# On the consistency of a spatial-type interval-valued median for random intervals

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### Abstract

The  $d_{\theta}$ -median is a much more robust estimator of the location of a random interval than the mean. We show that under general conditions the sample  $d_{\theta}$ -median is a strongly consistent estimator of the  $d_{\theta}$ -median.

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## 1. Introduction

In this data driven era, the amount and complexity of the available data grows at an almost incredible speed. Therefore, there is a high need to develop novel tools to cope with complex data structures, such as incomplete/missing data, functional data, interval valued or fuzzy data, and several other types of data.

Interval-valued data may arise for different reasons. The data may come from intrinsically interval-valued random elements (e.g. the daily fluctuation of the systolic blood pressure) or from random elements derived from an underlying real-valued random variable to preserve a level of confidentiality

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(e.g. when indicating an interval containing the real salary) or due to measurement imprecision (e.g. censored data), or from aggregation of a typically large dataset, etc. In this work no assumption about the source of the data is needed.

Interval-valued data are a type of complex data that requires specific statistical techniques. The main issue is that the space of intervals is only semilinear, but not linear due to the lack of the opposite of an interval. Therefore, although intervals can be identified with two-dimensional vectors (with first component the mid-point/centre and second component the non-negative spread/radius), it is not advisable to treat them as regular bivariate data. Indeed, common assumptions for multivariate techniques do not hold in this case.

Statistical procedures for random interval-valued data have already been proposed in the literature for different purposes, such as regression analysis (e.g. Gil *et al.*, 2002); testing hypotheses (e.g. González-Rodríguez *et al.*, 2012), clustering (e.g. D'Urso and Giordani *et al.*, 2006), principal component analysis (e.g. D'Urso and Giordani, 2004) and modelling distributions (see Brito and Duarte Silva, 2012).

A common location measure in this setting is the Aumann mean (Aumann, 1965). It is supported by several valuable properties and is also coherent with the interval arithmetic. The main disadvantage is that it is strongly influenced by outliers or small data changes, which can make it unreliable as a measure of the location of a random interval. In fact, it inherits this drawback from the standard real/vectorial-valued case.

In the real case, the most popular robust alternative is the median. For multivariate data the spatial median (or  $L_1$ -median, as introduced by Weber 1909) is a popular robust alternative. It is defined as the point in multivariate space with minimal average Euclidean distance to the observations. For more details and extensions, see for instance Gower (1974), Brown (1983), Milasevic and Ducharme (1987), Zuo (2013).

Sinova and Van Aelst (2014) adapted the spatial median to interval-valued data (see also Sinova *et al.* 2013), by using on this space as  $L^2$  distance the versatile generalized metric introduced by Bertoluzza *et al.* (1995). The resulting  $d_{\theta}$ -median estimator has been shown to be robust with high break-down point and good finite-sample properties. In this paper we show another important property of the estimator, which is its strong consistency.

#### 2. The $d_{\theta}$ -median of a random interval

Let  $\mathcal{K}_c(\mathbb{R})$  denote the class of nonempty compact intervals. Any interval K in the space  $K_c(\mathbb{R})$  can be characterized in terms of either its infimum and supremum,  $K = [\inf K, \sup K]$ , or its mid-point and spread or radius,  $K = [\min K - \operatorname{spr} K, \min K + \operatorname{spr} K]$ , where

$$\operatorname{mid} K = \frac{\inf K + \sup K}{2}, \quad \operatorname{spr} K = \frac{\sup K - \inf K}{2} \ge 0$$

The usual interval arithmetic provides the addition, i.e.  $K + K' = [\inf K + \inf K', \sup K + \sup K']$  with  $K, K' \in \mathcal{K}_c(\mathbb{R})$  and the product by a scalar, i.e.  $\gamma \cdot K = [\gamma \cdot \min K - |\gamma| \cdot \operatorname{spr} K, \gamma \cdot \min K + |\gamma| \cdot \operatorname{spr} K]$  with  $K \in \mathcal{K}_c(\mathbb{R})$  and  $\gamma \in \mathbb{R}$ . With these two operations the space  $\mathcal{K}_c(\mathbb{R})$  is only semilinear, so statistical techniques for interval-valued data will be based on distances.

