

Identifiable Sets in Optimization

Dmitriy Drusvyatskiy (joint work with A. S. Lewis),
School of ORIE, Cornell University

October 15, 2011

Goals

- 1 Intuitive notion of **identifiable sets**.
- 2 Characterizations.
- 3 Existence, calculus.
- 4 Connection to previous work (**partial smoothness**, **prox-regularity**).
- 5 Generic existence (**semi-algebraic setting**).

Preliminaries

For a function $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$, a vector v is a **Frechét subgradient** at \bar{x} , denoted $v \in \hat{\partial}f(x)$, if

$$f(x) \geq f(\bar{x}) + \langle v, x - \bar{x} \rangle + o(|x - \bar{x}|).$$

Preliminaries

For a function $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$, a vector v is a **Frechét subgradient** at \bar{x} , denoted $v \in \hat{\partial}f(x)$, if

$$f(x) \geq f(\bar{x}) + \langle v, x - \bar{x} \rangle + o(|x - \bar{x}|).$$

The **limiting subdifferential** at \bar{x} is

$$\partial f(\bar{x}) = \left\{ \lim_{i \rightarrow \infty} v_i : v_i \in \hat{\partial}f(x_i), x_i \rightarrow \bar{x}, f(x_i) \rightarrow f(\bar{x}) \right\}.$$

Preliminaries

For a function $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$, a vector v is a **Frechét subgradient** at \bar{x} , denoted $v \in \hat{\partial}f(x)$, if

$$f(x) \geq f(\bar{x}) + \langle v, x - \bar{x} \rangle + o(|x - \bar{x}|).$$

The **limiting subdifferential** at \bar{x} is

$$\partial f(\bar{x}) = \left\{ \lim_{i \rightarrow \infty} v_i : v_i \in \hat{\partial}f(x_i), x_i \rightarrow \bar{x}, f(x_i) \rightarrow f(\bar{x}) \right\}.$$

Definition (critical points)

\bar{x} is a **critical point** of f if $0 \in \partial f(\bar{x})$.

Preliminaries

For a function $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$, a vector v is a **Frechét subgradient** at \bar{x} , denoted $v \in \hat{\partial}f(x)$, if

$$f(x) \geq f(\bar{x}) + \langle v, x - \bar{x} \rangle + o(|x - \bar{x}|).$$

The **limiting subdifferential** at \bar{x} is

$$\partial f(\bar{x}) = \left\{ \lim_{i \rightarrow \infty} v_i : v_i \in \hat{\partial}f(x_i), x_i \rightarrow \bar{x}, f(x_i) \rightarrow f(\bar{x}) \right\}.$$

Definition (critical points)

\bar{x} is a **critical point** of f if $0 \in \partial f(\bar{x})$.

- For **convex** f , critical points are global minimizers.

Preliminaries

For a function $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$, a vector v is a **Frechét subgradient** at \bar{x} , denoted $v \in \hat{\partial}f(x)$, if

$$f(x) \geq f(\bar{x}) + \langle v, x - \bar{x} \rangle + o(|x - \bar{x}|).$$

The **limiting subdifferential** at \bar{x} is

$$\partial f(\bar{x}) = \left\{ \lim_{i \rightarrow \infty} v_i : v_i \in \hat{\partial}f(x_i), x_i \rightarrow \bar{x}, f(x_i) \rightarrow f(\bar{x}) \right\}.$$

Definition (critical points)

\bar{x} is a **critical point** of f if $0 \in \partial f(\bar{x})$.

- For **convex** f , critical points are global minimizers.
- If f is **C^1 -smooth**, criticality reduces to the classical condition $\nabla f(x) = 0$.

Sensitivity Analysis

Suppose $\bar{v} \in \partial f(\bar{x})$ for a function $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$. Consider the perturbed functions

$$f_v(x) = f(x) - \langle v, x \rangle.$$

Sensitivity Analysis

Suppose $\bar{v} \in \partial f(\bar{x})$ for a function $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$. Consider the perturbed functions

$$f_v(x) = f(x) - \langle v, x \rangle.$$

Sensitivity question: How do **critical points** of f_v , near \bar{x} , behave as v varies near \bar{v} ?

