|$, hence either $F_x(u) \leq \sum_{i=1}^n a_i$ or $F_y(v) \leq \sum_{i=1}^n a_i$, but in every case the inequality $\min(F_x(u), F_y(v)) \leq F_{x+y}(u+v)$ holds. The case $F_x(u) = 1$ is investigated in a similar way.

Example 3. Let (Ω, \mathcal{A}, P) be a probability space. Two random variables ξ, η on Ω with $P\{\omega : \xi(\omega) = \eta(\omega)\} = 1$ shall belong to the same class of equivalence. Let S denote these classes of equivalence on Ω . Evidently, S is a linear space. Let us define a mapping \mathcal{F} in the following way:

$$\mathcal{F}(\xi) [u] = P\{\omega : |\xi(\omega)| < u\} = F_\xi(u), \quad \xi \in S, \quad u \in \mathbb{R},$$

As a t -norm we choose $m(a, b) = \max(a + b + b - 1, 0)$. Then the triple (S, \mathcal{F}, m) is an *SLM*-space.

It is clear that for every $\lambda \neq 0$ and $\xi \in S$ it holds

$$P\{\omega : |\lambda \xi(\omega)| < u\} = P\left\{\omega : |\xi(\omega)| < \frac{u}{|\lambda|}\right\}$$

and hence $F_{\lambda \xi}(u) = F_\xi(u/|\lambda|)$. Similarly, $P\{\omega : |\xi(\omega)| < u\} = 0$ for $u \leq 0$ gives $F_\xi(u) = 0$ for every $u \leq 0$. Surely, $F_\xi(u) = H(u)$ for every $u \in \mathbb{R}$ if and only if $\xi = 0$. The validity of the generalized triangular inequality is based on the results in [10]. It holds that the joint distribution function $G_{\xi, \eta}(\cdot, \cdot)$ of $\xi, \eta \in S$ can be expressed as a function of their marginal distribution functions $g_\xi(\cdot), g_\eta(\cdot)$ $G_{\xi, \eta}(u, v) = C(g_\xi(u), g_\eta(v))$ where C is a 2-dimensional copula generally depending on a couple ξ, η . This copula C is a function defined on $\langle 0, 1 \rangle \times \langle 0, 1 \rangle$ satisfying the following inequality

$$\min(a, b) \geq C(a, b) \geq m(a, b).$$

The inclusions $\{\omega : |\xi(\omega) + \eta(\omega)| < u + v\} \supset \{\omega : |\xi(\omega)| + |\eta(\omega)| < u + v\} \supset \{\omega : |\xi(\omega)| < u, |\eta(\omega)| < v\}$ give

$$\begin{aligned} F_{\xi+\eta}(u+v) &= P\{\omega : |\xi(\omega) + \eta(\omega)| < u + v\} \geq \\ &\geq P\{\omega : |\xi(\omega)| < u, |\eta(\omega)| < v\} = C(F_{\xi}(u), F_{\eta}(v)) \geq m(F_{\xi}(u), F_{\eta}(v)). \end{aligned}$$

It proves the validity of the generalized triangular inequality with the t -norm m .

Theorem 1. Every SLM -space is an SMM -space with the same t -norm.

Proof. Let (S, \mathcal{F}, T) be an SLM -space. Let us define the mapping $\mathcal{K}(x, y) = \mathcal{F}(x - y)$, $\mathcal{K} : S \times S \rightarrow \mathcal{F}$. Then the triple (S, \mathcal{K}, T) is an SMM -space. $\mathcal{F}(x) = H$ if and only if $x = 0$. The mapping \mathcal{K} is surely symmetric, because $\mathcal{F}(x - y) = \mathcal{F}(y - x)$. If we denote $\mathcal{K}(x, y) = F_{xy}$, $\mathcal{F}(x) = F_x$, then the generalized triangular inequality holds, because

$$T(F_{xy}(u), F_{yz}(v)) = T(F_{x-y}(u), F_{y-z}(v)) \leq F_{x-z}(u+v) = F_{xz}(u+v). \quad \square$$

Remark. Let S be an n -dimensional real linear space. Then the triple (S, \mathcal{F}, T) is an SLM -space if and only if to every n -tuple of real numbers $(\lambda_1, \lambda_2, \dots, \lambda_n)$ a probability distribution function $F_{(\lambda_1, \lambda_2, \dots, \lambda_n)}$ corresponds such that

1. $F_{(\lambda_1, \lambda_2, \dots, \lambda_n)} = H$ if and only if $\lambda_1 = \lambda_2 = \dots = \lambda_n = 0$
2. $F_{(\mu\lambda_1, \mu\lambda_2, \dots, \mu\lambda_n)}(u) = F_{(\lambda_1, \lambda_2, \dots, \lambda_n)}(u/|\mu|)$ for every $\mu \neq 0$, $u \in \mathbb{R}$ and every n -tuple $(\lambda_1, \lambda_2, \dots, \lambda_n)$
3. $F_{(\lambda_1, \lambda_2, \dots, \lambda_n)}(0) = 0$ for every n -tuple $(\lambda_1, \lambda_2, \dots, \lambda_n)$
4. $T(F_{(\lambda_1, \lambda_2, \dots, \lambda_n)}(u), F_{(\mu_1, \mu_2, \dots, \mu_n)}(v)) \leq F_{(\lambda_1 + \mu_1, \lambda_2 + \mu_2, \dots, \lambda_n + \mu_n)}(u+v)$ for every n -tuple $(\lambda_1, \lambda_2, \dots, \lambda_n)$ and $(\mu_1, \mu_2, \dots, \mu_n)$ and every $u, v \in \mathbb{R}$ (T is a t -norm).

