

On optimal potential problems with changing sing data

Faustino Maestre

Departamento de Ecuaciones Diferenciales y Análisis Numérico
Universidad de Sevilla. Spain.

IX Partial differential equations, optimal design and numerics

Mass optimization problems and homogenization
Benasque, España. 2022

Joint work: Giuseppe Buttazzo, Università di Pisa,
Juan Casado, Universidad de Sevilla,
Bozhidar Velichkov, Università di Pisa.

- 1 Introduction
 - Statement of the problem
 - Capacitary measures

- 2 Existence results
 - The admissible class of potentials and its relaxation
 - Optimality conditions
 - Saturation of the Constraint

- 3 Unbounded domain
 - Existence Result

- 4 Numerical experiments
 - Bounded domain
 - Unbouded domain

On optimal potential problems

1 Introduction

- Statement of the problem
- Capacitary measures

2 Existence results

- The admissible class of potentials and its relaxation
- Optimality conditions
- Saturation of the Constraint

3 Unbounded domain

- Existence Result

4 Numerical experiments

- Bounded domain
- Unbounded domain

On optimal potential problems

1 Introduction

- Statement of the problem
- Capacitary measures

2 Existence results

- The admissible class of potentials and its relaxation
- Optimality conditions
- Saturation of the Constraint

3 Unbounded domain

- Existence Result

4 Numerical experiments

- Bounded domain
- Unbounded domain

We consider $D \subset \mathbb{R}^d$ a fixed open bounded set. We are interested in the optimization problem:

$$\min_{V \in \mathcal{V}} \int_D g(x) u(x) \, dx$$

subject to

$$\begin{cases} -\Delta u + V u = f & \text{in } D, \\ u = 0 & \text{on } \partial D, \end{cases}$$

where,

$$\mathcal{V} = \left\{ V : D \rightarrow [0, +\infty] : V \text{ Lebesgue measurable, } \int_D \psi(V(x)) \, dx \leq 1 \right\}$$

and ψ satisfying some appropriate qualitative conditions.

The function $\psi : [0, +\infty] \rightarrow [0, +\infty]$ we assume that:

- (i) ψ is strictly decreasing;
- (ii) there exist $p > 1$ such that the function $s \mapsto \psi^{-1}(s^p)$ is convex.

For instance the following functions:

- 1 $\psi(s) = s^{-p}$, for any $p > 0$,
- 2 $\psi(s) = e^{-\alpha s}$, for any $\alpha > 0$,

The choice $\psi(s) = e^{-\alpha s}$ was proposed in [Buttazzo et al ,2014], in order to approximate shape optimization problems with Dirichlet condition on the free boundary.

Moreover, as $\alpha \rightarrow 0$ the problems with the parameter α were shown to Γ -converge to the shape optimization problem with a volume constraint $|\Omega| \leq 1$ being Ω the shape variable.

The function $\psi : [0, +\infty] \rightarrow [0, +\infty]$ we assume that:

- (i) ψ is strictly decreasing;
- (ii) there exist $p > 1$ such that the function $s \mapsto \psi^{-1}(s^p)$ is convex.

For instance the following functions:

- 1 $\psi(s) = s^{-p}$, for any $p > 0$,
- 2 $\psi(s) = e^{-\alpha s}$, for any $\alpha > 0$,

The choice $\psi(s) = e^{-\alpha s}$ was proposed in [Buttazzo et al ,2014], in order to approximate shape optimization problems with Dirichlet condition on the free boundary.

Moreover, as $\alpha \rightarrow 0$ the problems with the parameter α were shown to Γ -converge to the shape optimization problem with a volume constraint $|\Omega| \leq 1$ being Ω the shape variable.

The function $\psi : [0, +\infty] \rightarrow [0, +\infty]$ we assume that:

- (i) ψ is strictly decreasing;
- (ii) there exist $p > 1$ such that the function $s \mapsto \psi^{-1}(s^p)$ is convex.

For instance the following functions:

- 1 $\psi(s) = s^{-p}$, for any $p > 0$,
- 2 $\psi(s) = e^{-\alpha s}$, for any $\alpha > 0$,

The choice $\psi(s) = e^{-\alpha s}$ was proposed in [Buttazzo et al ,2014], in order to approximate shape optimization problems with Dirichlet condition on the free boundary.

Moreover, as $\alpha \rightarrow 0$ the problems with the parameter α were shown to Γ -converge to the shape optimization problem with a volume constraint $|\Omega| \leq 1$ being Ω the shape variable.

The function $\psi : [0, +\infty] \rightarrow [0, +\infty]$ we assume that:

- (i) ψ is strictly decreasing;
- (ii) there exist $p > 1$ such that the function $s \mapsto \psi^{-1}(s^p)$ is convex.

For instance the following functions:

- 1 $\psi(s) = s^{-p}$, for any $p > 0$,
- 2 $\psi(s) = e^{-\alpha s}$, for any $\alpha > 0$,

The choice $\psi(s) = e^{-\alpha s}$ was proposed in [Buttazzo et al ,2014], in order to approximate shape optimization problems with Dirichlet condition on the free boundary.

Moreover, as $\alpha \rightarrow 0$ the problems with the parameter α were shown to Γ -converge to the shape optimization problem with a volume constraint $|\Omega| \leq 1$ being Ω the shape variable.

