

Air Gap Shape Optimization for Minimizing Proximity Effect Losses in an Inductor

Stéphane Gaydier^{1,2,3}, Itaï Zehavi⁴, Théodore Cherrière⁵, Peter Gangl⁶

[1] Schneider Electric, Global Technology Eybens, France

[2] Univ. Grenoble Alpes, CNRS, Grenoble INP, G2ELab, F-38000, Grenoble, France

[3] Univ. Grenoble Alpes, CNRS, Grenoble INP, LJK, 38000 Grenoble, France

[4] Electrical Engineering Department, ENS Paris-Saclay, Université Paris-Saclay, 91190 Gif-sur-Yvette, France

[5] Université Paris-Saclay, CentraleSupélec, Sorbonne Université, CNRS, GeePs, 91192 Gif-sur-Yvette, France

[6] Johann Radon Institute for Computational and Applied Mathematics (RICAM), ÖAW, 4040 Linz, Austria



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Introduction

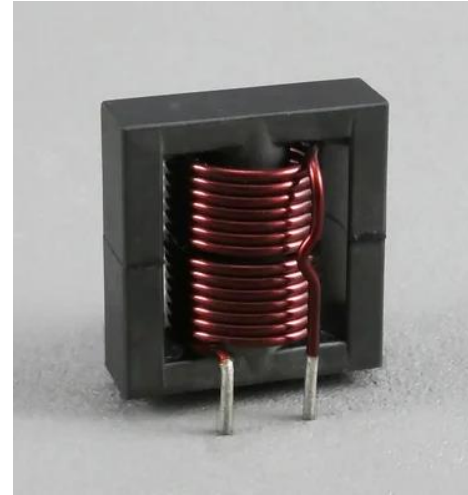
We want to optimize an **inductor** to minimize the AC losses due to the **fringing effect**.

Possible solutions :

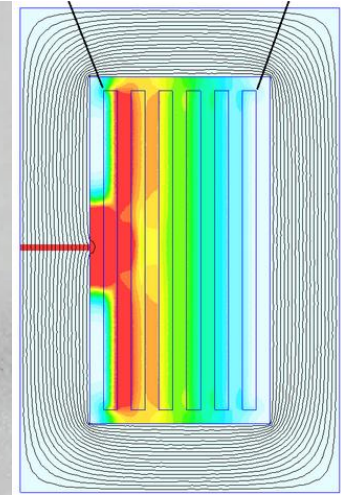
- Distributed air gaps
- Change winding distribution of the conductor [1]
- Optimize air gap profile (e.g. with heuristic methods [2])

Our proposition : We optimize the **air gap profile** using **gradient descent methods**.

- Converge faster
- More flexible parametrization



An inductor [3]



Fringing Effect [4]

[1] R.A. Jensen and C.R. Sullivan, "Optimal core dimensional ratios for minimizing winding loss in high-frequency gapped-inductor windings,"

[2] D.I. Zaikin, S. Jonassen, and S.L. Mikkelsen, "An Air-Gap Shape Optimization for Fringing Field Eddy Current Loss Reductions in Power Magnetics,"

[3] <https://info.triadmagnetics.com/blog/basics-of-inductors>

[4] https://www.e-magnetica.pl/doku.php/proximity_effect

Outline

Introduction

Part I – Optimization Problem

Part II – Methodology

Part III – Optimization Results

Conclusion

Part I – Optimization Problem

Device and modeling

Device: A simplified inductor composed of a coil and a magnetic core

Goal: Minimizing the **AC losses** in the conductor keeping the **inductance** at $L_0 = 1$ mH.

Magneto-harmonic equation: $\boxed{-div(\underline{\nu} \nabla \underline{a}) = j}$,

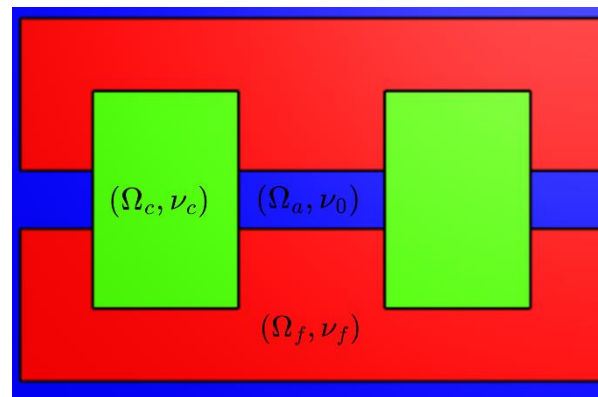
where :

\underline{a} is magnetic vector potential out-of-plane component

j is real current density,

and $\underline{\nu}$ is the reluctivity which takes values :

Reluctivity value (ν)	Subdomain
$\nu_0 = 1/\mu_0$	Air region Ω_a
$\nu_f = \nu_0/1000$	Core region (ferrite) Ω_f
$\boxed{\nu_c = \nu_0 \exp(i\delta)}$ with $\delta = 0.1$ rad to model proximity effects at $f = 50$ kHz	Conductor region (copper) Ω_c



Reference design

Objective and Constraint

The retained formulas are:

AC Losses:

$$P(\underline{a}) = l_z \pi f \int_{\Omega_c} \text{Im}(\underline{v}_c) |\nabla \underline{a}|^2 dx$$

Inductance:

$$L(\underline{a}) = \frac{l_z}{I^2} \int_{\Omega} \text{Re}(\underline{v}) |\nabla \underline{a}|^2 dx$$

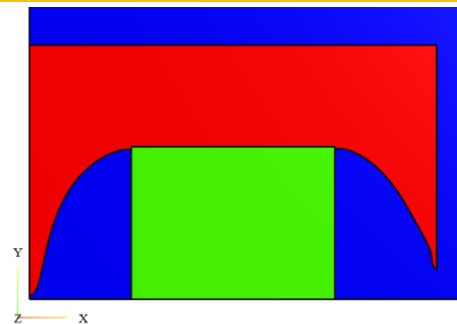
with $l_z = 1 \text{ cm}$ (thickness) and $I = 2 \text{ A}$ (current, 200 turns).

Parametrization

Let h be a **parametrization** of the air gap profile.

The following are dependent on h :

- the partition of the domain $\Omega = \Omega_{a,h} \cup \Omega_{f,h} \cup \Omega_{c,h}$,
- the reluctivity \underline{v}_h ,
- the magnetic potential \underline{a}_h solution of (E_h)



Parametrized air gap

$$\text{Find } \underline{a}_h \in H_0^1(\Omega) \text{ such that } \forall \underline{v} \in H_0^1(\Omega), \int_{\Omega} \underline{v}_h \nabla \underline{a}_h \cdot \nabla \underline{v} \, dx = \int_{\Omega_{c,h}} j \underline{v} \, dx \quad (E_h)$$

Then, the objective and constraint rewrites into :

AC Losses:

$$P(h, \underline{a}_h) = l_z \pi f \int_{\Omega_{c,h}} \text{Im}(\underline{v}_c) |\nabla \underline{a}_h|^2 \, dx$$

Inductance:

$$L(h, \underline{a}_h) = \frac{l_z}{I^2} \int_{\Omega} \text{Re}(\underline{v}_h) |\nabla \underline{a}_h|^2 \, dx$$

Constrained Minimization Problem

Let h be a **parametrization** of the air gap profile.

