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Article

Solid Isotropic Material with Penalization-Based Topology Optimization of Three-Dimensional Magnetic Circuits with Mechanical Constraints

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Abstract: Topology optimization is currently enjoying renewed interest thanks to the recent development of 3D printing techniques, which offer the possibility of producing these new complex designs. One of the difficulties encountered in manufacturing topologically optimized magnetostatic structures is that they are not necessarily mechanically stable. In order to take this mechanical constraint into account, we have developed a SIMP-based topology optimization algorithm which relies on numerical simulations of both the mechanical deformation and the magnetostatic behavior of the structure. Two variants are described in this paper, respectively taking into account the compliance or the von Mises constraint. By comparing the designs obtained with those from magnetostatic optimization alone, our approach proves effective in obtaining efficient and robust designs.

Keywords: topology optimization; sensitivity analysis; magnetostatics; mechanical constraint

MSC: 65K10; 78A30



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1. Introduction

Topology optimization was first introduced to solve design problems in mechanics [1–3]. One of the most efficient approaches is based on the adjoint method [1–3] associated with SIMP penalization techniques [2,4]. From these works, topology optimization based on SIMP and adjoint approaches was extended to deal with design problems in electromagnetism such as magnetostatic issues [5,6] in order to design, for example, circuits of Hall effect thrusters [5,7] or electric machines [6,8]. Note that these design problems are mainly simplified by 2D approximations. Magnetostatic topology optimization based on the adjoint method associated with SIMP penalization techniques has been recently extended from 2D to 3D problems by some of this paper's authors [9].

In all of these works, only one type of physics is considered in the optimization problem: either mechanics or magnetostatics. Thus, the designs obtained in magnetostatic problems are not always mechanically robust. They are sometimes not manufacturable and sometimes unable to withstand the mechanical stress experienced over their lifetime [10,11]. Note that a first study was recently conducted in [12] to develop a multiphysical topology optimization code to design electrical machines using 2D simplification.

Therefore, the purpose of this work is to address the 3D design problem of magnetostatic circuits, taking into account their mechanical robustness. Thus, this problem deals with two types of physics at the same time: mechanics and magnetostatics. The proposed approach combines the two physics through the use of a unique design variable. A continuous adjoint method associated with a SIMP penalization technique is developed to provide a topology optimization code which makes it possible to solve 3D design problems in magnetostatics while including mechanical considerations. In order to show the

efficiency of this topology optimization code, a 3D design problem of a magnetic circuit is solved. Note that this simple magnetic circuit is similar to the circuit of Hall effect thrusters in [5,13].

In Section 2, we present the formulation of the 3D design problem we address in this work. This includes the definition of what variables and objective and constraint functions we will have to deal with. In Section 3, the SIMP penalization techniques as well as the continuous adjoint method are recalled, and the approach linking the mechanics and magnetostatics is detailed. In Section 4, this topology optimization code dealing with two types of physics is validated through solving a 3D design example of a magnetic circuit. This design problem is detailed and the different results are discussed. Section 5 concludes the paper.

2. Formulation of the Coupled Mechanical and Magnetostatic Problem

In this paper, we focus on the optimization of a magnetic circuit similar to that of a Hall-effect thruster, as represented in Figure 1. It should be noted that the “toy” problem presented in this study is not intended to be fully representative of the real world, but rather is intended to allow a 3D magnetostatic problem to be considered without oversimplification.

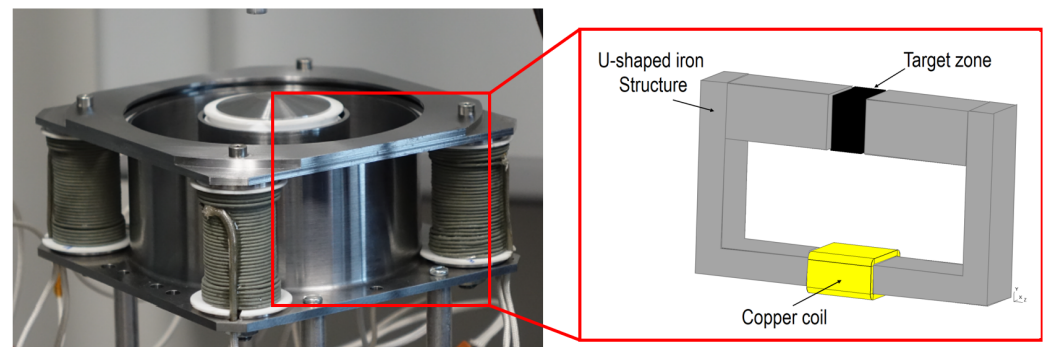


Figure 1. (left) A Hall-effect thruster whose shape must be optimized to obtain a target magnetic field \mathbf{B}_0 inside the cavity. (right) A cross-section of this device is approximated into the following “toy” problem: a U-shaped iron structure is obtained (the coil was moved down from one side of the thruster to the center of the structure) with a cubic target zone in which the field must be controlled.

This first section focuses on a comprehensive description of the physics involved in this study and is organized as follows: First, in Section 2.1, we describe the physics (PDE and objective function) behind the studied magnetic circuit. Second, in Section 2.2, we then present two mechanical constraints that are intended to yield manufacturable designs. Finally, in Section 2.3, we show how the two types of physics are coupled to define a unique topology optimization problem.

2.1. Magnetostatic Problem

The magnetic circuit consists of a copper coil enclosing a U-shaped iron structure surrounded by air. From a numerical point of view, the air is represented by a cube of finite size whose walls are sufficiently far from the structure so that they do not influence the simulation results. The full simulation domain is denoted by $\Omega \subset \mathbb{R}^3$, the coil by $\Omega_C \subset \Omega$, and the iron parts by $\Omega_I \subset \Omega$. The air regions are then given by $\Omega_A = \Omega \setminus (\Omega_C \cup \Omega_I)$; see Figure 2.

From the *Maxwell–Ampere equation*, the magnetic field \mathbf{H} inside Ω is a solution to $\nabla \times \mathbf{H} = \mathbf{J}$, with \mathbf{J} being a prescribed electric current which is equal to zero outside the coil,

where $\mathbf{J}(\mathbf{x}) = \mathbf{0}$ for $\mathbf{x} \in \Omega \setminus \Omega_C$. In this work, the materials are assumed to be linear; hence, we use the following constitutive relation

$$\mathbf{B} = \mu_0 \mu_r (\mathbf{H} + \mathbf{M}) \tag{1}$$

between the magnetic flux density \mathbf{B} , magnetization \mathbf{M} , relative permeability μ_r , and \mathbf{H} . Note that $\mathbf{M} = \mathbf{0}$ in this problem because there is no magnet. The permeability μ_r is equal to either μ_{air} in the air and coil or μ_{iron} inside the iron parts of the structure:

$$\mu_r(\mathbf{x}) = \begin{cases} \mu_{\text{iron}} & \text{if } \mathbf{x} \in \Omega_I \\ \mu_{\text{air}} & \text{elsewhere} \end{cases} \tag{2}$$

From the Gauss law for magnetism $\nabla \cdot \mathbf{B} = 0$, there exists a vector potential \mathbf{A} such that $\mathbf{B} = \nabla \times \mathbf{A}$ (unique up to a Gauge condition such as $\nabla \cdot \mathbf{A} = 0$). To summarize, \mathbf{A} is solution to

$$\begin{cases} -\nabla \times \left(\frac{1}{\mu_0 \mu_r} \nabla \times \mathbf{A} \right) + \mathbf{J} = \mathbf{0} & \text{in } \Omega \\ \mathbf{n} \times \mathbf{A} = \mathbf{0} & \text{on } \partial\Omega \end{cases} \tag{3}$$

in which a Dirichlet boundary condition on the borders $\partial\Omega$ of the domain is imposed to simulate the fact that the potential \mathbf{A} is set to zero at infinity.

