

# Multi-material shape optimization via a level set method

Grégoire Allaire<sup>1</sup>, Charles Dapogny<sup>2</sup>, Gabriel Delgado<sup>1,3</sup>, Georgios Michailidis<sup>1</sup>

<sup>1</sup> CMAP, UMR 7641 École Polytechnique, Palaiseau, France

<sup>2</sup> Laboratoire Jean Kuntzmann, Université Joseph Fourier, Grenoble, France

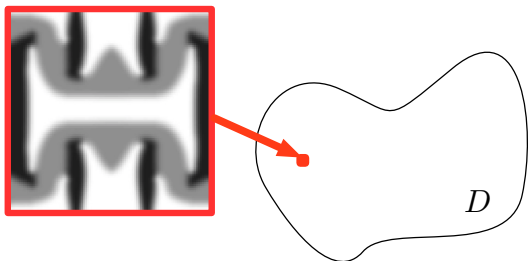
<sup>3</sup> Laboratoire J.-L. Lions, UPMC, Paris, France

29<sup>th</sup> January, 2015

## A foreword around multi-phase optimization

**Multi-phase optimization** is about finding the **optimal repartition** of two, or several, materials with conflicting properties within a fixed set. This problem has multiple applications in industrial design:

- At the macroscopic level: Repartition of several materials within a given structure to combine their respective assets.
- At the microscopic level: Mixture of several phases to achieve new materials with unique features (e.g. design of materials with **negative Poisson's ratio**, or **negative coefficient of thermal expansion...**).



Design of a material with negative CTE [Mi].

- 1 Introduction and definitions
  - Foreword
  - A short reminder of 'classical' shape optimization in linear elasticity
- 2 The (exact) sharp-interface multi-phase problem
- 3 The smoothed-interface approach
  - Setting of the smoothed-interface approach
  - A digression around the signed distance function
  - Shape derivatives in the smoothed-interface setting
- 4 Numerical study
  - Presentation of the numerical algorithm
  - Numerical examples

- 1 Introduction and definitions
  - Foreword
  - A short reminder of ‘classical’ shape optimization in linear elasticity
- 2 The (exact) sharp-interface multi-phase problem
- 3 The smoothed-interface approach
  - Setting of the smoothed-interface approach
  - A digression around the signed distance function
  - Shape derivatives in the smoothed-interface setting
- 4 Numerical study
  - Presentation of the numerical algorithm
  - Numerical examples

## Preliminaries: the usual linear elasticity setting (I)

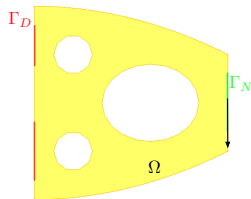
A **shape** is a bounded domain  $\Omega \subset \mathbb{R}^d$ , which is

- **fixed** on a part  $\Gamma_D$  of its boundary,
- submitted to **surface loads**  $g$ , applied on  $\Gamma_N \subset \partial\Omega$ ,  $\Gamma_D \cap \Gamma_N = \emptyset$ .

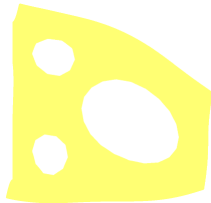
The displacement vector field  $u_\Omega : \Omega \rightarrow \mathbb{R}^d$  is governed by the **linear elasticity system**:

$$\left\{ \begin{array}{ll} -\operatorname{div}(Ae(u_\Omega)) & = 0 \quad \text{in } \Omega \\ u_\Omega & = 0 \quad \text{on } \Gamma_D \\ Ae(u_\Omega)n & = g \quad \text{on } \Gamma_N \\ Ae(u_\Omega)n & = 0 \quad \text{on } \Gamma := \partial\Omega \setminus (\Gamma_D \cup \Gamma_N) \end{array} \right. ,$$

where  $e(u) = \frac{1}{2}(\nabla u^T + \nabla u)$  is the strain tensor, and  $A$  is the Hooke's law of the material.



A 'Cantilever'



The deformed cantilever

## Preliminaries: the usual linear elasticity setting (II)

**Goal:** Starting from an initial structure  $\Omega_0$ , find a new one  $\Omega$  that minimizes a certain functional of the domain  $J(\Omega)$ .

### Examples:

- The work of the external loads  $g$  or **compliance**  $C(\Omega)$  of domain  $\Omega$ :

$$C(\Omega) = \int_{\Omega} A e(u_{\Omega}) : e(u_{\Omega}) dx = \int_{\Gamma_N} g \cdot u_{\Omega} ds$$

- A **least-square error** between  $u_{\Omega}$  and a target displacement  $u_0 \in H^1(\Omega)^d$  (useful when designing micro-mechanisms):

$$D(\Omega) = \left( \int_{\Omega} k(x) |u_{\Omega} - u_0|^{\alpha} dx \right)^{\frac{1}{\alpha}},$$

where  $\alpha$  is a fixed parameter, and  $k(x)$  is a weight factor.

A **volume constraint** may be enforced with a fixed penalty parameter  $\ell$ :

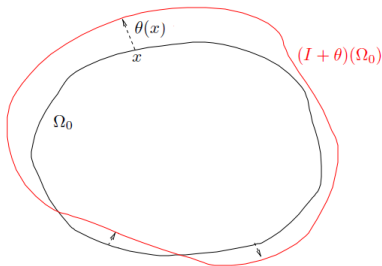
Minimize  $J(\Omega) := C(\Omega) + \ell \text{Vol}(\Omega)$ , or  $D(\Omega) + \ell \text{Vol}(\Omega)$ .

## Differentiation with respect to the domain: Hadamard's method

**Hadamard's boundary variation method** describes variations of a reference, Lipschitz domain  $\Omega$  of the form:

$$\Omega \rightarrow \Omega_\theta := (I + \theta)(\Omega),$$

for 'small'  $\theta \in W^{1,\infty}(\mathbb{R}^d, \mathbb{R}^d)$ .



### Definition 1.

Given a smooth domain  $\Omega$ , a functional  $F(\Omega)$  of the domain is **shape differentiable** at  $\Omega$  if the function

$$W^{1,\infty}(\mathbb{R}^d, \mathbb{R}^d) \ni \theta \mapsto F(\Omega_\theta)$$

is Fréchet-differentiable at 0, i.e. the following expansion holds around 0:

$$F(\Omega_\theta) = F(\Omega) + F'(\Omega)(\theta) + o(\|\theta\|_{W^{1,\infty}(\mathbb{R}^d, \mathbb{R}^d)}).$$

## Differentiation with respect to the domain: Hadamard's method

Techniques close to optimal control theory make it possible to compute shape gradients; in the case of 'many' functionals of the domain  $J(\Omega)$ , the shape derivative has the particular **structure**:

$$J'(\Omega)(\theta) = \int_{\Gamma} v_{\Omega} \theta \cdot n \, ds,$$

where  $v_{\Omega}$  is a scalar field depending on  $u_{\Omega}$ , and possibly on an **adjoint state**  $p_{\Omega}$ .

