

# Introductory Course on Non-smooth Optimisation

## Lecture 05 - Peaceman–Rachford, Douglas–Rachford splitting

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## Problem

Find  $x \in \mathbb{R}^n$  such that  $0 \in A(x) + B(x)$ .

## Assumptions

- $A, B : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$  are maximal monotone.
- the resolvents of  $A, B$  are simple, *i.e.* easy to compute.
- $\text{zer}(A + B) \neq \emptyset$ .

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## Peaceman-Rachford splitting

Let  $z_0 \in \mathbb{R}^n$ ,  $\gamma > 0$ :

$$x_k = \mathcal{J}_{\gamma B}(z_k),$$

$$y_k = \mathcal{J}_{\gamma A}(2x_k - z_k),$$

$$z_{k+1} = z_k + 2(y_k - x_k).$$

- dates back to 1950s for solving numerical PDEs.
- the resolvents of  $A, B$  are evaluated separately.

- given  $x^* \in \text{zer}(A + B)$ , there exists  $z^* \in \mathbb{R}^n$  such that

$$\begin{cases} z^* - x^* \in \gamma A(x^*) \\ x^* - z^* \in \gamma B(x^*) \end{cases} \implies \begin{cases} z^* \in x^* + \gamma A(x^*), \\ 2x^* - z^* \in x^* + \gamma B(x^*). \end{cases}$$

- apply the resolvent

$$\begin{cases} x^* = \mathcal{J}_{\gamma A}(z^*), \\ x^* = \mathcal{J}_{\gamma B}(2x^* - z^*). \end{cases}$$

- equivalent formulation

$$\begin{cases} x^* = \mathcal{J}_{\gamma A}(z^*), \\ z^* = z^* + 2(\mathcal{J}_{\gamma B}(2x^* - z^*) - x^*). \end{cases}$$

- fixed-point iteration

$$\begin{cases} x_k = \mathcal{J}_{\gamma A}(z_k), \\ z_{k+1} = z_k + 2(\mathcal{J}_{\gamma B}(2x_k - z_k) - x_k). \end{cases}$$

**Fixed-point formulation** Recall reflection operator  $\mathcal{R}_{\gamma A} = 2\mathcal{J}_{\gamma A} - \text{Id}$ .

- $y_k = \mathcal{J}_{\gamma A}(2x_k - z_k) = \mathcal{J}_{\gamma A} \circ (2\mathcal{J}_{\gamma B} - \text{Id})(z_k)$ .
- For  $z_k$ ,

$$\begin{aligned} z_{k+1} &= z_k + 2(y_k - x_k) \\ &= z_k + 2(\mathcal{J}_{\gamma A} \circ (2\mathcal{J}_{\gamma B} - \text{Id})(z_k) - \mathcal{J}_{\gamma B}(z_k)) \\ &= 2\mathcal{J}_{\gamma A} \circ (2\mathcal{J}_{\gamma B} - \text{Id})(z_k) - (2\mathcal{J}_{\gamma B} - \text{Id})(z_k) \\ &= (2\mathcal{J}_{\gamma A} - \text{Id}) \circ (2\mathcal{J}_{\gamma B} - \text{Id})(z_k). \end{aligned}$$

## Property

- $\mathcal{R}_{\gamma A} = 2\mathcal{J}_{\gamma A} - \text{Id}$ ,  $\mathcal{R}_{\gamma B} = 2\mathcal{J}_{\gamma B} - \text{Id}$  are non-expansive.
- $\mathcal{T}_{\text{PR}} = \mathcal{R}_{\gamma A} \circ \mathcal{R}_{\gamma B}$  is non-expansive.

