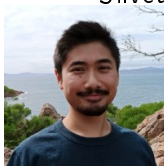


Nonsmooth Implicit Differentiation for Machine Learning

Jérôme Bolte, Tam Le, Edouard Pauwels, and Antonio Silveti-Falls



Neurips Poster (2021)

- 1 Smooth implicit function theorem.
- 2 Nonsmooth implicit function theorem of Clarke.
- 3 Path differentiable nonsmooth implicit function theorem (with calculus).
- 4 Applications
- 5 What can go wrong?

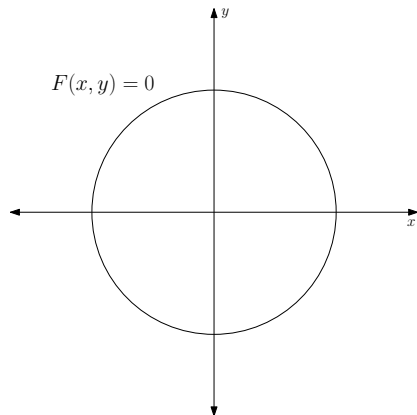
Implicit Functions

Consider the smooth function

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and the equation

$$F(x, y) = 0.$$



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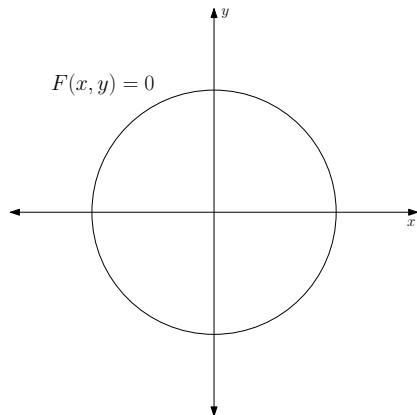
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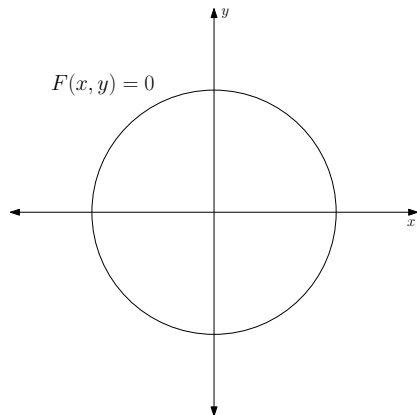
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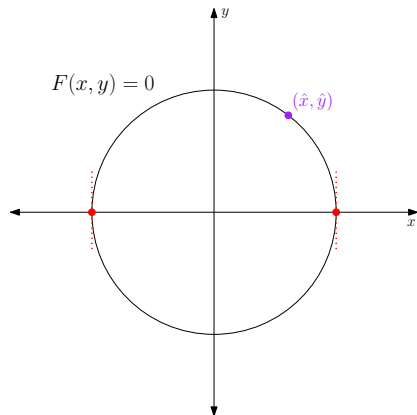
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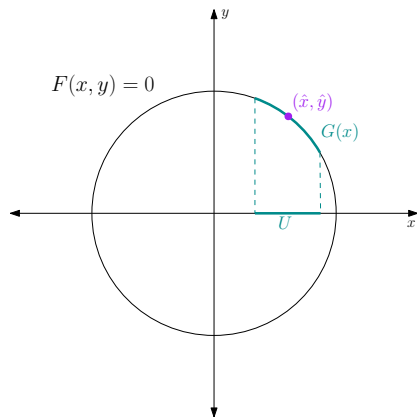
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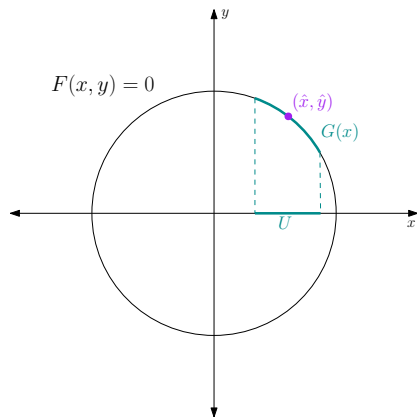
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Definition (Locally Lipschitz-continuous)

