

## A-Statistical Supremum-Infimum and A-Statistical Convergence

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**Abstract.** In this paper, the concept of A-statistical supremum ( $\sup_A x$ ) and A-statistical infimum ( $\inf_A x$ ) for real valued sequences  $x = (x_n)$  are defined and studied. It is mainly shown that, the equality of  $\sup_A x$  and  $\inf_A x$  is necessary but not sufficient for to existence of usual limit of the sequence. On the other hand, the equality of  $\sup_A x$  and  $\inf_A x$  is necessary and sufficient for to existence of A-statistical limit of the real valued sequences.

**Key Words and Phrases:** statistical supremum, statistical infimum, statistical convergence, upper (lower) peak point

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The concept of statistical convergence is introduced by Fast and Steinhaus in [7] and [14], respectively. The idea of this concept based on asymptotic density of the subset  $K$  of natural numbers  $\mathbb{N}$  (see [3]).

Over the years, by using asymptotic density some concepts in mathematics are generalized.

Let  $K$  be a subset of natural numbers  $\mathbb{N}$  and

$$K(n) := \{k : k \leq n, k \in \mathbb{K}\}.$$

Then, the asymptotic density of  $K \subseteq \mathbb{N}$  is defined by

$$\delta(K) := \lim_{n \rightarrow \infty} \frac{1}{n} |K(n)|$$

if the limit exists. The symbol  $|K(n)|$  indicates the cardinality of the set  $K(n)$ .

A real or complex valued sequence  $x = (x_n)$  is said to be statistically convergent to the number  $L$ , if for every  $\varepsilon > 0$ , the set

$$K(n, \varepsilon) := \{k : k \leq n, |x_k - L| \geq \varepsilon\}$$

has zero asymptotic density, i. e.,

$$\delta(K(n, \varepsilon)) := \lim_{n \rightarrow \infty} \frac{1}{n} |K(n, \varepsilon)| = 0,$$

and it is denoted by  $x_n \rightarrow L(S)$ .

Let  $A = (a_{n,k})$  be a non-negative matrix. If  $A = (a_{n,k})$  transforms all convergent sequences to convergent sequences with the same limit, then it is called regular matrix transformation. The theorem in (1.3.3 Theorem [13]) gives the conditions for a matrix to be regular:

- (a) There exists a constant  $K$  such that  $\sum_{k=1}^{\infty} |a_{n,k}| < K$  for all  $n$ ,
- (b) For every  $k$ ,  $\lim_{n \rightarrow \infty} a_{nk} = 0$ ,
- (c)  $\lim_{n \rightarrow \infty} \sum_{k=1}^{\infty} a_{n,k} = 1$ .

A-density of the set  $K \subset \mathbb{N}$  is defined by

$$\delta_A(K) := \lim_{n \rightarrow \infty} \sum_{k \in K} a_{n,k}$$

if the limit exists.

The sequence  $x = (x_n)$  is A-statistical convergent to  $L \in \mathbb{R}$ , if for every  $\varepsilon > 0$  the set  $K(n, \varepsilon) := \{k : k \leq n, |x_k - L| \geq \varepsilon\}$  has A-density zero. It is denoted by  $x_n \rightarrow L(A-st)$ .

By using A-density, matrix characterization of statistical convergence has been given in [8]. After this study, some concepts in classical analysis has been generalized [1, 2, 4, 5, 6, 9, 10, 12], etc.

In this study, A-statistical lower and upper bound of real valued sequences will be defined. By using this concept, A-statistical supremum and A-statistical infimum will be investigated and mainly their relations between A-statistical convergence will be given.

**Definition 1.** (*A-Statistical Lower Bound*) The point  $l \in \mathbb{R}$  is called A- statistical lower bound of the sequence  $x = (x_n)$ , if the following

$$\delta_A(\{k : x_k \geq l\}) = 1 \quad (\text{or} \quad \delta_A(\{k : x_k < l\}) = 0) \quad (1)$$

hold.