The  $d_{\theta}$  metric (Bertoluzza et al., 1995, Gil et al., 2002) can be defined as:

$$d_{\theta}(K, K') = \sqrt{(\operatorname{mid} K - \operatorname{mid} K')^2 + \theta \cdot (\operatorname{spr} K - \operatorname{spr} K')^2},$$

where  $K, K' \in \mathcal{K}_c(\mathbb{R})$  and  $\theta \in (0, \infty)$ . A random interval can be defined as a Borel measurable mapping  $X : \Omega \to \mathcal{K}_c(\mathbb{R})$ , where  $(\Omega, \mathcal{A}, P)$  is a probability space, with respect to  $\mathcal{A}$  and the Borel  $\sigma$ -field generated by the topology induced on  $\mathcal{K}_c(\mathbb{R})$  by the  $d_{\theta}$  metric.

The well-known Aumann mean value is the interval, if it exists, given by

$$E[X] = [E(\operatorname{mid} X) - E(\operatorname{spr} X), E(\operatorname{mid} X) + E(\operatorname{spr} X)].$$

Moreover, it is the Fréchet expectation with respect to the  $d_{\theta}$  metric, i.e., it is the unique interval that minimizes, over  $K \in \mathcal{K}_c(\mathbb{R})$ , the expression  $E[(d_{\theta}(X, K))^2].$ 

In Sinova and Van Aelst (2014) the  $d_{\theta}$ -median of a random interval X is defined as the interval(s)  $M_{\theta}[X] \in \mathcal{K}_{c}(\mathbb{R})$  such that

$$E(d_{\theta}(X, \mathcal{M}_{\theta}[X])) = \min_{K \in \mathcal{K}_{c}(\mathbb{R})} E(d_{\theta}(X, K)),$$

whenever the involved expectations exist. Analogously, the sample  $d_{\theta}$ -median statistic is defined as follows.

Let  $(X_1, \ldots, X_n)$  be iid random intervals associated with a probability space  $(\Omega, \mathcal{A}, P)$  and with realizations  $\mathbf{x}_n = (x_1, \ldots, x_n)$ . The sample  $d_{\theta}$ median (or medians)  $\widehat{M}_{\theta}[X]_n$  is (are) the random interval that for any  $\mathbf{x}_n$  is (are) the solution(s) of the following optimization problem:

$$\min_{K \in \mathcal{K}_c(\mathbb{R})} \frac{1}{n} \sum_{i=1}^n d_\theta(x_i, K) = \min_{(y, z) \in \mathbb{R} \times [0, \infty)} \frac{1}{n} \sum_{i=1}^n \sqrt{(\min x_i - y)^2 + \theta \cdot (\operatorname{spr} x_i - z)^2}$$

Sinova and Van Aelst (2014) showed the existence of the sample  $d_{\theta}$ -median estimator and its uniqueness whenever not all the two-dimensional sample points  $\{(\min x_i, \operatorname{spr} x_i)\}_{i=1}^n$  are collinear. Moreover, the robustness was shown by its finite sample breakdown point (Donoho and Huber, 1983), which is given by  $\frac{1}{n} \cdot \lfloor \frac{n+1}{2} \rfloor$ , where  $\lfloor \cdot \rfloor$  denotes the floor function.

#### 3. Consistency of the sample $d_{\theta}$ -median

In this section we investigate the strong consistency of the sample  $d_{\theta}$ median under general conditions.

**Theorem 1.** Let X be a random interval associated with a probability space  $(\Omega, \mathcal{A}, P)$  such that the  $d_{\theta}$ -median exists and is unique. Then, the sample  $d_{\theta}$ -median is a strongly consistent estimator of the  $d_{\theta}$ -median, that is,

$$\lim_{n \to \infty} d_{\theta}(\widehat{\mathcal{M}}_{\theta}[X]_n, M_{\theta}[X]) = 0 \quad a.s.[P].$$

*Proof.* Sufficient conditions for the strong consistency of an estimator are given in Huber (1967). We will check that these conditions, detailed below, are satisfied in our case and, hence, the theorem follows directly from Huber's general result.

First, note that the interval space  $(\mathbb{R} \times [0, \infty))$  with the topology induced by the  $d_{\theta}$ -metric) is a locally compact space with a countable base and  $(\Omega, \mathcal{A}, P)$ is a probability space.

