



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

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# Introduction

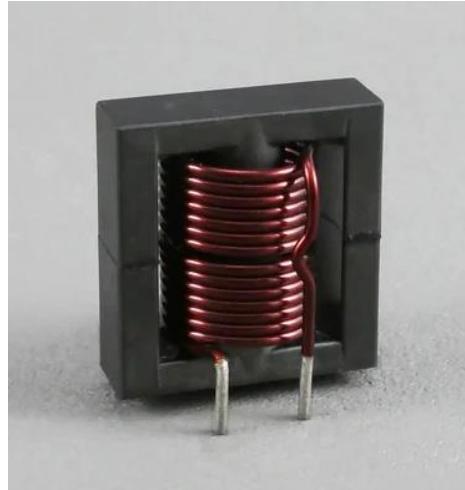
We want to optimize an **inductor** to minimize the AC losses due to the **fringing effect**.

Possible solutions :

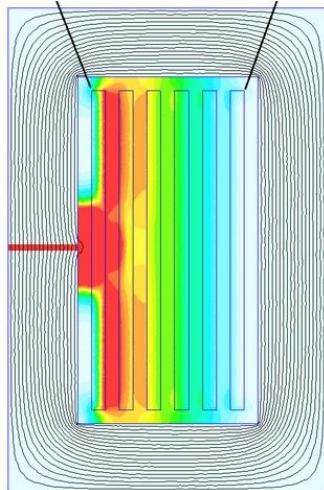
- Distributed air gaps
- Change winding distribution of the conductor [1]
- Optimize air gap profile (e.g. with heuristic methods [2])

**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](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  $L_0 = 1 \text{ mH}$ .

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

where :

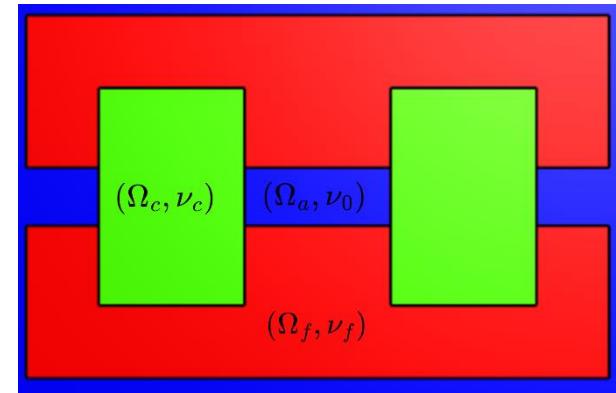
$\underline{a}$  is magnetic vector potential out-of-plane component

$\underline{j}$  is real current density,

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

Reluctivity value ( $\nu$ )	Subdomain
$\nu_0 = 1/\mu_0$	Air region $\Omega_a$
$\nu_f = \nu_0/1000$	Core region (ferrite) $\Omega_f$
$\nu_c = \nu_0 \exp(i\delta)$	Conductor region (copper) $\Omega_c$

with  $\delta = 0.1 \text{ rad}$  to model proximity effects at  $f = 50 \text{ kHz}$



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:

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

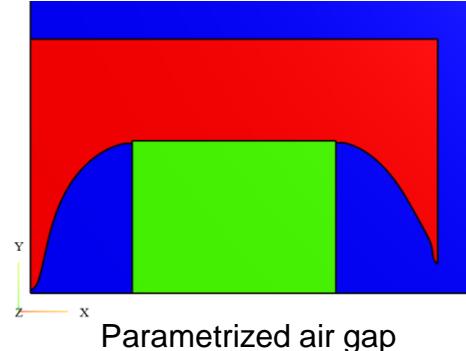
with  $l_z = 1 \text{ cm}$  (thickness) and  $I = 2 \text{ A}$  (current, 200 turns).

# 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  $\underline{\nu}_h$ ,
- the magnetic potential  $\underline{a}_h$  solution of  $(E_h)$



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

Then, the objective and constraint rewrites into :

AC Losses:

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

Inductance:

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

# 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  $h$ :

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

# Challenges and Solutions

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

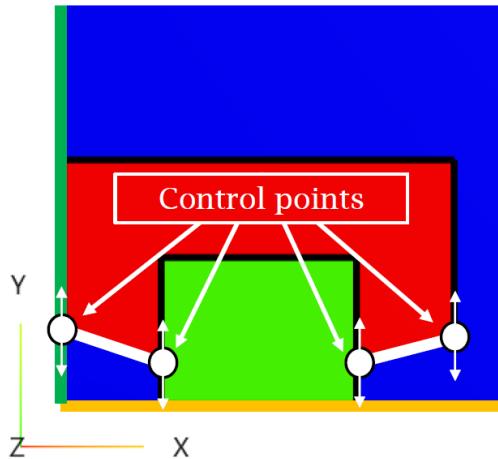
We are using a handwritten **augmented Lagrangian** algorithm to solve this constrained optimization problem.