2. TOPOLOGY IN SLM -SPACES

We shall use usual notions in the topology and in the theory of linear topological spaces; see, e.g. [11]. Only the notions important for us shall be defined explicitly.

Definition 4. Let (S, \mathcal{F}, T) be a statistical linear space in the sense of Menger, let $x \in S$, $0 < \varepsilon \leq 1$, $\eta > 0$. Then the subset of S

$$O(x, \varepsilon, \eta) = \{z \in S : F_{x-z}(\eta) > 1 - \varepsilon\}$$

is called the ε, η -neighbourhood of the point x .

As the space S is linear, it is sufficient to introduce neighbourhoods of the zero element only, i.e. the neighbourhoods of the form $O(\varepsilon, \eta) = \{z : F_z(\eta) > 1 - \varepsilon\}$. We shall assume the continuity of the t -norm T on $\langle 0, 1 \rangle \times \langle 0, 1 \rangle$. Under this assumption it is possible to prove that the collection of ε, η -neighbourhoods forms

a base of a topology in the space (S, \mathcal{F}, T) . It is clear that $0 \in O(\varepsilon, \eta)$ for every $0 < \varepsilon \leq 1, \eta > 0$, because $F_0(\eta) = H(\eta) = 1 > 1 - \varepsilon$. Further, if two ε, η -neighbourhoods $O(\varepsilon, \eta), O(\varepsilon', \eta')$ are given, then there exists a neighbourhood $O(\varepsilon^*, \eta^*)$ such that

$$O(\varepsilon^*, \eta^*) \subset O(\varepsilon, \eta) \cap O(\varepsilon', \eta').$$

It is sufficient to put $\varepsilon^* = \min(\varepsilon, \varepsilon'), \eta^* = \min(\eta, \eta')$ because

$$\begin{aligned} O(\varepsilon, \eta) \cap O(\varepsilon', \eta') &= \{z \in S : F_z(\eta) > 1 - \varepsilon, F_z(\eta') > \\ &> 1 - \varepsilon'\} \supset \{z : F_z(\min(\eta, \eta')) > 1 - \min(\varepsilon, \varepsilon')\} = O(\varepsilon^*, \eta^*). \end{aligned}$$

Similarly, if $\varepsilon \leq \varepsilon', \eta \leq \eta'$ then

$$O(\varepsilon, \eta) \subset O(\varepsilon', \eta').$$

The last property which is necessary for a base of neighbourhoods in a topology is that for every ε, η -neighbourhood $O(\varepsilon, \eta)$ and every $y \in O(\varepsilon, \eta)$ there exists such an ε, η -neighbourhood that $O(y, \varepsilon^*, \eta^*) \subset O(\varepsilon, \eta)$. Let $O(\varepsilon, \eta)$ and y be given. As the function F_y being a probability distribution function is left continuous at η , there exist $\eta_0 < \eta, \varepsilon_0 < \varepsilon$ that $F_y(\eta_0) > 1 - \varepsilon_0 > 1 - \varepsilon$. Now, we choose η^* such that $0 < \eta^* < \eta - \eta_0$ and ε^* such that $T(1 - \varepsilon_0, 1 - \varepsilon^*) > 1 - \varepsilon$ (such an ε^* exists because the t -norm T is assumed continuous and $T(a, 1) = a$). Let $s \in O(y, \varepsilon^*, \eta^*)$ then $F_s(\eta) \geq T(F_y(\eta_0), F_{y-\cdot}(\eta - \eta_0)) \geq T(F_y(\eta_0), F_{y-\cdot}(\eta^*)) \geq T(1 - \varepsilon_0, 1 - \varepsilon^*) > 1 - \varepsilon$ and $s \in O(\varepsilon, \eta)$.

Definition 5. The topology generated under the continuity of the t -norm T by the base $\mathcal{Q} = \{O(\varepsilon, \eta) : 0 < \varepsilon \leq 1, \eta > 0\}$ of the neighbourhoods of the zero element in (S, \mathcal{F}, T) will be called the ε, η -topology.

Definition 6. A sequence $\{x_n\}_{n=1}^\infty \subset (S, \mathcal{F}, T)$ will be called F -convergent at $x \in S$, if

$$\lim_{n \rightarrow \infty} F_{x_n - x}(u) = H(u)$$

for every $u \in \mathbb{R}$ (in symbols $x_n \xrightarrow{F} x$).