**NB:** Cannot guarantee convergence in general.

- Uniform monotonicity:  $\phi : \mathbb{R}_+ \rightarrow [0, +\infty]$  is increasing and vanishes only at 0

$$\langle u - v, x - y \rangle \geq \phi(\|x - y\|), \quad (x, u), (y, v) \in \text{gra}(B).$$

- If  $B$  is uniformly monotone, then  $\text{zer}(A + B) = \{x^*\}$  and  $\text{fix}(\mathcal{T}_{\text{PR}}) \neq \emptyset$ . Moreover

$$\langle x - y, \mathcal{J}_{\gamma B}(x) - \mathcal{J}_{\gamma B}(y) \rangle \geq \|\mathcal{J}_{\gamma B}(x) - \mathcal{J}_{\gamma B}(y)\|^2 + \gamma\phi(\|\mathcal{J}_{\gamma B}(x) - \mathcal{J}_{\gamma B}(y)\|).$$

- Let  $z^* \in \text{fix}(\mathcal{T}_{\text{PR}})$ , then  $x^* = \mathcal{J}_{\gamma A}(z^*)$ , and

$$\begin{aligned} \|z_{k+1} - z^*\|^2 &= \|\mathcal{R}_{\gamma A}\mathcal{R}_{\gamma B}(z_k) - \mathcal{R}_{\gamma A}\mathcal{R}_{\gamma B}(z^*)\|^2 \\ &\leq \|(2\mathcal{J}_{\gamma B} - \text{Id})(z_k) - (2\mathcal{J}_{\gamma B} - \text{Id})(z^*)\|^2 \\ &= \|z_k - z^*\|^2 - 4\langle z_k - z^*, \mathcal{J}_{\gamma B}(z_k) - \mathcal{J}_{\gamma B}(z^*) \rangle + 4\|\mathcal{J}_{\gamma B}(z_k) - \mathcal{J}_{\gamma B}(z^*)\|^2 \\ &\leq \|z_k - z^*\|^2 - 4\gamma\phi(\|\mathcal{J}_{\gamma B}(z_k) - \mathcal{J}_{\gamma B}(z^*)\|). \end{aligned}$$

- $\phi(\|z_k - z^*\|) \rightarrow 0$  and  $\|z_k - z^*\| \rightarrow 0$ .



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To overcome the drawback of Peaceman–Rachford splitting.

## Douglas–Rachford splitting

Let  $z_0 \in \mathbb{R}^n$ ,  $\gamma > 0$ ,  $\lambda \in ]0, 2[$ :

$$x_k = \mathcal{J}_{\gamma B}(z_k),$$

$$y_k = \mathcal{J}_{\gamma A}(2x_k - z_k),$$

$$z_{k+1} = z_k + \lambda(y_k - x_k).$$

- given  $x^* \in \text{zer}(A + B)$ , there exists  $z^* \in \mathbb{R}^n$  such that

$$\begin{cases} z^* - x^* \in \gamma A(x^*) \\ x^* - z^* \in \gamma B(x^*) \end{cases} \implies \begin{cases} z^* \in x^* + \gamma A(x^*), \\ 2x^* - z^* \in x^* + \gamma B(x^*). \end{cases}$$

- apply the resolvent

$$\begin{cases} x^* = \mathcal{J}_{\gamma A}(z^*), \\ x^* = \mathcal{J}_{\gamma B}(2x^* - z^*). \end{cases}$$

- equivalent formulation

$$\begin{cases} x^* = \mathcal{J}_{\gamma A}(z^*), \\ z^* = z^* + (\mathcal{J}_{\gamma B}(2x^* - z^*) - x^*). \end{cases}$$

- fixed-point iteration

$$\begin{cases} x_k = \mathcal{J}_{\gamma A}(z_k), \\ z_{k+1} = z_k + (\mathcal{J}_{\gamma B}(2x_k - z_k) - x_k). \end{cases}$$

