

# Generalized perspectives and applications

Pierre Maréchal

INSTITUT DE MATHÉMATIQUES DE TOULOUSE  
FRANCE

# Outline

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- Introduction
- Generalized perspectives
- Conjugacy
- Other properties
- Applications

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# Multiplicative barrier

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$$\mathcal{K}_n(x, \mathbf{y}) := \frac{x^{1+\alpha}}{\mathbf{y}^\beta} = \frac{x^{1+\alpha}}{y_1^{\beta_1} \cdots y_n^{\beta_n}}, \quad x \geq 0, \quad \mathbf{y} > \mathbf{0}$$

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H. Imai (1988): *On the convexity of the multiplicative version of Karmarkar's potential function*. *Mathematical Programming* **40**, 29-32

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J. Borwein (1998): *A generalization of Young's Inequality*. *Mathematical inequalities and Applications* **1**, 131-136

(Computation of the bi-conjugate)

# Perspectives

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**LEMMA.** Let  $f : \mathbb{R}^n \rightarrow [-\infty, \infty]$ . Then, the function  $\check{f}$  defined by

$$\check{f}(\mathbf{x}, y) = y f \left( \frac{\mathbf{x}}{y} \right), \quad \mathbf{x} \in \mathbb{R}^n, y \in \mathbb{R}_+^*$$

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The function  $\check{f}$  is referred to as the *perspective* of  $f$ .

# An inductive argument

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$$\mathcal{K}_{n+1}(x, y_1, \dots, y_{n+1}) = \frac{x^{1+\alpha}}{y_1^{\beta_1} \cdots y_{n+1}^{\beta_{n+1}}}$$

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$$\gamma := 1 + \alpha - |\beta| \geq 1)$$

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$$y_{n+1} \frac{\left(\frac{x}{y_{n+1}}\right)^{\frac{1+\alpha}{\gamma}}}{\left(\frac{y_1}{y_{n+1}}\right)^{\frac{\beta_1}{\gamma}} \cdots \left(\frac{y_n}{y_{n+1}}\right)^{\frac{\beta_n}{\gamma}}}$$

# Remark

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Imai also proved the convexity, for  $\alpha \geq |\beta|$ , of

$$\frac{x^\alpha}{\mathbf{y}^\beta} = \frac{x^\alpha}{y_1^{\beta_1} \cdots y_n^{\beta_n}}, \quad x \geq 0, \quad \mathbf{y} \in \Sigma_n$$

where  $\Sigma_n := \{\mathbf{y} \in \mathbb{R}_+^n \mid y_1 + \cdots + y_n = 1\}$ .

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$$g(y) f \left( \frac{x}{g(y)} \right) ?$$

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# Classes of convex sets

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$\Gamma(\mathbb{R}^n)$  nonempty convex subsets of  $\mathbb{R}^n$

$\bar{\Gamma}(\mathbb{R}^n)$  nonempty **closed** convex subsets of  $\mathbb{R}^n$