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

## Part II - Methodology

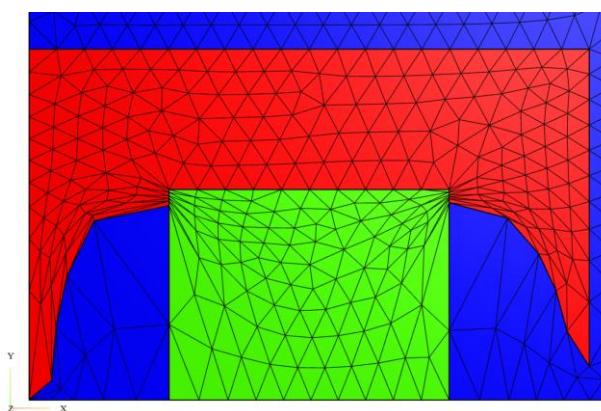
# Two Choices for the Parametrization $h$

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

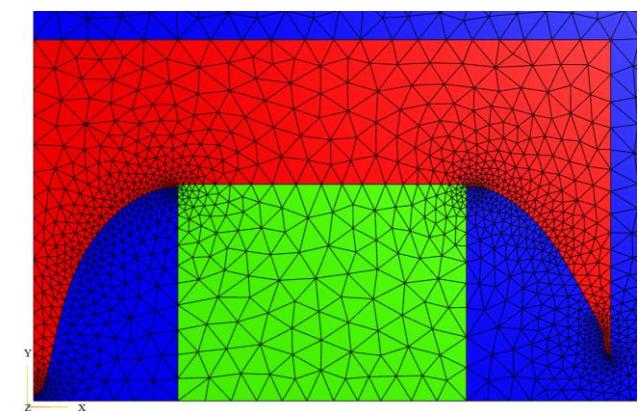


Parametric Shape Optimization using  
**Control points**

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



Geometric Shape Optimization by  
**mesh deformation**



Geometric Shape Optimization  
by **adaptative** 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  $\Omega$
- “Small and regular enough” **deformation fields**  $\phi$
- A real-value function  $J(\Omega)$  of the domain

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

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

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

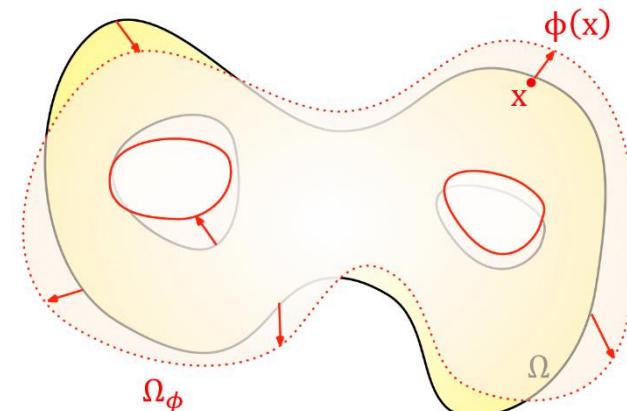
We are searching a  $\phi$  that is a **descent direction**:

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

**Two approaches:**

In our “control points” approach,  $\phi$  is defined on the control points

In our “adaptative mesh deformation” approach,  $\phi$  is defined on the mesh nodes.



Deformed domain

# Example of a Shape Derivative

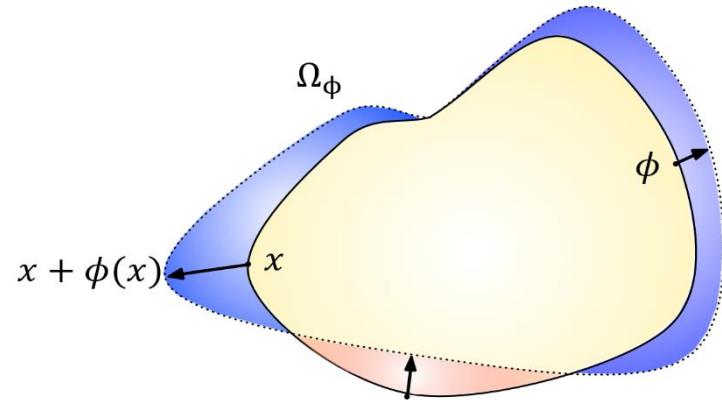
For example, if we take a function  $J(\Omega) = \int_{\Omega} f dx$ ,  
its shape derivative is:

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

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  $\Omega$  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** :

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

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  $2 \times 2$  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**

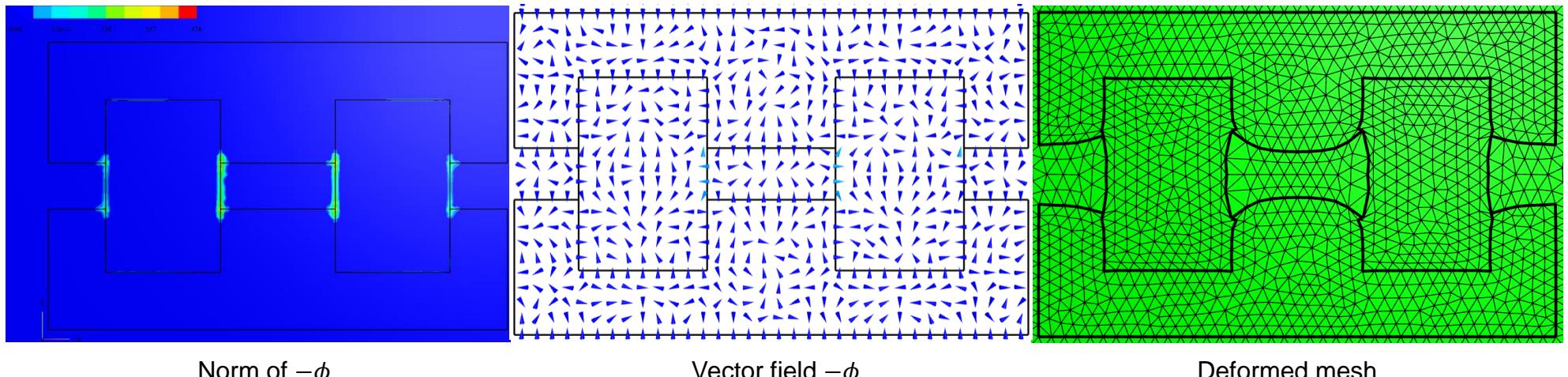
with a **structure** that can preserve the conductor's boundary.