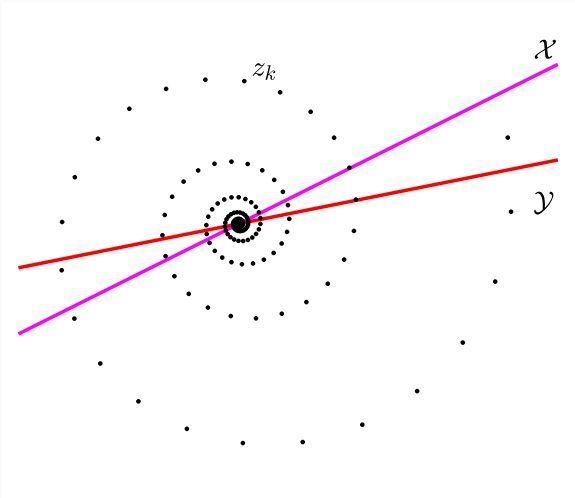
**Fixed-point formulation** Same as PR,  $y_k = \mathcal{J}_{\gamma A} \circ \mathcal{R}_{\gamma B}(z_k)$

$$\begin{aligned} z_{k+1} &= (1 - \lambda)z_k + \lambda(z_k + (y_k - x_k)) \\ &= (1 - \lambda)z_k + \lambda\left(\frac{1}{2}z_k + \frac{1}{2}(z_k + 2(y_k - x_k))\right) \\ &= (1 - \lambda)z_k + \lambda\frac{1}{2}(\text{Id} + \mathcal{R}_{\gamma A} \circ \mathcal{R}_{\gamma B})(z_k). \end{aligned}$$

## Property

- $\mathcal{T}_{\text{DR}} = \frac{1}{2}(\text{Id} + \mathcal{R}_{\gamma A} \circ \mathcal{R}_{\gamma B})$  is firmly non-expansive.
- $\mathcal{T}_{\text{DR}}^\lambda = (1 - \lambda)\text{Id} + \lambda\mathcal{T}_{\text{DR}}$  is  $\frac{\lambda}{2}$ -averaged non-expansive.
- Peaceman–Rachford is the limiting case of Douglas–Rachford,  $\lambda = 2$ .

**NB:** guaranteed convergence if  $\lambda(2 - \lambda) > 0$ .



- Let  $\mathcal{X}, \mathcal{Y}$  be two subspaces

$$\mathcal{X} = \{x : ax = 0\}, \quad \mathcal{Y} = \{x : bx = 0\}$$

and assume

$$1 \leq p \stackrel{\text{def}}{=} \dim(\mathcal{X}) \leq q \stackrel{\text{def}}{=} \dim(\mathcal{Y}) \leq n - 1.$$

- Projection onto subspace

$$\mathcal{P}_{\mathcal{X}}(x) = x - a^T (aa^T)^{-1} ax.$$

- Define diagonal matrices

$$c = \text{diag}(\cos(\theta_1), \dots, \cos(\theta_p)),$$

$$s = \text{diag}(\sin(\theta_1), \dots, \sin(\theta_p)).$$

- Suppose  $p + q < n$ , then there exists orthogonal matrix  $U$  such that

$$\mathcal{P}_{\mathcal{X}} = U \left[ \begin{array}{cc|cc} \text{Id}_p & 0 & 0 & 0 \\ 0 & 0_p & 0 & 0 \\ \hline 0 & 0 & 0_{q-p} & 0 \\ 0 & 0 & 0 & 0_{n-p-q} \end{array} \right] U^*$$

and

$$\mathcal{P}_{\mathcal{Y}} = U \left[ \begin{array}{cc|cc} c^2 & cs & 0 & 0 \\ cs & c^2 & 0 & 0 \\ \hline 0 & 0 & \text{Id}_{q-p} & 0 \\ 0 & 0 & 0 & 0_{n-p-q} \end{array} \right] U^*.$$

