

# Strong convergence and fast residual decay for monotone operator flows via Tikhonov regularization

Radu Ioan Boţ

University of Vienna  
Faculty of Mathematics  
Oskar-Morgenstern-Platz 1  
1090 Vienna  
Austria  
[www.mat.univie.ac.at/~rabot](http://www.mat.univie.ac.at/~rabot)

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## Monotone equation

- ▶ Consider the **monotone equation**

$$V(x) = 0,$$

where

- ▶  $V: \mathcal{H} \rightarrow \mathcal{H}$  is a continuous and monotone operator, i. e.

$$\langle z - u, V(z) - V(u) \rangle \geq 0 \quad \forall z, u \in \mathcal{H};$$

- ▶ the solution set  $\text{Zer } V$  is assumed to be nonempty.

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- ▶ Finding a **global minimum** of a convex and differentiable function  $f: \mathcal{H} \rightarrow \mathbb{R}$  reduces to solving the monotone equation

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- ▶ Finding a **global minimum** of a convex and differentiable function  $f: \mathcal{H} \rightarrow \mathbb{R}$  reduces to solving the monotone equation

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- ▶ Finding a **saddle point** of a convex-concave and differentiable function  $\Phi: \mathcal{H} \times \mathcal{G} \rightarrow \mathbb{R}$  reduces to solving the monotone equation

$$\begin{pmatrix} \nabla_x \Phi(x, \lambda) \\ -\nabla_\lambda \Phi(x, \lambda) \end{pmatrix} = 0.$$

## Monotone operator flows

- ▶ For  $t_0 \geq 0$ , consider on  $[t_0, +\infty)$  the dynamical system

$$\begin{cases} \dot{x}(t) + V(x(t)) = 0 \\ x(t_0) = x^0. \end{cases}$$

- ▶ (Baillon-Brézis, 1976): The **ergodic trajectory**  $t \mapsto \frac{1}{t} \int_0^t x(s) ds$  converges weakly to an element in  $\text{Zer } V$  as  $t \rightarrow +\infty$ .

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- ▶ Consider the **counterclockwise  $\pi/2$ -rotation operator**

$$V : \mathbb{R}^2 \rightarrow \mathbb{R}^2, \quad V(x, y) = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix},$$

and, for  $t_0 := 0$ ,  $(x(0), y(0)) = r_0(\cos(\theta_0), \sin(\theta_0))$ . Then  $\mathcal{S} = \{(0, 0)\}$  and  $(x(t), y(t)) = r_0(\cos(\theta_0 - t), \sin(\theta_0 - t))$ , which is bounded, but **does not have a limit** as  $t \rightarrow +\infty$ . However,  $t \mapsto \frac{1}{t} \int_0^t (x(s), y(s)) ds$  **converges to  $(0, 0)$**  as  $t \rightarrow +\infty$ .

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**Gradient Descent Ascent (GDA) method (does not converge in general)**

$$z^{k+1} := z^k - \theta V(z^k) \quad \forall k \geq 0.$$

- ▶ For the **counterclockwise  $\pi/2$ -rotation operator** it holds

$$\begin{pmatrix} x^{k+1} \\ y^{k+1} \end{pmatrix} := \begin{pmatrix} 1 & \theta \\ -\theta & 1 \end{pmatrix} \begin{pmatrix} x^k \\ y^k \end{pmatrix} \quad \forall k \geq 0.$$

Since  $\rho\left(\begin{pmatrix} 1 & \theta \\ -\theta & 1 \end{pmatrix}\right) = \max\{|\lambda_1|, |\lambda_2|\} = \sqrt{1 + \theta^2} > 1$ ,  $(x^k, y^k)_{k \geq 0}$  does not converge to  $(0, 0)$  as  $k \rightarrow +\infty$ .

## Monotone flows governed by cocoercive operators

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## Examples of cocoercive operators

- ▶  $\nabla f: \mathcal{H} \rightarrow \mathcal{H}$ , for  $f: \mathcal{H} \rightarrow \mathbb{R}$  a convex and differentiable function with  $\nabla f$  Lipschitz continuous (**Theorem of Baillon-Haddad**);

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Then it holds (Attouch-RIB-Nguyen, **MOR, 2023**):

- ▶  $x(t)$  converges weakly to a zero of  $V$  as  $t \rightarrow +\infty$ ;
- ▶  $\|V(x(t))\| = o\left(\frac{1}{\sqrt{t}}\right)$  as  $t \rightarrow +\infty$ .

## Monotone flows with corrector term for $V$ monotone and $L$ -Lipschitz continuous

- ▶ (Attouch-Svaiter, 2011): for  $\mu, \gamma > 0$ ,

$$\begin{cases} \dot{x}(t) + \mu \frac{d}{dt} V(x(t)) + \gamma V(x(t)) = 0 & \forall t > 0, \\ x(0) = x^0. \end{cases}$$

- ▶ existence and uniqueness of the solution trajectory, convergence of  $\|V(x(t))\|$  to 0 and weak convergence of  $x(t)$  to a zero of  $V$  as  $t \rightarrow +\infty$ ;

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- ▶ (RIB-Schindler, 2024):

$$\begin{cases} d(X(t) + \mu(t)V(X(t))) = -(\gamma(t) - \dot{\mu}(t))V(X(t))dt + \sigma(t, X(t))dW(t) & \forall t > 0, \\ X(0) = X^0, \end{cases}$$

where

- ▶  $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{t \geq 0}, \mathbb{P})$  is a filtered probability space;
- ▶ the diffusion term  $\sigma : \mathbb{R}_+ \times \mathbb{R}^d \rightarrow \mathbb{R}^{d \times m}$  is measurable;
- ▶  $X(\cdot)$  and  $V(X(\cdot))$  are stochastic Itô processes with the same  $m$ -dimensional Brownian motion  $W$ .
- ▶ existence and uniqueness of the trajectory process, **almost sure convergence** of  $\|V(X(t))\|$  to 0 and of  $X(t)$  to a zero of  $V$  as  $t \rightarrow +\infty$ , ergodic convergence rates in **expectation**.

