

On the behavior of the Douglas–Rachford algorithm in possibly nonconvex settings

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Priority Research Centre for Computer-Assisted Research Mathematics and its Applications
(CARMA)



Splitting Algorithms, Modern Operator Theory, and Applications

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Based on joint works with

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Outline

- 1 Introduction
- 2 Behavior of DR algorithm in possibly inconsistent case
- 3 Finite convergence
- 4 A Lyapunov-type approach to convergence theory
- 5 Local linear convergence

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Fundamental feasibility problem

Unless stated otherwise,

X : a real Hilbert space,
 A, B : closed (possibly nonconvex) subsets of X .

The **fundamental feasibility problem** asks to

$$\text{find } x \in A \cap B.$$

In the **inconsistent** case, i.e., $A \cap B = \emptyset$, it can be naturally formulated as finding a **best approximation pair** relative to A and B :

$$\text{find } (a, b) \in A \times B \text{ such that } \|a - b\| = \inf \|A - B\|.$$

- ▶ arises in a wide range of applications including image recovery and encoding algorithms.
- ▶ is typically approached by **projection methods** which combine projectors and their variants in a suitable way to generate a sequence converging to a solution of the problem.

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Projectors and relaxed projectors

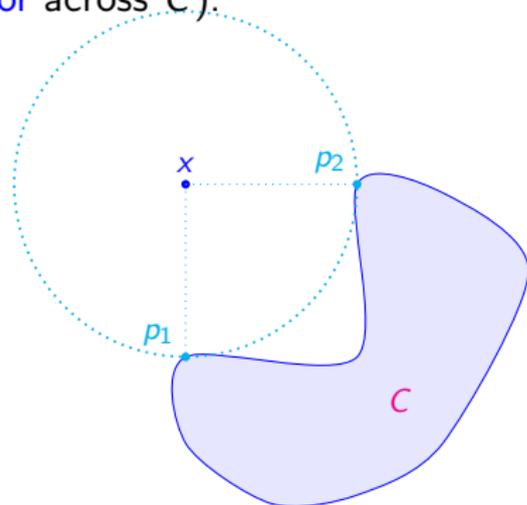
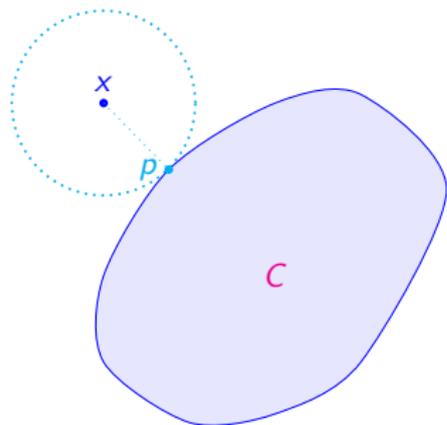
Let C be a nonempty closed set in X and $\lambda \in \mathbb{R}_+$. The **projector** onto C is

$$P_C: X \rightrightarrows C: x \mapsto P_C x := \operatorname{argmin}_{c \in C} \|x - c\|.$$

The **relaxed projector** for C with parameter λ is defined by

$$P_C^\lambda := (1 - \lambda) \operatorname{Id} + \lambda P_C.$$

- ▶ $P_C^0 = \operatorname{Id}$, $P_C^1 = P_C$,
- ▶ $P_C^2 = R_C := 2P_C - \operatorname{Id}$ (the **reflector** across C).



Projectors and relaxed projectors

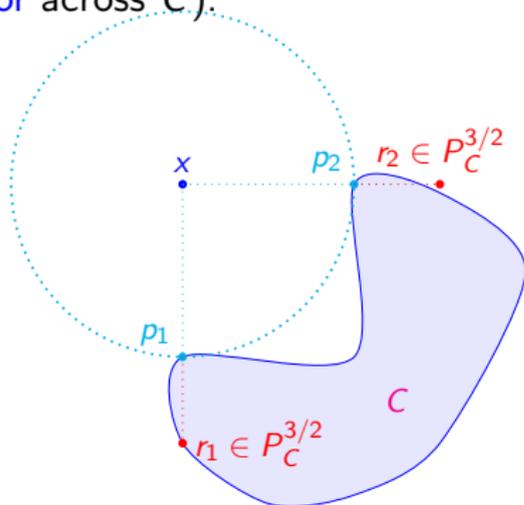
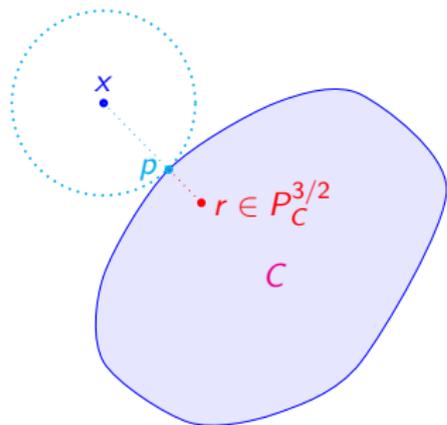
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Douglas–Rachford (DR) algorithm

Use the DR operator

$$T_{A,B} := \frac{1}{2}(\text{Id} + R_B R_A)$$

to generate a DR sequence $(z_n)_{n \in \mathbb{N}}$ by

$$(\forall n \in \mathbb{N}) \quad z_{n+1} \in T_{A,B} z_n, \quad \text{where } z_0 \in X.$$

- ▶ $T_{A,B}$ is single-valued when A and B are convex.
- ▶ $z \in \text{Fix } T_{A,B} := \{z \in X \mid z \in T_{A,B} z\} \Rightarrow (\exists a \in P_{Az}) a \in A \cap B.$

Douglas–Rachford (DR) algorithm

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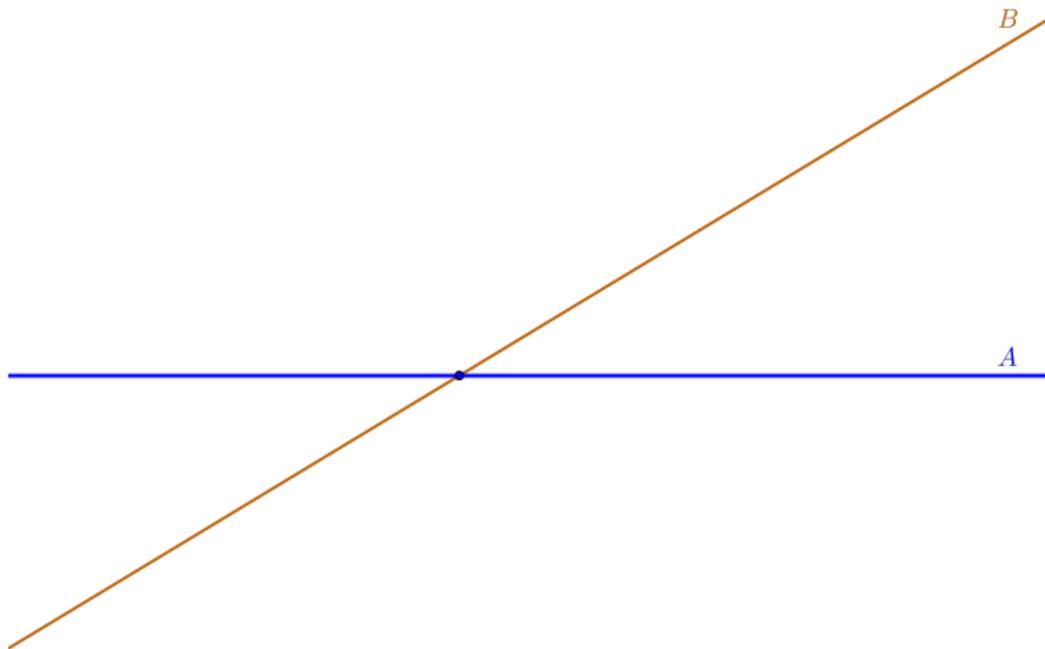
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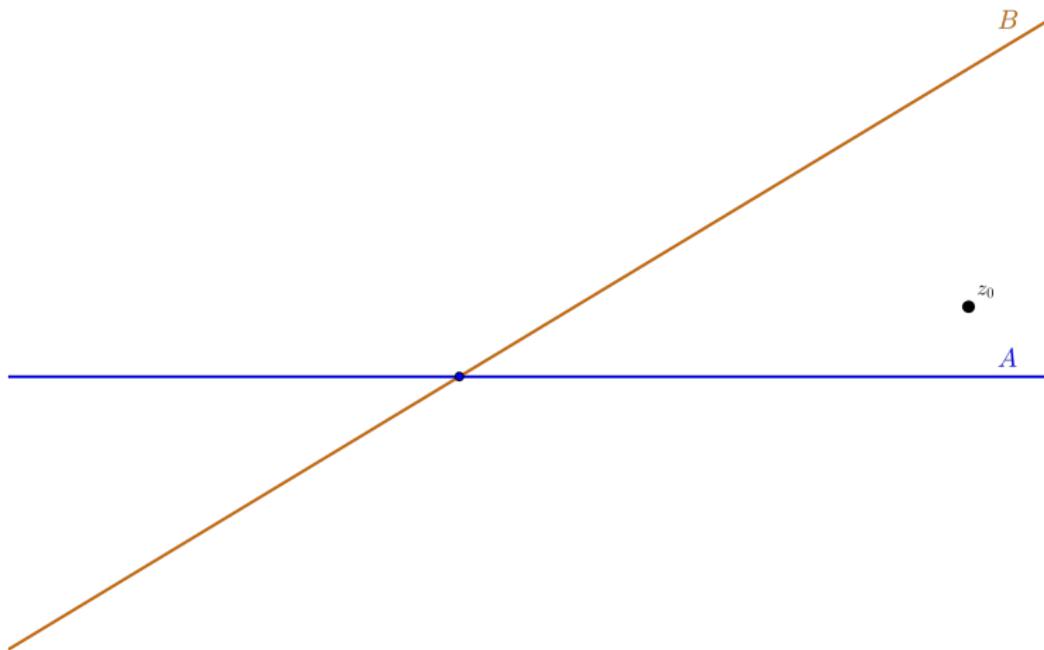
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Reflect, reflect, average. . .



