#### An Overview of Generalized Convexity

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Summer School, Kaohsiung, July 2008 -p.1/43

# Outline

- Brief history of convex analysis
- From convexity to generalized convexity
- Quasiconvex functions
- Pseudoconvex functions
- Invex functions
- Extensions of GC functions
- Generalized monotone operators
- References

# **Brief history of convex analysis**

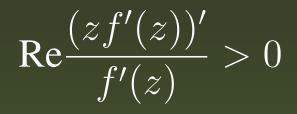
Ancient times : triangles, circles, polygons **Beginning of the 20th century (Brunn, Minkowski,** Hermite, Hadamard, Farkas, Steinitz, Jensen ...) Convex bodies Convex sets Linear inequalities Convex functions **From** the middle of the 20th century (Choquet, Klein, Milman, Hahn-Banach, Fenchel, Moreau, Rockafellar...): Convex sets, separation

Convex functions (continuity, differentiability, conjugate, subdifferential)

Convex inequalities, duality etc.

# **Brief history of convex analysis**

Attention : the same terminology in complex analysis A function f of a complex variable is called convex if



(the image of a convex domain is convex).

# convexity to generalized convexity

Convex sets :

 $A \subseteq \mathbb{R}^n$  is said to be convex  $\Leftrightarrow [a, b] \subseteq A \ \forall a, b \in A$ .

What happens if [a, b] is substituted by an arc from a family of functions?

- Two segments with a common fixed end-point  $\implies$  star-shaped sets
- Geodesic arcs  $\implies$  convex geodesic sets.
- Arcs parallel to coordinate axes => orthogonal convex sets.

# convexity to generalized convexity

**Conve**x functions :

 $f: A \subseteq \rightarrow \mathbb{R}$  is said to be convex  $\Leftrightarrow$ 

either of the following conditions holds

- Epi $f = \{(x,t) : x \in A, t \ge f(x)\}$  is convex
- $f(tx + (1 t)y) \le tf(x) + (1 t)f(y)$
- $f(y) f(x) \ge f'(x)(y x)$  when f is differentiable.

# From convexity to generalized convexity

What happens if

- Epi f is not convex, but a generalized convex set? (not so much studied except perhaps for IAR functions).
- Epi f is not convex but Lev(f) are convex ?  $\implies$  quasi-convex functions.
- f'(x)(y-x) is substituted by a function of (x, y)?  $\implies$  invex functions.
- f(tx + (1 ty)) is substituted by  $f(y) t\alpha$  for some  $\alpha > 0$  and for all t sufficiently small  $(f(y) \ge f(x))$  $\implies$  pseudo-convex functions.
- f(tx + (1 t)y) is substituted by the minimum of fover the integers whose coordinates are within unit distance from  $tx + (1 - t)y \Longrightarrow$  discretely convex functions.

# From convexity to generalized convexity

Why do we need generalized convex functions?

- Economics (analyse of microstructure, dislocation structure, demand functions, Cob-Douglas functions, equilibrium, noncooperative game ....)
- Mechanics (frictionless contact)
- Optimization (fractional programming, stochastic programming, multiobjective programming ...)
- Other areas of applied mathematics (statistics, differential equations, Hamilton-Jacobi equations, viscosity ...)

# From convexity to generalized convexity

Who have initialized the study of generalized convex functions? And recently?

- Von Neumann in 1929 with minimax theorem (according to Guerraggio and Molho)
- De Finetti in 1949.
- Arrow, Avriel, Crouzeix, Demianov, Enthoven, Ferland, Karamardian, Mangasarian, Martos, Rubinov, Schaible, Ziemba, Zang ....in the sixties and seventies of the last century.
- WGGC members and others.

A function  $f : \mathbb{R}^n \to \mathbb{R}$  is said to be quasiconvex if either of the following equivalent conditions holds (1) Sublevel sets of f are convex (2)  $f(tx + (1 - t)y) \le \max\{f(x), f(y))\}$ 

A function  $f : \mathbb{R}^n \to \mathbb{R}$  is said to be quasiconvex if either of the following equivalent conditions holds (1) Sublevel sets of f are convex (2)  $f(tx + (1 - t)y) \le \max\{f(x), f(y))\}$ Other variants

• strictly quasiconvex functions (semi-strictly quasiconvex functions)

(2) is strict when  $x \neq y$ 

• semi-strictly quasiconvex functions (explicitly quasiconvex functions)

(2) is strict when  $f(x) \neq f(y)$ 

Characterizations via derivatives (Arrow-Enthoven, 1961)

 $f \text{ is } QC \iff \nabla f \text{ is } QM$   $QM : \max\{\langle \nabla f(x), x - y \rangle, \langle \nabla f(y), y - x \rangle\} \ge 0$ Equivalent conditions  $(1) \ f(y) < f(x) \text{ implies } \langle \nabla f(x), x - y \rangle\} \ge 0$   $(2) \ f(y) \le f(x) \text{ implies } \langle \nabla f(x), x - y \rangle\} \ge 0$   $(3) \ f(y) < f(x) \text{ and } \langle \nabla f(x), x - y \rangle = 0 \text{ imply}$   $f(x) \le f(x + t(x - y)), \forall t > 0.$ 

Characterizations via directional derivatives (Komlosi, Luc ....)

$$f \text{ is QC} \iff f'(.;.) \text{ is QM}$$

Here : f'(x; u) is the directional derivative of f at x in direction u.

