Some results on uniform statistical cluster points

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Abstract: In this paper, we present some results linking the uniform statistical limit superior and inferior, almost convergence and uniform statistical convergence of a sequence. We also study the relationship between the set of uniform statistical cluster points of a given sequence and its subsequences. The results concerning uniform statistical convergence and uniform statistical cluster points presented here are also closely related to earlier results regarding statistical convergence and statistical cluster points of a sequence.

Key words: Uniform statistical convergence, subsequences, uniform statistical cluster points

1. Introduction

The convergence of sequences has undergone numerous generalizations in order to provide deeper insights into summability theory. Convergence of sequences has different generalizations. One of the most important generalizations is uniform statistical convergence. This type of convergence has been introduced by Brown and Freedman [3] by using uniform density and has been studied by many authors in various directions [2, 14, 15, 19, 20]. This type of convergence is stronger than ordinary convergence and so it is quite effective, especially when the classical limit does not exist.

Buck [5] initiated the study of the relationship between the convergence of a given sequence and the summability of its subsequences. Later Agnew [1], Buck [6], Buck and Pollard [7], Miller and Orhan [18], and Zeager [22] studied this relation changing the concept of convergence. Moreover, Dawson [8] and Fridy [11] have studied analogous results by replacing subsequences with stretching and rearrangements, respectively. In [21], we studied some relationships between convergence and uniform statistical convergence of a given sequence and its subsequences. The related notions of statistical limit superior and inferior and statistical cluster points have been studied in recent papers including [12, 13] by Fridy and Orhan and [16] by Miller and Miller-Van Wieren.

In the present paper, we are concerned with the relationships between the uniform statistical limit superior and inferior, almost convergence and uniform statistical convergence of a sequence. We also show some results about the set of uniform statistical cluster points of a given sequence and its subsequences and stretchings, including the discussion of the Lebesgue measure of the set of subsequences that retain the same set of uniform statistical cluster points.

Now let us recall some known notions. Let \( K \subseteq \mathbb{N} \), where \( \mathbb{N} \) is the set of natural numbers. If \( m, n \in \mathbb{N} \),

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by \( K(m, n) \) we denote the cardinality of the set of numbers \( i \) in \( K \) such that \( m \leq i \leq n \). Numbers

\[
d(K) = \liminf_{n \to \infty} \frac{K(1, n)}{n}, \quad \bar{d}(K) = \limsup_{n \to \infty} \frac{K(1, n)}{n}
\]

are called the lower and the upper asymptotic density of the set \( K \), respectively. If \( d(K) = \bar{d}(K) \) then it is said that \( d(K) = \bar{d}(K) \) is the asymptotic density of \( K \). The uniform density of \( K \subseteq \mathbb{N} \) has been introduced in [3, 4] as follows:

\[
\bar{u}(K) = \lim_{n \to \infty} \frac{\min_{i \geq 0} K(i + 1, i + n)}{n}, \quad \bar{u}(K) = \lim_{n \to \infty} \frac{\max_{i \geq 0} K(i + 1, i + n)}{n}
\]

are respectively called the lower and the upper uniform density of the set \( K \) (the existence of these bounds is also mentioned in [2]). If \( \bar{u}(K) = \bar{u}(K) \), then \( u(K) = \bar{u}(K) \) is called the uniform density of \( K \). It is clear that for each \( K \subseteq \mathbb{N} \) we have

\[
\bar{u}(K) \leq d(K) \leq \bar{d}(K) \leq \bar{u}(K).
\]

The concept of statistical convergence has been introduced in [10] as follows: Let \( x = \{x_n\} \) be a sequence of complex numbers. The sequence \( x \) is said to be statistically convergent to a complex number \( L \) provided that for every \( \varepsilon > 0 \) we have \( d(K_{\varepsilon}) = 0 \), where \( K_{\varepsilon} = \{ n \in \mathbb{N} : |x_n - L| \geq \varepsilon \} \). If \( x = \{x_n\} \) converges statistically to \( L \), then we write \( st - \lim x = L \).