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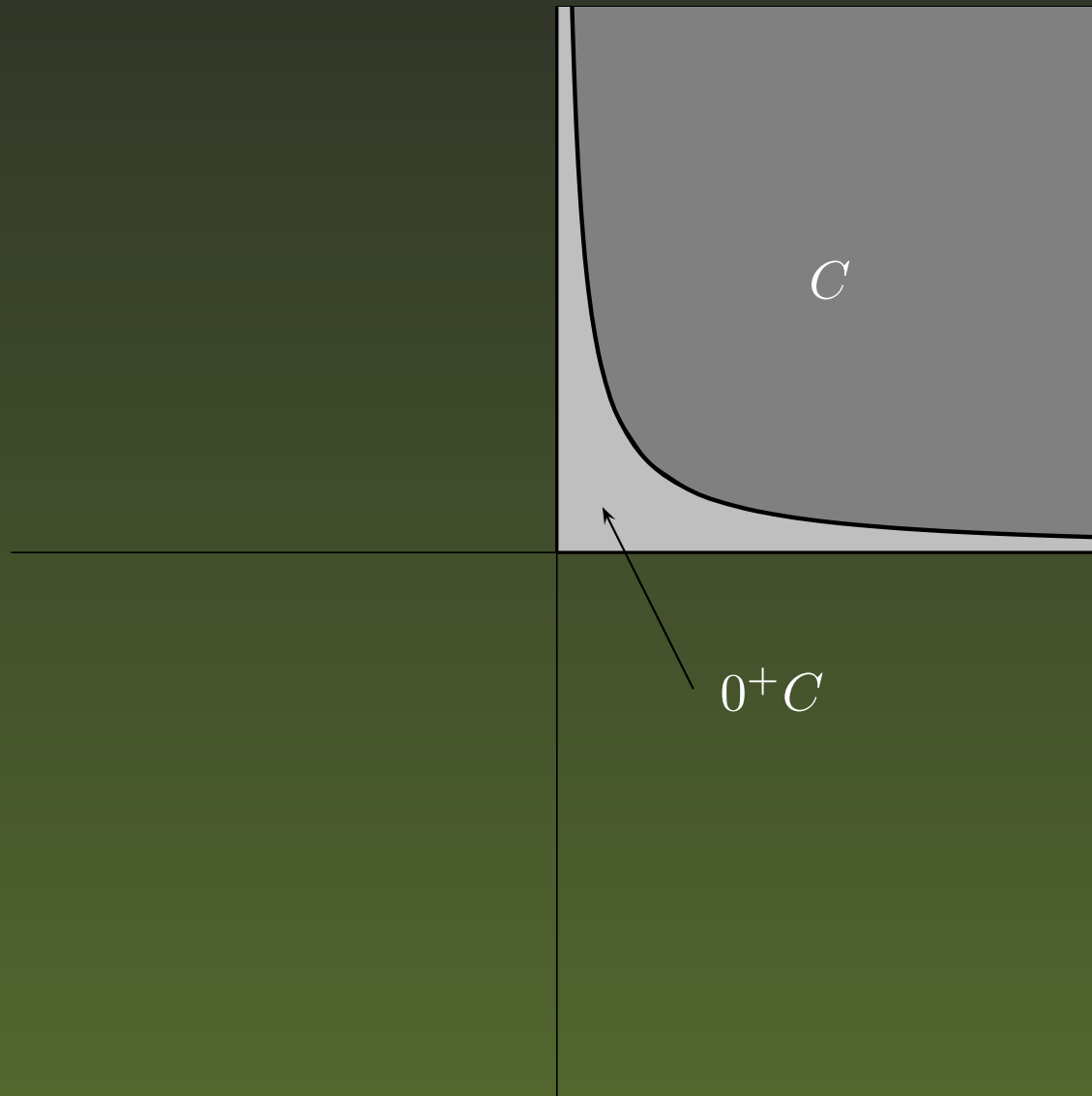
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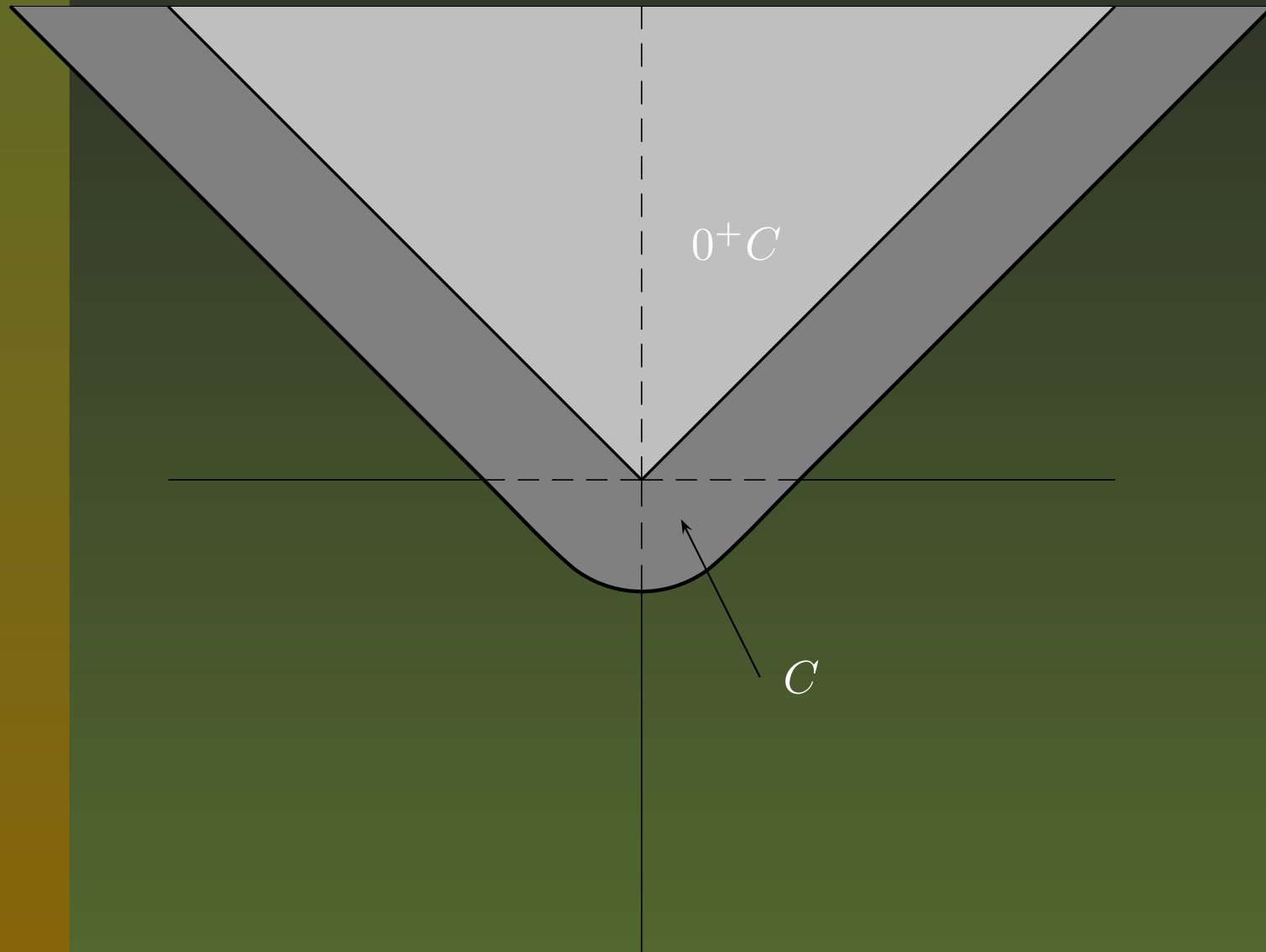
$$\Gamma_0(\mathbb{R}^n) = \{C \in \Gamma(\mathbb{R}^n) \mid 0^+C \subset C\}$$

