MAXIMIZATION OF THE STEKLOV EIGENVALUES WITH A DIAMETER CONSTRAINT

A. AL SAYED, B. BOGOSEL, A. HENROT, F. NACRY

ABSTRACT. In this paper, we address the problem of maximizing the Steklov eigenvalues with a diameter constraint. We provide an estimate of the Steklov eigenvalues for a convex domain in terms of its diameter and volume and we show the existence of an optimal convex domain. We establish that balls are never maximizers, even for the first non-trivial eigenvalue that contrasts with the case of volume or perimeter constraints. Under an additional regularity assumption, we are able to prove that the Steklov eigenvalue is multiple for the optimal domain. We illustrate our theoretical results by giving some optimal domains in the plane thanks to a numerical algorithm.

1. INTRODUCTION

Among classical questions in spectral geometry lies the problem of minimizing/maximizing under geometric constraints one (or several) eigenvalues of the Laplace operator with various boundary conditions. It has attracted much attention since the first conjecture by Lord Rayleigh stated in this famous book: *The Theory of Sound*. In particular, several important open problems have been solved these last twenty years. We refer the reader to the survey [2], the monograph [17] and the recent book [18] for a good overview on that topic.

In this paper, we deal with the eigenvalue problem for the Laplace operator with Steklov boundary conditions. For a nice survey covering many properties and questions related to these eigenvalues, we refer to [16] (see also [18, Chapter 5]). Recall that a real $\sigma \geq 0$ is a *Steklov eigenvalue* provided that there is $u \in H^1(\Omega)$ with $u \neq 0$ such that

$$
\begin{cases}
\Delta u = 0 & \text{in } \Omega, \\
\frac{\partial u}{\partial n} = \sigma u & \text{on } \partial\Omega.
\end{cases}
\tag{1}
$$

Here and below, $\frac{\partial}{\partial n}$ stands for the outward normal derivative and Ω is a smooth (say Lipschitzian) bounded and open set in $\mathbb{R}^d$. As usual, the problem (1) is considered in the weak sense, that is

$$
\int_\Omega \nabla u . \nabla v \, dx = \sigma \int_{\partial\Omega} uv \, ds \quad \forall v \in H^1(\Omega).
$$

In our framework, it is known that the so-called Steklov spectrum is nothing but a discrete sequence satisfying

$$
0 = \sigma_0(\Omega) \leq \sigma_1(\Omega) \leq \sigma_2(\Omega) \leq \ldots \nearrow +\infty.
$$

We also point out that each (Steklov) eigenvalue can be computed through the usual min-max formula:

$$
\sigma_k(\Omega) = \min_{S \in \mathcal{S}_{k+1}} \max_{v \in S \setminus \{0\}} \frac{\int_\Omega |\nabla v|^2 dx}{\int_{\partial\Omega} v^2 ds} = \min_{v \in [1, u_1, \ldots, u_{k-1}]^\perp} \frac{\int_\Omega |\nabla v|^2 dx}{\int_{\partial\Omega} v^2 ds},
\tag{2}
$$

where $\mathcal{S}_{k+1}$ denotes the set of subspaces of dimension $k+1$ of $H^1(\Omega)$.

In this work, we are interested in the maximization of $\sigma_k(\Omega)$ with a diameter constraint on the set Ω:

$$
\max_{\Omega \in \mathcal{C}, D(\Omega) = d_0} \sigma_k(\Omega),
\tag{3}
$$

for a suitable class $\mathcal{C}$ of open sets in $\mathbb{R}^d$ and where $D(\Omega)$ denotes the diameter of the open set Ω. Thanks to the positive homogeneity of the Steklov eigenvalues (i.e., $\sigma_k(t\Omega) = \sigma_k(\Omega)/t$ for

every real $t > 0$) we can replace the problem (3) by any of the following ones:

$$\max_{\Omega \in \mathcal{C}, D(\Omega) \geq d_0} \sigma_k(\Omega) \tag{4}$$

and

$$\max_{\Omega \in \mathcal{C}} D(\Omega)\sigma_k(\Omega). \tag{5}$$

More precisely, problems (3) and (4) have the same set of solutions while (3) and (5) are equivalent (that is, any solution of one problem is a solution to the other one up to a suitable dilatation).

The study of the maximization problem for Steklov eigenvalues under a diameter constraint is quite natural in view of the work [5]. Indeed, B. Bogosel, D. Bucur and A. Giacomini have established ([5, Proposition 4.3]) an isodiametric control for Steklov eigenvalues, namely the existence of a positive constant $C(d)$ (depending only on the dimension d) such that for every (smooth, bounded and connected) open set $\Omega \subset \mathbb{R}^d$

$$D(\Omega)\sigma_k(\Omega) \leq C(d)k^{\frac{2}{d}+1} \quad k = 1, 2, \ldots \tag{6}$$

It should be noted that the problem (5) merely reduces in finding the optimal upper bound in the estimate (6).

Besides the later general result, a particular attention has been devoted over the years to the case $k = 1$, i.e., to the first (non-trivial) Steklov eigenvalue $\sigma_1(\Omega)$. In 1954, R. Weinstock ([25]) proved that the disk maximizes σ_1 among simply connected plane domains of a given perimeter. In fact, for such a maximization problem, the diameter constraint is stronger than the perimeter constraint (itself stronger than the volume constraint) in the sense that if we show that the ball maximizes σ_1 with a diameter constraint, it would entail that it also maximizes σ_1 with a perimeter constraint and then implies Weinstock's result for simply connected plane domains. Very surprisingly, we establish in any dimension (see Theorem 3.1) that the ball is **never a maximizer of σ_1** under a diameter constraint.

For the sake of completeness, let us mention that F. Brock in [7] has proved that the ball in $\mathbb{R}^d$ is always a maximizer of $\sigma_1(\Omega)$ with *a volume constraint*. A. Girouard and I. Polterovich ([16]) observed that the disk is not a maximizer in the plane under a *perimeter constraint* whenever we remove the simple connectedness assumption: an annulus with a small inner radius provides a better value than the disk. Nevertheless, recently in [10], D. Bucur, V. Ferone, C. Nitsch and C. Trombetti extended Weinstock's result to convex domains in $\mathbb{R}^d$ proving that the ball maximizes σ_1 with a perimeter constraint among convex domains.

The paper is organized as follows: in Section 2, we give an estimate of $\sigma_k(\Omega)$ for a convex domain in terms of its diameter and volume and we prove existence of an optimal convex domain. Let us mention here that we do not address the question of regularity which seems to be very difficult as it is often the case for such problems. Assuming regularity of the optimal set, we recall the shape derivative of the Steklov eigenvalue and the shape derivative of the diameter which will be useful for the numerical simulation provided in Section 4 in order to perform some gradient-type algorithm. Section 3 is devoted to qualitative results. First, we show that the ball is never a maximizer for σ_k with $k = 1, 2, \ldots$. Then, we state and prove (see Theorem 3.3) that a (regular) optimal domain in the plane has necessarily a multiple eigenvalue. This is an important result which is suspected to hold for most optimization problems related to eigenvalues. To the best of our knowledge, Theorem 3.3 provides the first proof of such a multiplicity property. However, the result still remains a conjecture in other situations (see, e.g., [17, Open problem 1]). At last, we illustrate our theoretical results in Section 4 by giving some optimal domains in the plane thanks to a numerical algorithm.

2. Existence, optimality conditions

2.1. **Existence.** To prove the existence of a maximizer, we will use the classical method of calculus of variations. Compactness of any class of open sets is almost for free when we work with a diameter constraint, since, by translation invariance, we can assume that our maximizing sequence lies in a given ball and then, by [20, Theorem 2.2.25] we can extract a subsequence converging with respect to the Hausdorff metric to some open set. Now, we have to deal with two major issues:

(1) In general, the diameter is not (sequentially) continuous for the Hausdorff convergence of open sets, see e.g. [20, Figure 2.4].
(2) The continuity of Steklov eigenvalues requires additional assumptions as uniform regularity (see [4] for the use of the so-called ε-cone property) or a uniform control of norm of the trace operator (see [12]). Let us note that we can also work in a relaxed setting as in [5].

The two above remarks naturally lead us to work in the setting of convex domains, that is,

$$\mathcal{C}_d := \{\Omega \subset \mathbb{R}^d : \Omega \text{ open and convex}, D(\Omega) = d_0\},$$

where $d_0 \geq 0$ is fixed. It is well known (see, e.g., [20]) that the convexity property is preserved by the Hausdorff convergence. Now, let us assume that a sequence $(\Omega_n)_{n \geq 1}$ of open convex sets of diameter d_0 converges to a convex open set Ω which is nonempty. We are going to prove that $D(\Omega) = d_0$.