Lemma 1. A sequence $\{x_n\}_{n=1}^\infty \subset (S, \mathcal{F}, T)$ is F -convergent at $x \in S$ if and only if

$$(\forall \varepsilon \in (0, 1) \forall \eta > 0 \exists n_0 \forall n \geq n_0) \Rightarrow (x_n \in O(\varepsilon, \eta)).$$

Proof. If $\lim_{n \rightarrow \infty} F_{x_n}(u) = H(u), u > 0$, it is $\lim_{n \rightarrow \infty} F_{x_n}(u) = 1$, then

$$(\forall u > 0 \forall \varepsilon \in (0, 1) \exists n_0 \forall n \geq n_0) \Rightarrow F_{x_n}(u) > 1 - \varepsilon \Leftrightarrow x_n \in O(\varepsilon, u).$$

Conversely, if $(\forall \varepsilon \in (0, 1) \forall \eta > 0 \exists n_0 \forall n \geq n_0) \Rightarrow x_n \in O(\varepsilon, \eta) \Leftrightarrow F_{x_n}(\eta) > 1 - \varepsilon$, it is precisely that $\lim_{n \rightarrow \infty} F_{x_n}(\eta) = 1$ for every $\eta > 0$. If $u \leq 0$ we have $F_{x_n}(u) = 0$ for every n . \square

Theorem 2. Every SLM -space (S, \mathcal{F}, T) with a continuous t -norm is with respect

to the ε, η -topology a Hausdorff linear topological space with a countable base of neighbourhoods of the zero element and hence it is metrizable.

Proof. If we choose any sequences $\{\varepsilon_n\}_1^\infty, \{\eta_n\}_1^\infty$ such that $\varepsilon_n \downarrow 0, \eta_n \downarrow 0$ then $\{O(\varepsilon_n, \eta_n)\}_1^\infty$ is a base of neighbourhoods of the origin for the ε, η -topology, because for every $O(\varepsilon, \eta)$ we can find a pair $\varepsilon_{n_0}, \eta_{n_0}$ such that $\varepsilon_{n_0} \leq \varepsilon, \eta_{n_0} \leq \eta$ and hence $O(\varepsilon_{n_0}, \eta_{n_0}) \subset O(\varepsilon, \eta)$.

This space will be a Hausdorff space if and only if $\bigcap_{U \in \mathcal{B}(0)} U = \{0\}$ where $\mathcal{B}(0)$ is a base of neighbourhoods of the origin for the ε, η -topology. In our case it is necessary to prove that $\bigcap_{0 < \varepsilon \leq 1, \eta > 0} O(\varepsilon, \eta) = \{0\}$. Let us suppose that $x \in \bigcap_{\varepsilon, \eta} O(\varepsilon, \eta)$. Then for every $\eta > 0$ and every $\varepsilon \in (0, 1)$ $F_x(\eta) > 1 - \varepsilon$, in other words $F_x(\eta) = 1$ for every $\eta > 0$. It implies that $x = 0$ in S . We have proved that a countable base of the origin for the ε, η -topology exists and hence the ε, η -topology is metrizable.

Using Lemma 1 and the existence of a countable base for the ε, η -topology at the origin we can easily prove that linear operations and the ε, η -topology are consistent. Let $\lambda_n \rightarrow \lambda$ in reals, let $x_n \rightarrow x$ in S in the ε, η -topology. Then $\lambda_n x_n - \lambda x = \lambda_n(x_n - x) + (\lambda_n - \lambda)x$ and the generalized triangular inequality proves immediately continuity of scalar multiplication in the product topology. In a similar way, using the generalized triangular inequality again, one can prove continuity of addition in S in the product topology. \square

Theorem 3. Let (S, \mathcal{F}, T) be a statistical linear space with the t -norm T satisfying $\lim_{a \uparrow 1, b \uparrow 1} T(a, b) = 1$. Then (S, \mathcal{F}, T) with the topology defined by the F -convergence is a linear topological space.

Proof. When $x_n \xrightarrow{F} x$ then evidently for every subsequence $\{x_{n_k}\}_1^\infty \subset \{x_n\}_1^\infty$ $x_{n_k} \xrightarrow{F} x$ also. Further, for every stationary sequence $\{x_n\}_1^\infty$, i.e. $x_n = x$ for every $n \geq n_0$, it holds that $x_n \xrightarrow{F} x$.

If $x_n \xrightarrow{F} x$, i.e. there exists at least one $u_0 > 0$ that $F_{x_n - x}(u_0) \rightarrow 1$, then an $\varepsilon_0 > 0$ and subsequence $\{x_{n_k}\}_1^\infty \subset \{x_n\}$ must exist such that for every subsequence $\{x_k^*\}_1^\infty \subset \{x_{n_k}\}_1^\infty$ $F_{x_k^* - x}(u_0) \leq 1 - \varepsilon_0$, in other words $x_k^* \not\xrightarrow{F} x$.