# Unregularized the Deformation Field

With a deformation field  $\phi$  coming from:

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

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



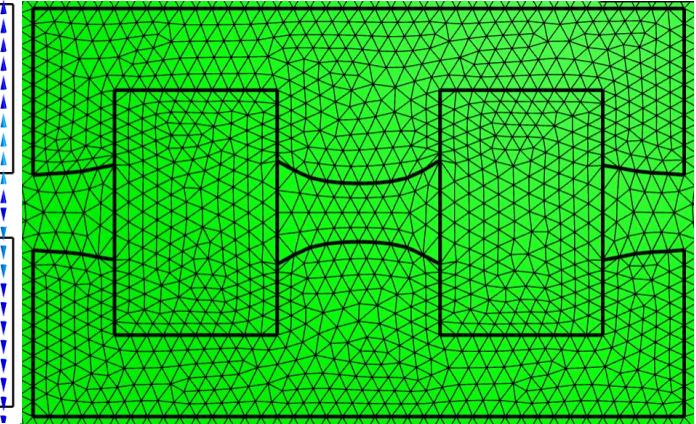
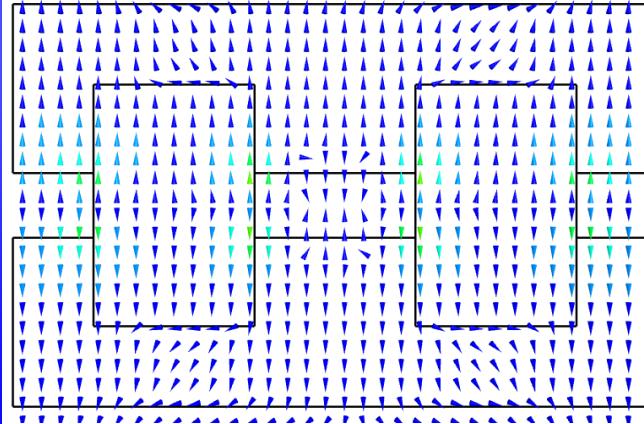
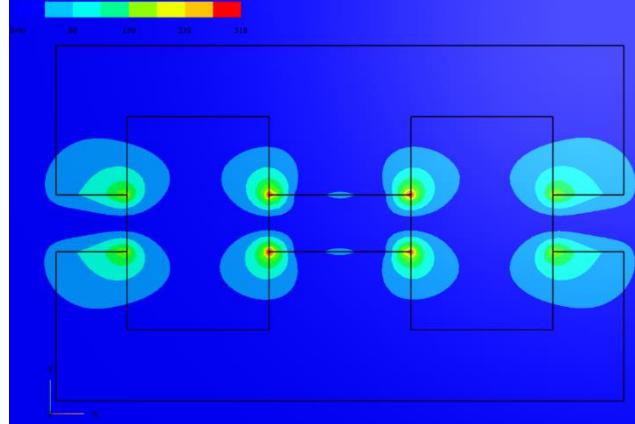
# Regularized Deformation Field

With a deformation field  $\phi$  coming from:

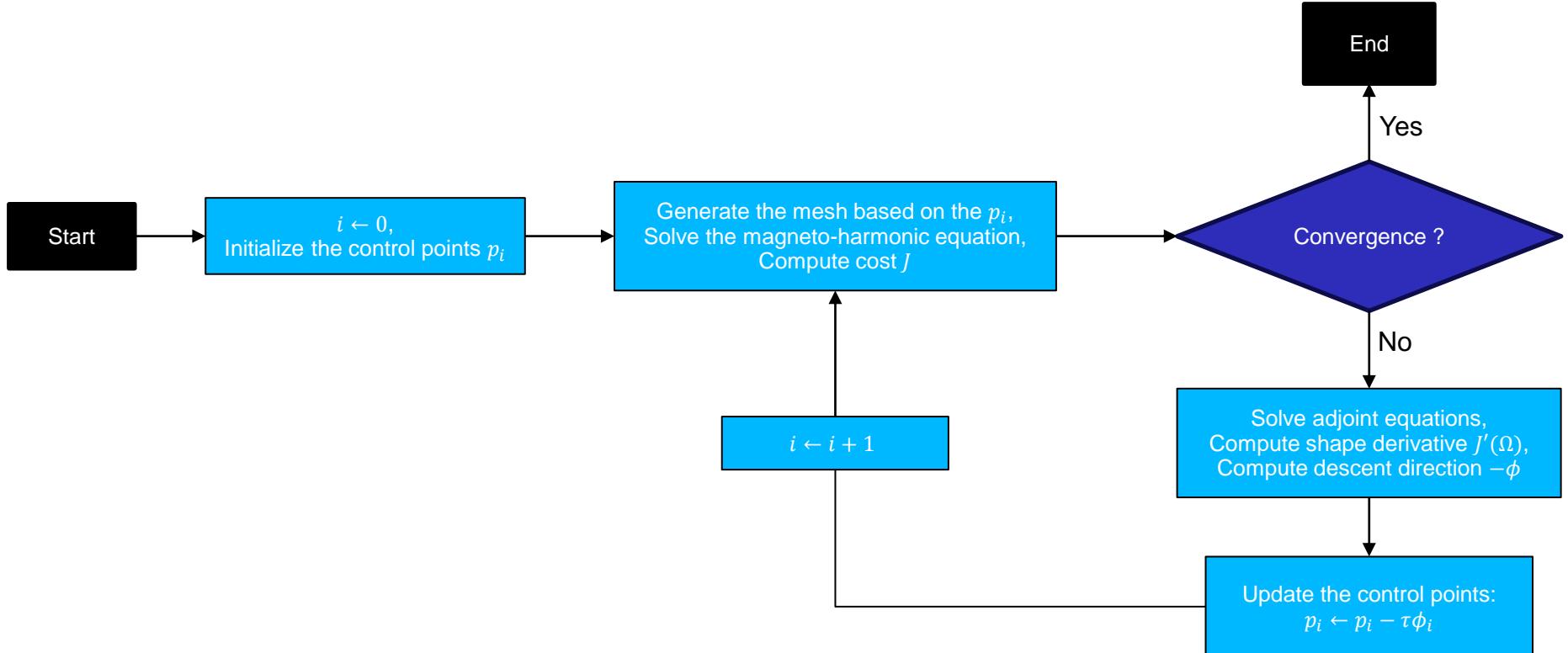
Find  $\phi \in V$  s.t.  $\forall \hat{\phi} \in V$ ,