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$$0^+C \subset C$$



# A fundamental and easy result

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$$\alpha \cdot C := \begin{cases} \alpha C & \text{if } \alpha > 0, \\ 0^+ C & \text{if } \alpha = 0 \end{cases}$$

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**PROPOSITION.** Let  $C$  be a convex subset of  $\mathbb{R}^n$ . Then

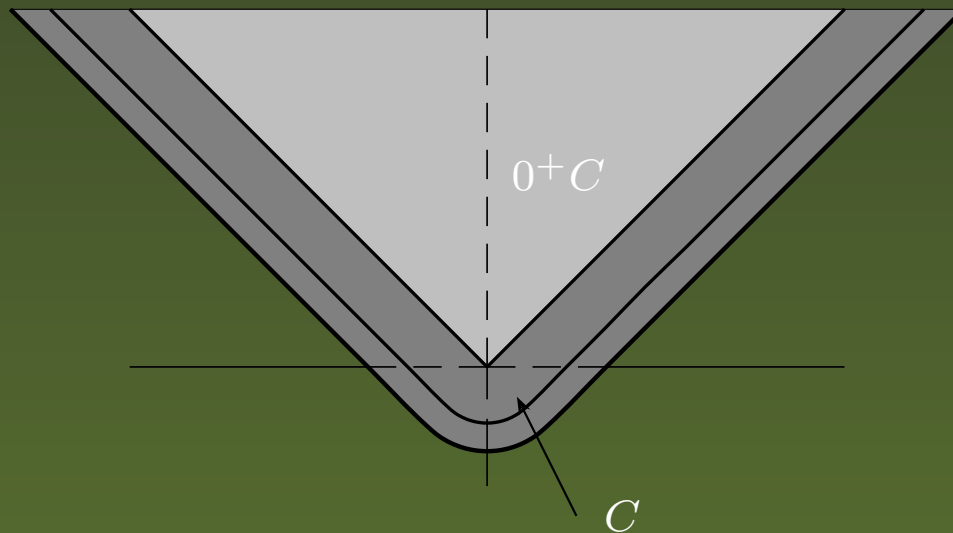
- (1) the mapping  $\alpha \mapsto \alpha \cdot C$  is increasing if and only if  $0^+ C \subset C$ ;
- (2) the mapping  $\alpha \mapsto \alpha \cdot C$  is decreasing if and only if  $C \subset 0^+ C$ .

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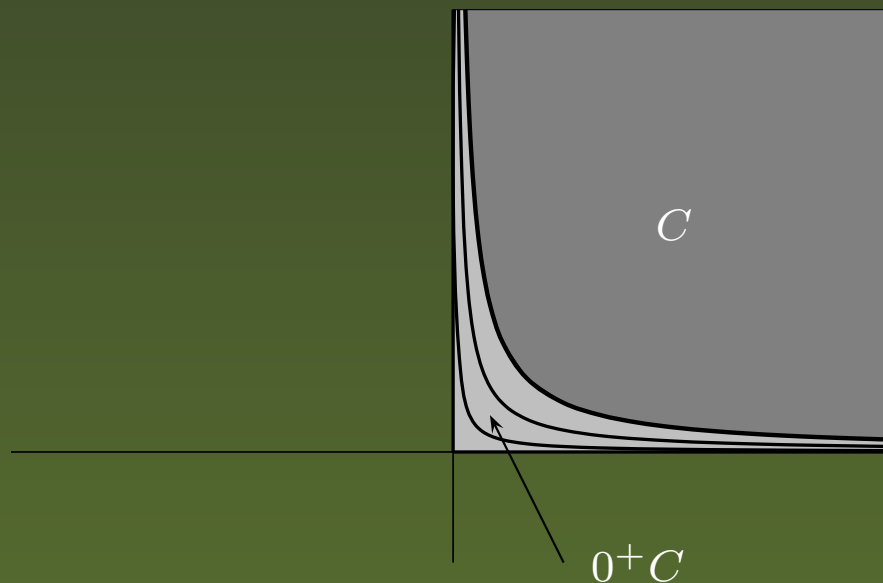


