

Approximate Calculation of the Chebyshev Center for a Convex Compact Set in $\mathbb{R}^n$

Maxim V. Balashov

V. A. Trapeznikov Institute of Control Sciences, Moscow, Russia 117997
balashov73@mail.ru

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We present an approximate algorithm for calculation of the Chebyshev center for a convex compact subset from $\mathbb{R}^n$ which is given via its supporting function. We reduce the problem to the solution of a linear programming problem and estimate the error between an approximate and the exact solutions in terms of the step of a grid.

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1. Introduction and main notations

Let $\mathbb{R}^n$ be an n -dimensional Euclidean space with the inner product $(\cdot, \cdot)$, the Euclidean norm $\|\cdot\|$ and $B_r(a) = \{x \in \mathbb{R}^n \mid \|x - a\| \leq r\}$ for $a \in \mathbb{R}^n$, $r > 0$. If $M \subset \mathbb{R}^n$ is a symmetric with respect to zero compact subset with nonempty interior, then we can consider a particular norm $n_M(x) = \mu(x, M) = \inf\{t > 0 \mid \frac{x}{t} \in M\}$ for all $x \in \mathbb{R}^n$ as Minkowski functional of M . We shall define the particular ball of radius r with center $a \in \mathbb{R}^n$: $rM + a = \{x \in \mathbb{R}^n \mid n_M(x - a) \leq r\}$. Denote by $\text{int } A$ and ∂A the interior and the boundary of the set A .

Let $A \subset \mathbb{R}^n$ be a convex compact subset. The *Chebyshev center* of the set A is the set of solutions $C_M(A)$ of the problem

$$R_M(A) = \min_{x \in \mathbb{R}^n} \max_{a \in A} n_M(x - a), \quad (1)$$

i.e. for any $x \in C_M(A)$ we have $\max_{a \in A} n_M(x - a) = R_M(A)$. Obviously $C_M(A) \neq \emptyset$ due to compactness arguments. So, $C_M(A)$ is the set of centers of balls $R_M(A)M + x$ of the smallest radius $R_M(A)$ each of which contains A . Note that if a compact set A is not convex, then $C_M(A) = C_M(\text{co } A)$. Further we shall assume that the set A is convex.

In general, $C_M(A)$ is not necessarily a singleton and does not belong to the set A . But in the case of Euclidean norm $n_M(x) = \|x\|$ the set $C(A)$, the solution of the problem

$$R(A) = \min_{x \in \mathbb{R}^n} \max_{a \in A} \|x - a\|, \quad (2)$$

is singleton and $C(A) \in A$ [6, Theorem 1].

The concept of Chebyshev center is an important tool in different applications [4, 5, 7, 10, 11], for example in some identification problems. It is known, that the problem of finding the Chebyshev center (2) is NP-hard and there are only few situation (in certain cases on the plane, for special sets: finite sets, intersection of two ellipsoids) when the Chebyshev center can be found numerically efficiently, see details in [10].

Recall that the *Minkowski-Pontryagin difference* [9] for subsets $P, Q \subset \mathbb{R}^n$ is

$$P \overset{*}{-} Q = \{x \in \mathbb{R}^n \mid x + Q \subset P\} = \bigcap_{x \in Q} (P - x),$$

and the *Hausdorff distance* is

$$h(P, Q) = \max \left\{ \max_{x \in P} \varrho_Q(x), \max_{x \in Q} \varrho_P(x) \right\}.$$

Here $\varrho_Q(x) = \inf_{y \in Q} \|x - y\|$ is the distance function.

Then we can rewrite $C_M(A)$ as follows

$$\forall x \in C_M(A) \quad A \subset R_M(A) \cdot M + x,$$

$$\forall x \in C_M(A) \quad -x \in R_M(A) \cdot M \overset{*}{-} A,$$

$$\text{i.e.} \quad -C_M(A) = (R_M(A) \cdot M) \overset{*}{-} A = \bigcap_{a \in A} (R_M(A) \cdot M - a). \quad (3)$$

We plan to find an approximate solution for problem (2). The idea is to approximate the unit ball $B_1(0)$ on a grid $\mathbb{G}_1 = \{p_i\}_{i=1}^I$ of unit vectors and to solve problem (1) with the “ball” $M = \{x \in \mathbb{R}^n \mid (p_i, x) \leq 1, i = 1, \dots, I\}$. Note that approximation M is not necessary a symmetric body, thus M is a quasiball. But this fact does not matter for future speculations.