**Example:** If  $J(\Omega) = C(\Omega) = \int_{\Gamma_N} g \cdot u_{\Omega} \, ds$  is the **compliance**,  
 $v_{\Omega} = -Ae(u_{\Omega}) : e(u_{\Omega})$ .



## The generic numerical algorithm

This shape gradient provides a natural **descent direction** for functional  $J$ : *for instance*, defining  $\theta$  as

$$\theta = -v_{\Omega} n$$

yields, for  $t > 0$  sufficiently small (*to be found numerically*):

$$J((I + t\theta)(\Omega)) = J(\Omega) - t \int_{\Gamma} v_{\Omega}^2 ds + o(t) < J(\Omega)$$

**Gradient algorithm:** For  $n = 0, \dots$  convergence,

1. Compute the solution  $u_{\Omega^n}$  (and  $p_{\Omega^n}$ ) of the elasticity system on  $\Omega^n$ .
2. Compute the shape gradient  $J'(\Omega^n)$  thanks to the previous formula, and infer a descent direction  $\theta^n$  for the cost functional.
3. **Advect** the shape  $\Omega^n$  according to  $\theta^n$ , so as to get  $\Omega^{n+1} := (I + \theta^n)(\Omega^n)$ .

## 1 Introduction and definitions

- Foreword
- A short reminder of 'classical' shape optimization in linear elasticity

## 2 The (exact) sharp-interface multi-phase problem

## 3 The smoothed-interface approach

- Setting of the smoothed-interface approach
- A digression around the signed distance function
- Shape derivatives in the smoothed-interface setting

## 4 Numerical study

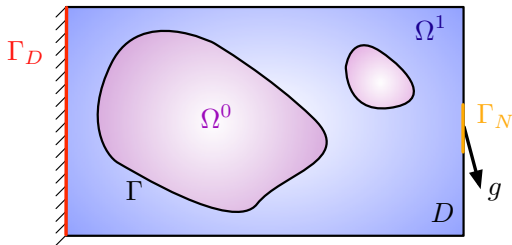
- Presentation of the numerical algorithm
- Numerical examples

## The multi-material shape optimization setting (I)

- A fixed working domain  $D \subset \mathbb{R}^d$  is occupied by two complementary phases  $\Omega^0$  and  $\Omega^1$ , filled with elastic materials with Hooke's laws  $A_0, A_1$ .
- The structure  $D$  is clamped on a region  $\Gamma_D \subset \partial D$ , surface loads are applied on  $\Gamma_N \subset \partial D$ , as well as body forces  $f$ .
- The total, **discontinuous** Hooke's law in  $D$  is:

$$A_{\Omega^0} := A_0 \chi_0 + A_1 \chi_1,$$

where  $\chi_i$  is the characteristic function of the phase  $\Omega^i$ .



## The multi-material shape optimization setting (II)

- The **displacement**

$$u_{\Omega^0} \in H_{\Gamma_D}^1(D)^d := \{u \in H^1(D)^d, u = 0 \text{ on } \Gamma_D\}$$

of the total structure  $D$  satisfies:

$$\begin{cases} -\operatorname{div}(A_{\Omega^0} e(u)) = f & \text{in } D \\ u = 0 & \text{on } \Gamma_D \\ A_1 e(u)n = g & \text{on } \Gamma_N \end{cases}.$$

- Goal:** Minimize a functional of the mixture of the form:

$$J(\Omega^0) = \int_D j(x, u_{\Omega^0}) \, dx + \int_{\Gamma_N} k(x, u_{\Omega^0}) \, ds,$$

under **constraints** (e.g. on the volume of one of the phases).

- Example:** The **compliance** of the total structure  $D$ :

$$C(\Omega^0) = \int_D A_{\Omega^0} e(u_{\Omega^0}) : e(u_{\Omega^0}) \, dx = \int_D f \cdot u_{\Omega^0} \, dx + \int_{\Gamma_N} g \cdot u_{\Omega^0} \, ds.$$

## The multi-material shape optimization setting (III)

- The material properties are different from either side of  $\Gamma \Rightarrow$  some quantities are discontinuous across  $\Gamma$ .
- If  $\alpha$  is a discontinuous quantity, with values  $\alpha^0, \alpha^1$  in  $\overline{\Omega^0}, \overline{\Omega^1}$  respectively,  $[\alpha] := \alpha^1 - \alpha^0$  is the **jump** of  $\alpha$  across  $\Gamma$ .
- If  $\mathcal{M}$  is a tensor-valued function, denote as:

$$\forall x \in \Gamma, \mathcal{M} = \begin{pmatrix} \mathcal{M}_{\tau\tau}(x) & \mathcal{M}_{\tau n}(x) \\ \mathcal{M}_{n\tau}(x) & \mathcal{M}_{nn}(x) \end{pmatrix}$$

its representation in a local basis  $(\tau, n)$  of  $\mathbb{R}^d$ .

- Difficulty: The strain tensor  $e \equiv e(u_{\Omega^0})$  has continuous components  $e_{\tau\tau}$ , but discontinuous components  $e_{\tau n}, e_{n\tau}, e_{nn}$ . The stress tensor  $\sigma \equiv \sigma(u_{\Omega^0})$  has continuous components  $\sigma_{n\tau}, \sigma_{\tau n}$  and  $\sigma_{nn}$ , but  $\sigma_{\tau\tau}$  is discontinuous.

# Shape derivative in the sharp-interface context (I)

## Theorem 1 ([AlJouVG]).

The functional  $J(\Omega^0)$  is shape differentiable, and its derivative reads:

$$\forall \theta \in W^{1,\infty}(\mathbb{R}^d, \mathbb{R}^d), \quad J'(\Omega^0)(\theta) = - \int_{\Gamma} \mathcal{D}(x, u_{\Omega^0}, p_{\Omega^0}) \theta \cdot n \, ds,$$

where the integrand factor  $\mathcal{D}(x, u, p)$  is defined as:

$$\mathcal{D}(x, u, p) = -\sigma(p)_{nn} : [e(u)_{nn}] - 2\sigma(u)_{n\tau} : [e(p)_{n\tau}] + [\sigma(u)_{\tau\tau}] : e(p)_{\tau\tau},$$

and  $p_{\Omega^0} \in H_{\Gamma_D}^1(D)^d$  is an *adjoint state*, defined as the solution to:

$$\begin{cases} -\operatorname{div}(A_{\Omega^0} e(p)) &= -j'(x, u_{\Omega^0}) & \text{in } D, \\ p &= 0 & \text{on } \Gamma_D, \\ (A_1 e(p)) n &= -k'(x, u_{\Omega^0}) & \text{on } \Gamma_N, \end{cases}$$

## Shape derivative in the sharp-interface context (II)

This formula is difficult to use in numerical practice, since it involves the jumps of discontinuous quantities across  $\Gamma$ .

