

# Nonsmooth critical values and Sard type results

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VARIATIONAL ANALYSIS, OPTIMIZATION AND  
QUANTITATIVE FINANCE

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## The (classical) Sard theorem

### CRITICAL SET (SMOOTH CASE)

- $F : \mathbb{R}^n \rightarrow \mathbb{R}^m$  differentiable function

#### Definition (Critical point)

A point  $x_0$  is called *critical*, if  $dF(x_0)$  is not surjective.

- $S :=$  set of critical points
- $F(S) :=$  set of critical values

## Theorem (Sard theorem)

For every  $C^k$ -smooth function  $F : \mathbb{R}^n \rightarrow \mathbb{R}^m$  we have  $m(F(S)) = 0$ , provided  $k \geq n - m + 1$ .

### APPLICATIONS (DIFFERENTIAL GEOMETRY)

- Immersed submanifolds  $\mathcal{N} \subset \mathcal{M}$  of positive co-dimension have zero measure in  $\mathcal{M}$ .
- (*Whitney embedding*)  
 $C^\infty$  mnfds of dimension  $d$  embed into  $\mathbb{R}^{2d+1}$ .

## APPLICATIONS (OPTIMIZATION)

- Constraint qualification condition (*genericity result*)

$$\begin{cases} \min & f(x) \\ & h_i(x) = r_i \\ & i \in \{1, \dots, m\} \end{cases} \longleftrightarrow \begin{cases} H : \mathbb{R}^n \rightarrow \mathbb{R}^m \\ H = (h_1, \dots, h_m) \end{cases} \quad (\mathcal{P}_r)$$

The set of data  $r = (r_1, \dots, r_m) \in \mathbb{R}^m$  for which the constraints **do not** satisfy the *qualification condition* at the solution  $x_*$  is of **measure zero**.

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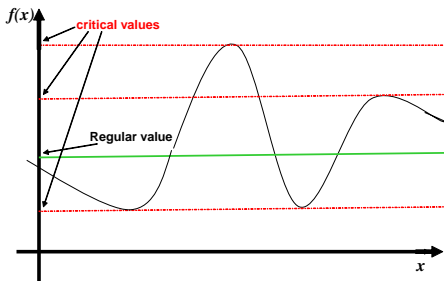
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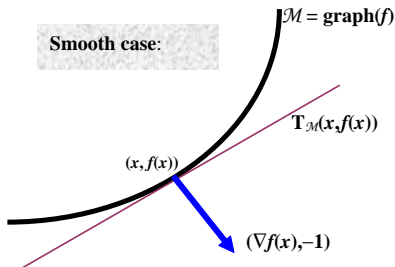
$$x_* \text{ solution of } (\mathcal{P}_r) \implies \nabla f(x_*) = \sum_{i=1}^m \lambda_i \nabla h_i(x_*)$$

## REAL-VALUED CASE

### Theorem (Morse theorem)

For every  $C^k$  function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  we have  $m(f(S)) = 0$ , provided  $k \geq n$ .





tangent space (to the graph)

$$T_{\text{graph}(f)}(u)$$



tangent cone (to the epigraph)

$$T_{\text{epi}(f)}(u)$$

normal space (to the graph)

$$N_{\text{graph}(f)}(u)$$



normal cone (to the epigraph)