We shall assume that we know the supporting function of the set A , i.e. for any $p \in \mathbb{R}^n$ we know

$$f(p) = \max_{a \in A} (p, a).$$

This is quite natural for a number of applications. In particular, if

$$A = \{x \in \mathbb{R}^n \mid (q_j, x) \leq \mu_j, j = 1, \dots, N\}, \quad \|q_j\| = 1, \mu_j \in \mathbb{R},$$

then we can find $f(p)$ for any $\|p\| = 1$ as the solution of the next linear programming problem

$$f(p) = \inf \left\{ \sum_{j=1}^N \lambda_j \mu_j \mid \sum_{j=1}^N \lambda_j q_j = p, \quad \lambda_j \geq 0, j = 1, \dots, N \right\}.$$

Further we shall consider an external approximation of the set

$$A = \{x \in \mathbb{R}^n \mid (p, x) \leq f(p), \|p\| = 1\}$$

on a grid $\mathbb{G}_2$ of the form

$$\hat{A} = \{x \in \mathbb{R}^n \mid (q_j, x) \leq f(q_j), \{q_j\}_{j=1}^N = \mathbb{G}_2\}.$$

We shall assume that $\mathbb{G}_2 = \mathbb{G}_1$, but the last is not necessary.

Considering an approximation M of the ball $B_1(0)$ and an approximation of the set A on a grid $\mathbb{G}_1$ we obtain the next problem

$$R \rightarrow \min, \quad R + (p_i, x) \geq f(p_i), \quad i = 1, \dots, I. \quad (4)$$

Any solution of problem (4) is a pair $(x, R) \in \mathbb{R}^n \times \mathbb{R}$ which gives the center and the smallest radius of a ball containing $\hat{A}$:

$$\hat{A} \subset R \cdot M + x.$$

Our purpose is to estimate the difference $\|x - C(A)\|$, where $C(A)$ is the solution (2), via the Hausdorff distance $h(A, \hat{A})$ and $h(M, B_1(0))$ and finally via step of the grid and geometrical properties of the set A .

There are few possibilities to estimate the error. We want to recall one result about stability of minimization problems [8, Theorem 1.19.3].

Proposition 1.1. *Let $A_1, A_2 \subset \mathbb{R}^n$ be compact convex subsets, $A = \text{co}(A_1 \cup A_2)$. We consider two minimization problems $\min_{x \in A_i} f_i(x)$, $i = 1, 2$, with the following properties:*

- (1) $f_1 : A \rightarrow \mathbb{R}$ is Lipschitz continuous with constant $L > 0$,
- (2) $f_2 : A \rightarrow \mathbb{R}$ is continuous and strongly convex with constant $\kappa > 0$,
- (3) $\exists \varepsilon > 0$ such that $\forall x \in A$ we have $|f_1(x) - f_2(x)| \leq \varepsilon$.

Suppose that $u_1 \in A_1$ is some solution of the first problem, $u_2 \in A_2$ is the solution of the second problem and $f_1(u_1) \leq f_1(u_2)$. Then

$$\|u_1 - u_2\| \leq \sqrt{\frac{2Lh(A_1, A_2) + 4\varepsilon}{\kappa}} + h(A_1, A_2).$$

Proof. Put $h = h(A_1, A_2)$. Fix a solution u_1 . There exists $x_2 \in A_2$: $\|u_1 - x_2\| \leq h$.