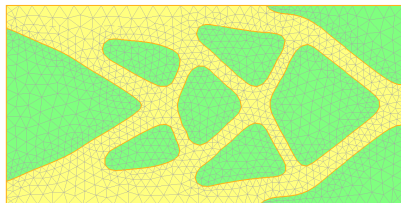
### Potential remedies:

- Discrete approach: [AIDaDeIMi]

Consider the shape derivative of the **discretization**  $J_h(\Omega^0)$  of  $J(\Omega^0)$  on the actual mesh, which features the numerical solution  $u_h$  (resp.  $p_h$ ) of the state (resp. adjoint) elasticity system.

- Body-fitted approach: [AIDaFr]

The interface  $\Gamma$  is **explicitly** discretized at each step of the process.



- ① Introduction and definitions
  - Foreword
  - A short reminder of 'classical' shape optimization in linear elasticity
- ② The (exact) sharp-interface multi-phase problem
- ③ The smoothed-interface approach
  - Setting of the smoothed-interface approach
  - A digression around the signed distance function
  - Shape derivatives in the smoothed-interface setting
- ④ Numerical study
  - Presentation of the numerical algorithm
  - Numerical examples



- ① Introduction and definitions
  - Foreword
  - A short reminder of 'classical' shape optimization in linear elasticity
- ② The (exact) sharp-interface multi-phase problem
- ③ The smoothed-interface approach
  - Setting of the smoothed-interface approach
  - A digression around the signed distance function
  - Shape derivatives in the smoothed-interface setting
- ④ Numerical study
  - Presentation of the numerical algorithm
  - Numerical examples

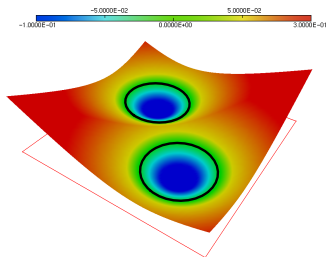
## The signed distance function

### Definition 2.

The **signed distance function**  $d_\Omega$  to a bounded domain  $\Omega \subset \mathbb{R}^d$  is defined as:

$$\forall x \in \mathbb{R}^d, \quad \begin{cases} -d(x, \partial\Omega) & \text{if } x \in \Omega \\ 0 & \text{if } x \in \partial\Omega \\ d(x, \partial\Omega) & \text{if } x \in \mathbb{R}^d \setminus \Omega \end{cases},$$

where  $d(\cdot, \partial\Omega)$  stands for the usual Euclidean distance function to  $\partial\Omega$ .



Graph of the signed distance function to a union of two disks (in black)

# The smoothed-interface setting (I)

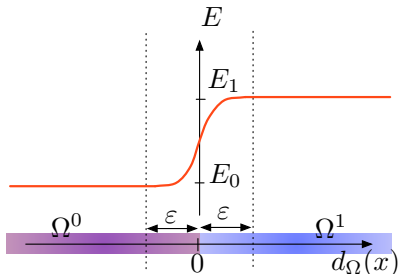
- The discontinuous tensor  $A_{\Omega^0}$  is approximated by:

$$\forall x \in D, \quad A_{\Omega^0, \varepsilon}(x) := A_0 + h_\varepsilon(d_{\Omega^0}(x))(A_1 - A_0),$$

where  $h_\varepsilon$  is a smooth approximation of the [Heaviside function](#):

$$h_\varepsilon(t) = \begin{cases} 0 & \text{if } t < -\varepsilon \\ \frac{1}{2} \left( 1 + \frac{t}{\varepsilon} + \frac{1}{\pi} \sin \left( \frac{\pi t}{\varepsilon} \right) \right) & \text{if } -\varepsilon \leq t \leq \varepsilon \\ 1 & \text{if } t > \varepsilon \end{cases}.$$

- This accounts for a [smooth interpolation](#) of the material properties between the two phases over a tubular neighborhood of  $\Gamma$  of [fixed](#) width  $2\varepsilon$ .



## The smoothed-interface setting (II)

- The **smoothed-interface problem** is then that of minimizing:

$$J_\varepsilon(\Omega^0) = \int_D j(x, u_{\Omega^0, \varepsilon}) dx + \int_{\Gamma_N} k(x, u_{\Omega^0, \varepsilon}) ds$$

(under constraints), where  $u_{\Omega^0, \varepsilon}$  arises as the solution to:

$$\begin{cases} -\operatorname{div}(A_{\Omega^0, \varepsilon} e(u)) = f & \text{in } D \\ u = 0 & \text{on } \Gamma_D \\ A_1 e(u) n = g & \text{on } \Gamma_N \end{cases}.$$

- It is worth considering for at least two reasons:
  - It is an **approximation of the sharp-interface problem**, and is easier to handle numerically.
  - It has some interest on its own, especially when it comes to **modelling interfaces**: interfaces may involve complex and ill-understood processes, which are better described e.g. by non monotone transition regions.

- ① Introduction and definitions
  - Foreword
  - A short reminder of 'classical' shape optimization in linear elasticity
- ② The (exact) sharp-interface multi-phase problem
- ③ The smoothed-interface approach
  - Setting of the smoothed-interface approach
  - A digression around the signed distance function
  - Shape derivatives in the smoothed-interface setting
- ④ Numerical study
  - Presentation of the numerical algorithm
  - Numerical examples

# Signed distance function and geometry (I)

## Definition 3.

Let  $\Omega \subset \mathbb{R}^d$  be a Lipschitz, bounded open set;

- Let  $x \in \mathbb{R}^d$ ; the *set of projections*  $\Pi_{\partial\Omega}(x)$  of  $x$  onto  $\partial\Omega$  is:

$$\Pi_{\partial\Omega}(x) = \{y \in \partial\Omega, d(x, \partial\Omega) = |x - y|\}.$$

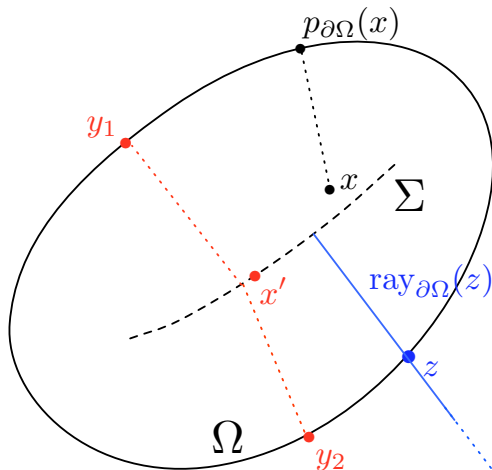
- When this set is a singleton,  $p_{\partial\Omega}(x)$  is *the projection* of  $x$  onto  $\partial\Omega$ .
- The *skeleton*  $\Sigma$  of  $\partial\Omega$  is:

$$\Sigma := \{x \in \mathbb{R}^d, d_{\Omega}^2 \text{ is not differentiable at } x\}.$$

- For  $x \in \partial\Omega$ , the *ray* emerging from  $x$  is:

$$\text{ray}_{\partial\Omega}(x) := p_{\partial\Omega}^{-1}(x).$$

## Signed distance function and geometry (II)



$x$  has a unique projection over  $\partial\Omega$ , whereas  $x'$  has two such points  $y_1, y_2$ .