Sensitivity Analysis

Suppose $\bar{v} \in \partial f(\bar{x})$ for a function $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$. Consider the perturbed functions

$$f_v(x) = f(x) - \langle v, x \rangle.$$

Sensitivity question: How do **critical points** of f_v , near \bar{x} , behave as v varies near \bar{v} ? Observe

$$0 \in \partial f_v(x) \iff v \in \partial f(x).$$

Sensitivity Analysis

Suppose $\bar{v} \in \partial f(\bar{x})$ for a function $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$. Consider the perturbed functions

$$f_v(x) = f(x) - \langle v, x \rangle.$$

Sensitivity question: How do **critical points** of f_v , near \bar{x} , behave as v varies near \bar{v} ? Observe

$$0 \in \partial f_v(x) \iff v \in \partial f(x).$$

Thus given $\bar{v} \in \partial f(\bar{x})$, we want to understand how solutions x_v of

$$v \in \partial f(x),$$

vary, as we perturb v near \bar{v} .

Motivating Example

Motivating example

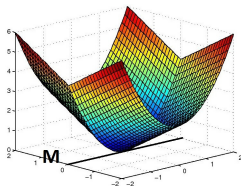


Figure: $f(x, y) = x^2 + |y|$, $M = \{(t, 0) : -1 < t < 1\}$

Motivating Example

Motivating example

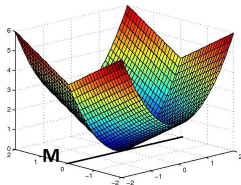


Figure: $f(x, y) = x^2 + |y|$, $M = \{(t, 0) : -1 < t < 1\}$

- Observe $(0, 0) \in \partial f(0, 0)$.

Motivating Example

Motivating example

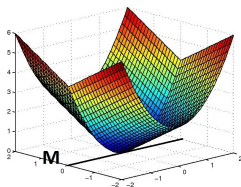


Figure: $f(x, y) = x^2 + |y|$, $M = \{(t, 0) : -1 < t < 1\}$

- Observe $(0, 0) \in \partial f(0, 0)$.

All perturbed solutions x_v of $v \in \partial f(x)$ lie on M

Motivating Example

Motivating example

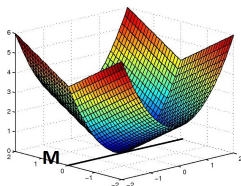


Figure: $f(x, y) = x^2 + |y|$, $M = \{(t, 0) : -1 < t < 1\}$

- Observe $(0, 0) \in \partial f(0, 0)$.

All perturbed solutions x_v of $v \in \partial f(x)$ lie on $M \implies M$ captures **all** the sensitivity information!

Motivating Example

Motivating example

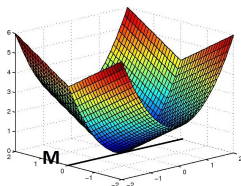


Figure: $f(x, y) = x^2 + |y|$, $M = \{(t, 0) : -1 < t < 1\}$

- Observe $(0, 0) \in \partial f(0, 0)$.

All perturbed solutions x_v of $v \in \partial f(x)$ lie on $M \implies M$ captures **all** the sensitivity information!

- Only the restriction $f|_M$ matters!

Motivating Example

Motivating example

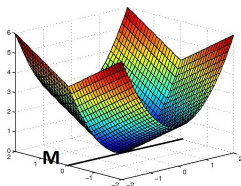


Figure: $f(x, y) = x^2 + |y|$, $M = \{(t, 0) : -1 < t < 1\}$

- Observe $(0, 0) \in \partial f(0, 0)$.

All perturbed solutions x_v of $v \in \partial f(x)$ lie on $M \implies M$ captures **all** the sensitivity information!

- Only the restriction $f|_M$ matters!
- **Goal:** Look for small, well-behaved sets capturing only the essential information.

Finite identification

Consider the system

$$v \in \partial f(x),$$

where $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$ and $\bar{v} \in \partial f(\bar{x})$.

Finite identification

Consider the system

$$v \in \partial f(x),$$

where $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$ and $\bar{v} \in \partial f(\bar{x})$.

Definition (Identifiable sets)

A set $M \subset \mathbf{R}^n$ is **identifiable** at \bar{x} for \bar{v} if

$$\left. \begin{array}{l} x_i \rightarrow \bar{x}, v_i \rightarrow \bar{v} \\ v_i \in \partial f(x_i) \end{array} \right\} \implies x_i \in M \text{ for all large } i,$$

Finite identification

Consider the system

$$v \in \partial f(x),$$

where $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$ and $\bar{v} \in \partial f(\bar{x})$.

Definition (Identifiable sets)

A set $M \subset \mathbf{R}^n$ is **identifiable** at \bar{x} for \bar{v} if

$$\left. \begin{array}{l} x_i \rightarrow \bar{x}, v_i \rightarrow \bar{v} \\ v_i \in \partial f(x_i) \end{array} \right\} \implies x_i \in M \text{ for all large } i,$$

Example (Normal cone map)

Let $\partial f = N_Q$ for a cube $Q \subset \mathbf{R}^3$.