In this way we have verified all demands put on the topological convergence and we must prove further that this convergence and linear operations defined on S are in accordance. When $x_n \xrightarrow{F} x, y_n \xrightarrow{F} y$ then using the generalized triangular inequality we obtain

$$F_{x_n + y_n}(2\eta) \geq T(F_{x_n}(\eta), F_{y_n}(\eta)) \geq T(1 - \varepsilon, 1 - \varepsilon)$$

for a suitable large n and the left continuity at $[1, 1]$ of the t -norm implies that $T(1 - \varepsilon, 1 - \varepsilon) \rightarrow 1$ if $\varepsilon \rightarrow 0$. Similarly, as it was done in the proof of Theorem 3 we can prove that $x_n \xrightarrow{F} x, \lambda_n \rightarrow \lambda$ imply that $\lambda_n x_n \xrightarrow{F} \lambda x$, too. It follows from the left continuity at $[1, 1]$ of the t -norm T that every F -convergent sequence has a unique limit

point, because

$$F_{x_0-y_0}(2\eta) \cong T(F_{x_n-x_0}(\eta), F_{x_n-y_0}(\eta)) > T(1-\varepsilon, 1-\varepsilon)$$

for a suitable large natural n and every $\eta > 0$. \square

Remark. If the t -norm T is continuous then as we proved in Lemma 1 and Theorem 3, the ε, η -topology and the F -convergence are equivalent. Generally, this equivalence need not hold without the assumption of the continuity of the t -norm T , because ε, η -neighbourhoods need not form a base of neighbourhoods of the origin in S for the topology generated by the F -convergence.

In further considerations we shall deal with continuous t -norms only. In this case every statistical linear space (S, \mathcal{F}, T) has the metrizable ε, η -topology and the question of its normability is interesting for us.

Definition 7. A subset $A \subset S$ where (S, τ) is a linear topological space with a topology τ is called bounded in topology τ if for every τ -neighbourhood U of the origin in S there exists $\lambda > 0$ that

$$A \subset \lambda U.$$

In our case of an SLM -space (S, \mathcal{F}, T) a subset $A \subset S$ is ε, η -bounded if and only if for every $O(\varepsilon, \eta)$ there exists $\lambda(\varepsilon, \eta) > 0$ that

$$A \subset \lambda(\varepsilon, \eta) \cdot O(\varepsilon, \eta) = O(\varepsilon, \lambda(\varepsilon, \eta) \cdot \eta).$$

In other words, the ε, η -boundedness of A can be expressed as follows: a subset A is ε, η -bounded if and only if for every sequence $\{x_n\}_1^\infty \subset A$ and every sequence $\{\lambda_n\}_1^\infty, \lambda_n \rightarrow 0$ of reals $\lambda_n x_n \xrightarrow{F} 0$ also in S .

Now, we use very important criterion of normability of linear topological spaces due to Kolmogorov, see [11]. A Hausdorff linear topological space is normable if and only if there exists a bounded convex neighbourhood of the origin in it. If U is such a neighbourhood then the norm in question can be expressed as

$$\|x\| = \inf \{ \lambda > 0 : x \in \lambda U \}, \quad x \in S.$$

In the case of an SLM -space (S, \mathcal{F}, T) if such a neighbourhood $O(\varepsilon_0, \eta_0)$ exists, then a possible norm $\|\cdot\|$ has the form

$$\begin{aligned} \|x\| &= \inf \{ \lambda > 0 : x \in \lambda O(\varepsilon_0, \eta_0) \} = \\ &= \inf \{ \lambda > 0 : x \in O(\varepsilon_0, \lambda \eta_0) \} = \\ &= \inf \{ \lambda > 0 : F_x(\lambda \eta_0) > 1 - \varepsilon_0 \}. \end{aligned}$$

With this question of normability an important property is connected as the following Theorem 4 states.

In the next Theorem 4 we shall need the following notation:

$\overline{\text{conv}} A$ is the absolutely convex hull of A , $\text{conv} A$ is the convex hull of A .

Theorem 4. Let an SLM -space (S, \mathcal{F}, T) be finite-dimensional. Then the ε, η -topology is normable and is equivalent to the usual Euclidean topology.

Proof. We suppose that the space (S, \mathcal{F}, T) is finite-dimensional and hence every $x \in S$ can be expressed in the form

$$x = \sum_{i=1}^n \lambda_i e_i ;$$

(e_1, e_2, \dots, e_n) is any linear base in S . As the number of the elements in a base is finite, we can find an ε, η -neighbourhood $O(\varepsilon, \eta)$ which contains all elements of the base. Further, every $x \in \overline{\text{conv}}(e_1, e_2, \dots, e_n)$ can be expressed as an absolutely convex combination of e_1, e_2, \dots, e_n , i.e. $x = \sum_{i=1}^n \mu_i e_i, \sum_{i=1}^n |\mu_i| \leq 1$, and because $\text{conv} O(\varepsilon, \eta)$ is also absolutely convex in S then $\overline{\text{conv}}(e_1, e_2, \dots, e_n) \subset \text{conv} O(\varepsilon, \eta)$.