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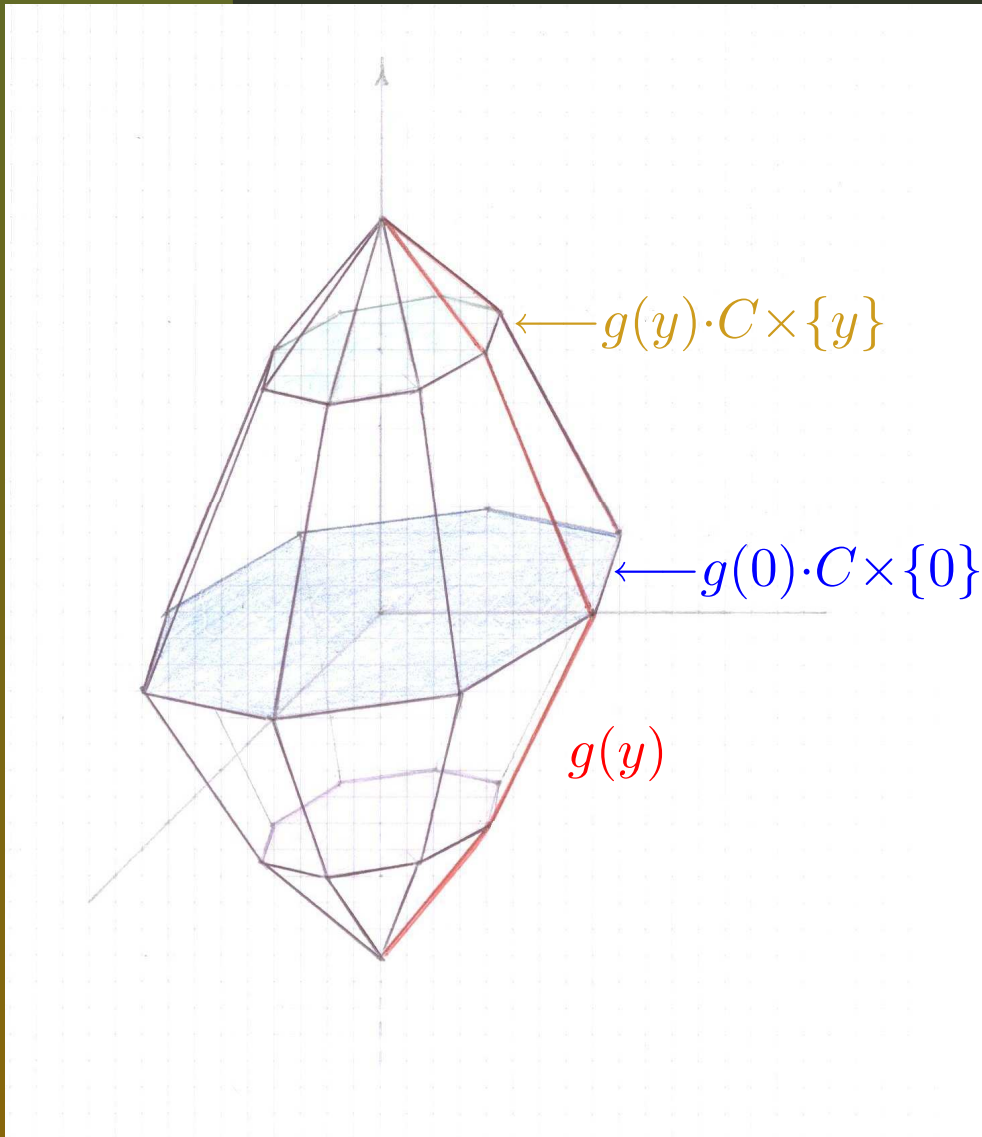
# Building convex sets on $\mathbb{R}^n \times \mathbb{R}^n$

**PROPOSITION.** Let  $C \in \Gamma_0(\mathbb{R}^n)$  and let  $g: \mathbb{R}^m \rightarrow \{-\infty\} \cup [0, \infty)$  be a proper concave function. Then, the set

$$C_g := \bigcup_{\mathbf{y} \in \text{dom } g} (g(\mathbf{y}) \cdot C \times \{\mathbf{y}\})$$

is a convex subset of  $\mathbb{R}^{n+m}$ .