Existence results:

- $f \geq 0$ and $g \leq 0$ (o reverse case) cost is monotonically increasing (maximum principle) and volume constraint saturated ([Buttazzo et al., 2014])
- Optimal domains with f and g are allowed to change sing ([Buttazzo and Velichkov, 2018])
- We analyze the existence of optimal potentials when f and g are allowed to change sing. We expect no saturation of volume constraint.

Main assumptions:

- linear cost (otherwise simple examples show optimal solution only exists in a relaxed sense).
- D bounded (characterization of the relaxed formulation).

Existence results:

- $f \geq 0$ and $g \leq 0$ (o reverse case) cost is monotonically increasing (maximum principle) and volume constraint saturated ([Buttazzo et al., 2014])
- **Optimal domains** with f and g are allowed to change sing ([Buttazzo and Velichkov, 2018])
- We analyze the existence of **optimal potentials** when f and g are allowed to change sing. We expect no saturation of volume constraint.

Main assumptions:

- linear cost (otherwise simple examples show optimal solution only exists in a relaxed sense).
- D bounded (characterization of the relaxed formulation).

Existence results:

- $f \geq 0$ and $g \leq 0$ (o reverse case) cost is monotonically increasing (maximum principle) and volume constraint saturated ([Buttazzo et al., 2014])
- **Optimal domains** with f and g are allowed to change sing ([Buttazzo and Velichkov, 2018])
- We analyze the existence of **optimal potentials** when f and g are allowed to change sing. We expect no saturation of volume constraint.

Main assumptions:

- linear cost (otherwise simple examples show optimal solution only exists in a relaxed sense).
- D bounded (characterization of the relaxed formulation).

Existence results:

- $f \geq 0$ and $g \leq 0$ (o reverse case) cost is monotonically increasing (maximum principle) and volume constraint saturated ([Buttazzo et al., 2014])
- **Optimal domains** with f and g are allowed to change sing ([Buttazzo and Velichkov, 2018])
- We analyze the existence of **optimal potentials** when f and g are allowed to change sing. We expect no saturation of volume constraint.

Main assumptions:

- linear cost (otherwise simple examples show optimal solution only exists in a relaxed sense).
- D bounded (characterization of the relaxed formulation).

Existence results:

- $f \geq 0$ and $g \leq 0$ (o reverse case) cost is monotonically increasing (maximum principle) and volume constraint saturated ([Buttazzo et al., 2014])
- **Optimal domains** with f and g are allowed to change sing ([Buttazzo and Velichkov, 2018])
- We analyze the existence of **optimal potentials** when f and g are allowed to change sing. We expect no saturation of volume constraint.

Main assumptions:

- linear cost (otherwise simple examples show optimal solution only exists in a relaxed sense).
- D bounded (characterization of the relaxed formulation).

On optimal potential problems

1 Introduction

- Statement of the problem
- **Capacitary measures**

2 Existence results

- The admissible class of potentials and its relaxation
- Optimality conditions
- Saturation of the Constraint

3 Unbounded domain

- Existence Result

4 Numerical experiments

- Bounded domain
- Unbounded domain

A *capacitary measure* μ is a nonnegative Borel measure on D , possibly taking the value $+\infty$, that vanishes on all sets of capacity zero. Notation $\mu \in \mathcal{M}_{cap}$.

Capacity is intended with respect to the H^1 norm

$$cap(E, D) = \inf \left\{ \int_D |\nabla u|^2 dx + \int_D u^2 dx : u \in H_0^1(D), \right. \\ \left. u \geq 1 \text{ in a neighborhood of } E \right\}$$

We consider the Hilbert space $H_0^1(D) \cap L^2(\mu)$ endowed with norm:

$$\|u\| = \left(\|\nabla u\|_{L^2(D)}^2 + \|u\|_{L^2(\mu)}^2 \right)^{1/2}.$$

We say that $u \in H_0^1(D) \cap L^2(\mu)$ is a solution of the problem

$$-\Delta u + \mu u = f, \quad \text{for a function } f \in L^2(D),$$

if

$$\int_D \nabla u \nabla \phi \, dx + \int_D u \phi \, d\mu = \int_D f \phi \, dx \quad \forall \phi \in H_0^1(D) \cap L^2(\mu),$$

On optimal potential problems

- 1 Introduction
 - Statement of the problem
 - Capacitary measures

- 2 Existence results
 - The admissible class of potentials and its relaxation
 - Optimality conditions
 - Saturation of the Constraint

- 3 Unbounded domain
 - Existence Result

- 4 Numerical experiments
 - Bounded domain
 - Unbounded domain

On optimal potential problems

- 1 Introduction
 - Statement of the problem
 - Capacitary measures

- 2 Existence results
 - The admissible class of potentials and its relaxation
 - Optimality conditions
 - Saturation of the Constraint

- 3 Unbounded domain
 - Existence Result

- 4 Numerical experiments
 - Bounded domain
 - Unbounded domain

For $V \in \mathcal{V}$ the state equation

$$-\Delta u + Vu = f, \quad u \in H_0^1(D) \cap L^2(V).$$

The capacitary measure μ associated to V is defined as:

$$\mu(A) = \begin{cases} \int_A V(x) dx & \text{if } \text{cap}(A \cap \{V = +\infty\}) = 0 \\ +\infty & \text{if } \text{cap}(A \cap \{V = +\infty\}) > 0, \end{cases}$$

which implies $u = 0$ quasi-everywhere on the set $\{V = +\infty\}$.
Abusing the terminology we will identify μ and V .

