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# Abstract

Monotone operators play important roles in optimization and convex analysis. We define a new average of monotone operators by using resolvents. The new average enjoys self-duality and inherits many nice features of given monotone operators. When the monotone operators are positive definite matrices, the new average lies between the harmonic average and arithmetic average. Appropriate limits of resolvent average lead to both harmonic average and arithmetic average. Consequences on matrix functions are also given.

# Outline of Topics

- 1 Resolvent averages
- 2 Dominant properties of  $\mathcal{R}_\mu(A, \lambda)$ 
  - At most single-valued or strict monotonicity
  - Uniform monotonicity
- 3 Recessive properties of  $\mathcal{R}_\mu(A, \lambda)$ 
  - Paramonotonicity and rectangularity
  - Nonexpansive monotonicity and displacement mapping
- 4 Graphical limits of resolvent averages
- 5 Extensions and relationships

## 5 Extensions and relationships

# What is a resolvent average?

$\mathcal{H}$ : a real Hilbert space with inner product  $\langle \cdot, \cdot \rangle$ .

For a set-valued operator  $A : \mathcal{H} \rightrightarrows \mathcal{H}$ ,

$$\text{dom } A = \{x \mid Ax \neq \emptyset\}, \quad \text{ran } A = \bigcup_{x \in \text{dom } A} Ax,$$

The set-valued inverse  $A^{-1}$  of  $A$ :

$$(y, x) \in \text{gra } A^{-1} \Leftrightarrow (x, y) \in \text{gra } A.$$

The operator  $A$  is called monotone if  $\forall (x_i, x_i^*) \in \text{gra } A, i = 1, 2$ ,

$$\langle x_2^* - x_1^*, x_2 - x_1 \rangle \geq 0,$$

and strictly monotone if this inequality is strict whenever  $x_1 \neq x_2$ .

$$S_A = \text{Id} - 2(\text{Id} + A)^{-1} : \quad \text{Rockafellar-Wets regularization of } A.$$

## Resolvent average

For monotone operators  $A_i, i = 1, \dots, n$  and  $\lambda_i > 0$  with  $\sum_{i=1}^n \lambda_i = 1$ , define

$$A = (A_1, A_2, \dots, A_n),$$

$$\lambda = (\lambda_1, \dots, \lambda_n).$$

**Resolvent average of  $A_j, j = 1, \dots, n$**

$$\mathcal{R}_\mu(A, \lambda) = [\lambda_1(A_1 + \mu^{-1}\text{Id})^{-1} + \cdots + \lambda_n(A_n + \mu^{-1}\text{Id})^{-1}]^{-1} - \mu^{-1}\text{Id},$$





# Harmonic average and arithmetic average

Well-known harmonic average and arithmetic average are

$$\mathcal{H}(A, \lambda) = (\lambda_1 A_1^{-1} + \cdots + \lambda_n A_n^{-1})^{-1},$$

$$\mathcal{A}(A, \lambda) = \lambda_1 A_1 + \cdots + \lambda_n A_n.$$

- ▶ If  $\bigcap_{i \in I} \text{ran } A_i = \emptyset$ , then  $\mathcal{H}(A, \lambda)$  is empty-valued.
- ▶ If  $\bigcap_{i \in I} \text{dom } A_i = \emptyset$ , then  $\mathcal{A}(A, \lambda)$  is empty-valued.

# Why study resolvent averages?

- $\mathcal{R}_\mu(A, \lambda)$  provides a novel method to obtain set-valued operators from known operators  $A_i$ 's.
- It is very interesting to ask what properties  $\mathcal{R}_\mu(A, \lambda)$  has and inherits from those  $A_i$ 's.
- What are the relationships among  $\mathcal{R}_\mu(A, \lambda)$ ,  $\mathcal{H}(A, \lambda)$  and  $\mathcal{A}(A, \lambda)$ ?

# Proximal average of functions: BGLW'08

The proximal average of convex functions is defined by

$$\mathcal{P}_\mu(f, \lambda) = [\lambda_1(f_1 + \mu^{-1}j)^* + \cdots + \lambda_n(f_n + \mu^{-1}j)^*]^* - \mu^{-1}j, \quad (2)$$

for  $f = (f_1, \dots, f_n)$  with  $(\forall i)$   $f_i$  being convex functions.

The proximal mapping of  $\mathcal{P}_\mu(f, \lambda)$  is the average of proximal mappings of  $f_i$ 's, namely

$$\text{Prox}_\mu \mathcal{R}_\mu(A, \lambda) = \lambda_1 \text{Prox}_\mu f_1 + \cdots + \lambda_n \text{Prox}_\mu f_n,$$

with the proximal mapping  $\text{Prox}_\mu f_i = (\mu \partial f_i + \text{Id})^{-1}$ .