## Numerical algorithms for $V$ monotone and $L$ -Lipschitz continuous

### Extragradient (EG) method (Korpelevich, 1976; Antipin, 1976)

$$(\forall k \geq 0) \quad \begin{cases} y^{k+1} & := x^k - \theta V(x^k) \\ x^{k+1} & := x^k - \theta V(y^{k+1}). \end{cases}$$

- ▶ For  $0 < \theta < \frac{1}{L}$ , convergences weakly to a zero of  $V$ .
- ▶ Last iterate convergence rate (Gorbunov-Loizou-Gidel, 2022)

$$\|V(x^k)\| = O\left(\frac{1}{\sqrt{k}}\right) \text{ and } \text{Gap}(x^k) = \sup_{u \in \mathbb{B}(x^*; \delta(x^0))} \langle x^k - u, V(u) \rangle = O\left(\frac{1}{\sqrt{k}}\right) \text{ as } k \rightarrow +\infty.$$

### Optimistic Gradient Descent Ascent (OGDA) method (Popov, 1980)

$$(\forall k \geq 1) \quad x^{k+1} := x^k - 2\theta V(x^k) + \theta V(x^{k-1})$$

- ▶ For  $0 < \theta < \frac{1}{2L}$ , convergences weakly to a zero of  $V$ .
- ▶ Last iterate convergence rate (Gorbunov-Taylor-Gidel, 2022; Cai-Oikonomou-Zheng, 2022)

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# Numerical algorithms for $V$ monotone and $L$ -Lipschitz continuous

## Extra-Anchored-Gradient and Popov-type methods

$$(\forall k \geq 0) \quad \begin{cases} y^{k+1} & := x^k - \varepsilon_k(x^k - v) - \eta_k V(z^{k+1}) \\ x^{k+1} & := x^k - \varepsilon_k(x^k - v) - \theta_k V(y^{k+1}). \end{cases}$$

- ▶ **EAG-C** (Yoon-Ryu, 2021):  $z^{k+1} = x^k$ ,  $\varepsilon_k = \frac{1}{k+2}$ , and  $\eta_k = \theta_k = \eta \in (0, \frac{1}{8L}]$ ;
- ▶ **EAG-V** (Yoon-Ryu, 2021):  $z^{k+1} = x^k$ ,  $\varepsilon_k = \frac{1}{k+2}$ , and  $\eta_k = \theta_k$  fulfilling

$$\theta_0 \in \left(0, \frac{3}{4L}\right) \text{ and } \theta_{k+1} := \theta_k \left(1 - \frac{1}{(k+1)(k+3)} \frac{\theta_k^2 L^2}{1 - \theta_k^2 L^2}\right) \quad \forall k \geq 0;$$

- ▶ **FEG** (Lee-Kim, 2024):  $z^{k+1} = x^k$ ,  $\varepsilon_k = \frac{1}{k+1}$ ,  $\eta_k = \frac{1}{L} \left(1 - \frac{1}{k+1}\right)$ ,  $\theta_k = \frac{1}{L}$ ;
- ▶ **APV** (Tran-Dinh-Luo, 2024):  $z^{k+1} = y^k$ ,  $\varepsilon_k = \frac{1}{k+2}$ , and  $\eta_k = \theta_k$  fulfilling

$$\theta_0 \in \left(0, \frac{1}{2L\sqrt{3}}\right) \text{ and } \theta_{k+1} := \frac{(1 - (\varepsilon_k)^2 - (2L\theta_k)^2)\varepsilon_{k+1}\theta_k}{(1 - (2L\theta_k)^2)(1 - \varepsilon_k)\varepsilon_k} \quad \forall k \geq 0.$$

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The four methods exhibit a convergence rate of

$$\|V(x^k)\| = O\left(\frac{1}{k}\right) \quad \text{as } k \rightarrow +\infty.$$

## Fast OGDA: continuous time approach

- ▶ For  $t_0 > 0$  we consider on  $[t_0, +\infty)$  the dynamical system

$$\begin{cases} \ddot{x}(t) + \frac{\alpha}{t} \dot{x}(t) + \frac{d}{dt} V(x(t)) + \frac{\alpha}{2t} V(x(t)) = 0 \\ x(t_0) = x^0 \text{ and } \dot{x}(t_0) = \dot{x}^0, \end{cases}$$

for which

- ▶  $\alpha \geq 2$  and  $(x^0, \dot{x}^0) \in \mathcal{H} \times \mathcal{H}$ ;
- ▶ we assume that admits a unique strong global solution  $x : [t_0, +\infty) \rightarrow \mathcal{H}$  with the property that  $t \mapsto V(x(t))$  is absolutely continuous on every compact interval.

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- ▶ For  $x^* \in \text{Zer } V$  and  $1 \leq \lambda \leq \alpha - 1$ , let the **energy function**  $\mathcal{E}_\lambda : [t_0, +\infty) \rightarrow [0, +\infty)$ ,

$$\begin{aligned} \mathcal{E}_\lambda(t) := & \frac{1}{2} \left\| 2\lambda(x(t) - x^*) + t(2\dot{x}(t) + V(x(t))) \right\|^2 + 2\lambda(\alpha - 1 - \lambda) \|x(t) - x^*\|^2 \\ & + 2\lambda t \langle x(t) - x^*, V(x(t)) \rangle + \frac{1}{2} t^2 \|V(x(t))\|^2. \end{aligned}$$

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- ▶  $t \mapsto \mathcal{E}_{\alpha-1}(t)$  is **nonincreasing** on  $[t_0, +\infty)$ , thus

$$0 \leq \|V(x(t))\| \leq \sqrt{2\mathcal{E}_{\alpha-1}(t_0)} \frac{1}{t} \quad \text{and} \quad 0 \leq \langle x(t) - x^*, V(x(t)) \rangle \leq \frac{\mathcal{E}_{\alpha-1}(t_0)}{2(\alpha-1)} \frac{1}{t}.$$

# Convergence of trajectories and faster convergence rates

## Theorem

If  $\alpha > 2$ , and  $x: [t_0, +\infty) \rightarrow \mathcal{H}$  be a solution of the dynamical system, and  $x^* \in \text{Zer } V$ , then it holds

▶  $x(t)$  converges weakly to a zero of  $V$  as  $t \rightarrow +\infty$ ;

▶

$$\|\dot{x}(t)\| = o\left(\frac{1}{t}\right) \quad \text{as } t \rightarrow +\infty,$$

$$\langle x(t) - x^*, V(x(t)) \rangle = o\left(\frac{1}{t}\right) \quad \text{and} \quad \|V(x(t))\| = o\left(\frac{1}{t}\right) \quad \text{as } t \rightarrow +\infty,$$

and

$$0 \leq \text{Gap}(x(t)) = \sup_{u \in \mathbb{B}(x^*; \delta(x^0))} \langle x(t) - u, V(u) \rangle = o\left(\frac{1}{t}\right) \quad \text{as } t \rightarrow +\infty.$$

