NEW ZEALAND JOURNAL OF MATHEMATICS Volume 50 (2020), 21-27

GENERALIZATION OF A REAL-ANALYSIS RESULT TO A CLASS OF TOPOLOGICAL VECTOR SPACES

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(Received 8 July, 2019)

Abstract. In this paper, we generalize an elementary real-analysis result to a class of topological vector spaces. We also give an example of a topological vector space to which the result cannot be generalized.

1. Introduction

This paper draws its inspiration from the following result, which appears to be a popular real-analysis exam problem (see [3], for example):

Let $(x_n)_{n \in \mathbb{N}}$ be a sequence in \mathbb{R} . If $\lim_{n \to \infty} (2x_{n+1} - x_n) = x$ for some $x \in \mathbb{R}$, then

 $\lim_{n \to \infty} x_n = x.$ An expedient proof can be given using the Stolz-Cesàro Theorem as follows: **Proof.** Define sequences $(a_n)_{n \in \mathbb{N}}$ and $(b_n)_{n \in \mathbb{N}}$ in \mathbb{R} by

$$\forall n \in \mathbb{N}: \quad a_n \stackrel{\mathrm{df}}{=} 2^n x_n \quad \text{and} \quad b_n \stackrel{\mathrm{df}}{=} 2^n.$$

Then $(b_n)_{n\in\mathbb{N}}$ is strictly increasing and diverges to ∞ , and as

$$\lim_{n \to \infty} \frac{a_{n+1} - a_n}{b_{n+1} - b_n} = \lim_{n \to \infty} \frac{2^{n+1}x_{n+1} - 2^n x_n}{2^{n+1} - 2^n}$$
$$= \lim_{n \to \infty} \frac{2^{n+1}x_{n+1} - 2^n x_n}{2^n}$$
$$= \lim_{n \to \infty} (2x_{n+1} - x_n)$$
$$= x,$$

the Stolz-Cesàro Theorem immediately tells us that $\lim_{n\to\infty} x_n = \lim_{n\to\infty} \frac{a_n}{b_n} = x.$

A natural question to ask is: Is this result still valid if \mathbb{R} is replaced by another topological vector space? The answer happens to be affirmative for a wide class of topological vector spaces that includes all the locally convex ones.

We will also exhibit a topological vector space for which the result is not valid, which indicates that it is rather badly behaved.

In this paper, we adopt the following conventions:

- \mathbb{N} denotes the set of all positive integers, and for each $n \in \mathbb{N}$, let $[n] \stackrel{\text{df}}{=} \mathbb{N}_{< n}$.
- All vector spaces are over the field $\mathbb{K} \in \{\mathbb{R}, \mathbb{C}\}$.

²⁰¹⁰ Mathematics Subject Classification 28A20, 46A16, 60A10.

Key words and phrases: Topological vector spaces, locally convex topological vector spaces, phomogeneous seminorms, random variables, convergence in probability.

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2. Good Topological Vector Spaces

Recall that a topological vector space is an ordered pair (V, τ) , where:

- V is a vector space, and
- τ is a topology on V, under which vector addition and scalar multiplication are continuous operations.

Definition 2.1. Let (V, τ) be a topological vector space, and $(x_{\lambda})_{\lambda \in \Lambda}$ a net in V. Then $x \in V$ is called a τ -*limit* for $(x_{\lambda})_{\lambda \in \Lambda}$ — which we write as $(x_{\lambda})_{\lambda \in \Lambda} \xrightarrow{\tau} x$ — if and only if for each τ -neighborhood U of x, there is a $\lambda_0 \in \Lambda$ such that $x_{\lambda} \in U$ for all $\lambda \in \Lambda_{\geq \lambda_0}$.

Remark 2.2. We do not assume that τ is a Hausdorff topology on V.

Definition 2.3. A topological vector space (V, τ) is said to be *good* if and only if any sequence $(x_n)_{n \in \mathbb{N}}$ in V has a τ -limit whenever $(2x_{n+1} - x_n)_{n \in \mathbb{N}}$ has a τ -limit. A topological vector space that is not good is said to be *bad*.

Proposition 2.4. Let (V, τ) be a topological vector space, and $(x_n)_{n \in \mathbb{N}}$ a sequence in V such that $(2x_{n+1} - x_n)_{n \in \mathbb{N}} \xrightarrow{\tau} x$ for some $x \in V$. Then either

- $(x_n)_{n\in\mathbb{N}} \xrightarrow{\tau} x \text{ also, or }$
- $(x_n)_{n\in\mathbb{N}}$ has no τ -limit.