Finite identification

Consider the system

$$v \in \partial f(x),$$

where $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$ and $\bar{v} \in \partial f(\bar{x})$.

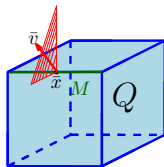
Definition (Identifiable sets)

A set $M \subset \mathbf{R}^n$ is **identifiable** at \bar{x} for \bar{v} if

$$\left. \begin{array}{l} x_i \rightarrow \bar{x}, v_i \rightarrow \bar{v} \\ v_i \in \partial f(x_i) \end{array} \right\} \implies x_i \in M \text{ for all large } i,$$

Example (Normal cone map)

Let $\partial f = N_Q$ for a cube $Q \subset \mathbf{R}^3$.



Finite identification

Consider the system

$$v \in \partial f(x),$$

where $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$ and $\bar{v} \in \partial f(\bar{x})$.

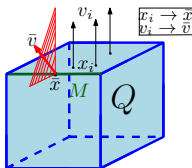
Definition (Identifiable sets)

A set $M \subset \mathbf{R}^n$ is **identifiable** at \bar{x} for \bar{v} if

$$\left. \begin{array}{l} x_i \rightarrow \bar{x}, v_i \rightarrow \bar{v} \\ v_i \in \partial f(x_i) \end{array} \right\} \implies x_i \in M \text{ for all large } i,$$

Example (Normal cone map)

Let $\partial f = N_Q$ for a cube $Q \subset \mathbf{R}^3$.



Finite identification

Consider the system

$$v \in \partial f(x),$$

where $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$ and $\bar{v} \in \partial f(\bar{x})$.

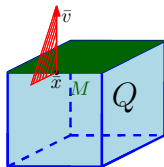
Definition (Identifiable sets)

A set $M \subset \mathbf{R}^n$ is **identifiable** at \bar{x} for \bar{v} if

$$\left. \begin{array}{l} x_i \rightarrow \bar{x}, v_i \rightarrow \bar{v} \\ v_i \in \partial f(x_i) \end{array} \right\} \implies x_i \in M \text{ for all large } i,$$

Example (Normal cone map)

Let $\partial f = N_Q$ for a cube $Q \subset \mathbf{R}^3$.



Prototypical results

Prototypical results

Proposition (Order of growth)

Suppose M is an *identifiable set* at \bar{x} for $0 \in \hat{\partial}f(\bar{x})$.

Prototypical results

Proposition (Order of growth)

Suppose M is an *identifiable set* at \bar{x} for $0 \in \hat{\partial}f(\bar{x})$.

- \bar{x} is a (strict) local minimizer of $f \iff \bar{x}$ is a (strict) local minimizer of f *on* M .

Prototypical results

Proposition (Order of growth)

Suppose M is an *identifiable set* at \bar{x} for $0 \in \hat{\partial}f(\bar{x})$.

- \bar{x} is a (strict) local minimizer of $f \iff \bar{x}$ is a (strict) local minimizer of f *on M* .
- f grows quadratically near $\bar{x} \iff f$ grows quadratically *on M* near \bar{x} .

Prototypical results

Proposition (Order of growth)

Suppose M is an *identifiable set* at \bar{x} for $0 \in \hat{\partial}f(\bar{x})$.

- \bar{x} is a (strict) local minimizer of $f \iff \bar{x}$ is a (strict) local minimizer of f *on M* .
- f grows quadratically near $\bar{x} \iff f$ grows quadratically *on M* near \bar{x} .

Proposition (Uniform projections)

Suppose M is an *identifiable set* at \bar{x} for \bar{v} .

f is *prox-regular* at \bar{x} for $\bar{v} \iff f|_M$ is *prox-regular* at \bar{x} for \bar{v} .

Locally minimal identifiable sets

Clearly all of \mathbf{R}^n is identifiable at \bar{x} for \bar{v} (not interesting). So...

Locally minimal identifiable sets

Clearly all of \mathbf{R}^n is identifiable at \bar{x} for \bar{v} (not interesting). So...

Question: What are the smallest possible identifiable sets?

Locally minimal identifiable sets

Clearly all of \mathbf{R}^n is identifiable at \bar{x} for \bar{v} (not interesting). So...

Question: What are the smallest possible identifiable sets?

Definition

An identifiable set M at \bar{x} for \bar{v} is **locally minimal** if

$$M' \text{ identifiable at } \bar{x} \text{ for } \bar{v} \implies M \subset M', \text{ locally near } \bar{x}.$$

Existence and calculus

Locally minimal identifiable sets exist for

Locally minimal identifiable sets exist for

- piecewise quadratic functions,

Existence and calculus

Locally minimal identifiable sets exist for

- piecewise quadratic functions,
- max-type functions: $f(x) = \max\{g_1(x), \dots, g_k(x)\}$ for \mathbf{C}^1 -smooth g_i .