We aim to solve the following minimization problem to determine the optimal h :

$$\min_h P(h, \underline{a_h}) \text{ subject to } \begin{cases} L(h, \underline{a_h}) = 1 \text{ mH} \\ \underline{a_h} \text{ solution of } (E_h) \end{cases}$$

Challenges and Solutions

Challenges :	Proposed solutions [1]:
Find good parametrizations h of the air gap.	Control point vs Adaptive Mesh deformation.
Differentiate P and L with respect to h ...	Shape derivatives...
...and to the PDE constraint !	...obtained by the adjoint method .
Find descent directions (How to find d such that $\frac{dP}{dh}(h, a_h)d < 0$?) that preserve the conductor domain.	Hilbertian extension-regularization.

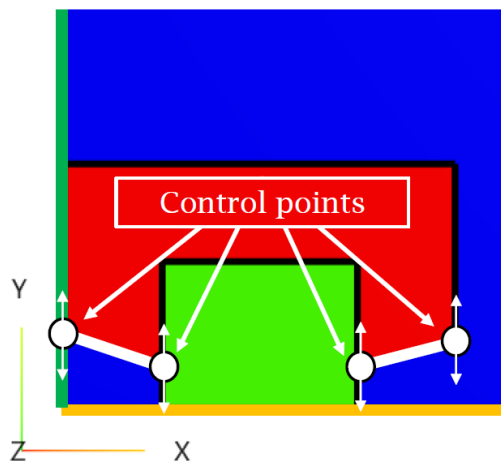
We are using a handwritten **augmented Lagrangian** algorithm to solve this constrained optimization problem.

[1] G. Allaire, C. Dapogny, and F. Jouve, "Shape and topology optimization," in Handbook of Numerical Analysis

Part II - Methodology

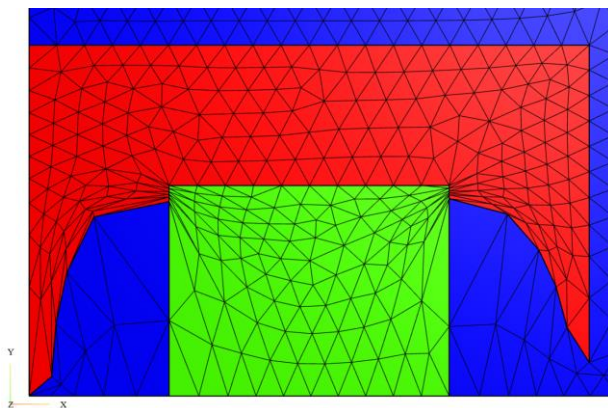
Two Choices for the Parametrization h

$$h = (p_1, p_2, p_3, p_4)$$

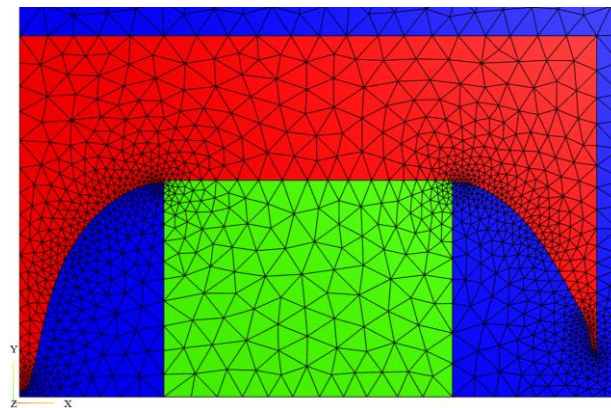


Parametric Shape Optimization using
Control points

$$h = (\Omega_a, \Omega_f, \Omega_c)$$



Geometric Shape Optimization by
mesh deformation



Geometric Shape Optimization
by **adaptive** mesh deformation
using MMG [1]

[1] <https://www.mmgtools.org/>

Shape derivatives

Shape derivative \approx derivative of a functional w.r.t. the deformation of a domain

Considering:

- A bounded Lipschitz domain Ω
- “Small and regular enough” **deformation fields** ϕ
- A real-value function $J(\Omega)$ of the domain

We define the **deformed domain** with respect to ϕ to be $\Omega_\phi = (Id + \phi)(\Omega)$

The **shape derivative** of $J(\Omega)$ is the derivative of $\phi \mapsto J(\Omega_\phi)$ at $\phi = 0$:

$$J(\Omega_\phi) = J(\Omega) + J'(\Omega)(\phi) + o(\phi)$$

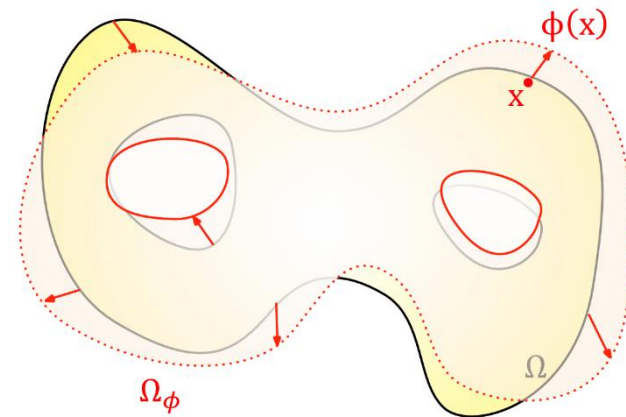
We are searching a ϕ that is a **descent direction**:

$$J'(\Omega)(\phi) < 0$$

Two approaches:

In our “control points” approach, ϕ is defined on the control points

In our “adaptive mesh deformation” approach, ϕ is defined on the mesh nodes.



Deformed domain

Example of a Shape Derivative

For example, if we take a function $J(\Omega) = \int_{\Omega} f dx$,

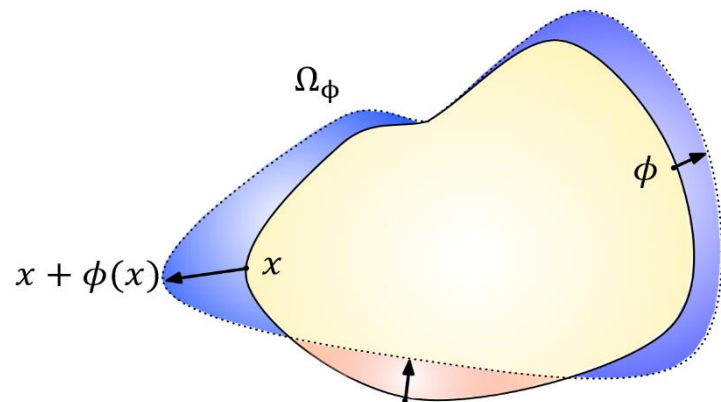
its shape derivative is:

$$J'(\Omega)(\phi) = \int_{\Omega} \operatorname{div}(f\phi) dx = \int_{\partial\Omega} f\phi \cdot n ds$$

This is the **Reynolds transports theorem!**

Finding a descent direction is obvious! For example:

$$\phi = -fn$$



Minimization of the integral.