# Explicit temporal discretization/Explicit Fast OGDA method

## Algorithm

Let  $\alpha > 2$ ,  $x^0, x^1, \bar{x}^0 \in \mathcal{H}$ , and  $0 < \theta < \frac{1}{4L}$ . For every  $k \geq 1$  we set

$$\begin{aligned}\bar{x}^k &:= x^k + \left(1 - \frac{\alpha}{k + \alpha}\right) (x^k - x^{k-1}) - \frac{\alpha\theta}{k + \alpha} V(\bar{x}^{k-1}) \\ x^{k+1} &:= \bar{x}^k - \theta \left(1 + \frac{k}{k + \alpha}\right) (V(\bar{x}^k) - V(\bar{x}^{k-1})).\end{aligned}$$

## Convergence of the iterates and fast convergence rates

### Theorem

Let  $x^* \in \text{Zer } V$  and  $(x^k)_{k \geq 0}$  the sequence generated by the **Explicit Fast OGDA method**. Then it holds:

- ▶ the sequence  $(x^k)_{k \geq 0}$  converges weakly to a zero of  $V$ ;
- ▶

$$\|x^k - x^{k-1}\| = o\left(\frac{1}{k}\right), \quad \langle x^k - x^*, V(x^k) \rangle = o\left(\frac{1}{k}\right),$$
$$\|V(x^k)\| = o\left(\frac{1}{k}\right), \quad \|V(\bar{x}^k)\| = o\left(\frac{1}{k}\right) \text{ as } k \rightarrow +\infty.$$

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RIB, E.R. Csetnek, D.-K. Nguyen, *Fast Optimistic Gradient Descent Ascent (OGDA) method in continuous and discrete time*, Foundations of Computational Mathematics 25(1), 162–222, 2025



RIB, D.-K. Nguyen, *Fast Krasnosel'skiĭ-Mann algorithm with a convergence rate of the fixed point iteration of  $o(1/k)$* , SIAM Journal on Numerical Analysis 61(6), 2813–2843, 2023

## Numerical experiments

- ▶ We consider a minmax problem studied in (Ouyang-Xu, 2021)

$$\min_{u \in \mathbb{R}^n} \max_{v \in \mathbb{R}^n} \mathcal{L}(u, v) := \frac{1}{2} \langle u, Hu \rangle - \langle u, h \rangle - \langle v, Au - b \rangle,$$

where

$$A := \frac{1}{4} \begin{pmatrix} & & & -1 & 1 \\ & & & \ddots & \ddots \\ & & -1 & 1 & \\ -1 & 1 & & & \\ 1 & & & & \end{pmatrix} \in \mathbb{R}^{n \times n},$$

$$H := 2A^T A, \quad b := \frac{1}{4} \begin{pmatrix} 1 \\ 1 \\ \vdots \\ 1 \\ 1 \end{pmatrix} \in \mathbb{R}^n \quad \text{and} \quad h := \frac{1}{4} \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix} \in \mathbb{R}^n.$$

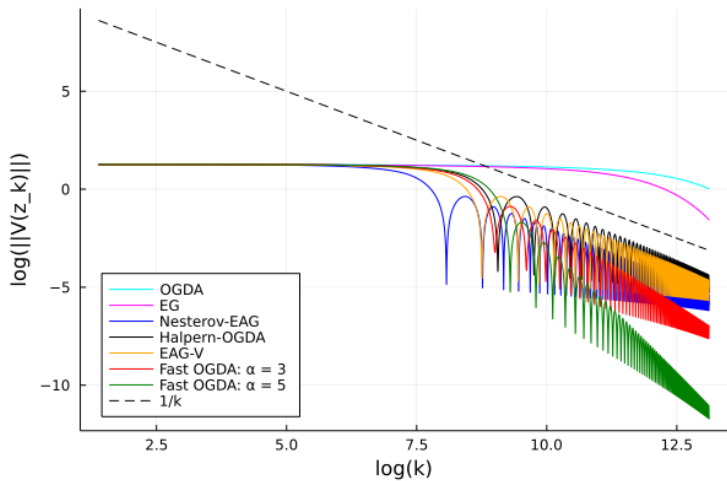
- ▶ Since  $\|A\| \leq \frac{1}{2}$ , one can take for the monotone mapping

$$V(u, v) = \begin{pmatrix} \nabla_u \mathcal{L}(u, v) \\ -\nabla_v \mathcal{L}(u, v) \end{pmatrix}$$

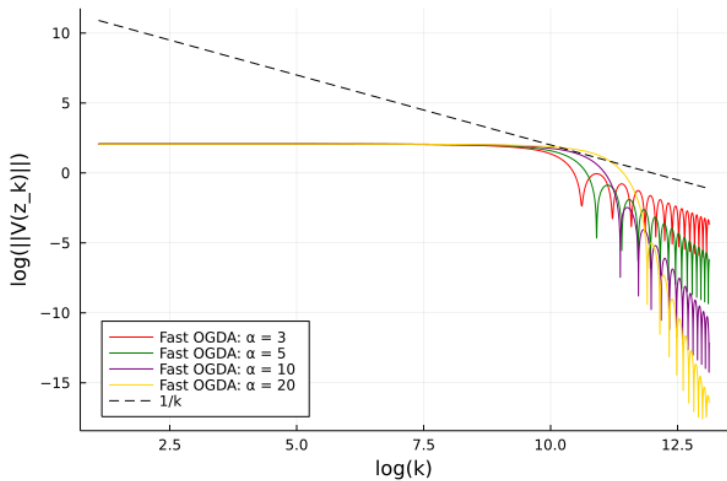
as Lipschitz constant  $L = 1$ .

- ▶ The following numerical algorithms are used in the numerical experiments:
  1. Optimistic Gradient Descent Ascent (OGDA) method (Popov, 1980) with  $s := \frac{0.48}{L}$ ;
  2. Extragradient (EG) method (Korpelevich, 1976; Antipin, 1976)) with  $s := \frac{0.96}{L}$ ;
  3. Extra Anchored Gradient (EAG-V) method (Yoon-Ryu, 2021) with variable step sizes  $(s_k)_{k \geq 0}$ ;
  4. Nesterov's accelerated variant (Nesterov-EAG) by (Tran-Dinh, 2022) of the Extra Anchored Gradient method;
  5. OGDA combined with Halpern anchoring scheme (Halpern-OGDA) by (Tran-Dinh-Luo, 2021);
  6. our explicit algorithm (Fast OGDA) with  $s := \frac{0.24}{L}$  and various choices for  $\alpha$ .