# Buiding $C_g$



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# Proof of the proposition

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(2) A desirable property:

$$C, g \text{ closed} \implies C_g \text{ closed.}$$

This can fail *only* in the case where  $C = 0^+C$  and  $g$  is closed proper convex.

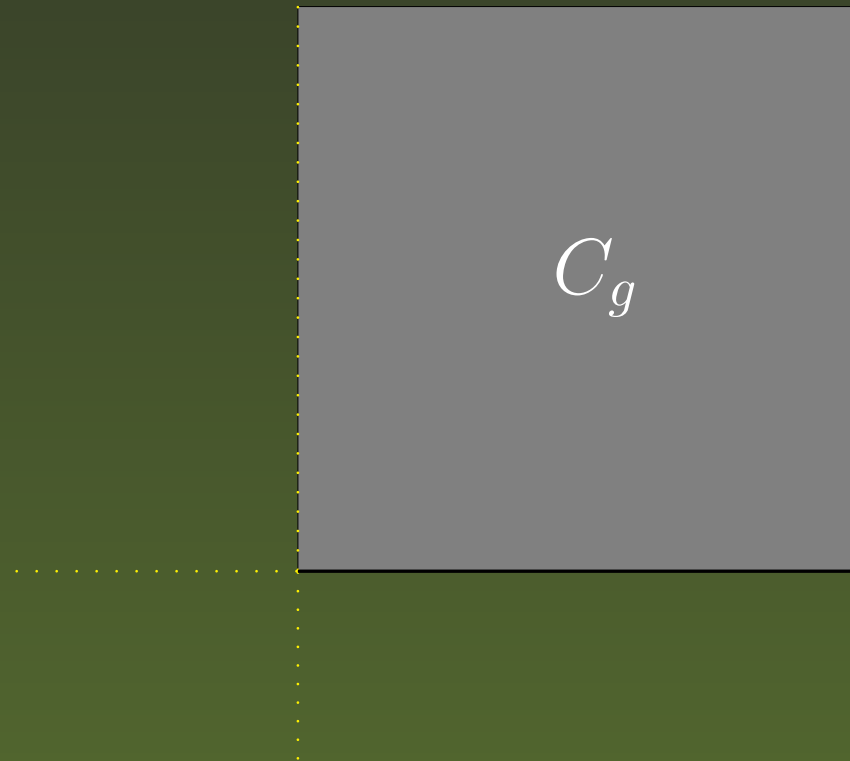
# Example

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- (1) If  $C \in \Gamma_0(\mathbb{R}^n)$  and  $g$  is **proper concave and non-negative** on its domain, we denote by  $C \triangle g$  the subset of  $\mathbb{R}^{n+m}$  defined by

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With this definition,  $C \triangle g$  is closed whenever  $C$  and  $g$  are closed.

# Characterizing $\bar{\Gamma}(\mathbb{R}^n)$

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**THEOREM [PM, 2005].** Let  $C \in \Gamma(\mathbb{R}^n)$ .