The set of A-statistical lower bound of the sequence  $x = (x_n)$  is denoted by  $L_A(x)$ :

$$L_A(x) := \{l \in \mathbb{R} : l \text{ satisfies (1)}\}$$

Let us denote the set of usual lower bound of the sequence  $x = (x_n)$  by  $L(x)$ :

$$L(x) := \{l \in \mathbb{R} : l \leq x_n \text{ for all } n \in \mathbb{N}\}.$$

From the above definition we have following simple result:

**Theorem 1.** If  $l \in \mathbb{R}$  is an usual lower bound of the sequence  $x = (x_n)$ , then  $l \in \mathbb{R}$  is an A-statistical lower bound of the sequence  $x = (x_n)$ .

*Proof.* Let us assume  $l \in \mathbb{R}$  is a lower bound of the sequence  $x = (x_n)$ . From the definition of usual lower bound we have  $l \leq x_n$  for all  $n \in \mathbb{N}$ . So, the set

$$\{k : x_k \geq l\} = \mathbb{N}.$$

Therefore,

$$\delta_A(\{k : x_k \geq l\}) = 1$$

holds. This shows that every usual lower bound is an A-statistical lower bound,

i. e,  $(L(x) \subseteq L_A(x))$ . ◀

**Remark 1.** *The inverse of Theorem 1 is not true in general.*

Let us consider the sequence  $x = (x_n) = (-\frac{1}{n})$  and  $l = -\frac{1}{2} \in \mathbb{R}$ . If we choose a regular matrix as

$$a_{nk} = \begin{cases} \frac{2k}{n^2}, & 1 \leq k \leq n, \\ 0, & k > n. \end{cases}$$

It is clear that  $l = -\frac{1}{2}$  is an A-statistical lower bound but it is not usual lower bound for the sequence  $x = (x_n) = (-\frac{1}{n})$ .

**Definition 2.** (*A-statistical Upper Bound*) *The point  $m \in \mathbb{R}$  is called A-statistical upper bound of the sequence  $x = (x_n)$ , if the following*

$$\delta_A(\{k : x_k \leq m\}) = 1 \text{ (or } \delta_A(\{k : x_k > m\}) = 0) \quad (2)$$

hold.

The set of A-statistical upper bound of the sequence  $x = (x_n)$  is denoted by  $U_A(x)$ :

$$U_A(x) := \{m \in \mathbb{R} : m \text{ satisfies (2)}\}.$$

Let us denote the set of usual upper bound of the sequence  $x = (x_n)$  by  $U(x)$ :

$$U(x) := \{m \in \mathbb{R} : x_n \leq m \text{ for all } n \in \mathbb{N}\}.$$

From the above definition we have following simple result:

**Theorem 2.** *If  $m \in \mathbb{R}$  is an usual upper bound of the sequence  $x = (x_n)$ , then it is an A-statistical upper bound of the sequence  $x = (x_n)$ .*

*Proof.* Let us assume  $m \in \mathbb{R}$  is an usual upper bound of the sequence  $x = (x_n)$ . From the definition of usual upper bound we have  $x_n \leq m$  for all  $n \in \mathbb{N}$ . So, the set

$$\{k : x_k \leq m\} = \mathbb{N}.$$

Therefore,

$$\delta_A(\{k : x_k \leq m\}) = 1$$

holds. This shows that every usual upper bound is an A-statistical upper bound, i. e,  $U(x) \subset U_A(x)$ . ◀

**Remark 2.** *The inverse of Theorem 2 is not true in general.*

Let us consider the sequence  $x = (x_n) = (\frac{1}{n})$  and  $l = \frac{1}{2} \in \mathbb{R}$ . If we choose

$$a_{nk} = \begin{cases} \frac{2k-1}{n^2}, & 1 \leq k \leq n, \\ 0, & k > n. \end{cases}$$

regular matrix. It is clear that  $l = \frac{1}{2}$  is an A-statistical upper bound but it is not usual upper bound for the sequence  $x = (x_n) = (\frac{1}{n})$ .