- f' may be
- Dini lower, upper directional derivatives for f lower semi-continuous on segments
- Clarke-Rockafellar's directional derivatives for f lower semicontinuous.

QM:  $\max\{f'(x, x - y), f'(y, y - x)\} \ge 0$ 

Characterizations via generalized derivatives (Elleia-Hassouni, Luc, Penot-Quang, Aussel-Lassonde...)

#### $f \text{ is QC} \iff \partial f(.;.) \text{ is QM}$

Here :  $\partial f$  is a subdifferential of f.  $\partial$  may be

• Clarke's subdifferential for lower semi-continuous functions;

• Pseudo-Jacobian for continuous functions. QM :  $\max\{\langle x^*, x - y \rangle, \langle y^*, y - x \rangle\} \ge 0$ for  $x^* \in \partial f(x)$  and  $y^* \in \partial f(y)$ PROOF. Mean-value theorems. Characterizations via normal cones(Aussel, Crouzeix, Danidiilis, Hadjisavvas, Lassonde ...) (f is continuous, or the space admits a Gâteaux smooth renorm)

$$f \text{ is } \mathrm{QC} \iff N_f \text{ is } \mathrm{QM}.$$

Here  $N_f$  is Clarke's normal cone to sublevel sets of fQM :  $\max\{\langle x^*, x - y \rangle, \langle y^*, y - x \rangle\} \ge 0$ for  $x^* \in \partial f(x)$  and  $y^* \in \partial f(y)$  Characterizations via normal cones(Aussel, Crouzeix, Daniliidis, Hadjisavvas, Lassonde ...) Clarke's tangent cone

$$T^{\circ}(C;x) = \liminf_{y \in C \to x; t \downarrow 0} \frac{C-y}{t}$$

i.e.  $v \in T \circ (C; x)$  iff  $\forall t_n \downarrow 0, y_n \in C \to x$  there are  $x_n \in C$  such that  $v = \lim(x_n - y_n)/t_n$ . Clarke's normal cone

$$N^{\circ}(C;x) = \{x^* : \langle x^*, v \rangle \le 0, \forall v \in T \circ (C;x)\}.$$

Quasiconvex subdifferential(Greenberg-Pierskalla, Crouzeix, Daniilidis, Hadjisavvas, Martinez-Legaz, Sach...)

$$\partial^{qcv} f(x) = \partial f(x) \cap N(f;x)$$

#### Here

N(f;x) is the convex normal cone to the sublevel set of f at x, i.e.  $x^* \in N(f;x)$  iff  $\langle x^*, y - x \rangle \leq 0, \forall y$ with  $f(y) \leq f(x)$ .  $\partial f(x)$  is any subdifferential satisfying the mean-value inequality :  $f(y) > f(x) \Longrightarrow \exists x_n \to z \in [x, y), x_n^* \in \partial f(x_n)$ such that  $\langle x_n^*, z + t(y - z) - x_n \rangle > 0 \forall t$ . Quasiconvex subdifferential(Greenberg-Pierskalla, Crouzeix, Daniilidis, Hadjisavvas, Martinez-Legaz, Sach...)

$$\partial^{qcv} f(x) = \partial f(x) \cap N(f;x)$$

Utility of quasiconvex subdifferential THEOREM : If f is lsc, radially continuous, then the following conditions are equivalent

(i) f is quasi-convex

(ii) 
$$\partial^{qcv} f = \partial f$$

(iii)  $\partial^{qcv} f$  satisfies the mean-value inequality

(iv) The domain of  $\partial^{qvc}$  is dense in the domain of f.

Quasiconvex conjugate (Martinez-Legaz, Penot, Rubinov-Dutta, Singer, Volle ...)

*c*-conjugate for  $f : \mathbb{R}^n \to \overline{\mathbb{R}}$ :

$$f^{c}(x^{*},t) = -\inf\{f(x) : \langle x, x^{*} \rangle \ge t\}$$

*c'*-conjugate for  $g : \mathbb{R}^n \times \mathbb{R} \to \overline{\mathbb{R}}$ :

$$g^{c'}(x) = -\inf\{g(x^*, t) : \langle x, x^* \rangle \ge t\}$$

Quasiconvex conjugate (Martinez-Legaz, Penot, Rubinov-Dutta, Singer, Volle ...)