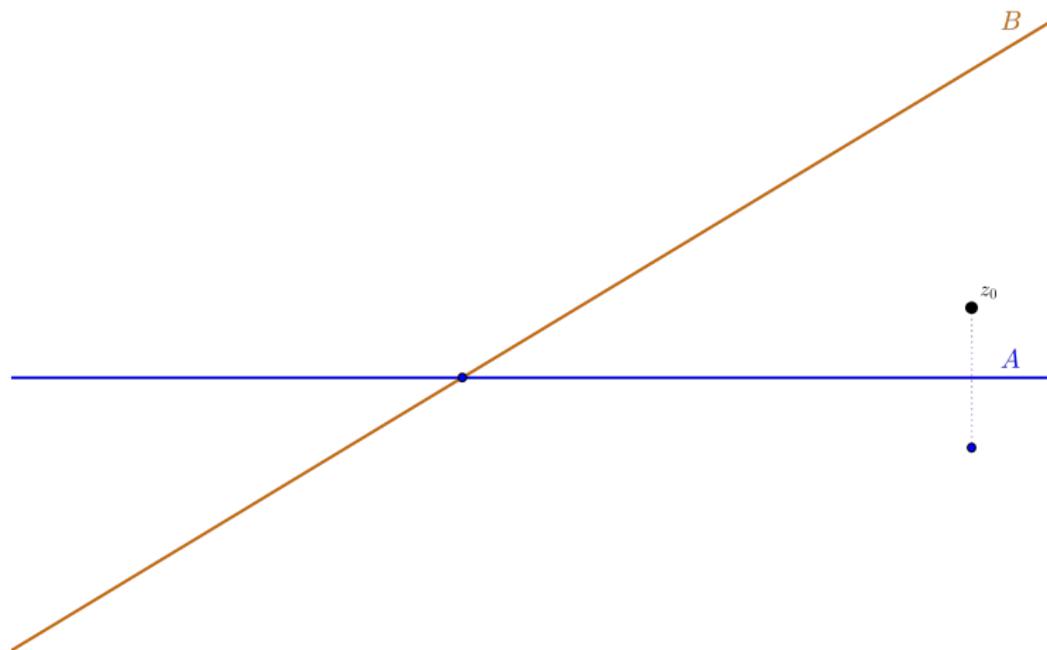
While $(z_n)_{n \in \mathbb{N}}$ “spirals” towards the origin, the “shadow sequence”
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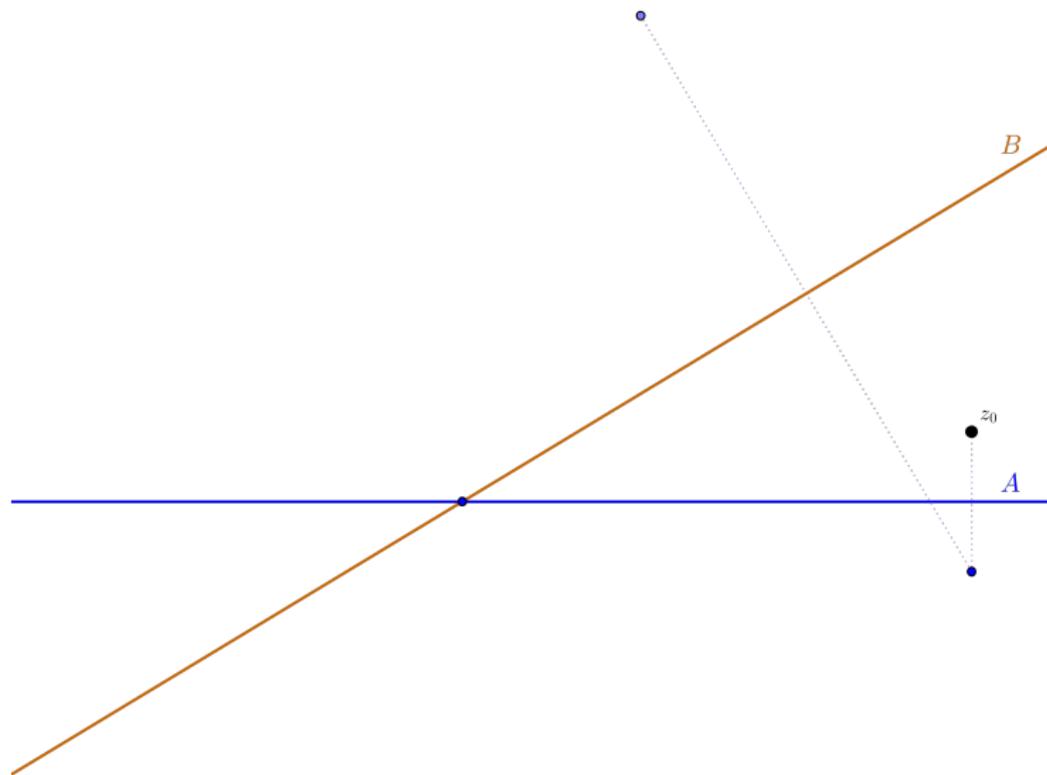
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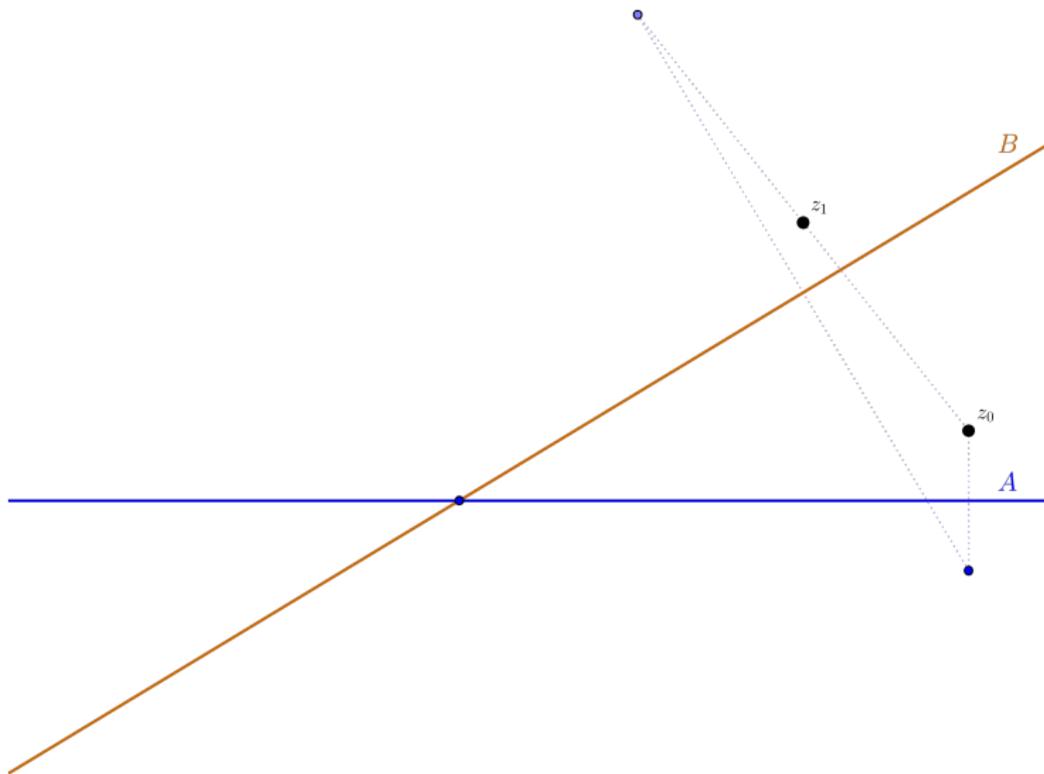
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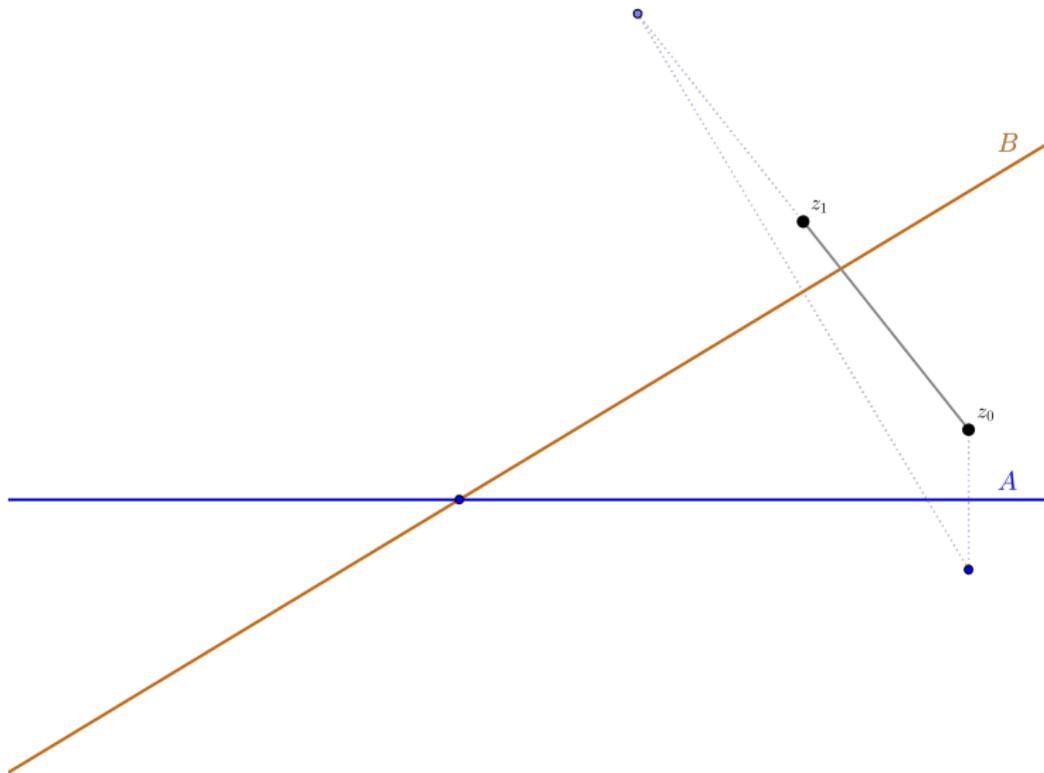