**Corollary 1.** *If  $l \in \mathbb{R}$  is an A-statistical lower bound of the sequence  $x = (x_n)$  and  $l' < l$ , then  $l' \in \mathbb{R}$  is also A-statistical lower bound of the sequence  $x = (x_n)$ .*

*Proof.* Assume that  $l \in \mathbb{R}$  is an A-statistical lower bound of the sequence  $x = (x_n)$ . Then,

$$\delta_A(\{k : x_k \geq l\}) = 1.$$

Since  $l' < l$ , then the inclusion

$$\{k : x_k \geq l\} \subset \{k : x_k \geq l'\}$$

holds. So, we have

$$1 \leq \delta_A(\{k : x_k \geq l'\}).$$

Therefore,

$$\delta_A(\{k : x_k \geq l'\}) = 1.$$

So,  $l' \in \mathbb{R}$  is an A-statistical lower bound of the sequence  $x = (x_n)$ . ◀

**Corollary 2.** *If  $m \in \mathbb{R}$  is an A-statistical upper bound of the sequence  $x = (x_n)$  and  $m < m'$ , then  $m' \in \mathbb{R}$  is also A-statistical upper bound of the sequence  $x = (x_n)$ .*

*Proof.* Assume that  $m \in \mathbb{R}$  is an A-statistical upper bound of the sequence  $x = (x_n)$ . Then,

$$\delta_A(\{k : x_k \leq m\}) = 1.$$

Since  $m < m'$ , then the inclusion

$$\{k : x_k \leq m\} \subset \{k : x_k \leq m'\}$$

holds. So, we have

$$1 \leq \delta_A(\{k : x_k \leq m'\}).$$

Therefore,

$$\delta_A(\{k : x_k \leq m'\}) = 1.$$

So,  $m' \in \mathbb{R}$  is an A-statistical upper bound of the sequence  $x = (x_n)$ . ◀

**Remark 3.** *If the sequence  $x = (x_n)$  has an A-statistical lower (upper) bound, then it has infinitely many A-statistical lower (upper) bounds.*

**Definition 3.** (*A-Statistical Infimum* ( $\inf_A$ )) A number  $s \in \mathbb{R}$  is called *A-statistical infimum* of the sequence  $x = (x_n)$  if  $s \in \mathbb{R}$  is supremum of  $L_A(x)$ . That is

$$\inf_A x := \sup L_A(x).$$

**Definition 4.** (*A-Statistical Supremum* ( $\sup_A$ )) A number  $s' \in \mathbb{R}$  is called *A-statistical supremum* of the sequence  $x = (x_n)$  if  $s' \in \mathbb{R}$  is infimum of  $U_A(x)$ . That is

$$\sup_A x := \inf U_A(x).$$

**Theorem 3.** Let  $x = (x_n)$  be a sequence of real numbers. Then,

$$\inf x_n \leq \inf_A x_n \leq \sup_A x_n \leq \sup x_n$$

hold.

*Proof.* From the definition of usual infimum we have

$$\delta_A(\{k : \inf x_n \leq x_k\}) = \delta_A(\mathbb{N}) = 1.$$

This gives  $\inf x_n \in L_A(x)$ . Since  $\inf_A x = \sup L_A(x)$ , then  $\inf_A x \geq \inf x_n$  hold. From the definition of usual supremum we have

$$\delta_A(\{k : \sup x_n \geq x_k\}) = \delta_A(\mathbb{N}) = 1.$$

This gives  $\sup x_n \in U_A(x)$ . Since  $\sup_A x = \inf U_A(x)$ , then  $\sup_A x \leq \sup x_n$  hold. For to completion of the proof it is enough to show that the inequality

$$l \leq m \tag{3}$$

holds for any  $l \in L_A(x)$  and  $m \in U_A(x)$ .

Let us assume (3) is not true. That is there exist a  $l' \in L_A(x)$  and  $m' \in U_A(x)$  such that  $m' < l'$  is satisfied. Since  $m'$  is an A-statistical upper bound, then from Corollary 1 (II)  $l'$  is also A-statistical upper bound of the sequence. This is the contradiction to the assumption of  $l'$ . Therefore,  $l \leq m$  hold. ◀