$n = 200$



$n = 1000$ : the influence of the parameter  $\alpha$



## First order dynamical system with Tikhonov regularization

- ▶ Consider again the **monotone equation**

$$V(x) = 0,$$

where

- ▶  $V: \mathcal{H} \rightarrow \mathcal{H}$  is a **continuous and monotone operator** with  $\text{Zer } V$  nonempty.

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- ▶ For  $t_0 \geq 0$  and  $\varepsilon: [t_0, +\infty) \rightarrow (0, +\infty)$ , consider on  $[t_0, +\infty)$  the **Tikhonov regularized dynamical system**

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- ▶ The **regularization function**  $\varepsilon: [t_0, +\infty) \rightarrow (0, +\infty)$  is required to be continuously differentiable and to satisfy

$$\lim_{t \rightarrow +\infty} \varepsilon(t) = 0 \quad \text{and} \quad \int_{t_0}^{+\infty} \varepsilon(t) dt = +\infty.$$

- ▶ The dynamical system is assumed to admit a **unique strong global solution**  $x: [t_0, +\infty) \rightarrow \mathcal{H}$  with the property that  $t \mapsto V(x(t))$  is absolutely continuous on every compact interval.

## A discussion of the underlying assumptions

- ▶ If  $\varepsilon(t) \equiv \varepsilon_0 > 0$ , then the trajectory  $x(t)$  exhibits **strong convergence** towards the unique zero of the strongly monotone operator  $V + \varepsilon_0 \text{Id}$ .
- ▶ If  $V \equiv 0$ , then the unique solution

$$x(t) = x^0 \exp\left(-\int_{t_0}^t \varepsilon(u) du\right)$$

of the system **converges to 0** as  $t \rightarrow +\infty$  if and only if  $\int_{t_0}^{+\infty} \varepsilon(t) dt = +\infty$ .

- ▶ If  $V : \mathcal{H} \rightarrow \mathcal{H}$  is **Lipschitz continuous**, then the system has for every  $x_0 \in \mathcal{H}$  a unique strong global solution  $x : [t_0, +\infty) \rightarrow \mathcal{H}$ . In this case,  $t \mapsto V(x(t))$  is absolutely continuous on every compact interval and therefore **almost everywhere differentiable** on  $[t_0, +\infty)$ .
- ▶ We define
$$\gamma_\varepsilon : [t_0, +\infty) \rightarrow \mathbb{R}_{++}, \quad \gamma_\varepsilon(t) := \exp\left(\int_{t_0}^t \varepsilon(u) du\right).$$
- ▶ For every  $t \geq t_0$  it holds  $\dot{\gamma}_\varepsilon(t) = \varepsilon(t) \gamma_\varepsilon(t)$ .
- ▶ In addition,  $\gamma_\varepsilon$  is strictly increasing, thus  $\gamma_\varepsilon(t) \geq \gamma_\varepsilon(t_0) = 1$  for every  $t \geq t_0$ , and  $\lim_{t \rightarrow +\infty} \gamma_\varepsilon(t) = +\infty$ .
- ▶ For every  $t \geq t_0$  it holds  $\ddot{\gamma}_\varepsilon(t) = [\dot{\varepsilon}(t) + \varepsilon^2(t)] \gamma_\varepsilon(t)$ .

► Given a trajectory solution  $x: [t_0, +\infty) \rightarrow \mathcal{H}$  and  $x^* \in \text{Zer } V$ , we consider the energy functions

$$\varphi: [t_0, +\infty) \rightarrow [0, +\infty) \quad \varphi(t) := \frac{1}{2} \|x(t) - x^*\|^2,$$

and

$$\psi: [t_0, +\infty) \rightarrow [0, +\infty), \quad \psi(t) := \frac{1}{2} \|V(x(t)) + \varepsilon(t)x(t)\|^2.$$

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- For a triple  $(x^0, v, x^*) \in \mathcal{H} \times \mathcal{H} \times \mathcal{H}$ , we denote

$$D_0(x^0, v, x^*) := \max \{ \|x^0 - x^*\|, \|v - x^*\| \}.$$

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- ▶ The **trajectory is bounded**; for every  $t \geq t_0$  it holds

$$\|x(t)\| \leq 2D_0(x^0, 0, x^*).$$

## Proposition

If  $\|V(x(t))\| \rightarrow 0$  as  $t \rightarrow +\infty$ , then  $x(t)$  converges strongly to the minimum norm solution  $x^* := \text{proj}_{\text{Zer } V}(0)$  as  $t \rightarrow +\infty$ .

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Sketch of a proof.

► Choosing  $x^* := \text{proj}_{\text{Zer } V}(0)$ , it holds for almost every  $t \geq t_0$

$$\dot{\varphi}(t) + \varepsilon(t) \varphi(t) \leq \varepsilon(t) \langle x(t) - x^*, 0 - x^* \rangle.$$

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## Lemma (Gronwall's inequality in continuous time)

Let  $h: [t_0, +\infty) \rightarrow \mathbb{R}$  be a locally absolutely continuous function,  $g: [t_0, +\infty) \rightarrow \mathbb{R}$  a bounded function and  $\varepsilon: [t_0, +\infty) \rightarrow [0, +\infty)$  a locally integrable function such that

$$\dot{h}(t) + \varepsilon(t) h(t) \leq \varepsilon(t) g(t) \quad \text{for almost every } t \geq t_0.$$

Then  $h$  is also bounded and, if  $\int_{t_0}^{+\infty} \varepsilon(t) dt = +\infty$ , then

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► It holds

$$\limsup_{t \rightarrow +\infty} \varphi(t) \leq \limsup_{t \rightarrow +\infty} \langle x(t) - x_*, 0 - x_* \rangle.$$

The assumption  $\|V(x(t))\| \rightarrow 0$  as  $t \rightarrow +\infty$  yields that **every weak cluster point of  $x(\cdot)$  belongs to  $\text{Zer } V$** . Thus,  $\limsup_{t \rightarrow +\infty} \varphi(t) \leq 0$ , which is nothing else than  $x(t) \rightarrow x_*$  as  $t \rightarrow +\infty$ . □

## Theorem (strong convergence of the trajectory)

If

$$\lim_{t \rightarrow +\infty} \frac{|\dot{\varepsilon}(t)|}{\varepsilon(t)} = 0 \quad \text{or} \quad \int_{t_0}^{+\infty} |\dot{\varepsilon}(t)| dt < +\infty,$$