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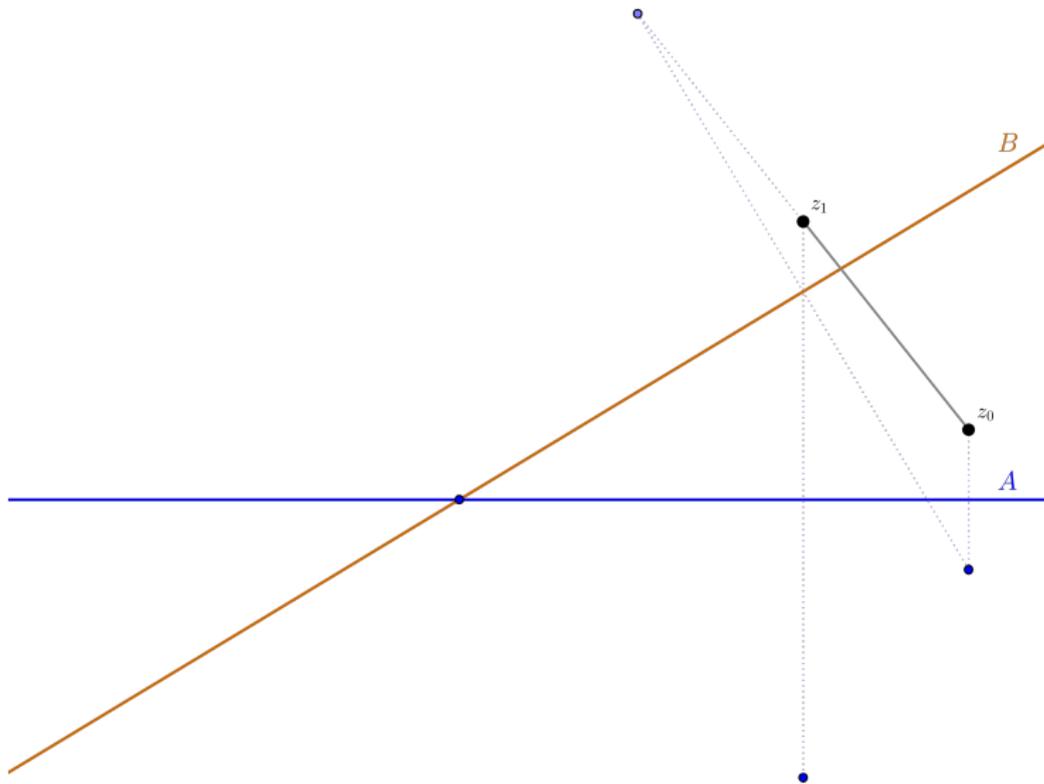
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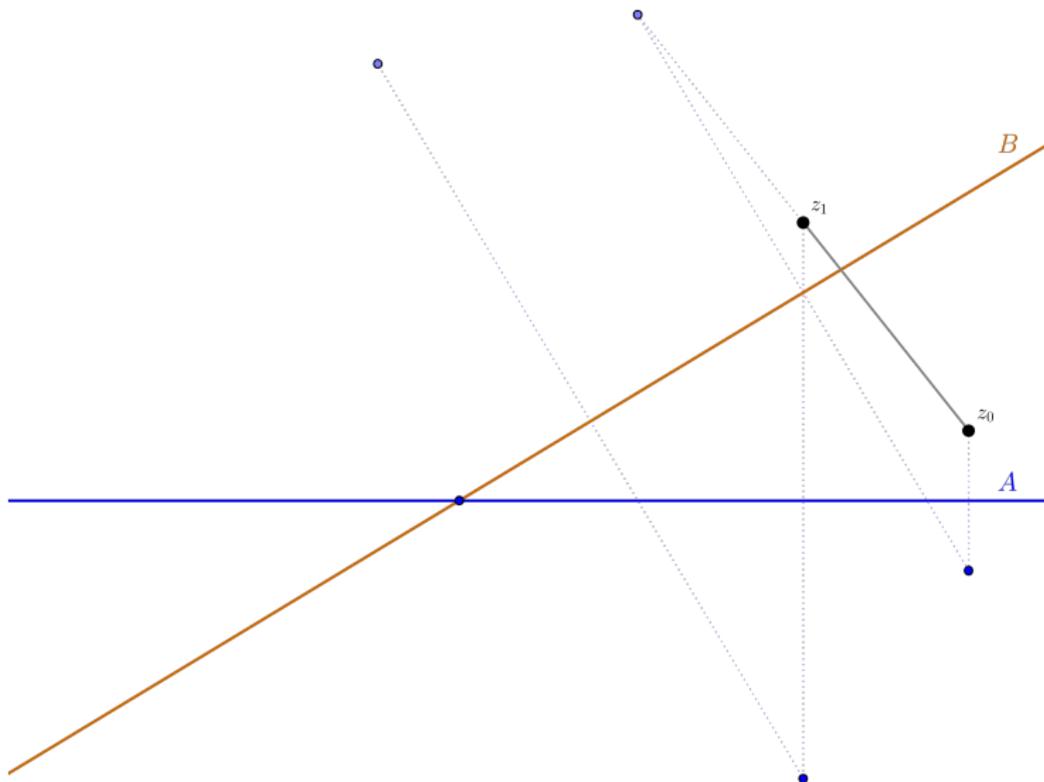
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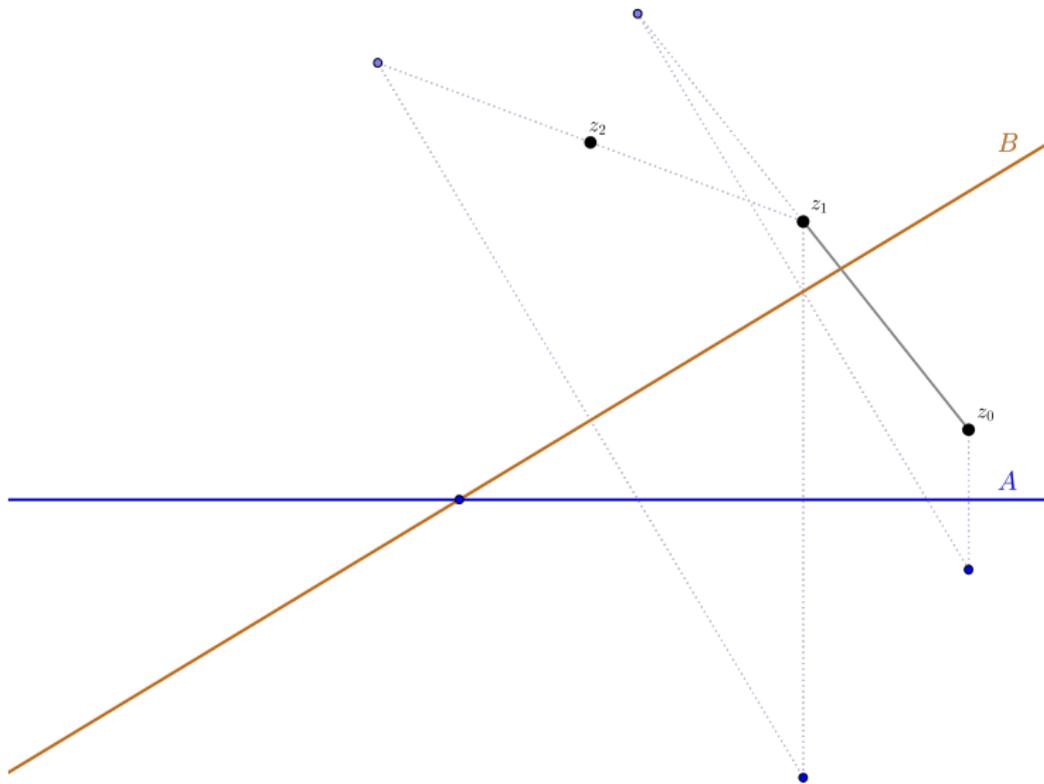
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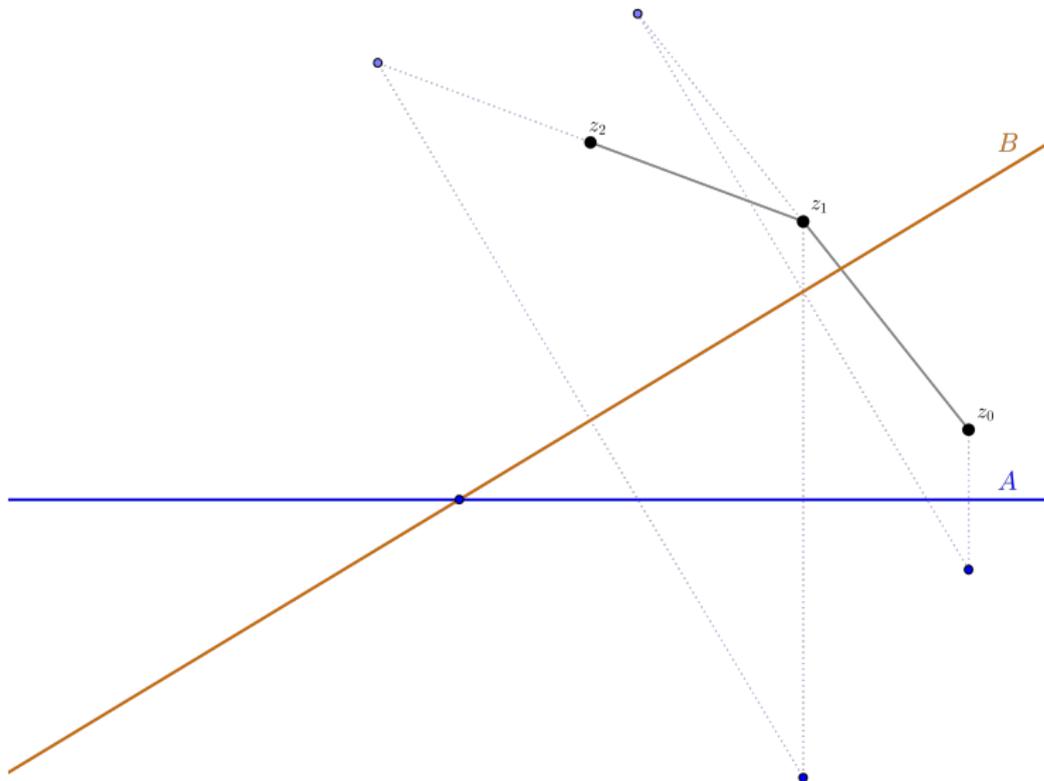