Fix any real number $\varepsilon > 0$. Choose two points $x, y \in \Omega$ such that $|x - y| > D(\Omega) - \varepsilon$. By virtue of [20, Proposition 2.2.17], we know that the points $x, y \in \Omega_n$ for $n \in \mathbb{N}$ large enough. Therefore, we see that

$$\liminf_{n \to \infty} D(\Omega_n) \geq \liminf_{n \to \infty} |x - y| \geq D(\Omega) - \varepsilon.$$

Then, letting $\varepsilon \downarrow 0$ gives the estimate

$$\liminf_{n \to \infty} D(\Omega_n) \geq D(\Omega). \tag{7}$$

Now, fix some increasing function $s : \mathbb{N} \to \mathbb{N}$ such that

$$\limsup_{n \to \infty} D(\Omega_n) = \lim_{n \to \infty} D(\Omega_{s(n)}).$$

For each integer $n \geq 1$, let us choose $x_{s(n)}, y_{s(n)} \in$ such that $|x_{s(n)} - y_{s(n)}| \geq D(\Omega_{s(n)}) - 1/s(n)$. Keeping in mind that $(\Omega_{s(n)})_{n \geq 1}$ is a sequence of convex sets contained in a fixed ball B, we can write

$$[x_{s(n)}, y_{s(n)}] \subset \Omega_{s(n)} \subset B \quad \text{for all } n \geq 1.$$

Hence, there is no loss of generality to assume that $x_{s(n)} \to x$ and $y_{s(n)} \to y$ for some $x, y \in \mathbb{R}^d$. Since the Hausdorff convergence preserves the inclusion, we must have $x, y \in \Omega$. Thus, we arrive to the inequality

$$\limsup_{n \to \infty} D(\Omega_n) = \lim_{n \to \infty} D(\Omega_{s(n)}) \leq |x - y| \leq D(\Omega). \tag{8}$$

It remains to put together (7) and (8) to get

$$\lim_{n \to \infty} D(\Omega_n) = D(\Omega).$$

We are now in position to prove the following result:

Theorem 2.1. *For any $k \geq 1$, the problem*

$$\max_{\Omega \in \mathcal{C}} \sigma_k(\Omega)$$

has a solution.

Proof. The proof will follow the same lines as in [4]. Let $(\Omega_n)_{n \geq 1}$ be a maximizing sequence (of open convex sets with diameter d_0). We recall that, by translation invariance, we can assume that the sequence $(\Omega_n)_{n \geq 1}$ lies in a given ball and then, by [20, Theorem 2.2.25] we can extract a subsequence converging with respect to the Hausdorff metric to some open set. There are two possibilities:

(1) there is a subsequence that converges to a (nonempty) open convex set Ω. Moreover, by the continuity property proved above $D(\Omega) = d_0$;
(2) the sequence $(\Omega_n)_{n \geq 1}$ converges to the empty set. This means that it shrinks to a convex body of dimension at most $d - 1$ and $|\Omega_n| \to 0$.

Due to the maximizing property of $(\Omega_n)_{n \geq 1}$, the case (2) cannot occur. In fact, we are going to prove that if $D(\Omega_n) = d_0$ and $|\Omega_n| \to 0$, then $\sigma_k(\Omega_n) \to 0$. In order to prove such a convergence result, the following proposition will be needed. It provides an estimate which can be seen a counterpart of the classical estimate of Steklov eigenvalues by Colbois, El Soufi, Girouard in terms of the isoperimetric ratio (see [13]). For convex domains, we are able to get a more precise estimate involving volume and diameter.

Proposition 2.2. *Let Ω be a convex domain of diameter $D(\Omega) = D$ in $\mathbb{R}^d$. Then, there exists an explicit constant $C = C(d, k)$ depending only on the dimension d and on k such that*

$$\sigma_k(\Omega) \leq C \frac{|\Omega|^{\frac{1}{d-1}}}{D^{\frac{2d-1}{d-1}}}.$$

Proof of the proposition. We will proceed as in the proof of [4, Proposition 4.2]. Let us denote $D := D(\Omega)$. Pick any diameter ρ of Ω. We associate to it a set Ω_0 (called *region*) which is defined as the part of Ω contained between two hyperplanes orthogonal to the diameter ρ. The width of the region is denoted by L.

Step 1. Following Part 1 of the proof of [4, Proposition 4.2], we can get through elementary geometric arguments, the following estimate

$$|\Omega_0| \geq \frac{L^d}{D^d} |\Omega|. \tag{9}$$

More precisely, the basic idea to get the latter inequality is to make a comparison with a cone: the smallest volume for a portion of a cone is near its vertex for which we get exactly this estimate.

Step 2. We also need a lower bound of the (lateral) perimeter of the region Ω_0. To that purpose, we first perform a Steiner symmetrization Ω_0^* of Ω_0 with respect to the direction of the choosen diameter. This preserves the volume and decreases the perimeter. Also, all sections of Ω_0^* orthogonal to the diameter are $(d-1)$-dimensional balls. Among these ones, pick the one of maximal radius r. Obviously, the cylinder of radius r and height L contains the region Ω_0^*, so its volume given by $\omega_{d-1} L r^{d-1}$ is greater than $|\Omega_0^*|$. Here and below, ω_k denotes the volume of the unit ball in $\mathbb{R}^k$. Using (9), this allows us to obtain a lower bound for r

$$r^{d-1} \geq \frac{L^{d-1} |\Omega|}{\omega_{d-1} D^d},$$

in particular,

$$r \geq L \left(\frac{|\Omega|}{\omega_{d-1} D^d} \right)^{1/(d-1)}. \tag{10}$$

On the other hand, note that we can always include in Ω_0^* two cones with basis balls of radius r and heights which sum up to L. The perimeter of convex sets is monotone with respect to inclusion (see, e.g., [9, Lemma 2.2.2.]) therefore the (lateral) perimeter of Ω_0^* can be bounded from below by the sum of the ones for the two cones, and a lower bound of the following form can be found:

$$P(\Omega_0^*) \geq \frac{\omega_{d-2}}{d-1} L r^{d-2}.$$

Using (10) we arrive to

$$P(\Omega_0) \geq \frac{\omega_{d-2}}{d-1} L^{d-1} \left(\frac{|\Omega|}{\omega_{d-1} D^d} \right)^{(d-2)/(d-1)}. \tag{11}$$

Step 3. Finally we obtain an upper bound for the Steklov eigenvalues by using the min-max formula (2). Assume that the diameter is in the direction of the first coordinate x_1. Let us divide the diameter D of Ω into $k+1$ equal parts and build a test function u_i, depending only on x_1 in the region defined by each one of these segments.

In a segment S_i of length $D/(k+1)$ consider $a = D/(4(k+1))$ and define the function u_i piecewise affine as follows:

- on the segment of length $2a$ whose midpoint coincides with the middle of S_i define $u_i = 1$.
- on the outer segments of length a let the function u_i goes to zero with gradient $1/a$.

Now, let us estimate the Rayleigh quotient associated to u_i:

- $\int_\Omega |\nabla u_i|^2 \leq \frac{1}{a^2} |\Omega|$.
- $\int_{\partial\Omega} u_i^2 \geq P(\{u_i = 1\})$. Since the set $\{u_i = 1\}$ is a *region* of width $2a$, we can use the estimate (11) to get

$$P(\{u_i = 1\}) \geq C_d (2a)^{d-1} \left(\frac{|\Omega|}{D^d} \right)^{\frac{d-2}{d-1}},$$

with $C_d := \omega_{d-2}/(d-1)\omega_{d-1}^{(d-2)/(d-1)}$. Therefore, we have

$$\frac{\int_\Omega |\nabla u_i|^2}{\int_{\partial\Omega} u_i^2} \leq \frac{[2(k+1)]^{d+1}|\Omega|^{\frac{1}{d-1}}}{4C_d D^{\frac{2d-1}{d-1}}}.$$

Since we can construct $k+1$ such functions with disjoint supports in Ω, we conclude that this also gives an upper bound for $\sigma_k(\Omega)$. $\qquad\square$

Now, let us come back to the proof of the existence result. We have established (thanks to the latter proposition) that the maximizing sequence $(\Omega_n)_{n\geq 1}$ converges (up to a subsequence) with respect to the Hausdorff distance to some open set Ω. Since the convexity and the diameter are preserved, Ω belongs to the class $\mathcal{C}$. Let B be a (compact) ball included in Ω. By [20, Proposition 2.2.17], B is also included into Ω_n for n large enough. By [20, Proposition 2.4.4], all the sets Ω_n and Ω satisfy the ε-cone property with the same constant ε (related to this ball B). Moreover, we also have (due to the convexity) the convergence of the involved perimeters, i.e., $P(\Omega_n) \to P(\Omega)$. Thus, by [4, Theorem 3.5], we have $\sigma_k(\Omega_n) \to \sigma_k(\Omega)$ and the existence follows. $\qquad\square$

Remark 2.3. The diameter constraint is, in some sense, more flexible that the volume or the perimeter constraint. Let us illustrate this by considering a domain Ω with holes (that is, the complement $\mathbb{R}^d \setminus \Omega$ is disconnected). Filling those holes does not modify the diameter but it would **increase** the associated Steklov eigenvalues (this can be seen through the Rayleigh quotient of any test function: the numerator will increase while the denominator will decrease). As a consequence, there is no loss of generality to state the maximization problem on the class of domains without holes (i.e., simply connected domains in the plane). Nevertheless, the existence of a maximizer is far being clear in such a class.