## Signed distance function and geometry (III)

### Proposition 2.

Let  $\Omega \subset \mathbb{R}^d$  be a Lipschitz, bounded open set;

- A point  $x \in \mathbb{R}^d$  has a unique projection point  $p_{\partial\Omega}(x)$  iff  $x \notin \Sigma$ . In such a case,  $d_\Omega$  is differentiable at  $x$ , and its gradient reads:

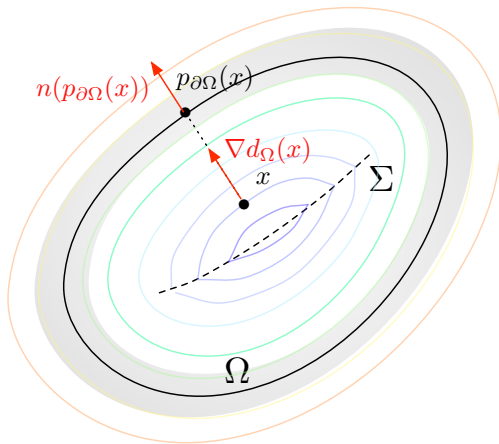
$$\nabla d_\Omega(x) = \frac{x - p_{\partial\Omega}(x)}{d_\Omega(x)}.$$

In particular,  $|\nabla d_\Omega(x)| = 1$  wherever it makes sense.

- If  $\Omega$  is of class  $\mathcal{C}^1$ , this last quantity equals  $\nabla d_\Omega(x) = n(p_{\partial\Omega}(x))$ .
- If  $\Omega$  is of class  $\mathcal{C}^k$ ,  $k \geq 2$ , then  $d_\Omega$  is also of class  $\mathcal{C}^k$  on a neighborhood of  $\partial\Omega$ .



## Signed distance function and geometry (IV)



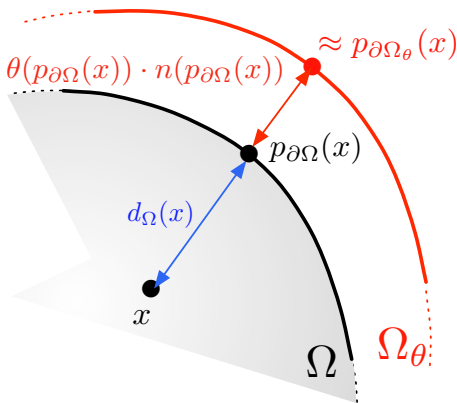
*Some level sets of  $d_\Omega$  are depicted in color;  $d_\Omega$  is as smooth as the boundary  $\partial\Omega$  on the shaded area (at least).*

# Shape differentiability of the signed distance function (I)

## Lemma 3.

Let  $\Omega \subset \mathbb{R}^d$  be a  $\mathcal{C}^1$  bounded domain, and  $x \notin \Sigma$ . The function  $\theta \mapsto d_{\Omega_\theta}(x)$ , from  $W^{1,\infty}(\mathbb{R}^d, \mathbb{R}^d)$  into  $\mathbb{R}$  is Gâteaux-differentiable at  $\theta = 0$ , with derivative:

$$d'_\Omega(\theta)(x) = -\theta(p_{\partial\Omega}(x)) \cdot n(p_{\partial\Omega}(x)).$$

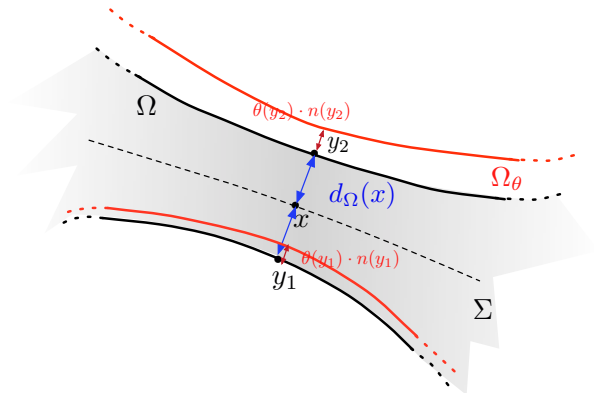


## Shape differentiability of the signed distance function (II)

**Remark:** A more general formula holds, which encompasses the case  $x \in \Sigma$ :

$$\text{If } x \in \Omega, \quad d'_\Omega(\theta)(x) = - \inf_{y \in \Pi_{\partial\Omega}(x)} \theta(y) \cdot n(y),$$

$$\text{If } x \in {}^c\overline{\Omega}, \quad d'_\Omega(\theta)(x) = - \sup_{y \in \Pi_{\partial\Omega}(x)} \theta(y) \cdot n(y).$$



## Shape differentiability of the signed distance function (II)

- **Formal clue:** Taking the shape derivative in

$$|\nabla d_{\Omega}(x)|^2 = 1$$

yields:

$$\nabla d'_{\Omega}(\theta)(x) \cdot \nabla d_{\Omega}(x) = 0.$$

$\Rightarrow$  The shape derivative of  $d_{\Omega}$  is constant along the rays.

- **Rigorous proof:** Use of the definition:

$$d_{\Omega}^2(x) = \min_{y \in \partial\Omega} |x - y|^2$$

in combination to a theorem for differentiating a minimum value with respect to a parameter.

## Shape differentiability of the signed distance function (III)

### Lemma 4.

Let  $\Omega$  be a  $\mathcal{C}^1$  bounded domain, enclosed in a large computational domain  $D$ , and  $j : \mathbb{R}_x \times \mathbb{R}_s \rightarrow \mathbb{R}$  be of class  $\mathcal{C}^1$ ; define the functional:

$$J(\Omega) = \int_D j(x, d_\Omega(x)) \, dx.$$

Then  $\theta \mapsto J(\Omega_\theta)$  is Gâteaux-differentiable at  $\theta = 0$  with derivative:

$$J'(\Omega)(\theta) = - \int_D \frac{\partial j}{\partial s}(x, d_\Omega(x)) \, \theta(p_{\partial\Omega}(x)) \cdot n(p_{\partial\Omega}(x)) \, dx.$$

This formula is awkward insofar it is not easily put under the form:

$$J'(\Omega)(\theta) = \int_\Gamma v \, \theta \cdot n \, ds,$$

and does not lend itself to the inference of a 'natural' descent direction for  $J$ .

