

# Some shape optimization problems with a polygonal solution

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# Introduction

In this talk, we consider the class

$$\mathcal{K} = \{K \subset \mathbb{R}^2; K \text{ convex set}\}$$

with various other constraints.

We are interested in minimization (or maximization) problems on various subsets of the class  $\mathcal{K}$  whose solutions are **singular**, actually polygons.

# Example 1: the farthest convex set

We consider the class

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where  $P(K)$  denotes the perimeter of  $K$  and  $s(K)$  its Steiner point.

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More precisely, we want to answer the following

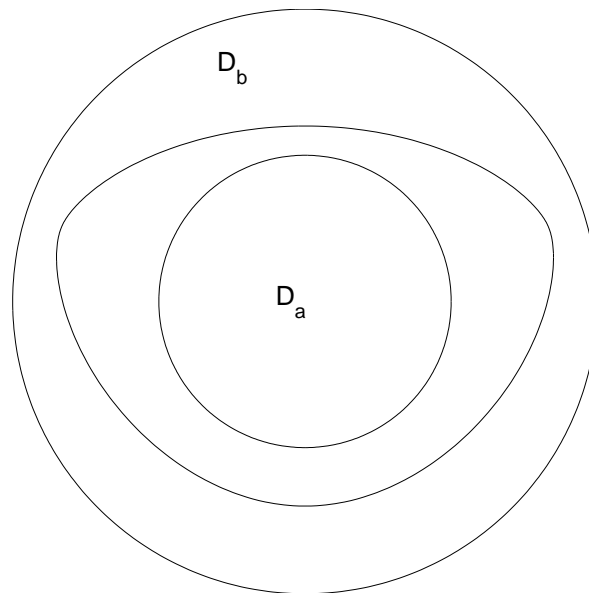
**Question 1:** Let  $C$  be given in  $\mathcal{A}$ , what is the farthest convex set  $K_C$  such that  $d(K_C, C) = \max_{K \in \mathcal{A}} d(K, C)$ .

# Example 2: a geometric problem

Let  $D_a$  and  $D_b$  be the disks, centered at  $O$  of radius  $a, b$ . Let  $\mathcal{B}$  be the class  $\mathcal{B} = \{D_a \subset K \subset D_b, K \text{ convex}\}$ . Let  $\alpha > 0$  be given. We want to minimize

$$J_\alpha(K) = \alpha|K| - P(K).$$

**Question 2:** What is the solution for any  $\alpha$ ?



# Example 3: the first eigenvalue

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The problem  $\max\{\lambda_1(\Omega), \Omega \text{ convex}, |\Omega| = V_0, P(\Omega) \geq P_0\}$  is well-posed.

**Question 3:** What is the maximizer of the first eigenvalue among convex domains with perimeter and area constraints?

# Example 4: the Mahler conjecture

Let  $K$  be a symmetric convex body and  $K^\circ$  its polar body:  
 $K^\circ = \{\xi; |x.\xi| < 1, \forall x \in K\}$ . The Mahler volume is  
 $M(K) = |K||K^\circ|$ . It is invariant by affine transformation.

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**Question 4:** What are the **minimizers** of the Mahler volume?

In 2D, it is the square (Mahler). In higher dimensions, it is a famous conjecture: it should be the cube (Difficult question: see T. Tao's blog).

# A geometric approach

**Classical fact:** If we **maximize** a **strictly convex** function over a **convex** domain  $\mathcal{A}$ , the maximum is attained at **extreme** points of  $\mathcal{A}$ .

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For domains, convexity or concavity properties are known as **Brunn-Minkowski** inequalities. For example, in the plane,  $|K|^{1/2}$  or  $\lambda_1(\Omega)^{-1/2}$  are strictly concave:

$$|(1-t)K_0 + tK_1|^{1/2} \geq (1-t)|K_0|^{1/2} + t|K_1|^{1/2}$$

with **equality** iff  $K_0, K_1$  are **homothetic**.



# Indecomposability

**Definition**  $K$  is **indecomposable** (in  $\mathcal{M}$ ) if

$$K = (1 - t)K_0 + tK_1 \text{ (with } K_0, K_1 \in \mathcal{M}\text{)}$$

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**Example** Minimizers of the logarithm capacity in  $\mathbb{R}^2$  among convex sets of given perimeter are triangles or segments.

# The support function(1)

Let  $K$  be a plane convex set.