Next we introduce the concept of uniform statistical convergence, which is the primary topic of this paper. A generalized approach to convergence has been obtained by means of the notion of an ideal \( I \) of subsets of \( \mathbb{N} \), i.e. \( I \) is an additive and hereditary class of sets. A sequence \( x \) is said to be \( I \)-convergent to \( L \) if for every \( \varepsilon \) the set \( K_{\varepsilon} = \{ n \in \mathbb{N} : |x_n - L| \geq \varepsilon \} \) belongs to \( I \), and we write \( I - \lim x = L \). If \( I = I_d = \{ A \subseteq \mathbb{N} : d(A) = 0 \} \), then \( I_d \)-convergence coincides with statistical convergence. In the case \( I = I_u = \{ A \subseteq \mathbb{N} : u(A) = 0 \} \) we obtain uniform statistical convergence to \( L \) or \( I_u \) convergence to \( L \). Then we write \( st_u - \lim x = L \).

In [18], Orhan and Miller have studied the concept of almost convergence of sequences and have obtained some results regarding subsequences. Namely a bounded sequence \( x = \{x_n\} \) is almost convergent to \( L \) if

\[
\lim_{n \to \infty} \frac{\sum_{i=m+1}^{m+n} x_i}{n} = L
\]

uniformly in \( m \) (see [18]). Orhan and Miller [18] have shown that if \( x \) almost converges to \( L \), then the set of \( t \in (0, 1] \) for which the associated subsequence \( (x_t) \) almost converges to \( L \) must have measure \( 0 \) or \( 1 \) (both values may occur).

**Definition 1** \( \gamma \) is called a uniform statistical cluster point of \( x = \{x_k\} \) if for every \( \varepsilon > 0 \) the set \( \{k : |x_k - \gamma| < \varepsilon\} \) does not have uniform density \( 0 \).

Let \( \Gamma_u \) denote the set of all uniform statistical cluster points of \( x \) and \( \Gamma_s \) denote the set of all statistical cluster points. Clearly \( \Gamma_s \subseteq \Gamma_u \). It can happen that \( \Gamma_s \nsubseteq \Gamma_u \), for example if

\[
x = 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, ..., 0, 0, 0, 1, 1, 1, ...
\]

where segments of 0’s of length \( 2^k \), \( k = 0, 1, 2, 3, ... \) and 1’s of length \( k + 1 \), \( k = 0, 1, 2, 3, ... \) alternate. Then one can see that \( x \) is statistically convergent to 0 and so \( \Gamma_s = \{0\} \) but \( \Gamma_u = \{0, 1\} \).

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In [12], Fridy and Orhan introduced the definitions of the statistical limit superior and statistical limit inferior of a sequence and proved some results concerning these notions. Here we proceed analogously for the case of uniform statistical convergence.

For a real sequence \( x \), let \( A_x, B_x \) denote
\[
B_x = \{ b \in \mathbb{R} : u(\{ k : x_k > b \}) \neq 0 \},
\]
\[
A_x = \{ a \in \mathbb{R} : u(\{ k : x_k < a \}) \neq 0 \}.
\]

Definition 2 For \( x = \{ x_k \} \) the uniform statistical limit superior and uniform statistical limit inferior are given by
\[
\text{st}_u \limsup x := \begin{cases} 
\sup B_x & \text{if } B_x \neq \emptyset \\
-\infty & \text{if } B_x = \emptyset
\end{cases}
\]
\[
\text{st}_u \liminf x := \begin{cases} 
\inf A_x & \text{if } A_x \neq \emptyset \\
\infty & \text{if } A_x = \emptyset
\end{cases}
\]

Theorem 1 If \( \beta = \text{st}_u \limsup x \) is finite then for every positive \( \varepsilon \)
\[ u(\{ k : x_k > \beta - \varepsilon \}) \neq 0 \text{ and } u(\{ k : x_k > \beta + \varepsilon \}) = 0. \tag{1.1} \]
Conversely if (1.1) holds for every positive \( \varepsilon \) then \( \beta = \text{st}_u \limsup x \).