$$N_{\text{epi}(f)}(u)$$

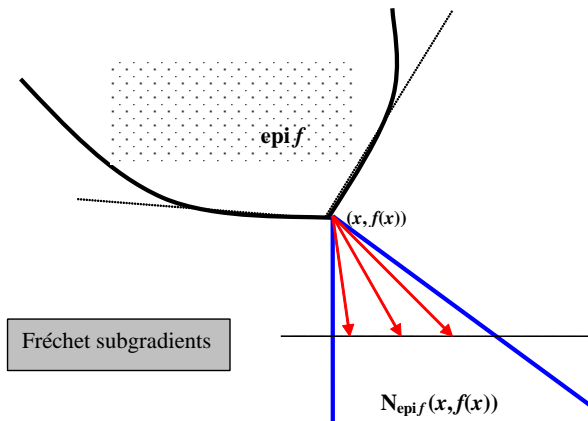
gradient

$$\nabla f(x) \text{ of } f \text{ at } x$$



subgradient

$$x^* \in \partial f(x) \text{ of } f \text{ at } x$$



### Unilateral definition:

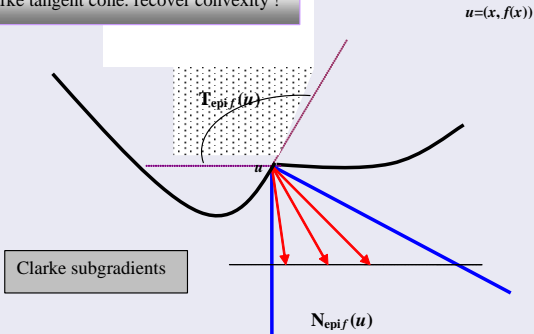
$\partial f(x)$  may be empty (if  $f$  presents an “inward” corner).



## Definition (Clarke subgradient - Lipschitz case)

$$\partial^C f(x) = \{p \in \mathbb{R}^n : (p, -1) \in N_{\text{epi}f}(x, f(x))\}.$$

Clarke tangent cone: recover convexity !



Clarke subgradients

$D_f :=$  differentiability points (dense by Rademacher thm)

Clarke subdifferential

$$\partial f(x_0) = \text{conv} \left\{ \lim_{x_n \rightarrow x_0} \nabla f(x_n) : \{x_n\}_n \subset D_f \right\}$$

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CRITICAL SET (NONSMOOTH CASE)

- $f : \mathbb{R}^n \rightarrow \mathbb{R}$  (nonsmooth) Lipschitz

Definition (Critical point)

A point  $x_0$  is called *critical*, if  $0 \in \partial f(x_0)$ .

- If  $f$  is  $C^1$  then (it is locally Lipschitz and)

$$\partial f(x) = \{\nabla f(x)\}, \quad \text{for all } x \in \mathbb{R}^n$$

- No hope to treat the general nonsmooth case if  $n > 1$  !

### Theorem (Recall – smooth case)

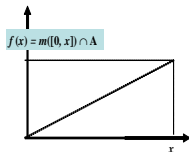
Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  be a  $C^k$ -smooth function. Then  $m(f(S)) = 0$  provided  $k \geq n$  (or  $k = n - 1$  and  $D^{n-1}f$  is Lipschitz).

## STRONG FAILURE OF SARD

$A \subset \mathbb{R}_+$  measurable:  $\forall$  (nontrivial) interval  $I$

$$0 < m(A \cap I) < m(I)$$

$$f(x) = m(A \cap [0, x]) = \int_0^x \chi_A(t) dt$$



$\partial f(x) = [0, 1], \forall x \implies S = [0, 1]$  and  $f$  is  $\nearrow$  (!)

## GENERIC PATHOLOGY

Theorem (X. Wang, 1998)

*All points of a generic Lipschitz function are critical.*

## SEEKING A GOOD STRUCTURE

Recompense *nonsmoothness* assuming **structure**:

- Convex paradigm (convexity, generalizations)
- Semialgebraic paradigm (semialgebraicity, tameness)

## The convex paradigm does not lead far...

### Fact (trivial)

*Convex functions satisfy Morse-Sard theorem.*

### Fact (Failure for DC functions)

*Every  $C^2$  function can be written as a difference of a  $C^2$  convex and a convex quadratic function (on compact sets)*

### Corollary (Failure in 3-dim)

*The Morse-Sard theorem fails for semiconvex (and thus DC) functions in  $\mathbb{R}^3$ .*

## Semialgebraic paradigm: strong conclusions

- Semialgebraicity (tameness)  $\implies C^p$ -Whitney stratifiability

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- Semialgebraicity (tameness)  $\implies C^p$ -Whitney stratifiability
- The set of critical values is semialgebraic  
(Quantifier elimination principle)

### Corollary (Tame case)

*(Locally) finite nonsmooth critical values.*



## Semialgebraic paradigm: strong conclusions

- Semialgebraicity (tameness)  $\implies C^p$ -Whitney stratifiability
- The set of critical values is semialgebraic  
Quantifier elimination principle (★)

### Corollary (Tame case)

*(Locally) finite nonsmooth critical values.*

- (★) The set of critical points is semialgebraic  
( $\rightarrow$  finite union of arc-connected parts)

## Further extensions:

### nonsmooth (and nontame!) case

- $f$  is a *maximum* of smooth functions  
(*scalar case*)
- $f$  is a (Lipschitz) *selection* of smooth functions  
(*vector case*)