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

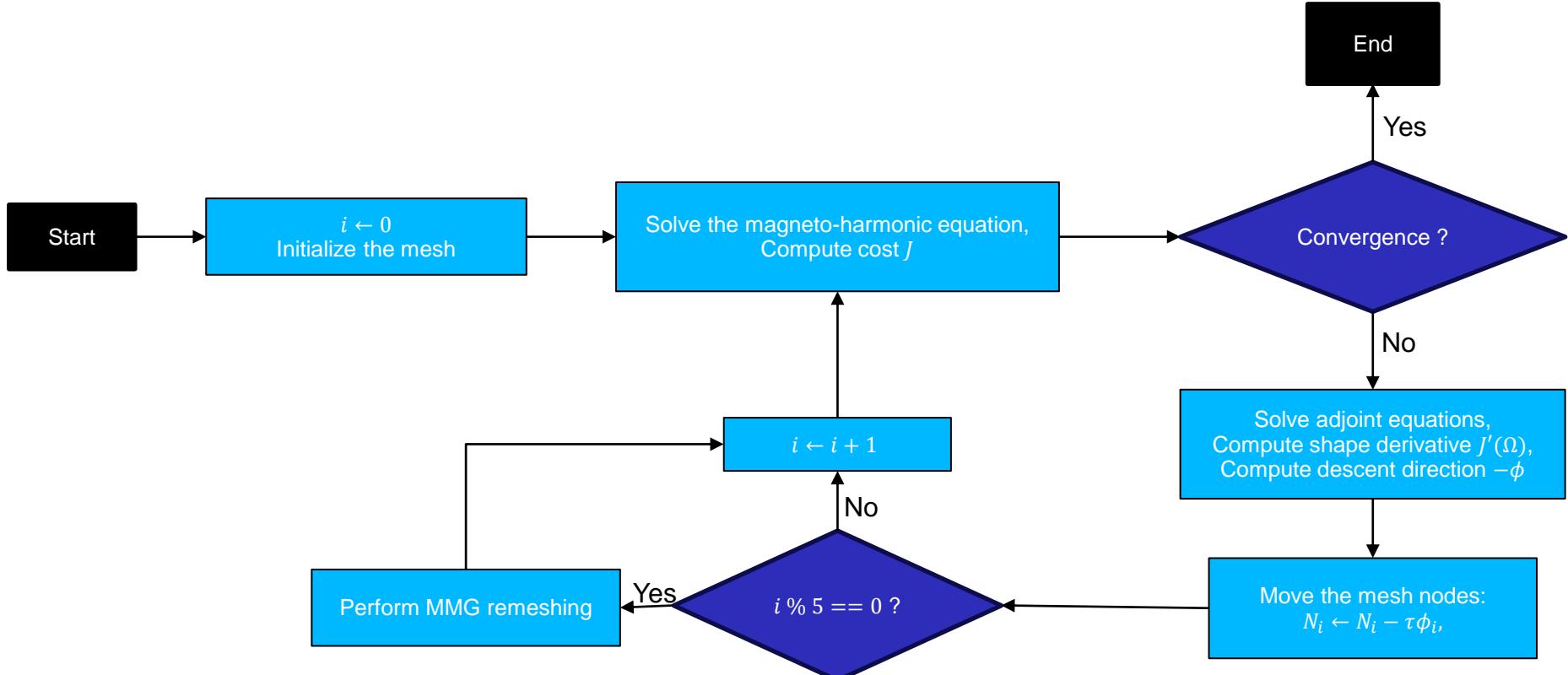
with  $V = \{\phi \in H^1(\Omega) | \phi \cdot \vec{e}_x = 0 \text{ on } \Gamma_{D,x}, \phi \cdot \vec{e}_y = 0 \text{ on } \Gamma_{D,y}\}$