Relaxed problem

We put $\overline{\mathcal{V}}$ the family of capacitary measures μ obtained as limits of sequences (V_n) of potentials in \mathcal{V} . Relaxed problem:

$$\min_{\mu \in \overline{\mathcal{V}}} \int_D g(x) u(x) dx$$

subject to

$$\begin{cases} u \in H_0^1(D) \cap L^2(\mu) \\ -\Delta u + \mu u = f \end{cases} \quad \text{in } D,$$

Theorem

Let $D \subset \mathbb{R}^d$ be a bounded open set and let ψ satisfy the assumptions i) and ii) above. Then, for every $f, g \in L^2(D)$, the original optimization problem has a solution.

Relaxed problem

We put $\overline{\mathcal{V}}$ the family of capacitary measures μ obtained as limits of sequences (V_n) of potentials in \mathcal{V} . Relaxed problem:

$$\min_{\mu \in \overline{\mathcal{V}}} \int_D g(x) u(x) dx$$

subject to

$$\begin{cases} u \in H_0^1(D) \cap L^2(\mu) \\ -\Delta u + \mu u = f \end{cases} \quad \text{in } D,$$

Theorem

Let $D \subset \mathbb{R}^d$ be a bounded open set and let ψ satisfy the assumptions i) and ii) above. Then, for every $f, g \in L^2(D)$, the original optimization problem has a solution.

On optimal potential problems

- 1 Introduction
 - Statement of the problem
 - Capacitary measures

- 2 Existence results
 - The admissible class of potentials and its relaxation
 - **Optimality conditions**
 - Saturation of the Constraint

- 3 Unbounded domain
 - Existence Result

- 4 Numerical experiments
 - Bounded domain
 - Unbounded domain

We consider $u, p \in H_0^1(D) \cap L^2(\mu)$ solutions of:

$$\begin{cases} -\Delta u + \mu u = f & \text{in } D, \\ u = 0 & \text{on } \partial D, \end{cases} \quad \begin{cases} -\Delta p + \mu p = g & \text{in } D, \\ p = 0 & \text{on } \partial D, \end{cases}$$

Proposition

Suppose that μ is a solution of the relaxed optimization problem on the bounded domain $D \subset \mathbb{R}^d$. Then

$$u p \leq 0 \quad \text{a.e. on } D.$$

Moreover, the above inequality holds quasi-everywhere on D .

We consider $u, p \in H_0^1(D) \cap L^2(\mu)$ solutions of:

$$\begin{cases} -\Delta u + \mu u = f & \text{in } D, \\ u = 0 & \text{on } \partial D, \end{cases} \quad \begin{cases} -\Delta p + \mu p = g & \text{in } D, \\ p = 0 & \text{on } \partial D, \end{cases}$$

Proposition

Suppose that μ is a solution of the relaxed optimization problem on the bounded domain $D \subset \mathbb{R}^d$. Then

$$u p \leq 0 \quad \text{a.e. on } D.$$

Moreover, the above inequality holds quasi-everywhere on D .

On optimal potential problems

- 1 Introduction
 - Statement of the problem
 - Capacitary measures

- 2 Existence results
 - The admissible class of potentials and its relaxation
 - Optimality conditions
 - **Saturation of the Constraint**

- 3 Unbounded domain
 - Existence Result

- 4 Numerical experiments
 - Bounded domain
 - Unbounded domain

Saturation volume constraint

One cannot expect that the constraint $\int_D \psi(V) dx \leq 1$ is saturated.

Moreover, we will show that, if this is not the case, then the optimal potential V can be reduced to a domain Ω , that is, V is a potential of the form:

$$V = 0 \text{ on } \Omega \quad \text{and} \quad V = +\infty \text{ on } \Omega^c.$$

On optimal potential problems

- 1 Introduction
 - Statement of the problem
 - Capacitary measures

- 2 Existence results
 - The admissible class of potentials and its relaxation
 - Optimality conditions
 - Saturation of the Constraint

- 3 **Unbounded domain**
 - Existence Result

- 4 Numerical experiments
 - Bounded domain
 - Unbounded domain

Unbounded domain

We consider $D = \mathbb{R}^d$. We are interested in the optimization problem:

$$\min_{\mu \in \mathcal{M}_{cap}} \int_{\mathbb{R}^d} j(x, u(x), \nabla u(x)) \, dx$$

subject to

$$-\Delta u + \mu u = f \quad \text{in } \mathbb{R}^d,$$

and

$$\Psi(\mu) \leq 1,$$

some difficulties

- Which is good space X for the solutions of $-\Delta u + \mu u = f$, ???

$$\int_{\mathbb{R}^d} \nabla u \cdot \nabla v \, dx + \int_{\mathbb{R}^d} u v \, d\mu = \int_{\mathbb{R}^d} f v, \quad v \in X,$$

We consider:

$$W(x) = \frac{1}{1 + |x|} \text{ if } d \neq 2, \quad W(x) = \frac{1}{(1 + |x|) \log(2 + |x|)} \text{ if } d = 2.$$

and we put:

$$L = \{u : \mathbb{R}^d \rightarrow \mathbb{R} : Wu \in L^2(\mathbb{R}^d)\}$$

$$H = \{u \in L \cap H_{loc}^1(\mathbb{R}^d) : \nabla u \in L^2(\mathbb{R}^d)^d\}$$

and one gets

$$\|u\|_L \leq C \left(\|\nabla u\|_{L^2(\mathbb{R}^d)} + \|u\|_{L_\mu^2} \right)$$