# Reformulations of $\mathcal{R}_\mu(A, \lambda)$

## Proposition 1

We have

$${}^\mu[\mathcal{R}_\mu(A, \lambda)] = \lambda_1 {}^\mu A_1 + \cdots + \lambda_n {}^\mu A_n. \quad (3)$$

$$S_{\mu\mathcal{R}_\mu(A, \lambda)} = \lambda_1 S_{\mu A_1} + \cdots + \lambda_n S_{\mu A_n}. \quad (4)$$

In terms of Yosida  $\mu$ -regularization of  $A_i$ 's, we have

## Theorem 2

$$\mathcal{R}_\mu(A, \lambda) = - {}^\mu[-(\lambda_1 {}^\mu A_1 + \cdots + \lambda_n {}^\mu A_n)].$$

### Proposition 3

$x^* \in \mathcal{R}_\mu(A, \lambda)(x)$  if and only if  $(\forall i) \exists x_i \in \text{dom } A_i$  such that

$$\begin{cases} x = \lambda_1 x_1 + \cdots + \lambda_n x_n \\ x^* \in \bigcap_{i=1}^n (A_i + \mu^{-1} \text{Id})(x_i) - \mu^{-1} x. \end{cases} \quad (5)$$

Consequently,  $\forall x \in \mathbb{R}^N$ ,

$$\mathcal{R}_\mu(A, \lambda)(x) = \bigcup \left\{ \bigcap_{i=1}^n (A_i + \mu^{-1} \text{Id})(x_i) - \mu^{-1} x : \sum_{i=1}^n \lambda_i x_i = x \right\}, \quad (6)$$

$$\text{dom } \mathcal{R}_\mu(A, \lambda) \subset \lambda_1 \text{dom } A_1 + \cdots + \lambda_n \text{dom } A_n. \quad (7)$$

Furthermore,  $\forall x \in \mathbb{R}^N, \forall \mu > 0$ ,

$$\bigcap_{i=1}^n A_i(x) \subset \mathcal{R}_\mu(A, \lambda)(x). \quad (8)$$

## Proposition 4

Let  $z^*, z \in \mathbb{R}^N$ . Then

$$\mathcal{R}_\mu((A_1 - z^*, \dots, A_n - z^*), \lambda) = \mathcal{R}_\mu(A, \lambda) - z^*, \quad (9)$$

$$\mathcal{R}_\mu((A_1(\cdot - z), \dots, A_n(\cdot - z)), \lambda) = \mathcal{R}_\mu(A, \lambda)(\cdot - z). \quad (10)$$

## Proposition 5

Let  $\alpha > 0$ . Then

$$\mathcal{R}_\mu(\alpha A, \lambda) = \alpha \mathcal{R}_{\alpha\mu}(A, \lambda). \quad (11)$$

In particular,

$$\mathcal{R}_\mu(A, \lambda) = \mu^{-1} \mathcal{R}_1(\mu A, \lambda). \quad (12)$$

### Example 6

Let  $A_i = N_{C_i} \forall i$ . Then

$$\mathcal{R}_\mu(A, \lambda) = \mu^{-1} [(\lambda_1 P_{C_1} + \cdots + \lambda_n P_{C_n})^{-1} - \text{Id}], \quad (13)$$



# Inverse formula

## Fact 7 (Poliquin-Rockafellar'96)

*Every mapping  $A : \mathcal{H} \rightrightarrows \mathcal{H}$  obeys the identity*

$$\text{Id} - (\text{Id} + A)^{-1} = (\text{Id} + A^{-1})^{-1}. \quad (14)$$

*Indeed, the Yosida regularizations of  $A$  are related to the resolvents of  $A$  by*

$${}^{\mu}A = (\mu \text{Id} + A^{-1})^{-1} = \mu^{-1}[\text{Id} - (\text{Id} + \mu A)^{-1}] \quad \forall \mu > 0. \quad (15)$$

## Theorem 8 (inverse formula: self duality)

Let  $A_i : \mathcal{H} \rightrightarrows \mathcal{H}$  be any set-valued mapping and  $\mu > 0$ . Assume that  $\sum_{i=1}^n \lambda_i = 1$  with  $\lambda_i > 0$ . Then

$$[\mathcal{R}_\mu(A, \lambda)]^{-1} = \mathcal{R}_{\mu^{-1}}(A^{-1}, \lambda), \text{ i.e.,} \quad (16)$$

$$\left[ \left( \lambda_1(A_1 + \mu^{-1}\text{Id})^{-1} + \cdots + \lambda_n(A_n + \mu^{-1}\text{Id})^{-1} \right)^{-1} - \mu^{-1}\text{Id} \right]^{-1} =$$

$$\left( \lambda_1(A_1^{-1} + \mu\text{Id})^{-1} + \cdots + \lambda_n(A_n^{-1} + \mu\text{Id})^{-1} \right)^{-1} - \mu\text{Id}.$$

# Who cares about resolvent averages?

## Theorem 9 (common solutions to monotone inclusions)

Suppose that for each  $i \in I$ ,  $A_i : \mathcal{H} \rightrightarrows \mathcal{H}$  is maximally monotone. Let  $x$  and  $u$  be points in  $\mathcal{H}$ . If  $\bigcap_{i \in I} A_i(x) \neq \emptyset$ , then

$$\mathcal{R}_\mu(A, \lambda)(x) = \bigcap_{i \in I} A_i(x). \quad (17)$$

If  $\bigcap_{i \in I} A_i^{-1}(u) \neq \emptyset$ , then

$$\mathcal{R}_\mu(A, \lambda)^{-1}(u) = \bigcap_{i \in I} A_i^{-1}(u). \quad (18)$$