Let  $\rho(\omega, (y, z))$  be the following real-valued function on  $\Omega \times (\mathbb{R} \times [0, \infty))$ :

$$\rho: \Omega \times (\mathbb{R} \times [0,\infty)) \longrightarrow \mathbb{R}$$
$$(\omega, (y,z)) \longmapsto d_{\theta}(X(\omega), [y-z, y+z]).$$

Assuming that  $\omega_1, \omega_2 \dots$  are independent  $\Omega$ -valued random elements with common probability distribution P, the sequence of functions  $\{T_n\}_{n\in\mathbb{N}}$ , defined as  $T_n(\omega_1, \dots, \omega_n) = \widehat{M_{\theta}}[(X(\omega_1), \dots, X(\omega_n))]_n$ , satisfies that  $\frac{1}{n} \sum_{i=1}^n d_{\theta}(X(\omega_i), T_n(\omega_1, \dots, \omega_n)) - \inf_{(y,z)\in\mathbb{R}\times[0,\infty)} \frac{1}{n} \sum_{i=1}^n d_{\theta}(X(\omega_i), [y-z, y+z]) \xrightarrow[n\to\infty]{} 0$ 

almost surely (obviously because of the definition of the sample  $d_{\theta}$ -median).

We now recall the sufficient conditions for the strong consistency in Huber (1967).

Condition (A-1). For each fixed  $(y_0, z_0) \in \mathbb{R} \times [0, \infty)$ , the function

$$\begin{array}{rccc} \rho_0: & \Omega & \longrightarrow & \mathbb{R} \\ & \omega & \longmapsto & \rho(\omega, (y_0, z_0)) & = d_{\theta}(X(\omega), [y_0 - z_0, y_0 + z_0]) \end{array}$$

is  $\mathcal{A}$ -measurable and separable in Doob's sense, i.e. there is a P-null set N and a countable subset  $S \subset \mathbb{R} \times [0, \infty)$  such that for every open set  $U \subset \mathbb{R} \times [0, \infty)$  and every closed interval A, the sets

$$V_1 = \{ \omega : \rho(\omega, (y, z)) \in A, \forall (y, z) \in U \}$$
$$V_2 = \{ \omega : \rho(\omega, (y, z)) \in A, \forall (y, z) \in U \cap S \}$$

differ by at most a subset of N.

Condition (A-2). The function  $\rho$  is a.s. lower semicontinuous in  $(y_0, z_0)$ , i.e.

$$\inf_{(y,z)\in U} \rho(\omega, (y,z)) \longrightarrow \rho(\omega, (y_0, z_0)), \tag{1}$$

as the neighborhood U of  $(y_0, z_0)$  shrinks to  $\{(y_0, z_0)\}$ .

Condition (A-3). There is a measurable function  $a: \Omega \to \mathbb{R}$  such that

$$E[\rho(\omega, (y, z)) - a(\omega)]^{-} < \infty \quad \text{for all } (y, z) \in \mathbb{R} \times [0, \infty),$$
$$E[\rho(\omega, (y, z)) - a(\omega)]^{+} < \infty \quad \text{for some } (y, z) \in \mathbb{R} \times [0, \infty).$$

Thus,  $\gamma((y, z)) = E[\rho(\omega, (y, z)) - a(\omega)]$  is well-defined for all (y, z). *Condition (A-4).* There is a  $(y_0, z_0) \in \mathbb{R} \times [0, \infty)$  such that  $\gamma((y, z))$  $> \gamma((y_0, z_0))$  for all  $(y, z) \neq (y_0, z_0)$ .

Condition (A-5). There is a continuous function b((y, z)) > 0 such that

- for some integrable h,  $\inf_{(y,z)\in\mathbb{R}\times[0,\infty)}\frac{\rho(\omega,(w,z))-a(\omega)}{b((y,z))}\geq h(\omega).$
- $\liminf_{(y,z)\to\infty} b((y,z)) > \gamma((y_0,z_0)).$

• 
$$E\left[\liminf_{(y,z)\to\infty}\frac{\rho(\omega,(y,z))-a(\omega)}{b((y,z))}\right] \ge 1.$$

We now verify these conditions.