Now, it is necessary to prove that $\text{conv}(e_1, e_2, \dots, e_n)$ is at the same time a neighbourhood of the zero element in the ε, η -topology; for this fact it is sufficient to find $O(\varepsilon^*, \eta^*)$ such that

$$O(\varepsilon^*, \eta^*) \subset \overline{\text{conv}}(e_1, e_2, \dots, e_n).$$

Let us suppose, that such a neighbourhood does not exist, i.e. for every $O(\varepsilon, \eta)$ there exists at least one point $x_0 \in O(\varepsilon, \eta)$ so that $x_0 \notin \overline{\text{conv}}(e_1, e_2, \dots, e_n)$. Taking $\varepsilon_n \downarrow 0, \eta_n \downarrow 0$ we can construct a sequence $\{x_m\}_1^\infty$ which has the zero element as its limit point, let us say $x_m = \sum_{i=1}^n \lambda_i^m e_i$, but $x_m \notin \overline{\text{conv}}(e_1, e_2, \dots, e_n)$, i.e. $\sum_{i=1}^n |\lambda_i^m| > 1$. First, we can suppose that $M \geq \sum_{i=1}^n |\lambda_i^m| > 1$ for all m , where $M < +\infty$. Then there exists a subsequence $\{\lambda_1^{m_k}, \lambda_2^{m_k}, \dots, \lambda_n^{m_k}\}$ that is convergent and hence

$$x_{m_k} = \sum_{i=1}^n \lambda_i^{m_k} e_i \xrightarrow{F} x_0 \quad \text{but} \quad x_0 \neq 0 \quad \text{because}$$

$$x_0 = \sum_{i=1}^n \lambda_i^0 e_i, \quad \lambda_i^0 = \lim_k \lambda_i^{m_k} \quad \text{and} \quad \sum_{i=1}^n |\lambda_i^0| \geq 1.$$

If there exists a subsequence $\sum_{i=1}^n |\lambda_i^{m_k}|$ unbounded from above, i.e.

$$\lim_k \sum_{i=1}^n |\lambda_i^{m_k}| = +\infty,$$

then we can consider the sequence

$$x_{m_k}^* = \sum_{i=1}^n \frac{\lambda_i^{m_k}}{\sum_{j=1}^n |\lambda_j^{m_k}|} e_i = \frac{1}{\sum_{j=1}^n |\lambda_j^{m_k}|} x_{m_k},$$

instead of the original $\{x_m\}_m$. However, at the same time, we have $x_{m_k}^* = \sum_{i=1}^n \mu_i^{m_k} e_i$ with $\sum_{i=1}^n |\mu_i^{m_k}| = 1$ and this case can be transformed to the previous one. This fact proves that $\overline{\text{conv}}(e_1, e_2, \dots, e_n)$ must be a neighbourhood of the zero element in the ε, η -topology. The boundedness of $\overline{\text{conv}}(e_1, e_2, \dots, e_n)$ is clear, because if $\{x_m\}_1^\infty$ is any sequence from $\overline{\text{conv}}(e_1, e_2, \dots, e_n)$, $\lim_m \varrho_m = 0$, $\varrho_m \in \mathbb{R}$ then

$$\varrho_m x_m = \varrho_m \sum_{i=1}^n \lambda_i^m e_i, \quad \sum_{i=1}^n |\lambda_i^m| \leq 1 \quad \text{and}$$

$$F_{\varrho_m x_m}(u) \geq T^{(n)}\left(F_{e_1}\left(\frac{u}{|\varrho_m \lambda_1^m|}\right), \dots, F_{e_n}\left(\frac{u}{|\varrho_m \lambda_n^m|}\right)\right);$$

$$(T^{(n)}(a_1, a_2, \dots, a_n) = T(a_1, T(a_2, \dots, T(a_{n-1}, a_n) \dots)),$$

with $|\lambda_i^m| \leq 1$ and this fact implies that $\varrho_m x_m \xrightarrow{F} 0$. We proved that in the case of a finite dimensional *SLM*-space (S, \mathcal{F}, T) the ε, η -topology is equivalent to the topology generated by the coordinate convergence and the ε, η -topology is normable. \square

Lemma 2. Every *SLM*-space (S, \mathcal{F}, T) where $T(a, b) = \min(a, b)$ is a locally convex linear topological space.

Proof. The proof is very simple. Let us consider any ε, η -neighbourhood $O(\varepsilon, \eta)$ in (S, \mathcal{F}, T) and let $x, y \in O(\varepsilon, \eta)$, $\alpha \in \langle 0, 1 \rangle$, then $F_x(\eta) > 1 - \varepsilon$, $F_y(\eta) > 1 - \varepsilon$ and hence

$$F_{\alpha x + (1-\alpha)y}(\eta) \geq \min(F_{\alpha x}(\alpha\eta), F_{(1-\alpha)y}((1-\alpha)\eta)) = \min(F_x(\eta), F_y(\eta)) > 1 - \varepsilon. \quad \square$$

3. PROPERTIES OF ε, η -NEIGHBOURHOODS

Lemma 3. Let $O(\varepsilon, \eta)$ be an ε, η -neighbourhood of the zero element in an *SLM*-space (S, \mathcal{F}, T) . Then for every $|\lambda| \leq 1$, $\lambda \in \mathbb{R}$ and every $x \in O(\varepsilon, \eta)$

$$\lambda x \in O(\varepsilon, \eta).$$

Proof. Let $x \in O(\varepsilon, \eta)$, i.e. $F_x(\eta) > 1 - \varepsilon$ then $F_{\lambda x}(\eta) = F_x(\eta/|\lambda|) \geq F_x(\eta) > 1 - \varepsilon$ and hence $\lambda x \in O(\varepsilon, \eta)$. \square

Lemma 4. Every ε, η -neighbourhood $O(\varepsilon, \eta)$ is a symmetric set.