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$$\dot{\psi}(t) + \varepsilon(t)\psi(t) \leq \varepsilon(t) \left( 2D_0^2(x^0, 0, x^*) \frac{|\dot{\varepsilon}(t)|^2}{\varepsilon(t)^2} \right).$$

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- ▶ In the second scenario, for  $s \geq t_0$  and every  $t \geq s$ , it holds

$$\gamma_\varepsilon^2(t)\psi(t) \leq \gamma_\varepsilon^2(s)\psi(s) + 2D_0^2(x^0, 0, x^*) \int_s^t |\dot{\varepsilon}(u)| \gamma_\varepsilon^2(u) du + \int_s^t |\dot{\varepsilon}(u)| \gamma_\varepsilon^2(u) \psi(u) du.$$

According to **Gronwall's Lemma**, we obtain for every  $t \geq s$

$$\gamma_\varepsilon^2(t)\psi(t) \leq \left( \gamma_\varepsilon^2(s)\psi(s) + 2D_0^2(x^0, 0, x^*) \int_s^t |\dot{\varepsilon}(u)| \gamma_\varepsilon^2(u) du \right) \exp\left( \int_s^t |\dot{\varepsilon}(u)| du \right).$$

From here it yields for every  $s \geq t_0$

$$\limsup_{t \rightarrow +\infty} \psi(t) \leq 2D_0^2(x^0, 0, x^*) \left( \int_s^{+\infty} |\dot{\varepsilon}(u)| du \right) \exp\left( \int_s^{+\infty} |\dot{\varepsilon}(u)| du \right).$$

Finally, letting  $s \rightarrow +\infty$ , we obtain  $\lim_{t \rightarrow +\infty} \psi(t) = 0$ . □

## Remarks

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► The condition  $\lim_{t \rightarrow +\infty} \frac{|\dot{\varepsilon}(t)|}{\varepsilon(t)} = 0$  represents an enhancement of the condition (Israel, Reich, 1981; Cominetti, Peypouquet, Sorin, 2008)

$$t \mapsto \varepsilon(t) \text{ is decreasing} \quad \text{and} \quad \lim_{t \rightarrow +\infty} \frac{|\dot{\varepsilon}(t)|}{\varepsilon^2(t)} = 0.$$

For  $\varepsilon(t) = \frac{\alpha}{t^q}$ , where  $0 < q \leq 1$  and  $\alpha > 0$ , it allows for the critical choice  $q := 1$ .

► We denote

$$\varepsilon_i := \inf_{t \geq t_0} \frac{d}{dt} \left( \frac{1}{\varepsilon(t)} \right) = \inf_{t \geq t_0} \frac{-\dot{\varepsilon}(t)}{\varepsilon^2(t)} \quad \text{and} \quad \varepsilon_s := \sup_{t \geq t_0} \frac{d}{dt} \left( \frac{1}{\varepsilon(t)} \right) = \sup_{t \geq t_0} \frac{-\dot{\varepsilon}(t)}{\varepsilon^2(t)}.$$

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$$\gamma_\varepsilon(t) := \exp \left( \int_{t_0}^t \varepsilon(u) du \right).$$

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► Recall that

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## Theorem (convergence rates)

Let  $x: [t_0, +\infty) \rightarrow \mathcal{H}$  be the trajectory solution of the first-order dynamical system with Tikhonov regularization. Suppose that the regularization function  $\varepsilon(\cdot)$  also satisfies

$$1 < \varepsilon_i \quad \text{or} \quad -1 < \varepsilon_i \leq \varepsilon_s < 1 \quad \text{or} \quad \varepsilon_i = \varepsilon_s = 1.$$

Then, the following statements are true:

1. as  $t \rightarrow +\infty$  it holds

$$\|\dot{x}(t)\| = \begin{cases} \mathcal{O} \left( \frac{1}{\gamma_\varepsilon(t)} \right) & \text{if } 1 < \varepsilon_i \text{ or } \varepsilon_i = \varepsilon_s = 1, \\ \mathcal{O}(\varepsilon(t)) & \text{if } -1 < \varepsilon_i \leq \varepsilon_s < 1, \end{cases}$$
$$\|V(x(t))\| = \begin{cases} \mathcal{O} \left( \frac{1}{\gamma_\varepsilon(t)} \right) & \text{if } 1 < \varepsilon_i \text{ or } \varepsilon_i = \varepsilon_s = 1, \\ \mathcal{O}(\varepsilon(t)) & \text{if } -1 < \varepsilon_i \leq \varepsilon_s < 1; \end{cases}$$

2. the trajectory  $x(t)$  **converges strongly** to the minimum norm solution  $x^* := \text{proj}_{\text{Zer } V}(0)$  as  $t \rightarrow +\infty$ .

## Theorem (convergence rates in the case $\varepsilon(t) := \frac{\alpha}{t}$ )

Let  $t_0 > 0$ ,  $\alpha > 0$ , and  $x: [t_0, +\infty) \rightarrow \mathcal{H}$  be the trajectory solution of the dynamical system

$$\dot{x}(t) + V(x(t)) + \frac{\alpha}{t}x(t) = 0,$$

with initial condition  $x(t_0) := x^0 \in \mathcal{H}$ . Then, the following statements are true:

1. as  $t \rightarrow +\infty$  it holds

$$\|\dot{x}(t)\| = \mathcal{O}\left(\frac{1}{t^{\min\{\alpha, 1\}}}\right) \quad \text{and} \quad \|V(x(t))\| = \mathcal{O}\left(\frac{1}{t^{\min\{\alpha, 1\}}}\right);$$

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## Theorem (the case $\varepsilon(t) := \frac{\alpha}{t^q}$ , $0 < q < 1$ )

Let  $t_0 > 0$ ,  $0 < q < 1$ ,  $\alpha > 0$  such that  $\alpha t_0^{1-q} \geq 2q$ , and  $x: [t_0, +\infty) \rightarrow \mathcal{H}$  be the trajectory solution of the dynamical system

$$\dot{x}(t) + V(x(t)) + \frac{\alpha}{t^q}x(t) = 0,$$

with initial condition  $x(t_0) := x^0 \in \mathcal{H}$ . Then, the following statements are true:

1. as  $t \rightarrow +\infty$  it holds

$$\|\dot{x}(t)\| = \mathcal{O}\left(\frac{1}{t^q}\right) \quad \text{and} \quad \|V(x(t))\| = \mathcal{O}\left(\frac{1}{t^q}\right);$$

2. the trajectory  $x(t)$  **converges strongly** to the minimum norm solution  $x^* := \text{proj}_{\text{ZER } V}(0)$  as  $t \rightarrow +\infty$ .