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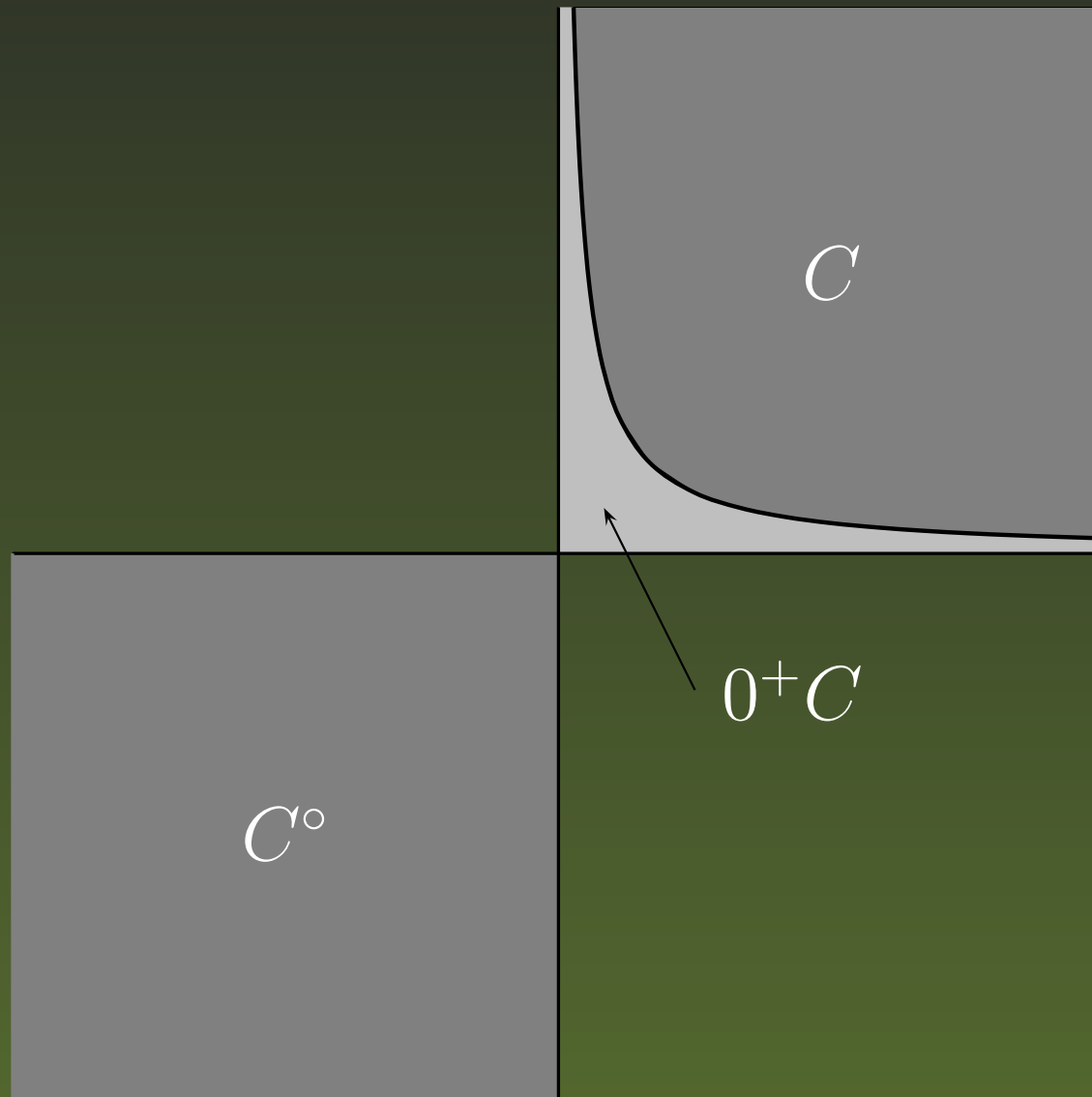
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- (2) If  $C$  is closed, then  $C \subset 0^+C$  if and only if  $C^\circ$  is a cone.

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$$0^+C \subset C \iff f(\mathbf{0}) \leq 0$$

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$$\text{epi } (f \Delta g) := (\text{epi } f) \Delta g = \bigcup_{y \in \text{dom } g} (g(y) \cdot \text{epi } f \times \{y\})$$

$$\alpha \cdot \text{epi } f = \text{epi } f \cdot \alpha := \begin{cases} \text{epi } f\alpha & \text{if } \alpha > 0 \\ \text{epi } f0^+ & \text{if } \alpha = 0 \end{cases}$$

$$(f\alpha)(\mathbf{x}) := \alpha f \begin{pmatrix} \mathbf{x} \\ - \\ \alpha \end{pmatrix}$$

# Definition 1

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- $f : \mathbb{R}^n \rightarrow (-\infty, \infty]$  proper **convex** with  $f(\mathbf{0}) \leq 0$

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$$(f \triangle g)(\mathbf{x}, \mathbf{y}) := \begin{cases} g(\mathbf{y}) f\left(\frac{\mathbf{x}}{g(\mathbf{y})}\right) & \text{if } g(\mathbf{y}) \in (0, \infty), \\ f0^+(\mathbf{x}) & \text{if } g(\mathbf{y}) = 0, \\ \infty & \text{if } g(\mathbf{y}) = -\infty. \end{cases}$$

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# Outline

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- Introduction
- Generalized perspectives
- Conjugacy
- Other properties
- Applications

# Theorem

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# Theorem

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(1) Assume that  $(f, g)$  is of type I, and that  $f$  and  $g$  are closed. Then  $((-g)^*, f^*)$  is of type II, and

$$\begin{aligned}(f \triangle g)^*(\xi, \eta) &= ((-g)^* \triangle f^*)(\eta, \xi), \\ ((-g)^* \triangle f^*)^*(\mathbf{y}, \mathbf{x}) &= (f \triangle g)(\mathbf{x}, \mathbf{y}).\end{aligned}$$

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(2) Assume that  $(f, g)$  is of type II, and that  $f$  and  $g$  are closed. Then  $(g^*, -f^*)$  is of type I, and

$$\begin{aligned}(f \triangle g)^*(\xi, \eta) &= (g^* \triangle (-f^*))(\eta, \xi), \\ (g^* \triangle (-f^*))^*(\mathbf{y}, \mathbf{x}) &= (f \triangle g)(\mathbf{x}, \mathbf{y}).\end{aligned}$$