Let  $f_1, f_2 : \mathbb{R}^n \rightarrow \mathbb{R}$   $C^k$ -smooth,  $k \geq n$ .

$$f(x) = \max\{f_1(x), f_2(x)\},$$

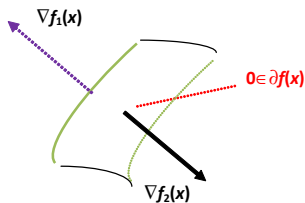
Clarke subdifferential:

$$\partial f(x) = \begin{cases} \nabla f_1(x) & \text{if } f_1(x) > f_2(x) \\ \nabla f_2(x) & \text{if } f_1(x) < f_2(x) \\ \text{co}\{\nabla f_1(x), \nabla f_2(x)\} & \text{if } f_1(x) = f_2(x) \end{cases}$$

Clarke critical points:

$$S = \{x \in \mathbb{R}^n : 0 \in \partial f(x)\}$$

A SIMPLE OBSERVATION



$$\Phi = f_1 - f_2 \implies \nabla \Phi(x) \neq 0.$$

Set

$$S_i := \{x \in \mathbb{R}^n : \nabla f_i(x) = 0\}$$

and

$$\mathcal{M} = \{x \in \mathbb{R}^n : f(x) = f_1(x) = f_2(x)\}$$

Fact

If  $x \in \mathcal{M}$  is Clarke critical and  $x \notin S_1 \cup S_2$  (i.e.  $\nabla f_i(x) \neq 0 \forall i$ )  
then

- $\mathcal{M}$  is a  $C^k$  smooth submanifold of  $\mathbb{R}^n$  around  $x$
- $f|_{\mathcal{M}}$  is  $C^k$ -smooth on  $\mathcal{M}$  and  $\nabla_{\mathcal{R}}(f|_{\mathcal{M}})(x) = 0$ .

## Deduce a nonsmooth result from the classical Morse-Sard theorem

Apply classical Morse-Sard theorem to  $f_1, f_2$  and  $f|_{\mathcal{M}}$

$$\implies f_1(S_1), f_2(S_2) \text{ and } f|_{\mathcal{M}}(S) \text{ null.}$$

- Conclude:

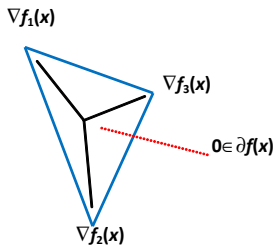
$$f(S) \subset f_1(S_1) \cup f_2(S_2) \cup f|_{\mathcal{M}}(S)$$

is null.

## Fact

*This argument extends to finite continuous selections*

$$f(x) \in \{f_1(x), \dots, f_k(x)\}.$$



$$\mathcal{M} = \{x : f_1(x) = f_2(x) = f_3(x)\}$$

Assume

$$\bar{x} \in \mathcal{M}, \quad 0 \in \text{co}\{\nabla f_1(\bar{x}), \nabla f_2(\bar{x}), \nabla f_3(\bar{x})\}$$

and

$$0 \notin \text{co}\{\nabla f_1(\bar{x}), \nabla f_2(\bar{x})\} \cup \text{co}\{\nabla f_2(\bar{x}), \nabla f_3(\bar{x})\} \cup \text{co}\{\nabla f_1(\bar{x}), \nabla f_3(\bar{x})\}$$

Then:

- $\mathcal{M}$  is a  $C^k$  submanifold of  $\mathbb{R}^n$  around  $\bar{x}$
- $f|_{\mathcal{M}}$  (is  $C^k$ -smooth and)  $\nabla_{\mathcal{R}}(f|_{\mathcal{M}})(x) = 0$ .



## GENERAL NONSMOOTH MORSE-SARD RESULT

Consider  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  *continuous* selection:

$$f(x) \in \{F(x, t) : t \in T\}, \quad \text{for all } x \in \mathbb{R}^n$$

- $T$  compact **countable**
- $x \mapsto F(x, t)$   $C^k$ -smooth ( $k \geq n$ )
- $F, \nabla_x F$  jointly continuous.

Theorem (Barbet, Dambrine, Daniilidis (Adv. Math. 2013))

*$f$  is locally Lipschitz and satisfies the Morse-Sard theorem.*

## Corollary (lower- $C^k$ functions over a *countable* set)

Assume :

- $T$  *countable* compact,
- $F(\cdot, t)$   $C^k$ -smooth ( $k \geq n$ ) and
- $F, \nabla_x F$  (jointly) continuous.