**Remark 4.** Let  $A = (a_{nk})$  be a non-negative regular matrix.

a) If  $x = (x_n)$  is a constant sequence then,

$$\inf x_n = \inf_A x_n = \sup_A x_n = \sup x_n.$$

b) If we consider the sequence  $x = (x_n)$  as

$$x_n = \begin{cases} x_n, & n \leq n_0, n_0 \in \mathbb{N} \text{ fixed} \\ a, & n > n_0 \end{cases}$$

such that  $x_n \leq a$  for all  $n \in \{1, 2, 3, \dots, n_0\}$ , then

$$\inf x_n \leq \inf_A x_n \leq \sup_A x_n = \sup x_n.$$

c) If we consider the sequence  $x = (x_n)$  as

$$x_n = \begin{cases} x_n, & n \leq n_0, n_0 \in \mathbb{N} \text{ fixed} \\ a, & n > n_0 \end{cases}$$

such that  $x_n \geq a$  for all  $n \in \{1, 2, 3, \dots, n_0\}$ , then

$$\inf x_n = \inf_A x_n \leq \sup_A x_n \leq \sup x_n.$$

**Theorem 4.** Let  $x = (x_n)$  be a real valued sequence and  $A = (a_{nk})$  be a regular matrix. Then,

$$\delta_A(\{k : x_k \notin [\inf_A x_n, \sup_A x_n]\}) = 0$$

and

$$\delta_A(\{k : x_k \in [\inf_A x_n, \sup_A x_n]\}) = 1$$

hold.

*Proof.* Let us assume for simplicity  $\inf_A x_n = l$  and  $\sup_A x_n = m$ . That is  $l = \sup L_A(x)$  and  $m = \inf U_A(x)$ . From the definition of infimum and supremum we have  $l - \varepsilon \in L_A(x)$ ,  $m + \varepsilon \in U_A(x)$  and

$$[l, m] \subset [l - \varepsilon, m + \varepsilon] \quad (4)$$

It is clear from (4) that we have

$$\begin{aligned} \delta_A(\{k : x_k \notin [l, m]\}) &\leq \delta_A(\{k : x_k \notin [l - \varepsilon, m + \varepsilon]\}) = \\ &= \delta_A(\{k : x_k < l - \varepsilon\}) + \delta_A(\{k : x_k > m + \varepsilon\}) \end{aligned} \quad (5)$$

Since  $\delta_A(\{k : x_k < l - \varepsilon\}) = 0$  and  $\delta_A(\{k : x_k > m + \varepsilon\}) = 0$ , then from (5) we have

$$\delta_A(\{k : x_k \notin [\inf_A x_n, \sup_A x_n]\}) = 0.$$

It is clear that the following equality

$$\{k : x_k \in [\inf_A x_n, \sup_A x_n]\} = \mathbb{N} - \{k : x_k \notin [\inf_A x_n, \sup_A x_n]\}$$

hold and we have

$$\delta_A(\{k : x_k \in [\inf_A x_n, \sup_A x_n]\}) = \delta_A(\mathbb{N}) - \delta_A(\{k : x_k \notin [\inf_A x_n, \sup_A x_n]\}).$$

This gives the desired result. ◀

**Theorem 5.** *If  $\lim_{n \rightarrow \infty} x_n = l$ , then  $\sup_A x_n = \inf_A x_n = l$ .*

*Proof.* Assume  $\lim_{n \rightarrow \infty} x_n = l$ , i.e.,

For every  $\varepsilon > 0$ , there exist  $n_0 = n_0(\varepsilon) \in \mathbb{N}$  such that

$$|x_n - l| < \varepsilon, \quad (6)$$

hold for all  $n \geq n_0$ . Therefore, the following inclusion deduced from (6)

$$\mathbb{N} - \{1, 2, 3, \dots, n_0\} \subset \{k : x_k \geq l - \varepsilon\}, \quad (7)$$

$$\mathbb{N} - \{1, 2, 3, \dots, n_0\} \subset \{k : x_k \leq l + \varepsilon\}. \quad (8)$$