Proof. If $(x_n)_{n \in \mathbb{N}}$ has no τ -limit, then we are done.

Next, suppose that $(x_n)_{n \in \mathbb{N}} \xrightarrow{\tau} y$ for some $y \in V$. Then

$$(2x_{n+1} - x_n)_{n \in \mathbb{N}} \xrightarrow{\tau} 2y - y = y,$$

so y is a τ -limit for $(2x_{n+1} - x_n)_{n \in \mathbb{N}}$ in addition to x. It follows that

$$(0_V)_{n\in\mathbb{N}} = \left((2x_{n+1} - x_n) - (2x_{n+1} - x_n) \right)_{n\in\mathbb{N}} \xrightarrow{\tau} x - y,$$

which yields

$$(y)_{n \in \mathbb{N}} = (0_V + y)_{n \in \mathbb{N}} \xrightarrow{\tau} (x - y) + y = x.$$

Therefore, any τ -neighborhood of x also contains y, giving us $(x_n)_{n \in \mathbb{N}} \xrightarrow{\tau} x$. \Box

Proposition 2.4 tells us: To prove that a topological vector space (V, τ) is good, it suffices to prove that for each sequence $(x_n)_{n \in \mathbb{N}}$ in V, if $(2x_{n+1} - x_n)_{n \in \mathbb{N}} \xrightarrow{\tau} x$ for some $x \in V$, then $(x_n)_{n \in \mathbb{N}} \xrightarrow{\tau} x$ also.

Definition 2.5. Let $p \in (0, 1]$. A *p*-homogeneous seminorm on a vector space V is then a function $\sigma : V \to \mathbb{R}_{>0}$ with the following properties:

(1) The Triangle Inequality: $\sigma(x+y) \leq \sigma(x) + \sigma(y)$ for all $x, y \in V$.

(2) *p*-Homogeneity: $\sigma(kx) = |k|^p \sigma(x)$ for all $k \in \mathbb{K}$ and $x \in V$.

Remark 2.6. • By letting k = 0 and $x = 0_V$ in (2), we find that $\sigma(0_V) = 0$.

- A 1-homogeneous seminorm is the same as a seminorm in the ordinary sense.
- No extra generality is gained by postulating that $\sigma(kx) \leq |k|^p \sigma(x)$ for all $k \in \mathbb{K}$ and $x \in V$. If $k \in \mathbb{K} \setminus \{0\}$, then replacing k by $\frac{1}{k}$ gives us the reverse inequality, which leads to equality; if k = 0, then equality automatically holds.

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• We do not consider $p \in (2, \infty)$ because

$$\forall x \in V : \quad 2^p \sigma(x) = \sigma(2x) \qquad \text{(By p-homogeneity.)}$$
$$= \sigma(x+x)$$
$$\leq 2\sigma(x), \qquad \text{(By the Triangle Inequality.)}$$

so if σ is non-trivial, then $2^p \leq 2$, which implies that $p \in (0,1]$ if $p \in \mathbb{R}_{>0}$.

Let V be a vector space, and S a collection of p-homogeneous seminorms on V where $p \in (0,1]$ may not be fixed. Define a function $\mathcal{U}: V \times \mathcal{S} \times \mathbb{R}_{>0} \to \mathcal{P}(V)$ by

$$\forall x \in V, \ \forall \sigma \in \mathcal{S}, \ \forall \epsilon \in \mathbb{R}_{>0}: \quad \mathcal{U}_{x,\sigma,\epsilon} \stackrel{\mathrm{df}}{=} \{ y \in V \mid \sigma(y-x) < \epsilon \}.$$

Then let $\tau_{\mathcal{S}}$ denote the topology on V that is generated by the sub-base

$$\{\mathcal{U}_{x,\sigma,\epsilon} \in \mathcal{P}(V) \mid (x,\sigma,\epsilon) \in V \times \mathcal{S} \times \mathbb{R}_{>0}\}.$$

Proposition 2.7. The following statements about τ_S hold:

(1) $\tau_{\mathcal{S}}$ is a vector-space topology on V.

(2) Let $(x_{\lambda})_{\lambda \in \Lambda}$ be a net in V. Then for each $x \in V$, we have

$$(x_{\lambda})_{\lambda \in \Lambda} \xrightarrow{\tau_{\mathcal{S}}} x \qquad \Longleftrightarrow \qquad \lim_{\lambda \in \Lambda} \sigma(x_{\lambda} - x) = 0 \text{ for all } \sigma \in \mathcal{S}.$$

Proof. One only has to imitate the proof in the case of locally convex topological vector spaces that the initial topology generated by a collection of seminorms is a vector-space topology. We refer the reader to Chapter 1 of [2] for details. \square

Proposition 2.8. (V, τ_S) is a good topological vector space.

