

Context and setting

Shape optimization problems arise in many engineering applications where the objective is to design optimal structures. In this context, the level-set approach have become a widely used tool to describe evolving complex geometries during the optimization process (see [1]).

In this work, we couple the level-set framework with φ -FEM, an unfitted Finite Element Method based on an implicate description of the geometry of the domain by a level-set (see [3]). Such FEM scheme avoids remeshing as the domain evolves, making it particularly well suited for shape optimization problems.

We focus on the minimization of the compliance functional for a elastic cantilever (see Figure 1). We use a shape optimization algorithm based on the level-set method and the gradient descent, working with FormOpt (see [4]). Our long-term objective is to develop efficient numerical tools for studies in three-dimensions and design of patient-specific angioplastic balloons.

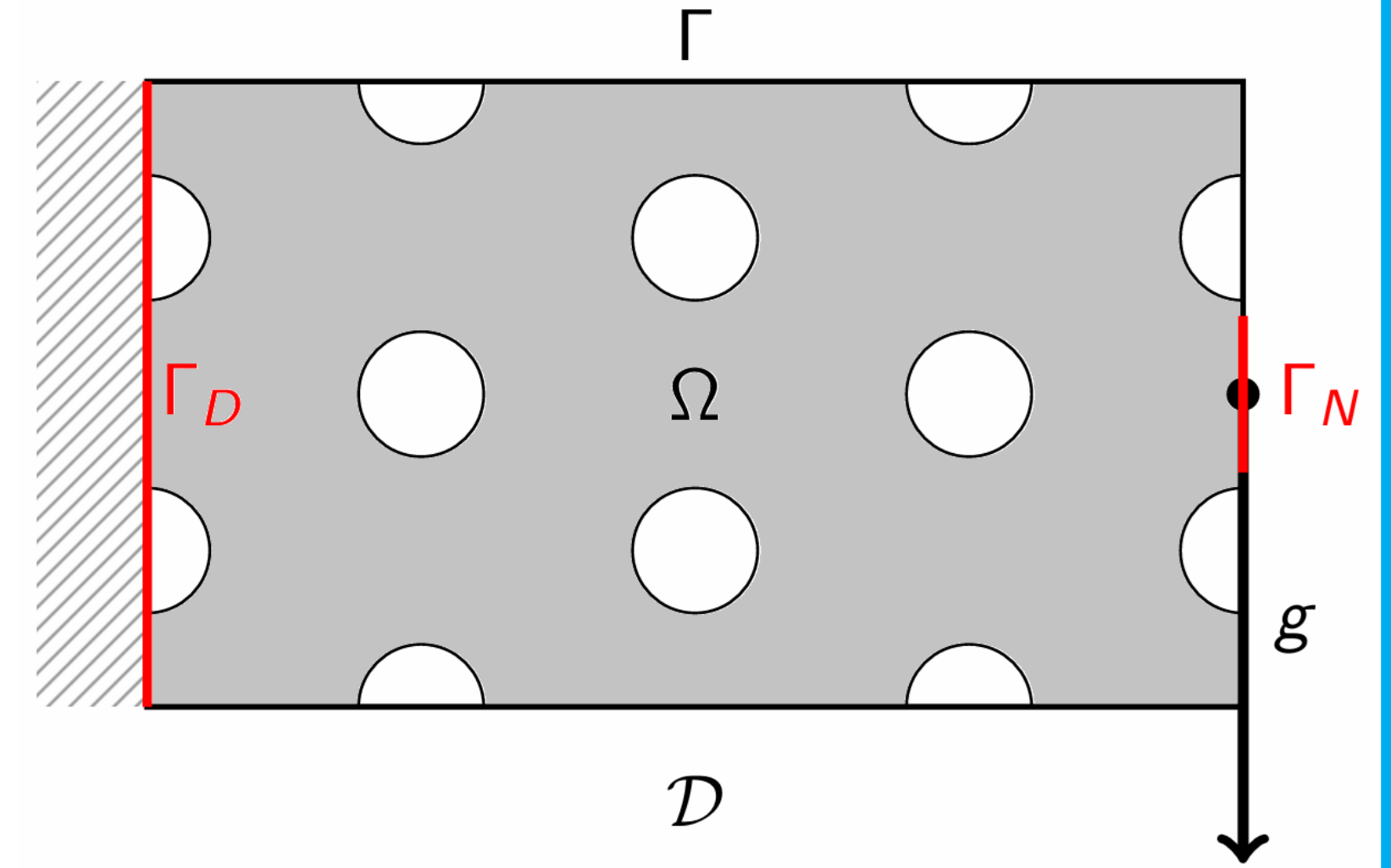


Fig. 1: The cantilever $\Omega \subset \mathcal{D} \subset \mathbb{R}^2$.
 $\Gamma := \partial\Omega \setminus (\Gamma_D \cup \Gamma_N)$.

I. Shape optimization problem

The displacement u satisfies:

$$-\operatorname{div}(\sigma(u)) = 0 \quad \text{in } \Omega, \quad (1)$$

$$u = 0 \quad \text{on } \Gamma_D, \quad (2)$$

$$\sigma(u)n = g \quad \text{on } \Gamma_N, \quad (3)$$

$$\sigma(u)n = 0 \quad \text{on } \Gamma, \quad (4)$$

with:

- $g \in L^2(\Gamma_N)^2$ a given traction force on Γ_N ;

- n the exterior unit normal to $\partial\Omega$;