2.2. **Derivative of Steklov eigenvalues.** We are interested in writing optimality conditions for our maximization problem involving a diameter constraint. For that purpose, we use the classical notion of shape derivative (see, e.g., [20, Chapter 5] for more details on that concept). The theorem below gives the formulae for the shape derivative of Steklov eigenvalues. It is a particular case of a more general result which appears in the paper by Dambrine, Kateb, Lamboley [14] devoted to the so-called Wentzell operator and its eigenvalues.

Theorem 2.4. *Let Ω be a nonempty open bounded set of class C^3. The following hold for any $V \in W^{3,\infty}(\Omega, \mathbb{R}^d)$.*
(a) If $\sigma_k := \sigma_k(\Omega)$ is a simple eigenvalue of the Steklov problem, then the application $t \mapsto \sigma_k(t) := \sigma_k(\Omega_t)$ (where as usal $\Omega_t := (I + tV)(\Omega)$) is differentiable and the derivative at 0 is

$$(\sigma_k)'(0) = \int_{\partial\Omega} (|\nabla_\tau u|^2 - \left|\frac{\partial u}{\partial n}\right|^2 - \sigma_k H |u|^2)V.n,$$

where $u(\cdot)$ is the normalized (Steklov) eigenfunction associated to σ_k.
(b) Let $(u_k)_{1\leq k\leq m}$ be the family of (Steklov) eigenfunctions associated to a multiple eigenvalue σ of order $m \geq 2$. Then, there exists m functions $t \mapsto \sigma_k(t)$ defined in a neighborhood of 0 such that

(1) $\sigma_k(0) = \sigma_k$;
(2) For every t near 0, $\sigma_k(t)$ is an eigenvalue of $\Omega_t := (I + tV)(\Omega)$;
(3) The functions $t \mapsto \sigma_k(t)$ admit derivatives and their values at 0 are eigenvalues of the matrix $M = (M_{i,j})_{1\leq i,j\leq m}$ defined by

$$M_{i,j} = \int_{\partial\Omega} (\nabla_\tau u_i.\nabla_\tau u_j - \frac{\partial u_i}{\partial n}\frac{\partial u_j}{\partial n} - \sigma H u_i u_j)V.n$$

2.3. **Shape derivatives of the diameter.** The development of optimality conditions for our maximization problem also requires the shape derivative of the diameter. This is the aim of what follows.

We work here in the context of a general (real) normed space $(X, \|\cdot\|)$. Let us consider a multimapping (i.e., a set-valued mapping) $C : I \rightrightarrows X$ with bounded values defined on a real interval $I := [T_0, T]$ with $T_0 < T$. We introduce the function $\delta : I \to \mathbb{R}$ defined by

$$\delta(t) := D(C(t)) := \sup_{(x,y)\in C(t)^2} \|x - y\| \quad \text{for all } t \in I.$$

Observe first that the function $\delta(\cdot)$ has no differentiability properties in general since for a given function $f : I \to \mathbb{R}_+$, we obviously have

$$\delta(t) = D([0, f(t)]) = f(t) \quad \text{for all } t \in I.$$

This leads to require some regularity assumptions on the multimapping $C(\cdot)$. Assume that $C(\cdot)$ is γ-Lipschitz relative to the Hausdorff distance for some real $\gamma \geq 0$, i.e.,

$$C(t) \subset C(s) + \gamma |t - s| \mathbb{B} \quad \text{for all } s, t \in I,$$

where $\mathbb{B}$ stands for the closed unit ball of $(X, \|\cdot\|)$. Such an hypothesis entails for every $s, t \in I$,

$$
\begin{aligned}
\delta(t) = \sup_{x,y \in C(t)} \|x - y\| &\leq \sup_{x,y \in C(s)+\gamma|t-s|\mathbb{B}} \|x - y\| \\
&= \sup_{x,y \in C(s), b_1, b_2 \in \mathbb{B}} \|x - y + \gamma |t - s| (b_1 - b_2)\| \\
&\leq 2\gamma |t - s| + \sup_{x,y \in C(s)} \|x - y\| \\
&= 2\gamma |t - s| + \delta(s),
\end{aligned}
$$

hence the mapping $\delta(\cdot)$ is 2γ-Lipschitz continuous on I. In particular, if $\dim X < \infty$, Rademacher's theorem says that $\delta(\cdot)$ is almost everywhere differentiable on I.

Coming back to shape optimization, we are going to assume that $0 \in \operatorname{int} I$ along with

$$C(t) := \{x + tV(x) : x \in \Omega\} = (\operatorname{Id}_X + tV)(\Omega) =: \Omega_t \quad \text{for all } t \in I,$$

where Id_X denotes the identity mapping on X and where Ω is a given nonempty relatively compact subset of X and $V : \Omega \to X$ is a bounded continuous mapping. Writing

$$x + tV(x) = x + sV(x) + (t - s)V(x) \quad \text{for all } x \in \Omega, \text{ all } t, s \in I$$

we then see

$$\Omega_t \subset \Omega_s + \sup_{x \in \Omega} \|V(x)\| |t - s| \mathbb{B}.$$

According to what precedes, we know that the diameter $\delta(\cdot)$ is differentiable almost everywhere on I whenever $X = \mathbb{R}^d$. Besides the latter differentiability property, we are going to establish the existence of the one-sided limit

$$\lim_{t \downarrow 0} \frac{\delta(t) - \delta(0)}{t}.$$

This amounts to say that $D(\cdot)$ has a shape derivative in the direction V. Let us introduce the set of *diameter points* of the set Ω:

$$\mathcal{D}_\Omega := \{(x, y) \in \overline{\Omega} : \|x - y\| = D(\Omega)\}.$$

As usual, here and below, $\overline{\Omega}$ denotes the closure of Ω in X. First, note that

$$\overline{C(t)} = \{x + tV(x) : x \in \overline{\Omega}\} \quad \text{for all } t \in I.$$

Fix any $(x_0, y_0) \in \mathcal{D}_\Omega$. We obviously have for any $t \in I$,

$$
\begin{aligned}
\frac{1}{2}\left(\delta^2(t) - \delta^2(0)\right) &\geq \frac{1}{2} \|x_0 + tV(x_0) - (y_0 + tV(y_0))\|^2 - \frac{1}{2} \|x_0 - y_0\|^2 \\
&= t \langle x_0 - y_0, V(x_0) - V(y_0) \rangle + \frac{t^2}{2} \|V(x_0) - V(y_0)\|^2,
\end{aligned}
$$

hence

$$\liminf_{t \downarrow 0} \frac{\delta^2(t) - \delta^2(0)}{2t} \geq \langle x_0 - y_0, V(x_0) - V(y_0) \rangle.$$

Since (x_0, y_0) has been arbitrarily chosen in the set $\mathcal{D}_\Omega$, we get

$$\liminf_{t \downarrow 0} \frac{\delta^2(t) - \delta^2(0)}{2t} \geq \sup_{(x,y) \in \mathcal{D}_\Omega} \langle x - y, V(x) - V(y) \rangle.$$

Now, let $(t_n)_{n \geq 1}$ be a sequence of positive real numbers such that $t_n \to 0$ and

$$\limsup_{t \downarrow 0} \frac{\delta^2(t) - \delta^2(0)}{2t} = \lim_{n \to \infty} \frac{\delta^2(t_n) - \delta^2(0)}{2t_n}.$$