The **support function**  $h_K$  of  $K$  is defined by:

$$h_K(\theta) := \max\{x \cdot e^{i\theta} : x \in K\}.$$

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The **Steiner point**  $s(K)$  of the convex set is defined by:

$$s(K) = \frac{1}{\pi} \int_0^{2\pi} h_K(\theta) e^{i\theta} d\theta.$$

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The **polygons** are also well characterized

$$K \text{ is a polygon} \iff h_K'' + h_K = \sum_{j=1}^n a_j \delta_{\theta_j}$$

where  $a_1, a_2, \dots, a_n$  and  $\theta_1, \theta_2, \dots, \theta_n$  denote the **lengths** of the sides and the **angles** of the corresponding outer **normals**.

# Support function and distances

The **Hausdorff distance** can be defined using the support functions:

$$d_H(K, L) = \|h_K - h_L\|_\infty.$$

We can also define a  **$L^2$  distance** (Mc Clure and Vitale) by

$$d_2(K, L) := \left( \int_0^{2\pi} |h_K - h_L|^2 d\theta \right)^{1/2}.$$

# The farthest convex set ( $L^2$ )

(Joint work with Evans Harrell)

**Theorem** Let  $J$  be a functional defined by

$$J(K) := \int_0^{2\pi} a h_K^2 + b h_K'^2 + c h_K + d h_K' d\theta$$

(example the  $L^2$  distance:  $J(K) = \int_0^{2\pi} (h_K - h_C)^2 d\theta$ ). Then every local maximizer of the functional  $J$  within the class  $\mathcal{A}$  is either a **segment** or a **triangle**.

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**Corollary** The farthest convex set for the  $L^2$  distance is either a segment or a triangle.

Actually, we can prove that it is a **segment**

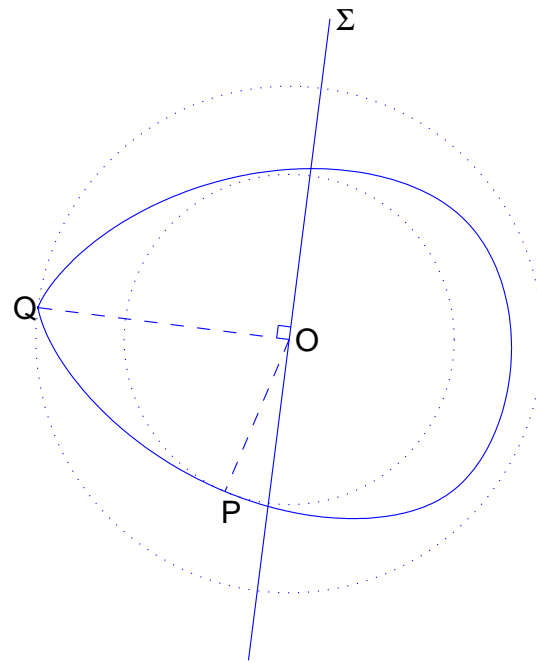
# The farthest convex set (Hausdorff)

**Theorem** [farthest convex set for Hausdorff distance]

If  $C$  is a given convex set in the class  $\mathcal{A}$ , then the convex set  $K_C$  for which

$$d_H(C, K_C) = \max\{d_H(C, K) : K \in \mathcal{A}\}$$

is a **segment**.



# Ingredient: a geometric inequality

**Theorem** Let  $K$  be any plane convex set with its Steiner point at the origin. Then

$$\max h_K \leq \frac{P(K)}{4} \leq \min h_K + \max h_K,$$

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The first inequality is due to P. Mc Mullen. It implies that the **diameter** of  $\mathcal{A}$  is less than  $\pi/2$ .

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**Question:** Does it imply that it is a triangle? If yes, it can be proved that the maximizer is an **isosceles** triangle.

# Analytic approach

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**Lemma**[T. Lachand-Robert, M. Peletier, J. Lamboley, A. Novruzzi] If  $\text{suppt}(h'' + h)$  has at least **3 points** in  $(0, \varepsilon)$ , there exists  $v$  compactly supported in  $(0, \varepsilon)$  such that  $h + tv$  is the support function of a convex set.



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**Consequence:** If  $J(K) = j(h_K)$  is strictly "locally concave" in  $h$ , the minimizers have to be **polygons**.

# A geometric example

(Work with Chiara Bianchini)

We recall that we want to minimize  $J_\alpha(K) = \alpha|K| - P(K)$  in the class  $\mathcal{B} = \{D_a \subset K \subset D_b, K \text{ convex}\}$ .