By using (7) and (8) we obtain

$$\delta_A(\{k : x_k \geq l - \varepsilon\}) = 1,$$

and

$$\delta_A(\{k : x_k \leq l + \varepsilon\}) = 1.$$

This discussion gives

$$l - \varepsilon \in L_A(x), \quad l + \varepsilon \in U_A(x)$$

for all  $\varepsilon > 0$  such that

$$L_A(x) = (-\infty, l) \text{ and } U_A(x) = (l, \infty).$$

Therefore,

$$\inf_A x_n = \sup(-\infty, l) = l = \inf(l, \infty) = \sup_A x_n$$

is obtained. ◀

**Remark 5.** *Theorem 5 stays true when  $l = \mp\infty$ .*

*Proof.* We shall give the proof only for  $l = +\infty$ . Lets take arbitrary  $M \in \mathbb{R}$ . From the assumption,  $\exists n_0 \equiv n_0(M) \in \mathbb{N}$  such that  $x_n > M$  for all  $n > n_0$ . So,

$$\delta_A(\{k : x_k \geq M\}) \geq \delta_A(\mathbb{N} - \{1, 2, \dots, n_0\}) = 1$$

and

$$\delta_A(\{k : x_k \leq M\}) \leq \delta_A(\{1, 2, \dots, n_0\}) = 0.$$

Therefore,  $M \in L_A(x)$ ,  $M \notin U_A(x)$ , i.e.  $L_A(x) = (-\infty, +\infty)$  and  $U_A(x) = \emptyset$ .

Hence,  $\inf_A x = \sup L_A(x) = \infty$ ,  $\sup_A x = \inf U_A(x) = \inf_A \emptyset = \infty$ . ◀

The following Corollary is a simple consequence of Theorem 5. So, the proof is omitted here.

**Corollary 3.** *Let  $x = (x_n)$  be a real valued sequence. The following statements are true.*

I) *If the sequence  $x = (x_n)$  is monotone increasing, then  $\inf_A x_n = \sup x_n$ .*

II) *If the sequence  $x = (x_n)$  is monotone decreasing, then  $\sup_A x_n = \inf x_n$ .*

**Remark 6.** Let  $A = (a_{nk})$  be a non-negative regular matrix.

a) If  $x = (x_n)$  is monotone increasing, then

$$\inf x_n \leq \inf_A x_n = \sup_A x_n = \sup x_n.$$

b) If  $x = (x_n)$  is monotone decreasing, then

$$\inf x_n = \inf_A x_n = \sup_A x_n \leq \sup x_n.$$

**Remark 7.** The inverse of Theorem 5 is not true.

For to see this let us consider the sequence  $x = (x_n)$  as

$$x_n = \begin{cases} 1, & n = k^2, k = 1, 2, \dots, \\ 0, & \text{otherwise} \end{cases}$$

and the matrix  $a = (a_{nk})$  as

$$a_{nk} = \begin{cases} \frac{2k}{n^2}, & 1 \leq k \leq n, \\ 0, & k > n. \end{cases}$$

It is clear that  $\sup_A x_n = \inf_A x_n = 0$  but the sequence is not convergent to 0.

**Theorem 6.**  $\lim_{n \rightarrow \infty} x_n = l(A - st)$  if and only if  $\sup_A x_n = \inf_A x_n = l$ .

*Proof.* "  $\Rightarrow$  " Assume that  $\lim_{n \rightarrow \infty} x_n = l(A - st)$ . We have for every  $\varepsilon > 0$ ,

$$\delta_A(\{k : k \leq n, |x_k - l| \geq \varepsilon\}) = 0 \quad (9)$$

hold. Since,

$$\{k : k \leq n, |x_k - l| \geq \varepsilon\} = \{k : k \leq n, x_k \geq l + \varepsilon\} \cup \{k : k \leq n, x_k \leq l - \varepsilon\}$$

and from (9) we have

$$\delta_A(\{k : x_k \geq l + \varepsilon\}) = 0 \quad (10)$$

and

$$\delta_A(\{k : x_k \leq l - \varepsilon\}) = 0. \quad (11)$$

Also, from (9) we have

$$\delta_A(\{k : k \leq n, |x_k - l| < \varepsilon\}) = 1 \quad (12)$$

and (12) gives

$$\delta_A(\{k : x_k < l + \varepsilon\}) = 1 \quad (13)$$

and

$$\delta_A(\{k : x_k > l - \varepsilon\}) = 1. \quad (14)$$



The equation (10) and (13) gives  $l + \varepsilon$  is an A-statistical upper bound and (11) and (14) gives  $l - \varepsilon$  is an A-statistical lower bound for the sequence.