- For the composition

$$\mathcal{P}_{\mathcal{X}} \circ \mathcal{P}_{\mathcal{Y}} = U \left[ \begin{array}{cc|cc} c^2 & cs & 0 & 0 \\ 0 & 0_p & 0 & 0 \\ \hline 0 & 0 & 0_{q-p} & 0 \\ 0 & 0 & 0 & 0_{n-p-q} \end{array} \right] U^*$$

and

$$\mathcal{P}_{\mathcal{X}^\perp} \circ \mathcal{P}_{\mathcal{Y}^\perp} = U \left[ \begin{array}{cc|cc} 0_p & 0 & 0 & 0 \\ -cs & c^2 & 0 & 0 \\ \hline 0 & 0 & 0_{q-p} & 0 \\ 0 & 0 & 0 & \text{Id}_{n-p-q} \end{array} \right] U^*.$$



- Fixed-point operator

$$\begin{aligned}\mathcal{T}_{\text{DR}} &= \mathcal{P}_{\mathcal{X}} \circ \mathcal{P}_{\mathcal{Y}} + \mathcal{P}_{\mathcal{X}^\perp} \circ \mathcal{P}_{\mathcal{Y}^\perp} \\ &= U \left[ \begin{array}{cc|cc} c^2 & cs & 0 & 0 \\ -cs & c^2 & 0 & 0 \\ \hline 0 & 0 & 0_{q-p} & 0 \\ 0 & 0 & 0 & \text{Id}_{n-p-q} \end{array} \right] U^*.\end{aligned}$$

- Consider relaxation

$$\begin{aligned}\mathcal{T}_{\text{DR}}^\lambda &= (1 - \lambda)\text{Id} + \lambda\mathcal{T}_{\text{DR}} \\ &= U \left[ \begin{array}{cc|cc} \text{Id}_p - \lambda s^2 & \lambda cs & 0 & 0 \\ -\lambda cs & \text{Id}_p - \lambda s^2 & 0 & 0 \\ \hline 0 & 0 & (1 - \lambda)\text{Id}_{q-p} & 0 \\ 0 & 0 & 0 & \text{Id}_{n-p-q} \end{array} \right] U^*.\end{aligned}$$

- Eigenvalues

$$\sigma(\mathcal{T}_{\text{DR}}^\lambda) = \left\{ \begin{array}{l} \{1 - \lambda \sin^2(\theta_i) \pm i\lambda \cos(\theta_i) \sin(\theta_i) \mid i = 1, \dots, p\} \cup \{1\} : q = p, \\ \{1 - \lambda \sin^2(\theta_i) \pm i\lambda \cos(\theta_i) \sin(\theta_i) \mid i = 1, \dots, p\} \cup \{1\} \cup \{1 - \lambda\} : q > p. \end{array} \right.$$

- Complex eigenvalues

$$|1 - \lambda \sin^2(\theta_i) \pm i\lambda \cos(\theta_i) \sin(\theta_i)| = \sqrt{\lambda(2 - \lambda) \cos^2(\theta_i) + (1 - \lambda)^2}$$

and

$$1 \geq \sqrt{\lambda(2 - \lambda) \cos^2(\theta_i) + (1 - \lambda)^2} \geq |1 - \lambda|.$$

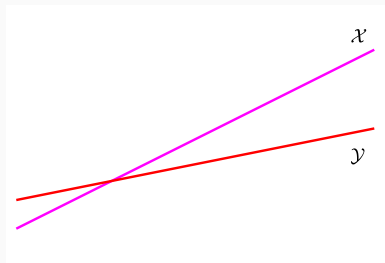
- $\lim_{k \rightarrow +\infty} \mathcal{T}_{\text{DR}}^k = \mathcal{T}_{\text{DR}}^\infty$  and  $\mathbf{z}_k - \mathbf{z}^* = (\mathcal{T}_{\text{DR}} - \mathcal{T}_{\text{DR}}^\infty)(\mathbf{z}_{k-1} - \mathbf{z}^*)$ .