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Using the support function, we have

$$J_\alpha(K) = \alpha \int_0^{2\pi} h_K^2 - h_K'^2 - \int_0^{2\pi} h_K$$

Let  $h(= h_K)$  be a minimizer and assume that the support of  $h'' + h$  has at least 3 points in some interval  $(0, \varepsilon)$ .

According to the previous Lemma, there exists  $v$  compactly supported in  $(0, \varepsilon)$  such that  $h + tv$  is the support function of a convex set.

# A geometric example (2)

Then the second derivative  $J''_{\alpha}$  must be non negative at  $v$ :

$$\langle J''_{\alpha}(h), v, v \rangle = \int_0^{2\pi} v^2 - v'^2 \geq 0.$$

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$$0 \leq \int_0^\varepsilon v^2 - v'^2 \leq \left(\frac{\varepsilon^2}{\pi^2} - 1\right) \int_0^\varepsilon v'^2$$

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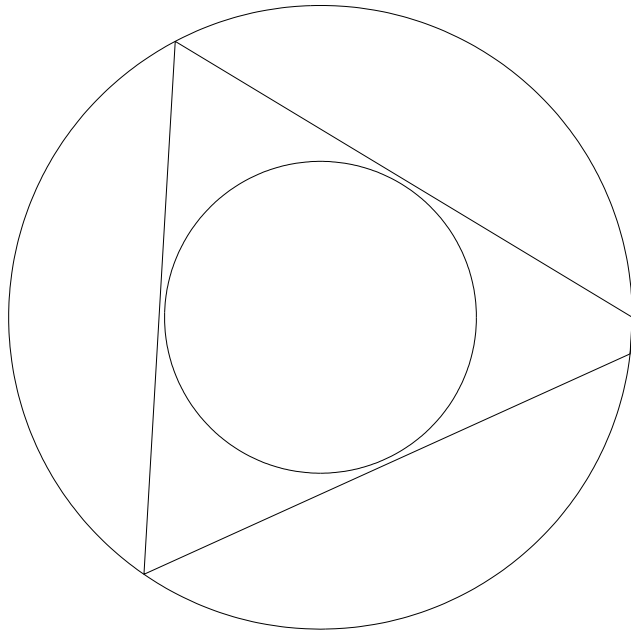
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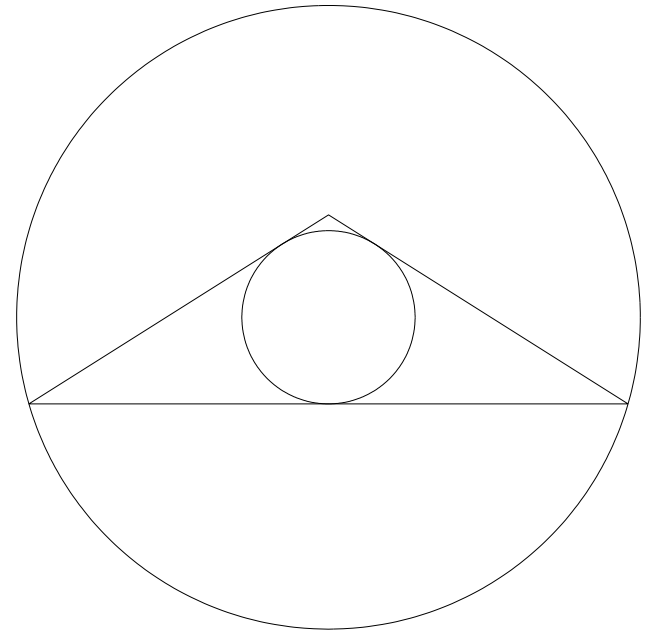
**Consequence:** Any (local) minimizer is a **polygon** inside the ring

# A geometric example (3)

What are the minimizer(s)? Of course, it depends on the parameters  $a, b, \alpha$ .



$$a = 1, b = 2, \alpha = 0.33$$



$$a = 1, b = 3, \alpha = 1.5$$

# The Mahler conjecture (1)

(work in progress with Evans Harrell and Jimmy Lamboley)

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In terms of the support function, the (2-D) Mahler volume can be expressed as

$$M(h_K) = \left( \int_0^{2\pi} h_K^2 - h'_K{}^2 \right) \left( \int_0^{2\pi} \frac{1}{2h_K^2} \right)$$

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Aim: apply this technique in higher dimensions to make progress in the general conjecture.