Theorem 2 If \( \alpha = \text{st}_u \liminf x \) is finite then for every positive \( \varepsilon \)
\[ u(\{ k : x_k < \alpha + \varepsilon \}) \neq 0 \text{ and } u(\{ k : x_k < \alpha - \varepsilon \}) = 0. \tag{1.2} \]
Conversely if (1.2) holds for every positive \( \varepsilon \) then \( \alpha = \text{st}_u \liminf x \).

Furthermore, it is easy to see that \( \Gamma_u \) is a closed set and if \( \beta = \text{st}_u \limsup x \) is finite then \( \beta = \max \Gamma_u \) (and vice versa) and if \( \alpha = \text{st}_u \liminf x \) is finite then \( \alpha = \min \Gamma_u \) (and vice versa).

In [9], the concepts of \( I \)-limit superior and inferior and \( I \)-core have been defined and studied in a different way.

Theorem 3 \( x \) converges uniformly statistically to \( \alpha \) if and only if \( \alpha = \text{st}_u \limsup x = \text{st}_u \liminf x \).

2. Results

Fridy and Orhan [12] proved the following theorem.

Theorem 4 If the sequence \( x \) is bounded above and \( C_1 \) summable to \( \beta = \text{st} \limsup x \), then \( x \) is statistically convergent to \( \beta \).

Here we prove the following analogue.

Theorem 5 If the sequence \( x \) is bounded and almost convergent to \( \beta = \text{st}_u \limsup x \), then \( x \) is uniformly statistically convergent to \( \beta \).
Proof  Let us use the following notation:

\[ K_{m+1}^{m+n} = K \cap \{m + 1, m + 2, ..., m + n\} \text{ for } K \subseteq \mathbb{N}. \]

Suppose \( x \) is not uniformly statistically convergent to \( \beta \). Then \( \text{st}_n \liminf x < \beta \text{ and so there exists } \mu < \beta \text{ such that } u(\{k : x_k < \mu\}) \neq 0 \). Let \( K' = \{k : x_k < \mu\} \). Since \( u(K') \neq 0 \) there exists \( d > 0 \) and infinitely many \( n, n \to \infty \) such that for each \( n \) there exists \( m = m(n) \) so that

\[
\left| K_{m+1}^{m+n} \right| = \left| \left\{ k \in K' : m + 1 \leq k \leq m + n \right\} \right| \geq d.
\]

Suppose \( \varepsilon \) is arbitrary fixed. Then \( u(\{k : x_k > \beta + \varepsilon\}) = 0 \). Define

\[ K'' = \{k : \mu \leq x_k \leq \beta + \varepsilon\}, K''' = \{k : x_k > \beta + \varepsilon\} \text{ and let } B = \sup_k x_k < \infty. \]

Then for each previously mentioned \( n, m = m(n) \) we have

\[
\frac{1}{n} \sum_{k=m+1}^{m+n} x_k = \frac{1}{n} \sum_{k \in K_{m+1}^{m+n}} x_k + \frac{1}{n} \sum_{k \in K_{m+1}''^{m+n}} x_k + \frac{1}{n} \sum_{k \in K_{m+1}'''^{m+n}} x_k
\]

\[
< \frac{\mu}{n} K_{m+1}^{m+n} + \frac{\beta + \varepsilon}{n} \left| K_{m+1}^{m+n} \right| + \frac{B}{n} \left| K_{m+1}''^{m+n} \right|
\]

\[
\leq \frac{\mu}{n} K_{m+1}^{m+n} + (\beta + \varepsilon) \left( 1 - \frac{\left| K_{m+1}^{m+n} \right|}{n} \right) + \frac{B}{n} \left| K_{m+1}'''^{m+n} \right|
\]