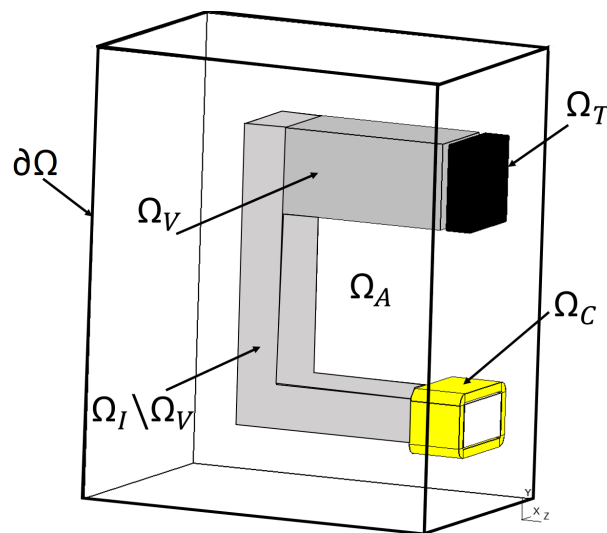


Figure 2. Half of the magnetic circuit studied in this article (the other half is obtained by symmetry), and the different domains that make it up: the iron domain Ω_I , the coil Ω_C , the air Ω_A , the target area Ω_T , and the variable region Ω_V , in which we will optimize the structure.

The aim of the magnetic circuit is to obtain a certain given field \mathbf{B}_0 within a target zone Ω_T located between the two “arms” of the structure. Indeed, in some applications, the aim is to impose a magnetic field in certain regions in order to improve the performance of the magnetic device. This is the case, for example, with the Hall-effect thruster shown in Figure 2, where the magnetic field traps electrons in certain regions and prevents them from striking and damaging the walls of the thruster. The problem presented in this article is a “toy” problem representative of this type of application, where a certain \mathbf{B}_0 field must be imposed in certain regions. To achieve this objective, we will optimize the distribution of matter (iron and air) within a region $\Omega_V \subset \Omega$ (the index V refers to the variable nature

of this region). To accomplish this, we define a distribution $\varrho : \Omega_V \rightarrow \{0, 1\}$ equal to either one or zero at each position $\mathbf{x} \in \Omega_V$ such that

$$\mu_r(\mathbf{x}) = \mu_{\text{air}} + (\mu_{\text{iron}} - \mu_{\text{air}})\varrho(\mathbf{x}). \tag{4}$$

We then search for the optimal distribution ϱ^* (which define an optimal shape $\Omega_{V,I,\text{opt}} = \{\mathbf{x}, \varrho^*(\mathbf{x}) = 1\}$ for the iron part inside the variable region) which minimizes the error between the actual field $\mathbf{B}(\varrho) = \nabla \times \mathbf{A}(\varrho)$ solution of Equation (3) using Equation (4) (note that to simplify the notations, the dependency on ϱ is not always specified hereafter) and the target one \mathbf{B}_0 :

$$\varrho^* = \arg \min_{\varrho} F_1(\varrho) \quad \text{where} \quad F_1(\varrho) = \int_{\Omega_T} \|\mathbf{B}(\varrho) - \mathbf{B}_0\|^2 \, d\mathbf{x}. \tag{5}$$

2.2. Mechanical Problem

The main objective of introducing mechanics in the optimization problem is to design magnetic parts that are mechanically robust. This is achieved by taking into account the mechanical loads that the parts will be subjected to during their future use. The load considered in this article is an arbitrary load that the part will encounter over its life time. Moreover, the mechanical considerations will tackle the problem of shapes that are mechanically unstable, in particular regions floating in the air, as seen in Section 4. We therefore present here two measures of the mechanical strength of the optimized structure based on simulations of the mechanical deformations caused by an arbitrary load \mathbf{g} applied to both end Γ_L of the structure arms assuming that the optimized shape is clamped on the non-variable iron parts Γ_I ; see Figure 3.

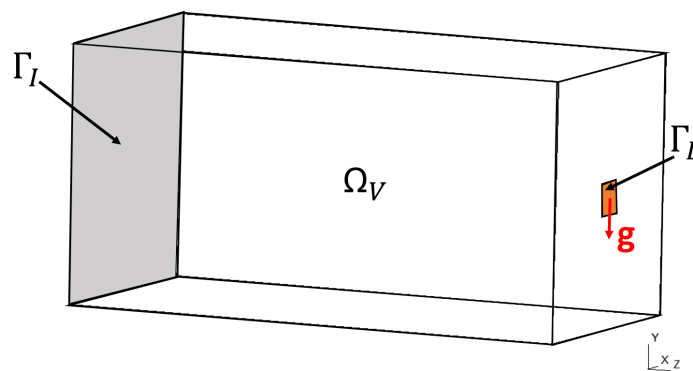


Figure 3. Zoom in on the Ω_V region in which we are searching for the optimal shape of the iron structure and the different notations used to simulate its mechanical deformation: Γ_I is a fixed surface on which the structure is clamped and Γ_L the surface on which a load \mathbf{g} is applied.

According to the principle of conservation of linear momentum, when no forces are applied to the structure, the stress tensor $\bar{\bar{\sigma}}$ inside the optimized region verifies $\nabla \cdot \bar{\bar{\sigma}} = \mathbf{0}$. Under *Hooke's law*, the displacement vector \mathbf{u} is related to $\bar{\bar{\sigma}}$ through

$$\bar{\bar{\sigma}}(\mathbf{u}) = 2\mu\bar{\bar{e}}(\mathbf{u}) + \lambda\text{tr}(\bar{\bar{e}}(\mathbf{u}))\mathbf{I}_3 = A\bar{\bar{e}}(\mathbf{u}) \tag{6}$$

where λ and μ are the *Lamé coefficients*, and can be expressed in terms of the *Young's modulus* E and the *Poisson ratio* ν by

$$\lambda = \frac{\nu E}{(1 + \nu)(1 - 2\nu)} \quad \text{and} \quad \mu = \frac{E}{2(1 + \nu)}$$

while A is a fourth-order elasticity tensor containing these mechanical properties. In the context of small perturbations, the deformation tensor $\bar{\bar{e}}(\mathbf{u})$ is also given by $\bar{\bar{e}}(\mathbf{u}) = (\nabla\mathbf{u} + \nabla\mathbf{u}^T)/2$. We denote by E_{iron} and $\nu = 0.3$ the mechanical properties in-

side the iron material. As is common in topology optimization of mechanical structures, we consider air to be a very soft material with a small Young’s modulus $E_{\text{air}} \simeq 0$ and $\nu = 0.3$. Using the distribution of matter $\varrho : \Omega_V \rightarrow \{0, 1\}$ inside the variable region again, we then have the following for $\mathbf{x} \in \Omega_V$:

$$E(\mathbf{x}) = E_{\text{air}} + (E_{\text{iron}} - E_{\text{air}})\varrho(\mathbf{x}). \tag{7}$$

This makes it possible to define the displacement field $\mathbf{u}(\varrho)$ as the only solution of the following problem:

$$\begin{cases} -\nabla \cdot [A(\varrho)\bar{\boldsymbol{\varepsilon}}(\mathbf{u}(\varrho))] = \mathbf{0} & \text{in } \Omega_V \\ \mathbf{u}(\varrho) = \mathbf{0} & \text{on } \Gamma_I \\ A(\varrho)\bar{\boldsymbol{\varepsilon}}(\mathbf{u}(\varrho))\mathbf{n} = \mathbf{g} & \text{on } \Gamma_L \\ A(\varrho)\bar{\boldsymbol{\varepsilon}}(\mathbf{u}(\varrho))\mathbf{n} = \mathbf{0} & \text{on } \partial\Omega_V \setminus (\Gamma_I \cup \Gamma_L) \end{cases}. \tag{8}$$

We now define two figures of merit that will be taken into account in order to ensure the mechanical strength of the topologically optimized design:

1. The first criterion to ensure robustness of the structure is to maximize its rigidity, which is equivalent to minimizing its compliance. In other words, we want a distribution ϱ which minimizes

$$F_2(\varrho) = \int_{\Gamma_L} \mathbf{g} \cdot \mathbf{u}(\varrho) \, ds. \tag{9}$$

2. The second criterion is based on the *von Mises stress* σ_{VM} expressed in terms of the components of the stress tensor $\bar{\boldsymbol{\sigma}}$ as follows:

$$\sigma_{VM} = \left(\sigma_{xx}^2 + \sigma_{yy}^2 + \sigma_{zz}^2 - \sigma_{xx}\sigma_{yy} - \sigma_{yy}\sigma_{zz} - \sigma_{zz}\sigma_{xx} + 3\sigma_{xy}^2 + 3\sigma_{yz}^2 + 3\sigma_{zx}^2 \right)^{\frac{1}{2}}. \tag{10}$$