### Example 10 (convex feasibility problem)

Let  $C_i \subset \mathbb{R}^N$  be non-empty closed convex, and  $A_i = N_{C_i}$ . If  $\bigcap_{i=1}^n C_i \neq \emptyset$ , then

$$\mathcal{R}_\mu(A, \lambda)^{-1}(0) = \bigcap_{i=1}^n C_i.$$

# Who cares about resolvent averages?

## Example 11 (homotopy transform)

Let  $A_1, A_2 : \mathbb{R}^N \rightrightarrows \mathbb{R}^N$  be maximal monotone operators. The mapping  $(\forall \lambda \in [0, 1])$   $h_\lambda : \mathbb{R}^N \rightrightarrows \mathbb{R}^N$  given by

$$\mathbb{R}^N \ni x \mapsto \begin{cases} A_1 x & \text{if } \lambda = 0; \\ A_2 x & \text{if } \lambda = 1; \\ \mathcal{R}_1(A, \lambda)x & \text{if } 0 < \lambda < 1. \end{cases} \quad (19)$$

is a homotopy in the graphical convergence topology. More precisely,  $\lambda \mapsto h_\lambda$  is continuous on  $[0, 1]$  in the graphical convergence topology.

# Example 1

Define  $A_1(x) = x$  and  $A_2(x) = \begin{cases} -1 & \text{if } x < 0 \\ [-1, 1] & \text{if } x = 0. \\ 1 & \text{if } x > 0 \end{cases}$

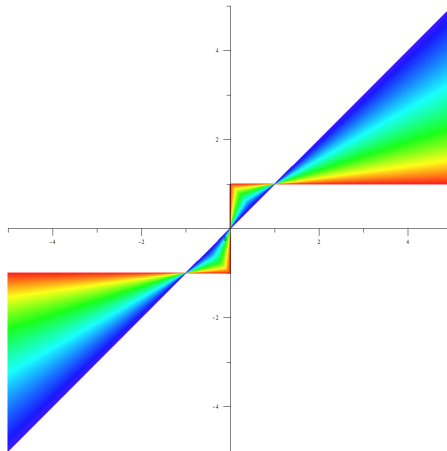
Then,

$$\lambda J_{A_1}(x) = \frac{\lambda}{2}x \text{ and } (1 - \lambda)J_{A_2}(x) = \begin{cases} (1 - \lambda)(x + 1) & \text{if } x < -1 \\ 0 & \text{if } -1 \leq x \leq 1 \\ (1 - \lambda)(x - 1) & \text{if } x > 1 \end{cases}$$

Now to see the graph of  $\mathcal{R}_1(A, \lambda)$ , we use the Minty Parameterization,

$$(J_{\mathcal{R}(A, \lambda)}, x - J_{\mathcal{R}(A, \lambda)})$$

# Example 1 cont.



## Example 2

Define  $A_1(x) = e^x$  and  $A_2(x) = -e^{-x}$ . Solving for  $J_{A_1}, J_{A_2}$ ,

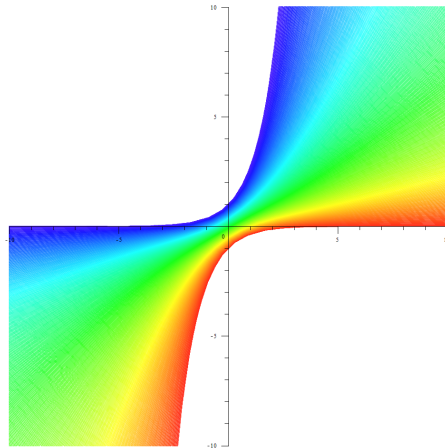
$$J_{A_1}(x) = -W(e^x) + x \text{ and } J_{A_2}(x) = W\left(\frac{1}{e^x}\right) + x$$

Where  $W$  is the Lambert  $W$  function. Now to see the graph of  $\mathcal{R}_1(A, \lambda)$  we use the Minty Parameterizations,

$$(J_{\mathcal{R}(A, \lambda)}, x - J_{\mathcal{R}(A, \lambda)})$$



# Example 2 cont.



# Inheritance of properties

## Definition 12

We say that a property  $(p)$  is

- ① **dominant** if the existence of  $i_0 \in I$  such that  $A_{i_0}$  has property  $(p)$  implies that  $\mathcal{R}_\mu(A, \lambda)$  has property  $(p)$ ;
- ② **recessive** if  $(p)$  is not dominant, and for all  $i \in I$ ,  $A_i$  having property  $(p)$  implies that  $\mathcal{R}_\mu(A, \lambda)$  has property  $(p)$ .
- ③ **indeterminate** if  $(p)$  is neither dominant nor recessive.

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# What have we gained?

## Theorem 13

Let  $A_1, \dots, A_n : \mathbb{R}^N \rightrightarrows \mathbb{R}^N$  be monotone. Then  $\mathcal{R}_\mu(A, \lambda)$  is monotone. Moreover,

$$\text{dom } J_{\mu \mathcal{R}_\mu(A, \lambda)} = \text{dom } J_{\mu A_1} \cap \dots \cap \text{dom } J_{\mu A_n}, \text{ i.e.,} \quad (20)$$

Consequently,  $\mathcal{R}_\mu(A, \lambda)$  is maximal monotone if and only if  $(\forall i) A_i$  is maximal monotone.