Existence and calculus

Locally minimal identifiable sets exist for

- piecewise quadratic functions,
- max-type functions: $f(x) = \max\{g_1(x), \dots, g_k(x)\}$ for \mathbf{C}^1 -smooth g_i .
- fully amenable functions: $f(x) = g(F(x))$ where
 - 1 F is \mathbf{C}^2 -smooth,
 - 2 g is (convex) piecewise quadratic,
 - 3 transversality condition holds.

Existence and calculus

Locally minimal identifiable sets exist for

- piecewise quadratic functions,
- max-type functions: $f(x) = \max\{g_1(x), \dots, g_k(x)\}$ for \mathbf{C}^1 -smooth g_i .
- fully amenable functions: $f(x) = g(F(x))$ where
 - 1 F is \mathbf{C}^2 -smooth,
 - 2 g is (convex) piecewise quadratic,
 - 3 transversality condition holds.

A strong chain rule is available.

Intriguing Question: Do locally minimal identifiable sets exist for any convex function?

Topological regularity

Intriguing Question: Do locally minimal identifiable sets exist for any convex function? **No.**

Topological regularity

Intriguing Question: Do locally minimal identifiable sets exist for any convex function? **No.**

This will follow from the following observation.

Topological regularity

Intriguing Question: Do locally minimal identifiable sets exist for any convex function? **No.**

This will follow from the following observation.

Proposition (Topological regularity)

Consider decreasing sequence of open neighborhoods

$$V_1 \supset V_2 \supset V_3 \supset \dots, \text{ with } V_i \downarrow \{\bar{v}\}.$$

*Then M is a **locally minimal identifiable set** at \bar{x} for $\bar{v} \iff M$ coincides with $(\partial f)^{-1}(V_i)$, near \bar{x} , for all large i .*

Topological regularity

Intriguing Question: Do locally minimal identifiable sets exist for any convex function? **No.**

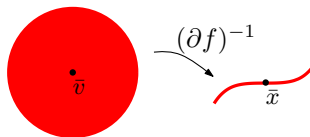
This will follow from the following observation.

Proposition (Topological regularity)

Consider decreasing sequence of open neighborhoods

$$V_1 \supset V_2 \supset V_3 \supset \dots, \text{ with } V_i \downarrow \{\bar{v}\}.$$

*Then M is a **locally minimal identifiable set** at \bar{x} for $\bar{v} \iff M$ coincides with $(\partial f)^{-1}(V_i)$, near \bar{x} , for all large i .*



Topological regularity

Intriguing Question: Do locally minimal identifiable sets exist for any convex function? **No.**

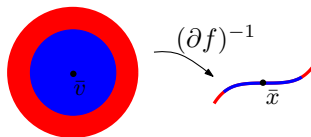
This will follow from the following observation.

Proposition (Topological regularity)

Consider decreasing sequence of open neighborhoods

$$V_1 \supset V_2 \supset V_3 \supset \dots, \text{ with } V_i \downarrow \{\bar{v}\}.$$

*Then M is a **locally minimal identifiable set** at \bar{x} for $\bar{v} \iff M$ coincides with $(\partial f)^{-1}(V_i)$, near \bar{x} , for all large i .*



Topological regularity

Intriguing Question: Do locally minimal identifiable sets exist for any convex function? **No.**

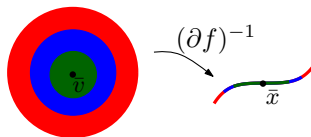
This will follow from the following observation.

Proposition (Topological regularity)

Consider decreasing sequence of open neighborhoods

$$V_1 \supset V_2 \supset V_3 \supset \dots, \text{ with } V_i \downarrow \{\bar{v}\}.$$

*Then M is a **locally minimal identifiable set** at \bar{x} for $\bar{v} \iff M$ coincides with $(\partial f)^{-1}(V_i)$, near \bar{x} , for all large i .*



Existence and calculus

Convex functions may fail to admit locally minimal identifiable sets!

Existence and calculus

Convex functions may fail to admit locally minimal identifiable sets!

Example

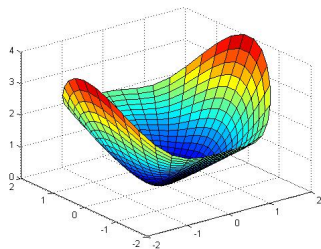


Figure: $f(x, y) = \sqrt{x^4 + y^2}$.