For each integer $n \geq 1$, pick any $x_n, y_n \in C(t_n)$ such that

$$\delta^2(t_n) - t_n^2 < \|x_n - y_n\|^2 \leq \delta^2(t_n).$$

According to the definition of $C(\cdot)$, we may write for every integer $n \geq 1$, $x_n = u_n + t_n V(u_n)$ and $y_n = v_n + t_n V(v_n)$ for some $u_n, v_n \in \Omega$. From the compactness of $\overline{\overline{\Omega}}$, we may suppose without loss of generality that $u_n \to u$ and $v_n \to v$ for some $u, v \in \overline{\Omega}$. It is straightforward to check that

$$\delta^2(t_n) - \delta^2(0) < \|x_n - y_n\|^2 + t_n^2 - \|u_n - v_n\|^2$$
$$< 2t_n \langle u_n - v_n, V(u_n) - V(v_n) \rangle + t_n^2 \|V(u_n) - V(v_n)\|^2 + t_n^2,$$

in particular

$$\limsup_{t \downarrow 0} \frac{\delta^2(t) - \delta^2(0)}{2t} \leq \langle u - v, V(u) - V(v) \rangle.$$

We claim that $(u, v) \in \mathcal{D}_\Omega$. Indeed, we have

$$\delta^2(t_n) \geq \left\| x - y + t_n (V(x) - V(y)) \right\|^2 \quad \text{for all } (x, y) \in \mathcal{D}_\Omega, \text{ all } n \geq 1,$$

which gives the inequality

$$\liminf_{n \to \infty} \delta^2(t_n) \geq \delta^2(0) = D(\Omega)^2$$

and

$$\delta^2(0) = \liminf_{n \to \infty} (\delta^2(t_n) - t_n^2) \leq \liminf_{n \to \infty} \|x_n - y_n\|^2 = \lim_{n \to \infty} \|u_n - v_n + t_n(V(u_n) - V(v_n))\|^2 = \|u - v\|^2.$$

Putting what precedes together, we arrive to

$$\sup_{(x,y) \in \mathcal{D}_\Omega} \langle x - y, V(x) - V(y) \rangle \leq \liminf_{t \downarrow 0} \frac{\delta^2(t) - \delta^2(0)}{2t} \leq \limsup_{t \downarrow 0} \frac{\delta^2(t) - \delta^2(0)}{2t} \leq \langle u - v, V(u) - V(v) \rangle$$

Consequently, the function $\frac{1}{2} \delta^2(\cdot)$ has a right derivative at 0 given by

$$\lim_{t \downarrow 0} \frac{\delta^2(t) - \delta^2(0)}{2t} = \sup_{(x,y) \in \mathcal{D}_\Omega} \langle x - y, V(x) - V(y) \rangle.$$

We summarize those features in the following proposition.

Proposition 2.5. *Let Ω be a nonempty open relatively compact subset of a real normed space $(X, \|\cdot\|)$ and let $V : \Omega \to X$ be a bounded and continuous mapping. Then, one has*

$$\lim_{t \downarrow 0} \frac{D(\Omega_t) - D(\Omega)}{t} = \frac{1}{D(\Omega)} \sup_{(x,y) \in \mathcal{D}_\Omega} \langle x - y, V(x) - V(y) \rangle, \tag{12}$$

where $\mathcal{D}_\Omega := \left\{ (x, y) \in \overline{\Omega} : \|x - y\| = D(\Omega) \right\}$ and $\Omega_t := \{ x + tV(x) : x \in \Omega \}$ for every $t > 0$.

3. Qualitative properties

3.1. **Case of the ball.** As recalled in the introduction, the ball maximizes the quantity $\sigma_1(\Omega)$ with a volume constraint (see [7]). It also maximizes $\sigma_1(\Omega)$ with a perimeter constraint among planar simply connected domains ([25]) and in any dimension among convex domains ([10]). Therefore, it is quite natural to expect that the ball is also a maximizer (or at least a local maximizer) for σ_1 in the setting of a diameter constraint. The following theorem shows that it is not the case!

Theorem 3.1. *The ball is not a local maximizer for Problem (5) for σ_1 in any dimension.*

Proof. The idea of the proof is simply to find a perturbation of the ball which increases the product $D(B)\sigma_1(B)$. Without loss of generality, we work with the unit ball and we use the usual spherical coordinates, that is,

$$\begin{cases} x_1 = \cos \varphi_1 \\ x_2 = \sin \varphi_1 \cos \varphi_2 \\ \vdots \\ x_{d-1} = \sin \varphi_1 \sin \varphi_2 \ldots \sin \varphi_{d-2} \cos \varphi_{d-1} \\ x_d = \sin \varphi_1 \sin \varphi_2 \ldots \sin \varphi_{d-2} \sin \varphi_{d-1} \end{cases}$$

where $\varphi_j \in [0, \pi]$ for $j \leq d - 2$ while $\varphi_{d-1} \in [0, 2\pi]$. At last, the area element is given by

$$ds = \sin^{d-2}(\varphi_1) \sin^{d-1}(\varphi_2) \ldots \sin(\varphi_{d-2}) d\varphi_1 d\varphi_2 \ldots d\varphi_{d-1}.$$

Let us consider a perturbation driven by the vector field defined in a neighbourhood of the unit sphere by

$$V(X) = \big(a_2 \cos(2\varphi_{d-1}) + a_4 \cos(4\varphi_{d-1})\big) X,$$

for some positive coefficients a_2, a_4 which will be chosen later. This means that, for every $\varepsilon > 0$ small enough, we consider some perturbations of the unit ball B defined by $B_\varepsilon := \{X + \varepsilon V(X), X \in B\}$. We have recalled in Theorem 2.4 (see also [14, Corollary 3.8]) that the first Steklov eigenvalue $\sigma_1(B_\varepsilon)$ has a directional derivative (even if this eigenvalue is multiple). Such directional derivatives are given by the eigenvalues of the $d \times d$ matrix $\mathcal{M}$ whose entries are (as usual ω_d denotes the volume of the unit ball while δ_{jk} is the Kronecker symbol)

$$\mathcal{M}_{j,k} = \frac{\delta_{jk}}{\omega_d} \int_{\partial B} V.n ds - \frac{d+1}{\omega_d} \int_{\partial B} x_j x_k V.n ds.$$

In our case, since $n = X$ on the unit sphere, we have $V.n = a_2 \cos(2\varphi_{d-1}) + a_4 \cos(4\varphi_{d-1})$. It is then not difficult to check that all the coefficients of the above matrix are zero except $\mathcal{M}_{d-1,d-1}$ and $\mathcal{M}_{d,d}$ which are respectively given by

$$\mathcal{M}_{d-1,d-1} = -\frac{d+1}{\omega_d} \int_{\partial B} \prod_{j=1}^{d-2} \sin^2(\varphi_j) \cos^2(\varphi_{d-1}) \big(a_2 \cos(2\varphi_{d-1}) + a_4 \cos(4\varphi_{d-1})\big) ds$$

$$\mathcal{M}_{d,d} = -\frac{d+1}{\omega_d} \int_{\partial B} \prod_{j=1}^{d-2} \sin^2(\varphi_j) \sin^2(\varphi_{d-1}) \big(a_2 \cos(2\varphi_{d-1}) + a_4 \cos(4\varphi_{d-1})\big) ds$$

Let us denote by $j_p := \int_0^\pi \sin^p t \, dt$ (twice the classical Wallis' integrals). The previous formulae can be rewritten as

$$\mathcal{M}_{d-1,d-1} = -\frac{d+1}{\omega_d} \prod_{p=3}^{d} j_p a_2 \frac{\pi}{2} \quad \text{and} \quad \mathcal{M}_{d,d} = \frac{d+1}{\omega_d} \prod_{p=3}^{d} j_p a_2 \frac{\pi}{2}.$$

Therefore, the eigenvalues of $\mathcal{M}$ are 0 of order $d-2$, $-Ka_2$ and Ka_2 where K is the positive constant (which is explicitly computable) $K := \frac{(d-1)\pi}{2\omega_d} \prod_{p=3}^{d} j_p$. In other words, the smallest eigenvalue $\sigma_1(B_\varepsilon)$ has the following expansion (keep in mind that $a_2 > 0$)

$$\sigma_1(B_\varepsilon) = 1 - \varepsilon K a_2 + o(\varepsilon).$$

Now, let us introduce the two antipodal points $N = (0, 0, \ldots, 1)$ and $S = (0, 0, \ldots, -1)$. Through the deformation they are sent to

$$N_\varepsilon = (0, 0, \ldots, 1 + \varepsilon(a_2 + a_4)) \quad \text{and} \quad S_\varepsilon = (0, 0, \ldots, -1 - \varepsilon(a_2 + a_4)).$$

Thus, the diameter of B_ε is greater than $N_\varepsilon S_\varepsilon = 2 + 2\varepsilon(a_2 + a_4)$ and then

$$D(B_\varepsilon)\sigma_1(\varepsilon) \geq 2\big(1 + \varepsilon(a_2 + a_4)\big)\big(1 - \varepsilon K a_2 + o(\varepsilon)\big) = 2\big(1 + \varepsilon(a_4 - (K-1)a_2) + o(\varepsilon)\big).$$

It remains to choose the coefficient a_4 such that $a_4 > (K-1)a_2$ to get the claim. $\square$

Let us point out here that the main idea of the latter proof is quite elementary. Indeed, we construct a perturbation with two trigonometric terms, the first one seen by the eigenvalue and the other one only seen by the diameter. Then, a suitable combination allows us to get a positive first derivative. In a same way, we are able to extend Theorem 3.1 to any Steklov eigenvalue. The details of the proof are left to the reader.