The deformation $\phi = -fn$ makes Ω shrink in the (red) region where $f(x) > 0$, and expand in the (blue) region where $f(x) < 0$

Shape Derivatives of Losses & Inductance

Recalling the formulas for the AC losses and the inductance :

- Losses : $P(\underline{a}) = l_z \pi f \int_{\Omega_c} \text{Im}(\underline{v}_c) |\nabla \underline{a}|^2 dx$
- Inductance : $L(\underline{a}) = \frac{l_z}{I^2} \int_{\Omega} \text{Re}(\underline{v}) |\nabla \underline{a}|^2 dx$

We introduce $\underline{\lambda}_P$ and $\underline{\lambda}_L$ the **adjoint states** given by the **adjoint equations** :

- $-\text{div}(\underline{v}^* \nabla \underline{\lambda}_P) = -\frac{dP}{d\underline{a}}(\underline{a})$
- $-\text{div}(\underline{v}^* \nabla \underline{\lambda}_L) = -\frac{dL}{d\underline{a}}(\underline{a})$

where the right-hand members are complex Fréchet-derivatives of real functions.

We compute the **shape derivatives** using the **adjoint method** and obtain :

- $P'(\Omega)(\phi) = \text{Re} \left(l_z \pi f \int_{\Omega_c} \text{Im}(\underline{v}_c) A_\phi \nabla \underline{a} \cdot (\nabla \underline{a})^* dx + \int_{\Omega} \underline{v} A_\phi \nabla \underline{a} \cdot (\nabla \underline{\lambda}_P)^* dx - \int_{\Omega_c} j \text{div} \phi \underline{\lambda}_P^* dx \right)$
- $L'(\Omega)(\phi) = \text{Re} \left(\frac{l_z}{I^2} \int_{\Omega} \underline{v} A_\phi \nabla \underline{a} \cdot (\nabla \underline{a})^* dx + \int_{\Omega} \underline{v} A_\phi \nabla \underline{a} \cdot (\nabla \underline{\lambda}_L)^* dx - \int_{\Omega_c} j \text{div} \phi \underline{\lambda}_L^* dx \right)$

with the real 2×2 matrix $A_\phi = \text{div} \phi I_2 - \partial \phi - \partial \phi^\top$

Those derivative formulas don't directly reveal the descent directions; we need another step.

Hilbertian Extension-Regularization

The “Hilbertian extension-regularization” technique consists in solving a Lax-Milgram-type equation of the form:

$$\text{Find } \phi \in V \text{ such that } \forall \hat{\phi} \in V, \quad \langle \phi, \hat{\phi} \rangle_V = J'(\Omega)(\hat{\phi})$$

where :

- V is any subspace of $H^1(\Omega)$ (with wisely defined Dirichlet conditions),
- $\langle \cdot, \cdot \rangle_V$ is any inner product of V .

Then $-\phi$ is a descent direction : $J'(\Omega)(-\phi) = -\|\phi\|_V^2$.

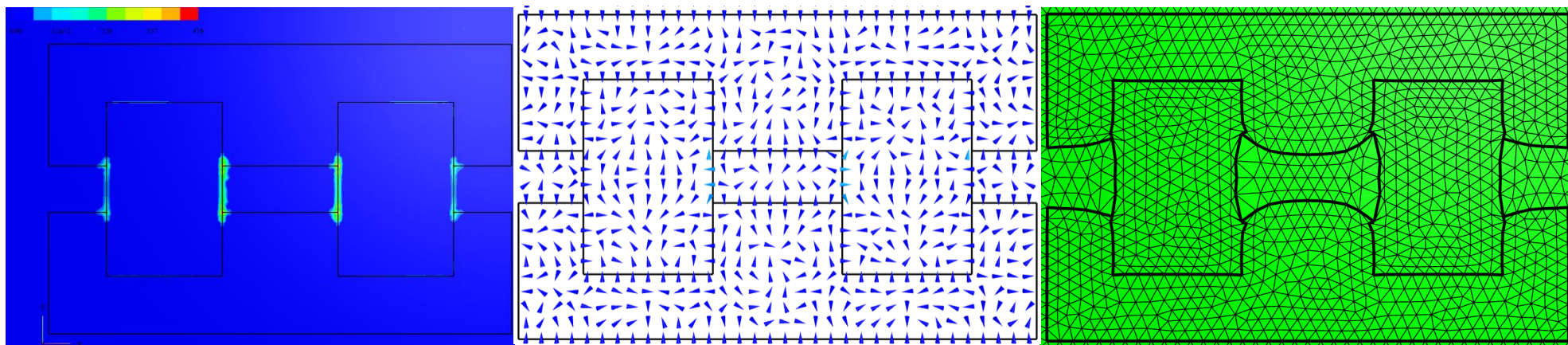
This procedure yields a **regularized descent direction**
with a **structure** that can preserve the conductor's boundary.

Unregularized the Deformation Field

With a deformation field ϕ coming from:

Find $\phi \in L^2(\Omega)$ s. t. $\forall \hat{\phi} \in L^2(\Omega)$,

$$\langle \phi, \hat{\phi} \rangle_{L^2(\Omega)} = J'(\Omega)(\hat{\phi})$$



Norm of $-\phi$

Vector field $-\phi$

Deformed mesh

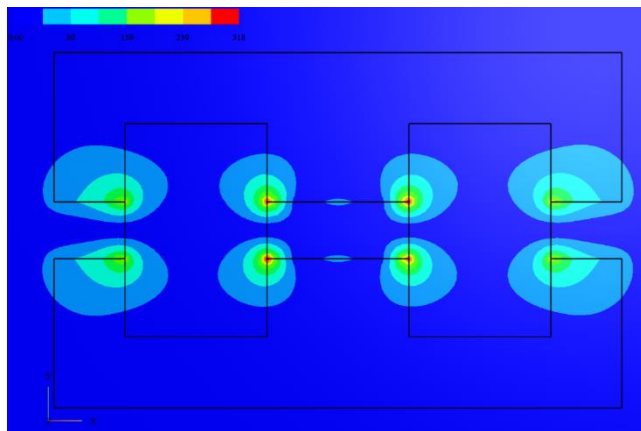
Regularized Deformation Field

With a deformation field ϕ coming from:

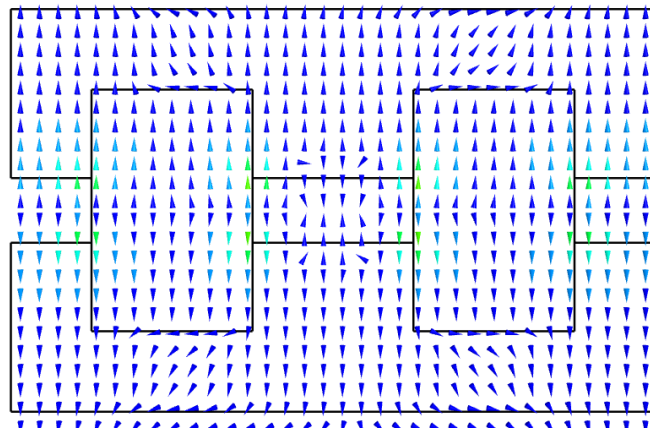
Find $\phi \in V$ s. t. $\forall \hat{\phi} \in V$,

$$\langle \phi, \hat{\phi} \rangle_{L^2(\Omega)} + \langle \nabla \phi, \nabla \hat{\phi} \rangle_{L^2(\Omega)} = J'(\Omega)(\hat{\phi})$$