## A coarea formula

### Proposition 5.

Let  $\Omega \subset D$  be a bounded domain of class  $\mathcal{C}^2$ , and let  $\varphi \in L^1(D)$ . Then,

$$\int_D \varphi(x) dx = \int_{\partial\Omega} \left( \int_{\text{ray}_{\partial\Omega}(y) \cap D} \varphi(z) \prod_{i=1}^{d-1} (1 + d_\Omega(z) \kappa_i(y)) dz \right) dy,$$

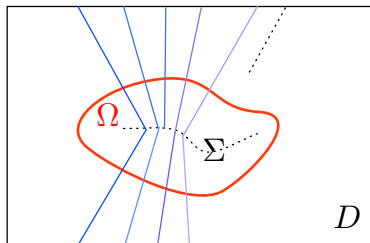
where  $z$  denotes a point in the ray emerging from  $y \in \partial\Omega$  and  $dz$  is the line integration along that ray.

Hint of proof:

Apply the **coarea formula** to the mapping:

$$p_{\partial\Omega} : D \setminus \Sigma \rightarrow \partial\Omega$$

to recast the integration over  $D \approx D \setminus \Sigma$  as an integration over  $\partial\Omega$  composed with an integration over the pre-images  $p_{\partial\Omega}^{-1}(x) = \text{ray}_{\partial\Omega}(x)$ ,  $x \in \partial\Omega$ .



- ① Introduction and definitions
  - Foreword
  - A short reminder of 'classical' shape optimization in linear elasticity
- ② The (exact) sharp-interface multi-phase problem
- ③ The smoothed-interface approach
  - Setting of the smoothed-interface approach
  - A digression around the signed distance function
  - Shape derivatives in the smoothed-interface setting
- ④ Numerical study
  - Presentation of the numerical algorithm
  - Numerical examples

# Shape derivative of the smoothed-interface functional (I)

## Theorem 6.

*The objective function*

$$J_\varepsilon(\Omega^0) = \int_D j(x, u_{\Omega^0, \varepsilon}) dx + \int_{\Gamma_N} k(x, u_{\Omega^0, \varepsilon}) ds,$$

is s.t.  $\theta \mapsto J_\varepsilon(\Omega_\theta^0)$  admits a *Gâteaux-derivative* at  $\theta = 0$ , which is

$$\forall \theta \in W^{1,\infty}(D, \mathbb{R}^d), \quad J'_\varepsilon(\Omega^0)(\theta) = - \int_\Gamma j(x) \theta(x) \cdot n(x) ds(x).$$

Here,  $n$  is the outer unit normal to  $\Omega^0$  and  $j$  is the scalar function defined by

$$j(x) = \int_{\text{ray}_\Gamma(x) \cap D} h'_\varepsilon(d_{\Omega^0}(z)) (A_1 - A_0) e(u)(z) : e(p)(z) \prod_{i=1}^{d-1} (1 + d_{\Omega^0}(z) \kappa_i(x)) dz,$$

where  $u \equiv u_{\Omega^0, \varepsilon}$  and the *adjoint state*  $p \equiv p_{\Omega^0, \varepsilon}$  is the solution to:

$$\begin{cases} -\text{div}(A_{\Omega^0, \varepsilon} e(p)) &= -j'(x, u_{\Omega^0, \varepsilon}) & \text{in } D, \\ p &= 0 & \text{on } \Gamma_D, \\ (A_1 e(p)) n &= -k'(x, u_{\Omega^0, \varepsilon}) & \text{on } \Gamma_N, \end{cases}$$



## Shape derivative of the smoothed-interface functional (II)

Sketch of (formal) proof: For functions  $v, q \in H_{\Gamma_D}^1(D)^d$ , define the **Lagrangian functional**  $\mathcal{L}(\Omega^0, v, q)$  as:

$$\begin{aligned}\mathcal{L}(\Omega^0, v, q) = & \int_D j(x, v) \, dx + \int_{\Gamma_N} k(x, v) \, ds \\ & + \int_D A_{\Omega^0, \varepsilon}(x) e(v) : e(q) \, dx - \int_D f \cdot q \, dx - \int_{\Gamma_N} g \cdot q \, ds.\end{aligned}$$

By definition,

$$\forall q \in H_{\Gamma_D}^1(D)^d, \quad J_\varepsilon(\Omega^0) = \mathcal{L}(\Omega^0, u_{\Omega^0, \varepsilon}, q).$$

Let us search for the **critical points**  $(u, p)$  of  $\mathcal{L}(\Omega^0, \cdot, \cdot)$ .

- Expressing  $\frac{\partial \mathcal{L}}{\partial p}(\Omega^0, u, p) = 0$  yields  $u = u_{\Omega^0, \varepsilon}$ .
- Expressing  $\frac{\partial \mathcal{L}}{\partial u}(\Omega^0, u, p) = 0$  yields  $p = p_{\Omega^0, \varepsilon}$ .

## Shape derivative of the smoothed-interface functional (III)

Thus, for any  $q \in H_{\Gamma_D}^1(D)^d$ , assuming that  $u_{\Omega^0, \varepsilon}$  is differentiable with respect to the domain,

$$J'_\varepsilon(\Omega^0)(\theta) = \frac{\partial \mathcal{L}}{\partial \Omega}(\Omega^0, u_{\Omega^0, \varepsilon}, q) + \frac{\partial \mathcal{L}}{\partial u}(\Omega^0, u_{\Omega^0, \varepsilon}, q)(u'_{\Omega^0, \varepsilon}(\theta)).$$

Now choosing  $q = p_{\Omega^0, \varepsilon}$ , and using  $\frac{\partial \mathcal{L}}{\partial u}(\Omega^0, u, p) = 0$  yield:

$$J'_\varepsilon(\Omega^0)(\theta) = \frac{\partial \mathcal{L}}{\partial \Omega}(\Omega^0, u_{\Omega^0, \varepsilon}, p_{\Omega^0, \varepsilon}),$$

which can be calculated thanks to Lemma 4.



## Approximate formulae

The formula of Theorem 6 can be given consistent and convenient approximations in two important limits in applications:

- Jacobian-free formula: If the interface  $\Gamma$  is **approximately plane**, that is  $d_{\Omega^0} \kappa_i \approx 0$ , we obtain:

$$J'_\varepsilon(\Omega^0)(\theta) = - \int_\Gamma j(x) \theta(x) \cdot n(x) ds(x),$$

with

$$j(x) \approx \int_{\text{ray}_\Gamma(x) \cap D} h'_\varepsilon(d_{\Omega^0}(z)) (A_1 - A_0) e(u)(z) : e(p)(z) dz.$$

- Thin-interface formula: If the transition layer is **very thin**, i.e.  $\varepsilon$  is very small,

$$J'_\varepsilon(\Omega^0)(\theta) \approx - \int_\Gamma (A_1 - A_0) e(u)(x) : e(p)(x) \theta(x) \cdot n(x) ds(x).$$