Proof. If $x \in O(\varepsilon, \eta)$ then $F_{-x}(\eta) = F_x(\eta) > 1 - \varepsilon$ also, what implies that $-x \in O(\varepsilon, \eta)$. \square

Lemma 5. Let an ε, η -neighbourhood $O(\varepsilon, \eta)$ be given. Then for every $x \in (S, \mathcal{F}, T)$ there exists a $\lambda > 0$ such that $x \in \mu O(\varepsilon, \eta)$ for every μ , $|\mu| \geq \lambda$. This property is called the absorbing property of ε, η -neighbourhoods.

Proof. Since for every $x \in (S, \mathcal{J}, T)$ $\lim_{n \rightarrow \infty} F_x(u) = 1$, i.e. for every $\varepsilon > 0$ there exists $u_x(\varepsilon) > 0$ such that for every $u \geq u_x(\varepsilon)$ we have $F_x(u) > 1 - \varepsilon$, it is evident to put $\lambda = u_x(\varepsilon)/\eta$. If μ is an arbitrary real number with $|\mu| \geq \lambda$ then $F_x(|\mu| \eta) \geq F_x(u_x(\varepsilon)) > 1 - \varepsilon$ and hence $x \in O(\varepsilon, |\mu| \lambda)$. As every $O(\varepsilon, \eta)$ is a symmetric set, then $O(\varepsilon, |\mu| \eta) = \mu \cdot O(\varepsilon, \eta)$. \square

Lemma 6. If an ε, η -neighbourhood $O(\varepsilon, \eta)$ is a convex set, then it is an absolutely convex set in (S, \mathcal{J}, T) .

Proof. It follows immediately from Lemma 3 and Lemma 4.

Lemma 7. For every ε, η -neighbourhood of the zero element in (S, \mathcal{J}, T)

$$S = \bigcup_{n=1}^{\infty} O(\varepsilon, n \cdot \eta).$$

Proof. Let $x \in (S, \mathcal{J}, T)$ and let $O(\varepsilon, \eta)$ be an arbitrary ε, η -neighbourhood of the zero element in S . As Lemma 5 states for the chosen $\varepsilon > 0$ there exists $u(\varepsilon) > 0$ such that $F_x(u(\varepsilon)) > 1 - \varepsilon$. Now, it is sufficient to choose a natural n in such a way that $n \cdot \eta \geq u(\varepsilon)$, at this moment $x \in O(\varepsilon, n\eta) = n \cdot O(\varepsilon, \eta)$. This proves that $S = \bigcup_{n=1}^{\infty} n \cdot O(\varepsilon, \eta)$. \square

Lemma 8. Let x_0 be a cluster point of an ε, η -neighbourhood $O(\varepsilon, \eta)$ in an *SLM*-space (S, \mathcal{J}, T) . Then

$$\lim_{u \rightarrow \eta^+} F_{x_0}(u) \geq 1 - \varepsilon.$$

Proof. Let $\{x_n\} \subset O(\varepsilon, \eta)$, $x_n \xrightarrow{F} x_0$, let $\lambda > 1$. Then, according to the generalized triangular inequality

$$F_{x_0}(\lambda\eta) \geq T(F_{x_n - x_0}((\lambda - 1)\eta), F_{x_n}(\eta)) \geq T(F_{x_n - x_0}((\lambda - 1)\eta), 1 - \varepsilon)$$

for every natural n because $x_n \in O(\varepsilon, \eta)$. But $x_n \xrightarrow{F} x_0$, i.e. $F_{x_n - x_0}((\lambda - 1)\eta) > 1 - \varepsilon'$ for a suitable large n and hence $F_{x_0}(\lambda\eta) \geq T(1 - \varepsilon', 1 - \varepsilon)$. As ε' is quite arbitrary, the t -norm T is continuous and $T(a, 1) = a$ for $a > 0$, this implies $F_{x_0}(\lambda\eta) \geq 1 - \varepsilon$ for every $\lambda > 1$. $F_{x_0}(\cdot)$ is a probability distribution function, therefore the limit $\lim_{u \rightarrow \eta^+} F_{x_0}(u)$ must exist and in this case $\lim_{u \rightarrow \eta^+} F_{x_0}(u) \geq 1 - \varepsilon$. \square

Lemma 9. If $O(\varepsilon, \eta)$ is a convex set in an *SLM*-space (S, \mathcal{J}, T) then its closure $\overline{O(\varepsilon, \eta)}$ in the ε, η -topology can be described as

$$\overline{O(\varepsilon, \eta)} = \{x \in S : \inf \{\lambda > 0 : F_x(\lambda\eta) > 1 - \varepsilon\} \leq 1\}.$$