## Theorem 14 (domain and range of $\mathcal{R}_\mu$ )

Suppose that for each  $i \in I$ ,  $A_i : \mathcal{H} \rightrightarrows \mathcal{H}$  is maximally monotone. Then

$$\text{ran } \mathcal{R}_\mu(A, \lambda) \simeq \sum_{i \in I} \lambda_i \text{ran } A_i, \quad \text{dom } \mathcal{R}_\mu(A, \lambda) \simeq \sum_{i \in I} \lambda_i \text{dom } A_i. \quad (21)$$

**Theorem 15** (nonempty interior of the domain, fullness of the domain and surjectivity are dominant)

*Suppose that for each  $i \in I$ ,  $A_i : \mathcal{H} \rightrightarrows \mathcal{H}$  is maximally monotone.*

- ① *If there exists  $i_0 \in I$  such that  $\text{int dom } A_{i_0} \neq \emptyset$ , then  $\text{int dom } \mathcal{R}_\mu(A, \lambda) \neq \emptyset$ .*
- ② *If there exists  $i_0 \in I$  such that  $\text{dom } A_{i_0} = \mathcal{H}$ , then  $\text{dom } \mathcal{R}_\mu(A, \lambda) = \mathcal{H}$ .*
- ③ *If there exists  $i_0 \in I$  such that  $A_{i_0}$  is surjective, then  $\mathcal{R}_\mu(A, \lambda)$  is surjective.*

# Convex combinations of NE or FNE

## Lemma 16

Suppose that for each  $i \in I$ ,  $N_i : \mathcal{H} \rightarrow \mathcal{H}$  is nonexpansive,  $T_i : \mathcal{H} \rightarrow \mathcal{H}$  is firmly nonexpansive and set  $N = \sum_{i \in I} \lambda_i N_i$  and  $T = \sum_{i \in I} \lambda_i T_i$ . Let  $x$  and  $y$  be points in  $\mathcal{H}$  such that  $\|Tx - Ty\|^2 = \langle x - y, Tx - Ty \rangle$ . Then  $T_i x - T_i y = Tx - Ty$  for every  $i \in I$ . As a consequence, the following assertions hold:

- ① If there exists  $i_0 \in I$  such that  $T_{i_0}$  is injective, then  $T$  is injective.
- ② If  $x$  and  $y$  are points in  $\mathcal{H}$  such that  $\|Nx - Ny\| = \|x - y\|$ , then  $N_i x - N_i y = Nx - Ny$  for every  $i \in I$ .
- ③ (Reich' 83) If  $\bigcap_{i \in I} \text{Fix } N_i \neq \emptyset$ , then  $\text{Fix } N = \bigcap_{i \in I} \text{Fix } N_i$ .

►  $T_{i_0}$  being injective:  $T_{i_0}(x) = T_{i_0}(y) \Rightarrow x = y$ .

## Lemma 17

Suppose that for each  $i \in I$ ,  $T_i : \mathcal{H} \rightarrow \mathcal{H}$  is firmly nonexpansive and set  $T = \sum_{i \in I} \lambda_i T_i$ . If there exists  $i_0 \in I$  such that

$$T_{i_0}x \neq T_{i_0}y \quad \Rightarrow \quad \|T_{i_0}x - T_{i_0}y\|^2 < \langle x - y, T_{i_0}x - T_{i_0}y \rangle, \quad (22)$$

then  $T$  has property (22) as well.

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Lemma 16 gives:

### Theorem 18

- ① Assume that some  $A_i$  is at most single-valued. Then  $\mathcal{R}_\mu(A, \lambda)$  is also at most single-valued.
- ② Assume that some  $A_{i_0}$  is strictly monotone. Then  $\mathcal{R}_\mu(A, \lambda)$  is also strictly monotone.

► A mapping  $A : \mathcal{H} \rightrightarrows \mathcal{H}$  is at most single-valued if for every  $x \in \mathcal{H}$ ,  $Ax$  is either empty or a singleton.

► A mapping  $A : \mathcal{H} \rightrightarrows \mathcal{H}$  is said to be strictly monotone if whenever  $u \in Ax$  and  $v \in Ay$  are such that  $x \neq y$ , then  $0 < \langle u - v, x - y \rangle$ .

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# Uniform monotonicity and uniform FNE

## Definition 19

A mapping  $A : \mathcal{H} \rightrightarrows \mathcal{H}$  is monotone with modulus  $\phi : [0, \infty[ \rightarrow [0, \infty]$  if for every two points  $(x, u)$  and  $(y, v)$  in  $\text{gra } A$ ,

$$\phi(\|x - y\|) \leq \langle u - v, x - y \rangle.$$

The mapping  $A$  is said to be uniformly monotone with modulus  $\phi$  if  $\phi(t) = 0 \Leftrightarrow t = 0$ .