- $\sigma(u) := 2\mu\varepsilon(u) + \lambda\operatorname{div}(u)I$, where λ, μ are the Lamé parameters, $\varepsilon(u) = \frac{1}{2}(\nabla u + \nabla u^T)$ and I the identity matrix of $\mathcal{M}_2(\mathbb{R})$.

Variational formulation:

Find $u_\Omega \in H(\Omega) := \{u \in H^1(\Omega)^2 \mid u|_{\Gamma_D} = 0\}$:

$$\forall v \in H(\Omega), \quad \int_{\Omega} \sigma(u_\Omega) : \varepsilon(v) dx = \int_{\Gamma_N} g \cdot v ds.$$

Shape optimization problem:

$$\min_{\Omega \in \mathcal{O}} J(\Omega),$$

where $J(\Omega) = \int_{\Omega} \sigma(u_\Omega) : \varepsilon(u_\Omega) dx$ and \mathcal{O} the shape admissible set.

The shape derivative of J is (see [4]):

$$dJ_\Omega(\theta) = \int_{\Omega} [2\nabla u_\Omega^T \sigma(u_\Omega) - \sigma(u_\Omega) : \varepsilon(u_\Omega)I] : \nabla \theta.$$

Remark. We penalize the volume by adding the term $\Lambda|\Omega|$, where Λ acts as a material cost and $|\Omega|$ is the volume of Ω .

II. Shape optimization algorithm

Via the gradient descent and the level-set method: assume $\Omega = \{\varphi < 0\}$, $\partial\Omega = \{\varphi = 0\}$.

1. Elastic problem approximation:

φ -FEM scheme given in Section III.

2. Perturbation computation: find θ s.t.

$$\forall \xi, \quad \mathcal{B}(\theta, \xi) = -dJ_{\Omega_0}(\xi),$$

where \mathcal{B} is a continuous coercive bilinear form. Thus, $dJ_{\Omega_0}(\theta) = -\mathcal{B}(\theta, \theta) < 0$.

See Section IV.

3. Level-set transportation: find ϕ s.t.

$$\partial_t \phi + \theta \cdot \nabla \phi = 0, \quad \phi(0, \cdot) = \varphi.$$

4. Level-set reinitialization: find ψ s.t.

$$\partial_t \psi + \mathcal{S}(\phi)(|\nabla \psi| - 1) = 0, \quad \psi(0, \cdot) = \phi,$$

where \mathcal{S} is the sign function.