The aim here is to minimize the maximal value taken by σ_{VM} inside the optimized iron shape where $\varrho(\mathbf{x}) = 1$, i.e.,

$$\|\varrho\sigma_{VM}\|_{\infty} = \max_{\mathbf{x} \in \Omega_V} \varrho(\mathbf{x})\sigma_{VM}(\mathbf{x}).$$

Note that here the stress measure is defined by the relaxed stress $\varrho\sigma_{VM}$ instead of σ_{VM} in order to ensure that the stress measure is zero in the air region where $\varrho(\mathbf{x}) = 0$ [14]. As the use of an infinite norm raises difficulties for the optimization method, we will instead use the following *Kreisselmeier–Steinhauser (KS) function*, $F_{3,p}(\varrho) \simeq \|\varrho\sigma_{VM}(\varrho, \mathbf{u}(\varrho))\|_{\infty}$, which is valid for large values of p :

$$F_{3,p}(\varrho) = \left(\int_{\Omega_V} [\varrho\sigma_{VM}(1, \mathbf{u}(\varrho))]^p \, d\mathbf{x} \right)^{\frac{1}{p}}. \tag{11}$$

2.3. Link between the Two Types of Physics

We have seen in the previous sections that the material distribution $\varrho(\mathbf{x})$ is responsible for both the magnetic $\mu_r(\mathbf{x})$ (with Equation (4)) and mechanical $E(\mathbf{x})$ (with Equation (7)) properties by defining iron (resp., air) regions where $\varrho(\mathbf{x}) = 1$ (resp., $\varrho(\mathbf{x}) = 0$). We can thus define a coupled optimization problem in which an optimal distribution ϱ is sought such that it simultaneously minimizes the error between \mathbf{B} and \mathbf{B}_0 (Equation (5)) while simultaneously minimizing one of the two mechanical criteria of Equation (9) or Equation (11).

To achieve this, we will first optimize the magnetic circuit alone without any mechanical constraint in order to find an unstable optimal distribution ϱ_0^* which minimizes F_1 . In a second step, we will consider the following problem, which consists of minimizing one

of the two mechanical criteria F_2 or F_3 under the constraint of preserving good magnetic performance close to those of ϱ_0^* :

$$(\mathcal{P}) \quad \begin{cases} \min_{\varrho} & F_2(\varrho) \quad (\text{or } F_{3,p}(\varrho)) \\ \text{u.c.} & F_1(\varrho) \leq F_1(\varrho_0^*) + \epsilon \end{cases} \quad (12)$$

where $\epsilon > 0$ is a small value.

3. A SIMP-Based Optimization Problem

Solving problem (\mathcal{P}) of Equation (12), in which $\varrho : \Omega_V \rightarrow \{0, 1\}$ is a distribution that can only take the values 0 or 1, is a numerically difficult task. Instead, we can relax the latter constraint and look for a material distribution $\rho : \Omega_V \rightarrow [0, 1]$ that accepts any intermediate value between 0 and 1. Thus, a continuous optimization problem is obtained (for which an efficient gradient-based method can be used) rather than a discrete one. To achieve a true material distribution with only iron ($\rho(\mathbf{x}) = 1$) or air ($\rho(\mathbf{x}) = 0$) at the end of the optimization process, it is necessary to penalize intermediate values of ρ during the gradient descent. This is the principle behind the well-known solid isotropic material with penalization (SIMP) method [2,4,9].

In the next Section 3.1, the main method to deal with the penalization issue is recalled. In Section 3.2, we then show how computing the gradients of the functions F_1 , F_2 , and F_3 presented in Section 2 is possible using adjoint states obtained via the *Céa method* introduced in [15]. These gradients are then used in Section 3.3 to implement an optimization algorithm which can be used to solve problem (\mathcal{P}) .

3.1. Penalization with Material Density Method

As previously explained, the constraint on the values taken by $\rho(\mathbf{x})$ has been relaxed to the whole $[0, 1]$ interval to simplify the optimization process. However, to prevent the appearance of intermediate values in the optimal distribution $\rho^*(\mathbf{x})$, the SIMP method suggests penalizing the values of ρ , which are used in the definition of the material properties in the following way:

$$\begin{aligned} \mu_r(\mathbf{x}) &= \mu_{\text{air}} + (\mu_{\text{iron}} - \mu_{\text{air}})g(\rho(\mathbf{x})) \\ E(\mathbf{x}) &= E_{\text{air}} + (E_{\text{iron}} - E_{\text{air}})g(\rho(\mathbf{x})) \end{aligned}$$

with $g : [0, 1] \rightarrow [0, 1]$ being a non-decreasing interpolation function verifying $g(0) = 0$ and $g(1) = 1$. In the literature, there are several choices of interpolation functions g that influence the final result of topology optimization [2,4,16]. In this work, we use the polynomial interpolation function for its simplicity [4], which is defined by

$$g(x) = x^n \quad \text{and} \quad g'(x) = nx^{n-1} \quad (13)$$

where $n > 0$ is a penalty parameter that will be incremented during the optimization process. As n goes up, the optimal solution tends to become discrete, with values closer to either 0 or 1. It is worth noting that this interpolation function is not symmetrical, and hence it favors materials with low values of ρ (i.e., air).

Remark 1. In the definition of the objective function $F_{3,p}$ described in Equation (11), we consider the von Mises stress inside the iron structure. Now that intermediate materials can be present in the optimized design, we need to relax the definition of $F_{3,p}$. Following [17], Equation (11) is modified into:

$$\hat{F}_{3,p}(\rho) = \left(\int_{\Omega_V} [\rho^m \sigma_{VM}(1, \mathbf{u}(\rho))]^p d\mathbf{x} \right)^{\frac{1}{p}} \quad (14)$$

with $m > 0$ ($m = 1/2$ in the numerical examples of Section 4).

3.2. Sensitivity Analysis

In this section, the sensitivities of the three objective functions F_1 , F_2 and $\hat{F}_{3,p}$ with respect to the design variable ρ are provided. Regarding the magnetostatic problem, the following proposition is verified:

Proposition 1 (Sensitivity of F_1 with respect to ρ). *The expression of the sensitivity of F_1 with respect to the design variable ρ is:*

$$\langle F'_1(\rho), \tilde{\rho} \rangle = \int_{\Omega_V} \frac{(\mu_{iron} - \mu_{air})g'(\rho)\tilde{\rho}}{(\mu_{air} + (\mu_{iron} - \mu_{air})g(\rho))^2} \mathbf{B}_{adj} \cdot \mathbf{B} \, dx \tag{15}$$

in which \mathbf{B} is given by the direct problem of Equation (3) and $\mathbf{B}_{adj} = \nabla \times \mathbf{A}_{adj}$ is an “adjoint” field solution of

$$\begin{cases} -\nabla \times \left(\frac{1}{\mu_0 \mu_r} \nabla \times \mathbf{A}_{adj} - \mathbf{M}_{adj} \right) = \mathbf{0} & \text{in } \Omega \\ \mathbf{n} \times \mathbf{A}_{adj} = \mathbf{0} & \text{on } \partial\Omega \end{cases} \tag{16}$$

where \mathbf{M}_{adj} is a magnetization source which is non-zero inside Ω_T and defined as follows:

$$\mathbf{M}_{adj} = \frac{2}{\mu_0} (\mathbf{B} - \mathbf{B}_0) \mathbb{1}_{\Omega_T}.$$

We already proved this proposition in [9]. Similarly, for the first mechanical constraint F_2 , we have:

Proposition 2 (Sensitivity of F_2 with respect to ρ). *The expression of the sensitivity of F_2 with respect to the design variable ρ in a direction $\tilde{\rho}$ is given by*

$$\langle F'_2(\rho), \tilde{\rho} \rangle = - \int_{\Omega_V} A^*(\rho) \bar{\mathbf{e}}(\mathbf{u}) : \bar{\mathbf{e}}(\mathbf{u}) \tilde{\rho} \, dx \tag{17}$$

where $A^*(\rho) = \left[\frac{E_{air}}{E_{iron}} + \left(1 - \frac{E_{air}}{E_{iron}}\right)g(\rho) \right] A(1)$.