Proof. If $O(\varepsilon, \eta)$ is a convex set in (S, \mathcal{J}, T) then it is at the same time absolutely convex and absorbing. Let us define a functional (Minkowski functional)

$$\begin{aligned} p_{\varepsilon\eta}(x) &= \inf \{\lambda > 0 : x \in O(\varepsilon, \lambda\eta)\} = \\ &= \inf \{\lambda > 0 : F_x(\lambda\eta) > 1 - \varepsilon\}. \end{aligned}$$

From the properties of the ε, η -neighbourhood $O(\varepsilon, \eta)$ mentioned above it follows that $p_{\varepsilon\eta}(\cdot)$ is a seminorm defined on S . As $O(\varepsilon, \eta)$ is a neighbourhood in the ε, η -topology this seminorm $p_{\varepsilon\eta}(\cdot)$ is continuous in the ε, η -topology, and the closure $\overline{O(\varepsilon, \eta)}$ can be expressed as

$$\overline{O(\varepsilon, \eta)} = \{x \in S : \inf \{\lambda > 0 : F_x(\lambda\eta) > 1 - \varepsilon\} \leq 1\} = \{x : n_{1-\varepsilon}(x) \leq \eta\}$$

where $n_{1-\varepsilon}(x) = \inf \{\lambda > 0 : F_x(\lambda) > 1 - \varepsilon\}$. □

4. PROPERTIES OF MAPPING \mathcal{J}

Let an *SLM*-space (S, \mathcal{F}, T) be given. The mapping \mathcal{J} is defined on the linear space S with values in the set \mathcal{F} of all probability distribution functions defined on real numbers. In \mathcal{F} we can introduce a metric L defined by

$$L(F, G) = \inf \{h > 0 : F(u - h) - h \leq G(u) \leq F(u + h) + h \text{ for every } u \in \mathbb{R}\};$$

this metric is called Lévy's metric and the pair (\mathcal{F}, L) is a complete metric space.

Definition 9. Let (S, \mathcal{F}, T) and (S, \mathcal{F}', T') be two *SLM*-spaces defined on the same linear space S . We shall say that (S, \mathcal{F}, T) and (S, \mathcal{F}', T') are topologically equivalent if the mappings $\mathcal{J}, \mathcal{J}'$ define equivalent ε, η -topologies.

Theorem 5. *SLM*-spaces $(S, \mathcal{F}, T), (S, \mathcal{F}', T')$ are topologically equivalent if and only if the mapping $L(\mathcal{J}(\cdot), \mathcal{J}'(\cdot))$ defined on S is continuous at 0 in both the ε, η -topologies.

Proof. If the ε, η -topologies are equivalent, i.e. if $x_n \xrightarrow{F} 0$ in (S, \mathcal{F}, T) then $x_n \xrightarrow{F'} 0$ in (S, \mathcal{F}', T') and vice versa, then $\mathcal{J}(x_n)(u) = F_{x_n}(u) \rightarrow H(u), \mathcal{J}'(x_n)(u) = F'_{x_n}(u) \rightarrow H(u)$ for every $u \in \mathbb{R}$ what can be expressed also in the form $L(\mathcal{J}(x_n), H) \xrightarrow{n \rightarrow \infty} 0, L(\mathcal{J}'(x_n), H) \rightarrow 0$. From the triangular inequality in the metric space (\mathcal{F}, L)

$$L(\mathcal{J}(x_n), \mathcal{J}'(x_n)) \leq L(\mathcal{J}(x_n), H) + L(\mathcal{J}'(x_n), H)$$

it immediately follows that

$$\lim_{n \rightarrow \infty} L(\mathcal{J}(x_n), \mathcal{J}'(x_n)) = 0.$$

Conversely, if $x_n \xrightarrow{F} 0$ in (S, \mathcal{F}, T) , i.e. $L(F_{x_n}, H) \rightarrow 0$ and we assume that $L(\mathcal{J}(x_n), \mathcal{J}'(x_n)) \rightarrow 0$ also, then $L(\mathcal{J}'(x_n), H) \leq L(\mathcal{J}(x_n), H) + L(\mathcal{J}(x_n), \mathcal{J}'(x_n))$ for every n and hence $\lim_{n \rightarrow \infty} L(\mathcal{J}'(x_n), H) = 0$. This fact says that $x_n \xrightarrow{F'} 0$ in (S, \mathcal{F}', T') and the ε, η -topology in (S, \mathcal{F}, T) is stronger than the ε, η -topology in (S, \mathcal{F}', T') . In a similar way we can prove the opposite implication what completes the proof of Theorem 5. □

Theorem 6. Let an *SLM*-space (S, \mathcal{F}, T) be given. Then the mapping $\mathcal{J} : S \rightarrow (\mathcal{F}, L)$ is uniformly continuous in the ε, η -topology.