III. Resolution of the elastic problem

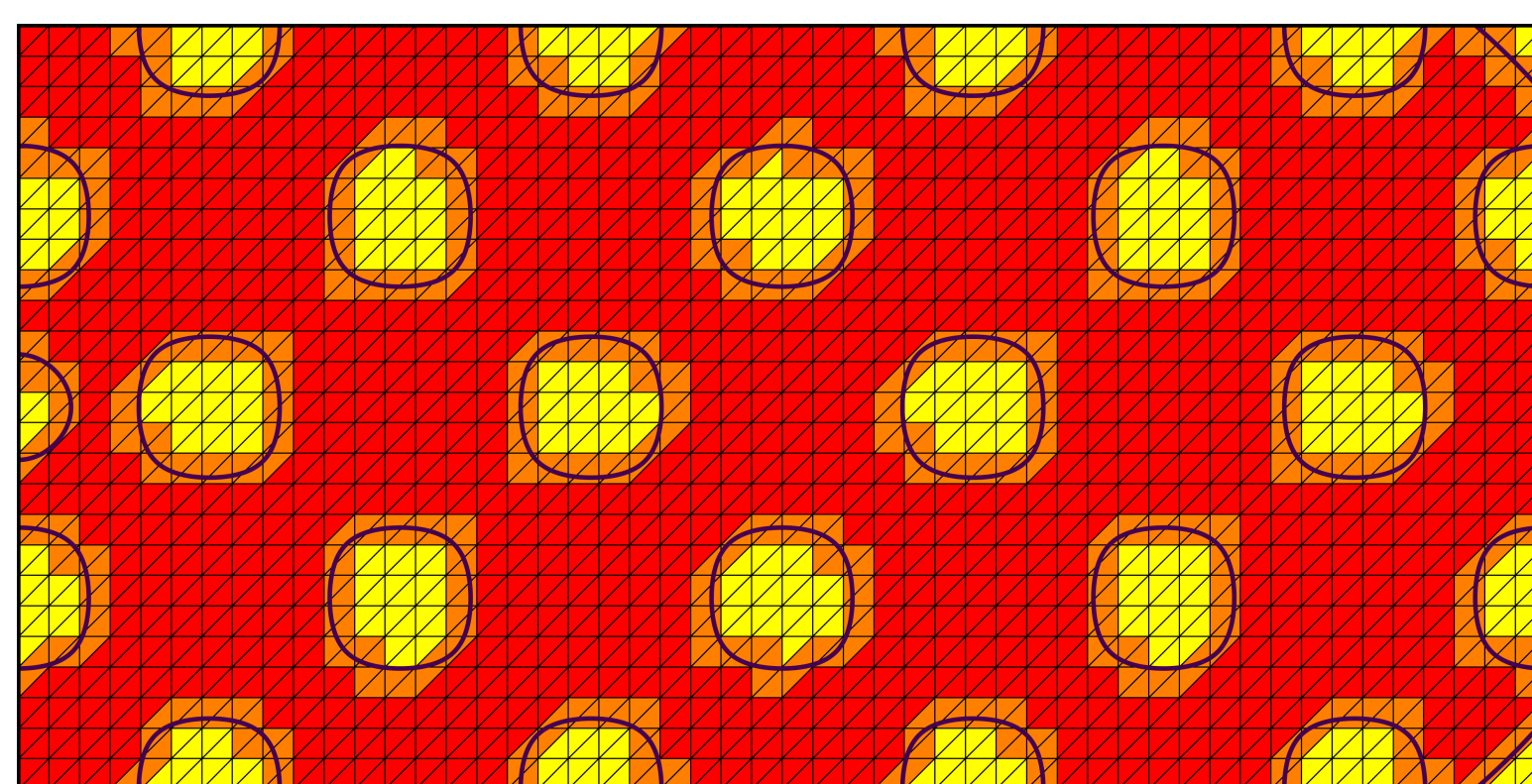


Fig. 2: Ω_h : red + orange | Ω_h^Γ : orange
Initialization mesh.