# Control Points Flowchart

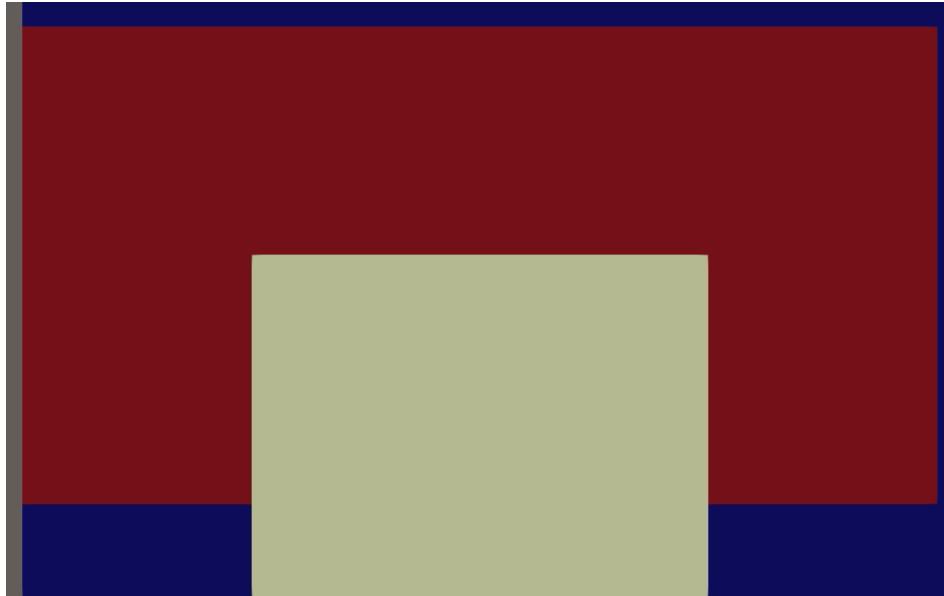


# Adaptative Mesh Deformation Flowchart

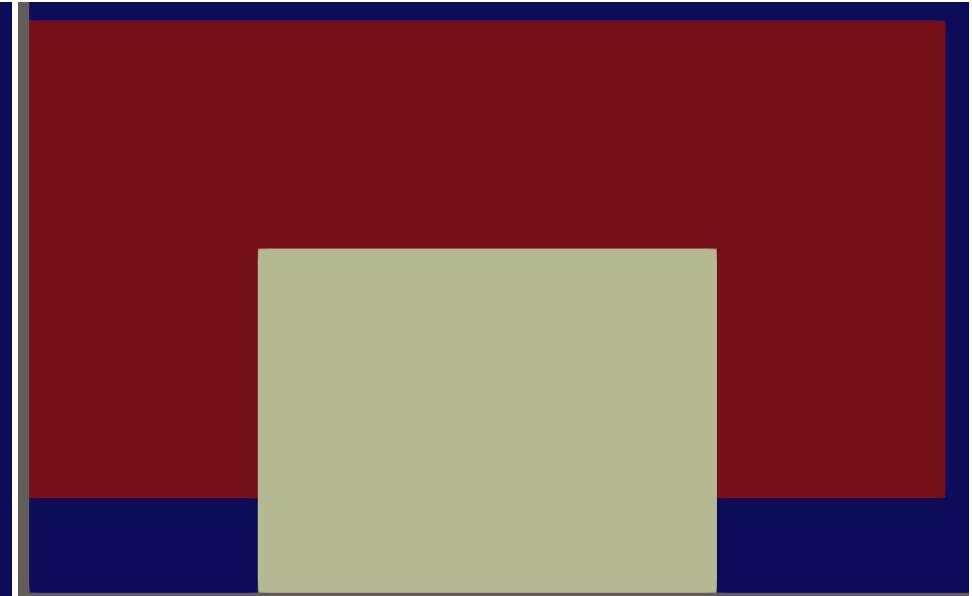


# Part III – Optimization Results

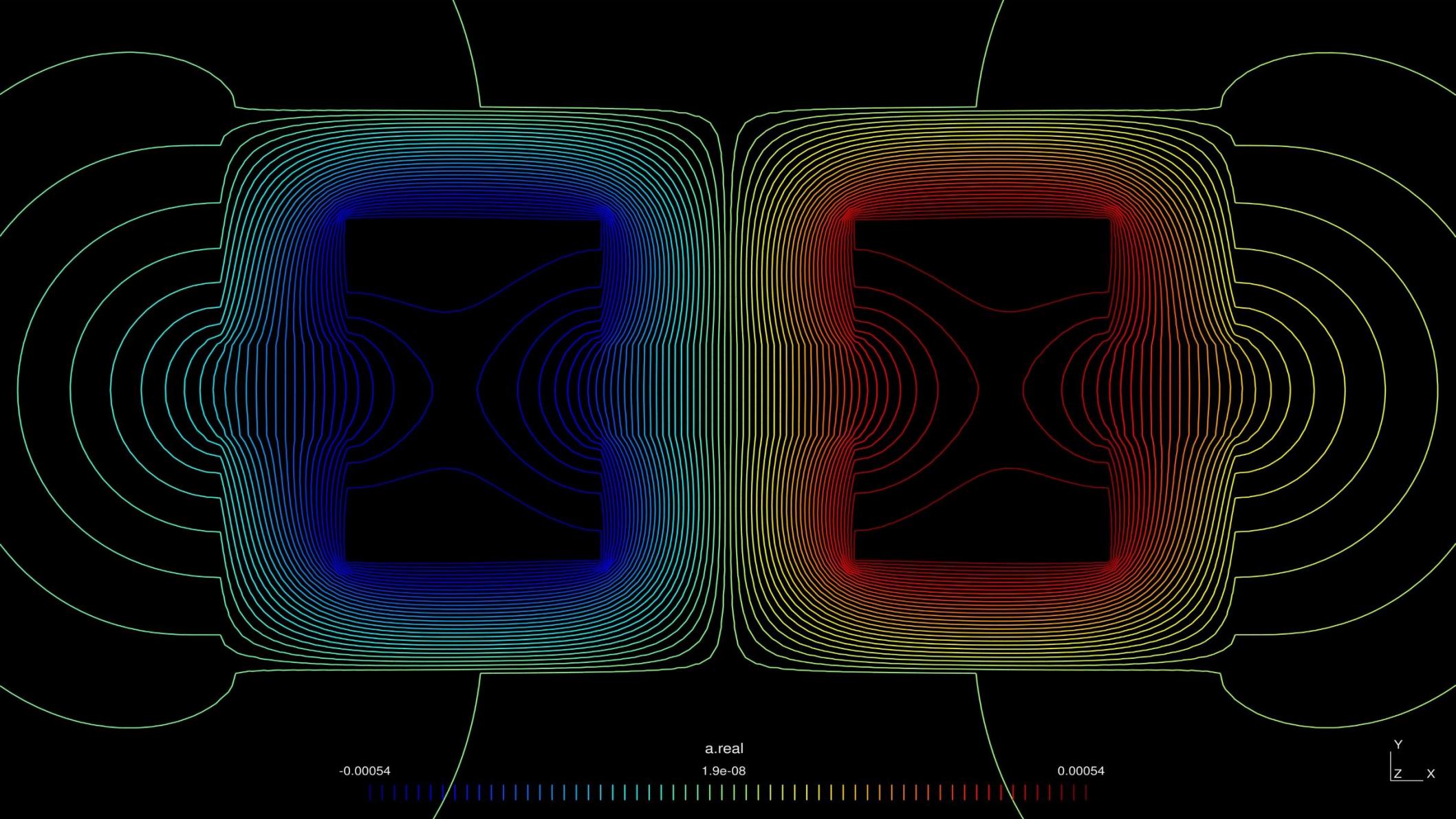
# Results



Control points optimization results