Existence and calculus

Convex functions may fail to admit locally minimal identifiable sets!

Example

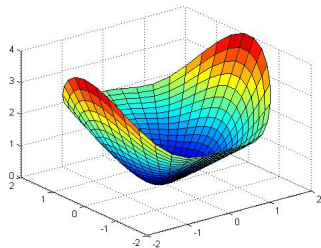


Figure: $f(x, y) = \sqrt{x^4 + y^2}$.

Existence and calculus

Convex functions may fail to admit locally minimal identifiable sets!

Example

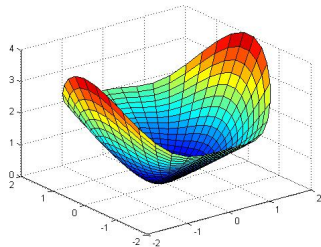


Figure: $f(x, y) = \sqrt{x^4 + y^2}$.

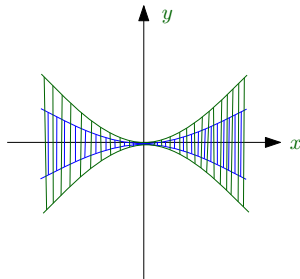


Figure: Level sets: $[|\nabla f| < \epsilon]$.

Existence and calculus

Convex functions may fail to admit locally minimal identifiable sets!

Example

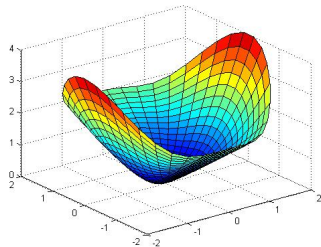


Figure: $f(x, y) = \sqrt{x^4 + y^2}$.

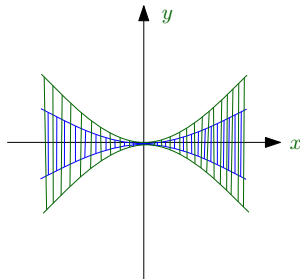


Figure: Level sets: $[|\nabla f| < \epsilon]$.

Existence and calculus

Convex functions may fail to admit locally minimal identifiable sets!

Example

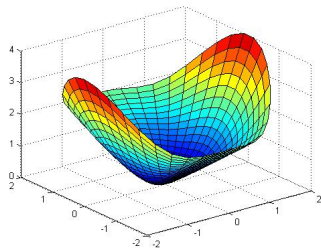


Figure: $f(x, y) = \sqrt{x^4 + y^2}$.

Level sets of $|\nabla f|$ get “pinched”.

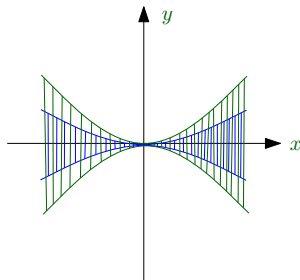


Figure: Level sets: $[|\nabla f| < \epsilon]$.

Existence and calculus

Convex functions may fail to admit locally minimal identifiable sets!

Example

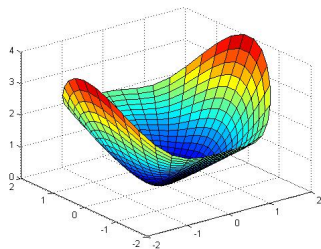


Figure: $f(x, y) = \sqrt{x^4 + y^2}$.

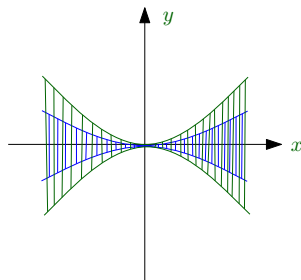


Figure: Level sets: $[|\nabla f| < \epsilon]$.

Level sets of $|\nabla f|$ get “pinched”. There is **no** locally minimal identifiable set at $\bar{x} = (0, 0)$ for $\bar{v} = (0, 0)$.

Algorithmic origins

- Intuitive idea of finite identification, in this setting, is old.

Algorithmic origins

- Intuitive idea of finite identification, in this setting, is old.
- Some algorithms for solving $\min_{x \in Q} f(x)$ would stop in finite time (proximal point algorithm [Rockafellar '76](#)).

Algorithmic origins

- Intuitive idea of finite identification, in this setting, is old.
- Some algorithms for solving $\min_{x \in Q} f(x)$ would stop in finite time (proximal point algorithm [Rockafellar '76](#)).
- Many algorithms would generate iterates that eventually lie on a distinguished subset of Q

Algorithmic origins

- Intuitive idea of finite identification, in this setting, is old.
- Some algorithms for solving $\min_{x \in Q} f(x)$ would stop in finite time (proximal point algorithm [Rockafellar '76](#)).
- Many algorithms would generate iterates that eventually lie on a distinguished subset of Q (subgradient projection [Calamai-Moré '87](#), Newton-like methods [Burke-Moré '88](#), stochastic gradient methods [Wright '11](#)).