## Definition 20

A mapping  $T : \mathcal{H} \rightarrow \mathcal{H}$  is firmly nonexpansive with modulus  $\phi : [0, \infty[ \rightarrow [0, \infty]$  if for every pair of points  $x$  and  $y$  in  $\mathcal{H}$ ,

$$\|Tx - Ty\|^2 + \phi(\|Tx - Ty\|) \leq \langle Tx - Ty, x - y \rangle.$$

The mapping  $T : \mathcal{H} \rightarrow \mathcal{H}$  is said to be uniformly firmly nonexpansive with modulus  $\phi$  if  $\phi(t) = 0 \Leftrightarrow t = 0$ .

►  $A$  is  $\varphi$ -monotone  $\Leftrightarrow J_A$  is  $\varphi$ -FNE.

## Proposition 21

*Suppose that for each  $i \in I$ ,  $T_i : \mathcal{H} \rightarrow \mathcal{H}$  is firmly nonexpansive with modulus  $\phi_i$  which is lower semicontinuous and convex and set  $T = \sum_{i \in I} \lambda_i T_i$ . Then  $T$  is firmly nonexpansive with modulus  $\phi = p_{\frac{1}{2}}(\phi, \lambda)$  which is proper, lower semicontinuous and convex. In particular, if there exists  $i_0 \in I$  such that  $T_{i_0}$  is  $\phi_{i_0}$ -uniformly firmly nonexpansive, then  $T$  is  $\phi$ -uniformly firmly nonexpansive.*

This gives

## Theorem 22 (uniform monotonicity is dominant)

*Suppose that for each  $i \in I$ ,  $A_i : \mathcal{H} \rightrightarrows \mathcal{H}$  is maximally monotone with modulus  $\phi_i$  which is lower semicontinuous and convex. Then  $\mathcal{R}_\mu(A, \lambda)$  is monotone with modulus  $\phi = p_{\frac{\mu}{2}}(\phi, \lambda)$  which is lower semicontinuous and convex. In particular, if there exists  $i_0 \in I$  such that  $A_{i_0}$  is  $\phi_{i_0}$ -uniformly monotone, then  $\mathcal{R}_\mu(A, \lambda)$  is  $\phi$ -uniformly monotone.*

## Definition 23

A mapping  $A : \mathcal{H} \rightrightarrows \mathcal{H}$  is  $\epsilon$ -monotone, where  $\epsilon \geq 0$ , if  $A - \epsilon \text{Id}$  is monotone, that is, if for any two points  $(x, u)$  and  $(y, v)$  in  $\text{gra } A$ ,

$$\epsilon \|x - y\|^2 \leq \langle v - u, x - y \rangle.$$

We also say that  $A^{-1}$  is  $\epsilon$ -cocoercive.

Suppose that for each  $i \in I$ ,  $0 \leq \alpha_i \leq \infty$  and set  $\alpha = (\alpha_1 \cdots, \alpha_n)$ . Define

$$r_\mu(\alpha, \lambda) = \left[ \sum_{i \in I} \lambda_i (\alpha_i + \mu^{-1})^{-1} \right]^{-1} - \mu^{-1} \quad \text{and} \quad r(\alpha, \lambda) = r_1(\alpha, \lambda). \quad (23)$$

## Theorem 24 (strong monotonicity is dominant)

*Suppose that for each  $i \in I$ ,  $\epsilon_i \geq 0$ ,  $A_i : \mathcal{H} \rightrightarrows \mathcal{H}$  is maximally monotone and  $\epsilon_i$ -monotone. Then  $\mathcal{R}_\mu(A, \lambda)$  is  $\epsilon$ -monotone where  $\epsilon = r_\mu(\epsilon, \lambda)$ . In particular, if there exists  $i_0 \in I$ , such that  $A_{i_0}$  is  $\epsilon_{i_0}$ -strongly monotone, then  $\mathcal{R}_\mu(A, \lambda)$  is  $\epsilon$ -strongly monotone.*

## Corollary 25 (cocoerciveness is dominant)

*Suppose that for each  $i \in I$ ,  $\epsilon_i \geq 0$ ,  $A_i : \mathcal{H} \rightrightarrows \mathcal{H}$  is maximally monotone and  $A_i^{-1}$  is  $\epsilon_i$ -monotone. Then  $(\mathcal{R}_\mu(A, \lambda))^{-1}$  is  $\epsilon$ -monotone where  $\epsilon = r_{\mu^{-1}}(\epsilon, \lambda)$ . In particular, if there exists  $i_0 \in I$  such that  $A_{i_0}$  is  $\epsilon_{i_0}$ -cocoercive, then  $\mathcal{R}_1(A, \lambda)$  is  $\epsilon$ -cocoercive.*

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## Example 26

Let  $f = \|\cdot\|$ ,  $A_1 = \partial f$  and  $A_2 = \mathbf{0}$ . Then

$$J_{A_1}x = \begin{cases} \left(1 - \frac{1}{\|x\|}\right)x, & \text{if } \|x\| > 1; \\ 0, & \text{if } \|x\| \leq 1 \end{cases}$$

and  $J_{A_2} = \text{Id}$ .  $J_{A_1}$  is not an affine relation and  $J_{A_2}$  is linear. However, for  $0 < \lambda < 1$ ,  $\lambda_1 = \lambda$ ,

$$J_{\mathcal{R}_1(A, \lambda)}x = \lambda J_{A_1}x + (1 - \lambda)J_{A_2}x = \begin{cases} \left(1 - \lambda \frac{1}{\|x\|}\right)x, & \text{if } \|x\| > 1; \\ (1 - \lambda)x, & \text{if } \|x\| \leq 1, \end{cases}$$

is not an affine relation. Thus,  $\mathcal{R}_1(A, \lambda)$  is not an affine relation.