Theorem (the case  $\varepsilon(t) := \frac{1}{t \log(t)}$ )

Let  $t_0 > 1$  and  $x: [t_0, +\infty) \rightarrow \mathcal{H}$  be the trajectory solution of the dynamical system

$$\dot{x}(t) + V(x(t)) + \frac{1}{t \log(t)} x(t) = 0,$$

with initial condition  $x(t_0) := x^0 \in \mathcal{H}$ . Then, the following statements are true:

1. as  $t \rightarrow +\infty$  it holds

$$\|\dot{x}(t)\| = \mathcal{O}\left(\frac{1}{\log(t)}\right) \quad \text{and} \quad \|V(x(t))\| = \mathcal{O}\left(\frac{1}{\log(t)}\right).$$

2. the trajectory  $x(t)$  **converges strongly** to the minimum norm solution  $x^* := \text{proj}_{\text{Zer } V}(0)$  as  $t \rightarrow +\infty$ .

## First order dynamical system with anchor point

- ▶ For  $t_0 > 0$  and  $v \in \mathcal{H}$  an **anchor point**, consider on  $[t_0, +\infty)$  the dynamical system

$$\begin{cases} \dot{x}(t) + V(x(t)) + \varepsilon(t)(x(t) - v) = 0, \\ x(t_0) = x^0. \end{cases}$$

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- ▶  $x(\cdot)$  is a trajectory solution if and only if

$$z: [t_0, +\infty) \rightarrow \mathcal{H}, \quad z(t) := x(t) - v \quad \forall t \geq t_0,$$

is the trajectory solution of

$$\begin{cases} \dot{z}(t) + V^v(z(t)) + \varepsilon(t)z(t) = 0, \\ z(t_0) = x^0 - v, \end{cases}$$

where

$$V^v: \mathcal{H} \rightarrow \mathcal{H}, \quad V^v(x) = V(x + v) \quad \forall x \in \mathcal{H}.$$

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- ▶ The rates of convergence for the velocity and the operator norm follow from the results provided for the **first order system with Tikhonov regularization**, since  $\dot{z}(t) = \dot{x}(t)$  and  $V^v(z(t)) = V(x(t))$  for every  $s \geq s_0$ .

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- ▶ The strong convergence of the trajectory  $x(t) = z(t) + v$  to  $\text{proj}_{\text{Zer } V}(v) = \text{proj}_{\text{Zer } V^v}(0) + v$  as  $s \rightarrow +\infty$  follows also from the results provided for the **first order system with Tikhonov regularization and time rescaling**.

## From first order dynamics with Tikhonov regularization to second order dynamics with vanishing damping term

► For  $t_0 > 0$  and  $\alpha > 1$ , consider the second-order dynamical system

$$\begin{cases} \ddot{x}(t) + \frac{\alpha}{t}\dot{x}(t) + \frac{d}{dt}V(x(t)) + \frac{1}{t}V(x(t)) = 0, \\ x(t_0) = x^0 \text{ and } \dot{x}(t_0) = \dot{x}^0. \end{cases}$$

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### Proposition

Let  $t_0 > 0$  and  $\alpha > 1$ . Then  $x: [t_0, +\infty) \rightarrow \mathcal{H}$  is the trajectory solution of

$$\ddot{x}(t) + \frac{\alpha}{t} \dot{x}(t) + \frac{d}{dt} V(x(t)) + \frac{1}{t} V(x(t)) = 0$$

with initial conditions  $x(t_0) = x^0$  and  $\dot{x}(t_0) = \dot{x}^0$  if and only if it is the trajectory solution of

$$\dot{x}(t) + V(x(t)) + \frac{\alpha-1}{t} (x(t) - v) = 0,$$

with initial condition  $x(t_0) = x^0$  and anchor point

$$v := x^0 + \frac{t_0}{\alpha-1} (\dot{x}^0 + (t_0) V(x^0)).$$

## Second-order dynamics with vanishing damping term: strong convergence and convergence rates

### Theorem

Let  $t_0 > 0$ ,  $\alpha > 1$ , and  $x: [t_0, +\infty) \rightarrow \mathcal{H}$  the trajectory solution of the second-order dynamical system

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with initial conditions  $x(t_0) = x^0$  and  $\dot{x}(t_0) = \dot{x}^0$ . Then,  $x(t)$  converges strongly to  $\text{proj}_{\text{Zer } V} \left( x^0 + \frac{t_0}{\alpha - 1} (\dot{x}^0 + V(x^0)) \right)$  as  $t \rightarrow +\infty$ ,  $\|\dot{x}(t)\| = O\left(\frac{1}{t^{\min\{\alpha-1, 1\}}}\right)$ , and  $\|V(x(t))\| = O\left(\frac{1}{t^{\min\{\alpha-1, 1\}}}\right)$  as  $t \rightarrow +\infty$ .

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► If  $\alpha \geq 2$ , then  $x(t)$  converges strongly to  $\text{proj}_{\text{Zer } V} \left( x^0 + \frac{t_0}{\alpha - 1} (\dot{x}^0 + V(x^0)) \right)$ ,  $\|\dot{x}(t)\| = O\left(\frac{1}{t}\right)$ , and  $\|V(x(t))\| = O\left(\frac{1}{t}\right)$  as  $t \rightarrow +\infty$ .

## Remark

In case  $\alpha = 2$ , the **Fast OGD** dynamical system

$$\ddot{x}(t) + \frac{\alpha}{t} \dot{x}(t) + \frac{d}{dt} V(x(t)) + \frac{\alpha}{2t} V(x(t)) = 0$$

is nothing else than

$$\ddot{x}(t) + \frac{\alpha}{t} \dot{x}(t) + \frac{d}{dt} V(x(t)) + \frac{1}{t} V(x(t)) = 0,$$

thus it inherits its asymptotic properties.

