

# Variational analysis with substructure

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- Intuitive notion of **identifiable sets**.
- Existence, calculus, properties.
- Connection to critical cones (**Generalized Reduction Lemma**).
- Illustration: **Spectral functions**.
- **Generic** existence (**semi-algebraic** setting).

Definition (Generalized critical points)

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

## Motivation (Sensitivity Analysis)

Consider the perturbed functions

$$f_v(x) = f(x) - \langle v, x \rangle. \quad [\textit{For simplicity}],$$

and suppose  $\bar{x}$  is critical for  $f_{\bar{v}}$ , that is  $\bar{v} \in \partial f(\bar{x})$ .

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**Sensitivity question:** How do solutions  $x_v$  of

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

vary, as we perturb  $v$  near  $\bar{v}$ ?

# Motivating Example

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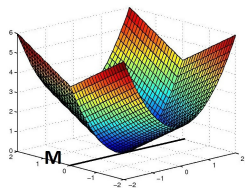


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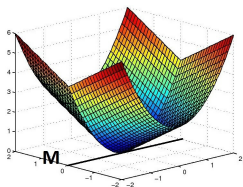


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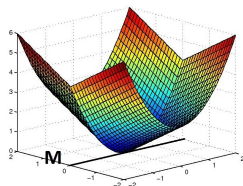


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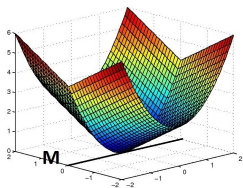


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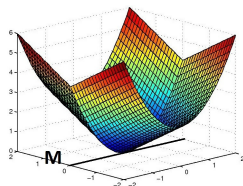


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- Only the restriction  $f|_M$  matters!
- **Goal:** Look for small, well-behaved sets capturing only the essential information.

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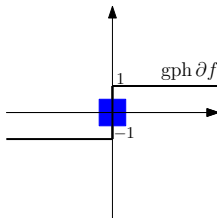
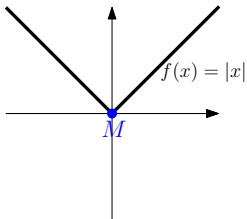
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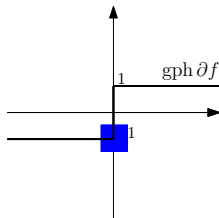
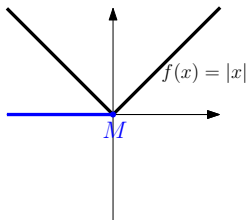
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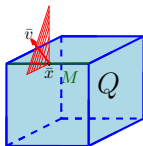
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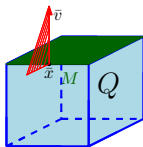
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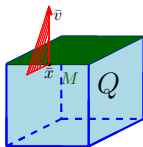
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Let  $\partial f = N_Q$  for a cube  $Q \subset \mathbf{R}^3$ .



In this case  $M = \bar{x} + K_Q(\bar{x}, \bar{v})$ .

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- $\bar{x}$  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  $\bar{x}$ .

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### Definition

An identifiable set  $M$  at  $(\bar{x}, \bar{v}) \in \text{gph } \partial f$  is **locally minimal** if

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



Locally minimal identifiable sets exist for

- fully amenable functions:  $f(x) = g(F(x))$  where
  - 1  $F$  is  $\mathbf{C}^2$ -smooth,
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A strong chain rule is available for composite functions

$$f(x) = g(F(x)).$$

The **critical cone** of a convex  $Q$  at  $\bar{x}$  for  $\bar{v} \in N_Q(\bar{x})$  is

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# Dimension Reduction

Proposition (Reduction Lemma due to Robinson)

*If  $Q$  is polyhedral, then*

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Let  $M$  be a (prox-regular) identifiable set at  $(\bar{x}, \bar{v}) \in \text{gph } N_Q(\bar{x})$ .

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May use this to study **nonpolyhedral** variational inequalities!

# Identifiable manifolds

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- When an identifiable manifold exists, **nonsmoothness** is **not** intrinsic...So can reduce analysis to the **classical** setting.

# Lifts of identifiable manifolds

Consider  $\mathbf{S}^n := \{n \times n \text{ symmetric matrices}\}$  and the **eigenvalue map**

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Identifiable manifolds “**lift**”: (D, Lewis), (Daniilidis, Malick, Sendov)

$M$  identifiable manifold at  $(\bar{x}, \bar{v}) \in \text{gph } \partial f$

$\implies \lambda^{-1}(M)$  identifiable manifold at  $(\bar{X}, \bar{V}) \in \text{gph } \partial(f \circ \lambda)$ .

# Generic Properties

History: Rockafellar-Spingarn '79, considered problems

$$P(\mathbf{v}, \mathbf{u}) : \quad \min_x f(x) - \langle \mathbf{v}, x \rangle,$$
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Theorem (Rockafellar-Spingarn '79)

- For almost all  $(\mathbf{v}, \mathbf{u})$ , at every minimizer of  $P(\mathbf{v}, \mathbf{u})$ :

*Active manifold*: active gradients are independent

*Strict complementarity*: multipliers are strictly positive and

*Quadratic growth*: objective function grows quadratically.

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*Active manifold:* existence of an **identifiable** manifold

*Strict complementarity:*  $0 \in \text{ri } \hat{\partial} f_{\mathbf{v}}(x_{\mathbf{v}})$

*Quadratic growth:*  $f_{\mathbf{v}}$  grows quadratically near  $x_{\mathbf{v}}$ .

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Theorem (Ioffe)

For semi-algebraic  $F: \mathbf{R}^n \rightrightarrows \mathbf{R}^m$ , for generic  $v \in \mathbf{R}^m$ , have

$$x \in F^{-1}(v) \implies F \text{ is } \textit{metrically regular} \text{ at } (x, v).$$

# Generic Properties

Semi-algebraic subdifferential graphs are **not** too big, **not** too small, but **just right**:

Theorem (D, Ioffe, Lewis)

For lsc, semi-algebraic  $f: \mathbf{R}^n \rightarrow \overline{\mathbf{R}}$ , we have

$$\dim \text{gph } \partial f = n,$$

even **locally** around any pair  $(x, v) \in \text{gph } \partial f$ .

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Further, if  $\dim \text{gph } F = n = m$ , then **strong metric regularity** is typical. **Strong metric regularity** of  $\partial f$  (i.e. **tilt-stability**) is equivalent to a uniform quadratic growth condition (D, Lewis '12).

# Summary

- Presented the intuitive notion of **identifiable sets**.
- Showed how identifiable sets capture the essence of previously developed concepts (**dimension reduction**, **critical cones**, **optimality conditions**).
- Illustration: **spectral functions**.
- Generic properties of **semi-algebraic** optimization problems.

Thank you.