# Left and right scalar multiplications

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$$(\alpha f)(\mathbf{x}) := \alpha f(\mathbf{x}) \quad \text{and} \quad (f\alpha)(\mathbf{x}) := \alpha f\left(\begin{array}{c} \mathbf{x} \\ - \\ \alpha \end{array}\right)$$

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$$(\alpha f)(\mathbf{x}) := \alpha f(\mathbf{x}) \quad \text{and} \quad (f\alpha)(\mathbf{x}) := \alpha f\left(\frac{\mathbf{x}}{\alpha}\right)$$

**LEMMA.** Let  $f: [-\infty, \infty]$  be any function. Then

$$(\alpha f)^*(\xi) = (f^*\alpha)(\xi),$$

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**LEMMA.** Let  $f: [-\infty, \infty]$  be any function. Then

$$(\alpha f)^*(\xi) = (f^*\alpha)(\xi),$$

$$(f\alpha)^*(\xi) = (\alpha f^*)(\xi).$$

$$\begin{aligned} (\alpha f)^*(\xi) &:= \sup\{\langle \xi, \mathbf{x} \rangle - \alpha f(\mathbf{x})\} \\ &= \sup\{\alpha(\langle \alpha^{-1}\xi, \mathbf{x} \rangle - f(\mathbf{x}))\} \\ &= \alpha \sup\{\langle \alpha^{-1}\xi, \mathbf{x} \rangle - f(\mathbf{x})\} \\ &= \alpha f^*(\alpha^{-1}\xi) \end{aligned}$$

# Sketch of a proof

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$$(f \triangle g)^*(\xi, \eta) = \sup_{\mathbf{x}, \mathbf{y}} \left\{ \langle \xi, \mathbf{x} \rangle + \langle \eta, \mathbf{y} \rangle - g(\mathbf{y}) f \left( \frac{\mathbf{x}}{g(\mathbf{y})} \right) \right\}$$

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$$\begin{aligned}(f \triangle g)^*(\xi, \eta) &= \sup_{\mathbf{x}, \mathbf{y}} \left\{ \langle \xi, \mathbf{x} \rangle + \langle \eta, \mathbf{y} \rangle - g(\mathbf{y}) f \left( \frac{\mathbf{x}}{g(\mathbf{y})} \right) \right\} \\ &= \sup_{\mathbf{y}} \left\{ \langle \eta, \mathbf{y} \rangle + \sup_{\mathbf{x}} \left\{ \langle \xi, \mathbf{x} \rangle - (fg(\mathbf{y}))(\mathbf{x}) \right\} \right\}\end{aligned}$$

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$$\begin{aligned}(f \triangle g)^*(\xi, \eta) &= \sup_{\mathbf{x}, \mathbf{y}} \left\{ \langle \xi, \mathbf{x} \rangle + \langle \eta, \mathbf{y} \rangle - g(\mathbf{y}) f \left( \frac{\mathbf{x}}{g(\mathbf{y})} \right) \right\} \\ &= \sup_{\mathbf{y}} \left\{ \langle \eta, \mathbf{y} \rangle + \sup_{\mathbf{x}} \left\{ \langle \xi, \mathbf{x} \rangle - (fg(\mathbf{y}))(\mathbf{x}) \right\} \right\} \\ &= \sup_{\mathbf{y}} \left\{ \langle \eta, \mathbf{y} \rangle + (g(\mathbf{y}) f^*)(\xi) \right\} \\ &= \sup_{\mathbf{y}} \left\{ \langle \eta, \mathbf{y} \rangle + f^*(\xi) g(\mathbf{y}) \right\} \\ &= \sup_{\mathbf{y}} \left\{ \langle \eta, \mathbf{y} \rangle - (f^*(\xi)(-g))(\mathbf{y}) \right\}\end{aligned}$$

$$[f^*(\xi) \geq 0 \text{ if } f(\mathbf{0}) \leq 0]$$

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- $f : \mathbb{R}^n \rightarrow [-\infty, \infty)$  proper **concave** with  $f(\mathbf{0}) \geq 0$

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# Theorem

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(1) Assume that  $(f, g)$  is of type  $I'$ , and that  $f$  and  $g$  are closed. Then  $(g_\star, -f_\star)$  is of type  $II'$ , and

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(2) Assume that  $(f, g)$  is of type  $II'$ , and that  $f$  and  $g$  are closed. Then  $((-g)_\star, f_\star)$  is of type  $I'$ , and

$$\begin{aligned}(f \nabla g)_\star(\xi, \eta) &= ((-g)_\star \nabla (f_\star))(\eta, \xi), \\ ((-g)_\star \nabla (f_\star))_\star(\mathbf{y}, \mathbf{x}) &= (f \nabla g)(\mathbf{x}, \mathbf{y}).\end{aligned}$$