Adaptative mesh deformation results

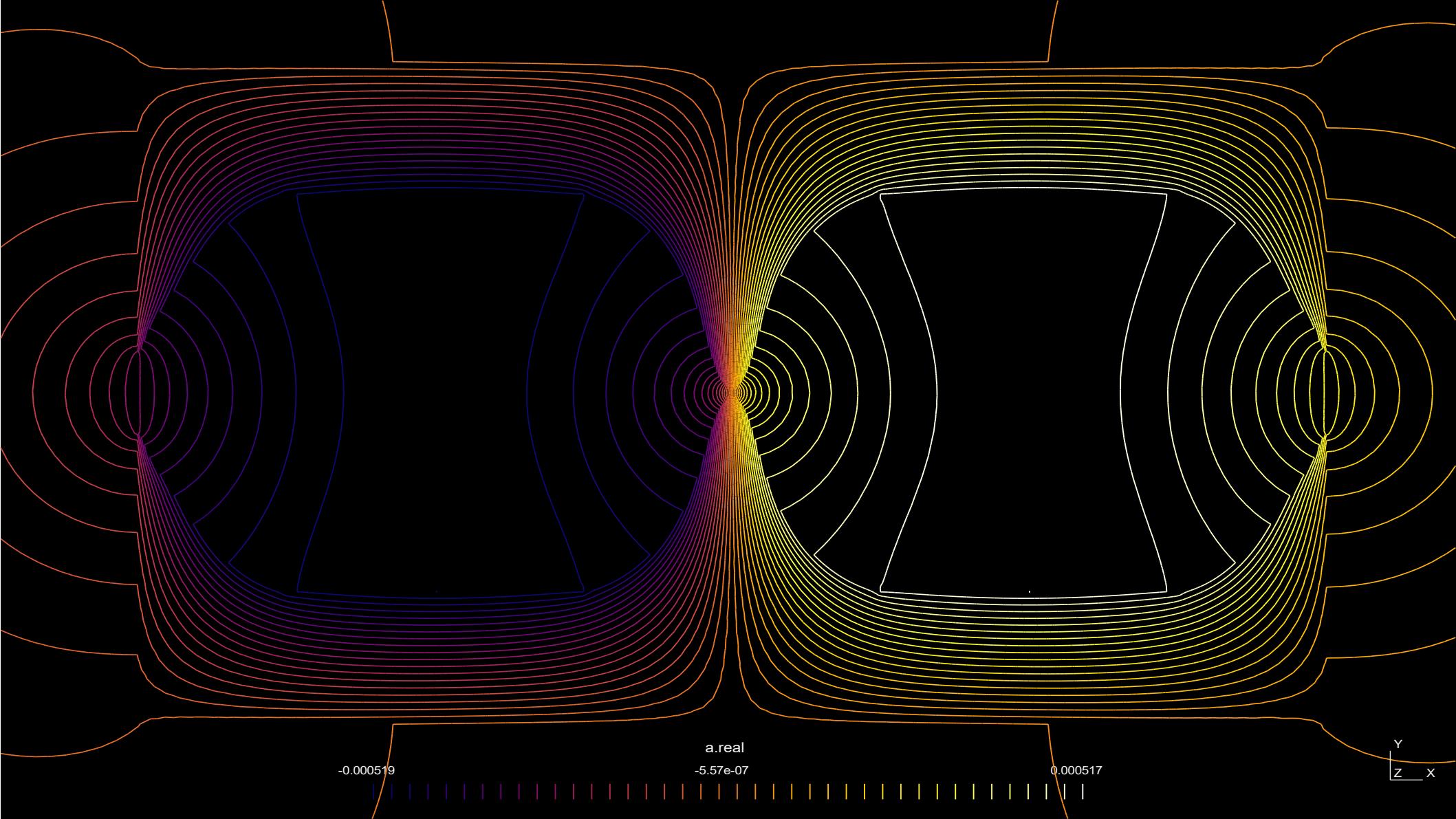


-0.00054

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

0.00054

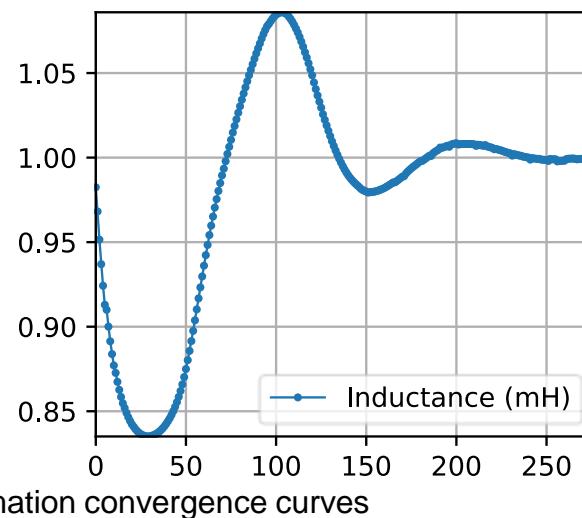
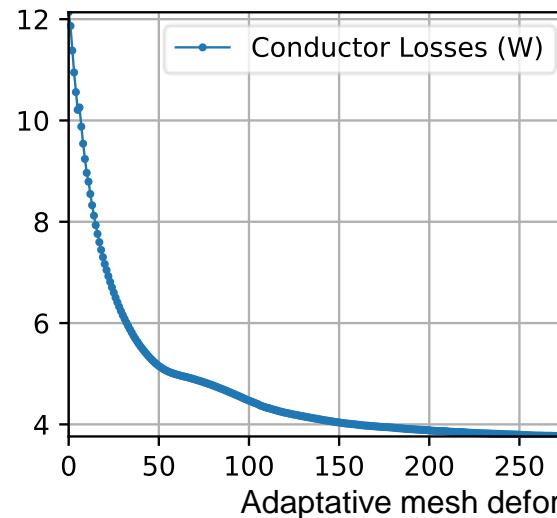
Y  
Z  
X