A function $F : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is locally Lipschitz-continuous if, $\forall x \in \mathbb{R}^n$, \exists a neighborhood $U \subset \mathbb{R}^n$ of x and $c > 0$ such that, $\forall y, z \in U$,

$$\|F(z) - F(y)\| \leq c \|z - y\| .$$

Locally Lipschitz Functions and the Clarke Subdifferential

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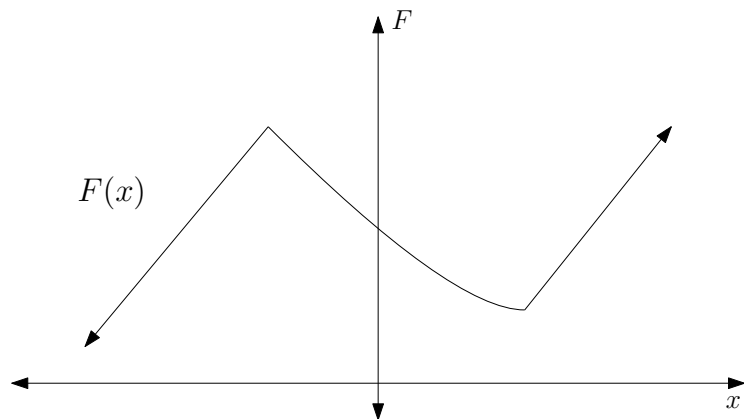
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Definition (Clarke subdifferential (1983))

Given a locally Lipschitz function $F : \mathbb{R}^n \rightarrow \mathbb{R}^m$, the Clarke subdifferential at a point $x \in \mathbb{R}^n$ is

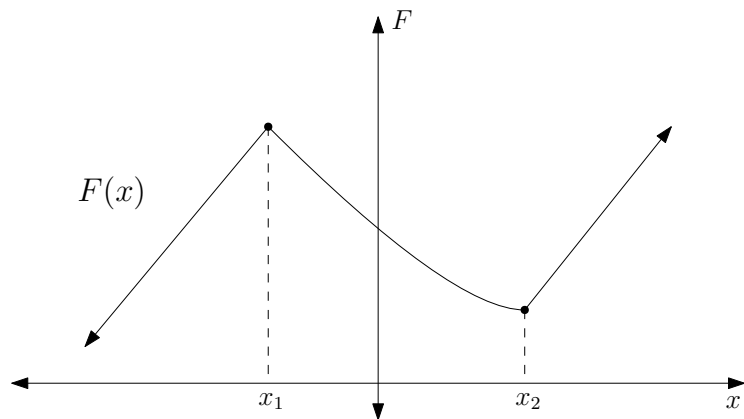
$$\partial^c F(x) = \text{conv} \left(\left\{ \lim_{k \rightarrow \infty} J_F(x_k) : x_k \in \text{diff}_F \text{ and } x_k \rightarrow x \right\} \right).$$

