**Analysis of the convergence rate for the cyclic projection algorithm applied to semi-algebraic convex sets**

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# **2 [Notation and auxiliary results](#page-3-0) •** [Notation](#page-3-0)



### **Motivation**

Given finitely many closed convex sets  $C_1, C_2, \cdots, C_m$  in  $\mathbb{R}^n$  with  $\bigcap_{i=1}^m C_i \neq \varnothing$ . Let  $x_0 \in \mathbb{R}^n$ . The sequence of *cyclic projections*,  $(x_k)_{k \in \mathbb{N}},$ is defined by

 $X_1 := P_1 X_0, X_2 := P_2 X_1, \cdots, X_m := P_m X_{m-1}, X_{m+1} := P_1 X_m, \ldots$ 

where *P<sup>i</sup>* denotes the Euclidean projection to the set *C<sup>i</sup>* .

Bregman showed that

<span id="page-2-0"></span>
$$
x_k \longrightarrow x_\infty \in \bigcap_{i=1}^m C_i.
$$

In this talk, we study

how fast 
$$
(x_k)_{k \in \mathbb{N}}
$$
 converges a point in  $\bigcap_{i=1}^m C_i$ .

Throughout this talk,

- $\mathbb{R}^n$  is a *Euclidean space* with the norm  $\|\cdot\|$  and the inner product  $\langle \cdot, \cdot \rangle$ .
- Let *<sup>C</sup>* <sup>⊆</sup> <sup>R</sup> *n* . The *interior* of *C* is int *C* and *C* is the *norm closure* of *C*.
- $\mathbb{B}(\mathsf{x}_0,\delta):=\big\{\mathsf{x}\in\mathbb{R}^n\mid\|\mathsf{x}-\mathsf{x}_0\|\leq\delta\big\}.$
- $\bullet$  [ $\alpha$ ]<sub>+</sub> := max{ $\alpha$ , 0}.
- The *distance function* to the set *C* is dist(*x*, *C*) :=  $\inf_{c \in C} ||x c||$ .
- <span id="page-3-0"></span>The *projector operator* to the set *C* is  $P_C(x) := \{c \in C \mid ||y - c|| = \text{dist}(x, C)\}.$

### **Notation and definitions**



**Figure :** The projection of the point  $x_0$  to the set C.

 $\beta(n) := {n \choose \lfloor n \rfloor}$ [*n*/2] is the *central binomial coefficient* with respect to *n*.  $f\colon \mathbb{R}^n \to \mathbb{R}$  is a *polynomial* if

$$
f(x) := \sum_{0 \leq |\alpha| \leq r} \lambda_{\alpha} x^{\alpha},
$$

where  $\lambda_{\alpha} \in \mathbb{R}$ ,  $x = (x_1, \dots, x_n)$ ,  $x^{\alpha} := x_1^{\alpha_1} \cdots x_n^{\alpha_n}$ ,  $\alpha_i \in \mathbb{N} \cup \{0\}$ , and  $|\alpha| := \sum_{j=1}^n \alpha_j$ . The corresponding constant *r* is called the *degree* of *f*.

A polynomial *f* is constant on  $[x, y] \Rightarrow f$  is constant on aff  $[x, y]$ .

We say that *C* is a *semi-algebraic convex set* if

$$
C=\{x\in\mathbb{R}^n\mid g_i(x)\leq 0, i=1,\cdots,N\},\
$$

where  $g_i$  is convex polynomial for every  $i=1,\ldots,N.$ 

• From now on we assume that

 $m \in \mathbb{N}, \gamma_i \in \mathbb{N}, i = 1, \ldots, m$  $g_{i,1}, g_{i,2}, \cdots, g_{i,\gamma_i}$  are convex polynomials on  $\mathbb{R}^n$  $C_i := \left\{x \in \mathbb{R}^n \mid g_{i,1}(x) \leq 0, g_{i,2}(x) \leq 0, \cdots, g_{i,\gamma_i}(x) \leq 0\right\}$  $P_i := P_{C_i}$  $C := \bigcap^m$ *i*=1  $C_i \neq \emptyset$ .