# Consistency of the smoothed-interface approach

## Theorem 7.

Assume that  $\Omega^0$  is of class  $\mathcal{C}^2$ . Then the smoothed-interface problem *converges* to its sharp-interface counterpart in the sense that:

$$J_\varepsilon(\Omega^0) \xrightarrow{\varepsilon \rightarrow 0} J(\Omega^0),$$

and, for any deformation field  $\theta \in W^{1,\infty}(\mathbb{R}^d, \mathbb{R}^d)$ ,

$$J'_\varepsilon(\Omega^0)(\theta) \xrightarrow{\varepsilon \rightarrow 0} J'(\Omega^0)(\theta).$$

- ① Introduction and definitions
  - Foreword
  - A short reminder of ‘classical’ shape optimization in linear elasticity
- ② The (exact) sharp-interface multi-phase problem
- ③ The smoothed-interface approach
  - Setting of the smoothed-interface approach
  - A digression around the signed distance function
  - Shape derivatives in the smoothed-interface setting
- ④ Numerical study
  - Presentation of the numerical algorithm
  - Numerical examples

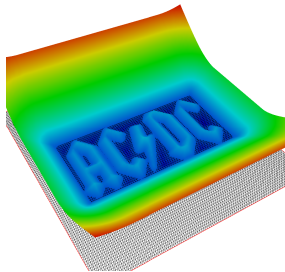
- ① Introduction and definitions
  - Foreword
  - A short reminder of 'classical' shape optimization in linear elasticity
- ② The (exact) sharp-interface multi-phase problem
- ③ The smoothed-interface approach
  - Setting of the smoothed-interface approach
  - A digression around the signed distance function
  - Shape derivatives in the smoothed-interface setting
- ④ Numerical study
  - Presentation of the numerical algorithm
  - Numerical examples

# The Level Set Method

**A paradigm:** [OSe] *the motion of an evolving domain is best described in an **implicit** way.*

A bounded domain  $\Omega \subset \mathbb{R}^d$  is equivalently defined by a function  $\phi : \mathbb{R}^d \rightarrow \mathbb{R}$  such that:

$$\phi(x) < 0 \quad \text{if } x \in \Omega \quad ; \quad \phi(x) = 0 \quad \text{if } x \in \partial\Omega \quad ; \quad \phi(x) > 0 \quad \text{if } x \in \mathbb{R}^d \setminus \bar{\Omega}$$



A bounded domain  $\Omega \subset \mathbb{R}^2$  (left); graph of an associated level set function (right).

# Surface evolution equations in the level set framework

The motion of an evolving domain  $\Omega(t) \subset \mathbb{R}^d$  along a velocity field  $v(t, x) \in \mathbb{R}^d$  translates in terms of an associated 'level set function'  $\phi(t, \cdot)$  into the **level set advection equation**:

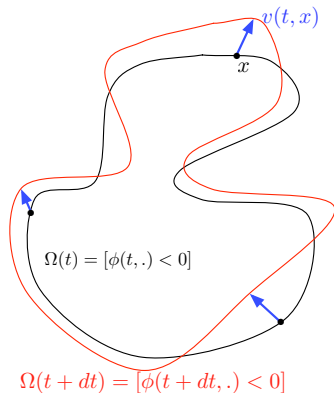
$$\forall t, \forall x \in \mathbb{R}^d, \frac{\partial \phi}{\partial t}(t, x) + v(t, x) \cdot \nabla \phi(t, x) = 0$$

In many applications, the velocity  $v(t, x)$  is normal to the boundary  $\partial\Omega(t)$ :

$$v(t, x) := V(t, x) \frac{\nabla \phi(t, x)}{|\nabla \phi(t, x)|}.$$

Then the evolution equation rewrites as a **Hamilton-Jacobi equation**:

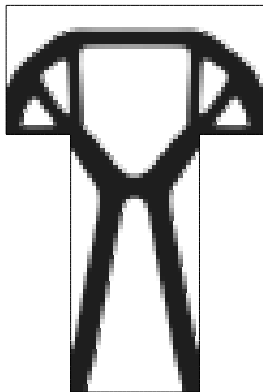
$$\forall t, \forall x \in \mathbb{R}^d, \frac{\partial \phi}{\partial t}(t, x) + V(t, x) |\nabla \phi(t, x)| = 0$$





# The level set method for shape optimization [AlJouToa]

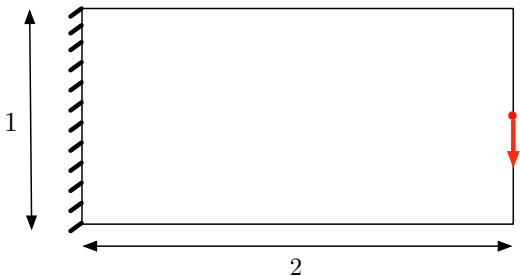
- The shapes  $\Omega^n$  under evolution are embedded in a working domain  $D$  equipped with a **fixed** mesh.
- The successive shapes  $\Omega^n$  are accounted for in the **level set** framework, i.e. via a function  $\phi^n : D \rightarrow \mathbb{R}$  which **implicitly** defines them.
- This approach is very versatile and does not require a mesh of the shapes at each iteration.



*Shape accounted for with a level set description*

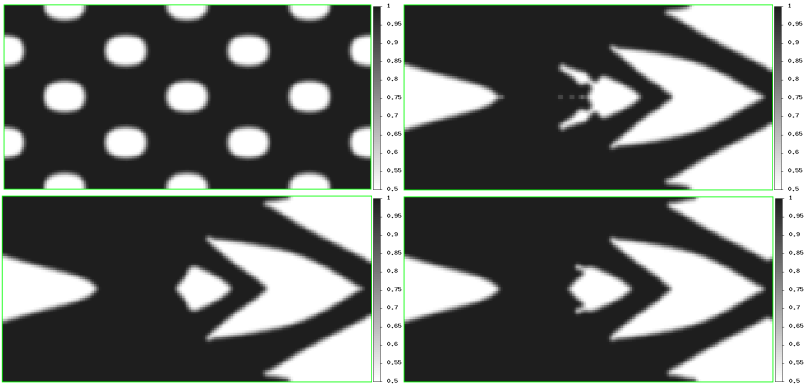
- ① Introduction and definitions
  - Foreword
  - A short reminder of 'classical' shape optimization in linear elasticity
- ② The (exact) sharp-interface multi-phase problem
- ③ The smoothed-interface approach
  - Setting of the smoothed-interface approach
  - A digression around the signed distance function
  - Shape derivatives in the smoothed-interface setting
- ④ Numerical study
  - Presentation of the numerical algorithm
  - Numerical examples