Algorithmic origins

- Intuitive idea of finite identification, in this setting, is old.
- Some algorithms for solving $\min_{x \in Q} f(x)$ would stop in finite time (proximal point algorithm [Rockafellar '76](#)).
- Many algorithms would generate iterates that eventually lie on a distinguished subset of Q (subgradient projection [Calamai-Moré '87](#), Newton-like methods [Burke-Moré '88](#), stochastic gradient methods [Wright '11](#)).
- One may try to exploit a nice identifiable set, if one exists; perhaps a \mathbf{C}^2 -manifold.

For simplicity, we work with **normal cones** to a set $Q \subset \mathbf{R}^n$, that is

$$\hat{N}_Q(x) := \hat{\partial}\delta_Q(x), \quad N_Q(x) := \partial\delta_Q(x).$$

Identifiable manifolds

For simplicity, we work with **normal cones** to a set $Q \subset \mathbf{R}^n$, that is

$$\hat{N}_Q(x) := \hat{\partial}\delta_Q(x), \quad N_Q(x) := \partial\delta_Q(x).$$

When there exists an identifiable subset $M \subset Q$ that is a **manifold**, things simplify drastically.

Identifiable manifolds

For simplicity, we work with **normal cones** to a set $Q \subset \mathbf{R}^n$, that is

$$\hat{N}_Q(x) := \hat{\partial}\delta_Q(x), \quad N_Q(x) := \partial\delta_Q(x).$$

When there exists an identifiable subset $M \subset Q$ that is a **manifold**, things simplify drastically.

Proposition (Identifiable manifolds)

Identifiable manifolds $M \subset Q$ are automatically locally minimal.

Partly smooth manifolds

To get a good handle on sensitivity analysis, Lewis '03 introduced partly smooth manifolds.

Partly smooth manifolds

To get a good handle on sensitivity analysis, Lewis '03 introduced partly smooth manifolds.

Definition

Q is partly smooth, with respect to M , at $\bar{x} \in M$ for $\bar{v} \in N_Q(\bar{x})$, if there exist open neighborhoods U of \bar{x} and V of \bar{v} satisfying

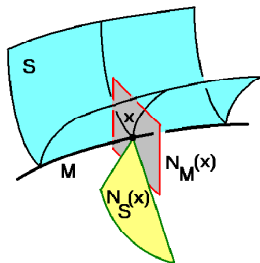
Partly smooth manifolds

To get a good handle on sensitivity analysis, Lewis '03 introduced partly smooth manifolds.

Definition

Q is partly smooth, with respect to M , at $\bar{x} \in M$ for $\bar{v} \in N_Q(\bar{x})$, if there exist open neighborhoods U of \bar{x} and V of \bar{v} satisfying

- 1 (regularity) $V \cap N_Q(x) \subset \hat{N}_Q(x)$ for each $x \in U \cap M$,
- 2 (sharpness) $\text{span } \hat{N}_Q(\bar{x}) = N_M(\bar{x})$.
- 3 (continuity) The mapping $x \mapsto V \cap N_Q(x)$ is continuous on M at \bar{x} .



Partly smooth manifolds

Theorem (Characterization)

Consider a \mathbf{C}^2 -manifold M with $\bar{x} \in M \subset Q$. Then the following are equivalent.

Partly smooth manifolds

Theorem (Characterization)

Consider a \mathbf{C}^2 -manifold M with $\bar{x} \in M \subset Q$. Then the following are equivalent.

- ① M is an identifiable manifold around \bar{x} for \bar{v} .
- ②
 - Q is partly smooth with respect to M at \bar{x} for \bar{v} .
 - Q is prox-regular at \bar{x} for \bar{v} .
 - the strong inclusion $\bar{v} \in \text{ri } \hat{N}_Q(x)$ holds.

Partly smooth manifolds

Theorem (Characterization)

Consider a \mathbf{C}^2 -manifold M with $\bar{x} \in M \subset Q$. Then the following are equivalent.

- ① M is an identifiable manifold around \bar{x} for \bar{v} .
- ②
 - Q is partly smooth with respect to M at \bar{x} for \bar{v} .
 - Q is prox-regular at \bar{x} for \bar{v} .
 - the strong inclusion $\bar{v} \in \text{ri } \hat{N}_Q(x)$ holds.
- Implication \Uparrow was mostly proven in [Hare-Lewis '04](#).