Consider the function g : $g(x) = f_2(x)$ for all $x \in A_2$ and $g(x) = +\infty$ otherwise. This function is strongly convex with constant κ and thus

$$f_2(x_2) - f_2(u_2) = g(x_2) - g(u_2) \geq \frac{\kappa}{2} \|x_2 - u_2\|^2.$$

By (3) we get

$$\frac{\kappa}{2} \|x_2 - u_2\|^2 \leq f_2(x_2) - f_2(u_2) \leq f_1(x_2) - f_1(u_2) + 2\varepsilon$$

and from (1) and inequality $f_1(u_1) \leq f_1(u_2)$ we obtain

$$\frac{\kappa}{2} \|x_2 - u_2\|^2 \leq f_1(x_2) - f_1(u_1) + 2\varepsilon \leq Lh + 2\varepsilon, \quad \|x_2 - u_2\| \leq \sqrt{\frac{2Lh + 4\varepsilon}{\kappa}}.$$

Hence $\|u_1 - u_2\| \leq \|u_1 - x_2\| + \|x_2 - u_2\| \leq h + \sqrt{\frac{2Lh + 4\varepsilon}{\kappa}}$. □

Taking $A_2 = A$, $A_1 = \hat{A}$, $f_2(x) = \max_{a \in A} \|x - a\|^2$, $f_1(x) = \max_{a \in \hat{A}} n_M^2(x - a)$ one can estimate constants κ (obviously $\kappa = 2$ for f_2), L , ε , etc. and obtain Holder dependence between $C(A)$ and an approximate Chebyshev center (which belongs to $\hat{A}$ in this situation) via $h(M, B_1(0))$, $h(A, \hat{A})$.

2. Approximation on a grid

We specify the definition of a grid and recall the results about the error of approximation of a set in the Hausdorff metric.

Definition 2.1. ([1],[2],[8]) Let $\Delta \in (0, \frac{1}{2})$. The set $\mathbb{G} = \{p_i\}_{i=1}^I$, $\|p_i\| = 1$ for all i , is called a *grid* with the step Δ if for any unit vector $p \in \mathbb{R}^n$ there exists $I_p \subset \{1, \dots, I\}$ and $\alpha_i > 0$, $i \in I_p$, such that

$$p = \sum_{i \in I_p} \alpha_i p_i, \quad \|p_i - p_j\| < \Delta \quad \forall i, j \in I_p.$$

Proposition 2.2. ([1, Remark 1]) Let $\mathbb{G}$ be a grid with the step $\Delta \in (0, \frac{1}{2})$ and $M = \{x \in \mathbb{R}^n \mid (p_i, x) \leq 1, i = 1, \dots, I\}$. Then $B_1(0) \subset M$ and

$$h(B_1(0), M) \leq \frac{\Delta^2}{1 - \frac{1}{2}\Delta^2}. \quad (5)$$

Analogously we have for any $R > 0$: $h(B_R(0), RM) \leq \frac{R\Delta^2}{1 - \frac{1}{2}\Delta^2}$. (6)

Proposition 2.3. [2], [1, Definition 3, Remark 1] Let $\mathbb{G}$ be a grid with the step $\Delta \in (0, \frac{1}{2})$ and the supporting function of the convex compact set A is uniformly continuous subdifferentiable with modulus $w(\cdot)$,

$$\hat{A} = \{x \in \mathbb{R}^n \mid (p_i, x) \leq s(p_i, A), i = 1, \dots, I\}.$$

Then $h(A, \hat{A}) \leq \frac{w(\Delta)}{1 - \frac{1}{2}\Delta^2}$. (7)

In particular, if $A = \bigcap_{x \in X} B_R(x)$ then $w(\Delta) = R\Delta^2$, if A is a polyhedron then $w(\Delta) = \Delta \cdot \text{diam } A$.

3. The distance between $C(A)$ and $C_M(\hat{A})$

Lemma 3.1. Suppose that $A \subset \mathbb{R}^n$ is a convex compact set, $R(A)$ is the solution of problem (2) and R_M is the solution of problem (4) on a grid $\mathbb{G}_1$ with the step $\Delta \in (0, \frac{1}{2})$. Then

$$R_M \leq R(A) + h(A, \hat{A}) \quad \text{and} \quad C(A) + (R(A) + \varepsilon)B_1(0) \supset \hat{A}, \quad (8)$$

where $\varepsilon = h(A, \hat{A}) + \frac{(R(A) + h(A, \hat{A}))\Delta^2}{1 - \frac{1}{2}\Delta^2}$.