Proof. Let $(x_n)_{n \in \mathbb{N}}$ be a sequence in V. Suppose that $(2x_{n+1} - x_n)_{n \in \mathbb{N}} \xrightarrow{\tau_S} x$ for some $x \in V$. Then without loss of generality, we may assume that $x = 0_V$. To see why, define a new sequence $(y_n)_{n \in \mathbb{N}}$ in V by $y_n \stackrel{\text{df}}{=} x_n - x$ for all $n \in \mathbb{N}$, so that

$$\forall n \in \mathbb{N} : \quad 2y_{n+1} - y_n = 2(x_{n+1} - x) - (x_n - x)$$
$$= 2x_{n+1} - 2x - x_n + x$$
$$= (2x_{n+1} - x_n) - x.$$

Hence,

$$(2y_{n+1} - y_n)_{n \in \mathbb{N}} = ((2x_{n+1} - x_n) - x)_{n \in \mathbb{N}} \xrightarrow{\tau_S} x - x = 0_V$$

so if we can prove that $(y_n)_{n \in \mathbb{N}} \xrightarrow{\tau_{\mathcal{S}}} 0_V$, then $(x_n)_{n \in \mathbb{N}} \xrightarrow{\tau_{\mathcal{S}}} x$ as desired. Let $\sigma \in \mathcal{S}$ and $\epsilon > 0$, and suppose that σ is *p*-homogeneous for some $p \in (0, 1]$. Then by (2) of Proposition 2.7, there is an $N \in \mathbb{N}$ such that

$$\forall n \in \mathbb{N}_{\geq N}$$
: $\sigma(2x_{n+1} - x_n) = \sigma((2x_{n+1} - x_n) - 0_V) < (2^p - 1)\epsilon$

By *p*-homogeneity, we thus have

$$\forall k \in \mathbb{N} : \quad \sigma \left(2^k x_{N+k} - 2^{k-1} x_{N+k-1} \right) = \sigma \left(2^{k-1} (2x_{N+k} - x_{N+k-1}) \right)$$
$$= 2^{(k-1)p} \sigma (2x_{N+k} - x_{N+k-1})$$
$$< 2^{(k-1)p} (2^p - 1) \epsilon.$$

Next, a telescoping sum in conjunction with the Triangle Inequality yields

$$\forall m \in \mathbb{N}: \quad \sigma(2^m x_{N+m} - x_N) = \sigma\left(\sum_{k=1}^m (2^k x_{N+k} - 2^{k-1} x_{N+k-1})\right)$$

$$\leq \sum_{k=1}^{m} \sigma \left(2^{k} x_{N+k} - 2^{k-1} x_{N+k-1} \right)$$

$$< \sum_{k=1}^{m} 2^{(k-1)p} (2^{p} - 1) \epsilon$$

$$= (2^{mp} - 1) \epsilon.$$

Then by p-homogeneity again,

$$\forall m \in \mathbb{N} : \quad \sigma \left(x_{N+m} - \frac{1}{2^m} x_N \right) = \sigma \left(\frac{1}{2^m} (2^m x_{N+m} - x_N) \right)$$
$$= \frac{1}{2^{mp}} \sigma (2^m x_{N+m} - x_N)$$
$$< \left(1 - \frac{1}{2^{mp}} \right) \epsilon.$$

Applying the Triangle Inequality and *p*-homogeneity once more, we get

$$\forall m \in \mathbb{N}: \quad \sigma(x_{N+m}) < \sigma\left(\frac{1}{2^m}x_N\right) + \left(1 - \frac{1}{2^{mp}}\right)\epsilon = \frac{1}{2^{mp}}\sigma(x_N) + \left(1 - \frac{1}{2^{mp}}\right)\epsilon.$$
Consequently

Consequently,

$$\limsup_{n \to \infty} \sigma(x_n) = \limsup_{m \to \infty} \sigma(x_{N+m}) \le \limsup_{m \to \infty} \left[\frac{1}{2^{mp}} \sigma(x_N) + \left(1 - \frac{1}{2^{mp}} \right) \epsilon \right] = \epsilon.$$

As $\epsilon > 0$ is arbitrary, we obtain

$$\lim_{n \to \infty} \sigma(x_n - 0_V) = \lim_{n \to \infty} \sigma(x_n) = 0.$$

Finally, as $\sigma \in \mathcal{S}$ is arbitrary, (2) of Proposition 2.7 says that $(x_n)_{n \in \mathbb{N}} \xrightarrow{\tau_{\mathcal{S}}} 0_V$. \Box

By Proposition 2.8, the class of good topological vector spaces includes:

- All locally convex topological vector spaces.
- All L^p -spaces for $p \in (0, 1)$, which are generally not locally convex.