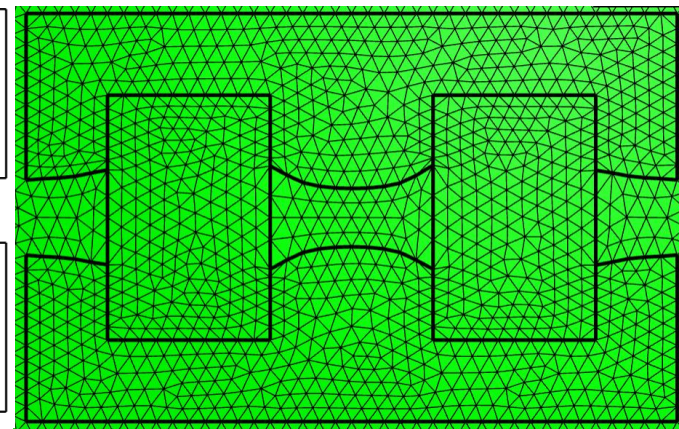
with $V = \{\phi \in H^1(\Omega) \mid \phi \cdot \vec{e}_x = 0 \text{ on } \Gamma_{D,x}, \phi \cdot \vec{e}_y = 0 \text{ on } \Gamma_{D,y}\}$



Norm of $-\phi$

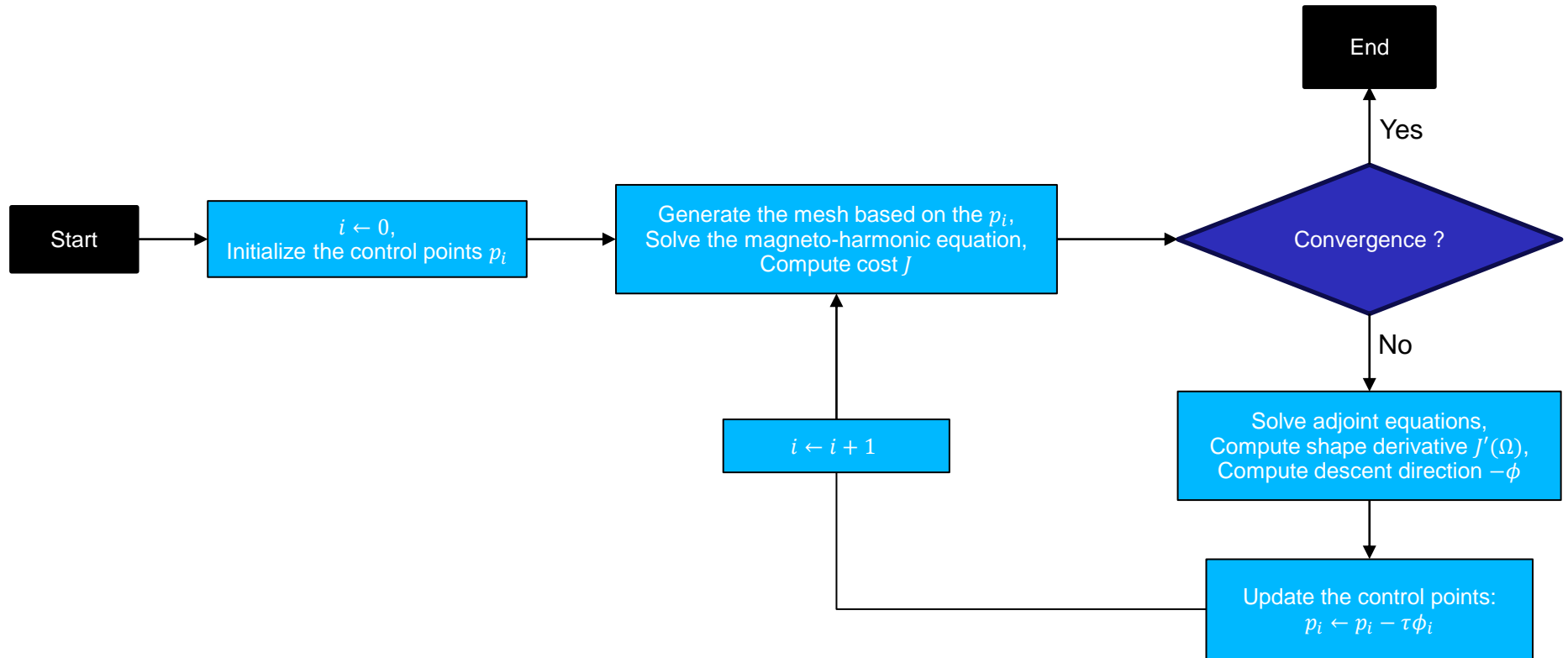


Vector field $-\phi$

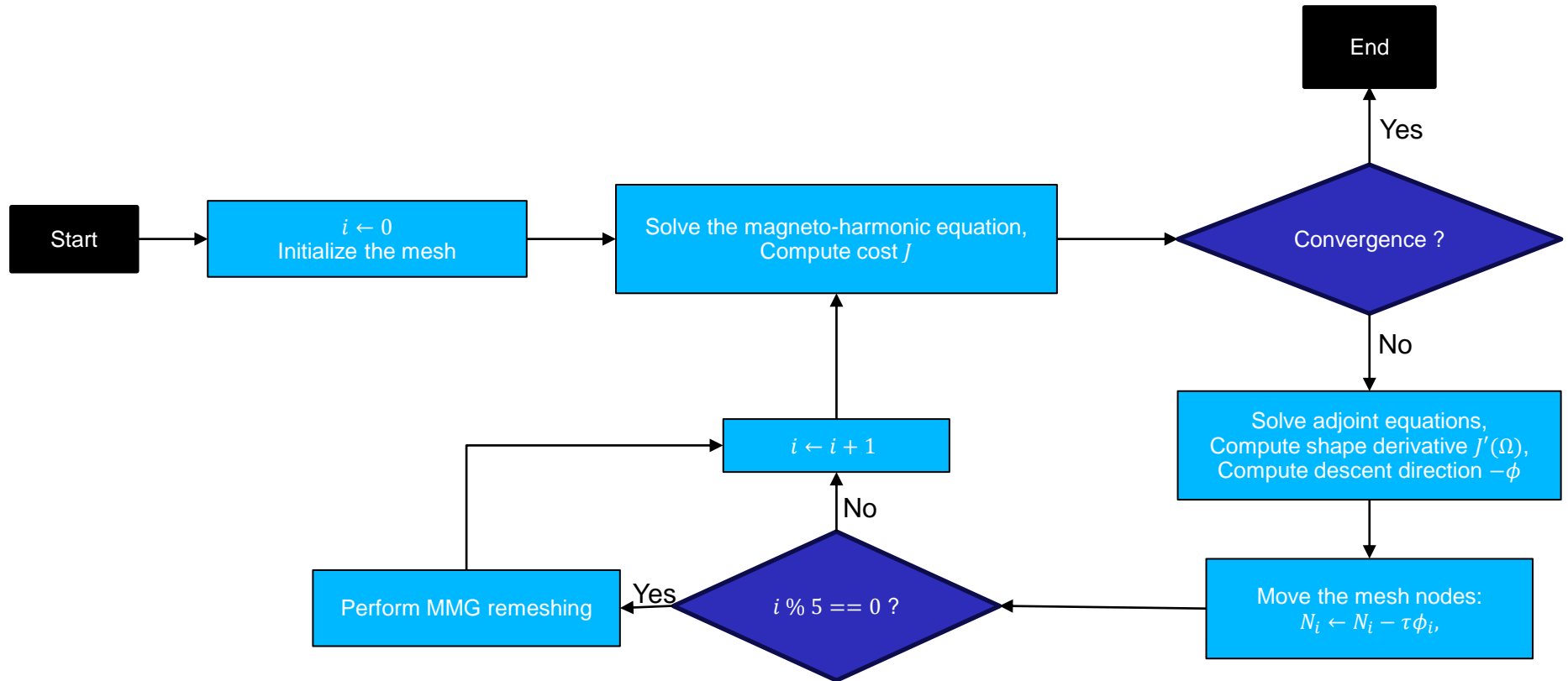


Deformed mesh

Control Points Flowchart