Here "c" is referred to a special coupling function from an abstract scheme of level set conjugation :

$$c(x, x^*, t) = \begin{cases} 0 & \text{if } \langle x, x^* \rangle \ge t \\ -\infty & \text{otherwise} \end{cases}$$

Quasiconvex conjugate (Martinez-Legaz, Penot, Rubinov-Dutta, Singer, Volle ...)

THEOREM : f is evenly quasiconvex  $\iff f = f^{cc'}$ . In particular, every evenly quasiconvex function is the pointwise supremum of a collection of evenly quasiaffine functions (quasiconvex and quasiconcave)

Quasiconvex conjugate (Martinez-Legaz, Penot, Rubinov-Dutta, Singer, Volle ...)

f is evenly quasiconvex if its sublevel sets are intersections of open halfspaces.Lower semicontinuous quasiconvex functions and upper semicontinuous quasiconvex functions are evenly quasiconvex.

A differentiable function  $f : \mathbb{R}^n \to \mathbb{R}$  is said to be pseudoconvex if either of the following equivalent conditions holds

(1)  $f(y) < f(x) \Longrightarrow \langle \nabla f(x), y - x \rangle < 0$ (2)  $\langle \nabla f(x), y - x \rangle > 0 \Longrightarrow \langle \nabla f(y), y - x \rangle > 0$ (3)  $\langle \nabla f(x), y - x \rangle \ge 0 \Longrightarrow \langle \nabla f(y), y - x \rangle \ge 0$ 

#### Other variants

- strictly pseudoconvex functions
  - $x \neq y \text{ and } f(y) \leq f(x) \Longrightarrow \langle \nabla f(x), y x \rangle < 0$
- nondifferentiable pseudoconvex functions  $f(y) > f(x) \Longrightarrow \exists \beta(x,y) > 0, \delta(x,y) \in (0,1]$  such that

$$f(y) - f(tx + (1 - t)y) \ge t\beta(x, y) \forall t \in (0, \delta(x, y))$$

• nondifferentiable strictly pseudoconvex functions Inequality holds whenever  $f(y) \ge f(x), x \ne y$ .

**THEO**REM : *f* is a differentiable function.

- If f is pseudoconvex, then f is quasiconvex and attains a global minimum at every x with  $\nabla f(x) = 0$ .
- If f is quasiconvex and attains a local minimum at any x with  $\nabla f(x) = 0$ , then f is pseudoconvex.
- f is pseudoconvex  $\iff \nabla f$  is PM (pseudomonotone).

Here  $\nabla f$  is PM if

 $\max\{\langle \nabla f(x), x - y \rangle, \langle \nabla f(x), x - y \rangle\} > 0$ 

when both of two terms are nonzero.

Nondifferentiable pseudoconvex functions THEOREM : f is pseudoconvex  $\iff f^+(.;.)$  is PM.

#### Here

- f<sup>+</sup>(x; u) is Dini upper directional derivative of f at x in direction u;
- $f^+$  is PM if

$$\max\{f^+(x; x - y), f^+(y; y - x)\} > 0$$

provided both of the two terms are nonzero.

#### Nondifferentiable pseudoconvex functions THEOREM : f is pseudoconvex $\implies$ (1) $\partial f$ is QM (2) $f(x) < f(y) \Rightarrow \exists y^* \in \partial f(y) : \langle y^*, x - y \rangle < 0.$ f is pseudoconvex $\iff \partial f$ is PM.

#### Here

- $\partial f$  may be Clarke's subdifferential when f is locally Lipschitz and Pseudo-Jacobian when f is continuous.
- ∂f is PM if max{⟨x\*, x y⟩, ⟨y\*, y x⟩} > 0 with x\* ∈ ∂f(x), y\* ∈ ∂f(y) provided both of the two terms are nonzero.

Nondifferentiable pseudoconvex functions (lower semicontinuous and continuous on segments) REMARK :

If one modifies the definition of pseudoconvexity by using a given subdifferential :

f is pseudoconvex if  $\exists x^* \in \partial f(x) : \langle x^*, y - x \rangle \ge 0 \Longrightarrow f(z) \le f(y) \forall z \in [x, y]$ , then

f is pseudoconvex  $\iff \partial f$  is PM

where  $\partial f$  is a subdifferential satisfying a mean value inequality.

f is differentiable (Hanson, Craven) DEFINITION : f is invex if there is a vector function  $\eta$ of two variables such that

$$f(x) - f(y) \ge \langle \nabla f(y), \eta(x, y) \rangle$$

THEOREM :  $f \text{ is invex} \iff \text{every } x \text{ with } \nabla f(x) = 0 \text{ is a global}$ minimum point.