Proof. Put $h = h(A, \hat{A})$. By the inclusion $A \subset C(A) + B_{R(A)}(0)$ we have

$$\hat{A} \subset A + hB_1(0) \subset C(A) + (R(A) + h)B_1(0) \subset C(A) + (R(A) + h)M.$$

Thus $R_M \leq R(A) + h$. From Proposition 2.2, formula (6) with $R = R(A) + h$, we get $(R(A) + h)M \subset (R(A) + \varepsilon)B_1(0)$ and we obtain the right inclusion in (8). $\square$

By formula (3) and Lemma 3.1 we obtain

$$-C_M(\hat{A}) = R_M M \overset{*}{-} \hat{A} \subset B_{R(A)+\varepsilon}(0) \overset{*}{-} A = \bigcap_{a \in A} B_{R(A)+\varepsilon}(-a), \quad (9)$$

where ε is from Lemma 3.1, simultaneously with the equality

$$B_{R(A)}(0) \overset{*}{-} A = -C(A).$$

Theorem 3.2. *Assume that conditions of Lemma 3.1 are fulfilled. Then*

$$h(C_M(\hat{A}), C(A)) \leq \sqrt{2R(A)\varepsilon + \varepsilon^2}, \quad \varepsilon = h(A, \hat{A}) + \frac{(R(A) + h(A, \hat{A}))\Delta^2}{1 - \frac{1}{2}\Delta^2}.$$

Proof. Without loss of generality put $C(A) = 0$; put $R = R(A)$. Geometrically by (9) $C_M(\hat{A})$ is contained in the collection of shifts $v \in \mathbb{R}^n$ such that $-v + A \subset B_{R+\varepsilon}(0)$ or $v - A \subset B_{R+\varepsilon}(0)$, moreover $0 = B_R(0) \overset{*}{-} A$.

By the Jung theorem [3, Theorem 11.5.8] there exists a simplex $S = \text{co}\{s_i\}_{i=1}^k$, $2 \leq k \leq n+1$, with $S \subset A$ and $0 = B_R(0) \overset{*}{-} S$. The latter means that there exist nonnegative numbers $\{\lambda_i\}_{i=1}^k$, $\sum_{i=1}^k \lambda_i = 1$ and $\sum_{i=1}^k \lambda_i s_i = 0 = C(A)$.

Fix $v \in C_M(\hat{A})$. Then $\sum_{i=1}^k \lambda_i(v, s_i) = 0$ and there exists i , $1 \leq i \leq k$, with $(v, -s_i) \geq 0$. Hence $v - s_i \in v - A$ and, on the other hand,

$$v - s_i \in B_{R+\varepsilon}(0) \cap \{x \in \mathbb{R}^n \mid (x, -s_i) \geq \|s_i\|^2\} = \Sigma.$$

The set Σ is a segment and any farthest from $-s_i$ point $a \in \Sigma$ is a point of the intersection $a \in \partial B_{R+\varepsilon}(0) \cap \{x \in \mathbb{R}^n \mid (x, -s_i) = \|s_i\|^2\}$. From the Pythagoras theorem we conclude

$$\|v\| \leq \|a - (-s_i)\| = \sqrt{\|a\|^2 - \|s_i\|^2} = \sqrt{(R+\varepsilon)^2 - R^2} = \sqrt{2R\varepsilon + \varepsilon^2}. \quad \square$$

Theorem 3.3. *Assume that the conditions of Lemma 3.1 hold. Let $D_A = A \cap \partial B_{R(A)}(C(A))$ and assume that there exists $\delta > 0$ with $B_\delta(C(A)) \subset \text{co } D_A$.*

Then $h(C_M(\hat{A}), C(A)) \leq \varepsilon \frac{R(A)}{\delta}$, where $\varepsilon = h(A, \hat{A}) + \frac{(R(A) + h(A, \hat{A}))\Delta^2}{1 - \frac{1}{2}\Delta^2}$.