Adaptative Mesh Deformation Flowchart

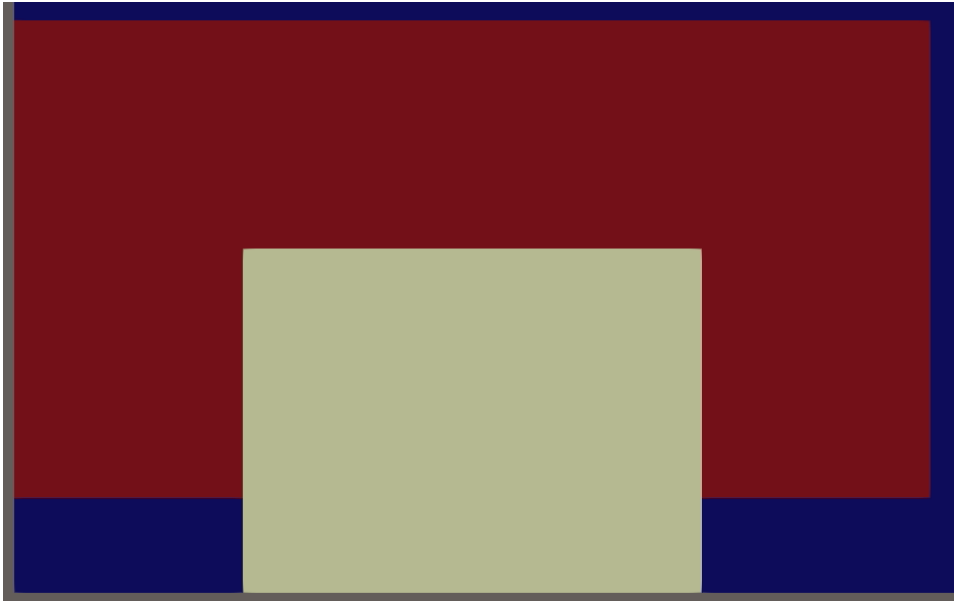


Part III – Optimization Results

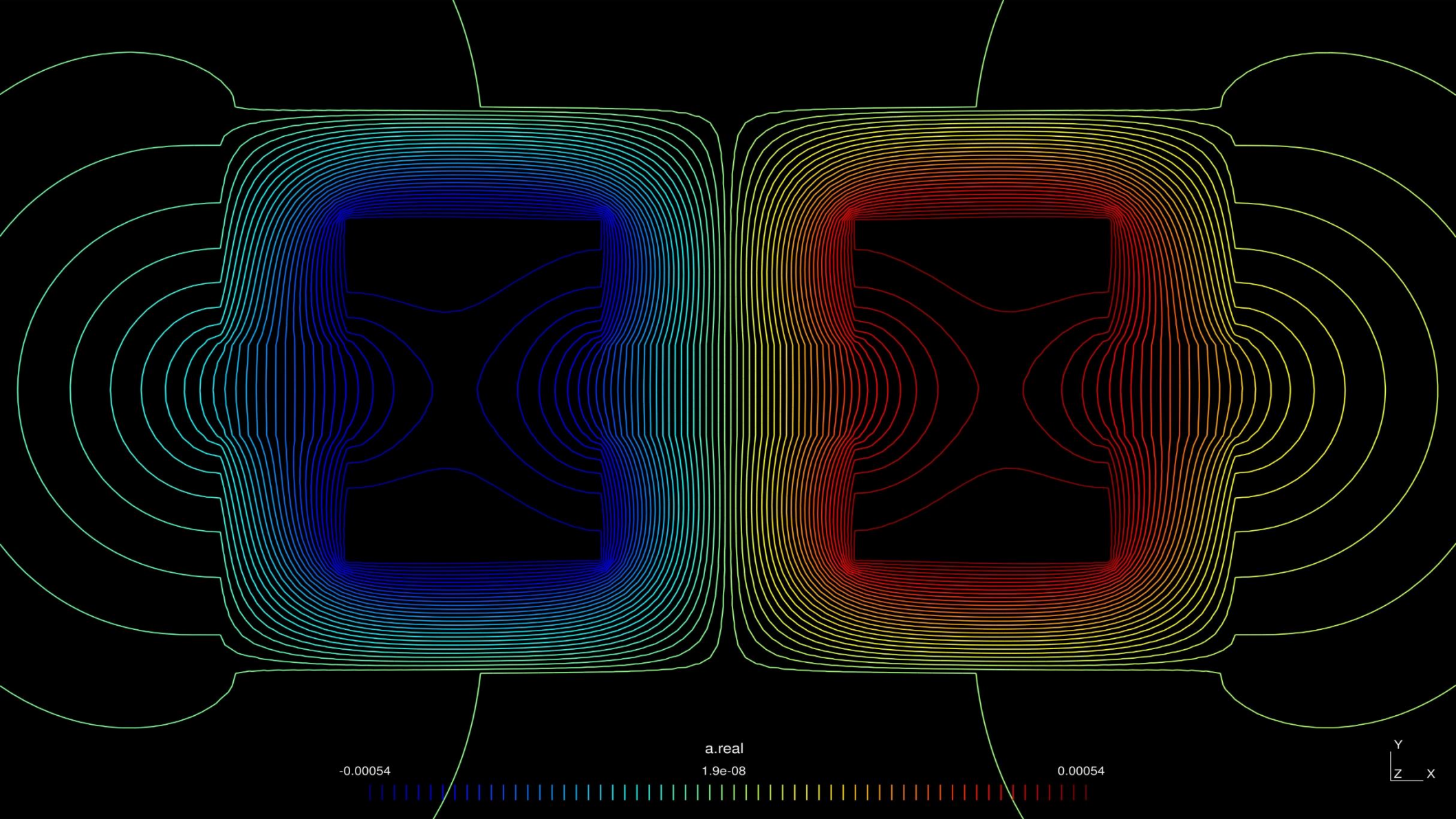
Results



Control points optimization results



Adaptive mesh deformation results

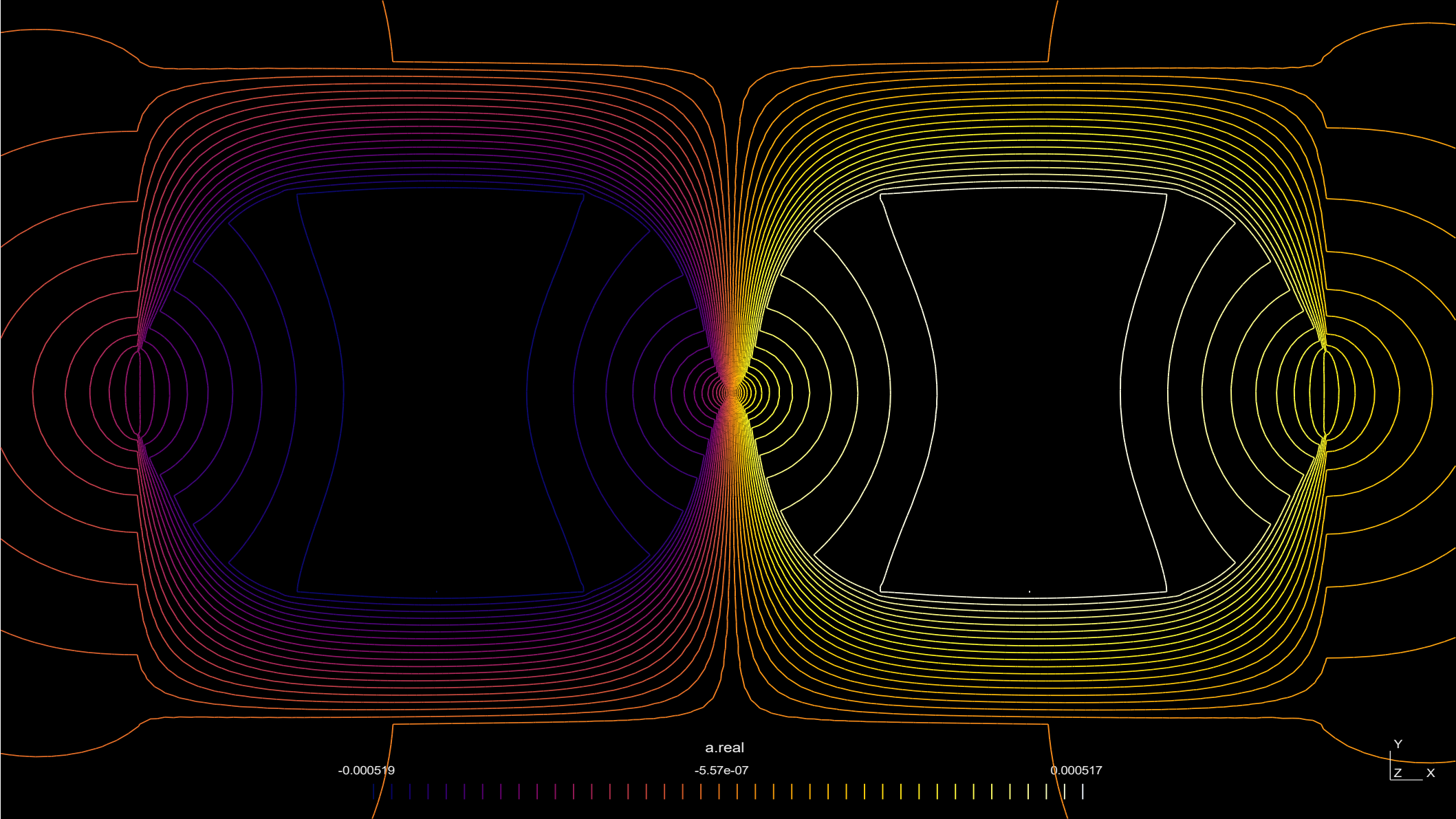


-0.00054

$a.\text{real}$
1.9e-08

0.00054

y
z x



-0.000519

a.real
-5.57e-07