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RIB, D.-K. Nguyen, *Tikhonov regularization of monotone operator flows not only ensures strong convergence of the trajectories but also speeds up the vanishing of the residuals*, Journal of Complexity 94, Article 102029, 2026

# Generalized Extra-Anchored-Gradient method

## Generalized Extra-Anchored-Gradient method (G-EAG)

Let  $x^0 \in \mathcal{H}$ ,  $v \in \mathcal{H}$ ,  $\theta \in (0, \frac{1}{L})$ , and  $(\varepsilon_k)_{k \geq 0}$  a positive sequence of parameters satisfying

$$\lim_{k \rightarrow +\infty} \varepsilon_k = 0, \quad \sum_{k=0}^{+\infty} \varepsilon_k = +\infty, \quad \text{and} \quad \lim_{k \rightarrow +\infty} \frac{|\varepsilon_{k+1} - \varepsilon_k|}{\varepsilon_k} = 0.$$

Consider the **explicit** iterative scheme

$$(\forall k \geq 0) \quad \begin{cases} y^{k+1} & := x^k - \theta \varepsilon_k (x^k - v) - \theta V(x^k) \\ x^{k+1} & := x^k - \theta \varepsilon_{k+1} (x^{k+1} - v) - \theta V(y^{k+1}). \end{cases}$$

# Generalized Extra-Anchored-Gradient method

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### Remark

Without loss of generality, we can assume  $v = 0$ . The second identity in (G-EAG) can be written as

$$\frac{x^{k+1} - x^k}{\theta} + V(y^{k+1}) + \varepsilon_{k+1} x^{k+1} = 0 \quad \text{for all } k \geq 0, \quad (1)$$

which shows that it can be understood as an explicit discretization of the monotone flow with Tikhonov regularization,  $y^{k+1}$  being an extrapolation point. If  $\varepsilon_k = 0$  for all  $k \geq 0$ , the method reduces to EG.

## Generalized Extra-Anchored-Gradient method: convergence analysis

- ▶ Given  $x^* \in \text{Zer } V$ , we consider the discrete energy functions

$$\varphi^k := \frac{1}{2} \|x^k - x^*\|^2 \quad \text{for all } k \geq 0,$$

and

$$\psi^k := \frac{1}{2} \|V(x^k) + \varepsilon_k x^k\|^2 \quad \text{for all } k \geq 0.$$

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**Theorem (Gronwall's inequality in discrete time; Xu, 2002)**

Let  $(f^k)_k$  and  $(h^k)_k$  be two sequences in  $\mathbb{R}$  and  $(\varepsilon_k)_k$  be a nonnegative sequence satisfying

$$f^{k+1} - f^k + \varepsilon_k f^k \leq \varepsilon_k h^k \quad \text{for all } k \geq 0. \quad (2)$$

Suppose further that  $\varepsilon_k < 1$  for all  $k \geq 0$ . Then it holds

$$f^{k+1} \leq \frac{f^0}{\gamma_k} + \frac{1}{\gamma_k} \sum_{j=0}^k \gamma_j \varepsilon_j h^j \quad \text{for all } k \geq 0,$$

where  $\gamma_k := \prod_{\ell=0}^k (1 - \varepsilon_\ell)^{-1}$ . In particular, if  $\sum_{k=0}^{+\infty} \varepsilon_k = +\infty$ , it holds

$$\limsup_{k \rightarrow +\infty} f^k \leq \limsup_{k \rightarrow +\infty} h^k.$$

## Generalized Extra-Anchored-Gradient method: convergence analysis

- ▶ There exist a vanishing real sequence  $(\eta_k)_k$  and  $k_1 \geq 0$ , such that

$$\varphi^{k+1} - \varphi^k + \frac{\theta\varepsilon_{k+1}}{2}\varphi^k \leq \frac{\theta\varepsilon_{k+1}}{2} \left( \|x^*\|^2 - \frac{1 - \eta_k}{1 + \theta\varepsilon_{k+1}} \|x^k\|^2 \right) \quad \text{for all } k \geq k_1.$$

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- ▶ Assume that  $(x^k)_{k \geq 0}$  is bounded. Then, for any  $\delta \in (0, 1)$ , there exist  $C_\delta \geq 0$  and  $k_0 \geq 0$  such that

$$\psi^{k+1} - \psi^k + 2(1 - \delta)\theta\varepsilon_k\psi^k \leq C_\delta\theta\varepsilon_k \left( \frac{\varepsilon_{k+1} - \varepsilon_k}{\theta\varepsilon_k} \right)^2 \quad \text{for all } k \geq k_0.$$

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### Theorem (strong convergence of the iterates and convergence rates for the residual)

The sequence  $(x^k)_{k \geq 0}$  **converges strongly** to the minimum norm solution  $\text{proj}_{\text{Zer } V}(0)$  as  $k \rightarrow +\infty$ , and the convergence rate of the residual results from the following statement: For any  $\delta \in (0, 1)$ , there exist  $\tilde{C}_\delta \geq 0$  and  $k_0 \geq 0$  such that

$$\psi^{k+1} \leq \frac{\psi^{k_0}}{\gamma_k} + \frac{\tilde{C}_\delta}{\gamma_k} \sum_{j=k_0}^k \gamma_j \varepsilon_j \left( \frac{\varepsilon_{j+1} - \varepsilon_j}{\theta\varepsilon_j} \right)^2 \quad \text{for all } k \geq k_0,$$

where  $(\gamma_k)_{k \geq k_0}$  is the sequence defined by  $\gamma_k := \prod_{l=k_0}^k (1 - 2(1 - \delta)\theta\varepsilon_l)^{-1}$ .

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### Remark

Since  $(x^k)_{k \geq 0}$  converges strongly to the minimum norm solution  $\text{proj}_{Z_{\text{er}} V}(0)$  as  $k \rightarrow +\infty$  and  $V(x^k) \rightarrow 0$  as  $k \rightarrow +\infty$ ,  $(y^{k+1})_{k \geq 0}$  converges also strongly to the same point.