Locally Lipschitz Functions and the Clarke Subdifferential



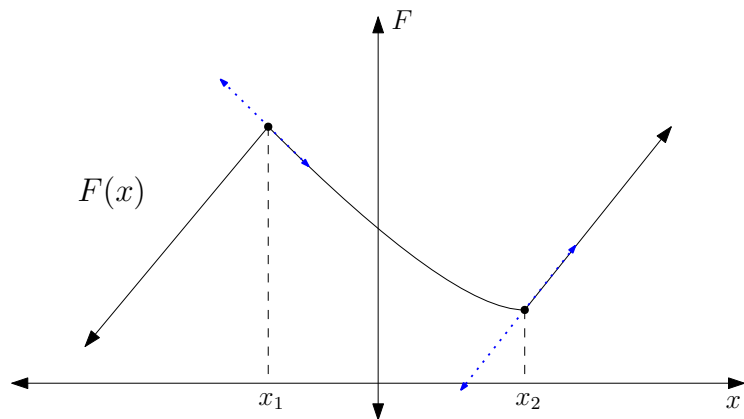
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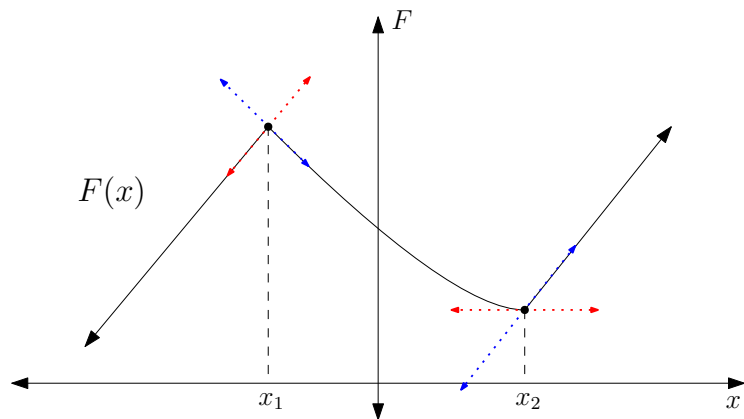
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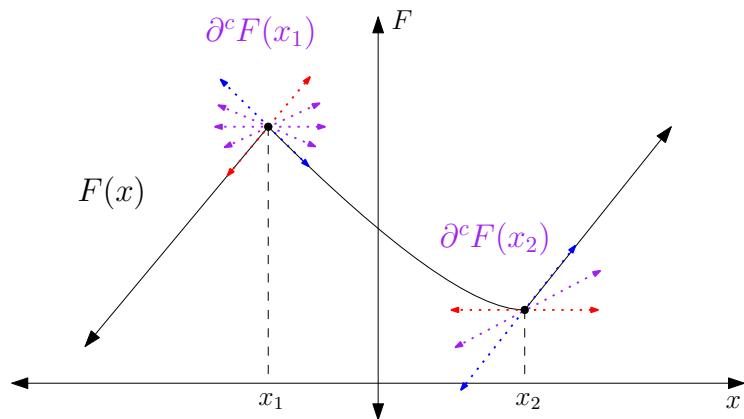
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Nonsmooth Implicit Function Theorem of Clarke

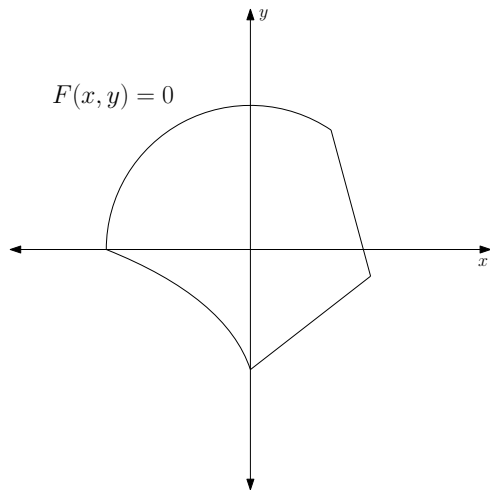
Theorem (Clarke 1983)

Let $F : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^m$ be locally Lipschitz and $(\hat{x}, \hat{y}) \in \mathbb{R}^n \times \mathbb{R}^m$ such that

$$F(\hat{x}, \hat{y}) = 0.$$

If, $\forall [A \ B] \in \partial^c F(\hat{x}, \hat{y})$, B is invertible, then $\exists U \subset \mathbb{R}^n$ a neighborhood of \hat{x} and a locally Lipschitz function $G(x)$ so that