In this case, the problem is said to be auto-adjoint (the adjoint is the same as the direct solution \mathbf{u}). The proof is detailed in Appendix A. And finally, for F_3 , we find the following proposition holds:

Proposition 3 (Sensitivity of $\hat{F}_{3,p}$ with respect to ρ). *The expression of the sensitivity of $\hat{F}_{3,p}$ with respect to the variable design ρ in a direction $\tilde{\rho}$ is given by*

$$\begin{aligned} \langle \hat{F}'_{3,p}(\rho), \tilde{\rho} \rangle &= \int_{\Omega_V} A^*(\rho) \bar{\mathbf{e}}(\mathbf{u}) : \bar{\mathbf{e}}(\mathbf{p}) \tilde{\rho} \, dx \\ &+ \left(\int_{\Omega_V} \rho^{mp} \sigma_{VM}(1, \mathbf{u})^p \, dx \right)^{\frac{1}{p}-1} \int_{\Omega_V} m \sigma_{VM}^p(1, \mathbf{u}) \rho^{mp-1} \tilde{\rho} \, dx \end{aligned} \tag{18}$$

where \mathbf{p} is the adjoint state, a solution for all $\mathbf{v} \in V$, where

$$V = (H_0^1(\Omega_V))^3 = \left\{ \mathbf{v} \in (H^1(\Omega_V))^3, \mathbf{v} = \mathbf{0} \text{ on } \Gamma_I \right\} \tag{19}$$

of the following variational formulation:

$$\left(\int_{\Omega_V} \rho^{mp} \sigma_{VM}^p(1, \mathbf{u}) \, d\mathbf{x} \right)^{\frac{1}{p}-1} \int_{\Omega_V} \rho^{mp} \sigma_{VM}^{p-1}(1, \mathbf{u}) \frac{\partial \sigma_{VM}^T}{\partial \vec{\sigma}} \vec{\sigma}(\mathbf{v}) \, d\mathbf{x} + \int_{\Omega_V} A^*(\rho) \bar{\mathbf{e}}(\mathbf{v}) : \bar{\mathbf{e}}(\mathbf{p}) \, d\mathbf{x} = 0 \quad (20)$$

where $\vec{\sigma}$ is the Voigt notation for constraints $\bar{\mathbf{e}}$ defined by

$$\vec{\sigma} = (\sigma_{xx}, \sigma_{yy}, \sigma_{zz}, \sigma_{xy}, \sigma_{yz}, \sigma_{zx})^T$$

and

$$\frac{\partial \sigma_{VM}}{\partial \vec{\sigma}} = \begin{pmatrix} \frac{\partial \sigma_{VM}}{\partial \sigma_{xx}} \\ \frac{\partial \sigma_{VM}}{\partial \sigma_{yy}} \\ \frac{\partial \sigma_{VM}}{\partial \sigma_{zz}} \\ \frac{\partial \sigma_{VM}}{\partial \sigma_{xy}} \\ \frac{\partial \sigma_{VM}}{\partial \sigma_{yz}} \\ \frac{\partial \sigma_{VM}}{\partial \sigma_{zx}} \end{pmatrix} = \begin{pmatrix} \frac{1}{2\sigma_{VM}} (2\sigma_{xx} - \sigma_{yy} - \sigma_{zz}) \\ \frac{1}{2\sigma_{VM}} (2\sigma_{yy} - \sigma_{xx} - \sigma_{zz}) \\ \frac{1}{2\sigma_{VM}} (2\sigma_{zz} - \sigma_{xx} - \sigma_{yy}) \\ \frac{3}{\sigma_{VM}} (\sigma_{xy}) \\ \frac{3}{\sigma_{VM}} (\sigma_{yz}) \\ \frac{3}{\sigma_{VM}} (\sigma_{zx}) \end{pmatrix}. \quad (21)$$

This proposition is also proven in Appendix A. Note that the proof of these three propositions are all based on the following four steps (which are theoretically based on [1,2] and detailed in either [9] or Appendix A):

1. Definition of a constrained optimization problem in which $J(\rho, \mathbf{w}) = F(\rho)$ is given by either F_1 , F_2 , or $\hat{F}_{3,p}$ and \mathbf{w} is either the magnetic potential \mathbf{A} or the displacement \mathbf{u} (solution of the PDE $L_\rho(\mathbf{w}) = f$):

$$\begin{cases} \min_{\rho, \mathbf{w}} & J(\rho, \mathbf{w}) \\ \text{u.c.} & L_\rho(\mathbf{w}) = f \quad \text{and} \quad \rho \in L^\infty(\Omega_V, [0, 1]) \end{cases}$$

2. Introduction of a Lagrange multiplier λ associated with the constraint $L_\rho(\mathbf{w}) = f$ to obtain a Lagrangian function $\mathcal{L}(\rho, \mathbf{w}, \lambda)$.
3. Determination of the adjoint state \mathbf{w}_{adj} solution of an adjoint equation $L_\rho(\mathbf{w}_{\text{adj}}) = f_{\text{adj}}$ from the Karush–Khun–Tucker theorem (partial derivatives of the Lagrangian with regard to \mathbf{w} and λ equal to zero).
4. Computation of $\langle F'(\rho), \bar{\rho} \rangle = \left\langle \frac{\partial \mathcal{L}}{\partial \rho}(\rho, \mathbf{w}, \mathbf{w}_{\text{adj}}), \bar{\rho} \right\rangle$ yields the sensitivity of the objective function with respect to the design variable ρ ; derivation of the Lagrangian with regard to ρ ; and evaluation using the solutions \mathbf{w} and \mathbf{w}_{adj} of the direct and adjoint problems.

Now that the derivatives of the three objective functions have been derived, we can define in the next section the general algorithm that will be used to solve problem (P).

3.3. Optimization Algorithm

From a numerical point of view, the distribution $\rho : \Omega_V \rightarrow [0, 1]$ will need to be discretized. Since the magnetostatic and mechanical PDEs of Equations (3) and (8) are solved using the Finite Element Method (FEM), the variable domain Ω_V is already discretized into N open sets (tetrahedrons) T_i such that $\bar{\Omega}_V = \cup_{i=1}^N \bar{T}_i$ and $T_i \cap T_j = \emptyset$ for all $i \neq j$. We then naturally discretize ρ into a vector $\boldsymbol{\rho} = (\rho_1, \dots, \rho_N)^T$ of N values such that for all $\mathbf{x} \in \Omega_V$:

$$\rho(\mathbf{x}) = \sum_{i=1}^N \rho_i \mathbb{1}_{T_i}(\mathbf{x}). \quad (22)$$

Using this discretization, the sensitivity of any objective function F with regard to $\rho_i \in \mathbb{R}$ in the direction $\tilde{\rho}_i$ (to simplify, we use the same notation for F depending on either ρ or $\boldsymbol{\rho}$) is given by:

$$\langle F'(\boldsymbol{\rho}), \tilde{\rho}_i \rangle = \langle F'(\rho), \mathbb{1}_{T_i} \rangle \tilde{\rho}_i$$

where each term $\langle F'(\rho), \mathbb{1}_{T_i} \rangle$ is an integral on the set T_i . In order to solve problem (\mathcal{P}) , we will use the MMA gradient-based algorithm [18], which requires the following ingredients:

- The mechanical objective function (Equations (9) and (14)) and its associated gradient with respect to each ρ_i (see Equations (17), (18) and (22)).
- The magnetostatic constraint (Equation (5)) and its associated gradient with respect to each ρ_i (see Equations (15) and (22)).
- The constraints on $\boldsymbol{\rho}$ which are only the lower and upper bounds of the material density: $\boldsymbol{\rho} \in [0, 1]^N$.

Like any local optimization algorithm based on gradient descent, this optimization module starts from a starting point denoted by $\boldsymbol{\rho}^{[0]}$, builds a sequence of points $\boldsymbol{\rho}^{[k]}$, and ends at $\boldsymbol{\rho}^*$. During the process of optimization, the penalty parameter n used in Equation (13) is increased when the objective function stagnates. This makes it possible to converge slowly but strongly to efficient design solutions [4,19,20].