We take

$$X = H \cap L_\mu^2$$

some difficulties

- Which is good space X for the solutions of $-\Delta u + \mu u = f$, ???

$$\int_{\mathbb{R}^d} \nabla u \cdot \nabla v \, dx + \int_{\mathbb{R}^d} u v \, d\mu = \int_{\mathbb{R}^d} f v, \quad v \in X,$$

We consider:

$$W(x) = \frac{1}{1 + |x|} \text{ if } d \neq 2, \quad W(x) = \frac{1}{(1 + |x|) \log(2 + |x|)} \text{ if } d = 2.$$

and we put:

$$L = \{u : \mathbb{R}^d \rightarrow \mathbb{R} : Wu \in L^2(\mathbb{R}^d)\}$$

$$H = \{u \in L \cap H_{loc}^1(\mathbb{R}^d) : \nabla u \in L^2(\mathbb{R}^d)^d\}$$

and one gets

$$\|u\|_L \leq C \left(\|\nabla u\|_{L^2(\mathbb{R}^d)} + \|u\|_{L_\mu^2} \right)$$

We take

$$X = H \cap L_\mu^2$$

some difficulties

- Which is good space X for the solutions of $-\Delta u + \mu u = f$, ???

$$\int_{\mathbb{R}^d} \nabla u \cdot \nabla v \, dx + \int_{\mathbb{R}^d} u v \, d\mu = \int_{\mathbb{R}^d} f v, \quad v \in X,$$

We consider:

$$W(x) = \frac{1}{1 + |x|} \text{ if } d \neq 2, \quad W(x) = \frac{1}{(1 + |x|) \log(2 + |x|)} \text{ if } d = 2.$$

and we put:

$$L = \{u : \mathbb{R}^d \rightarrow \mathbb{R} : Wu \in L^2(\mathbb{R}^d)\}$$

$$H = \{u \in L \cap H_{loc}^1(\mathbb{R}^d) : \nabla u \in L^2(\mathbb{R}^d)^d\}$$

and one gets

$$\|u\|_L \leq C \left(\|\nabla u\|_{L^2(\mathbb{R}^d)} + \|u\|_{L_\mu^2} \right)$$

We take

$$X = H \cap L_\mu^2$$

some difficulties

- Which is good space X for the solutions of $-\Delta u + \mu u = f$, ???

$$\int_{\mathbb{R}^d} \nabla u \cdot \nabla v \, dx + \int_{\mathbb{R}^d} u v \, d\mu = \int_{\mathbb{R}^d} f v, \quad v \in X,$$

We consider:

$$W(x) = \frac{1}{1 + |x|} \text{ if } d \neq 2, \quad W(x) = \frac{1}{(1 + |x|) \log(2 + |x|)} \text{ if } d = 2.$$

and we put:

$$L = \{u : \mathbb{R}^d \rightarrow \mathbb{R} : Wu \in L^2(\mathbb{R}^d)\}$$

$$H = \{u \in L \cap H_{loc}^1(\mathbb{R}^d) : \nabla u \in L^2(\mathbb{R}^d)^d\}$$

and one gets

$$\|u\|_L \leq C \left(\|\nabla u\|_{L^2(\mathbb{R}^d)} + \|u\|_{L_\mu^2} \right)$$

We take

$$X = H \cap L_\mu^2$$

- How is defined $\Psi(\mu)$???

We decompose $\mu = \mu^a + \mu^s + \infty_K$, and consider:

$$\Psi(\mu) = \int_{\mathbb{R}^d} \psi(\mu^a) dx + C_\psi \mu^s(\mathbb{R}^d) + \psi(\infty) \text{cap}(K),$$

where $\psi : \mathbb{R}^+ \rightarrow [0, +\infty]$ is a convex and lower semicontinuous function and

$$C_\psi = \lim_{t \rightarrow +\infty} \frac{\psi(t)}{t}$$

On optimal potential problems

- 1 Introduction
 - Statement of the problem
 - Capacitary measures

- 2 Existence results
 - The admissible class of potentials and its relaxation
 - Optimality conditions
 - Saturation of the Constraint

- 3 Unbounded domain
 - **Existence Result**

- 4 Numerical experiments
 - Bounded domain
 - Unbounded domain

Theorem

We consider $j : \mathbb{R}^d \times \mathbb{R} \times \mathbb{R}^d \rightarrow \mathbb{R} \cup \{+\infty\}$ measurable in $x \in \mathbb{R}$, lower semicontinuous in (s, ξ) and some growth conditions in (s, ξ) . We consider $\psi : \mathbb{R} \rightarrow [0, +\infty]$ convex and lower semicontinuous and a measure $\nu \in \mathcal{M}_{cap}$ such that there exists $\hat{\mu} \in \mathcal{M}_{cap}$ satisfying

$$\hat{\mu} \geq \nu, \quad \Psi(\hat{\mu}) \leq 1.$$

Moreover, if $d = 1, 2$ we assume that:

either $\psi(0) > 0$ or ν is not the null measure

Then, for every $f \in H'$ the optimization problem has at least one solution.