To solve (1)-(4), we use φ -FEM (Neumann case [3]).

Assume φ is a signed distance function: $n = \frac{\nabla \varphi}{|\nabla \varphi|}$ on Γ .

Formally, we introduce auxiliary variables:

- $y := \sigma(u)$ on Ω_h^Γ .

- Relaxation: $y \nabla \varphi + p \varphi = 0$ on Ω_h^Γ (so that $yn = 0$ on Γ).

Let $\mathcal{F}_{h,1}^\Gamma$ be the facets between orange and red cells.

Goal: find (u_h, y_h, p_h) such that for all (v_h, z_h, q_h) ,

$$\begin{aligned} & \int_{\Omega_h} \sigma(u_h) : \varepsilon(v_h) ds + \int_{\partial\Omega_h} (y_h n) \cdot v_h ds \\ & + \gamma_u \int_{\Omega_h^\Gamma} (y_h + \sigma(u_h)) : (z_h + \sigma(v_h)) dx + \frac{\gamma_p}{h^2} \int_{\Omega_h^\Gamma} \left(y_h \cdot \nabla \varphi_h + \frac{1}{h} p_h \varphi_h \right) \cdot \left(z_h \cdot \nabla \varphi_h + \frac{1}{h} q_h \varphi_h \right) dx \\ & + \gamma_{div} \int_{\Omega_h^\Gamma} \operatorname{div}(y_h) \operatorname{div}(z_h) dx + \sigma h \sum_{F \in \mathcal{F}_{h,1}^\Gamma} \int_F [\sigma(u_h) n_F] [\sigma(v_h) n_F] ds = \int_{\Gamma_N} g \cdot v_h ds. \end{aligned}$$

Classic terms | Penalization | Stabilization | Stabilization (Ghost penalty)

IV. Resolution of the perturbation problem

When u_h is obtained, we compute dJ_Ω . This part is based on gradient descent.

Goal: find $\theta \in \Theta := \{\theta \in W^{1,\infty}(\mathcal{D})^2, \theta|_{\Gamma_D \cup \Gamma_N} = 0\}$ such that

$$\forall \xi \in \Theta, \quad \mathcal{B}(\theta, \xi) = -dJ_{\Omega_0}(\xi),$$

where \mathcal{B} is a continuous coercive bilinear form. We take $\mathcal{B}(\theta, \xi) = \int_{\Omega} \left(\nabla \theta : \nabla \xi + \frac{1}{10} \theta \cdot \xi \right) dx$.

Thus, taking $\xi = \theta$, $dJ_{\Omega_0}(\theta) = -\mathcal{B}(\theta, \theta) < 0$ (coercivity of \mathcal{B}).

To solve this perturbation problem, we use classical FEM (unfitted).

V. Results and future work

We take $\mathcal{D} = [0, 2] \times [0, 1]$, $\lambda \approx 0.5769$, $\mu \approx 0.3846$, $g = (0, -2)$, $\Lambda = 1$.

We compute the optimal shape Ω_{opt} in Figure 3.

Using φ -FEM for the elastic problem, the optimal value of the functional is $J(\Omega_{\text{opt}}) \approx 3.0486$. Using FEM for both problems (ersatz material for $\bar{\mathcal{D}} \setminus \Omega$, see [4]), we obtain $J(\Omega_{\text{opt}}) \approx 3.1744$.

Future work:

- Implementation of φ -FEM for the perturbation problem (interface problem).
- Validation on other problems (e.g. L-shape, ...).
- Generalizing in three dimensions.
- Final goal: design of patient-specific angioplastic balloons by optimizing their thickness.



Fig. 3: Result of the shape optimization.

Bibliography

- [1] G. Allaire et al. A level-set method for shape optimization. *Comptes Rendus Mathématique*, 2002.
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