Theorem 3.2. *The ball is not a maximizer for Problem* (5) *for any σ_k and any dimension.*

3.2. **Multiplicity.** As explained in the introduction, it is suspected that most of optimization problems for eigenvalues have solution with multiplicity. For example, if we denote by Ω_k^* a domain which minimizes the k-th eigenvalue of the Laplacian with Dirichlet boundary conditions (see [8] and [21]) it is still an open problem (see, e.g., [17]) to prove that $\lambda_{k-1}(\Omega_k^*) = \lambda_k(\Omega_k^*)$ for any $k \geq 3$ (what we know so far is the case $k = 2$).

In our context, we are able to prove (by contradiction) such a result for a smooth optimal domain in the plane. We establish it in both situations: without any constraint or with a convexity constraint.

Theorem 3.3. *Let $\Omega^\star$ be a smooth (C^3) optimal domain in the plane with or without convexity constraints. Then, the k-th Steklov eigenvalue $\sigma_k(\Omega^\star)$ is multiple.*

Proof. Since there is no ambiguity here, we will denote by σ the Steklov eigenvalue and u a normalized associated Steklov eigenfunction. Let us start with the unconstrained case. First, note that the smoothness assumption on the optimal domain implies two things:

- we can use the shape derivative formulae stated in Theorem 2.4;
- by elliptic regularity, the eigenfunctions are at least C^2 up to the boundary.

We write the usual Rellich formula valid for any smooth function v (see e.g., [22, 23]):

$$2\int_{\partial\Omega^*}(x.\nabla v)\frac{\partial v}{\partial n} - \int_{\partial\Omega^*}(x.n)|\nabla v|^2 = 2\int_{\Omega^*}(x.\nabla v)\Delta v + (2-d)\int_{\Omega^*}|\nabla v|^2.$$

For our eigenfunction u in dimension 2, this yields

$$2\int_{\partial\Omega^*}(x.\nabla u)\frac{\partial u}{\partial n} - \int_{\partial\Omega^*}(x.n)|\nabla u|^2 = 0.$$

Decomposing the gradient in its tangential and normal component, that is, $\nabla u = u_\tau \tau + u_n n$ and using $u_n = \sigma u$ yields

$$2\sigma\int_{\partial\Omega^*}uu_\tau X.\tau + \sigma^2\int_{\partial\Omega^*}u^2 X.n - \int_{\partial\Omega^*}u_\tau^2 X.n = 0.$$

Now, assume by contradiction that the eigenvalue is simple. For any deformation field V the shape derivative of $D\sigma_K$ is nonpositive. Let us denote by $\mathcal{D}$ the set of diameter points (where the diameter is achieved, this is the projection of the set $\mathcal{D}_{\Omega^*}$ introduced in Section 2.3). Let $x \notin \mathcal{D}$ a point which does not belong to a diameter. A small perturbation V locally supported near x does not change the diameter and since both V and $-V$ are admissible, we infer $d\sigma_k(\Omega^*, V) = 0$ for all such V which implies, according to Theorem 2.4:

$$u_\tau^2 - \sigma^2 u^2 - H\sigma u^2 = 0 \quad \text{at any } x \notin \mathcal{D}. \tag{13}$$

We are going to distinguish two cases:

- **Case 1.** The set $\mathcal{D}$ is discrete. In that case, the relation (13) holds everywhere on the boundary by continuity of the function and its derivatives up to the boundary.
- **Case 2.** Now, let us assume that the set $\mathcal{D}$ (where the diameter is achieved) is not discrete. Let us consider a point $x \in \mathcal{D}$. If we perform a local perturbation driven by V near x, we can consider two situations: either $V.n$ is inward and then the diameter does not change and we recover $d\sigma_k(\Omega^*, V) \leq 0$ at such a point or $V.n$ is outward and the derivative of the diameter is positive, according to formulae (12) which implies

$$D(\Omega^*)d\sigma_k(\Omega^*, V) \leq d(D\sigma_k)(\Omega^*, V) = \sigma_k(\Omega^*)dD(\Omega^*, V) + D(\Omega^*)d\sigma_k(\Omega^*, V) \leq 0$$

  and then in any case, we have $d\sigma_k(\Omega^*, V) \leq 0$. Since this property holds for any V, we still infer $d\sigma_k(\Omega^*, V) = 0$ and the equality (13) holds true on the whole boundary.

Therefore, in both cases, we have

$$u_\tau^2 - \sigma^2 u^2 = H\sigma u^2 \text{ on } \partial\Omega^*.$$

Thus,

$$2\sigma\int_{\partial\Omega^*}uu_\tau X.\tau = \int_{\partial\Omega^*}u_\tau^2 - \sigma^2 u^2 X.nds = \sigma\int_{\partial\Omega^*}Hu^2 X.n\,ds. \tag{14}$$

Now let us compute the left-hand side of (14) integrating by parts. Since $X.\tau = xx' + yy'$ and $2uu_\tau = \frac{d}{ds}u^2$ we have

$$2\sigma\int_{\partial\Omega^*}uu_\tau X.\tau = -\sigma\int_{\partial\Omega^*}u^2[xx" + yy" + x'^2 + y'^2]ds. \tag{15}$$

Now since $\frac{d}{ds}\tau = -Hn$ and $x'^2 + y'^2 = 1$, (15) provides

$$2\sigma\int_{\partial\Omega^*}uu_\tau X.\tau = \sigma\int_{\partial\Omega^*}Hu^2 X.nds - \sigma\int_{\partial\Omega^*}u^2 ds$$

which, together with (14) would give

$$\sigma\int_{\partial\Omega^*}u^2 ds = 0$$

a contradiction since σ cannot be zero for the maximizer. This finishes the proof of the unconstrained case.

Now let us consider the case with a convexity constraint. On strictly convex parts, we can perform any deformation and then the identity $u_\tau^2 - \sigma^2 u^2 = H\sigma u^2$ still holds true on strictly convex parts of the boundary. The flat parts (or segments) require more attention. Let us consider such a segment, say $\Sigma \subset \partial\Omega^*$. In the spirit of [19] (see also [17, Theorem 4.2.2]) we can prove the following:

Lemma 3.4. *Let Ω^* be a smooth maximizer of $D(\Omega)\sigma_k(\Omega)$ among convex sets and let Σ be a segment of extremities A and B included in the boundary of Ω^*. Let $t \in [a, b]$, a parametrization of the segment (the boundary is assumed to be oriented in the clockwise sense). Then, there exists a nonnegative-valued function w defined on $[a, b]$ with triple roots at a and b, such that*

$$u_\tau^2 - \sigma^2 u^2 = w''(t). \tag{16}$$

Proof of the Lemma: The diameter constraint cannot be achieved at any point of the segment (but possibly its extremities). Thus, we just need to look at the derivative of the Steklov eigenvalue which is given by Theorem 2.4:

$$d\sigma_k(\Omega^*, V) = \int_{\partial\Omega^*} [u_\tau^2 - \sigma^2 u^2 - H\sigma u^2] V.n \, ds. \tag{17}$$

Recall that the the curvature $H = 0$ on Σ. In formula (17), the only perturbations V which are allowed are such that the deformed domain $(Id + \tau V)(\Omega^*)$ is still convex (for small τ). This holds true if and only if $t \mapsto V.n(t)$ is a concave function on $[a, b]$. Let us denote by $v = V.n$ such a concave function. Replacing in (17) and using the relation $u_\tau^2 - \sigma^2 u^2 - H\sigma u^2 = 0$ which holds on the strictly convex parts as explained above, yields on the segment Σ:

$$\int_a^b [u_\tau^2 - \sigma^2 u^2] v \, dt \leq 0. \tag{18}$$

Setting $w_2(t) := u_\tau^2 - \sigma^2 u^2$ this can also be rewritten

$$\int_a^b w_2(t) v(t) \, dt \leq 0. \tag{19}$$

The latter estimate (19) must be true for every (regular) concave function v. In particular, in the case $v(t) = 1$ and $v(t) = t$, both functions v and $-v$ are concave, therefore

$$\int_a^b w_2(t) \, dt = 0 \qquad \int_a^b t w_2(t) \, dt = 0. \tag{20}$$

Now, let us introduce the functions defined by

$$w_1(t) = \int_a^t w_2(s) \, ds \quad \text{and} \quad w(t) = \int_a^t w_1(s) \, ds = \int_a^t (t - s) w_2(s) \, ds.$$