## **Fact 1. (Kollár, 1999)**

Let  $g_i$  be polynomials on  $\mathbb{R}^n$  with degree  $\leq d$  for every  $i = 1, \cdots, m$ . Let  $g(x) := \max_{1 \le i \le m} g_i(x)$ . Suppose that there exists  $\varepsilon_0 > 0$  such that  $g(x) > 0$  for all  $x \in \mathbb{B}(0,\varepsilon_0) \setminus \{0\}$ . Then there exist constants  $c, \varepsilon > 0$ such that

$$
\|x\| \leq c \, g(x)^{\frac{1}{\beta(n-1)d^n}}, \quad \forall \|x\| \leq \varepsilon.
$$

### **Fact 2. (Li, 2010)**

Let *g* be a convex polynomial on  $\mathbb{R}^n$  with degree at most *d*. Let  $S := \{x \mid g(x) \leq 0\}$  and  $\bar{x} \in S$ . Then there exist constants  $c, \epsilon > 0$ such that

$$
\textnormal{dist}(x,S) \leq c \left[ g(x) \right]_+^{\frac{1}{(d-1)^n+1}}, \quad \forall \|x-\bar{x}\| \leq \varepsilon.
$$

**Theorem 1.** (Local error bounds for convex polynomial systems) Let  $g_i$  be convex polynomials on  $\mathbb{R}^n$  with degree at most  $d$  for every  $i = 1, \dots, m$ . Let  $S := \{x \in \mathbb{R}^n \mid g_i(x) \le 0, i = 1, \dots, m\}$  and  $\bar{x} \in S$ . Then there exist  $c, \varepsilon > 0$  such that

$$
\left|\text{dist}(x, S) \leq c \bigg( \max_{1 \leq i \leq m} [g_i(x)]_+ \bigg)^{\tau} \quad \forall \|x - \overline{x}\| \leq \varepsilon,
$$

where  $[\alpha]_+:=\max\{\alpha,0\},$   $\tau:=\max\big\{\frac{2}{(2d-1)^n+1}, \frac{1}{\beta(n-1)^n}\big\}$ β(*n*−1)*d n* , β(*n* − 1) is the central binomial coefficient with respect to  $n - 1$  which is given by  $\binom{n-1}{(n-1)}$  $\binom{n-1}{(n-1)/2}$ .

### **Theorem 2.** (Hölderian regularity)

Let  $\theta > 0$  and  $K \subseteq \mathbb{R}^n$  be a compact set. Then there exists  $c > 0$  such that

$$
\text{dist}^{\theta}(x, C) \leq c \bigg( \sum_{i=1}^{m} \text{dist}^{\theta}(x, C_{i}) \bigg)^{\tau}, \quad \forall x \in K,
$$

where  $\tau:=\frac{1}{\min\left\{\frac{(2d-1)^n+1}{2},\beta(n-1)d^n\right\}}$  and  $\beta(n-1)$  is the central binomial coefficient with respect to  $n-1$  which is given by  $\binom{n-1}{[(n-1)]}$  $\binom{n-1}{(n-1)/2}$ .

**Theorem 3.** (Cyclic convergence rate) Suppose that  $d > 1$ . Let  $x_0 \in \mathbb{R}^n$  and the sequence of cyclic projections,  $(x_k)_{k \in \mathbb{N}}$ , be defined by

$$
x_1 := P_1 x_0, x_2 := P_2 x_1, \cdots, x_m := P_m x_{m-1}, x_{m+1} := P_1 x_m \ldots
$$

Then  $x_k$  converges to  $x_{\infty} \in C$ , and there exists  $M > 0$  such that

$$
\|x_k-x_\infty\|\leq M\frac{1}{k^\rho},\quad \forall k\in\mathbb{N},
$$

where  $\rho := \frac{1}{\min\left\{(2d-1)^n-1, 2\beta(n-1)d^n-2\right\}}$  and  $\beta(n-1)$  is the central binomial coefficient with respect to  $n-1$  which is given by  $\binom{n-1}{[(n-1)]}$  $\binom{n-1}{(n-1)/2}$ . When  $m = 2$ , we can consider the general case where the intersection of these two sets is (possibly) empty.

We assume that

*gi* , *h<sup>j</sup>* are convex polynomials with degree at most *d*  $A := \{x \in \mathbb{R}^n \mid g_i(x) \leq 0, i = 1, \cdots, m\}$  $B := \{x \in \mathbb{R}^n \mid h_j(x) \leq 0, j = 1, \cdots, l\}$  $b_0 \in \mathbb{R}^n$ ,  $a_{k+1} := P_A b_k$ ,  $b_{k+1} := P_B a_{k+1}$ .