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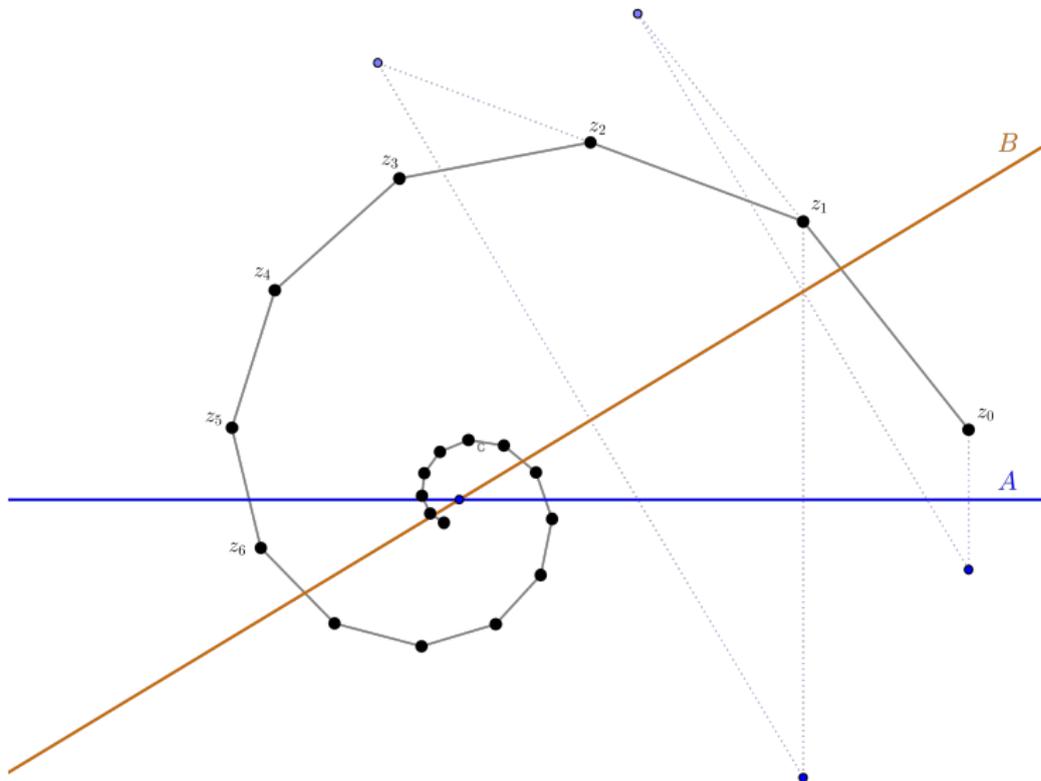
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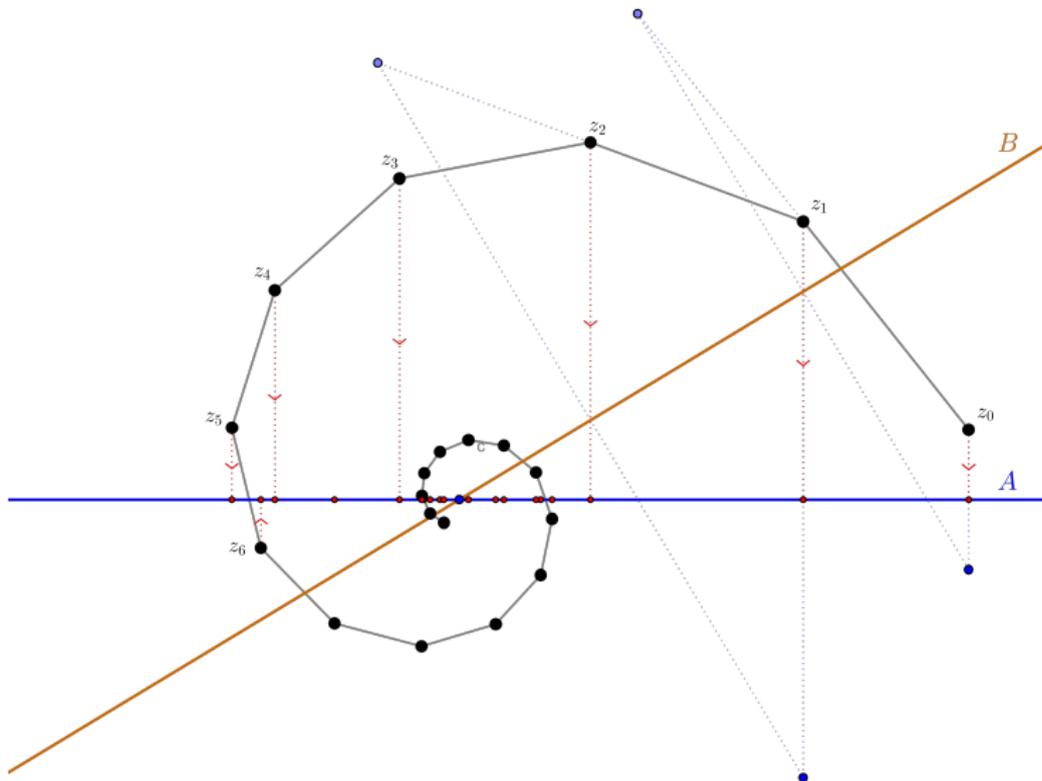
While $(z_n)_{n \in \mathbb{N}}$ “spirals” towards the origin, the “shadow sequence” $(P_{A \setminus B} z_n)_{n \in \mathbb{N}}$ occasionally gets very close to the origin!