On optimal potential problems

- 1 Introduction
 - Statement of the problem
 - Capacitary measures

- 2 Existence results
 - The admissible class of potentials and its relaxation
 - Optimality conditions
 - Saturation of the Constraint

- 3 Unbounded domain
 - Existence Result

- 4 Numerical experiments
 - Bounded domain
 - Unbounded domain

On optimal potential problems

- 1 Introduction
 - Statement of the problem
 - Capacitary measures

- 2 Existence results
 - The admissible class of potentials and its relaxation
 - Optimality conditions
 - Saturation of the Constraint

- 3 Unbounded domain
 - Existence Result

- 4 Numerical experiments
 - Bounded domain
 - Unbounded domain

Numerical Analysis

We propose the numerical analysis of the following problem. We consider $D = [0, 1]^2$:

$$\min_{V \in \mathcal{V}} \int_D g(x) u(x) dx$$

subject to

$$\begin{cases} -\Delta u + V u = f & \text{in } D, \\ u = 0 & \text{on } \partial D, \end{cases}$$

where,

$$\mathcal{V} = \left\{ V : D \rightarrow [0, +\infty] : V \text{ Lebesgue measurable, } \int_D \psi(V(x)) dx \leq 1 \right\}$$

$$\text{and } \psi(s) = \frac{1}{m} e^{-\alpha s}$$

We use Method of Moving Asymptotes (MMA).

The structure of the algorithm is as follows.

- Initialization of the potential $V^0 \in \mathcal{V}$;
- for $k \geq 0$, iteration until convergence as follows:
 - ▶ compute the state u_{V^k} and then the co-state p_{V^k} ,
 - ▶ compute the descent direction $\tilde{V}^k(x) = -u_{V^k} \cdot p_{V^k}$
 - ▶ update the potential V^k in D :

$$V^{k+1} = V^k + \ell_k \tilde{V}^k,$$

We use Method of Moving Asymptotes (MMA).

The structure of the algorithm is as follows.

- Initialization of the potential $V^0 \in \mathcal{V}$;
- for $k \geq 0$, iteration until convergence as follows:
 - ▶ compute the state u_{V^k} and then the co-state p_{V^k} ,
 - ▶ compute the descent direction $\tilde{V}^k(x) = -u_{V^k} \cdot p_{V^k}$
 - ▶ update the potential V^k in D :

$$V^{k+1} = V^k + \ell_k \tilde{V}^k,$$

We use Method of Moving Asymptotes (MMA).

The structure of the algorithm is as follows.

- Initialization of the potential $V^0 \in \mathcal{V}$;
- for $k \geq 0$, iteration until convergence as follows:
 - ▶ compute the state u_{V^k} and then the co-state p_{V^k} ,
 - ▶ compute the descent direction $\tilde{V}^k(x) = -u_{V^k} \cdot p_{V^k}$
 - ▶ update the potential V^k in D :

$$V^{k+1} = V^k + \ell_k \tilde{V}^k,$$

We use Method of Moving Asymptotes (MMA).

The structure of the algorithm is as follows.

- Initialization of the potential $V^0 \in \mathcal{V}$;
- for $k \geq 0$, iteration until convergence as follows:
 - ▶ compute the state u_{V^k} and then the co-state p_{V^k} ,
 - ▶ compute the descent direction $\tilde{V}^k(x) = -u_{V^k} \cdot p_{V^k}$
 - ▶ update the potential V^k in D :

$$V^{k+1} = V^k + \ell_k \tilde{V}^k,$$

We use Method of Moving Asymptotes (MMA).

The structure of the algorithm is as follows.

- Initialization of the potential $V^0 \in \mathcal{V}$;
- for $k \geq 0$, iteration until convergence as follows:
 - ▶ compute the state u_{V^k} and then the co-state p_{V^k} ,
 - ▶ compute the descent direction $\tilde{V}^k(x) = -u_{V^k} \cdot p_{V^k}$
 - ▶ update the potential V^k in D :

$$V^{k+1} = V^k + \ell_k \tilde{V}^k,$$

We use `FreeFem++ v 3.50` completed with the library `NLopt`.

In the following, we take $g \equiv 1$ and consider different choices for f and parameters α and m .

We use `FreeFem++ v 3.50` completed with the library `NLopt`.

In the following, we take $g \equiv 1$ and consider different choices for f and parameters α and m .

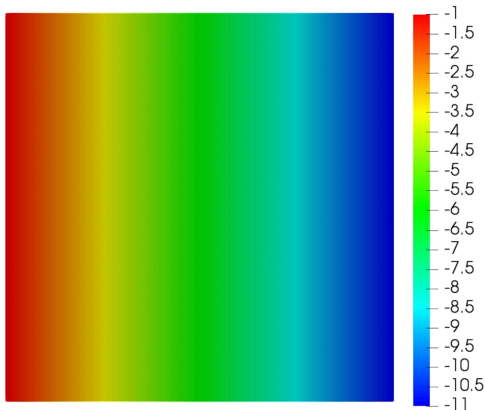
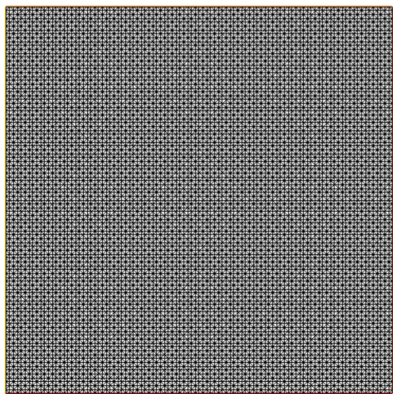


Figure: To the left: the domain D and its triangulation; number of nodes: 40401; number of triangles: 80000. To the right: the right-hand side function $f(x, y) = -(1 + 10x)$.