# 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 derivates
- Adjoint method
- Hilbertian extension-regularization

Perspective:

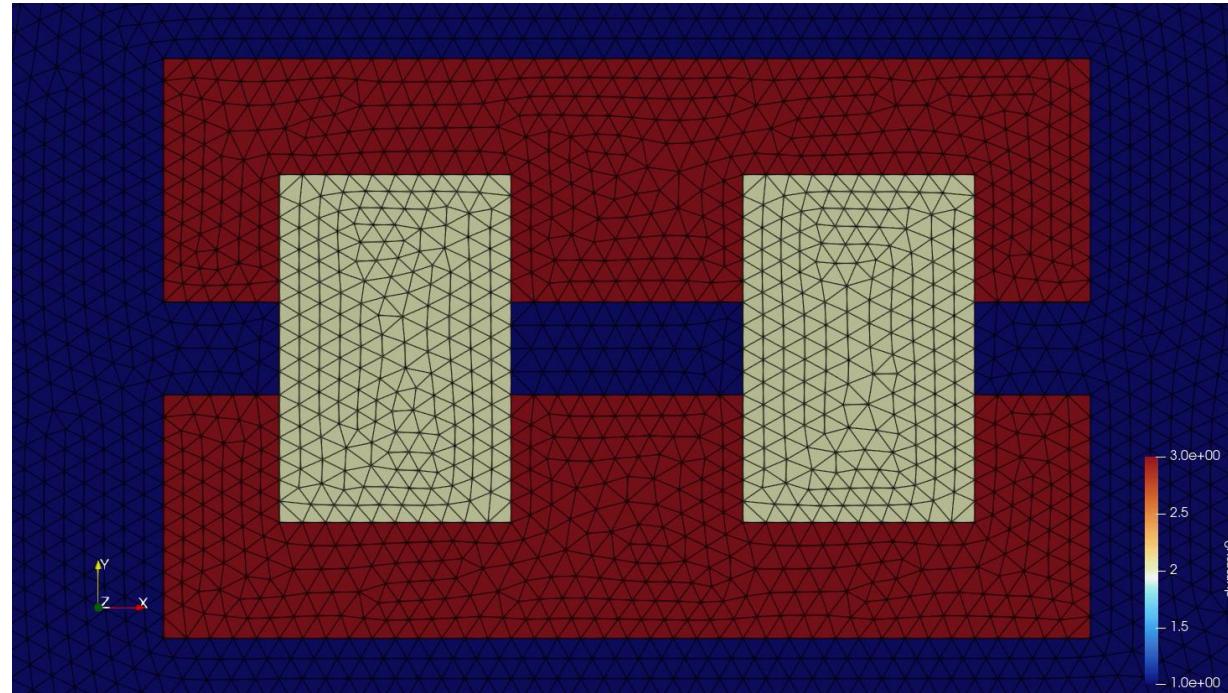
- Include magnetic saturation in the optimization
- Find better topology of the magnetic core
- Include iron losses