Proof. Without loss of generality suppose that $C(A) = 0$; put $R = R(A)$.

Fix $\tau \in (0, 1)$. Let $D_\tau = \{d_i\}_{i=1}^N \subset D_A$ be such a finite set that

$$h(\text{co } D_\tau, \text{co } D_A) < (1 - \tau)\delta.$$

Then $B_{\tau\delta}(0) \subset \text{co } D_\tau$.

Let $v \neq 0$ be a shift: $-v + D_\tau \subset -v + A \subset B_{R+\varepsilon}(0)$.

Suppose that $-v \in \text{cone}\{d_{i_1}, \dots, d_{i_k}\}$ and moreover $\text{co}\{d_{i_1}, \dots, d_{i_k}\}$ is an edge of $\text{co } D_\tau$.

Define L as the affine hull of the points $\{d_{i_j}\}_{j=1}^k$. We have $h = \varrho_L(0) \geq \tau\delta$. Define $e = P_L 0$.

The angle φ_0 between d_{i_j} (for any $1 \leq j \leq k$) and e fulfills

$$\cos \varphi_0 = \frac{h}{R} \geq \frac{\tau\delta}{R}.$$

We claim that there exists j , $1 \leq j \leq k$, with $\angle(d_{i_j}, -v) \leq \varphi_0$. Indeed, put $v_1 = (-L) \cap \{tv \mid t \geq 0\}$. We have $-v_1 \in \text{co}\{d_{i_1}, \dots, d_{i_k}\} = \mathcal{S}$. The point e is the center of the circumscribed ball for $\text{co}\{d_{i_1}, \dots, d_{i_k}\}$. Thus by the inclusion

$$\mathcal{S} \subset \bigcup_{j=1}^k \left(\mathcal{S} \cap B_{\|d_{i_j}-e\|}(d_{i_j}) \right)$$

there exists a number j , $1 \leq j \leq k$, with $-v_1 \in \mathcal{S} \cap B_{\|d_{i_j}-e\|}(d_{i_j})$ and hence $\|d_{i_j} - (-v_1)\| \leq \|d_{i_j} - e\|$. Denote $\|d_{i_j}\| = a$, $\|v_1\| = b$, $\|e\| = b_1$, $\|d_{i_j} - (-v_1)\| = c$, $\|d_{i_j} - e\| = c_1$. Note that $b_1 \leq b$, $c \leq c_1$ and $a^2 = b_1^2 + c_1^2$. Then we conclude from the Cosine theorem

$$\begin{aligned} \cos \varphi_0 &= \frac{2b_1b}{2ab} \leq \frac{b_1^2 + b^2}{2ab} = \frac{b_1^2 + c_1^2 + b^2 - c_1^2}{2ab} = \frac{a^2 + b^2 - c_1^2}{2ab} \\ &\leq \frac{a^2 + b^2 - c^2}{2ab} = \cos \angle(-v_1, d_{i_j}). \end{aligned}$$

Consider a cone K with the vertex $-d_{i_j}$, the axis $\{-td_{i_j} \mid t \geq 1\}$ and the angle between the axis and a generatrix equals φ_0 .