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While $(z_n)_{n \in \mathbb{N}}$ “spirals” towards the origin, the “shadow sequence” $(P_{AZ_n})_{n \in \mathbb{N}}$ occasionally gets very close to the origin!

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Convex-convex case

Suppose that A and B are **convex**.

Fact (Lions–Mercier 1979, Svaiter 2011, Bauschke–Combettes–Luke 2004)

If $A \cap B \neq \emptyset$, then $z_n \rightarrow z \in \text{Fix } T = (A \cap B) + N_{A-B}(0)$ and $P_{Az_n} \rightarrow P_{Az} \in A \cap B$; otherwise, $\|z_n\| \rightarrow +\infty$.

Now assume that $g := P_{B-A}0 \in B - A$, or equivalently,

$$E := A \cap (B - g) \neq \emptyset \quad \text{and} \quad F := (A + g) \cap B \neq \emptyset.$$

Fact (Bauschke–Combettes–Luke 2004)

The sequence $(P_{Az_n})_{n \in \mathbb{N}}$ is bounded and its weak cluster points lie in E .

Here $N_{A-B}(0)$ is the normal cone of the set $A - B = \{a - b \mid a \in A, b \in B\}$.

Affine-convex case

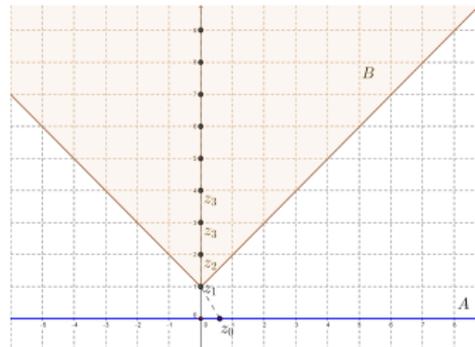
Theorem (Bauschke–D–Moursi 2016, When A is a closed affine subspace)

- 1 $P_A z_n \rightarrow a \in E = A \cap (B - g)$.
- 2 No general conclusion can be drawn about the sequence $(P_B z_n)_{n \in \mathbb{N}}$.

This is strengthened by Bauschke–Moursi 2017 for A convex.

Example ($X = \mathbb{R}^2$, $A = \mathbb{R} \times \{0\}$,
 $B = \text{epi}(|\cdot| + 1)$)

For $z_0 \in [-1, 1] \times \{0\}$, $z_n = (0, n) \in B$,
 and $\|P_B z_n\| = \|z_n\| = n \rightarrow \infty$.



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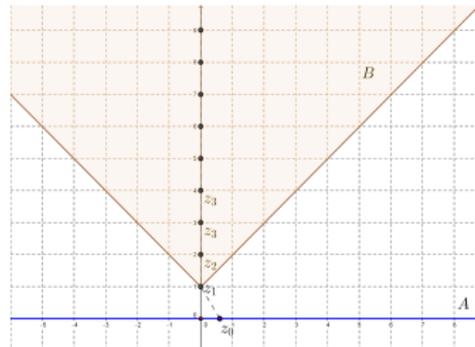
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Spingarn's method

Consider the problem to find a *least-squares solution* of $\bigcap_{j=1}^M C_j$, i.e., to

$$\text{find minimizers of } \sum_{j=1}^M d_{C_j}^2, \quad (1)$$

where C_1, \dots, C_M are nonempty closed convex (possibly nonintersecting) subsets of X with corresponding distance functions d_{C_1}, \dots, d_{C_M} .

Set $\mathbf{X} := X^M$, $\mathbf{A} := \{(x, \dots, x) \in \mathbf{X} \mid x \in X\}$, and $\mathbf{B} := C_1 \times \dots \times C_M$.

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Corollary

Let $(z_n)_{n \in \mathbb{N}}$ be a DR sequence for (\mathbf{A}, \mathbf{B}) . Then

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This generalizes work by Spingarn (1987) who considered **only halfspaces** and whose proof was **much more complicated**.

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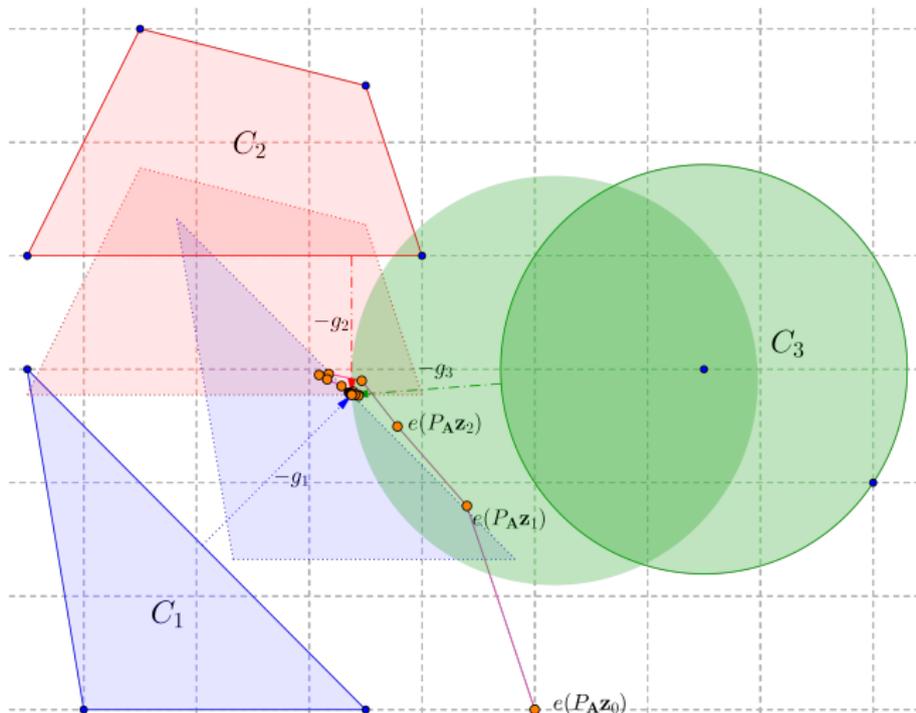
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Find a point in the generalized intersection



Three closed convex sets are shown along with their translations forming the generalized intersection. The first few terms of $(e(P_{\mathbf{A}} z_n))_{n \in \mathbb{N}}$ (yellow points) are also depicted. Here $e : \mathbf{A} \rightarrow \mathbb{R}^2 : (x, x, x) \mapsto x$.