**Theorem 4.** (von Neumann alternating projection) Assume  $d > 1$ . Then  $a_k \longrightarrow \widetilde{a} \in A$  and  $b_k \longrightarrow \widetilde{b} \in B$  with  $\widetilde{b} - \widetilde{a} = v$ where  $v := P_{B-A}0$ . Moreover, there exists  $M > 0$  such that

$$
\boxed{\|a_k-\widetilde{a}\|\leq M \frac{1}{k^\rho}\quad \text{ and }\quad \|b_k,-b\|\leq M \frac{1}{k^\rho},\quad \forall k\in\mathbb{N},}
$$

where  $\rho:=\frac{1}{\min\left\{(2d-1)^n-1,2\beta(n-1)d^n-2\right\}}$  and  $\beta(n-1)$  is the central binomial coefficient with respect to  $n-1$  which is given by  $\binom{n-1}{[(n-1)]}$  $\binom{n-1}{(n-1)/2}$ .

**Remark 1.** The intersection of these two sets *A* and *B* is (possibly) empty.

#### **Example 1.**

Let

$$
C_1 := \{ (x, y) \in \mathbb{R}^2 \mid (x + 1)^2 + y^2 - 1 \le 0 \}
$$
  
\n
$$
C_2 := \{ (x, y) \in \mathbb{R}^2 \mid x + y - 1 \le 0 \}
$$
  
\n
$$
C_3 := \{ (x, y) \in \mathbb{R}^2 \mid (x - 1)^2 + y^2 - 1 \le 0 \}
$$
  
\n
$$
C_4 := \{ (x, y) \in \mathbb{R}^2 \mid x + (y + 2)^2 - 4 \le 0 \}.
$$

Take  $x_0 \in \mathbb{R}^2$ . Let  $(x_k)_{k \in \mathbb{N}}$  be defined by

 $x_1 := P_1 x_0, x_2 := P_2 x_1, x_3 := P_3 x_2, x_4 := P_4 x_3, x_5 := P_1 x_4 \dots$ Then  $\|x_k\| = O(\frac{1}{\epsilon^{\frac{1}{\epsilon^{\prime}}}})$  $\frac{1}{k^{\frac{1}{6}}}$ ).

### **Examples and Applications**

### **Example 2.**

Let  $\alpha > 0$  and

 $A := (-1,0) + \mathbb{B}(0,1)$  and  $B := \{(x,y) \in \mathbb{R}^2 \mid -x + \alpha \leq 0\}.$ Let  $(a_k)_{k \in \mathbb{N}}$  and  $(b_k)_{k \in \mathbb{N}}$  be defined by

 $b_0 \in \mathbb{R}^2$ ,  $a_{k+1} := P_A b_k$ ,  $b_{k+1} := P_B a_{k+1}$ .

Then for every  $k > 2$ 

$$
b_k = \left(\alpha, \frac{t_1}{\sqrt{(1+\alpha)^{2(k-1)} + t_1^2 \sum_{i=0}^{k-2} ((1+\alpha)^{2i})}}\right)
$$
  
\n
$$
a_{k+1} = \left(-1 + \frac{\alpha + 1}{\sqrt{(\alpha + 1)^2 + \frac{t_1^2}{(1+\alpha)^{2(k-1)} + t_1^2 \sum_{i=0}^{k-2} ((1+\alpha)^{2i})}}}, \frac{t_1}{\sqrt{(1+\alpha)^{2k} + t_1^2 \sum_{i=0}^{k-1} ((1+\alpha)^{2i})}}\right).
$$

Consequently,

If  $\alpha = 0$ ,  $a_k \longrightarrow 0$  and  $b_k \longrightarrow (\alpha, 0)$  at the rate of  $k^{-\frac{1}{2}}$ if  $\alpha \neq 0$ (then  $A \cap B = \emptyset$ ),  $a_k \longrightarrow 0$  and  $b_k \longrightarrow (\alpha, 0)$  at the rate of  $(1 + \alpha)^{-k}$ .

### **Remark 2.**

According to Theorem 4, we can only deduce that  $(a_k)_{k \in \mathbb{N}}$  in Example 2 converges to (0, 0) and  $(b_k)_{k\in\mathbb{N}}$  converge to  $(\alpha, 0)$  at the rate of at least of  $\kappa^{-\frac{1}{6}}$  .