In the next section, we will give an example of a bad topological vector space.

3. A Bad Topological Vector Space from Probability Theory

Before we present the example, let us first fix some probabilistic terminology.

Definition 3.1. Let $(\Omega, \Sigma, \mathsf{P})$ be a probability space.

- A measurable function from (Ω, Σ) to $(\mathbb{R}, \mathscr{B}(\mathbb{R}))$ is called a *random variable*.¹
- The \mathbb{R} -vector space of random variables on (Ω, Σ) is denoted by $\mathsf{RV}(\Omega, \Sigma)$.
- Let $(X_{\lambda})_{\lambda \in \Lambda}$ be a net in $\mathsf{RV}(\Omega, \Sigma)$, and let $X \in \mathsf{RV}(\Omega, \Sigma)$. Then $(X_{\lambda})_{\lambda \in \Lambda}$ is said to converge in probability to X (for P) if and only if for each $\epsilon > 0$, we have

$$\lim_{\lambda \in \Lambda} \mathsf{P}(\{\omega \in \Omega \mid |X_{\lambda}(\omega) - X(\omega)| > \epsilon\}) = 0,$$

in which case, we write $(X_{\lambda})_{\lambda \in \Lambda} \xrightarrow{\mathsf{P}} X$.

The following theorem says that convergence in probability is convergence with respect to a vector-space topology on the vector space of random variables.

 $^{{}^1\}mathscr{B}(\mathbb{R})$ denotes the Borel $\sigma\text{-algebra generated}$ by the standard topology on $\mathbb{R}.$

Theorem 3.2. Let $(\Omega, \Sigma, \mathsf{P})$ be a probability space, and define a pseudo-metric ρ_{P} on $\mathsf{RV}(\Omega, \Sigma)$ by

$$\forall X, Y \in \mathsf{RV}(\Omega, \Sigma) : \quad \rho_{\mathsf{P}}(X, Y) \stackrel{\mathrm{df}}{=} \int_{\Omega} \frac{|X - Y|}{1 + |X - Y|} \, \mathrm{d}\mathsf{P}.$$

Then the topology τ_{P} on $\mathsf{RV}(\Omega, \Sigma)$ generated by ρ_{P} has the following properties: • $\tau_{\rm P}$ is a vector-space topology.

• Let $(X_{\lambda})_{\lambda \in \Lambda}$ be a net in $\widetilde{\mathsf{RV}}(\Omega, \Sigma)$. Then for each $X \in \mathsf{RV}(\Omega, \Sigma)$, we have

$$(X_{\lambda})_{\lambda \in \Lambda} \xrightarrow{\mathsf{P}} X \qquad \Longleftrightarrow \qquad (X_{\lambda})_{\lambda \in \Lambda} \xrightarrow{\tau_{\mathsf{P}}} X.$$

Proof. Please refer to Problems 6, 10 and 14 in Section 5.2 of [1].

Now, for each $k \in \mathbb{N}$, define a probability measure c_k on $([k], \mathcal{P}([k]))$ by

$$\forall A \subseteq [k]: \quad \mathsf{c}_k(A) \stackrel{\mathrm{df}}{=} \frac{\mathsf{Card}(A)}{k}$$

and let $(\Omega, \Sigma, \mathsf{P})$ denote the product probability space $\prod_{k=1}^{\infty} ([k], \mathcal{P}([k]), \mathsf{c}_k)$. Define a

sequence
$$(S_n)_{n\in\mathbb{N}}$$
 in Σ by

$$\forall n \in \mathbb{N} : \quad S_n \stackrel{\mathrm{df}}{=} \left\{ \mathbf{v} \in \prod_{k=1}^{\infty} [k] \; \middle| \; \mathbf{v}(n) = 1 \right\}.$$