► Example 26 says: linearity and affinity are not dominant properties w.r.t.  $\mathcal{R}_\mu(A, \lambda)$ .

### Theorem 27 (Linearity and affinity are recessive)

*Suppose that for each  $i \in I$ ,  $A_i : \mathcal{H} \rightrightarrows \mathcal{H}$  is a maximally monotone linear (resp. affine) relation. Then  $\mathcal{R}_\mu(A, \lambda)$  is a maximally monotone linear (resp. affine) relation.*

### Definition 28 (rectangular and paramonotone mappings)

① rectangular (also known as  $3^*$  monotone) if for every  $x \in \text{dom } A$  and every  $v \in \text{ran } A$  we have

$$\inf_{(z,w) \in \text{gra } A} \langle v - w, x - z \rangle > -\infty, \quad (24)$$

equivalently, if

$$\text{dom } A \times \text{ran } A \subseteq \text{dom } F_A. \quad (25)$$

- ② paramonotone if whenever we have a pair of points  $(x, v)$  and  $(y, u)$  in  $\text{gra } A$  such that  $\langle x - y, v - u \rangle = 0$ , then  $(x, u)$  and  $(y, v)$  are also in  $\text{gra } A$ .



## Fact 30

Let  $A \in \mathbb{R}^{N \times N}$  be monotone and set  $A_+ = \frac{1}{2}A + \frac{1}{2}A^\top$ . Then the following assertions are equivalent:

- ①  $A$  is paramonotone;
- ②  $A$  is rectangular;
- ③  $\text{rank } A = \text{rank } A_+$ ;
- ④  $\text{ran } A = \text{ran } A_+$ .

## Example 31

In  $\mathbb{R}^2$ , let  $A_1 = N_{\mathbb{R} \times \{0\}}$ . Then  $J_{A_1}$  is the projection on  $\mathbb{R} \times \{0\}$ . Since  $A_1$  is a subdifferential, it is rectangular and paramonotone. Let  $A_2 : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be the counterclockwise rotation by  $\pi/2$ . Then

$$J_{A_1} = P_{\mathbb{R} \times \{0\}} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \quad \text{and} \quad J_{A_2} = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{1}{2} & \frac{1}{2} \end{pmatrix}.$$

Since

$$A_2 = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}, \quad A_{2+} = \frac{1}{2}(A_2 + A_2^T) = 0,$$

$A_2$  is neither rectangular nor paramonotone.

Letting  $\lambda_1 = \lambda_2 = \frac{1}{2}$ , we obtain

$$\mathcal{R}(\mathbf{A}) = \left(\frac{1}{2}J_{A_1} + \frac{1}{2}J_{A_2}\right)^{-1} - \text{Id} = \begin{pmatrix} 0 & -1 \\ 1 & 2 \end{pmatrix} \quad \text{and}$$

$$\mathcal{R}(\mathbf{A})_+ = \frac{1}{2}(\mathcal{R}(\mathbf{A}) + \mathcal{R}(\mathbf{A})^\top) = \begin{pmatrix} 0 & 0 \\ 0 & 2 \end{pmatrix}.$$

By employing Fact 30 we see that  $\mathcal{R}_1(A, \lambda)$  is neither rectangular nor paramonotone.

# Outline

- 1 Resolvent averages
- 2 Dominant properties of  $\mathcal{R}_\mu(A, \lambda)$ 
  - At most single-valued or strict monotonicity
  - Uniform monotonicity
- 3 Recessive properties of  $\mathcal{R}_\mu(A, \lambda)$ 
  - Paramonotonicity and rectangularity
  - Nonexpansive monotonicity and displacement mapping
- 4 Graphical limits of resolvent averages
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### Lemma 32 (Fitzpatrick function of $\mathcal{R}_\mu$ )

Suppose that for each  $i \in I$ ,  $A_i : \mathcal{H} \rightrightarrows \mathcal{H}$  is maximally monotone. Then

$$F_{\mu\mathcal{R}_\mu}(A, \lambda) \leq \sum_{i \in I} \lambda_i F_{\mu A_i} \quad \text{in particular,} \quad F_{\mathcal{R}_1}(A, \lambda) \leq \sum_{i \in I} \lambda_i F_{A_i} \quad (27)$$

and

$$\sum_{i \in I} \lambda_i \operatorname{dom} F_{\mu A_i} \subseteq \operatorname{dom} F_{\mu\mathcal{R}_\mu}(A, \lambda) \quad \text{in particular,} \quad \sum_{i \in I} \lambda_i \operatorname{dom} F_{A_i} \subseteq \operatorname{dom} F_{\mathcal{R}_1}(A, \lambda). \quad (28)$$

### Theorem 33 (rectangularity is recessive)

Suppose that for each  $i \in I$ ,  $A_i : \mathcal{H} \rightrightarrows \mathcal{H}$  is rectangular and maximally monotone. Then  $\mathcal{R}_\mu(A, \lambda)$  is rectangular.