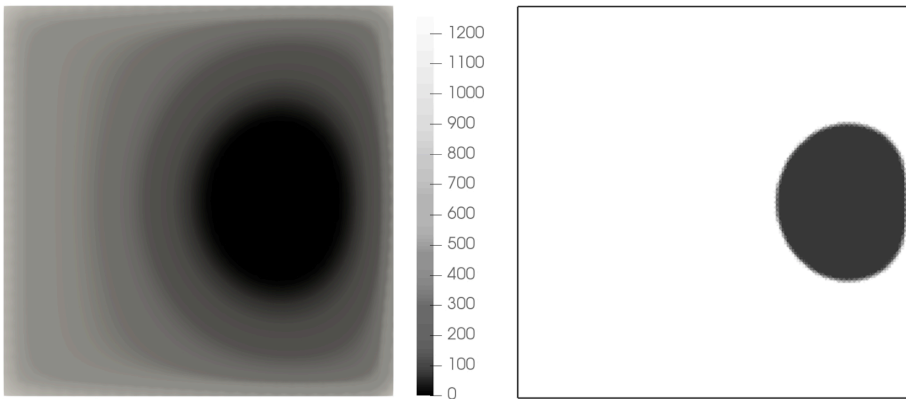


Figure: The optimal potential V_{opt} for volume constraint $m = 0.2 = m_{opt}$. Case $\alpha = 10^{-2}$ (left) and $\alpha = 3 \cdot 10^{-4}$ (right).

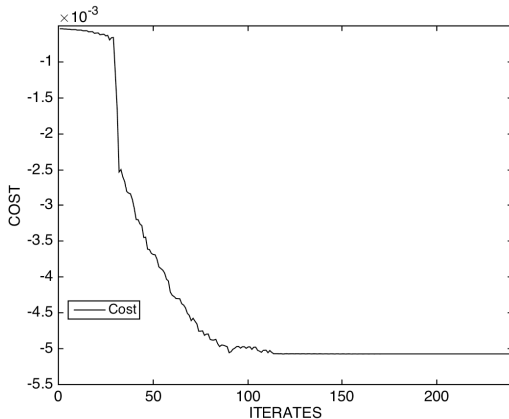


Figure: Cost evolution for the example from previous Figure, case $\alpha = 3 \cdot 10^{-4}$.

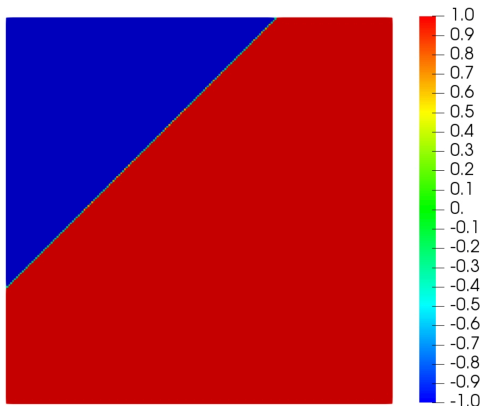


Figure: The right-hand side function f is given by $f(x, y) = -1$, if $y - 1.4x \geq 0.3$, and $f(x, y) = 1$, if $y - 1.4x < 0.3$

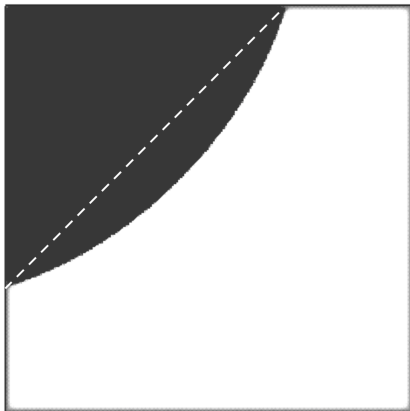
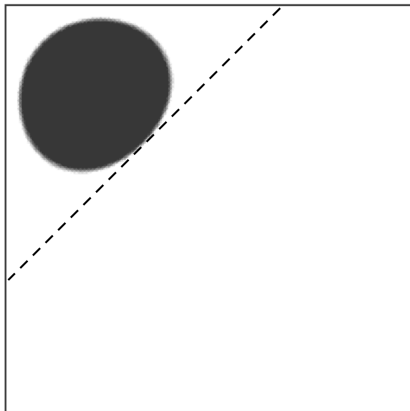


Figure: Optimal potential V_{opt} for $m = 0.2$ (left) and $m = 0.45$ (right). The occupied volume on the rig is 0.33276.

On optimal potential problems

- 1 Introduction
 - Statement of the problem
 - Capacitary measures

- 2 Existence results
 - The admissible class of potentials and its relaxation
 - Optimality conditions
 - Saturation of the Constraint

- 3 Unbounded domain
 - Existence Result

- 4 Numerical experiments
 - Bounded domain
 - Unbounded domain

We consider the sequence of problems

$$\min \left\{ \int_{D_n} j(x, u, \nabla u) dx : -\Delta u + \mu u = f \text{ in } D_n, u_n = 0 \text{ on } \partial D_n, \right. \\ \left. \int_{D_n} \psi(\mu) dx \leq 1, \mu \in L^\infty(D_n), \nu_n \leq \mu \leq k_n \right\}. \quad (1)$$

Where $\mathbb{R}^d = \cup_n D_n$ and $k_n \rightarrow +\infty$.