- Spectral radius, **minimises at  $\lambda = 1$**

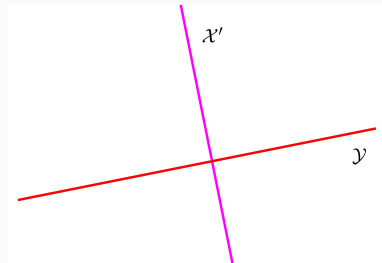
$$\rho(\mathcal{T}_{\text{DR}} - \mathcal{T}_{\text{DR}}^\infty) = \sqrt{\lambda(2 - \lambda) \cos^2(\theta_i) + (1 - \lambda)^2}.$$

- $\widetilde{\mathcal{T}}_{\text{DR}} = \mathcal{T}_{\text{DR}} - \mathcal{T}_{\text{DR}}^\infty$

$$\begin{aligned} \|\mathbf{z}_k - \mathbf{z}^*\| &= \|\widetilde{\mathcal{T}}_{\text{DR}} \mathbf{z}_{k-1} - \widetilde{\mathcal{T}}_{\text{DR}} \mathbf{z}^*\| = \dots = \|\widetilde{\mathcal{T}}_{\text{DR}}^k (\mathbf{z}_0 - \mathbf{z}^*)\| \\ &\leq C(\rho(\widetilde{\mathcal{T}}_{\text{DR}}))^k \|\mathbf{z}_0 - \mathbf{z}^*\|. \end{aligned}$$



$\mathcal{X}$  and  $\mathcal{Y}$



$\mathcal{X}'$  and  $\mathcal{Y}$

**Optimal metric** A invertable operation which makes the Friedrichs angle between  $\mathcal{X}'$  and  $\mathcal{Y}$  the largest, e.g.  $\frac{\pi}{2} \dots$

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**Problem**  $s \in \mathbb{N}_+$  and  $s \geq 2$

Find  $x \in \mathbb{R}^n$  such that  $0 \in \sum_i A_i(x)$ .

### Assumptions

- for each  $i = 1, \dots, s$ ,  $A_i : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$  is maximal monotone.
- $\text{zer}(\sum_i A_i) \neq \emptyset$ .

- Let  $\mathcal{H} = \underbrace{\mathbb{R}^n \times \cdots \times \mathbb{R}^n}_{s \text{ times}}$  endowed with the scalar inner-product and norm

$$\forall \mathbf{x}, \mathbf{y} \in \mathcal{H}, \langle \mathbf{x}, \mathbf{y} \rangle = \sum_{i=1}^s \langle x_i, y_i \rangle, \|\mathbf{x}\| = \sqrt{\sum_{i=1}^s \|x_i\|^2}.$$

- Let

$$\mathcal{S} = \{\mathbf{x} = (x_i)_i \in \mathcal{H} : x_1 = \cdots = x_s\}$$

and its orthogonal complement

$$\mathcal{S}^\perp = \{\mathbf{x} = (x_i)_i \in \mathcal{H} : \sum_{i=1}^s x_i = 0\}.$$

Define  $\mathbf{A}$  by

$$\mathbf{A}(\mathbf{x}) : \mathbf{x} \in \mathcal{H} \rightarrow A_1(x_1) \times \cdots \times A_s(x_s).$$

**Lifted problem**

Find  $\mathbf{x} \in \mathcal{H}$  such that  $0 \in \mathbf{A}(\mathbf{x}) + \mathcal{N}_{\mathcal{S}}(\mathbf{x})$ .

- the resolvent of  $\mathbf{A}$  is separable, i.e.  $\mathcal{J}_{\gamma \mathbf{A}} = (\mathcal{J}_{\gamma A_i})_i$ .
- define the canonical isometry,

$$\mathbf{C} : \mathbb{R}^n \rightarrow \mathcal{S}, \mathbf{x} \mapsto (x, \cdots, x),$$

then  $\mathcal{P}_{\mathcal{S}}(\mathbf{z}) = \mathbf{C}(\frac{1}{s} \sum_{i=1}^s z_i)$ .