According to (20), we have $w_1(a) = w_1(b) = w(a) = w(b) = 0$. Integrating twice by parts, it comes

$$\int_a^b w_2(t) v(t) \, dt = \int_a^b w(t) v''(t) \, dt.$$

This last integral must be nonpositive (according to (19)) for every function v concave, i.e., for every smooth function v such that $v'' \leq 0$. This guarantees that $w \geq 0$. At last a and b are triple roots of w because $w''(a) = w_2(a) = 0$ by continuity of the gradient. This finishes the proof of the Lemma. $\qquad\square$

Let us come back to the proof of Theorem 3.3. We have already seen that $u_\tau^2 - \sigma^2 u^2 = H\sigma u^2$ on the strictly convex parts of $\partial\Omega^*$. Now, let us consider a segment Σ. On such a segment $X.n$ is constant (equal to the distance, say δ, of the origin to the line supporting the segment). Therefore, according to (16), we have

$$\int_\Sigma [u_\tau^2 - \sigma^2 u^2] X.n \, dt = \delta \int_a^b w''(t) \, dt = w'(b) - w'(a) = 0 = \sigma \int_\Sigma H\sigma u^2 X.n \, ds.$$

Therefore, the relation

$$\int_{\partial\Omega^*} u_\tau^2 - \sigma^2 u^2 X.n \, ds = \sigma \int_{\partial\Omega^*} H u^2 X.n \, ds$$

holds true on the whole boundary and we can conclude as in the case without convexity constraint. $\qquad\square$

Remark 3.5. The numerical simulations of the next section suggest that the optimal domain is not exactly C^3 regular. It seems that its boundary has two singular points where the diameter is achieved. Nevertheless, it is straightforward to check that Theorem 3.3 remains true if we replace the C^3 regularity assumption by the following weaker assumptions that could be true for our optimal domains:

- the boundary of the optimal domain is C^3 except at a finite number of points;
- the curvature H is bounded;
- the eigenfunction u belongs to $C^1(\overline{\Omega})$.

4. Numerical simulations

In Section 3, we showed among other things that the disk is never a local maximizer of $\sigma_k(\Omega)$ under a diameter constraint. This leads us to provide some numerical computations in order to find some approximations of these maximizers in the plane. We point out that the diameter constraint is difficult to handle in a numerical point of view: this comes from the fact that on regions where this constraint is saturated not all arbitrarily small perturbations are admissible.

A good tool for investigating the diameter constraint in the convex setting is the support function. This is why, in a first stage we consider the maximization problem in the class of convex sets. The support function of a set $\Omega \subset \mathbb{R}^2$ is defined for each $\theta \in [0, 2\pi]$ by

$$p(\theta) = \max_{x \in \Omega} x \cdot (\cos\theta, \sin\theta),$$

where the dot $\cdot$ denotes as usual the Euclidean scalar product. An intuitive interpretation of $p(\theta)$ is the distance from the origin to the tangent orthogonal to θ (see Figure 1 for an illustration). With this geometric meaning of the support function in mind, it is obvious that the diameter or the width of Ω in the direction θ is given by $p(\theta) + p(\theta + \pi)$.
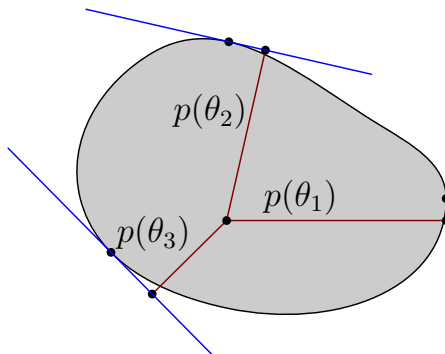

Figure 1. Geometric interpretation of the support function.

The support function has been successfully used in the approximation of optimal shapes under convexity constraint (see, e.g., [3, 6, 1]). In the paper [3], the authors investigate various functionals related to the volume and perimeter. The work [6] is devoted to the case of Dirichlet-Laplace eigenvalues and in [1] the method is extended to the dimension three and multiple case tests are provided. If p is the support function of a strictly convex domain Ω then, as recalled in these works, a parametrization of $\partial\Omega$ is given by

$$\begin{cases} x(\theta) = p(\theta)\cos\theta - p'(\theta)\sin\theta, \\ y(\theta) = p(\theta)\sin\theta + p'(\theta)\cos\theta. \end{cases} \tag{21}$$

This shows immediately that the radius of curvature is $\rho = p + p''$ and the convexity of Ω implies $p + p'' \geq 0$. Conversely, a classical fact recalled in all these cited works is the fact that a C^1 and 2π periodic real function p which satisfies (in the sense of distributions) $p + p'' \geq 0$ is the support function of a unique convex shape $\Omega \subset \mathbb{R}^2$.

In all the works cited above, the support function is discretized with the help of a truncated spectral decomposition (namely, Fourier series in dimension two and spherical harmonics decomposition in dimension three). This makes easier the treatment of some constraints, like the constant width constraint, but forces the support function to be smooth. On the other hand, as

already seen in [19], when dealing with spectral functionals under convexity constraints, minimizers often tend to contain segments in their boundary. In such cases, the support function is not smooth anymore and then its parametrization through spectral decomposition is no longer appropriate. This is why in the following we choose a different approach which can handle discontinuities in the derivative of the support function.

Consider N a positive integer and $\theta_i = 2\pi i/N$, $i = 0, ..., N-1$ angles in $[0, 2\pi]$. Then, the support function will be discretized by considering its values $p_i = p(\theta_i)$, $i = 0, ..., N-1$ at the angles chosen before. The first and second derivatives of p are approximated using finite differences in the following way:

$$p'(\theta_i) = \frac{p_{i+1} - p_{i-1}}{2h} \quad \text{and} \quad p''(\theta_i) = \frac{p_{i+1} + p_{i-1} - 2p_i}{h^2}, \tag{22}$$

for $i = 0, ..., N-1$ (indices considered modulo N) and $h = 2\pi/N$.

The computation of the Steklov eigenvalues is done using the software FreeFEM ([15]) while the constrained optimization is done with the algorithm IPOPT ([24]). In the FreeFEM software the domain is meshed and finite elements are used in the computations. The main components of the optimization algorithm are shown below.

Convexity and Diameter constraints. The convexity constraint is imposed pointwise for each θ_i, that is,

$$p(\theta_i) + p''(\theta_i) \geq 0 \quad i = 0, ..., N-1.$$

Taking into account the second equality of (22) then yields

$$p_i + \frac{1}{h^2}(p_{i+1} + p_{i-1} - 2p_i) \geq 0, \ i = 0, ..., N-1,$$

which can be translated into a set of N linear inequality constraints on the variables p_i, $i = 0, ..., N-1$.

We have seen before that diameter constraints can be imposed by controlling the quantity $p(\theta) + p(\theta + \pi)$. In practice, we consider N even, so that $\theta_i + \pi = \theta_{i+N/2}$ with indices considered modulo N. The fact that the shape has diameter at most d is expressed by

$$p_i + p_{i+N/2} \leq d, \ i = 0, ..., N/2 - 1.$$

In order to have diameter exactly equal to d, we impose the reverse inequality for one pair of opposite points:

$$p_0 + p_{N/2} \geq d.$$

At last, we obtain a set of $N/2 + 1$ linear inequality constraints.

Construction of the mesh. The inputs of the objective function are values of $p_0, ..., p_{N-1}$. Starting from these values and using (21) we can find points $Q_i(x(\theta_i), y(\theta_i))$ by approximating the derivatives $p'(\theta_i)$ using centered finite differences as shown above (see (22)). The points Q_i form a polygonal line whose interior is meshed in FreeFEM.

Note that the meshing algorithm in FreeFEM will give an error if the polygonal line contains self-intersections. In case such an error appears we reject the current computation. The algorithm IPOPT which deals with the optimization will eventually produce admissible vectors when imposing the convexity constraints shown above.

The discretization points may be close on the boundary of $\partial\Omega$, especially close to eventual angular points. On the other hand, on parts which are almost flat, the discretization points will be rather sparse. In order to have a good finite element approximation the quality of the resulting mesh is improved using the command `adaptmesh` with parameters `hmax=0.05*D`, `nbvx=50000`, `iso=1` refering to the maximal size of triangles, maximal number of vertices and the quality of the mesh. For more details, one should consult the FreeFEM documentation.