Remark 2. Using this classical method, the final solution generally has a checkerboard patterns [3]. In order to make the solution more manufacturable, a regularization method must be added. Thus, in this work, a density filtering method is used. It consists of modifying the value of ρ used inside each tetrahedron into a weighted sum of its current value and the values of its neighboring tetrahedrons. The implementation details and associated modified gradient follow the procedure described in [3].

4. Numerical Examples

The aim of this section is to validate our methodology on a U-shaped magnetic circuit. We begin by giving all the geometrical parameters of the structure in Section 4.1. We then move on to the solution of the magnetostatic problem alone in Section 4.2. Finally in Sections 4.3 and 4.4 we solve the hybrid problem (\mathcal{P}) and discuss the results in detail.

4.1. Design Problem

We consider here a simple magnetic circuit, shown in the right of Figure 1, with one coil and two distinct variable zones which can be either air or a ferromagnetic material (iron). For this magnetic circuit, the design variable is the material distribution ρ which varies between its maximal value of 1 (corresponding to ferromagnetic material: $\mu_{\text{iron}} = 3000$ and $E_{\text{iron}} = 0.21$ GPa) and minimal value of 0 (corresponding to air: $\mu_{\text{air}} = 1$ and $E_{\text{air}} = 10^{-10} E_{\text{iron}}$; this value is not taken to be zero in order to avoid the degeneration of the mechanical solutions).

For this example, the magnetic circuit can be divided into four parts, as follows:

- A fixed part of the U-shaped iron structure $\Omega_I \setminus \Omega_V$ represented on Figure 1. Its dimensions are 23 cm \times 14 cm \times 2 cm, the thickness of the bottom hyper-rectangle containing the coil is 2 cm, and the thickness of both sides of the U-shaped circuit is 4 cm.
- A coil which is represented in yellow on Figure 1. It is made up from two bottom and top bars of 4 cm \times 0.4 cm \times 5 cm and two lateral bars of 2 cm \times 0.4 cm \times 5 cm which are linked by four curved bars. This coil is crossed by a current of 4 A and has 300 windings.
- Two design variable zones Ω_V which are hyper-rectangles of size 8 cm \times 4 cm \times 4 cm. Both of these two parts are discretized into 144,800 tetrahedrons corresponding to the finite element mesh, which contains air or ferromagnetic material. Both of these parts are subjected to a surface force of $\mathbf{g} = -1\mathbf{y}$ (N/m²) on a small surface of size 0.5 cm \times 0.5 cm, as we can see in Figure 3. Note that the mechanical functions F_2 and

$\hat{F}_{3,p}$ considered in the article are linearly dependent on the amplitude of the surface force \mathbf{g} . Therefore, the optimal design ρ^* obtained by minimizing one of these two functions does not depend on this amplitude. Thus, this amplitude has been arbitrarily set to 1 (N/m²).

- The target zone Ω_T represented in black on Figure 1 is discretized into small cubes of size 0.1 cm, as we can see on Figure 4. These cubes are centered in the air gap between the two variable hyper-rectangles. The target zone Ω_T is composed by $216 = 6 \times 6 \times 6$ cubes, where the magnetic flux density is evaluated.

To simplify time-consuming calculations, we consider the plane of antisymmetry of the magnetic flux density \mathbf{B} : (Oxy), and restrict ourselves to the half-circuit shown in Figure 4.

In order to solve the main problems, see Equations (3) and (8), and the adjoint problems, defined in Equations (16) and (20), we use the FEM software GetDP (version 3.5.0, 13 May 2022) [21] together with the finite element mesh generator Gmsh (version 4.12.2, 21 January 2024) [22,23].

For this example, the given magnetic flux density \mathbf{B}_0 in the target zone Ω_T is first computed using a sample ferromagnetic shape, as represented in Figure 4. This corresponds to the magnetic flux density produced by a bar of dimensions $8 \text{ cm} \times \frac{4}{3} \text{ cm} \times \frac{4}{3} \text{ cm}$.

In Sections 4.2–4.4, we aim to solve the following problems: (1) minimizing only the magnetostatic function F_1 without constraints and (2) minimizing the mechanical functions F_2 and $\hat{F}_{3,p}$, respectively, with constraints on the magnetostatic function F_1 .

To solve these topology optimization problems, a gradient-based algorithm is used. At each iteration of the optimization algorithm, the gradient is computed using the adjoint method and, hence, a descent step is obtained. In this paper, we choose to use a MATLAB (version R2021b) implementation of the MMA (method of moving asymptots) [18] solver, which is a nonlinear gradient-based optimization method. The inputs of the MMA are the objective function, the constraints, and their respective gradients.

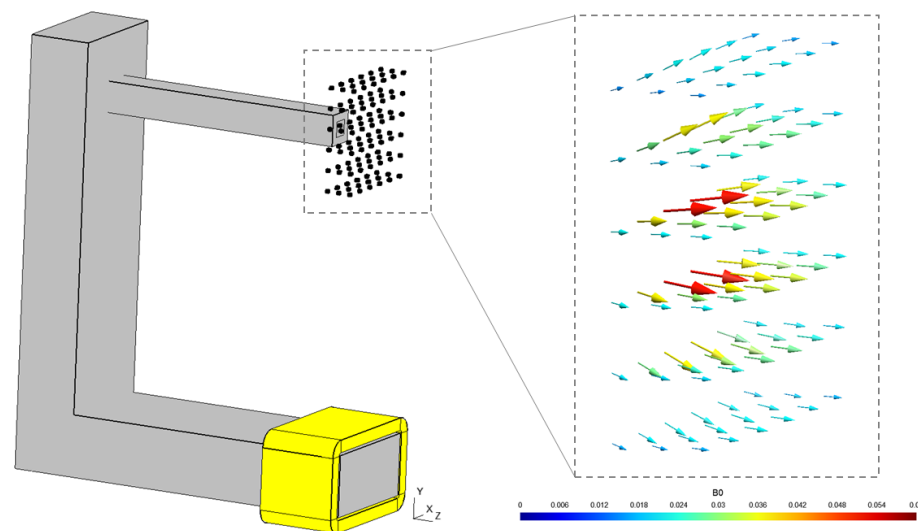


Figure 4. Material distribution providing the magnetic flux density \mathbf{B}_0 and zoom on this magnetic flux density \mathbf{B}_0 on the target zone.

4.2. Problem Considering Magnetostatics Only

In this section, the goal is to find the optimal distribution ρ_0^* (discretized via Equation (22)) which minimizes the function F_1 given by Equation (5):

$$(\mathcal{P}_1) \quad \rho_0^* = \arg \min_{\rho \in [0,1]^N} F_1(\rho) . \tag{23}$$

In our case, the variable zone Ω_V is discretized into $N = 144,800$ tetrahedrons. In order to find a design that looks like the bar in Figure 4 and which provides the target field \mathbf{B}_0 , it is essential to achieve large values of the penalty parameter n used by the SIMP method (see Equation (13)). This can be explained by the very large number of iterations required for the optimization method to converge, which comes from the fact that the gradient of the function F_1 in the region of Ω_V close to the fixed iron part is very small compared to the one in the area near Ω_T .

Note that we could have used a projection method such as one of the methods described in [24], but we have chosen to stick with the homotopy algorithm described in [9] where the penalization parameter is increased step by step. Therefore, the optimization problem is gradually forced to “chase” between 0 and 1 for each variable while ensuring that the objective function remains as low as possible.

Remark 3. *The choice of the penalty parameter n which appears in Equation (13) is challenging. Indeed, if n is very large at the start of the optimization, this leads to convergence towards a poor local minimum. Thus, we carry out a continuation process consisting of progressively increasing the penalty parameter starting from $n = 3$ and increasing it by 0.5 when, at an iteration $k + 1$, $\|F_1(\boldsymbol{\rho}^{[k+1]}) - F_1(\boldsymbol{\rho}^{[k]})\| \leq 10^{-4}$. Note that the optimization module is stopped after a fixed number of iterations (1000).*

The history of convergence is shown in Figure 5 starting from the initial distribution $\boldsymbol{\rho}^{[0]} = 0.95$ (an almost full material distribution). This optimization results in an optimal design $\boldsymbol{\rho}_0^*$ after 1000 iterations with a relative error between $\mathbf{B}(\boldsymbol{\rho}_0^*)$ and the target field \mathbf{B}_0 of no more than 1.5% inside the target area Ω_T , as shown by the boxplot in Figure 6 (right). Note that at the end of the optimization process, the penalty parameter reached a value $n > 50$.