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**DR in product space** for  $\mathbf{x}^* \in \mathcal{S}$ ,  $\exists -\mathbf{v} \in \mathcal{S}$  such that

$$-\mathbf{v} \in \mathcal{S}^\perp = \mathcal{N}_{\mathcal{S}}(\mathbf{x}^*) \quad \text{and} \quad \mathbf{v} \in \mathbf{A}(\mathbf{x}^*).$$

**Problem**  $V$  is a close subspace

$$\text{Find } \mathbf{x} \in V \text{ and } \mathbf{v} \in V^\perp \text{ such that } \mathbf{v} \in \mathbf{A}(\mathbf{x}).$$

## Assumptions

- $\mathbf{A} : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$  is maximal monotone.
- admits at least one solution.

## Partial inverse

Let  $A : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$  be set-valued and  $V \subseteq \mathbb{R}^n$  be a closed subspace. The partial inverse of  $A$  respect to  $V$  is the operator  $A_V : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$  define by

$$\text{gra}(A_V) = \{ (\mathcal{P}_V(x) + \mathcal{P}_{V^\perp}(u), \mathcal{P}_{V^\perp}(x) + \mathcal{P}_V(u)) : (x, u) \in \text{gra}(A) \}.$$

**Example** Let  $A : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$ , then  $A_{\mathbb{R}^n} = A$  and  $A_{\{0\}} = A^{-1}$ .

An application of Proximal Point Algorithm.

## Spingarn

Let  $x_0 \in V$ ,  $u_0 \in V^\perp$ :

$$y_k = \mathcal{J}_A(x_k + u_k),$$

$$v_k = x_k + u_k - y_k,$$

$$(x_{k+1}, u_{k+1}) = (\mathcal{P}_V(y_k), \mathcal{P}_{V^\perp}(v_k)).$$

- define mapping

$$L : \mathbb{R}^n \oplus \mathbb{R}^n \rightarrow \mathbb{R}^n \oplus \mathbb{R}^n : (x, u) \rightarrow (\mathcal{P}_V(x) + \mathcal{P}_{V^\perp}(u), \mathcal{P}_{V^\perp}(x) + \mathcal{P}_V(u)).$$

■

$$\begin{aligned} p = \mathcal{J}_{A_V}(x) &\iff (p, x - p) \in \text{gra}(A_V) \\ &\iff L(p, x - p) \in L(\text{gra}(A_V)) = \text{gra}(A) \\ &\iff (\mathcal{P}_V(p) + \mathcal{P}_{V^\perp}(x - p), \mathcal{P}_V(x - p) + \mathcal{P}_{V^\perp}(p)) \in \text{gra}(A). \end{aligned}$$

- let  $q = \mathcal{P}_V(p) + \mathcal{P}_{V^\perp}(x - p)$

$$\begin{aligned} p = \mathcal{J}_{A_V}(x) &\iff x - q = \mathcal{P}_V(x - p) + \mathcal{P}_{V^\perp}p \in A(q) \\ &\iff q = \mathcal{J}_A(x). \end{aligned}$$

- let  $z_k = x_k + u_k$ , since  $x_k \in V$  and  $u_k \in V^\perp$

$$\begin{aligned} \mathcal{P}_V(z_{k+1}) + \mathcal{P}_{V^\perp}(z_k - z_{k+1}) &= x_{k+1} + \mathcal{P}_{V^\perp}(u_k) - u_{k+1} \\ &= \mathcal{P}_V(y_k) + \mathcal{P}_{V^\perp}(v_k - x_k + y_k) - \mathcal{P}_{V^\perp}(v_k) \\ &= \mathcal{P}_V(y_k) + \mathcal{P}_{V^\perp}(v_k) + \mathcal{P}_{V^\perp}(y_k) - \mathcal{P}_{V^\perp}(v_k). \end{aligned}$$

- $z_{k+1} = \mathcal{J}_A(z_k)$ .

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## An inertial DR splitting

**Initial** :  $x_0 \in \mathbb{R}^n$ ,  $x_{-1} = x_0$  and  $\gamma > 0$ ;

$$\begin{aligned}y_k &= z_k + a_{0,k}(z_k - z_{k-1}) + a_{1,k}(z_{k-1} - z_{k-2}) + \cdots, \\z_{k+1} &= \mathcal{T}_{\text{DR}}(y_k)\end{aligned}$$

- relaxation can be applied.