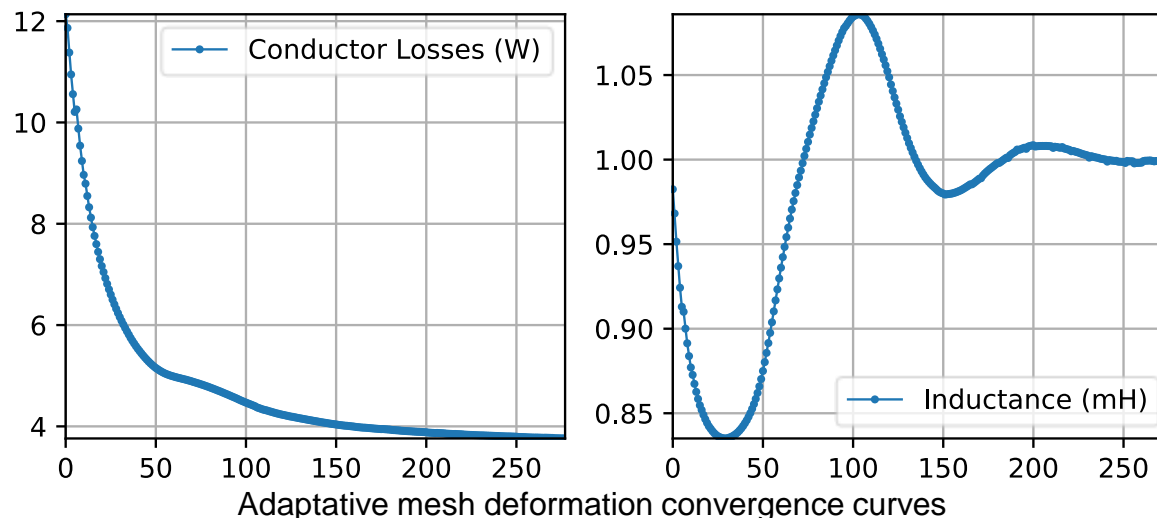
0.000517

Y
Z X

Results – Performance and Convergence

Optimization Type	Iterations	P (W)	L (mH)
Reference	-	13.16	1.00
Control points	92	6.00	1.00
Adaptative mesh deformation	277	3.76	1.00

Optimization results



Conclusion

We used **gradient-based** optimizations of the air gap profile of an inductor to minimize AC losses.