To study paramonotonicity, we need:

### Proposition 34

Suppose that for each  $i \in I$ ,  $T_i : \mathcal{H} \rightarrow \mathcal{H}$  is firmly nonexpansive and set  $T = \sum_{i \in I} \lambda_i T_i$ . Then:

- ① If for each  $i \in I$ , given points  $x$  and  $y$  in  $\mathcal{H}$ ,

$$\begin{aligned} \|T_i x - T_i y\|^2 = \langle x - y, T_i x - T_i y \rangle &\Rightarrow \\ \begin{cases} T_i x = T_i(T_i x + y - T_i y) \\ T_i y = T_i(T_i y + x - T_i x), \end{cases} & \quad (29) \end{aligned}$$

then  $T$  also has property (29).

- ② If there exists  $i_0 \in I$  such that  $T_{i_0}$  has property (29) and is injective, then  $T$  has property (29) and is injective.

### Theorem 35 (paramonotonicity is recessive)

*Suppose that for each  $i \in I$ ,  $A_i : \mathcal{H} \rightrightarrows \mathcal{H}$  is maximally monotone and paramonotone. Then  $\mathcal{R}_\mu(A, \lambda)$  is paramonotone.*

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## Lemma 39

*The maximally monotone mapping  $N : \mathcal{H} \rightarrow \mathcal{H}$  is nonexpansive if and only if  $N = 2J_B - \text{Id}$  for a maximally monotone and nonexpansive mapping  $B$ .*

## Theorem 40 (nonexpansiveness is recessive)

*Suppose that for each  $i \in I$ ,  $A_i : \mathcal{H} \rightarrow \mathcal{H}$  is a nonexpansive and monotone mapping. Then  $\mathcal{R}_1(A, \lambda)$  is nonexpansive. Furthermore, for each  $i \in I$ ,  $A_i = 2J_{B_i} - \text{Id}$  where  $B_i$  is maximally monotone, nonexpansive and  $\mathcal{R}_1(A, \lambda) = 2J_B - \text{Id}$  where  $B$  is the maximally monotone and nonexpansive mapping given by  $B = \sum_{i \in I} \lambda_i B_i$ .*

Theorem 41 (within the class of nonexpansive mappings, being a Banach contraction is dominant)

*Suppose that for each  $i \in I$ ,  $A_i : \mathcal{H} \rightarrow \mathcal{H}$  is nonexpansive and monotone. If there exists  $i_0 \in I$  such that  $A_{i_0}$  is a Banach contraction, then  $\mathcal{R}_1(A, \lambda)$  is a Banach contraction.*





## Indeterminate properties

- 1 Being a projection is indeterminate;
- 2 Being a normal cone operator is indeterminate.

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The graphical limits by Attouch, Rockafellar and Wets are effective for analyzing the convergence of sequences of resolvent averages.

### Definition 43

For a sequence of mappings  $S_k : \mathbb{R}^N \rightrightarrows \mathbb{R}^N$ , we say that  $S_k$  converges graphically to  $S$ , denoted by  $S_k \xrightarrow{g} S$ , if

$$\limsup_k (\text{gra } S_k) = \liminf_k (\text{gra } S_k) = \text{gra } S.$$

Equivalently,  $\forall x \in \mathbb{R}^N$  one has

$$\bigcup_{x_k \rightarrow x} \limsup_k S_k(x_k) \subset S(x) \subset \bigcup_{x_k \rightarrow x} \liminf_k S(x_k).$$



## Theorem 45

Let  $(A_{i,k})_{k \in \mathbb{N}}$  be sequences of maximal monotone mappings and let  $(\lambda_{i,k})_{k \in \mathbb{N}}$  and  $(\mu_k)_{k \in \mathbb{N}}$  be sequences in  $(0, +\infty)$  such that that  $(\forall i) A_{i,k} \xrightarrow{g} A_i$ ,  $\lambda_{i,k} \rightarrow \lambda_i > 0$  and  $\mu_k \rightarrow \mu > 0$ . Then

$$\mathcal{R}_{\mu_k}((A_{1,k}, \dots, A_{n,k}), (\lambda_{1,k}, \dots, \lambda_{n,k})) \xrightarrow{g} \mathcal{R}_{\mu}(A, \lambda), \text{ as } k \rightarrow \infty.$$

Moreover,  $\mathcal{R}_{\mu}(A, \lambda)$  is maximal monotone.

**Question:** What happens for  $\mu \downarrow 0$  or  $\mu \uparrow \infty$ ?

## Lemma 46

Let  $(A_n)_{n \in \mathbb{N}}, A$  be linear operators from  $\mathbb{R}^N$  to  $\mathbb{R}^N$ . If  $A_n \xrightarrow{g} A$ , then there exists  $M > 0$  such that

$$\|A_n\| < M \quad \forall n \in \mathbb{N}, \text{ and } \|A\| < M. \quad (30)$$

Consequently, for linear operators  $(A_n)_{n \in \mathbb{N}}$  and  $A$  on  $\mathbb{R}^N$ , the followings are equivalent:

- (i) graphical convergence:  $A_n \xrightarrow{g} A$ ;
- (ii) point-wise convergence:  $A_n \xrightarrow{p} A$ ;
- (iii) norm convergence:  $A_n \xrightarrow{n} A$ .