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### Basis pursuit

$$\min_{x \in \mathbb{R}^n} \|x\|_1$$

such that  $Ax = b$ ,

- $A : \mathbb{R}^n \rightarrow \mathbb{R}^m$  with  $m \ll n$ .
- $b \in \text{Img}(A)$ .



## Image inpainting

$$\min_{X \in \mathbb{R}^{n \times n}} \|WX\|_1$$

$$\text{such that } \mathcal{P}_\Omega(X) = \bar{X},$$

- $W$ : total variation, orthonormal basis, redundant wavelet frame.
- Observation constraint

$$(\mathcal{P}_\Omega(X))_{i,j} = \begin{cases} \bar{X}_{i,j} : (i,j) \in \Omega, \\ 0 : (i,j) \notin \Omega. \end{cases}$$

- Painting reconstruction in museum.

## Matrix completion

$$\min_{X \in \mathbb{R}^{n \times n}} \|X\|_*$$

such that  $\mathcal{P}_\Omega(X) = \bar{X}$ ,

- Observation constraint

$$(\mathcal{P}_\Omega(X))_{i,j} = \begin{cases} \bar{X}_{i,j} & : (i,j) \in \Omega, \\ 0 & : (i,j) \notin \Omega. \end{cases}$$

- Netflix prize, recommendation system.

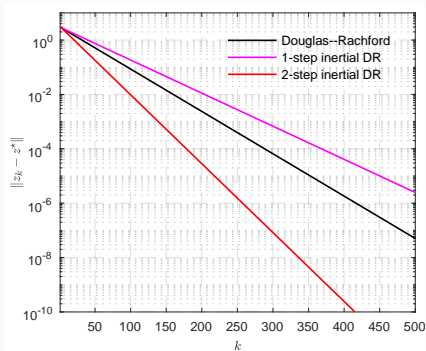
### Variation inequality

Find  $x \in \mathbb{R}^n$  such that  $\exists u \in A(x), \forall y \in \mathbb{R}^n : \langle x - y, u \rangle + R(x) \leq R(y)$ .

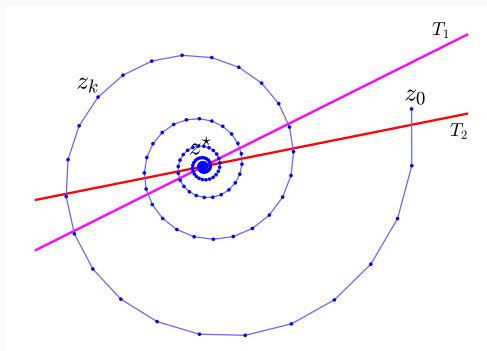
- $R \in \Gamma_0$ .
- $A : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$  is maximal monotone.

**Example** Let  $R, J \in \Gamma_0$ , and  $x^* \in \text{Argmin}(R + J)$ , then  $\exists u \in \partial J(x^*)$  s.t.  $-u \in \partial R(x^*)$  and

$$\begin{aligned} \langle y - x^*, -u \rangle + R(x^*) &\leq R(y) \\ \iff \langle x^* - y, u \rangle + R(x^*) &\leq R(y). \end{aligned}$$



Comparison



Trajectory

- H. H. Bauschke, J. Y. Bello Cruz, T. T. A. Nghia, H. M. Pha, and X. Wang. “Optimal rates of linear convergence of relaxed alternating projections and generalized Douglas–Rachford methods for two subspaces”. *Numerical Algorithms*, 73(1):33–76, 2016.
- H. Bauschke and P. L. Combettes. “Convex Analysis and Monotone Operator Theory in Hilbert Spaces”. Springer, 2011.
- J. Liang. “Convergence rates of first-order operator splitting methods”. Diss. Normandie Université; GREYC CNRS UMR 6072, 2016.