We leverage classical techniques such as:

- Control points, mesh deformations, remeshing
- Shape derivatives
- Adjoint method
- Hilbertian extension-regularization

Perspective:

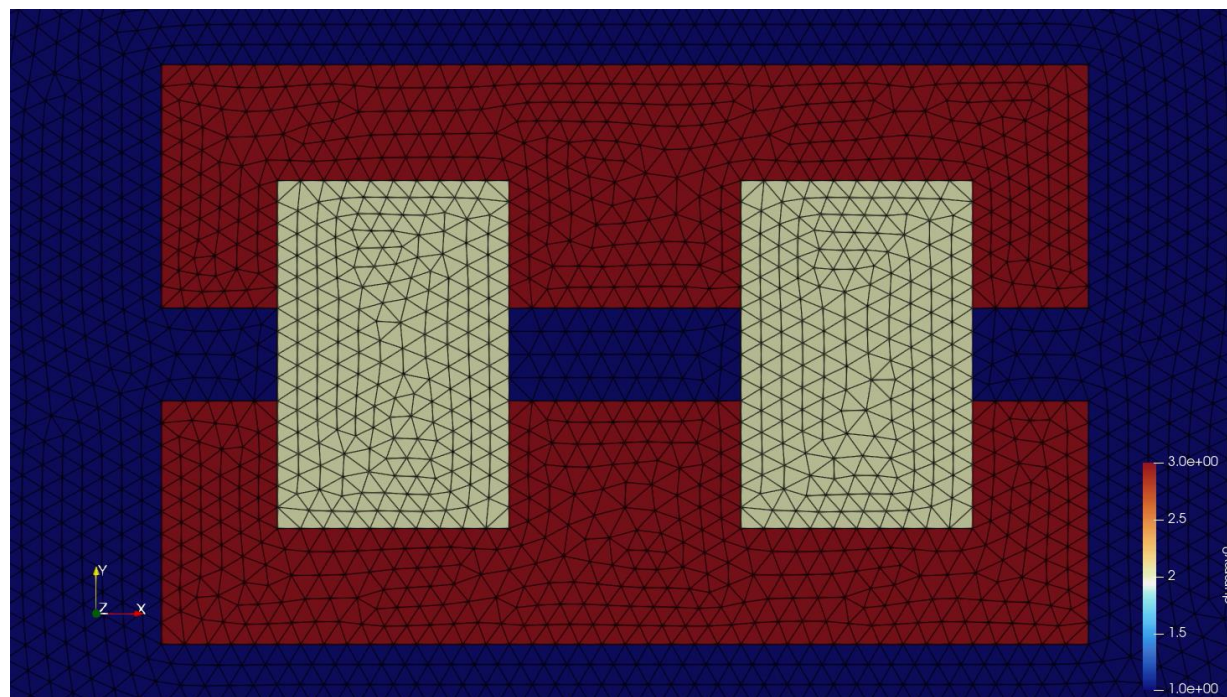
- Include magnetic saturation in the optimization
- Find better topology of the magnetic core
- Include iron losses

Thank You!

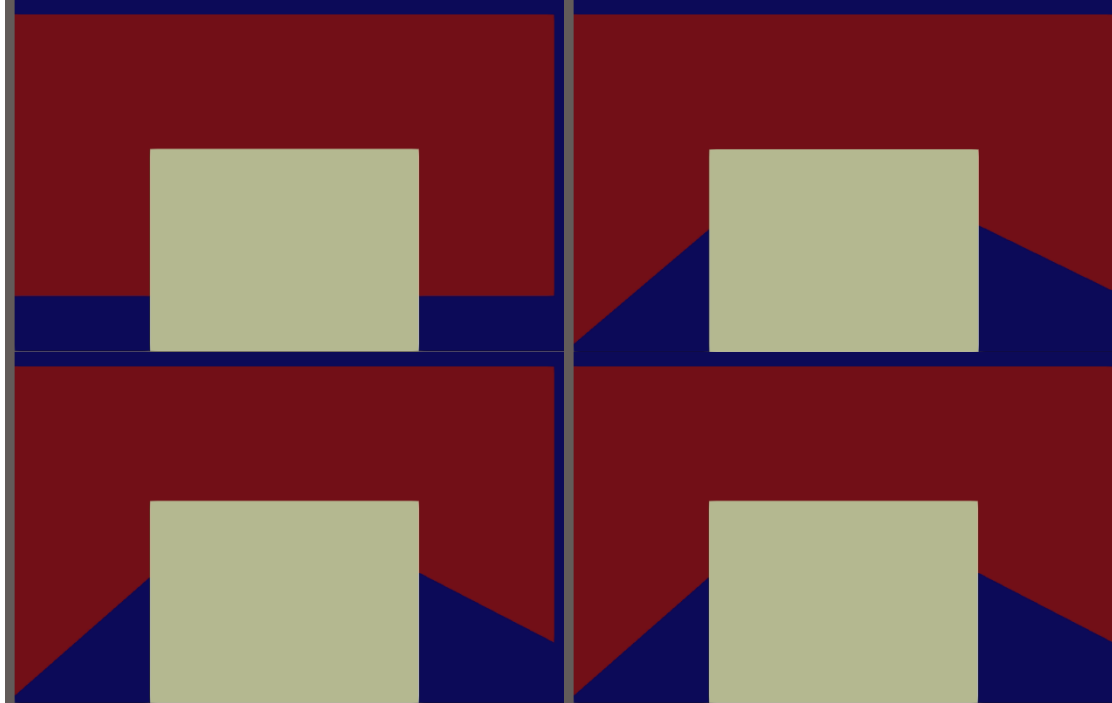
Code available on zenodo



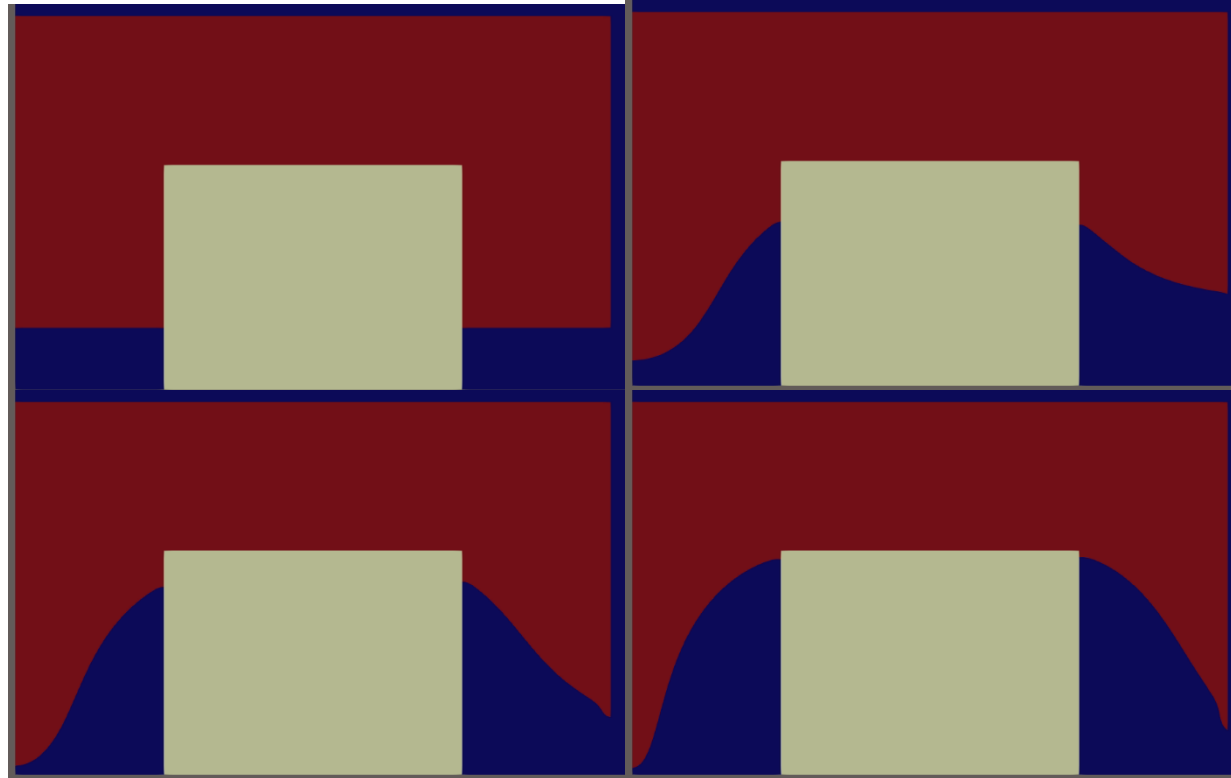
Results



Results – Control Points



Results – Adaptive Mesh Deformation



Shape derivative of $J(\Omega) = \int_{\Omega} f dx$

The change of variables $T = Id + \phi$ allows to transport the integral $J(\Omega_{\phi})$ onto the fixed reference domain Ω :