## Two-phase long cantilever (I)



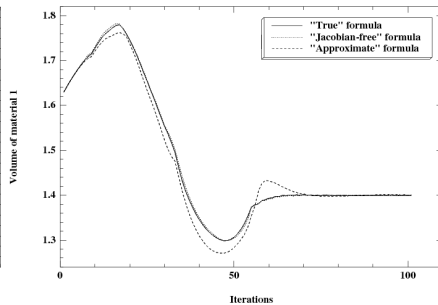
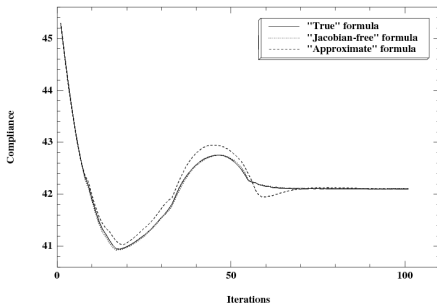
- Optimization of the repartition of two materials with the same Poisson ratio  $\nu_0 = \nu_1 = 0.3$ , but different Young's moduli  $E_0 = 0.5$ ,  $E_1 = 1$ .
- The thickness parameter  $\varepsilon$  is set to  $2\Delta_x$ .
- The **compliance** of the total structure  $D$  is minimized.
- A constraint is imposed on the volume of the stronger phase:  $V_T = 0.7|D|$ , owing to an **augmented Lagrangian algorithm**.

## Two-phase long cantilever (II)



*Initial shape, optimized shape using the 'true' formula, optimized shape using the 'Jacobian-free' formula, optimized shape using the 'thin-interface' formula.*

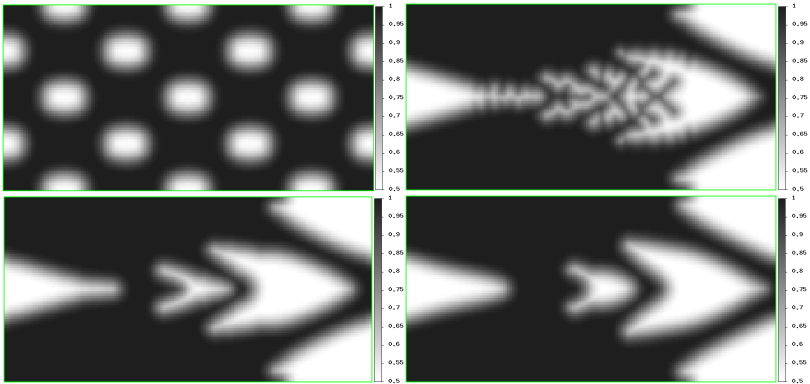
## Two-phase long cantilever (III)



*Convergence histories in the three cases of interest.*

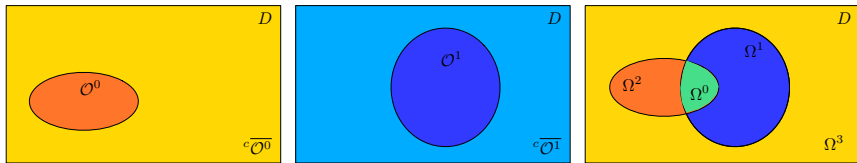
## Two-phase long cantilever (IV)

Use of a larger thickness parameter  $\varepsilon = 8\Delta x$  for the transition zone.



*Initial shape, optimized shape using the 'true' formula, optimized shape using the 'Jacobian-free' formula, optimized shape using the 'thin-interface' formula.*

## Extension to more than 2 (e.g. 3, 4) phases



Two subdomains  $\mathcal{O}^0, \mathcal{O}^1 \subset D$ , and the 4 phases derived by combining them.

- One subdomain  $\mathcal{O}_0 \subset D$  accounts for two phases  $\Omega^0 = \mathcal{O}^0$ ,  $\Omega^1 = {}^c\overline{\mathcal{O}^0}$ .
- Combining 2 subdomains  $\mathcal{O}^0, \mathcal{O}^1 \subset D$ , one can represent up to 4 phases:

$$\Omega^0 = \mathcal{O}^0 \cap \mathcal{O}^1, \quad \Omega^1 = {}^c\overline{\mathcal{O}^0} \cap \mathcal{O}^1, \quad \Omega^2 = \dots$$

- The previous framework can be easily extended to deal with multiple phases:

$\Rightarrow$  Using  $m$  different level set functions allows to account for up to  $2^m$  distinct phases.

## Multiple-phase short cantilever

- Two phases and void:

The Young's moduli of the different phases are:

$$E_0 = 0.5, E_2 = 1, E_1 = E_3 = 1e^{-3}.$$

(Phases 1 and 3 mimick void).

Volume constraint:

$$V_0 = 0.2|D|, V_2 = 0.1|D|.$$

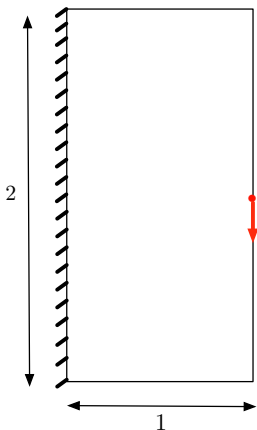
- Three phases and void:

The Young's moduli are:

$$E_0 = 0.5, E_1 = 0.25, E_2 = 1, E_3 = 1e^{-3}.$$

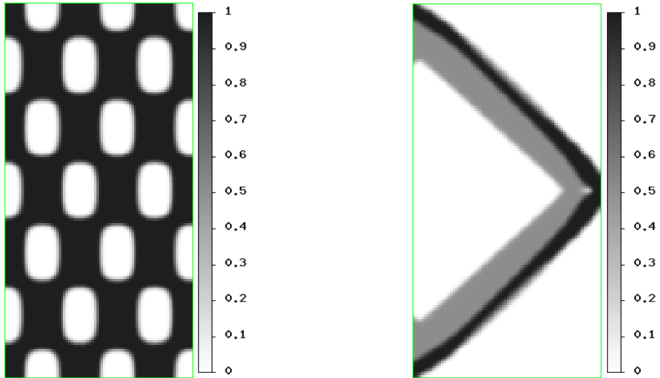
Volume constraint:

$$V_0 = V_1 = V_2 = 0.1|D|.$$



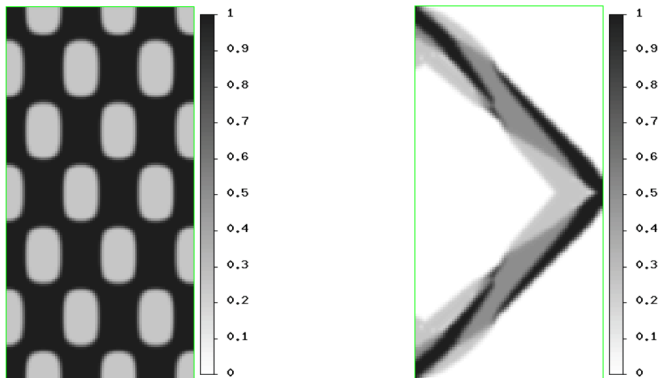


## Two-phase short cantilever



*Short cantilever using two phases and void; (left) initialization, (right) optimal shape.*