## **Extensions of GC functions**

Vector quasiconvex functions : DEFINITION :  $f : \mathbb{R}^n \to \mathbb{R}^m$  is QC if its sublevel sets are convex.

Other variants

- QC:  $f(x) \le f(y) \Longrightarrow f(tx + (1 t)y) \le f(y)$ (Cambini, Martein ...)
- Semistrictly quasiconvex functions, strictly quasiconvex functions.
- Set-valued quasiconvex maps.

## **Extensions of GC functions**

Vector quasiconvex functions : THEOREM : f is QC  $\iff \lambda \circ f$  is QC for all extreme directions  $\lambda$  of the ordering cone.

(Luc : polyhedral cone Benoist-Borwein-Popovici : convex cone)

# **Extensions of GC functions**

#### REMARK :

- Similar extensions for pseudoconvex, invex functions.
- Several applications in vector optimization (existence, optimality conditions, local-global property, topological properties of solution sets ...),

Monotone operators : DEFINITION :  $F : E \rightarrow E'$  is monotone if

$$\langle F(x), x - y \rangle + \langle F(y), y - x \rangle \le 0$$

#### **EXAMPLE** : Linear positive semi-definite operators

Monotone operators : DEFINITION :  $F : E \rightarrow E'$  is monotone if

$$\langle F(x), x - y \rangle + \langle F(y), y - x \rangle \le 0$$

#### **IMPORTANT FEATURES :**

- (Kachurovskii) The derivative of a convex function is monotone.
- $\diamond$  Equation Ax = b has a solution in a separable, reflexive B-space for every b if
  - A is monotone
  - *A* is continuous on finite dimensional subspaces
  - A is coercive:  $\langle Ax, x \rangle / \|x\| \to \infty as_{\text{Summer School, Kaohsiung, July 2008}} p.34/43$

Generalized monotonicity of the derivative of generalized convex functions

- quasimonotone/ semistrictly quasimonotone/ strictly quasimonotone operators
- pseudomonotone/ strictly pseudomonotone operators

Quasimonotone operator

 $\max\{\langle F(x), x - y\rangle, \langle F(y), y - x\rangle \ge 0$ 

 $\begin{array}{l} \diamondsuit f \text{ is } \operatorname{QC} \Longleftrightarrow \nabla f \text{ is } \operatorname{QM}/\operatorname{CQM} \\ \text{(cyclically quasimonotone :} \\ \min \{ \langle F(x_i), x_{i+1} - x_i \rangle : i = 1, ..., n; x_{n+1} = x_1 \} \leq 0 ) \end{array}$ 

Characterization of QM operators

 $\max\{\langle F(x), x - y\rangle, \langle F(y), y - x\rangle \ge 0$ 

theorem (Crouzeix-Ferland ; Luc-Schaible) COROLLARY : Second-order characterizations of quasiconvex functions (Hessian matrix of f instead of F'.) INTERESTING CASE : f is a quadratic function

(Ferland, Martos, Schaible)

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Generalized monotonicity for solving inequalities and inclusions

- proper quasimonotone/ pseudomonotone operators
- Brezis' pseudomonotone/ monotone-like operators
- KKM-operators

• properly quasimonotone :

$$x \in conv\{x_1, ..., x_k\} \Longrightarrow \exists i : \langle F(x_i), x - x_i \rangle \ge 0$$

Other name : diagonally quasiconvex function APPLICATIONS : Variational inequalities, equilibrium (Aussel, Flores-Bazan, Hadjisavvas, Luc, Tan,...)

• Brezis' pseudomonotone operator :

 $\limsup_{x_n \to x} \langle F(x_n), x_n - x \rangle \leq 0 \Longrightarrow$  $\liminf_{x_n \to x} \langle F(x_n), x_n - y \rangle \geq \langle F(x), x - y \rangle \forall y.$ Particular cases :

1) Monotone and hemicontinuous operators are B-pseudomonotone

2) Strongly continuous operators are B-pseudomonotone
3) Continuous operators in finite dimensional spaces are B-pseudomonotone.

APPLICATIONS : Variational inequalities (Aubin, Brezis, Browder, Minty, Chowdhury, Tan, Yao ....)

• KKM-relation :  $R \subseteq E \times E$  is KKM if

 $a \in conv\{a_1, ..., a_k\} \Longrightarrow \exists i : (a, a_i) \in R$ 

Particular cases :
1) Properly quasimonotone operator
2) Quasiconvex inclusions (Hai-Khanh ; Lin-Chen)
3) KKM mapping
APPLICATIONS : Variational relation problems including quasi-variational inequalities, inclusions, generalized quasi-equilibrium problems ... (Anh, Ansari, Chadli, Li, Lin, Chen, Huang, Hai, Khanh, Tan, Sach...)

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