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Presence of Slater's condition ($A \cap \text{int } B \neq \emptyset$)

From now on, X is finite-dimensional.

Lemma

If A and B are convex and $0 \in \text{int}(A - B)$, then $z_n \rightarrow z \in A \cap B$; the convergence is *finite* provided that $z \in A \cap \text{int } B$.

Theorem (Bauschke–D–Noll–Phan 2016, Bauschke–D 2017)

Suppose that $A \cap \text{int } B \neq \emptyset$. Then the DR algorithm *converges finitely* to a point in $A \cap B$ in each of the following cases:

- ① A is an *affine subspace* and B is a *polyhedron*.
- ② $A \in \{X \times \{0\}, X \times \mathbb{R}_+, X \times \mathbb{R}_-\}$ and $B = \text{epi } f$, where $f: X \rightarrow]-\infty, +\infty]$ is convex, l.s.c., and proper.
- ③ A is a *hyperplane/halfspace* and B is a *finite intersection of closed balls* B_j such that $(\forall x \in A \cap \text{bdry } B)(\exists ! B_j) x \in \text{bdry } B_j$.

$\text{int } C$: the interior of C .

Absence of Slater's condition

- ▶ In the case of an affine subspace and a polyhedron, if the Slater's condition is replaced by " $A \cap \text{ri } B \neq \emptyset$ ", then **finite convergence fails** in general, e.g., the case of two lines in \mathbb{R}^2 .
- ▶ If $A \in \{X \times \{0\}, X \times \mathbb{R}_-\}$ and $B = \text{epi } f$, where $\inf_X f \geq 0$ and f is differentiable at its minimizers, then $(P_A z_n)_{n \in \mathbb{N}}$ and hence $(z_n)_{n \in \mathbb{N}}$ **do not converge finitely** whenever $z_0 = (x_0, \rho_0) \in B$ with $x_0 \notin \text{argmin } f$.

Theorem (Bauschke–D 2017)

Suppose that A is a *hyperplane/halfspace* and that $A \cap B \neq \emptyset$. Then the DR sequence *converges finitely* to a point in $z \in \text{Fix } T_{A,B}$ with $P_A z \in A \cap B$ in each of the following cases:

- 1 B is a halfspace of X .
- 2 $X = \mathbb{R}^2$, and B is a polyhedral set.

$\text{ri } C$: the interior of C relative to the affine hull of C .

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When one set is finite

Suppose that B is a **finite subset** of X and let $(z_n)_{n \in \mathbb{N}}$ be a DR sequence for (A, B) .

Theorem (Bauschke–D 2017)

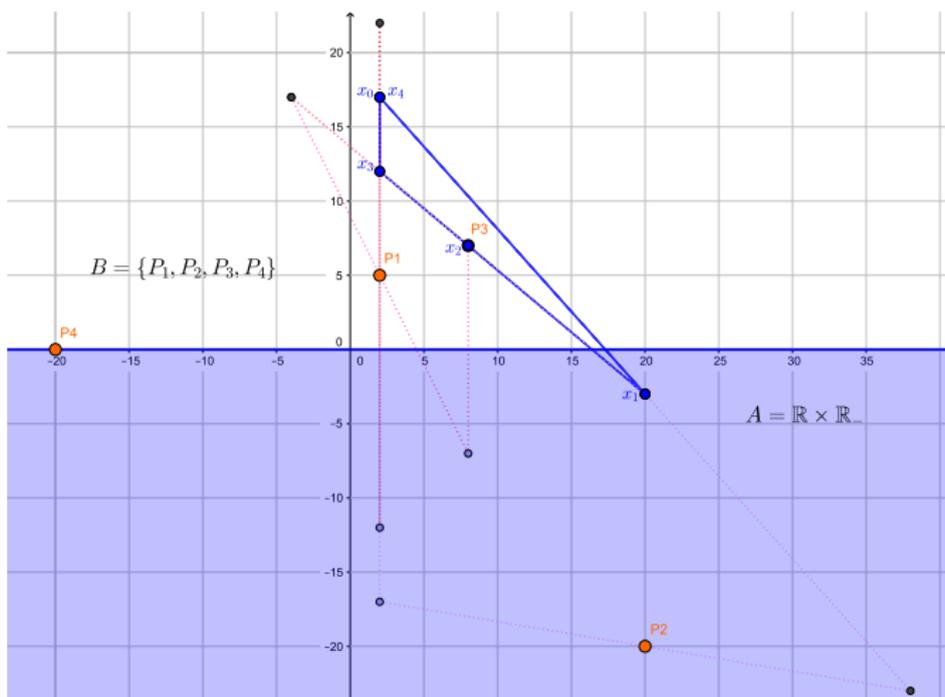
If A is *an affine subspace/a halfspace*, $A \cap B \neq \emptyset$, and the sequence $(z_n)_{n \in \mathbb{N}}$ is *asymptotically regular*, i.e., $z_n - z_{n+1} \rightarrow 0$, then $(z_n)_{n \in \mathbb{N}}$ converges in *finitely many steps* to a point $z \in \text{Fix } T_{A,B}$ with $P_{AZ} \in A \cap B$.

Theorem (Bauschke–D 2017)

If A is a *hyperplane/halfspace* and B is *contained in one of two halfspaces* generated by A , then either

- ① $(z_n)_{n \in \mathbb{N}}$ *converges finitely* to a point $z \in \text{Fix } T_{A,B}$ with $P_{AZ} \in A \cap B$,
or
- ② $A \cap B = \emptyset$ and $\|z_n\| \rightarrow +\infty$ in which case $(P_{AZ_n})_{n \in \mathbb{N}}$ *converges finitely* to a *best approximation solution* $a \in A$ relative to A and B .

Without asymptotic regularity or “one-side” property



A 4-cycle of the DR algorithm for a halfspace and a finite set.
 Interchanging the roles of two sets gives finite convergence, as shown by
[Aragón Artacho–Borwein–Tam 2016](#).

Periodic behavior

Theorem (Bauschke–D–Lindstrom 2017)

Suppose that A is a *hyperplane* and that $B = \{b_1, b_2\}$, where b_1 and b_2 do not belong to the same halfspace generated by A . Let $(z_n)_{n \in \mathbb{N}}$ be a DR sequence for (A, B) . Then

- 1 $(z_n)_{n \in \mathbb{N}}$ does not converge.
- 2 $(z_n)_{n \in \mathbb{N}}$ *cycles after certain steps* regardless the starting point if and only if there exist $k_1, k_2 \in \mathbb{N} \setminus \{0\}$ such that $k_1 d_A(b_1) = k_2 d_A(b_2)$.

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The case of a line and a circle

Fact (Borwein–Sims 2011)

Let $\alpha \in [0, 1[$. Then the DR algorithm for

$$A = \mathbb{R} \times \{\alpha\} \quad \text{and} \quad B = \{(x, \rho) \in \mathbb{R}^2 \mid x^2 + \rho^2 = 1\}$$

is *locally convergent* around $(\pm\sqrt{1-\alpha^2}, \alpha)$.

Conjecture (BS11): The DR algorithm is actually *globally convergent*. This has since been resolved in the affirmative by Benoist (2015).

Idea: Consider $V: \mathbb{R}^2 \rightarrow]-\infty, +\infty]$ given by

$$V(x, \rho) := \frac{1}{2}x^2 - (1-\alpha) \ln|x| + \alpha\sqrt{1-x^2} - \alpha \ln(1+\sqrt{1-x^2}) + \frac{1}{2}(\rho-\alpha)^2.$$

Then V decreases along DR sequences: $V(T_{A,B}z) \leq V(z)$ with equality if and only if $z \in \text{Fix } T_{A,B}$.

The case of a line and a circle

Fact (Borwein–Sims 2011)

Let $\alpha \in [0, 1[$. Then the DR algorithm for

$$A = \mathbb{R} \times \{\alpha\} \quad \text{and} \quad B = \{(x, \rho) \in \mathbb{R}^2 \mid x^2 + \rho^2 = 1\}$$

is *locally convergent* around $(\pm\sqrt{1-\alpha^2}, \alpha)$.