# Thank You!

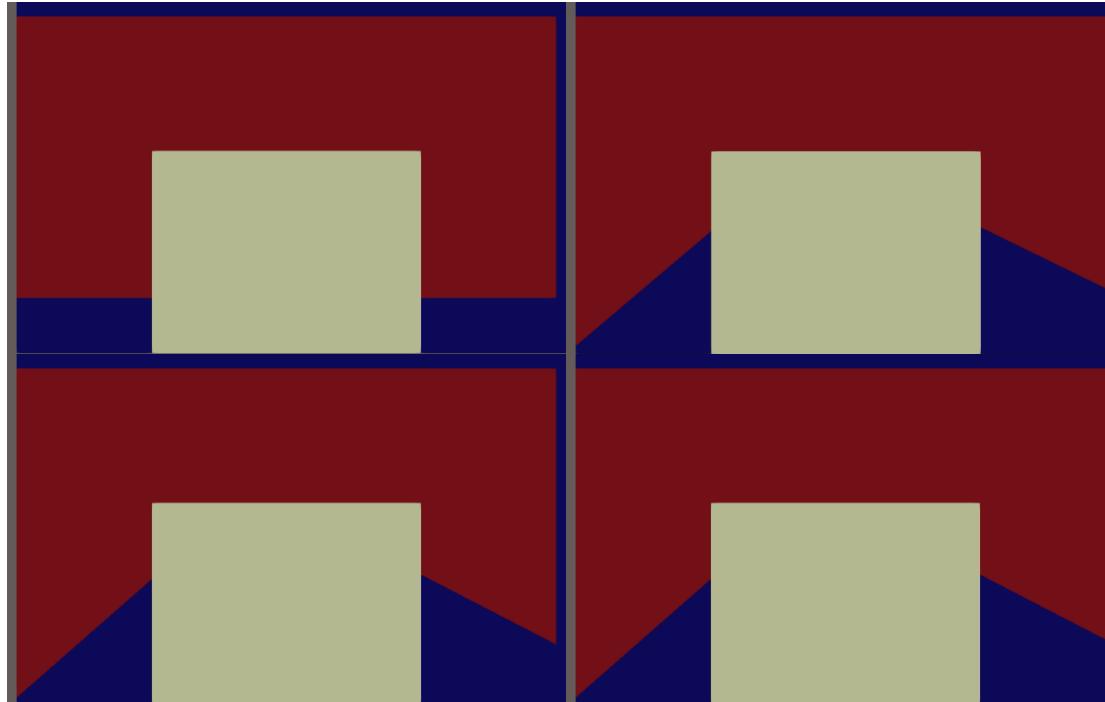
Code available on zenodo



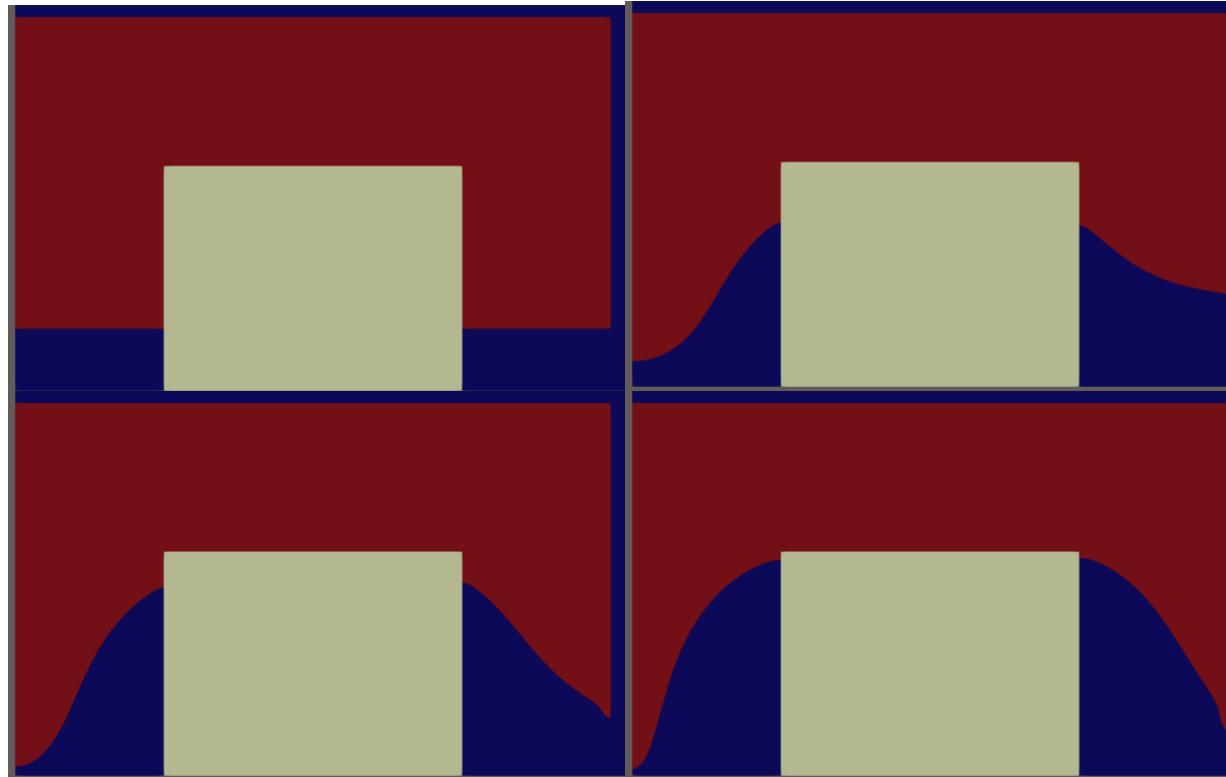
# Results



# Results – Control Points



# Results – Adaptative 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  $\Omega$ :

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

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,  $\phi \mapsto F(\Omega_\phi)$  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  $\Omega$  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  $\lambda_p$  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 Céa'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:

$$\begin{aligned} K(p) : H^1(D) &\rightarrow H^{-1}(D) \\ a &\mapsto \left( v \mapsto \int_D v(p) \nabla a \cdot \nabla v dx \right) \\ J : H^1(D) &\rightarrow \mathbb{R} \\ v &\mapsto \int_D j v dx \end{aligned}$$

# Glimpse of Jean Céa'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  $\lambda$ . 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 Céa's proposal is to choose  $\lambda$  such that the second term vanishes.

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

# Glimpse of Jean Céa's method 2/3

Recalling the lagrangian function we posed :

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

Cancelling with respect to  $\lambda$ , 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**  $K(p)a_p = J$ .

Cancelling 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 Céa'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.**