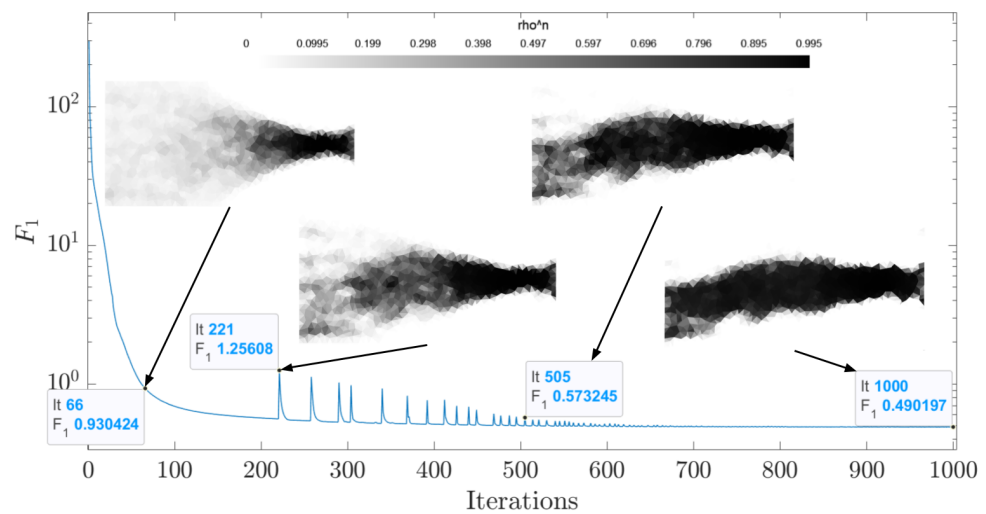


Figure 5. Evolution of the objective function F_1 during the optimization process described in Section 4.2 to solve problem (\mathcal{P}_1) . The small peaks correspond to the increase in the penalty parameter n .

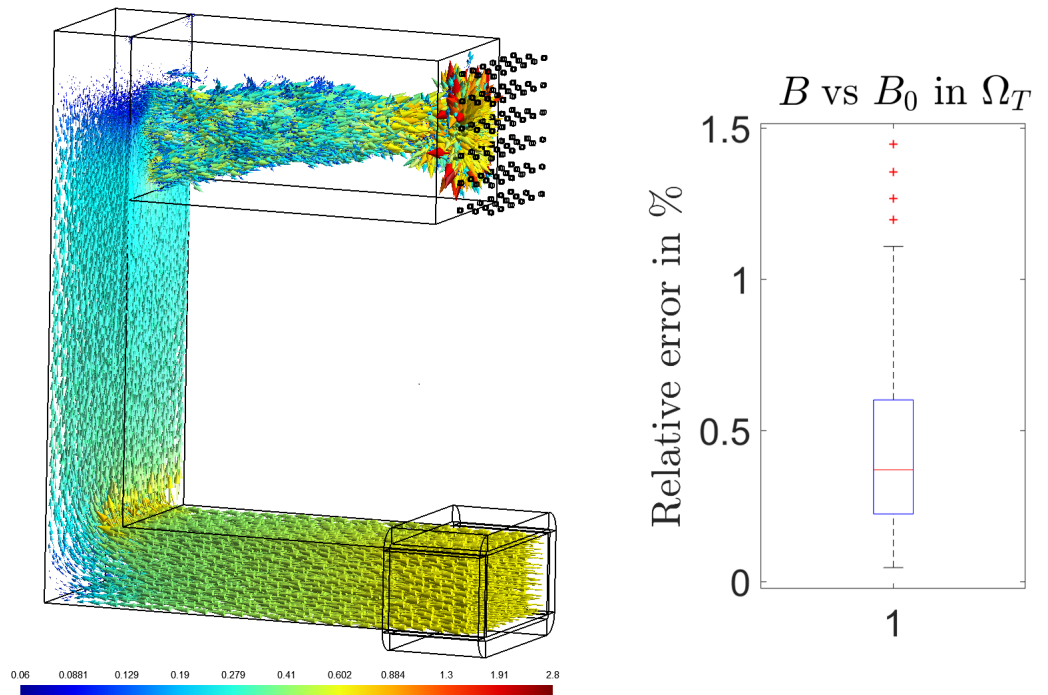


Figure 6. (left) Magnetic field inside the simulation domain for ρ_0^* . (right) Boxplot of the relative error $\|\mathbf{B}(\rho_0^*) - \mathbf{B}_0\| / \|\mathbf{B}_0\|$ between these two fields inside the 216 cubes which discretize Ω_T (as explained in Section 4.1).

Although this design is close to the optimal structure (the bar in Figure 4), it still contains a lot of intermediate densities ρ_i (corresponding to gray areas in Figure 6) near the fixed iron part Ω_I . As such, it is not mechanically rigid. Indeed, we can notice that the left-hand side of the final structure in Figure 5 does not fit perfectly with the rest of the magnetic circuit. We will now move on to the simultaneous optimization of both the magnetic circuit’s performance and its mechanical robustness.

4.3. Problem Combining Compliance Minimization and Magnetostatics

Now that the magnetic circuit has been optimized and a first design ρ_0^* has been obtained, we can move our attention to the solution of the coupled (magnetic and mechanical) problem (\mathcal{P}) . In this section, we thus consider the optimization of the following problem:

$$(\mathcal{P}_2) \begin{cases} \min_{\rho \in [0,1]^N} & F_2(\rho) \\ u.c. & F_1(\rho) \leq F_1(\rho_0^*) + \epsilon \end{cases} \quad (24)$$

in which F_2 is the compliance of the mechanical structure as defined in Equation (9) and ϵ is a small value taken such that $F_1(\rho_0^*) + \epsilon$ corresponds to an error of 3% for each cube in the target zone Ω_T . The same algorithm as in Section 4.2 was used following the continuation method explained in Remark 3 (except that only 500 iterations are necessary). Figure 7 shows the evolution of the objective function F_2 and constraint F_1 over the course of the optimization process.

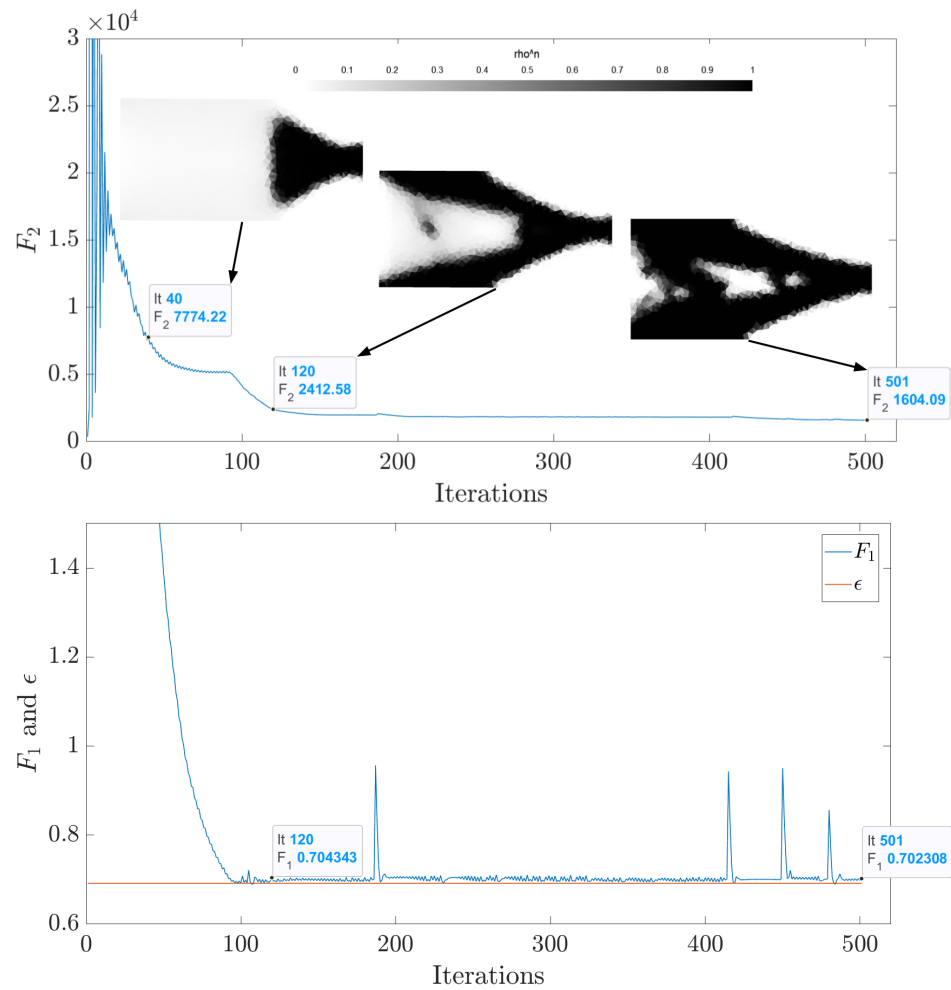


Figure 7. (top) Evolution of the objective function F_2 during the optimization process described in Section 4.3 to solve (P_2) . (bottom) Same for the constraint F_1 .