Then the locally Lipschitz function

$$f(x) = \max_{t \in T} F(x, t), \quad x \in \mathbb{R}^n$$

satisfies the Morse–Sard theorem.

## TOOLS OF THE PROOF

- Cantor-Bendixon index
- Formula for the Clarke subgradients:

$$\partial f(x) \subset \text{conv} \{ \nabla_x F(x, t) : t \in T(x) \}$$

where  $T(x)$  = active indices of  $f$  at  $x$ .

- Finite representation of critical points (*Caratheodory*)

### Fact

*The proof uses the Morse–Sard theorem, but it also recovers it.*

## VECTORIAL CASE (SARD THEOREM) ?

$f : \mathbb{R}^n \rightarrow \mathbb{R}^p$  *continuous* selection:

$$f(x) \in \{F(x, t) : t \in T\}, \quad \text{for all } x \in \mathbb{R}^n$$

where  $F : \mathbb{R}^n \times T \rightarrow \mathbb{R}^p$

- $T$  compact **countable**
- $x \mapsto F(x, t)$   $C^k$ -smooth ( $k \geq n$ )
- $F, \nabla_x F$  jointly continuous.

## VECTORIAL CASE (SARD THEOREM) ?

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where  $F : \mathbb{R}^n \times T \rightarrow \mathbb{R}^p$

Question (How to proceed ?)

*The previous approach (i.e. Riemann gradients to naturally arising submanifolds) requires  $f$  to be scalar valued.*

## PREPARATORY SARD THEOREM

$$\begin{cases} \Psi : \text{inn } \Delta^m \times \mathcal{M} \rightarrow \mathbb{R}^p \\ \Psi(\lambda, x) := \sum_{i=0}^m \lambda_i \phi^i(x). \end{cases}$$

where

$$\dim \mathcal{M} = n, \quad \phi^i : \mathcal{M} \rightarrow \mathbb{R}^p$$

$$\begin{cases} \Psi(\lambda, x) := \sum_{i=0}^m \lambda_i \phi^i(x), \\ (\lambda, x) \in \text{inn } \Delta^m \times \mathcal{M}. \end{cases}$$

Define  $\widehat{\text{Crit}}\Psi$  *strongly critical points*:

$$(\lambda, x) \in \widehat{\text{Crit}}\Psi \iff \begin{cases} \phi^i(x) = \phi^0(x), & i \in \{0, \dots, m\} \\ \text{rank} \left( \sum_{i=0}^m \lambda_i D\phi^i(x) \right) < p. \end{cases}$$

Theorem (“Preparatory Sard theorem”)

$\mathcal{M}$  is  $C^k$  mnfd,  $\dim \mathcal{M} = n$

$\phi^i : \mathcal{M} \rightarrow \mathbb{R}^p, i \in \{0, \dots, m\}$  of class  $C^k$  and

$$\begin{cases} \Psi : \text{inn } \Delta^m \times \mathcal{M} \rightarrow \mathbb{R}^p \\ \Psi(\lambda, x) := \sum_{i=0}^m \lambda_i \phi^i(x). \end{cases}$$

If  $k \geq n - p + 1$ , then  $\Psi(\widehat{\text{Crit}}\Psi)$  is null in  $\mathbb{R}^p$ .

## Proof of Preparatory Sard theorem.

- Wnlog  $\mathcal{M} = \mathbb{R}^n$  and  $\text{dom } f = \text{dom } \phi^i = [0, 1]^n$
- Divide  $[0, 1]^n$  in  $\ell$  subcubes.
- Approximate  $\phi^i$  by a polynomial  $P_j^i$   
(in the subcube  $j \in \{1, \dots, \ell\}$ )
- Apply the *Yomdin technique* to estimate the  $\varepsilon$ -critical values. □

## APPLICATION (NONSMOOTH SARD THEOREM - VECTOR CASE)

$$\begin{cases} f(x) \in \{F(x, t) : t \in T\} \\ F : \mathbb{R}^n \times T \rightarrow \mathbb{R}^p \end{cases}$$



Assume  $T = \{1, \dots, \ell\}$ .