$$J(\Omega_{\phi}) = \int_{\Omega_{\phi}} f dx = \int_{\Omega} |\det(I + \nabla \phi)| f \circ (Id + \phi) dx.$$

Considering that:

- $|\det(I + \nabla \phi)|$ is Fréchet differentiable at $\phi = 0$, expanding to $1 + \operatorname{div}(\phi) + o(\phi)$.
- $f \circ (Id + \phi)$ is Fréchet differentiable at $\phi = 0$, expanding to $f + \nabla f \cdot \phi + o(\phi)$.

By the product rule, $\phi \mapsto J(\Omega_{\phi})$ is Fréchet differentiable at 0, and its derivative is the **volume form** :

$$J'(\Omega)(\phi) = \int_{\Omega} (\operatorname{div}(\phi)f + \nabla f \cdot \phi) dx = \int_{\Omega} \operatorname{div}(f\phi) dx.$$

Alternatively, if Ω is a bounded and Lipschitz domain, an integration by parts gives the **surface form** :

$$J'(\Omega)(\phi) = \int_{\partial\Omega} \phi \cdot n ds.$$

Structural theorem : Under mild assumptions, if $\phi \cdot n = 0$ on $\partial\Omega$ then $J'(\Omega)(\phi) = 0$.

The derivation can be difficult for elaborate functions, particularly those constrained by PDEs.

Adjoint Method of $E(p) = F(a_p)$

We want to differentiate a $E(p) = F(a_p)$ (e.g., the magnetic energy $E(p) = \int_D j a_p dx$) with respect to p , an arbitrary parametrization of the air gap profile. Here, j is the current density, and a_p is the magnetic potential.

We introduce two equations:

- The **state equation** $K(p)a_p = J$
- The **adjoint equation** : $K(p)^*\lambda_p = -\frac{dF}{da}(a_p)$ where $K(p)^*$ is the adjoint operator of $K(p)$.

Then, the sensitivities of $E(p)$ with respect to p can be expressed as :

$$\frac{dE}{dp}(p) = \left\langle \lambda_p, \frac{dK}{dp}(p)a_p \right\rangle$$

The scalar product shown above varies based on how $K(p)$, a_p , and λ_p are represented:

- if they are in discretized form, it's an \mathbb{R}^n scalar product, reflecting a "discretize then optimize" approach,
- if they are in variational form, it's an $L^2(D)$ scalar product, reflecting an "optimize then discretize" approach.

Jean C  a's method offers an elegant way to obtain this result.

Adjoint Method of $E(p) = F(a_p)$

The state equation can be expressed in two ways:

- **Discretized Form:** A linear system where $K(p)$ is the stiffness matrix and J is the discretized current density vector:

$$K(p)a_p = J$$

- **Variational Form:** Where $K(p)$ is a linear operator and J is a linear form:

$$K(p): H^1(D) \rightarrow H^{-1}(D)$$

$$a \mapsto \left(v \mapsto \int_D v(p) \nabla a \cdot \nabla v dx \right)$$

$$J : H^1(D) \rightarrow \mathbb{R}$$

$$v \mapsto \int_D j v dx$$

Glimpse of Jean C ea's method 1/3

To find the derivative of $E(p)$ with respect to p , we introduce a lagrangian function:

$$L(p, a, \lambda) = F(a) + \langle \lambda, K(p)a - J \rangle$$

A key observation is that $E(p) = L(p, a_p, \lambda)$ for any λ . Applying the chain rule, we get:

$$\frac{dE}{dp}(p) = \frac{\partial L}{\partial p}(p, a_p, \lambda) + \left\langle \frac{\partial L}{\partial a}(p, a_p, \lambda), \frac{da_p}{dp}(p) \right\rangle, \quad \forall \lambda.$$

The challenge here is that $\frac{da_p}{dp}(p)$ is difficult to define rigorously and compute.

Jean C ea's proposal is to choose λ such that the second term vanishes.

This is achieved by finding a saddle point of $L(p, \cdot, \cdot)$.

Glimpse of Jean C ea's method 2/3

Recalling the lagrangian function we posed :

$$L(p, a, \lambda) = F(a) + \langle \lambda, K(p)a - J \rangle$$

Canceling with respect to λ , we obtain :

$$\left\langle \frac{\partial L}{\partial \lambda}(p, a, \lambda), \hat{\lambda} \right\rangle = \langle \hat{\lambda}, K(p)a - J \rangle = 0, \quad \forall \hat{\lambda},$$

which recovers our original **state equation** $K(p)a_p = J$.

Canceling with respect to a , we obtain

$$\left\langle \frac{\partial L}{\partial a}(p, a, \lambda), \hat{a} \right\rangle = \left\langle \frac{dF}{da}(a), \hat{a} \right\rangle + \langle \lambda, K(p)\hat{a} \rangle = 0, \quad \forall \hat{a}.$$

Using the definition of the adjoint operator $K(p)^*$, this restates to

$$\left\langle \frac{dF}{da}(a), \hat{a} \right\rangle + \langle K(p)^*\lambda, \hat{a} \rangle = 0, \quad \forall \hat{a},$$

which recovers our original **adjoint equation** $K(p)^*\lambda_p = -\frac{dF}{da}(a_p)$.

Thus, a saddle point of $L(p, \cdot, \cdot)$ is $(a, \lambda) = (a_p, \lambda_p)$.

Glimpse of Jean C ea's method 3/3

Recalling the lagrangian function we posed :

$$L(p, a, \lambda) = F(a) + \langle \lambda, K(p)a - J \rangle$$

At the saddle point $(a, \lambda) = (a_p, \lambda_p)$, we have:

$$\frac{dE}{dp}(p) = \frac{\partial L}{\partial p}(p, a_p, \lambda_p) + \left\langle \frac{\partial L}{\partial a}(p, a_p, \lambda_p), \frac{da_p}{dp}(p) \right\rangle$$

Thus, the formula for sensitivities is :

$$\boxed{\frac{dE}{dp}(p) = \frac{\partial L}{\partial p}(p, a_p, \lambda_p) = \left\langle \lambda_p, \frac{dK}{dp}(p)a_p \right\rangle}$$

**Applying this approach to shape derivatives demands special care
and yields more intricate expressions.**