(A-1) For each fixed  $(y_0, z_0) \in \mathbb{R} \times [0, \infty)$ , the function  $\rho_0$  is  $\mathcal{A}$ -measurable because mid X and spr X are measurable functions since X is a random interval.  $\rho_0$  is also separable in Doob's sense: choose  $S = \mathbb{Q} \times (\mathbb{Q} \cap [0, \infty))$ as countable subset, then for every open set  $U \subset \mathbb{R} \times [0, \infty)$  and every closed interval A, it can be seen that the following sets coincide.

$$V_1 = \{ \omega : \rho_0(\omega) \in A, \forall (y, z) \in U \}, \ V_2 = \{ \omega : \rho_0(\omega) \in A, \forall (y, z) \in U \cap S \}$$

Obviously,  $V_1 \subseteq V_2$ . By reductio ad absurdum, suppose that  $V_2 \cap V_1^c \neq \emptyset$ . Let  $\omega_0 \in V_2 \cap V_1^c$ :

- Since  $\omega_0 \in V_2$ ,  $\rho(\omega_0, (y, z)) \in A$  for all  $(y, z) \in U \cap S$ ;
- Since  $\omega_0 \in V_1^c$ , there exists  $(y_0, z_0) \in U$  such that  $\rho(\omega_0, (y_0, z_0)) \in A^c$ .  $A^c$  is an open set, so there exists a ball of radius r > 0 such that

$$(\rho(\omega_0, (y_0, z_0)) - r, \rho(\omega_0, (y_0, z_0)) + r) \subseteq A^c.$$

Now, notice that for a fixed  $\omega \in \Omega$ , the following function is continuous:

$$\begin{array}{rccc} \rho_{\omega}: & \mathbb{R} \times [0,\infty) & \longrightarrow & \mathbb{R} \\ & & (y,z) & \longmapsto & \rho(\omega,(y,z)), \end{array}$$

Thus,  $B = \rho_{\omega_0}^{-1}(\rho(\omega_0, (y_0, z_0)) - r, \rho(\omega_0, (y_0, z_0)) + r)$  is an open set of  $\mathbb{R} \times [0, \infty)$ and  $U \cap B \neq \emptyset$ . S is a dense set of  $\mathbb{R} \times [0, \infty)$ , so  $U \cap B \cap S \neq \emptyset$ . Let  $(y', z') \in U \cap B \cap S$ . Then,  $(y', z') \in U \cap S$ , so  $\rho(\omega_0, (y', z')) \in A$ . But also,

$$\rho(\omega_0, (y', z')) \in (\rho(\omega_0, (y_0, z_0)) - r, \rho(\omega_0, (y_0, z_0)) + r) \subset A^c.$$

This is a contradiction, so the conclusion is that also  $V_2 \subseteq V_1$ , which proofs this condition.

(A-2) Let  $\omega$  be any element of  $\Omega$  and let  $(y_0, z_0)$  be any (fixed) point of  $\mathbb{R} \times [0, \infty)$ . First, note that (1) is fulfilled for a sequence of neighborhoods  $\{U_n\}_{n\in\mathbb{N}}$  of  $(y_0, z_0)$  with  $U_n \supseteq U_{n+1}$  for all n. Since

$$\left\{\inf_{(y,z)\in U_n} d_{\theta}(X(\omega), [y-z, y+z])\right\}_{n\in\mathbb{N}}$$

is a monotonically increasing sequence and  $(y_0, z_0) \in \bigcap_{n \in \mathbb{N}} U_n$ , the sequence is bounded by  $d_{\theta}(X(\omega), [y_0 - z_0, y_0 + z_0])$  and, thus, it converges to its supremum. We now show that this supremum is indeed  $d_{\theta}(X(\omega), [y_0 - z_0, y_0 + z_0])$ .

By reductio ad absurdum, suppose there is a smaller upper bound c = $d_{\theta}(X(\omega), [y_0 - z_0, y_0 + z_0]) - \varepsilon$ , for an arbitrary  $\varepsilon > 0$ . Let  $U_{n_0}$  denote a neighborhood of  $(y_0, z_0)$  such that  $U_{n_0} \subseteq C = B((y_0, z_0), \frac{\varepsilon}{2})$ . It follows that  $c < \inf_{(y,z) \in U_{n_0}} d_{\theta}(X(\omega), [y-z, y+z])$ , so c cannot be the supremum. Indeed,

$$\inf_{(y,z)\in U_{n_0}} d_{\theta}(X(\omega), [y-z, y+z]) \ge \inf_{(y,z)\in C} d_{\theta}(X(\omega), [y-z, y+z])$$

$$\ge \inf_{(y,z)\in C} \left[ d_{\theta}(X(\omega), [y_0-z_0, y_0+z_0]) - d_{\theta}([y-z, y+z], [y_0-z_0, y_0+z_0]) \right]$$