Partly smooth manifolds

Theorem (Characterization)

Consider a \mathbf{C}^2 -manifold M with $\bar{x} \in M \subset Q$. Then the following are equivalent.

- ① M is an identifiable manifold around \bar{x} for \bar{v} .
- ②
 - Q is partly smooth with respect to M at \bar{x} for \bar{v} .
 - Q is prox-regular at \bar{x} for \bar{v} .
 - the strong inclusion $\bar{v} \in \text{ri } \hat{N}_Q(x)$ holds.
- Implication \uparrow was mostly proven in Hare-Lewis '04.
- Equivalence yields intuitive interpretation of **partial smoothness**, **prox-regularity**, and **nondegeneracy** (all sophisticated concepts).

Smooth dependence

Consider

Smooth dependence

Consider

$$\begin{aligned} P(v) : \quad & \max \quad \langle v, x \rangle, \\ & \text{s.t.} \quad x \in Q. \end{aligned}$$

Smooth dependence

Consider

$$\begin{aligned} P(v) : \quad & \max \quad \langle v, x \rangle, \\ & \text{s.t.} \quad x \in Q. \end{aligned}$$

- When is there a **smooth dependence** of critical points on v ?

Smooth dependence

Consider

$$\begin{aligned} P(v) : \quad & \max \quad \langle v, x \rangle, \\ & \text{s.t.} \quad x \in Q. \end{aligned}$$

- When is there a **smooth dependence** of critical points on v ?
- This is a difficult question in general,

Smooth dependence

Consider

$$\begin{aligned} P(v) : \quad & \max \quad \langle v, x \rangle, \\ & \text{s.t.} \quad x \in Q. \end{aligned}$$

- When is there a **smooth dependence** of critical points on v ?
- This is a difficult question in general, but when an identifiable manifold exists, it is straightforward!

Smooth dependence

Consider

$$\begin{aligned} P(v) : \quad & \max \quad \langle v, x \rangle, \\ & \text{s.t.} \quad x \in Q. \end{aligned}$$

- When is there a **smooth dependence** of critical points on v ?
- This is a difficult question in general, but when an identifiable manifold exists, it is straightforward! Just need to consider **curvature** of M .

Smooth dependence

Theorem (Sensitivity)

Suppose M is an identifiable manifold at \bar{x} for \bar{v} and \bar{x} is a local maximizer of $\langle \bar{v}, \cdot \rangle$ restricted to M . Then the following are equivalent.

Smooth dependence

Theorem (Sensitivity)

Suppose M is an identifiable manifold at \bar{x} for \bar{v} and \bar{x} is a local maximizer of $\langle \bar{v}, \cdot \rangle$ restricted to M . Then the following are equivalent.

- 1 In a localized sense, the *critical point map*,

$$v \mapsto N_Q^{-1}(v),$$

is single-valued, \mathbf{C}^1 -smooth, and onto a neighborhood of \bar{x} in M .

Smooth dependence

Theorem (Sensitivity)

Suppose M is an identifiable manifold at \bar{x} for \bar{v} and \bar{x} is a local maximizer of $\langle \bar{v}, \cdot \rangle$ restricted to M . Then the following are equivalent.

- ① In a localized sense, the **critical point map**,

$$v \mapsto N_Q^{-1}(v),$$

is single-valued, \mathbf{C}^1 -smooth, and onto a neighborhood of \bar{x} in M .

- ② **Second-order decay**: There exists $\rho > 0$ such that,

$$\langle \bar{v}, \bar{x} \rangle \geq \langle \bar{v}, x \rangle + \rho |x - \bar{x}|^2, \text{ for all } x \in M \text{ near } \bar{x}.$$

Generic Properties

- How **typical** are identifiable manifolds? Second-order growth at minimizers?

Generic Properties

- How **typical** are identifiable manifolds? Second-order growth at minimizers?
- We answer this question in the setting of **semi-algebraic sets**:

Generic Properties

- How **typical** are identifiable manifolds? Second-order growth at minimizers?
- We answer this question in the setting of **semi-algebraic sets**: represented as finite union of sets, each defined by finitely many polynomial inequalities.

Generic Properties

- How **typical** are identifiable manifolds? Second-order growth at minimizers?
- We answer this question in the setting of **semi-algebraic sets**: represented as finite union of sets, each defined by finitely many polynomial inequalities.
- Large class of sets for which the word **typical** has a canonical meaning.