Comparing the evolution of the structure in both Figures 5 and 7 (top), we can see that both optimizations start by designing the right-hand side of the structure. However, after a hundred or so iterations, the two optimized structures begin to diverge significantly from each other, with the coupled optimization resulting in the appearance of two slabs connecting the right optimized area to the mechanically fixed iron part on the left. The final design is then close to the classical topology-optimized “cantilever” structure [1]. The resulting field from the optimal structure can be seen in Figure 8. The relative error inside Ω_T has naturally increased (the maximum error is now 5%) but its mechanical rigidity has greatly improved.

Minimizing compliance has increased the rigidity of the structure. Nevertheless, minimizing the compliance does not strictly ensure the mechanical strength of the part. Indeed, the real criteria for mechanical strength is the von Mises stress (for ductile materials): the material is robust if the von Mises stress does not exceed the yield point of the material. That is why we are now going to look at the direct optimization of the von Mises stress.

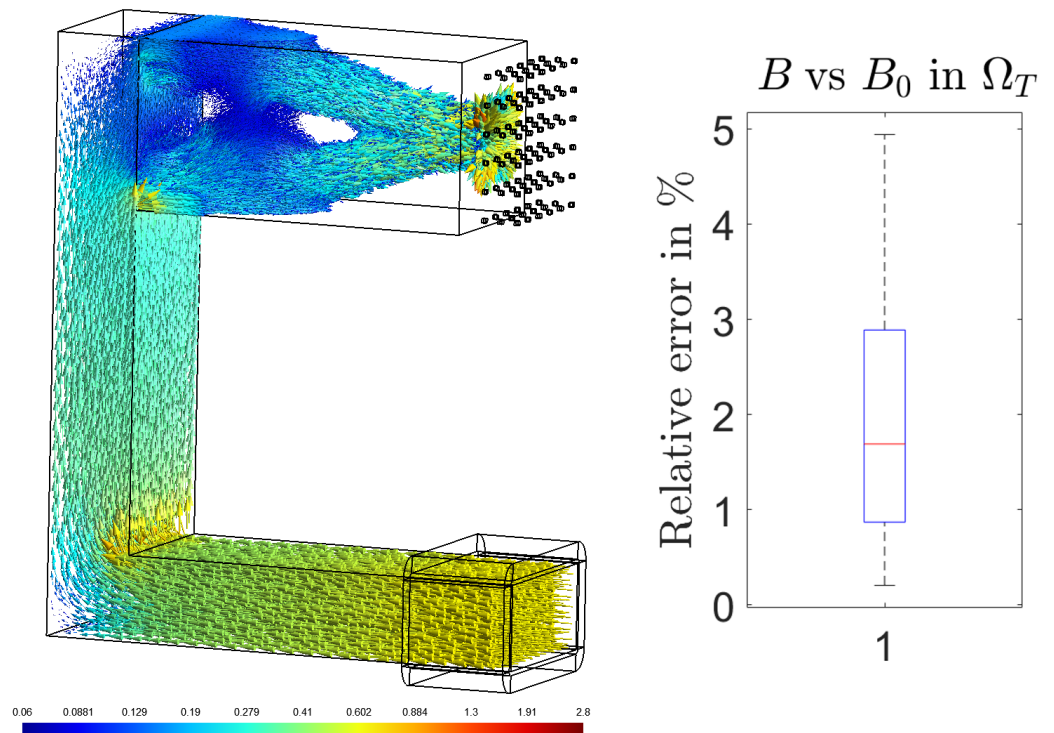


Figure 8. Same as in Figure 6 for the optimal structure resulting from the optimization of (24).

4.4. Problem Combining Global Stress Measure Minimization and Magnetostatics

In this section, we will solve the following optimization problem:

$$(\mathcal{P}_3) \quad \begin{cases} \min_{\boldsymbol{\rho} \in [0,1]^N} & \hat{F}_{3,p}(\boldsymbol{\rho}) \\ u.c. & F_1(\boldsymbol{\rho}) \leq F_1(\boldsymbol{\rho}_0^*) + \epsilon \end{cases} \quad (25)$$

where $\hat{F}_{3,p}$ is the global measure of the relaxed stress function defined in Equation (14). The parameter p is set equal to 6 and ϵ is the same value as in the previous Section 4.3.

Figure 9 shows the evolution of the objective function $\hat{F}_{3,p}$ and constraint F_1 , respectively, over the course of the iterations. The field associated with the final design is shown in Figure 10.

Once again, we can see significant differences between this design and the one previously obtained for the compliance minimization with a much finer shape joining the top and bottom slabs. We now have a relative magnetic error of no more than 4% in the target zone Ω_T . In detail, Figure 11 shows the von Mises stress inside the optimized design of (\mathcal{P}_2) and (\mathcal{P}_3) . If the objective function is the compliance (F_2), then the stress measure is higher and more concentrated than the case where the objective function is the relaxed von Mises stress ($\hat{F}_{3,p}$), which leads to a more uniform distribution of $\rho\sigma_{VM}$.

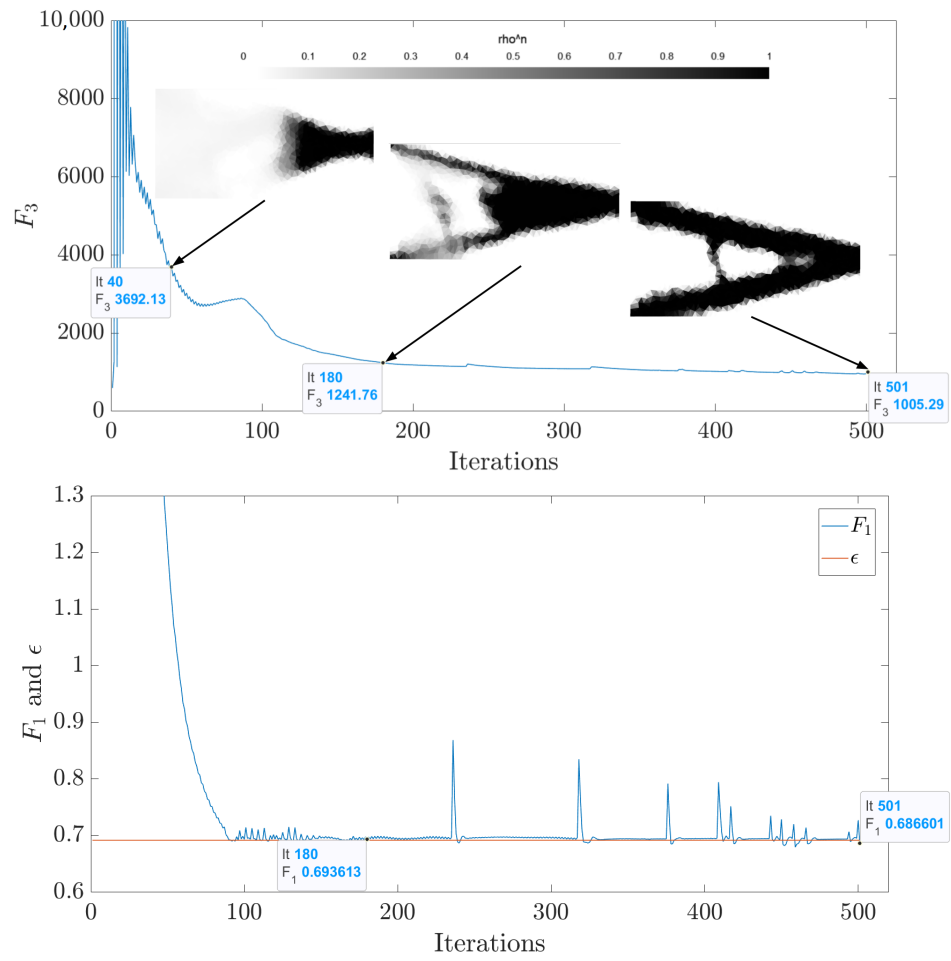


Figure 9. (top) Evolution of the objective function $\hat{F}_{3,p}$ during the optimization process described in Section 4.3 to solve (\mathcal{P}_2) . (bottom) Same for the constraint F_1 .