Eigenvalue problem and gradient of the objective function. Once the mesh is constructed, FreeFEM allows us to solve the eigenvalue problem starting from the variational formulation using finite elements. It is possible to recover the approximate eigenvalue and the associated eigenfunction. Concerning the finite element setup, P_2 finite elements are used for solving the eigenvalue problem and P_1 elements are used for evaluating the shape derivative (which contains derivatives of P_2 functions). In the discrete setting, the eigenvalue $\sigma_k(\Omega)$ is a function of the parameters p_i, $i = 0, ..., N-1$:

$$\sigma_k(\Omega) \approx F_k(p_0, ..., p_{N-1}).$$

In order to use a gradient based optimization algorithm, it is necessary to compute the gradient of F_k with respect to each one of the parameters. The classical method to handle this is to use the shape derivative formula given in Theorem 2.4. Then, for each one of the parameters p_i, we look at the boundary perturbation V_i obtained when considering perturbations $p_i + \delta t$ as $\delta t \to 0$. It suffices to put the perturbation V_i in the shape derivative formula to obtain the gradient with respect to the variable p_i. A straightforward computation shows that a perturbation of the form $p_i + \delta t$ induces a vector field V_i such that $V_i.n$ is equal to 1 at Q_i and is 0 for every other point in the discrete boundary. Define χ_i to be a function which is piecewise affine on the segments $Q_i Q_{i+1}$ and which is 1 at Q_i and 0 at Q_j for $j \neq i$. Then the gradient of F_k with respect to p_i is approximated by

$$\frac{\partial F_k}{\partial p_i} = \int_{\partial \Omega} \left(|\nabla u_k|^2 - (\partial_n u_k)^2 - H \sigma_k u_k^2 \right) \chi_i d\sigma$$

where, as usual, σ_k, u_k denote the k-th eigenvalue and associated eigenfunction and H denotes the curvature. The FreeFEM command `curvature` is used to approximate the discrete curvature of the polygonal line.

Optimization algorithm. As already mentioned before, the optimization is done in FreeFEM using the algorithm IPOPT. The inputs are the function F_k and its gradient, as well as the matrices involving the linear discrete constraints associated to the convexity and diameter constraints. In addition to the linear constraints, pointwise positivity constraints are imposed on p_i, since we can assume that the origin is strictly inside our shape. The discretization uses $N = 200$ angles in $[0, 2\pi]$ and the diameter is fixed to $D = 2$.

Results and remarks. The algorithm is run for $1 \leq N \leq 7$ and the resulting numerical optimal shapes are represented in Figure 2. The numerical results give rise to the following remarks:

- As predicted by the theoretical results, in each case the optimal eigenvalue is multiple: $\sigma_k(\Omega_k^*) = \sigma_{k+1}(\Omega_k^*)$
- In all the numerical results obtained the convexity constraint is saturated in some region, giving rise to segments in the boundary. Note that the direct discretization of the support function proposed here manages to properly capture this phenomenon, which was not the case for the Fourier decomposition used in [1].
- The diameter constraint seems to be saturated only at the two antipodal points included in the constraints. Moreover, angular points seem to be present at these antipodal points.
- The sequence of maximizers seems to become more and more flat as the index grows. We discuss that point below.
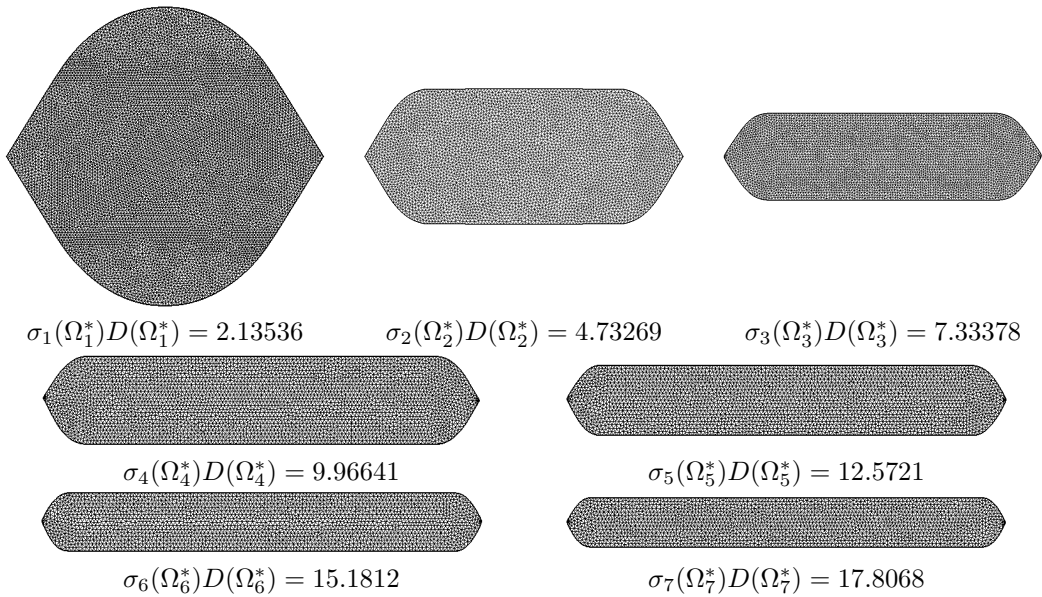

$\sigma_1(\Omega_1^*)D(\Omega_1^*) = 2.13536$ $\qquad$ $\sigma_2(\Omega_2^*)D(\Omega_2^*) = 4.73269$ $\qquad$ $\sigma_3(\Omega_3^*)D(\Omega_3^*) = 7.33378$

$\sigma_4(\Omega_4^*)D(\Omega_4^*) = 9.96641$ $\qquad\qquad$ $\sigma_5(\Omega_5^*)D(\Omega_5^*) = 12.5721$

$\sigma_6(\Omega_6^*)D(\Omega_6^*) = 15.1812$ $\qquad\qquad$ $\sigma_7(\Omega_7^*)D(\Omega_7^*) = 17.8068$

FIGURE 2. Results of the optimization algorithm: maximization of $\sigma_k(\Omega)D(\Omega)$ under convexity constraint.

Remark 4.1. The numerical results above suggest that the sequence of optimal domains converges to a segment. This would be really interesting since it would differ with the case of Dirichlet eigenvalues where it is shown in [11] (perimeter constraint) and in [6] (diameter constraint) that the sequence of optimal domains converges to a disk (or even a ball in any dimension in the second case). A possible strategy to prove that fact can be by contradiction:

(1) Assume that the sequence of optimal domains $(\Omega_k^*)_k$ converges for the Hausdorff metric to a convex open set Ω_∞ with diameter d_0. Then, provide a uniform control of the difference of eigenvalues of the kind $0 \leq \sigma_k(\Omega_k^*) - \sigma_k(\Omega_\infty) \leq \varepsilon(k)$ with $\varepsilon(k)/k \to 0$ as $k \to \infty$.

(2) Now fix a smooth convex set ω (like an elongated ellipse) of diameter d_0 and with a perimeter such that $P(\omega) < P(\Omega_\infty)$. Optimality of Ω_k^* then yields

$$\sigma_k(\Omega_\infty) + \varepsilon(k) \geq \sigma_k(\Omega_k^*) \geq \sigma_k(\omega).$$

(3) Finally apply Weyl's law for Steklov eigenvalues that writes (see [16])

$$\sigma_k(\Omega) \sim \frac{2\pi k}{P(\Omega)} \quad \text{as } k \to \infty \tag{23}$$

to get $P(\Omega_\infty) \leq P(\omega)$, which is a contradiction.

The flaw of this strategy is that Weyl's law (23) is only known for smooth open sets, see the discussion in [16]. Now, we have no guarantee that the limit convex domain Ω_∞ would be smooth!

Alternative approach and the non-convex case. The method described above imposed rigorously the convexity and diameter constraints. As can be seen in Figure 2 the numerical maximizers seem to saturate the diameter constraint at exactly two points. This suggests *a posteriori* that a simpler parametrization should work. Moreover, it seems to be enough to only impose the diameter condition for exactly two points.

One may consider the segment $[-D/2, D/2] \times \{0\}$ in $\mathbb{R}^2$ and the family of shapes defined as regions contained between the graphs of two functions $f_1, f_2 : [-D/2, D/2] \to \mathbb{R}$, $f_1 \leq f_2$. From a discrete point of view f_1 and f_2 are discretized at an equidistant family of points in $[-D/2, D/2]$ with values $p_1, ..., p_N$ and $q_1, ..., q_N$. The convexity of f_1 and the concavity of f_2 translate to the discrete inequalities

$$\frac{f_{i-1} + f_{i+1}}{2} \geq f_i, \ \frac{q_{i-1} + q_{i+1}}{2} \leq q_i, \ i = 1, ..., N$$

with the convention $p_0 = p_{N+1} = q_0 = q_{N+1} = 0$. Given values $p_i \leq q_i$, $i = 1, ..., N$, the discrete domain is meshed in FreeFEM and the Steklov eigenvalue problem is solved using finite elements as before. The computation of the gradient with respect to the variables p_i, q_i is similar to what was done with the support function. One only needs to keep in mind that a perturbation in these variables amounts to a perturbation in the y direction of the normal to the boundary of Ω.