Fix  $m_i \in \{0, 1, \dots, n\}$ ,  $i \in \{1, \dots, p\}$  and set:

$$\left\{ \begin{array}{l} G : \text{inn } \Delta^{m_1} \times \dots \times \text{inn } \Delta^{m_p} \times \mathbb{R}^n \longrightarrow \mathbb{R}^p \\ G(\lambda^1, \dots, \lambda^p, x) = \left( \sum_{j_1=0}^{m_1} \lambda_{j_1}^1 F_1(x, t_{j_1}^1), \dots, \sum_{j_p=0}^{m_p} \lambda_{j_p}^p F_p(x, t_{j_p}^p) \right) \end{array} \right.$$

Fact (transferring criticality from  $f$  to some  $G$ )

$\bar{x} \in \text{Crit } f \implies \exists \{m_1, \dots, m_p\}$  and  $\lambda^i \in \text{inn } \Delta^{m_i}$  (depending on  $\bar{x}$ ):

- $(\lambda^1, \dots, \lambda^p, \bar{x}) \in \widehat{\text{Crit}} G$ ;
- $f(\bar{x}) = G(\lambda^1, \dots, \lambda^p, \bar{x})$ .

- Transferring criticality from  $f$  to  $G$

$$(\bar{x} \in \text{Crit} f \implies) \quad (\lambda^1, \dots, \lambda^p, x) \in \widehat{\text{Crit}} G \iff$$

$$\iff \begin{cases} F_i(x, t_{j_i}^i) = F_i(x, t_0^i), & \begin{cases} \text{for all } i \in \{1, \dots, p\} \\ \text{for all } j_i \in \{0, \dots, m_i\} \end{cases} \\ \text{rank} \left( \sum_{j_1=0}^{m_1} \lambda_{j_1}^1 D_x F_1(x, t_{j_1}^1), \dots, \sum_{j_p=0}^{m_p} \lambda_{j_p}^p D_x F_p(x, t_{j_p}^p) \right) < p. \end{cases}$$

• Set:

$$m = \left( \prod_{i=1}^p (m_i + 1) \right) - 1 \quad \text{and} \quad \begin{cases} \vec{i} = (i_1, i_2, \dots, i_\ell) \\ i_j \in \{0, \dots, m_j\} \end{cases}$$

Describe  $\text{inn } \Delta^m$  by multi-indices, i.e.  $\lambda = (\lambda_{\vec{i}}) \in \text{inn } \Delta^m$ .

Set:

$$\begin{cases} \phi^{\vec{i}} : \mathbb{R}^n \rightarrow \mathbb{R}^p, \quad \vec{i} = (i_1, i_2, \dots, i_p) \\ \phi^{\vec{i}}(x) = (F_1(x, t_{i_1}^1), \dots, F_p(x, t_{i_p}^p)) \end{cases}$$

and

$$\Psi(\lambda, x) := \sum_{i_1=0}^{m_1} \cdots \sum_{i_p=0}^{m_p} \lambda_{\vec{i}} \phi^{\vec{i}}(x), \quad (\lambda, x) \in \text{inn } \Delta^m \times \mathbb{R}^n.$$

- Transferring strong criticality from  $G$  to  $\Psi$

$(\lambda^1, \dots, \lambda^p, x) \in \widehat{\text{Crit}}G \implies$  for  $\lambda = (\lambda_{\vec{i}}) \in \text{inn } \Delta^m$  with

$$\lambda_{(i_1, \dots, i_p)} = \lambda_{i_1}^1 \cdots \lambda_{i_p}^p$$

it holds

$$(\lambda, x) \in \widehat{\text{Crit}}\Psi \quad \text{and} \quad \Psi(\lambda, x) = G(\lambda^1, \dots, \lambda^p, x).$$

- By the Sard Preparatory theorem:

$$\Psi(\widehat{\text{Crit}}\Psi) \text{ is null in } \mathbb{R}^p.$$

## APPLICATION: SEMI-INFINITE PROGRAMMING

$$(\mathcal{P}_r) \quad \min_{g_t(x) \leq r} u(x)$$

- Necessary optimality conditions for  $(\mathcal{P}_r)$ .

For *a.a.*  $r \in \mathbb{R}$ , and solution  $\bar{x}$  of  $(\mathcal{P}_r)$  there exist  $\lambda_1, \dots, \lambda_n \geq 0$  and  $\{t_1, \dots, t_n\} \subset T(\bar{x})$  such that

$$0 \in \partial u(\bar{x}) + \sum_{i=1}^n \lambda_i \nabla g_{t_i}(\bar{x})$$

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