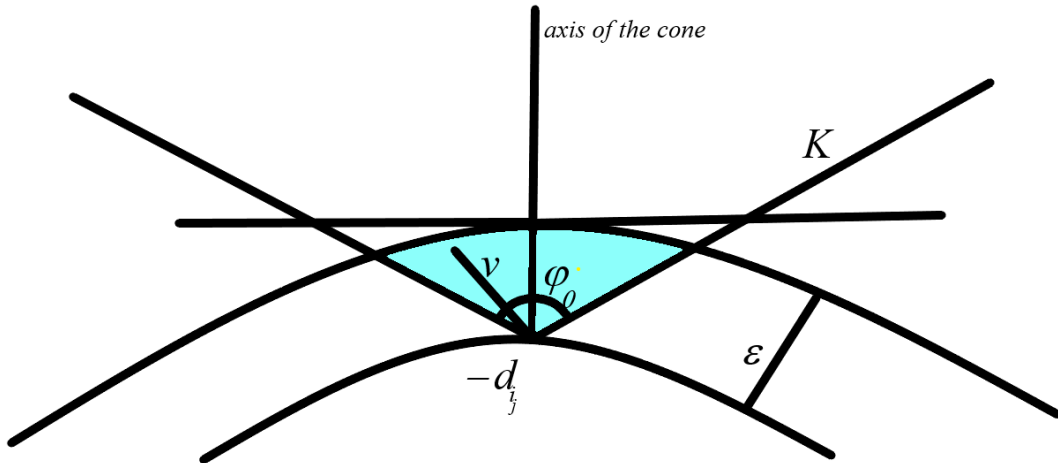


Figure 3.1: Intersection of the layer and the cone K .

Considering the intersection of the layer $B_{R+\varepsilon}(0) \setminus B_R(0)$, see Figure 3.1, with the cone K we get

$$\frac{\varepsilon}{\|v\|} \geq \cos \varphi_0 \geq \frac{\tau\delta}{R}.$$

Hence for any vector v with $v - A \subset B_{R+\varepsilon}(0)$ we obtain that $\|v\| \leq \varepsilon \frac{R}{\tau\delta}$. Thus $h(C_M(\hat{A}), C(A)) \leq \varepsilon \frac{R(A)}{\tau\delta}$ for all $\tau \in (0, 1)$. Passing $\tau \rightarrow 1 - 0$ we obtain

$$h(C_M(\hat{A}), C(A)) \leq \varepsilon \frac{R(A)}{\delta}. \quad \square$$

Remark 3.4. Theorem 3.3 gives a sufficient condition. Consider now in $\mathbb{R}^2$ the set $A = \text{co}\{(0, -1), (0, 1)\}$ and a grid $\mathbb{G} = \{(\cos \frac{2\pi m}{N}, \sin \frac{2\pi m}{N})\}_{m=0}^{N-1}$ which contains the vectors $\pm(1, 0)$, but does not contain the vectors $\pm(0, 1)$. Then $\hat{A} = A$ and $C_M(A) = C(A) = (0, 0)$.

Example 3.5. In $\mathbb{R}^2$ consider $A = \text{co}\{(0, -1), (0, 1)\}$ and a grid

$$\mathbb{G} = \{(\cos \frac{2\pi m}{N}, \sin \frac{2\pi m}{N})\}_{m=0}^{N-1}$$

which contains the vectors $\pm(0, 1)$ and $\pm(1, 0)$. Then $\hat{A} = A$, the length of a side of the regular N -polygon M equals $2 \tan \frac{\pi}{N}$ and $\Delta = 2 \sin \frac{\pi}{N}$. It is easy to see that $C(A) = (0, 0)$, $C_M(A) \supset [(-\frac{1}{2}\Delta, 0), (\frac{1}{2}\Delta, 0)]$. Thus $h(C(A), C_M(A)) \geq \frac{1}{2}\Delta$. We see that Theorem 3.2 is realized in the present example, but not Theorem 3.3.

Suppose that under the conditions of Theorem 3.3 $A \subset \mathbb{R}^n$ is a polyhedron and the grid $\mathbb{G}_1$ with the step Δ is such that $\hat{A} = A$. Then the number I of elements of the grid is $I \asymp O(\frac{1}{\Delta^{n-1}})$, $\Delta \rightarrow 0$. The error for approximate Chebyshev center is

$$h(C_M(\hat{A}), C(A)) \leq \frac{R R(A) \Delta^2}{\delta 1 - \frac{1}{2}\Delta^2} \asymp \Delta^2, \quad \Delta \rightarrow 0.$$

In other words, after $O(N)$ operations we get an estimate

$$h(C_M(\hat{A}), C(A)) \asymp \frac{1}{N^{2/(n-1)}}.$$

In the same situation under the conditions of Theorem 3.2 after $O(N)$ operations we get an estimate

$$h(C_M(\hat{A}), C(A)) \asymp \frac{1}{N^{1/(n-1)}}.$$

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