The resulting numerical algorithm gives exactly the same results as those shown in Figure 2. Moreover, even if the diameter constraint is not imposed during the optimization at other points than the endpoints of the segment $[-D/2, D/2]$, the numerical shapes obtained verify the diameter constraint everywhere.

This alternative method has the advantage that it can also handle the non-convex case. Indeed, if we do not impose that f_1 is convex and f_2 is concave during the optimization process we obtain the non-degenerate shapes shown in Figure 3. Note that for the first eigenvalue, the result is a slight loss of convexity near the two corners observed in the domain. However, the corresponding maximal eigenvalue is only a bit larger than the one obtained imposing the convexity constraint. For $k \in \{2, 3\}$ we observe obvious departs from the convexity near the parts where the results in Figure 2 contained segments in the boundary. One may note similarities between the maximizer of $\sigma_2(\Omega)D(\Omega)$ and the maximizer of $\sigma_2(\Omega)$ under area constraint shown in [5], but the case $k = 3$ is completely different.

The same remarks as in the convex case hold: the k-th eigenvalue is multiple at the optimum, the diameter constraint is saturated at exactly two points and the minimizers become flatter as k grows. The fact that the numerical algorithm does find non-degenerate shapes suggests that the existence of a maximizer should hold even without the convexity assumption.

$$\sigma_1(\omega_1^*)D(\omega_1^*) = 2.13623 \qquad \sigma_1(\omega_2^*)D(\omega_2^*) = 4.92925 \qquad \sigma_1(\omega_3^*)D(\omega_3^*) = 7.76108$$
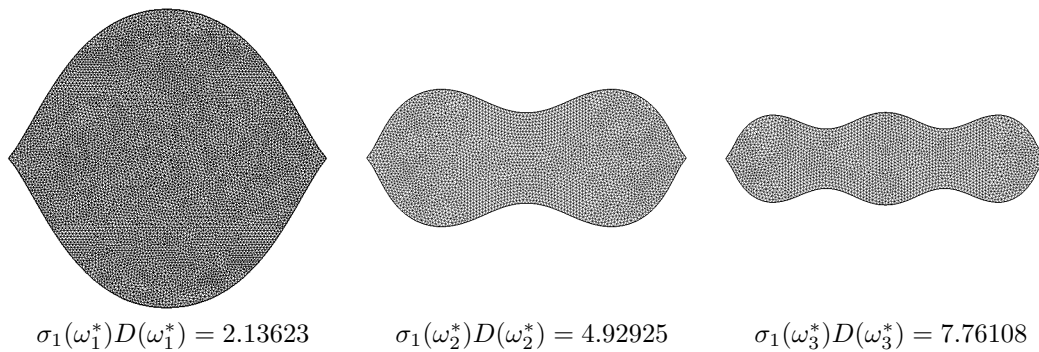
FIGURE 3. Numerical results obtained using the alternative method for $k \in \{1, 2, 3\}$: maximization of $\sigma_k(\Omega)D(\Omega)$ without the convexity constraint.

Acknowledgments

References

[1] P. Antunes and B. Bogosel, Parametric Shape Optimization using the Support Function Arxiv preprint: 1809.00254

[2] M. S. Ashbaugh, Isoperimetric and universal inequalities for eigenvalues, Spectral theory and geometry (Edinburgh, 1998), London Math. Soc. Lecture Note Ser., vol. 273, Cambridge Univ. Press, Cambridge, 1999, pp. 95139.

[3] T. Bayen and D. Henrion Semidefinite programming for optimizing convex bodies under width constraints. *Optim. Methods Softw.*, 27(6):1073–1099, 2012.

[4] B. Bogosel: The Steklov spectrum on moving domains, *Appl. Math. Optim.* 75 (2017), no. 1, 125.

[5] B. Bogosel, D. Bucur, A. Giacomini, Optimal shapes maximizing the Steklov eigenvalues, *SIAM J. Math. Anal.* 49 (2017), no. 2, 16451680.

[6] B. Bogosel, A. Henrot, I. Lucardesi: Minimization of the eigenvalues of the Dirichlet-Laplacian with a diameter constraint. SIAM J. Math. Anal. 50 (2018), no. 5, 53375361.

[7] F. Brock, An isoperimetric inequality for eigenvalues of the Stekloff problem, *ZAMM Z. Angew. Math. Mech.* 81 (2001), no. 1, 6971.

[8] D. Bucur: Minimization of the k-th eigenvalue of the Dirichlet Laplacian, *Arch. Ration. Mech. Anal.* **206**, no. 3, 1073–1083 (2012).

[9] D. Bucur, G. Buttazzo Variational methods in shape optimization problems. Progress in Nonlinear Differential Equations and their Applications, 65. Birkhäuser Boston, Inc., Boston, MA, 2005.

[10] D. Bucur, V. Ferone, C. Nitsch, C. Trombetti, Weinstock inequality in higher dimensions, to appear in *Journal of Differential Geometry*, preprint: arXiv:1710.04587

[11] D. Bucur, P. Freitas: Asymptotic behaviour of optimal spectral planar domains with fixed perimeter. J. Math. Phys. 54 (2013), no. 5, 053504, 6 pp.

[12] D. Bucur, A. Giacomini, P. Trebeschi: L^∞ bounds of Steklov eigenfunctions and spectrum stability under domain variation, preprint 2019 http://cvgmt.sns.it/paper/4268/

[13] B. Colbois, A. El Soufi, A. Girouard Isoperimetric control of the Steklov spectrum, J. Funct. Anal. 261 (2011), no. 5, 13841399.

[14] M. Dambrine, D. Kateb, J. Lamboley, An extremal eigenvalue problem for the Wentzell-Laplace operator, *Ann. Inst. H. Poincar Anal. Non Linaire* 33 (2016), no. 2, 409450.

[15] F. Hecht, New development in FreeFem++. *J. Numer. Math.*, 20(3-4):251–265, 2012.

[16] A. Girouard, I. Polterovich, Spectral geometry of the Steklov problem (survey article). *J. Spectr. Theory* 7 (2017), no. 2, 321359.

[17] A. Henrot: Extremum problems for eigenvalues of elliptic operators. Birkhäuser, Basel (2006).

[18] A. Henrot (ed): Shape Optimization and Spectral Theory. De Gruyter open (2017), freely downloadable at https://www.degruyter.com/view/product/490255

[19] A. Henrot, E. Oudet, Minimizing the second eigenvalue of the Laplace operator with Dirichlet boundary conditions, *Archive for Rational Mechanics and Analysis*, **169** (2003), 73-87.

[20] A. Henrot, M. Pierre: Shape variation and optimization. A geometrical analysis. EMS Tracts in Mathematics, 28. European Mathematical Society (EMS), Zrich, 2018.

[21] D. Mazzoleni, A. Pratelli: Existence of minimizers for spectral problems, *J. Math. Pures Appl.* **9**, no. 3, 433–453 (2013).

[22] J. Nečas, Les méthodes directes en théorie des équations elliptiques, Masson et Cie, Éditeurs, Paris; Academia, Prague 1967.

[23] F. Rellich, Darstellung der eigenwerte $\Delta u + \lambda u$ durch ein randintegral, *Math. Z.*, 46 (1940), 635-646.

[24] A. Wächter, L. Biegler On the implementation of an interior-point filter line-search algorithm for large-scale nonlinear programming, *Math. Program.*, 106 (2006), 25-57.

[25] R. Weinstock, Inequalities for a classical eigenvalue problem, *J. Rational Mech. Anal.* 3 (1954), 745753.

Abdelkader Al Sayed, Institut Élie Cartan de Lorraine, UMR 7502, Université de Lorraine CNRS, email: `alsayed.abdkader@gmail.com`

Beniamin Bogosel, Centre de Mathématiques Appliquées, Ecole Polytechnique, UMR CNRS 7641, email: `beniamin.bogosel@polytechnique.edu`

Antoine Henrot, Institut Élie Cartan de Lorraine, UMR 7502, Université de Lorraine CNRS, email: `antoine.henrot@univ-lorraine.fr` (corresponding author)

Florent Nacry, Laboratoire de Mathématiques, Physique et Systèmes, Université de Perpignan Via Domitia, 52 Avenue Paul Alduy, 66860 Perpignan, email: `florent.nacry@univ-perp.fr`