## Generalized Extra-Anchored-Gradient method: particular case

► Let

$$\varepsilon_k := \frac{\alpha}{\theta(k + \beta)} \quad \text{for all } k \geq 0, \text{ and } \alpha, \beta > 0.$$

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### Lemma (Chung, 1954; Polyak, 1987)

Let  $(\psi^k)_{k \geq 1}$  be a non-negative sequence and let  $p, c, C > 0$ . Suppose that

$$\psi^{k+1} \leq \left(1 - \frac{c}{k}\right) \psi^k + \frac{C}{k^{p+1}} \quad \text{for all } k \geq 1.$$

Then the following holds true as  $k \rightarrow +\infty$

$$\psi^k \leq \frac{C}{(c-p)k^p} + o\left(\frac{1}{k^p}\right), \quad \text{if } c > p;$$

$$\psi^k = O\left(\frac{\log(k)}{k^c}\right), \quad \text{if } c = p;$$

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### Theorem (strong convergence of the iterates and convergence rates for the residual)

The sequence  $(x^k)_{k \geq 0}$  **converges strongly** to the minimum norm solution  $\text{proj}_{\text{Zer } V}(0)$  as  $k \rightarrow +\infty$ , the residual showing the following convergence rate:

1. if  $\alpha > 1$ , then  $\|V(x^k)\| = O\left(\frac{1}{k}\right)$  as  $k \rightarrow +\infty$ ;
2. if  $\alpha \leq 1$ , then  $\|V(x^k)\| = O\left(\frac{1}{k^{\alpha(1-\delta)}}\right)$ , for any  $\delta \in (0, 1)$ , as  $k \rightarrow +\infty$ .

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$$\varepsilon_k := \frac{\alpha}{(k + \beta)^\eta} \quad \text{for all } k \geq 0, \text{ and } \alpha, \beta > 0, \eta \in (0, 1).$$

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Then the following holds true as  $k \rightarrow +\infty$

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$$\|V(x^k)\| = O(k^{-\eta}) \quad \text{as } k \rightarrow +\infty.$$



RIB, E. Chenchene, *Extra-Gradient Method with flexible anchoring: strong convergence and fast residual decay*, to appear in SIAM Journal on Optimization

## Numerical experiments

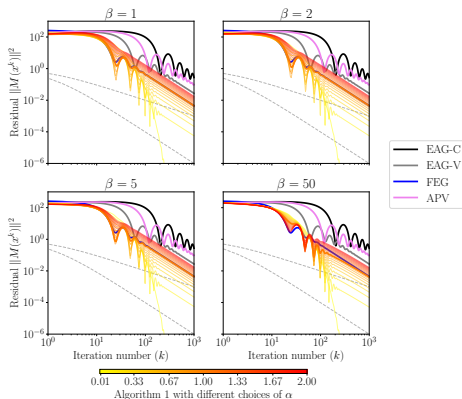
Let  $\mathcal{H} = \mathbb{R}^{2d}$  for  $d = 5$ ,

$$N := \begin{bmatrix} 0 & I \\ -I & 0 \end{bmatrix}, \quad \text{and} \quad x^* \in \mathbb{R}^{2d}, \quad b := Nx^*,$$

where  $x^*$  is randomly sampled. To induce some nonlinearity, consider  $P: H \rightarrow H$  to be projection onto the ball with radius 1 and center  $x^*$ . Then, we define

$$V(x) := Nx - b + P(x) \quad \text{for all } x \in \mathbb{R}^{2d},$$

which is monotone and Lipschitz continuous with  $V(x^*) = 0$ . We choose  $L := \|V\|_2 + 1$ , and  $\varepsilon_k = \frac{\alpha}{\theta(k+\beta)}$ , for  $\alpha$  in  $(10^{-3}, 2]$  and  $\beta = 1, 2, 5, 50$ .

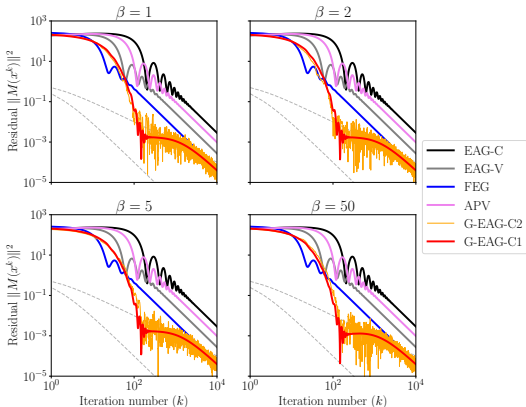


## Numerical experiments: a practical enhancement

For the same problem, we choose

$$\varepsilon_k = \frac{\alpha_k}{\theta(k + \beta)} \quad \text{for all } k \geq 0, \quad (3)$$

where  $\alpha_k$  is set to be either  $\alpha_k := \frac{2}{\pi} \arctan(Mk)$  (falls in the setting of the main convergence theorem), with  $M \in \mathbb{R}_+$  (we set  $M = 10^{-3}$ ), or  $\alpha_k$  to be equal to the same as before but further corrupted with Gaussian noise of variance  $\sigma_k = \frac{1}{k}$  for all  $k \geq 1$ .



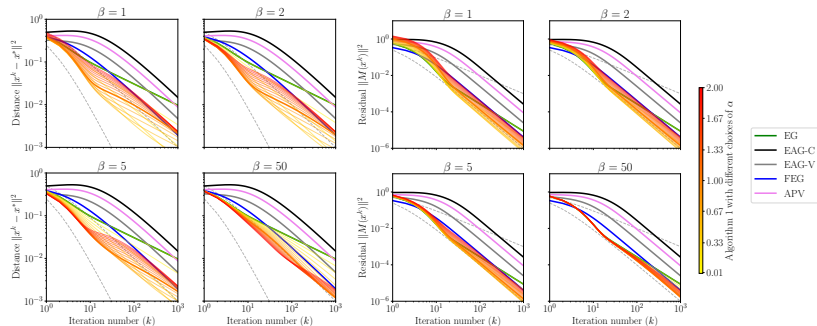
## Numerical experiments: an infinite dimensional example

- On  $\mathcal{H} := \ell^2$ , the space of square summable sequences, we consider the operator

$$V(x) = x - Sx - b, \quad \text{where } S(x_0, x_1, \dots) = (0, x_0, x_1, \dots),$$

for all  $x := (x_0, x_1, \dots) \in \mathcal{H}$ , i.e.  $S$  is the right-shift operator and  $b := (1, -1, 0, \dots) \in \mathcal{H}$ . Since  $S$  is nonexpansive,  $V$  is monotone and 2-Lipschitz continuous. We see that  $x^* := (1, 0, \dots)$  is the unique solution of  $V(x) = 0$ .

- We set a maximum number of iterations  $K := 10^3$  and initialize  $x^0 \in \text{span}\{\delta^0, \delta^1\}$  randomly. In this way,  $x_1, \dots, x_K$  will be always contained in  $\text{span}\{\delta^0, \dots, \delta^{2(K+1)}\} \simeq \mathbb{R}^{2(K+1)}$ .



- G-EAG is particularly competitive also in terms of distance to solution. The example seems to suggest that in this case EG only converges weakly.

Thank you for your attention!

`https://www.mat.univie.ac.at/~rabort/`