# When $A_i = \partial f_i$ :

## Theorem 49

Let  $f_i : X \rightarrow ]-\infty, +\infty]$ ,  $i = 1, \dots, n$  be proper lower semi-continuous convex functions. We have

(i) When  $\mu \downarrow 0$ ,

$$\mathcal{R}_\mu(\partial f, \lambda) \xrightarrow{\mathcal{G}} \partial(\lambda_1 f_1 + \dots + \lambda_n f_n).$$

If, in addition,  $\bigcap_{i=1}^n \text{ri}(\text{dom } f_i) \neq \emptyset$ , then

$$\mathcal{R}_\mu(\partial f, \lambda) \xrightarrow{\mathcal{G}} \lambda_1 \partial f_1 + \dots + \lambda_n \partial f_n \quad \text{when } \mu \downarrow 0;$$



(ii) When  $\mu \uparrow \infty$ ,

$$\mathcal{R}_\mu(\partial f, \lambda) \xrightarrow{g} \partial(\lambda_1 f_1^* + \cdots + \lambda_n f_n^*)^*.$$

If, in addition,  $\bigcap_{i=1}^n \text{ri}(\text{dom } f_i^*) \neq \emptyset$ , then

$$\mathcal{R}_\mu(\partial f, \lambda) \xrightarrow{g} [\lambda_1(\partial f_1)^{-1} + \cdots + \lambda_n(\partial f_n)^{-1}]^{-1} \quad \text{when } \mu \uparrow \infty.$$

When  $(\forall i) A_i$  are positive definite matrices:

Let  $S_+^N$  (resp.  $S_{++}^N$ ) be the set of positive semidefinite matrices (resp. positive definite matrices). For symmetric matrices  $X, Y$ , if  $X - Y \in S_+^N$  we write  $X \succeq Y$ .

## Theorem 50

Let  $A_1, \dots, A_n \in S_{++}^n$ . We have

$$\textcircled{1} \quad \mathcal{H}(A, \lambda) \preceq \mathcal{R}_\mu(A, \lambda) \preceq \mathcal{A}(A, \lambda); \quad (31)$$

②  $\mathcal{R}_\mu(A, \lambda) \rightarrow \mathcal{A}(A, \lambda)$  when  $\mu \rightarrow 0$ ;

③  $\mathcal{R}_\mu(A, \lambda) \rightarrow \mathcal{H}(A, \lambda)$  when  $\mu \rightarrow \infty$ .

## Corollary 51

Assume that  $(\forall i) A_i \in S_{++}^N$  and  $\sum_{i=1}^n \lambda_i = 1$  with  $\lambda_i > 0$ . Then

$$(\lambda_1 A_1 + \cdots + \lambda_n A_n)^{-1} \preceq \lambda_1 A_1^{-1} + \cdots + \lambda_n A_n^{-1}.$$

Consequently, the matrix function  $X \mapsto X^{-1}$  is matrix convex on  $S_{++}^N$ .

## Corollary 52

For every  $\mu > 0$ , the resolvent average matrix function  $A \mapsto \mathcal{R}_\mu(A, \lambda)$  given by

$$(A_1, \dots, A_n) \mapsto [\lambda_1 (A_1^{-1} + \mu^{-1} \text{Id})^{-1} + \cdots + \lambda_n (A_n^{-1} + \mu^{-1} \text{Id})^{-1}]^{-1} - \mu^{-1} \text{Id}$$

is matrix concave on  $S_{++}^N \times \cdots \times S_{++}^N$ .

For each  $\lambda = (\lambda_1, \dots, \lambda_n)$  with  $\sum_{i=1}^n \lambda_i = 1$  and  $\lambda_i > 0 \forall i$ , the harmonic average matrix function

$$(A_1, \dots, A_n) \mapsto (\lambda_1 A_1^{-1} + \dots + \lambda_n A_n^{-1})^{-1} \text{ is matrix concave}$$

on  $S_{++}^N \times \dots \times S_{++}^N$ . Consequently, the harmonic average function

$$(x_1, \dots, x_n) \mapsto \frac{1}{x_1^{-1} + \dots + x_n^{-1}} \text{ is concave} \quad (32)$$

on  $\mathbb{R}_{++} \times \dots \times \mathbb{R}_{++}$ .

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# How far does this take us?

- Extensions to averaged mappings by Combettes?
- Relationships to geometric averages of matrices, variational sums of monotone operators by Attouch, Baillon & Thera?
- For general monotone operators, under what conditions



$$\mathcal{R}_\mu(A, \lambda) \xrightarrow{\mathcal{G}} \mathcal{H}(A, \lambda) \text{ when } \mu \uparrow \infty,$$

$$\mathcal{R}_\mu(A, \lambda) \xrightarrow{\mathcal{G}} \mathcal{A}(A, \lambda) \text{ when } \mu \downarrow 0$$

?

# Thank You Very Much!

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