Generic Properties

Theorem

Suppose $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$ is semi-algebraic. Consider the perturbed functions

$$f_v(x) := f(x) - \langle v, x \rangle.$$

Then for a “typical” $v \in \mathbf{R}^n$,

Generic Properties

Theorem

Suppose $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$ is semi-algebraic. Consider the perturbed functions

$$f_v(x) := f(x) - \langle v, x \rangle.$$

Then for a “typical” $v \in \mathbf{R}^n$,

- 1 f_v has finitely many *critical points* x_v .

Generic Properties

Theorem

Suppose $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$ is semi-algebraic. Consider the perturbed functions

$$f_v(x) := f(x) - \langle v, x \rangle.$$

Then for a “typical” $v \in \mathbf{R}^n$,

- 1 f_v has finitely many *critical points* x_v .
- 2 f_v admits an *identifiable manifold* M_v near each critical point x_v for 0.

Generic Properties

Theorem

Suppose $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$ is semi-algebraic. Consider the perturbed functions

$$f_v(x) := f(x) - \langle v, x \rangle.$$

Then for a “typical” $v \in \mathbf{R}^n$,

- 1 f_v has finitely many **critical points** x_v .
- 2 f_v admits an **identifiable manifold** M_v near each critical point x_v for 0.
- 3 every local minimizer x_v of f_v restricted to M_v is a **strong local minimizer** of f_v , that is

$$f_v(x) > f_v(x_v) + \rho|x - x_v|^2, \text{ for all } x \text{ near } x_v.$$

Generic Properties

Theorem

Suppose $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$ is semi-algebraic. Consider the perturbed functions

$$f_v(x) := f(x) - \langle v, x \rangle.$$

Then for a “typical” $v \in \mathbf{R}^n$,

- 1 f_v has finitely many **critical points** x_v .
- 2 f_v admits an **identifiable manifold** M_v near each critical point x_v for 0.
- 3 every local minimizer x_v of f_v restricted to M_v is a **strong local minimizer** of f_v , that is

$$f_v(x) > f_v(x_v) + \rho|x - x_v|^2, \text{ for all } x \text{ near } x_v.$$

This extends a result of [Bolte, Daniilidis, Lewis '11](#).

Generic Properties

Theorem

Suppose $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$ is semi-algebraic. Consider the perturbed functions

$$f_v(x) := f(x) - \langle v, x \rangle.$$

Then for a “typical” $v \in \mathbf{R}^n$,

- 1 f_v has finitely many **critical points** x_v .
- 2 f_v admits an **identifiable manifold** M_v near each critical point x_v for 0.
- 3 every local minimizer x_v of f_v restricted to M_v is a **strong local minimizer** of f_v , that is

$$f_v(x) > f_v(x_v) + \rho|x - x_v|^2, \text{ for all } x \text{ near } x_v.$$

This extends a result of Bolte, Daniilidis, Lewis '11.

Second-order sufficient conditions for optimality are almost necessary (Spingarn-Rockafellar '79).

Summary

- Presented the intuitive notion of **identifiable sets**.
- Showed how these objects relate to previously developed concepts (Partial Smoothness, prox-regularity, etc).
- Existence of identifiable sets (manifolds) leads to significant insight about the problem.

Future directions

- Could consider **identifiability** for more general perturbations

$$\min_{x \in \mathbf{R}^n} f(x, v),$$

Future directions

- Could consider **identifiability** for more general perturbations

$$\min_{x \in \mathbf{R}^n} f(x, v),$$

or more generally

$$v \in G(x),$$

for some set-valued mapping G . (**Variational Inequalities, equilibria**)

Future directions

- Could consider **identifiability** for more general perturbations

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

or more generally

$$v \in G(x),$$

for some set-valued mapping G . (**Variational Inequalities, equilibria**)

- Compute identifiable manifolds for common convex cones (**positive polynomials** and the **SOS** cone).

Future directions

- Could consider **identifiability** for more general perturbations

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

or more generally

$$v \in G(x),$$

for some set-valued mapping G . (**Variational Inequalities, equilibria**)

- Compute identifiable manifolds for common convex cones (**positive polynomials** and the **SOS** cone).
- Explore **duality** of identifiable sets.

Future directions

- Could consider **identifiability** for more general perturbations

$$\min_{x \in \mathbf{R}^n} f(x, v),$$

or more generally

$$v \in G(x),$$

for some set-valued mapping G . (**Variational Inequalities, equilibria**)

- Compute identifiable manifolds for common convex cones (**positive polynomials** and the **SOS** cone).
- Explore **duality** of identifiable sets.

Example

Semi-definite cone \mathbf{S}_+^n stratifies into partly smooth manifolds

$$\{X \in \mathbf{S}_+^n : \text{rank } X = k\}, \text{ for } k = 0, \dots, n,$$

*and is self-dual. **Natural duality** between these manifolds.*

Thank you.