$$= d_{\theta}(X(\omega), [y_0-z_0, y_0+z_0]) - \sup_{(y,z)\in C} d_{\theta}([y-z, y+z], [y_0-z_0, y_0+z_0])$$

$$> d_{\theta}(X(\omega), [y_0-z_0, y_0+z_0]) - \varepsilon = c.$$

Now, we extend this result to general sequences  $\{U_n\}_{n\in\mathbb{N}}$ . Consider the suprema and the infima radii reached in every neighborhood, namely,

$$r_n = \sup_{(y,z)\in U_n} d_\theta([y_0 - z_0, y_0 + z_0], [y - z, y + z]),$$
  
$$s_n = \inf_{(y,z)\in U_n} d_\theta([y_0 - z_0, y_0 + z_0], [y - z, y + z]).$$

It follows that  $r_n \xrightarrow[n \to \infty]{} 0$ , since  $\{U_n\}_{n \in \mathbb{N}}$  shrinks to  $\{(y_0, z_0)\}$ . Hence, also  $s_n \xrightarrow[n \to \infty]{} 0$  since  $0 \leq s_n \leq r_n$  for all  $n \in \mathbb{N}$ . Choose  $\varepsilon > 0$  arbitrarily. As  $r_n \xrightarrow[n \to \infty]{} 0$ , there exists  $n_1 \in \mathbb{N}$  such that for

all  $n > n_1, r_n < \varepsilon$ . Then,  $U_n \subseteq B((y_0, z_0), r_n)$  and

$$\inf_{(y,z)\in U_n} d_{\theta}(X(\omega), [y-z, y+z]) \ge \inf_{(y,z)\in B((y_0,z_0),r_n)} d_{\theta}(X(\omega), [y-z, y+z])$$

$$\geq d_{\theta}(X(\omega), [y_0 - z_0, y_0 + z_0]) - \sup_{(y,z) \in B((y_0, z_0), r_n)} d_{\theta}([y_0 - z_0, y_0 + z_0], [y - z, y + z])$$
  
 
$$> d_{\theta}(X(\omega), [y_0 - z_0, y_0 + z_0]) - \varepsilon.$$

Analogously, as  $s_n \xrightarrow[n \to \infty]{} 0$ , there exists  $n_2 \in \mathbb{N}$  such that for all  $n > n_2$ ,  $s_n < \varepsilon$ . Therefore,  $U_n \supseteq \widetilde{B}((y_0, z_0), s_n)$  and

$$\inf_{\substack{(y,z)\in U_n}} d_{\theta}(X(\omega), [y-z, y+z]) \leq \inf_{\substack{(y,z)\in B((y_0,z_0),s_n)}} d_{\theta}(X(\omega), [y-z, y+z]) \\
\leq d_{\theta}(X(\omega), [y_0-z_0, y_0+z_0]) + \inf_{\substack{(y,z)\in B((y_0,z_0),s_n)}} d_{\theta}([y-z, y+z], [y_0-z_0, y_0+z_0]) \\
< d_{\theta}(X(\omega), [y_0-z_0, y_0+z_0]) + \varepsilon.$$

So for any  $\varepsilon > 0$ , there exists  $n_0 = \max\{n_1, n_2\}$  s.t. for all  $n > n_0$ ,  $\left| \inf_{(y,z)\in U_n} d_{\theta}(X(\omega), [y-z, y+z]) - d_{\theta}(X(\omega), [y_0 - z_0, y_0 + z_0]) \right| < \varepsilon$ . That is, the considered sequence converges to  $d_{\theta}(X(\omega), [y_0 - z_0, y_0 + z_0])$ , which completes the proof of (1).

(A-3) Let a be the measurable function (see (A-1)):

$$a: \Omega \longrightarrow \mathbb{R}$$
  
$$\omega \longmapsto d_{\theta}(X(\omega), [0, 0]) = \sqrt{(\operatorname{mid} X(\omega))^{2} + \theta \cdot (\operatorname{spr} X(\omega))^{2}}.$$

For any  $(y, z) \in \mathbb{R} \times [0, \infty)$ , we then have that  $E[\rho(\omega, (y, z)) - a(\omega)]^{-1}$ 

$$= \int_{\Omega} -\min\{d_{\theta}(X(\omega), [y-z, y+z]) - d_{\theta}(X(\omega), [0,0]), 0\} dP(\omega)$$
  