\[
\leq \mu \frac{K_{m+1}^{m+n}}{n} + (\beta + \varepsilon) \left( 1 - \frac{\left| K_{m+1}^{m+n} \right|}{n} \right) + M \frac{K_{m+1}'''^{m+n}}{n}, \text{ } M \text{ some constant}
\]

\[
\leq \beta + (\mu - \beta) \frac{K_{m+1}^{m+n}}{n} + \varepsilon (1 - d) + M \frac{K_{m+1}'''^{m+n}}{n}
\]

\[
\leq \beta - (\beta - \mu) d + \varepsilon (1 - d) + M \frac{K_{m+1}'''^{m+n}}{n}.
\]

Since for any \( \varepsilon > 0, M \frac{K_{m+1}'''^{m+n}}{n} \to 0 \text{ as } n \to \infty \) and \( \varepsilon \) can be chosen arbitrarily small we get that for infinitely many \( n, m = m(n) \)

\[
\frac{1}{n} \sum_{k=m+1}^{m+n} x_k \leq \beta - d(\beta - \mu) < \beta
\]

and so \( x \) does not almost converge to \( \beta \).

\[ \square \]

Remark 1  The assumption that \( x \) is bounded is necessary as of course almost convergent sequences must be bounded.
Remark 2 Clearly, the uniform statistical limit superior can be replaced by the uniform statistical limit inferior in the statement of the above theorem.

In \[16\] the following theorem has been proved.

**Theorem 6** Suppose \( x = \{x_k\} \) is bounded and has \( L \) as its set of limit points, and \( \Gamma \subseteq L \) is closed and nonempty. Then there exists a subsequence \( \{x_{n_k}\} \) of \( x \) such that \( \Gamma \) is the set of statistical cluster points of \( \{x_{n_k}\} \).

Using the same construction of subsequence as in that proof with \( \Gamma \), closed we get \( \{x_{n_k}\} \) whose set of uniform statistical cluster points is \( \Gamma \). Because the subsequence is constructed by taking "every other" repeatedly the asymptotic densities in the proof will be at the same time uniform densities. Thus we will have the following theorem.

**Theorem 7** If \( x = \{x_k\} \) is bounded and has \( L \) as its set of limit points, and \( \Gamma \subseteq L \) is closed and nonempty.

Then there exists a subsequence \( \{x_{n_k}\} \) of \( x \) such that \( \Gamma \) is the set of uniform statistical cluster points of \( \{x_{n_k}\} \).

However, the theorem about "stretching" from that paper cannot be generalized. Consider \( x = 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, ..., 0, 0, 0, 1, 1, 1, ... \), where segments of 0’s of length \( 2^k \), \( k = 0, 1, 2, 3, ... \) and 1’s of length \( k + 1, k = 0, 1, 2, 3, ..., \) alternate. There is no stretching of \( x \) with \( \{0\} \) (or \( \{1\} \)) as its set of uniform statistical cluster points.

If \( x = \{x_n\} \) is a sequence of reals, for \( t \in (0, 1] \) written in binary representation with infinitely many ones, let \((xt)\) denote the usual generated subsequence of \( x \).

The following theorem has been proved in \[17\].

**Theorem 8** Given a bounded sequence \( x \) if \( \Gamma_u(x) \) is the set of its statistical cluster points, the set \( \{t \in (0, 1] : \Gamma_u(xt) = \Gamma_u(x)\} \) has measure 1.

Let \( \Gamma_u(x) \) denote the set of uniform statistical cluster points of \( x \); likewise we can prove:

**Theorem 9** Suppose \( x \) is a bounded sequence of reals. Then \( T = \{t \in (0, 1] : \Gamma_u(xt) = \Gamma_u(x)\} \) has measure 0 or 1 (both can occur).