Then $\mathsf{P}(S_n) = \frac{1}{n}$ for all $n \in \mathbb{N}$, and the S_n 's form mutually-independent events. Next, define a sequence $(Y_n)_{n \in \mathbb{N}}$ in $\mathsf{RV}(\Omega, \Sigma)$ by

$$\forall n \in \mathbb{N} : \quad Y_n \stackrel{\mathrm{df}}{=} 2^n \chi_{S_n},$$

where χ_{S_n} denotes the indicator function of S_n . Then we get for each $\epsilon > 0$ that

$$\lim_{n \to \infty} \mathsf{P}(\{\omega \in \Omega \mid |Y_n(\omega)| > \epsilon\}) = \lim_{n \to \infty} \mathsf{P}(S_n) = \lim_{n \to \infty} \frac{1}{n} = 0.$$

The first equality is obtained because, for each $\epsilon > 0$, we have $2^n > \epsilon$ for all $n \in \mathbb{N}$ large enough. Consequently, $(Y_n)_{n \in \mathbb{N}} \xrightarrow{\mathsf{P}} 0_{\Omega \to \mathbb{R}}$. Define a new sequence $(X_n)_{n \in \mathbb{N}}$ in $\mathsf{RV}(\Omega, \Sigma)$ by

$$\forall n \in \mathbb{N} : \quad X_n \stackrel{\text{df}}{=} \begin{cases} 0_{\Omega \to \mathbb{R}} & \text{if } n = 1; \\ \sum_{k=1}^{n-1} \frac{1}{2^{n-k}} Y_k & \text{if } n \ge 2. \end{cases}$$

Then $2X_2 - X_1 = 2X_2 = Y_1$, and

$$\forall n \in \mathbb{N}_{\geq 2} : \quad 2X_{n+1} - X_n = 2\sum_{k=1}^n \frac{1}{2^{n+1-k}} Y_k - \sum_{k=1}^{n-1} \frac{1}{2^{n-k}} Y_k$$
$$= \sum_{k=1}^n \frac{1}{2^{n-k}} Y_k - \sum_{k=1}^{n-1} \frac{1}{2^{n-k}} Y_k$$
$$= Y_n.$$

It follows that $(2X_{n+1} - X_n)_{n \in \mathbb{N}} = (Y_n)_{n \in \mathbb{N}} \xrightarrow{\mathsf{P}} 0_{\Omega \to \mathbb{R}}.$

Gathering what we have thus far, observe that

$$\forall n \in \mathbb{N} : \quad X_{2n+1} = \sum_{k=1}^{2n} \frac{1}{2^{2n+1-k}} Y_k$$
$$= \sum_{k=1}^{2n} \frac{1}{2^{2n+1-k}} (2^k \chi_{S_k})$$
$$= \sum_{k=1}^{2n} 2^{2k-2n-1} \chi_{S_k}$$
$$\ge \sum_{k=n+1}^{2n} 2^{2k-2n-1} \chi_{S_k}$$
$$\ge \sum_{k=n+1}^{2n} \chi_{S_k}$$
$$\ge \chi_{\bigcup_{k=n+1}^{2n} S_k}.$$

As the ${\cal S}_k$'s are mutually independent, their complements are as well, so

$$\forall n \in \mathbb{N} : \quad \mathsf{P}\left(\left\{\omega \in \Omega \mid |X_{2n+1}(\omega)| > \frac{1}{2}\right\}\right) \ge \mathsf{P}\left(\bigcup_{k=n+1}^{2n} S_k\right) \\ = 1 - \mathsf{P}\left(\Omega \setminus \bigcup_{k=n+1}^{2n} S_k\right) \\ = 1 - \mathsf{P}\left(\bigcap_{k=n+1}^{2n} \Omega \setminus S_k\right) \\ = 1 - \prod_{k=n+1}^{2n} \mathsf{P}(\Omega \setminus S_k) \\ = 1 - \prod_{k=n+1}^{2n} \left(1 - \frac{1}{k}\right) \\ = 1 - \prod_{k=n+1}^{2n} \frac{k-1}{k} \\ = 1 - \frac{1}{2n} \\ = 1 - \frac{1}{2} \\ = \frac{1}{2}.$$

Hence, $(X_n)_{n \in \mathbb{N}}$ does not converge to $0_{\Omega \to \mathbb{R}}$ in probability. By Theorem 3.2: **Proposition 3.3.** $(\mathsf{RV}(\Omega, \Sigma), \tau_{\mathsf{P}})$ is therefore a bad topological vector space.

By Proposition 2.4, $(X_n)_{n \in \mathbb{N}}$ does not, in fact, converge in probability at all.

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4. Acknowledgments

The author would like to express his deepest thanks to Dr. Jochen Wengenroth for communicating his example above of a bad topological vector space.

Many thanks also go to the referee who meticulously proofread this paper.

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