## Three-phase short cantilever

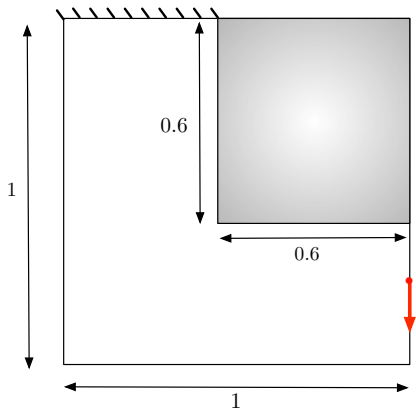


*Short cantilever with three phases and void; (left) initialization, (right) optimal shape.*

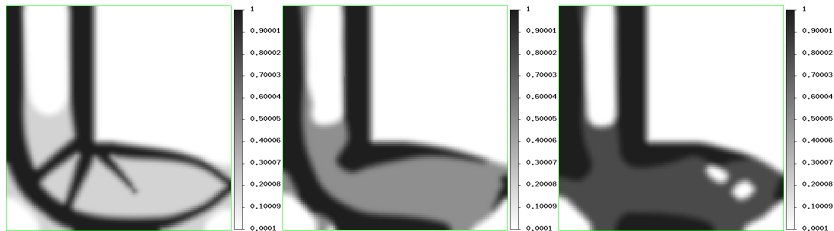
## Two-phase L-Beam

- Phase 0 has Young's modulus  $E_0 = 1$ .
- Phases 1 and 3 mimic void ( $E_1 = E_3 = 1e^{-3}$ ).
- Phase 2 has different Young's moduli depending on the considered example.
- A constraint on the volumes of phases 0 and 2 is imposed:

$$V_T^0 = V_T^2 = 0.25|D|.$$

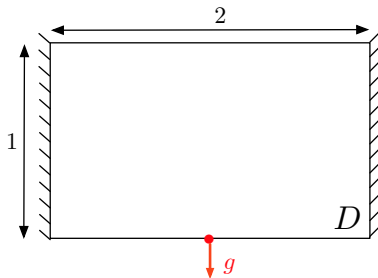


## Two-phase L-Beam



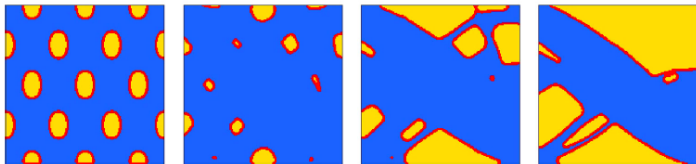
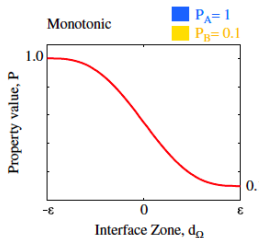
*Optimal designs for the two-phase L-Beam problem with (from left to right)  $E_2 = 0.2, 0.5, 0.8$ .*

## An example using non monotone interfaces



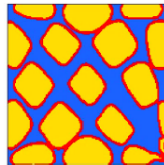
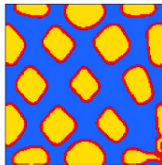
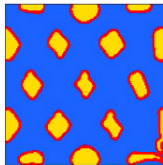
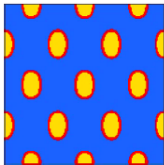
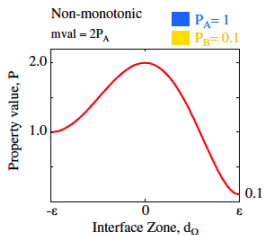
- Work carried out by G. Allaire, Y. Bréchet, R. Estevez, G. Michailidis, G. Parry and N. Vermaak [VerMi].
- Optimization of the repartition of two materials with the same Poisson ratio  $\nu_0 = \nu_1 = 0.3$ , but different Young's moduli  $E_0 = 0.1$ ,  $E_1 = 1$ .
- The **compliance** of the total structure  $D$  is minimized, under a constraint  $V_T^1 = 0.5|D|$  on the volume of the stronger phase.
- The properties of the material inside the transition layer are **non monotone**.

# An example using non monotone interfaces



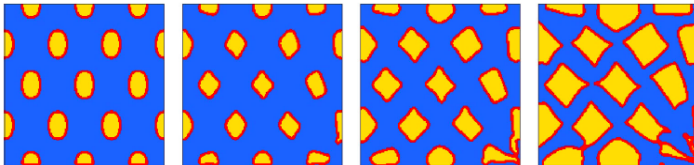
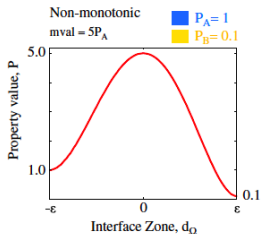
(Top-left) Profile of the Young's modulus in the transition layer, (top-right) final design, (bottom) iterations 1, 10, 25, 40.

# An example using non monotone interfaces



(Top-left) Profile of the Young's modulus in the transition layer, (top-right) final design, (bottom) iterations 1, 50, 75, 90.

# An example using non monotone interfaces







(Top-left) Profile of the Young's modulus in the transition layer, (top-right) final design, (bottom) iterations 1, 5, 50, 110.







Thank you !

Thank you for your attention!

## References I

-  [AIDaDeLMi] G. Allaire, C. Dapogny, G. Delgado and G. Michailidis, *Multi-phase optimization via a level set method*, ESAIM: Control, Optimisation and Calculus of Variations, vol 20, (2014) pp. 576–611
-  [AIDaFr] G. Allaire, C. Dapogny and P. Frey, *Shape optimization with a level set based mesh evolution method*, Comp. Meth. Appl. Mech. Eng., 281, (2014), pp. 22–53.
-  [AlJouToa] G. Allaire, F. Jouve and A.M. Toader, *Structural optimization using shape sensitivity analysis and a level-set method*, J. Comput. Phys., 194 (2004) pp. 363–393.
-  [AlJouVG] G. Allaire, F. Jouve and N. Van Goethem, *Damage evolution in brittle materials by shape and topological sensitivity analysis*, J. Comput. Phys., 230 (2011) pp. 5010–5044.

## References II

-  [DelZol] M. C. Delfour and J.-P. Zolesio, *Oriented distance function and its evolution equation for initial sets with thin boundary*, SIAM J. Control Optim. 42, No 6 (2004) pp. 2286–2304.
-  [Mi] G. Michailidis, *Thèse de l'École Polytechnique*
-  [OSe] S. J. Osher and J.A. Sethian, *Front propagating with curvature dependent speed: algorithms based on Hamilton-Jacobi formulations*, J. Comp. Phys. **78** (1988) pp. 12-49
-  [VerMi] N. Vermaak, G. Michailidis, G. Parry, R. Estevez, G. Allaire, Y. Bréchet, *Material interface effects on the topology optimization of multi-phase structures using a level set method*, Struct. Multidisc. Optim., 50, (2014) pp. 623–644.