So,

$$L_A(x) = (-\infty, l) \text{ and } U_A(x) = (l, \infty).$$

Therefore, we have

$$\sup L_A(x) = l, \quad \inf U_A(x) = l.$$

" $\Leftarrow$ " Assume that

$$\sup_A x_n = \inf_A x_n = l.$$

That is

$$l = \sup L_A(x) = \inf U_A(x).$$

From the definition of supremum and infimum, there exists at least one element  $l' \in L_A(x)$  and  $l'' \in U_A(x)$  for all  $\varepsilon > 0$  such that the inequality

$$l - \varepsilon < l' \text{ and } l'' < l + \varepsilon$$

hold. Since  $l'$  is an A-statistical lower bound then we have following inclusion

$$\{k : x_k \geq l + \varepsilon\} \subset \{k : x_k \geq l'\}.$$

So,

$$\delta_A(\{k : x_k \geq l + \varepsilon\}) = 0 \tag{15}$$

Since  $l''$  is an A-statistical upper bound then we have following inclusion

$$\{k : x_k \leq l - \varepsilon\} \subset \{k : x_k \leq l''\}.$$

So,

$$\delta_A(\{k : x_k \leq l - \varepsilon\}) = 0. \tag{16}$$

From the equations (15), (16) and

$$\{k : |x_k - l| \geq \varepsilon\} = \{k : x_k \geq l + \varepsilon\} \cup \{k : x_k \leq l - \varepsilon\},$$

we have

$$\delta_A(\{k : |x_k - l| \geq \varepsilon\}) = 0.$$

Therefore, the sequence  $x = (x_n)$  is A-statistical convergent to  $l \in \mathbb{R}$ .  $\blacktriangleleft$

**Definition 5.** The real valued sequences  $x = (x_n)$  and  $y = (y_n)$  are called A-statistical equivalent if the A-density of the set  $H = \{k : x_k \neq y_k\}$  is zero. It is denoted by  $x \asymp y$ .

**Theorem 7.** If the sequence  $x = (x_n)$  and  $y = (y_n)$  are equivalent, then

$$\inf_A x_n = \inf_A y_n \text{ and } \sup_A x_n = \sup_A y_n.$$

*Proof.* Since the sequence  $x = (x_n)$  and  $y = (y_n)$  are equivalent, then the set  $H = \{k : x_k \neq y_k\}$  has zero A-density. Let us consider an arbitrary element  $l \in L_A(x)$ . The element  $l \in \mathbb{R}$  is an A-statistical lower bound of the sequence  $x = (x_n)$ , then we have

$$\delta_A(\{k : x_k < l\}) = 0 \text{ and } \delta_A(\{k : x_k \geq l\}) = 1.$$

From the following inclusion

$$\{k : y_k < l\} = \{k : x_k \neq y_k < l\} \cup \{k : x_k = y_k < l\} \subset H \cup \{k : x_k = y_k < l\}$$

we have

$$\begin{aligned} 0 \leq \delta_A(\{k : y_k < l\}) &= \delta_A(\{k : x_k \neq y_k < l\}) + \delta_A(\{k : x_k = y_k < l\}) \\ &\leq \delta_A(H) + \delta_A(\{k : x_k = y_k < l\}) = 0 + 0 = 0. \end{aligned} \quad (17)$$

Since the inclusion

$$\begin{aligned} \{k : y_k \geq l\} &= \{k : x_k \neq y_k \geq l\} \cup \{k : x_k = y_k \geq l\} \\ &\supset \{k : x_k = y_k < l\} \end{aligned}$$

then we have

$$1 = \delta_A(\{k : y_k \geq l\}) \geq \delta_A(\{k : x_k = y_k \geq l\}) = 1. \quad (18)$$

From (17) and (18), the element  $l \in \mathbb{R}$  is an A-statistical lower bound of the sequence  $y = (y_n)$ . That is,  $L_A(x) \subset L_A(y)$ .

If we consider an arbitrary element  $l \in L_A(y)$ , it can be easily obtained that  $l \in L_A(x)$ . Then  $L_A(y) \subset L_A(x)$ . Therefore,

$$L_A(y) = L_A(x)$$

hold. Since  $\sup L_A(y) = \sup L_A(x)$ , then  $\inf_A x = \inf_A y$  is obtained.