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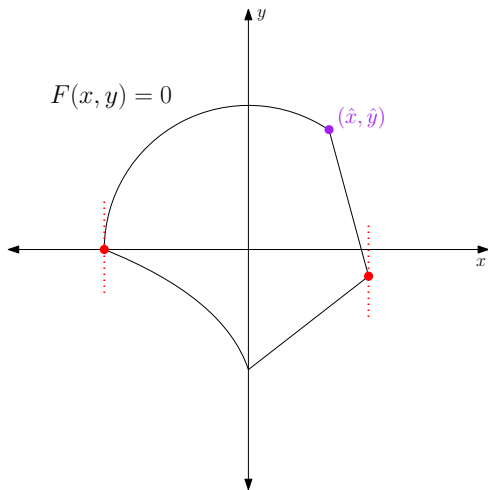
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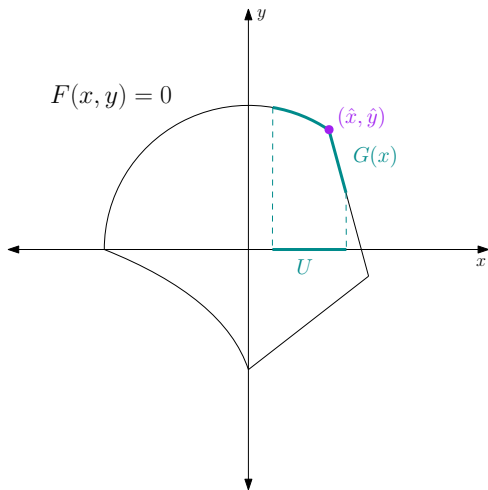
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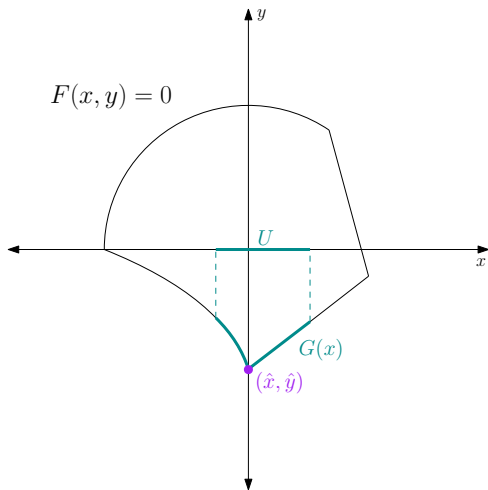
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Recall from smooth IFT: $J_G(x) = -B^{-1}A$ $[A \ B] = J_F(x, G(x))$.

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No - need something beyond ∂^c .

Definition (Conservative field (Bolte-Pauwels 2019))

A set valued mapping $D_F : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$ is a conservative field (or conservative Jacobian) for $F : \mathbb{R}^n \rightarrow \mathbb{R}$ locally Lipschitz if:

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- 3 For any absolutely continuous curve $\gamma : [0, 1] \rightarrow \mathbb{R}^n$,

$$\frac{d}{dt} F(\gamma(t)) = \langle u, \dot{\gamma}(t) \rangle \quad \forall u \in D_F(\gamma(t))$$

for almost all $t \in [0, 1]$.

We call F path differentiable.

Path Differentiable Nonsmooth Implicit Function Theorem

Theorem

Let $F : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^m$ be path differentiable and $(\hat{x}, \hat{y}) \in \mathbb{R}^n \times \mathbb{R}^m$ be such that

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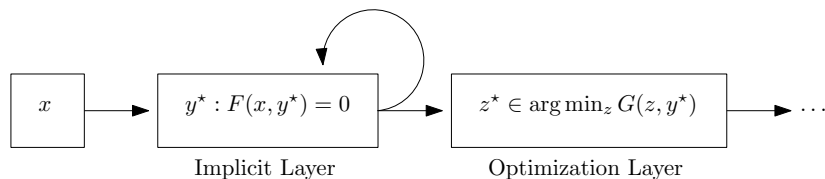
Then $\exists U \subset \mathbb{R}^n$ a neighborhood of \hat{x} and a path differentiable function G such that