# **Example 3.** Let

$$
A := \{(x, y) \in \mathbb{R}^2 \mid (x + 1)^2 + y^2 - 1 \le 0\}
$$
  

$$
B := \{(x, y) \in \mathbb{R}^2 \mid (x - 1)^2 + y^2 - 1 \le 0\}.
$$

Let  $(x_0, y_0) \in \mathbb{R}^2$  and  $(x_k, y_k)_{k \in \mathbb{N}}$  be defined by

 $(x_1, y_1) := P_A(x_0, y_0), (x_2, y_2) := P_B(x_1, y_1), (x_3, y_3) := P_A(x_2, y_2),$ 

Figure [2](#page-17-0) depicts the algorithm's trajectory with starting point (0, 2).

## **Example 3 continued**



<span id="page-17-0"></span>**Figure :** The iteration commencing at  $(0, 2)$ .

### **Example 3 continued**

 $\textsf{Since}~(\textsf{x}_k,\textsf{y}_k) \in \textsf{bdry}~\pmb{A} \cup \textsf{bdry}~\pmb{B},~\textsf{r}_k := \sqrt{\textsf{x}_k^2 + \textsf{y}_k^2}~\textsf{satisfies}~\textsf{r}_k^2 = 2\alpha_k,$ where  $\alpha_k := |x_k|$ . Hence

$$
1 - \alpha_{k+1} = \frac{1 + \alpha_k}{\sqrt{1 + 4\alpha_k}}.
$$

Linearizing the above equation, set  $w_k := 4\alpha_k$  to obtain

$$
w_{k+1} \approx w_k(1 - w_k)
$$
 and then  $\frac{1}{w_{k+1}} - \frac{1}{w_k} = \frac{1}{1 - w_k}$ .

When summing and dividing by *N*, leads to

$$
\lim_{N \to \infty} \frac{1}{N w_N} = \lim_{N \to \infty} \left( \frac{1}{N w_N} - \frac{1}{N w_0} \right) = \lim_{N \to \infty} \frac{1}{N} \sum_{k=0}^{N-1} \frac{1}{1 - w_k}
$$

$$
= \lim_{N \to \infty} \frac{1}{1 - w_N} = 1 \quad \text{(since } (x_k, y_k) \to 0\text{)}.
$$

Then we have

$$
\sqrt{x_k^2 + y_k^2} \sim \frac{1}{\sqrt{2k}}.
$$

Hence  $(x_k)_{k\in\mathbb{N}}$  and  $(y_k)_{k\in\mathbb{N}}$  converge to 0 and at the rate of  $k^{-\frac{1}{2}}.$ 

### **Remark 3.**

According to Theorem 4, we can only deduce that  $(x_k)_{k \in \mathbb{N}}$  and  $(y_k)_{k \in \mathbb{N}}$ in Example 3 converge to (0, 0) at the rate of at least of  $\kappa^{-\frac{1}{6}}.$ 

### **Examples and Applications**

#### **Example 4.**

Let *A*, *B* be defined by

$$
A:=\{(x,y)\in\mathbb{R}^2\;|\;x\leq 0\}\quad\text{and}\quad B:=\{(x,y)\in\mathbb{R}^2\;|\;y^2-x\leq 0\}.
$$

Let  $b_0 \in \mathbb{R} \times \mathbb{R}_+$  with  $||b_0|| \leq 1$ , and  $(a_k)_{k \in \mathbb{N}}$ ,  $(b_k)_{k \in \mathbb{N}}$  be defined by

$$
a_{k+1} := P_A b_k, \quad b_{k+1} := P_B a_{k+1}.
$$

### Then

 $(a_k)_{k \in \mathbb{N}}$  and  $(b_k)_{k \in \mathbb{N}}$  converge to 0 at the rate of at exactly of  $k^{-\frac{1}{2}}$ .

# **Remark 4.** According to Theorem 4, we can only deduce that  $(a_k)_{k \in \mathbb{N}}$  and  $(b_k)_{k \in \mathbb{N}}$ in Example 4 converge to (0, 0) at the rate of at least of  $\kappa^{-\frac{1}{6}}.$

- **•** Our explicit examples show that, in general, our estimate of the convergence rate of the cyclic projection algorithm will not be tight. It would be interesting to see how one can sharpen the estimate obtained in this talk and get a tight estimate for the cyclic projection algorithm.
- Can we extend the approach here to analyze the convergence rate of the Douglas-Rachford algorithm?
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# <span id="page-24-0"></span>**Thanks for your attention.**