By using the same argument as above it can be obtained  $\sup_A x = \sup_A y$ . ◀

**Definition 6.** (*Upper or Lower Peak Point*) [11] *The point  $x_k$  is called upper( or lower) peak point of the sequence  $x = (x_n)$  if the inequality  $x_k \geq x_l$  (or  $x_k \leq x_l$ ) holds for all  $l \geq k$ .*

**Theorem 8.** *Let  $x = (x_n)$  be a real valued sequence. If the element  $x_{n_0}$  is an upper(or lower) peak point of  $(x_n)$ , then the element  $x_{n_0}$  is an A-statistical upper (or A-statistical lower) bound.*

*Proof.* Assume the point  $x_{n_0}$  is an upper peak point of the sequence  $x = (x_n)$  that  $x_k \leq x_{n_0}$  holds for all  $k \geq n_0$ . So, the following inclusion

$$\{k : x_k \leq x_{n_0}\} \supset \mathbb{N} - \{1, 2, \dots, n_0\}$$

holds. From this inclusion and the properties of asymptotic density we have

$$1 \leq \delta_A(\{k : x_k \leq x_{n_0}\}) = 1.$$

This give us the point  $x_{n_0}$  is an A-statistical upper bound of the sequence  $x = (x_n)$ . ◀

**Theorem 9.** Let  $x = (x_n)$  be a real valued sequence and  $A = (a_{nk})$ ,  $B = (b_{nk})$  be regular matrix. If the condition

$$\limsup_{n \rightarrow \infty} \sum_{k=1}^{\infty} |a_{nk} - b_{nk}| = 0$$

hold. Then

$$\inf_A x_n = \inf_B x_n \text{ and } \sup_A x_n = \sup_B x_n.$$

*Proof.* Let  $K_1 = \{k : x_k < m\}$ ,  $K_2 = \{k : x_k \geq m\}$  be subsets of natural numbers  $\mathbb{N}$  for all  $m \in \mathbb{R}$ . For  $K = K_1$  (or  $K_2$ )

$$\begin{aligned} |\delta_A(K) - \delta_B(K)| &= \left| \lim_{n \rightarrow \infty} \sum_{k \in K} a_{nk} - \sum_{k \in K} b_{nk} \right| \\ &\leq \lim_{n \rightarrow \infty} \sum_{k \in K} |a_{nk} - b_{nk}| \\ &\leq \sum_{k=1}^{\infty} |a_{nk} - b_{nk}| \end{aligned}$$

hold. Namely  $\delta_A(K) = \delta_B(K)$ . So  $\inf_A x_n = \inf_B x_n$ .

**Remark 8.** The inverse of Theorem 9 is not true.

For to see this let us consider the sequence  $x = (x_n)$  where

$$x_n = \begin{cases} 1, & n = k^2, k = 1, 2, \dots, \\ 0, & \text{otherwise,} \end{cases}$$

and the matrices  $A = (a_{n,k})$  and  $B = (b_{n,k})$  as

$$a_{nk} = \begin{cases} \frac{n}{3(n+1)}, & k = n^2, \\ 1 - \frac{n}{3(n+1)}, & k = n^2 + 1, \\ 0, & \text{otherwise,} \end{cases}$$

and

$$b_{nk} = \begin{cases} \frac{n}{5(n+1)}, & k = n^2, \\ 1 - \frac{n}{5(n+1)}, & k = n^2 + 1, \\ 0, & \text{otherwise.} \end{cases}$$

The matrices  $A, B$  are non-negative and regular. This sequence and matrices  $A$  and  $B$  has been considered in [4]. It is clear that  $L_A(x) = (-\infty, 0]$ ,  $L_B(x) = (-\infty, 0]$ ,  $U_A(x) = (0, \infty)$  and  $U_B(x) = (0, \infty)$ . Therefore,

$$\sup L_A(x) = \sup L_B(x) = 0$$

and

$$\inf U_A(x) = \inf U_B(x) = 0.$$

That is,  $\sup_A x = \sup_B x$ ,  $\inf_A x = \inf_B x$  hold. Unfortunately, the condition given theorem doesn't hold for the matrices  $A$  and  $B$ . The other case is obtained by similar way. ◀

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