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The conservative Jacobian of G is given by

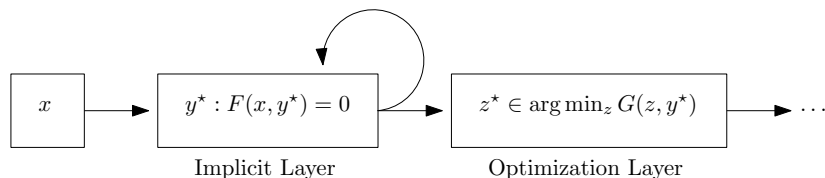
$$D_G(x) = \{-B^{-1}A : [A \ B] \in D_F(x, G(x))\}.$$

Deep Learning with Implicit Layers



- Deep equilibrium networks [Shaojie Bai, J. Zico Kolter, Vladlen Koltun 2019], Monotone deep equilibrium networks [Ezra Winston, J. Zico Kolter 2020].
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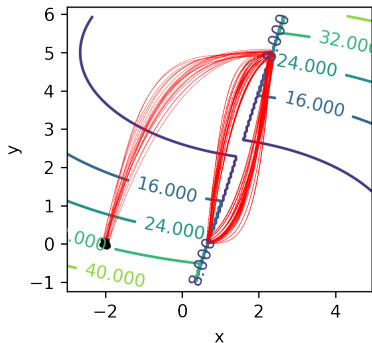
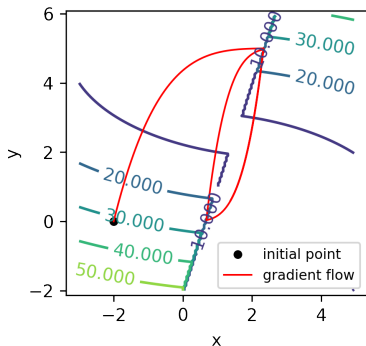
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- Convergence guarantees for training.

Pathological Examples - Cycles

Gradient descent type algorithm (using backprop) applied to:

$$\min_{x,y,s} \ell(x,y,s) \stackrel{\text{def}}{=} (x - s_1)^2 + 4(y - s_2)^2$$

$$\text{s.t. } s \in \arg \max \{(a + b)(-2x + y + 2) : a \in [0, 3], b \in [0, 5]\}.$$



Pathological Examples - Lorenz Attractor

For $u \in \mathbb{R}^3$, define $L(u) \stackrel{\text{def}}{=} (10(y-x), x(28-z)-y, xy - \frac{8}{3}z)$.

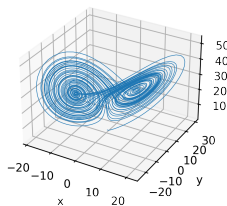
Implicit formulation

$$\begin{aligned} \max_{u \in \mathbb{R}^3} u^T z \quad \text{s.t.} \\ z \in \operatorname{argmin}_{s \in \mathbb{R}^3} \|s - L(u)\|^4 \end{aligned}$$

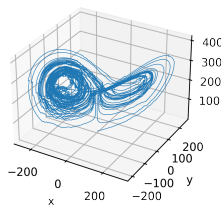
Explicit (vanilla) formulation

$$\max_{u \in \mathbb{R}^3} u^T L(u)$$

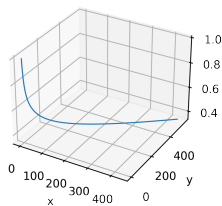
Lorenz attractor



Implicit gradient ascent



Vanilla gradient ascent



Thanks for listening.

Full paper available on arxiv: <https://arxiv.org/abs/2106.04350>

“Nonsmooth Implicit Differentiation for Machine Learning and Optimization” - Jérôme Bolte, Ngoc Tâm Lê, Edouard Pauwels, Antonio Silveti-Falls