= 
$$\int_{\substack{\{\omega \in \Omega : d_{\theta}(X(\omega), [0,0]) \\ > d_{\theta}(X(\omega), [y-z, y+z])\}}} \left[d_{\theta}(X(\omega), [0,0]) - d_{\theta}(X(\omega), [y-z, y+z])\right] dP(\omega)$$

By the triangular inequality,

$$\leq \int_{\substack{\{\omega \in \Omega : d_{\theta}(X(\omega), [0,0]) \\ > d_{\theta}(X(\omega), [y-z, y+z])\}}} \left[ d_{\theta}(X(\omega), [y-z, y+z]) + d_{\theta}([y-z, y+z], [0,0]) \right. \\ \left. - d_{\theta}(X(\omega), [y-z, y+z]) \right] dP(\omega)$$

$$= d_{\theta}([y-z, y+z], [0,0]) \cdot P\left(\omega : d_{\theta}(X(\omega), [0,0]) > d_{\theta}(X(\omega), [y-z, y+z])\right) < \infty.$$
Analogously,  $E[\rho(\omega, (y, z)) - a(\omega)]^+$ 

$$= \int_{\Omega} \max\{d_{\theta}(X(\omega), [y-z, y+z]) - d_{\theta}(X(\omega), [0, 0]), 0\} dP(\omega)$$

 $\leq d_{\theta}([0,0], [y-z, y+z]) \cdot P(\omega : d_{\theta}(X(\omega), [0,0]) \leq d_{\theta}(X(\omega), [y-z, y+z])) < \infty,$ for all  $(y,z) \in \mathbb{R} \times [0,\infty)$ . So, with this choice for the function a, both

inequalities in condition (A-3) hold.

(A-4) The  $d_{\theta}$ -median exists and is unique, so that

$$(\operatorname{mid} M_{\theta}[X], \operatorname{spr} M_{\theta}[X]) = \arg \min_{(y,z) \in \mathbb{R} \times [0,\infty)} E\left[d_{\theta}(X(\omega), [y-z, y+z])\right]$$
$$= \arg \min_{(y,z) \in \mathbb{R} \times [0,\infty)} E\left[d_{\theta}(X(\omega), [y-z, y+z])\right] - E\left[d_{\theta}(X(\omega), [0, 0])\right]$$

$$= \arg \min_{(y,z) \in \mathbb{R} \times [0,\infty)} \gamma((y,z)).$$

Thus,  $(y_0, z_0) := (\operatorname{mid} M_{\theta}[X], \operatorname{spr} M_{\theta}[X])$  fulfills this condition.

(A-5) Consider the continuous function b((y, z)) > 0 defined as

$$b: \mathbb{R} \times [0,\infty) \longrightarrow \mathbb{R}$$
  
(y,z) 
$$\longmapsto d_{\theta}([y-z,y+z],[0,0]) + 1.$$

• For the integrable function  $h(\omega) := -1$ , it then holds that

$$\inf_{(y,z)\in\mathbb{R}\times[0,\infty)}\frac{d_{\theta}(X(\omega),[y-z,y+z])-d_{\theta}(X(\omega),[0,0])}{d_{\theta}([y-z,y+z],[0,0])+1}\geq -1$$

because using the triangular inequality,

$$\geq \inf_{(y,z)\in\mathbb{R}\times[0,\infty)} \frac{d_{\theta}(X(\omega), [0,0]) - d_{\theta}([y-z,y+z], [0,0]) - d_{\theta}(X(\omega), [0,0])}{d_{\theta}([y-z,y+z], [0,0]) + 1}$$
$$= \inf_{(y,z)\in\mathbb{R}\times[0,\infty)} \frac{-d_{\theta}([y-z,y+z], [0,0])}{d_{\theta}([y-z,y+z], [0,0]) + 1} \geq -1.$$

