### Contraction of Continuous-Time Proximal Gradient Dynamics





Center for Control, Dynamical Systems & Computation University of California at Santa Barbara davydovalexander.github.io

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## Acknowledgments



Veronica Centorrino Scuola Sup Meridionale



Anand Gokhale UCSB



Giovanni Russo Univ Salerno



Francesco Bullo UCSB

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#### Convexity and contractivity

**Kachurovskii's Theorem**: For differentiable  $f : \mathbb{R}^n \to \mathbb{R}$ , equivalent statements:

- f is strongly convex with parameter m
- **2**  $-\nabla f$  is (strongly) infinitesimally contracting with respect to  $\|\cdot\|_2$  with rate m

Also: global minimum of f = globally-exponentially stable equilibrium of  $-\nabla f$ 

For strongly convex f, provides natural way to solve minimization with dynamical system

**1** The minimization problem

$$\min_{x\in\mathbb{R}^n}f(x)$$

Istrongly infinitesimally contracting dynamics

$$\dot{x} = -\nabla f(x)$$

How about more general minimization problems like  $\min_{x \in \mathbb{R}^n} f(x) + g(x)$ ?

R. I. Kachurovskii. Monotone operators and convex functionals. Uspekhi Matematicheskikh Nauk, 15(4):213-215, 1960

### Composite minimization and proximal gradient

For strongly convex + strongly smooth f, convex, closed, proper  $g: \mathbb{R}^n \to \overline{\mathbb{R}}$ ,

$$x^{\star} = \operatorname*{argmin}_{x \in \mathbb{R}^{n}} f(x) + g(x) \qquad \Longleftrightarrow \qquad x^{\star} = \operatorname{prox}_{\gamma g} (x^{\star} - \gamma \nabla f(x))$$
$$\operatorname{prox}_{\gamma g}(z) = \operatorname*{argmin}_{x \in \mathbb{R}^{n}} g(x) + \frac{1}{2\gamma} \|x - z\|_{2}^{2}.$$

Transcription from

The minimization problem

 $\min_{x\in\mathbb{R}^n}f(x)+g(x)$ 

**2** to strongly infinitesimally contracting *proximal gradient* dynamics

 $\dot{x} = -x + \operatorname{prox}_{\gamma g}(x - \gamma \nabla f(x))$ 

# Contractivity of proximal gradient

$$\dot{x} = -x + \operatorname{prox}_{\gamma g}(x - \gamma \nabla f(x))$$

#### Contractivity properties

**Q** For  $\gamma \in (0, 2/\ell)$ , prox. gradient is contracting w.r.t.  $\|\cdot\|_2$  with rate

$$c = 1 - \max\{|1 - \gamma m|, |1 - \gamma \ell|\}$$

Optimal 
$$\gamma^* = \frac{2}{m+\ell}$$
  
2 For  $f(x) = \frac{1}{2}x^{\top}Ax + b^{\top}x$ ,  $A \succ 0$  prox. gradient is contracting w.r.t.  $\|\cdot\|_{(\gamma A - I_n)}$  with rate

for every  $\gamma \in \left] 1/\lambda_{\mathsf{min}}(\mathsf{A}), +\infty \right[$ 

Thank you