# Outline

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- Introduction
- Generalized perspectives
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- Other properties
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# Associativity

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$(f, g)$  of type I

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$(f \triangle g)$  closed proper convex

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Additional assumption:  $\mathbf{0} \in \text{dom } g$ . Then:

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One can show that  $g \nabla h$  is actually nonnegative (on its domain), so we can also consider  $f \triangle (g \nabla h)$

Both  $(f \triangle g) \triangle h$  and  $f \triangle (g \nabla h)$  are closed proper convex

# Associative laws

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THEOREM [PM, 2005]. With the above assumptions,

$$(f \triangle g) \triangle h = f \triangle (g \nabla h).$$

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THEOREM [PM, 2005]. With the above assumptions,

$$(f \triangle g) \triangle h = f \triangle (g \nabla h).$$

$$\begin{aligned} [(f \triangle g) \triangle h]((\mathbf{x}, \mathbf{y}), \mathbf{z}) &= h(\mathbf{z})(f \triangle g) \left( \frac{\mathbf{x}}{h(\mathbf{z})}, \frac{\mathbf{y}}{h(\mathbf{z})} \right) \\ &= h(\mathbf{z})g(\mathbf{y}/h(\mathbf{z}))f \left( \frac{\mathbf{x}/h(\mathbf{z})}{g(\mathbf{y}/h(\mathbf{z}))} \right) \\ &= h(\mathbf{z})g(\mathbf{y}/h(\mathbf{z}))f \left( \frac{\mathbf{x}}{h(\mathbf{z})g(\mathbf{y}/h(\mathbf{z}))} \right) \\ &= (g \nabla h)(\mathbf{y}, \mathbf{z})f \left( \frac{\mathbf{x}}{(g \nabla h)(\mathbf{y}, \mathbf{z})} \right) \\ &= [f \triangle (g \nabla h)](\mathbf{x}, (\mathbf{y}, \mathbf{z})) \end{aligned}$$

# Commutative laws

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# Commutative laws

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## Infimal convolution

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Roughly speaking, addition and epigraphic sum are dual operations.

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Then  $(f_1 + f_2, g)$  is of type I, and

$$(f_1 + f_2) \triangle g = (f_1 \triangle g) + (f_2 \triangle g).$$

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If  $f \neq f0^+$ , or if  $f = f0^+$  and the sets  $\text{ri dom } g_k^*0^+$  have a point in common, then  $(f, g_1 \vee g_2)$  is of type II, and

$$f \triangle (g_1 \vee g_2) = (f \triangle g_1) \vee (f \triangle g_2).$$

# Outline

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- Introduction
- Generalized perspectives
- Conjugacy
- Other properties
- Applications

# Minimizing condition numbers (with Jane Ye)

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$\lambda_1(A) \geq \dots \geq \lambda_n(A)$  eigenvalues of  $A$

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Take restriction to  $\{A = B\}$

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$\kappa_p$  and  $\kappa_p^p$  have the same minimizers !

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Replace the original **quasiconvex** problem

$$(\mathcal{P}) \quad \left| \begin{array}{l} \text{Minimize } \kappa(A) \\ \text{s.t. } A \in \Omega \end{array} \right.$$

by the surrogate **convex** problem

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Question: can we approach a solution to Problem  $(\mathcal{P})$  with solutions to the surrogate problems  $(\mathcal{P}_p)$  ?

# Main results [PM & Jane Ye, 2009]

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**THEOREM 1 [Exact approximation].** Let  $(p_k)_{k \in \mathbb{N}^*} \subset [1, \infty)$  be a sequence which tends to infinity, and, for every  $k \in \mathbb{N}^*$ , let  $\bar{A}_{p_k}$  be a solution to problem  $(\mathcal{P}_{p_k})$ . Then every cluster point  $\bar{A}$  of the sequence  $(\bar{A}_{p_k})$  (there is at least one) is a global solution of problem  $(\mathcal{P})$ .

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$$\forall A \in \Omega, \quad \kappa(A) \geq \kappa(\bar{A}_k) - \varepsilon_k$$

# Back to multiplicative functions

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$$\mathcal{K}_n(x, \mathbf{y}) := \frac{x^{1+\alpha}}{\mathbf{y}^\beta} = \frac{x^{1+\alpha}}{y_1^{\beta_1} \cdots y_n^{\beta_n}}, \quad x \geq 0, \quad \mathbf{y} > \mathbf{0}$$

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**THEOREM [PM, 2010].** The function  $K_n$  is convex.

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## Some references

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Thank you for your attention !