• The condition  $\liminf_{(y,z)\to\infty} b((y,z)) > \gamma((y_0,z_0))$  is satisfied. Let  $\{(y_n,z_n)\} \subset \mathbb{R} \times [0,\infty)$  be any sequence with  $(y_n,z_n) \xrightarrow[n\to\infty]{} \infty$  in the sense that  $d_{\theta}([y_n-z_n,y_n+z_n],[0,0]) \xrightarrow[n\to\infty]{} \infty$ , and set  $M = E[d_{\theta}(X(\omega),[y_0-z_0,y_0+z_0]) - d_{\theta}(X(\omega),[0,0])] = \gamma((y_0,z_0)) \in \mathbb{R}$ , where  $(y_0,z_0)$  represents the minimum in (A-4). Then, there exists  $n_0 \in \mathbb{N}$  such that for all  $n \geq n_0$ ,  $d_{\theta}([y_n-z_n,y_n+z_n],[0,0]) > M$ . Thus,

$$\inf_{k \ge n_0} b((y_k, z_k)) = \inf_{k \ge n_0} \left( d_\theta([y_k - z_k, y_k + z_k], [0, 0]) + 1 \right) \ge M + 1,$$

so 
$$\liminf_{n \to \infty} b((y_n, z_n)) = \lim_{n \to \infty} (\inf_{k \ge n} b((y_k, z_k))) \ge M + 1 > M = \gamma((y_0, z_0)).$$

• It also holds that

$$E\left[\liminf_{(y,z)\to\infty} \frac{d_{\theta}(X(\omega), [y-z, y+z]) - d_{\theta}(X(\omega), [0,0])}{b((y,z))}\right] \ge 1, \quad (2)$$

because we have that

$$\liminf_{(y,z)\to\infty} \frac{d_{\theta}(X(\omega), [y-z, y+z]) - d_{\theta}(X(\omega), [0,0])}{d_{\theta}([y-z, y+z], [0,0]) + 1} \ge 1,$$

Indeed, for any  $\omega \in \Omega$ 

$$\lim_{(y,z)\to\infty} \inf \frac{d_{\theta}(X(\omega), [y-z, y+z]) - d_{\theta}(X(\omega), [0,0])}{d_{\theta}([y-z, y+z], [0,0]) + 1}$$
$$= \lim_{n\to\infty} \left( \inf_{k\geq n} \frac{d_{\theta}(X(\omega), [y_k - z_k, y_k + z_k]) - d_{\theta}(X(\omega), [0,0])}{d_{\theta}([y_k - z_k, y_k + z_k], [0,0]) + 1} \right).$$

The sequence

$$\left\{\inf_{k\geq n}\frac{d_{\theta}(X(\omega), [y_k - z_k, y_k + z_k]) - d_{\theta}(X(\omega), [0, 0])}{d_{\theta}([y_k - z_k, y_k + z_k], [0, 0]) + 1}\right\}_{n\in\mathbb{N}}$$
(3)

is monotonically increasing with upper bound 1, since for all  $k \in \mathbb{N}$ , using the triangular inequality,

$$\frac{d_{\theta}(X(\omega), [y_k - z_k, y_k + z_k]) - d_{\theta}(X(\omega), [0, 0])}{d_{\theta}([y_k - z_k, y_k + z_k], [0, 0]) + 1} \le \frac{d_{\theta}([y_k - z_k, y_k + z_k], [0, 0])}{d_{\theta}([y_k - z_k, y_k + z_k], [0, 0]) + 1} \le 1.$$

Hence, the sequence (3) converges to its supremum:

$$\lim_{n \to \infty} \left( \inf_{k \ge n} \frac{d_{\theta}(X(\omega), [y_k - z_k, y_k + z_k]) - d_{\theta}(X(\omega), [0, 0])}{d_{\theta}([y_k - z_k, y_k + z_k], [0, 0]) + 1} \right)$$
$$= \sup_{n} \left( \inf_{k \ge n} \frac{d_{\theta}(X(\omega), [y_k - z_k, y_k + z_k]) - d_{\theta}(X(\omega), [0, 0])}{d_{\theta}([y_k - z_k, y_k + z_k], [0, 0]) + 1} \right)$$

Finally, it will be checked this supremum is at least equal to 1. By *reductio ad absurdum*, we suppose that

$$\sup_{n} \left( \inf_{k \ge n} \frac{d_{\theta}(X(\omega), [y_k - z_k, y_k + z_k]) - d_{\theta}(X(\omega), [0, 0])}{d_{\theta}([y_k - z_k, y_k + z_k], [0, 0]) + 1} \right) = 1 - \varepsilon,$$

for some  $\varepsilon > 0$ . We need to show that we can find an  $n^* \in \mathbb{N}$  such that

$$\inf_{k \ge n^*} \frac{d_{\theta}(X(\omega), [y_k - z_k, y_k + z_k]) - d_{\theta}(X(\omega), [0, 0])}{d_{\theta}([y_k - z_k, y_k + z_k], [0, 0]) + 1} > 1 - \varepsilon.$$
(4)