Proof. The t -norm T is continuous on $\langle 0, 1 \rangle \times \langle 0, 1 \rangle$ and therefore T is uniformly continuous on $\langle 0, 1 \rangle \times \langle 0, 1 \rangle$ and $\lim_{x \uparrow 1} T(a, x) = a$ uniform in a . It means that $(\forall \eta > 0 \exists \varepsilon \in (0, 1) \forall a \in \langle 0, 1 \rangle) \Rightarrow T(a, 1 - \varepsilon) > a - \eta$. Let $x_n \rightarrow x_0$ in the ε, η -topology, we can find a natural number n_0 such that for every $n \geq n_0$

$$x_n \in O(x_0, \varepsilon, \eta) \Leftrightarrow F_{x_n - x_0}(\eta) > 1 - \varepsilon.$$

Let $u \in \mathbb{R}$ be arbitrary, then

$$F_{x_0}(u + \eta) \geq T(F_{x_0 - x_n}(\eta), F_{x_n}(u)) \geq T(F_{x_n}(u), 1 - \varepsilon) > F_{x_n}(u) - \eta.$$

From this inequality we obtain that $F_{x_0}(u + \eta) + \eta > F_{x_n}(u)$. In a similar way we can prove the opposite inequality $F_{x_n}(u) > F_{x_0}(u - \eta) - \eta$. Both the obtained inequalities express together that $L(F_{x_n}, F_{x_0}) < \eta$. The continuity of the mapping \mathcal{F} in the ε, η -topology is proved. It is necessary to note that a choice of ε and η does not depend on x_n, x_0 and the continuity of \mathcal{F} can be expressed in a stronger form as follows $(\forall \eta > 0 \forall \varepsilon \in (0, 1) \forall x, y \in S, x - y \in O(\varepsilon, \eta)) \Rightarrow L(F_x, F_y) < \eta$. This implication means, of course, the uniform continuity of the mapping \mathcal{F} in the ε, η -topology. \square

Theorem 7. A set $K \subset (S, \mathcal{F}, T)$ is bounded in the ε, η -topology if and only if the image $\mathcal{F}(K)$ in (\mathcal{F}, L) is compact.

Proof. Let K be a bounded subset in (S, \mathcal{F}, T) . It means that for every ε, η -neighbourhood $O(\varepsilon, \eta)$ there exists an $\alpha = \alpha(\varepsilon, \eta) \in \mathbb{R}$ such that for every real $\lambda, |\lambda| \geq \alpha$

$$K \subset \lambda O(\varepsilon, \eta) = O(\varepsilon, |\lambda| \eta).$$

Let $\mathcal{F}(K) = \{F_x : x \in K\}$. If we choose the neighbourhood $O(\varepsilon, 1)$ then for every $\lambda, |\lambda| \geq \alpha(\varepsilon, 1)$ $K \subset O(\varepsilon, |\lambda|)$. It implies that $\mathcal{F}(K) \subset \mathcal{F}(O(\varepsilon, |\lambda|))$ what means for every $|\lambda| \geq \alpha(\varepsilon, 1)$ and every $x \in K$ $F_x(|\lambda|) > 1 - \varepsilon$. We have proved that for every $F \in \mathcal{F}(K)$ and every $u \geq \alpha(\varepsilon, 1)$

$$F(u) > 1 - \varepsilon.$$

This fact can be expressed in the form $\lim_{u \rightarrow \infty} F_x(u) = 1$ uniformly in $x \in K$. As we know that the subset $\mathcal{F}(K)$ is compact in (\mathcal{F}, L) if and only if

$$\lim_{u \rightarrow \infty} F(u) = 1, \quad \lim_{u \rightarrow -\infty} F(u) = 0 \quad \text{uniformly in } \mathcal{F}(K)$$

the necessary part of the proof is finished. Let us suppose that $\mathcal{F}(K)$ is compact in (\mathcal{F}, L) , $K \subset (S, \mathcal{F}, T)$. Then $\lim_{u \rightarrow \infty} F_x(u) = 1$ uniformly in $x \in K$, i.e.

$$(\forall \varepsilon \in (0, 1) \exists \alpha = \alpha(\varepsilon) \forall u \geq \alpha \forall x \in K) \Rightarrow F_x(u) > 1 - \varepsilon.$$

Let $\{x_n\}_1^\infty$ be an arbitrary sequence in K and let $\lambda_n \rightarrow 0$ in reals. Then

$$F_{\lambda_n x_n}(u) = F_{x_n}\left(\frac{u}{|\lambda_n|}\right) > 1 - \varepsilon \quad \text{for } u \geq \alpha |\lambda_n|.$$

As $\lambda_n \rightarrow 0$, then for every $u > 0$ there exists such a natural n_0 that $u \geq \alpha|\lambda_n|$ for every $n \geq n_0$. So, for $u \geq u_0$ we have $\lambda_n x_n \in O(\varepsilon, u)$. The convergence $\lambda_n x_n \xrightarrow{F} 0$ is proved and hence the subset K is bounded in the ε, η -topology. \square

Theorem 8. An *SLM*-space (S, \mathcal{F}, T) with the t -norm $T = \min$ is normable if and only if there exists such an ε, η -neighbourhood $O(\varepsilon, \eta)$ of the zero element that its image $\mathcal{F}(O(\varepsilon, \eta))$ is compact in (\mathcal{F}, L) .

Proof. This statement immediately follows from Theorem 7 and Criterion of normability. \square

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RNDr. Jiří Michálek, CSc., Ústav teorie informace a automatizace ČSAV (Institute of Information Theory and Automation – Czechoslovak Academy of Sciences), Pod vodárenskou věží 4, 182 08 Praha 8, Czechoslovakia.