**Proof** First \( T \) is a tail set and so \( T \) has measure 0 or 1 or it is unmeasurable. First we will check \( T \) is measurable. For \( a \in [\lim \inf x, \lim \sup x] \), let

\[ \Gamma_u(a) = \{t \in (0, 1] : a \text{ is uniform statistical cluster point of } (xt)\}. \]

Then

\[ T = \bigcap_{a \in \Gamma_u(x)} \Gamma_u(a) \cap \bigcap_{a \in [\lim \inf x, \lim \sup x] \setminus \Gamma_u(x)} (0, 1 \setminus \Gamma_u(a)) \]

since \((xt)\) has the same uniform statistical cluster points as \( x \). Now \( \Gamma_u(x) \) is closed and separable and so there exists \( \{a_n\}, a_n \in \Gamma_u(x) \) dense in \( \Gamma_u(x) \).
Easily
$$\bigcap_{a \in \Gamma_n(x)} \Gamma_n(a) = \bigcap_n \Gamma_n(a_n).$$

For any \(a\)
$$\Gamma_n(a) = \bigcap_{k} \bigcup_{j} \bigcap_{n \geq N} \bigcup_{m} \left\{ t \in (0,1) : \frac{\{m + 1 \leq i \leq m + n : |(xt)_i - a| < \frac{1}{j}\}}{n} > \frac{1}{j} \right\}$$
is measurable and so
$$\bigcap_{a \in \Gamma_n(x)} \Gamma_n(a) = \bigcap_n \Gamma_n(a_n) \quad (2.1)$$
is measurable. Looking at the set \([\lim \inf x, \lim \sup x] \setminus \Gamma_n(x)\), since \(\Gamma_n(x)\) is closed, it can be represented as an union of countably many intervals \(\{I_k\}\) that are mutually disjoint and open (or half open/closed). Further each \(I_k = \bigcup_{i=1}^{\infty} J_{k_i}\) where \(J_{k_i}\) are closed intervals and \(J_{k_1} \subseteq J_{k_2} \subseteq \ldots \subseteq J_{k_i} \subseteq \ldots \subseteq I_k\). It is not hard to show that for \(t \in (0,1)\):
$$x \notin \Gamma_n(a), \forall a \in [\lim \inf x, \lim \sup x] \setminus \Gamma_n(x) \text{ if and only if } u(\{j : (xt)_j \in J_{k_i}\}) = 0, \forall k, \forall i.$$Therefore
$$\bigcap_{a \in [\lim \inf x, \lim \sup x] \setminus \Gamma_n(x)} (0,1) \setminus \Gamma_n(a) = \bigcap_{k} \bigcap_{i} \left\{ t \in (0,1) : u(\{j : (xt)_j \in J_{k_i}\}) = 0 \right\}. \quad (2.2)$$
Now
$$\left\{ t \in (0,1) : u(\{j : (xt)_j \in J_{k_i}\}) = 0 \right\} = \bigcap_{l} \bigcap_{N \geq N} \bigcap_{m} \left\{ t \in (0,1) : \frac{\{m + 1 \leq j \leq m + n : (xt)_j \in J_{k_i}\}}{n} < \frac{1}{l} \right\}$$
and so it is measurable and thus the set in (2.2) is measurable. From (2.1) and (2.2) finally we get that \(T\) is measurable.

Now \(T\) can have measure 1 since if \(x\) is a convergent sequence, then \(T = [0,1]\) and so it has measure 1. Moreover, let \(x\) be given by:
$$0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1; \ldots$$
with \(n_1, n_2, n_3, \ldots\) as in part b) of the proof of Theorem 2.3. in [18]. Then \(x\) is almost convergent and hence from Lemma 1 of [21] uniformly statistically convergent to 0 and from the proof of Theorem 1 of [21], the set
$$\left\{ t \in (0,1) : (xt) \text{ converges uniformly statistically to } 0 \right\}$$
has measure 0. However, that set is the same as the set \(T\), and so for this sequence \(m(T) = 0\). Therefore \(m(T) = 0\) or 1, and both occur. \(\square\)

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References