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Finding a zero of a function

In this section,

$f: X \rightarrow [-\infty, +\infty]$ is proper with closed graph.

Consider the feasibility problem in $X \times \mathbb{R}$ with constraints

$$A = X \times \{0\} \quad \text{and} \quad B = \text{gra } f := \{(x, \rho) \in X \times \mathbb{R} \mid f(x) = \rho\},$$

which can be cast as

find $x \in X$ such that $f(x) = 0$.

- ▶ B is generally not convex unless f is affine.
- ▶ **For a line and a circle:** Up to symmetry, take $f(x) = -\sqrt{1-x^2} + \alpha$.

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A Lyapunov-type approach

Definition (Lyapunov-type function)

A function $V: X \times \mathbb{R} \rightarrow]-\infty, +\infty]$ is a **Lyapunov-type function** for f on a nonempty convex subset D of X if it can be expressed in the form

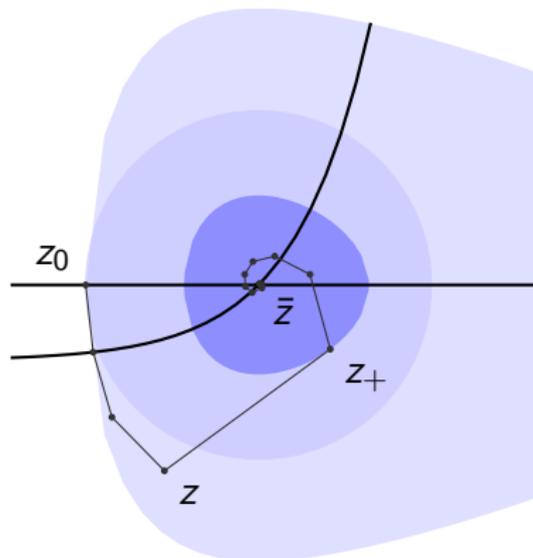
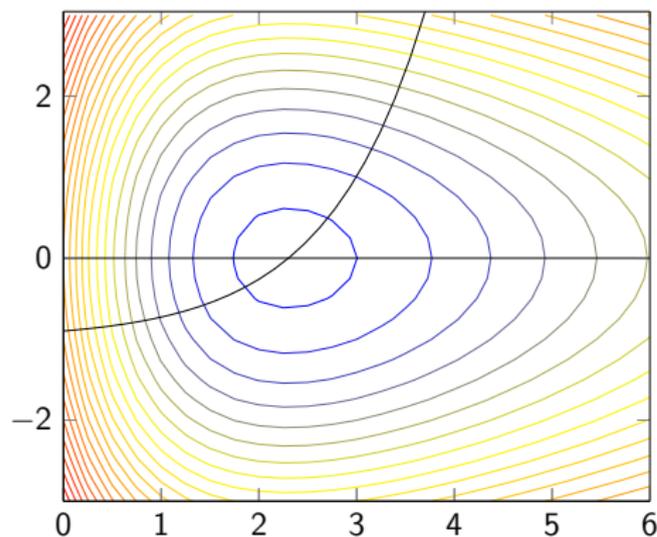
$$V(x, \rho) = F(x) + \frac{1}{2}\rho^2$$

for some proper coercive convex function $F: D \rightarrow]-\infty, +\infty]$ whose subdifferential satisfies

$$(\forall x \in D) \quad \partial F(x) \supseteq \begin{cases} \left\{ \frac{f(x)}{\|x^*\|^2} x^* \mid x^* \in \partial^0 f(x) \right\} & \text{if } 0 \notin \partial^0 f(x), \\ \{0\} & \text{if } f(x) = 0. \end{cases}$$

$\partial^0 f := \partial f \cup -\partial(-f)$: the symmetric (limiting) subdifferential of f .

A Lyapunov-type approach: Some intuition



A Lyapunov-type function for $f(x) = \frac{1}{10} \exp(x) - 1$, which guarantees global convergence of the DR algorithm to $\bar{z} := (\ln(10), 0)$.

Convergence theorem

Write $z_n = (x_n, \rho_n) \in X \times \mathbb{R}$. Suppose that there exists a **Lyapunov-type function** for f on D , that f is **locally Lipschitz continuous** on $D \setminus f^{-1}(0)$, and that

$$(\exists n_0 \in \mathbb{N})(\forall n \geq n_0) \quad x_n \in D \quad \text{and} \quad x_{n+1} \notin (\partial^0 f)^{-1}(0) \setminus f^{-1}(0).$$

Theorem (D–Tam 2017)

The DR sequence $(z_n)_{n \in \mathbb{N}}$ is *bounded and asymptotically regular*, and each of its cluster points \bar{z} satisfy $P_A \bar{z} \in A \cap B$. Suppose, in addition, that $\bar{D} \cap f^{-1}(0) = \{\bar{x}\}$ is contained in D . Then

- ① $z_n \rightarrow \bar{z}$ with $P_A \bar{z} \in A \cap B$.
- ② $z_n \rightarrow \bar{z} = (\bar{x}, 0) \in A \cap B$ provided that $0 \notin \partial^0 f(\bar{x})$ and $f|_D$ is continuous at \bar{x} .

Linear convergence

Corollary

Suppose that $\bar{D} \cap f^{-1}(0) = \{\bar{x}\} \subseteq D$. Then

- 1 If f is **continuously differentiable** around \bar{x} with $\nabla f(\bar{x}) \neq 0$, then $z_n \rightarrow \bar{z} = (\bar{x}, 0) \in A \cap B$ with **R -linear rate**.
- 2 If $X = \mathbb{R}$ and f is **twice strictly differentiable** at \bar{x} with $f'(\bar{x}) \neq 0$, then $z_n \rightarrow \bar{z} = (\bar{x}, 0) \in A \cap B$ with **Q -linear rate**

$$\kappa := \frac{1}{\sqrt{1 + |f'(\bar{x})|^2}}.$$

A sequence $(z_n)_{n \in \mathbb{N}}$ is said to converge to a point \bar{z}

- ▶ with **R -linear rate** $\kappa \in [0, 1[$ if $(\exists \eta \in \mathbb{R}_+)(\forall n \in \mathbb{N}) \|z_n - \bar{z}\| \leq \eta \kappa^n$;
- ▶ with **Q -linear rate** $\kappa \in [0, 1[$ if $\limsup_{n \rightarrow \infty} \frac{\|z_{n+1} - \bar{z}\|}{\|z_n - \bar{z}\|} \leq \kappa$.

Some examples

- $X = \mathbb{R}$, $f(x) = \alpha \exp(x) - \beta$ with $(\alpha, \beta) \in \mathbb{R}_{++}^2$. One possible F is

$$F(x) := \int \frac{f(x)}{f'(x)} dx = \int \left(1 - \frac{\beta}{\alpha} \exp(-x)\right) dx = x + \frac{\beta}{\alpha} \exp(-x)$$

→ Global Q-linear convergence with rate $\kappa = 1/\sqrt{1 + \beta^2}$.

- $X = \mathbb{R}$, $p \in]1, +\infty[$,

$$f(x) := \begin{cases} x^p & \text{if } x \geq 0, \\ x & \text{if } x < 0, \end{cases} \quad \partial^0 f(x) = \begin{cases} px^{p-1} & \text{if } x > 0, \\ [0, 1] & \text{if } x = 0, \\ 1 & \text{if } x < 0. \end{cases}$$

Note that f is nonconvex and nonsmooth at $x = 0$. Define F by

$$F(x) := \begin{cases} \frac{1}{2p} x^2 & \text{if } x \geq 0, \\ \frac{1}{2} x^2 & \text{if } x < 0, \end{cases}$$

then V is a Lyapunov-type function for f on \mathbb{R} .