Recall that  $(y_n, z_n) \xrightarrow[n \to]{n \to} \infty$ , so for any  $M \in \mathbb{R}$ , there exists  $n^* \in \mathbb{N}$  such that for all  $n \ge n^*$ ,  $d_{\theta}([y_n - z_n, y_n + z_n], [0, 0]) > M$ . Then,

$$d_{\theta}([y_n - z_n, y_n + z_n], X(\omega)) \ge d_{\theta}([y_n - z_n, y_n + z_n], [0, 0]) - d_{\theta}(X(\omega), [0, 0])$$

 $> M - d_{\theta}(X(\omega), [0, 0]).$ 

Take  $M := \frac{2}{\varepsilon} - 1 + \frac{4}{\varepsilon} \cdot d_{\theta}(X(\omega), [0, 0]) \in \mathbb{R}$  (for the fixed arbitrary  $\omega \in \Omega$ ), then for any  $k \ge n^*$ , we have that

$$d_{\theta}(X(\omega), [y_{k} - z_{k}, y_{k} + z_{k}]) - d_{\theta}(X(\omega), [0, 0])$$

$$= \left(1 - \frac{\varepsilon}{2}\right) d_{\theta}(X(\omega), [y_{k} - z_{k}, y_{k} + z_{k}]) + \frac{\varepsilon}{2} d_{\theta}(X(\omega), [y_{k} - z_{k}, y_{k} + z_{k}]) - d_{\theta}(X(\omega), [0, 0])$$

$$\geq \left(1 - \frac{\varepsilon}{2}\right) d_{\theta}([y_{k} - z_{k}, y_{k} + z_{k}], [0, 0]) - \left(1 - \frac{\varepsilon}{2}\right) d_{\theta}(X(\omega), [0, 0]) + \frac{\varepsilon}{2} d_{\theta}(X(\omega), [y_{k} - z_{k}, y_{k} + z_{k}]) - d_{\theta}(X(\omega), [0, 0])$$

$$> \left(1 - \frac{\varepsilon}{2}\right) d_{\theta}([y_{k} - z_{k}, y_{k} + z_{k}], [0, 0]) + \frac{\varepsilon}{2} (M - d_{\theta}(X(\omega), [0, 0])) - \left(2 - \frac{\varepsilon}{2}\right) d_{\theta}(X(\omega), [0, 0])$$

$$\geq \left(1 - \frac{\varepsilon}{2}\right) d_{\theta}([y_{k} - z_{k}, y_{k} + z_{k}], [0, 0]) + \frac{\varepsilon}{2} \left(2 - 1 + \left(\frac{4}{2} - 1\right) d_{\theta}(X(\omega), [0, 0])\right)$$

$$> \left(1 - \frac{\varepsilon}{2}\right) d_{\theta}([y_k - z_k, y_k + z_k], [0, 0]) + \frac{\varepsilon}{2} \left(\frac{2}{\varepsilon} - 1 + \left(\frac{4}{\varepsilon} - 1\right) d_{\theta}(X(\omega), [0, 0])\right) - \left(2 - \frac{\varepsilon}{2}\right) d_{\theta}(X(\omega), [0, 0]) = \left(1 - \frac{\varepsilon}{2}\right) d_{\theta}([y_k - z_k, y_k + z_k], [0, 0]) + 1 - \frac{\varepsilon}{2} = \left(1 - \frac{\varepsilon}{2}\right) \left(d_{\theta}([y_k - z_k, y_k + z_k], [0, 0]) + 1\right).$$

Hence, for all  $k \ge n^*$ ,

$$\frac{d_{\theta}(X(\omega), [y_k - z_k, y_k + z_k]) - d_{\theta}(X(\omega), [0, 0])}{d_{\theta}([y_k - z_k, y_k + z_k], [0, 0]) + 1} \ge 1 - \frac{\varepsilon}{2} > 1 - \varepsilon,$$

which implies (4) and thus the inequality (2) follows. Hence, the three inequalities in condition (A-5) are satisfied, which completes the proof.

## 4. Concluding remarks

This paper complements the study of the properties of the  $d_{\theta}$ -median as a robust estimator of the center of a random interval by showing its strong consistency, which is one of the most important basic properties of an estimator. This result open the door to further develop robust statistical inference for random intervals based on the  $d_{\theta}$ -median.

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