Theorem

Assume the conditions of the existence Theorem with j such that $j(x, s, \xi)$ is continuous in (s, ξ) and there exist $k \in L^1$, $l_1, l_2 \in L^\infty$ such that

$$j(x, s, \xi) \leq k(x) + l_1 W^2 |s|^2 + l_2 |\xi|^2, \quad \forall (s, \xi) \in \mathbb{R} \times \mathbb{R}^d, \text{ a.e. } x \in \mathbb{R}^d.$$

Then there exists a sequence $\tilde{k}_n \rightarrow +\infty$ with $\tilde{k}_n > \|\varphi_n\|_{L^\infty}$ such that taking $k_n \geq \tilde{k}_n$ problem (1) has at least a solution μ_n . Extending μ_n by zero outside D_n and extracting a subsequence which γ -converges to a measure μ we have that μ is a solution of (1). Moreover, defining u_n as the solution of

$$-\Delta u_n + \mu_n u_n = f \text{ in } D_n, \quad u_n = 0 \text{ on } \partial D_n,$$

we have

$$\lim_{n \rightarrow \infty} \int_{D_n} j(x, u_n, \nabla u_n) dx = \int j(x, u, \nabla u) dx.$$

Numerical Simulations

We consider

- $j(x, u, \nabla u) = g \cdot u$
- $g(x, y) = \frac{1}{1+\epsilon(x^2+y^2)^3}$, such that $W^{-1}g \in L^2$ we have considered $\epsilon = 10^{-10}$.
- $\psi(s) = \frac{1}{m}e^{-\alpha s}$ with $\alpha = 3 \cdot 10^{-4}$.
- $f(x, y) = \begin{cases} x^2 + y^2 - 1 & \text{if } x^2 + y^2 < 11, \\ \frac{10}{1+\epsilon(x^2+y^2)^3} & \text{if } x^2 + y^2 > 11. \end{cases}$

The solution for these data can be explicitly obtained and it is given by $\Omega = B(0, R)$.

$$R = \sqrt{m/\pi} \wedge \sqrt{2}.$$

Numerical Simulations

We consider

- $j(x, u, \nabla u) = g \cdot u$
- $g(x, y) = \frac{1}{1+\epsilon(x^2+y^2)^3}$, such that $W^{-1}g \in L^2$ we have considered $\epsilon = 10^{-10}$.
- $\psi(s) = \frac{1}{m}e^{-\alpha s}$ with $\alpha = 3 \cdot 10^{-4}$.
- $f(x, y) = \begin{cases} x^2 + y^2 - 1 & \text{if } x^2 + y^2 < 11, \\ \frac{10}{1+\epsilon(x^2+y^2)^3} & \text{if } x^2 + y^2 > 11. \end{cases}$

The solution for these data can be explicitly obtained and it is given by $\Omega = B(0, R)$.

$$R = \sqrt{m/\pi} \wedge \sqrt{2}.$$

Numerical Simulations

We consider

- $j(x, u, \nabla u) = g \cdot u$
- $g(x, y) = \frac{1}{1+\epsilon(x^2+y^2)^3}$, such that $W^{-1}g \in L^2$ we have considered $\epsilon = 10^{-10}$.
- $\psi(s) = \frac{1}{m}e^{-\alpha s}$ with $\alpha = 3 \cdot 10^{-4}$.
- $f(x, y) = \begin{cases} x^2 + y^2 - 1 & \text{if } x^2 + y^2 < 11, \\ \frac{10}{1+\epsilon(x^2+y^2)^3} & \text{if } x^2 + y^2 > 11. \end{cases}$

The solution for these data can be explicitly obtained and it is given by $\Omega = B(0, R)$.

$$R = \sqrt{m/\pi} \wedge \sqrt{2}.$$

Numerical Simulations

We consider

- $j(x, u, \nabla u) = g \cdot u$
- $g(x, y) = \frac{1}{1+\epsilon(x^2+y^2)^3}$, such that $W^{-1}g \in L^2$ we have considered $\epsilon = 10^{-10}$.
- $\psi(s) = \frac{1}{m}e^{-\alpha s}$ with $\alpha = 3 \cdot 10^{-4}$.
- $f(x, y) = \begin{cases} x^2 + y^2 - 1 & \text{if } x^2 + y^2 < 11, \\ \frac{10}{1+\epsilon(x^2+y^2)^3} & \text{if } x^2 + y^2 > 11. \end{cases}$

The solution for these data can be explicitly obtained and it is given by $\Omega = B(0, R)$.