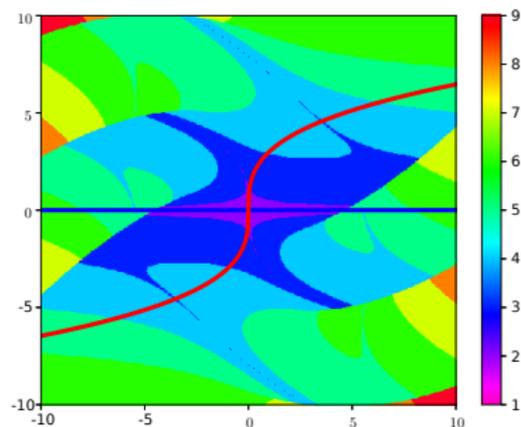
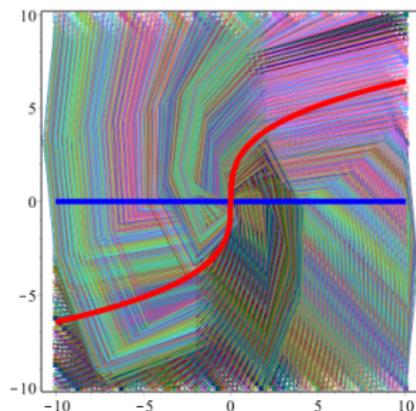
Some examples

Suppose $f = \alpha \|\cdot\|^p$ for $\alpha \in \mathbb{R} \setminus \{0\}$ and $p \in]0, +\infty[$. Then whenever $x \neq 0$, we have $\partial^0 f(x) = \{\alpha p \|x\|^{p-2} x\}$ and

$$\frac{f(x)}{\|\nabla f(x)\|} \nabla f(x) = \frac{\alpha \|x\|^p}{\alpha^2 p^2 \|x\|^{2p-2}} \alpha p \|x\|^{p-2} x = \frac{1}{p} x,$$

which leads to $F(x) = \frac{1}{2p} \|x\|^2$. The global convergence follows.

The same function F works for $f = \alpha |\cdot|^p \operatorname{sgn}(\cdot)$ on \mathbb{R} .



Illustrations of the DR algorithm for $f(x) = 3\sqrt[3]{x}$ on $[-10, 10] \times [-10, 10]$.

Outline

- 1 Introduction
- 2 Behavior of DR algorithm in possibly inconsistent case
- 3 Finite convergence
- 4 A Lyapunov-type approach to convergence theory
- 5 Local linear convergence**

Generalized DR operator

Let $\lambda, \mu \in]0, 2]$ and $\alpha \in]0, 1]$. The **generalized DR operator** for (A, B) with parameters (λ, μ, α) is defined by

$$T_{\lambda, \mu}^{\alpha} := (1 - \alpha) \text{Id} + \alpha P_B^{\mu} P_A^{\lambda}.$$

- ▶ $T_{1,1}^1 = P_B P_A$ is the **classical alternating projection (AP) operator**.
- ▶ $T_{2,2}^{1/2} = \frac{1}{2}(\text{Id} + R_B R_A)$ is the **classical DR operator**.
- ▶ $T_{2,2\alpha}^{1/2} = (1 - \alpha)P_A + \frac{\alpha}{2}(\text{Id} + R_B R_A)$ is the **relaxed averaged alternating reflection (RAAR) operator**.
- ▶ If B is an affine subspace of X , then

$$T_{1+\alpha, 1+\alpha}^{1/(1+\alpha)} = (1 - \alpha)P_B P_A + \frac{\alpha}{2}(\text{Id} + R_B R_A)$$

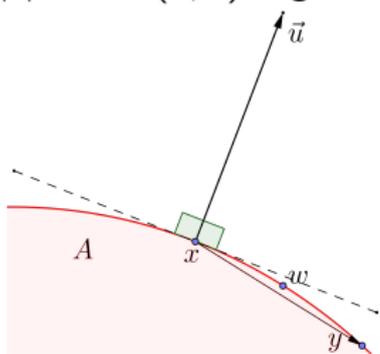
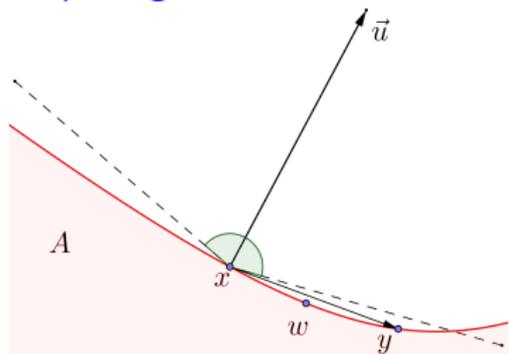
(a convex combination of the classical AP and DR operators).

Regularity of sets

Let $\varepsilon \in \mathbb{R}_+$ and $\delta \in \mathbb{R}_{++}$. A set C is said to be (ε, δ) -regular at $w \in X$ if

$$\forall x, y \in C \cap \mathbb{B}_\delta(w), \forall u \in N_C^{\text{prox}}(x) : \quad \langle u, y - x \rangle \leq \varepsilon \|u\| \cdot \|y - x\|$$

and **superregular** at w if $\forall \varepsilon \in \mathbb{R}_{++}, \exists \delta \in \mathbb{R}_{++} : C$ is (ε, δ) -regular at w .



► Convex sets and sets with “smooth” boundary are superregular.

$$N_C^{\text{prox}}(x) := \text{cone}(P_C^{-1}(x) - x) = \{ \lambda(z - x) \mid z \in P_C^{-1}(x), \lambda \in \mathbb{R}_+ \}.$$

Key properties

Lemma

Let $\varepsilon_1 \in [0, 1/3]$, $\varepsilon_2 \in [0, 1[$ and set $\gamma := 1 - \alpha + \alpha \left(1 + \frac{\lambda \varepsilon_1}{1 - \varepsilon_1}\right) \left(1 + \frac{\mu \varepsilon_2}{1 - \varepsilon_2}\right)$, $\beta := \frac{1 - \alpha}{\alpha}$. If A and B are (ε_1, δ) - and $(\varepsilon_2, \sqrt{2}\delta)$ -regular at $w \in A \cap B$, then $T_{\lambda, \mu}^\alpha$ is $(A \cap B \cap \mathbb{B}_\delta(w), \gamma, \beta)$ -quasi firmly Fejér monotone on $\mathbb{B}_{\delta/2}(w)$ in the sense that

$$\forall x \in \mathbb{B}_{\delta/2}(w), \forall x_+ \in T_{\lambda, \mu}^\alpha x, \forall \bar{x} \in A \cap B \cap \mathbb{B}_\delta(w) :$$

$$\|x_+ - \bar{x}\|^2 + \beta \|x - x_+\|^2 \leq \gamma \|x - \bar{x}\|^2.$$

Lemma

Let $\varepsilon \in [0, 1/3]$. If A is *superregular* at w and $\{A, B\}$ is *strongly regular* at $w \in A \cap B$, then there exist $\delta \in \mathbb{R}_{++}$, $\nu \in \mathbb{R}_{++}$ such that $T_{\lambda, \mu}^\alpha$ is $(A \cap B, \nu)$ -quasi coercive on $\mathbb{B}_{\delta/2}(w)$ in the sense that

$$\forall x \in \mathbb{B}_{\delta/2}(w), \forall x_+ \in T_{\lambda, \mu}^\alpha x : \quad \|x - x_+\| \geq \nu d_{A \cap B}(x).$$

$\{A, B\}$ is strongly regular at $w \in A \cap B$ if $N_A(w) \cap (-N_B(w)) = \{0\}$.

Local linear convergence

Let A and B be closed subsets of X with $A \cap B \neq \emptyset$. Suppose that $\{A, B\}$ is **superregular** and **strongly regular** at some point $w \in A \cap B$.

Fact (Phan 2016)

When started at a point sufficiently close to w , the DR sequence converges R -linearly to a point in $A \cap B$.

Theorem (D–Phan 2016)

*Let $\lambda, \mu \in]0, 2]$ and $\alpha \in]0, 1[$. Then when started at a point sufficiently close to w , the **generalized DR sequence** generated by $T_{\lambda, \mu}^{\alpha}$ converges R -linearly to a point in $A \cap B$.*

Some key references

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Contact

The manuscripts corresponding to this talk are



H.H. Bauschke and MND, On the finite convergence of the Douglas–Rachford algorithm for solving (not necessarily convex) feasibility problems in Euclidean spaces, *SIAM Journal on Optimization* 27 (2017), 507–537.



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H.H. Bauschke, MND, D. Noll, and H.M. Phan, On Slater's condition and finite convergence of the Douglas–Rachford algorithm for solving convex feasibility problems in Euclidean spaces, *Journal of Global Optimization* 65 (2016), 329–349.



MND and H.M. Phan, Linear convergence of projection algorithms, (2016), arXiv:1609.00341.



MND and M.K. Tam, A Lyapunov-type approach to convergence of the Douglas–Rachford algorithm, (2017), arXiv:1706.04846.

THANK YOU VERY MUCH!

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