$$R = \sqrt{m/\pi} \wedge \sqrt{2}.$$

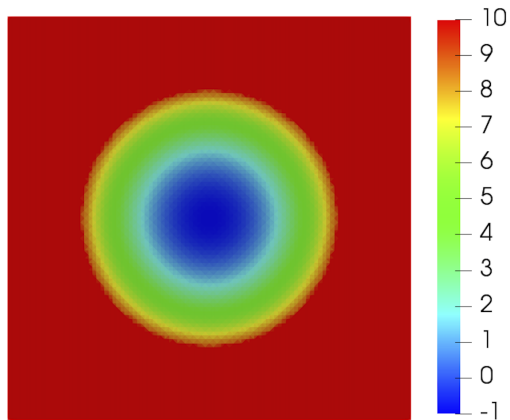


Figure: Right-hand side function f in $D = (-5, 5) \times (-5, 5)$

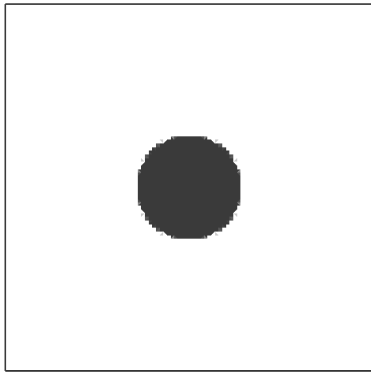
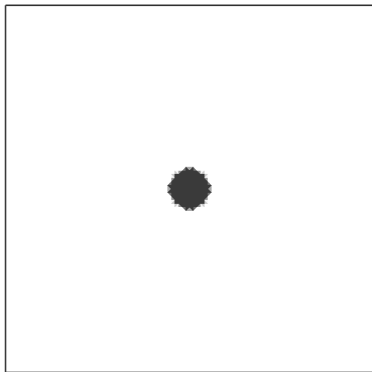


Figure: The optimal potential μ_{opt} for volume constraint $m = 2 = m_{opt}$ (left) and $m = 20 > 6.367 = m_{opt}$ (right).

We consider now:

$$f(x, y) = \begin{cases} -10 & \text{if } (x - 2)^2 + (y + 1)^2 < 1, \\ 10 & \text{if } (x + 2)^2 + (y - 0.5)^2 < 1 \\ 0 & \text{otherwise.} \end{cases}$$

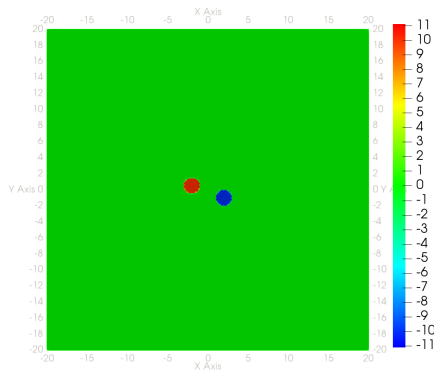
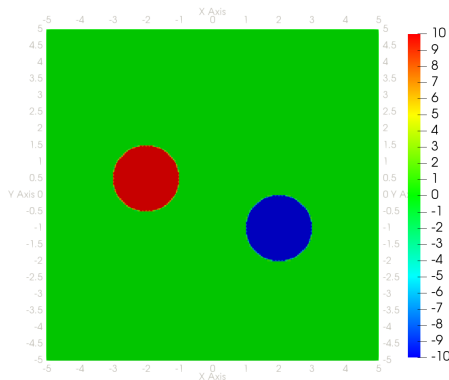


Figure: The right-hand side function f in $D = (-5, 5) \times (-5, 5)$ (left) and in $D = (-20, 20) \times (-20, 20)$ (right).

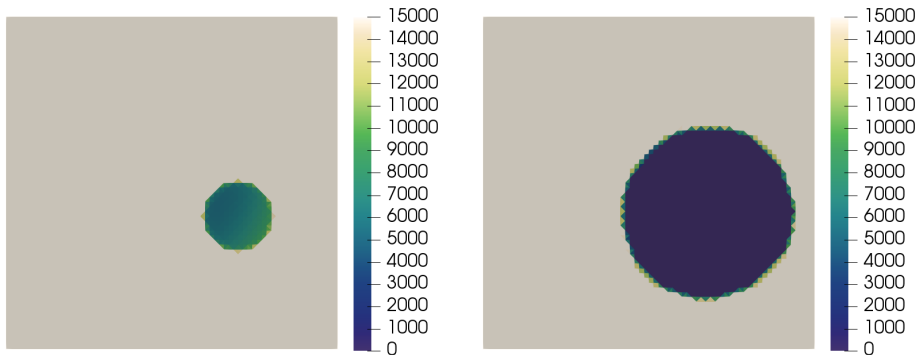


Figure: The optimal potential μ_{opt} . $D = (-5, 5) \times (-5, 5)$, volume constraint $m = 0.2$ (left) and $m = 10$ (right).

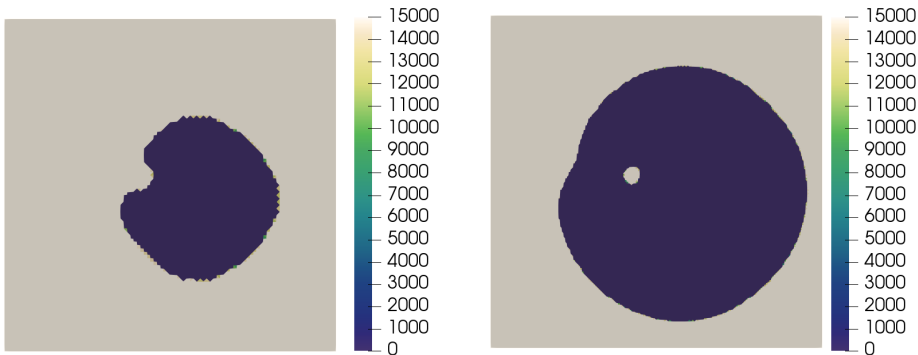


Figure: The optimal potential μ_{opt} . $D = (-12.5, 12.5) \times (-12.5, 12.5)$ and volume constraint $m = 110$ (left). $D = (-20, 20) \times (-20, 20)$ and $m = 400$ (right).

THANK YOU
FOR YOUR
ATTENTION!!!