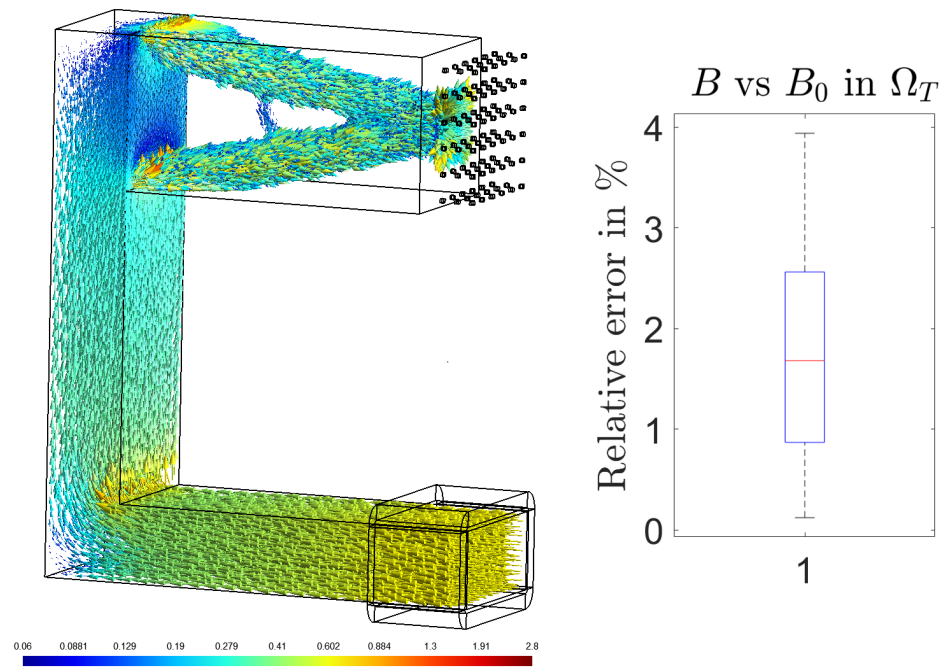


Figure 10. Same as in Figure 6 for the optimal structure resulting from the optimization of (25).

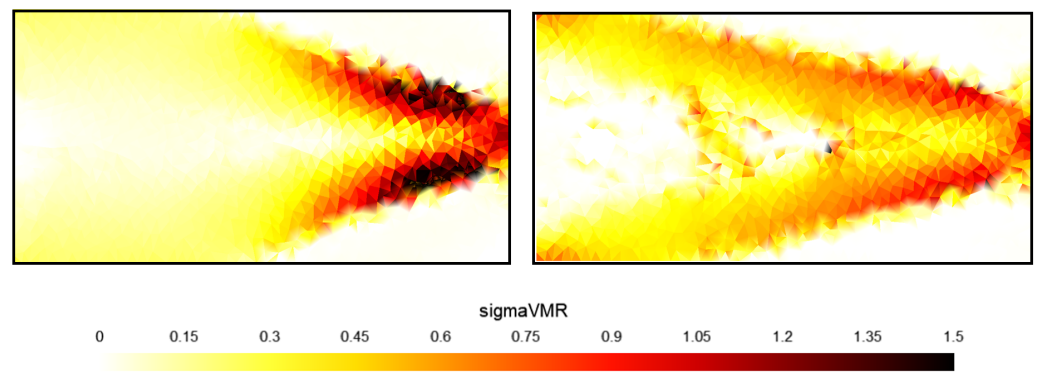


Figure 11. (left) Relaxed von Mises stress distribution of the solution of problem (24). (right) Same for the solution of problem (25).

5. Conclusions

In this paper, we have extended topology optimization based on the adjoint method and the SIMP methodology to address 3D design problems involving both magnetic and mechanical considerations. This coupled optimization proved interesting for achieving manufacturable designs, as a first optimization without taking mechanical considerations into account showed that the magnetic circuit could break if any force acted on it. We implemented two methods to maximize the mechanical rigidity of the structure. The first, based on the compliance of structure, resulted in a rigid design but with locally high stress. The second, based on the maximal von Mises stress, resulted in locally robust designs while maintaining good magnetic performance.

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Conflicts of Interest: The authors declare no conflicts of interest.

Appendix A. Proof of Propositions 2 and 3

The method summarized here is based on the *Céa method* introduced in [15], which is a formal method for calculating the sensitivity of a shape or topological optimization problem. Generally speaking, and by considering two functions $j : \mathbb{R} \times \mathbb{R}^3 \rightarrow \mathbb{R}$ and $l : \mathbb{R}^3 \rightarrow \mathbb{R}$, the optimization problem in terms of mechanic can be formulated as follows:

$$\min_{\rho} F(\rho) = J\left(\int_{\Omega_V} j(\rho, \mathbf{u}(\rho)) \, dx\right) + L\left(\int_{\Gamma_L} l(\mathbf{u}(\rho)) \, ds\right) \tag{A1}$$

with $\mathbf{u} \in V$ being the displacement solution of the following variational formulation (in the absence of volumetric forces) for all $\mathbf{v} \in V$:

$$\int_{\Omega_V} A^*(\rho) \bar{\epsilon}(\mathbf{u}) : \bar{\epsilon}(\mathbf{v}) \, dx - \int_{\Gamma_L} \mathbf{g} \cdot \mathbf{v} \, ds - \int_{\Omega_V} \mathbf{f}(\rho) \cdot \mathbf{v} \, dx = 0 \tag{A2}$$

with $A^*(\rho) = \left[\frac{E_{\text{air}}}{E_{\text{iron}}} + \left(1 - \frac{E_{\text{air}}}{E_{\text{iron}}}\right)g(\rho) \right] A(1)$.

We introduce the Lagrangian $\mathcal{L} : (\rho, \mathbf{u}, \mathbf{p}) \in L^\infty_{[0,1]}(\Omega) \times V \times V \mapsto \mathbb{R}$ defined by

$$\begin{aligned} \mathcal{L}(\rho, \mathbf{u}, \mathbf{p}) = & J\left(\int_{\Omega_V} j(\rho, \mathbf{u}) \, d\mathbf{x}\right) + L\left(\int_{\Gamma_L} l(\mathbf{u}) \, d\mathbf{s}\right) \\ & + \int_{\Omega_V} A^*(\rho)\bar{\mathbf{e}}(\mathbf{u}) : \bar{\mathbf{e}}(\mathbf{p}) \, d\mathbf{x} - \int_{\Gamma_L} \mathbf{g} \cdot \mathbf{p} \, d\mathbf{s} - \int_{\Omega_V} \mathbf{f}(\rho) \cdot \mathbf{p} \, d\mathbf{x}. \end{aligned} \tag{A3}$$

Note that $F(\rho) = \mathcal{L}(\rho, \mathbf{u}(\rho), \mathbf{p})$ for all $\mathbf{p} \in V$ if $\mathbf{u}(\rho)$ verifies the weak formulation of the elastic problem described in Equation (A2). In this case, the derivative of F with respect to ρ in the direction $\tilde{\rho}$ is written using the derivative chain rule:

$$\langle F'(\rho), \tilde{\rho} \rangle = \left\langle \frac{\partial \mathcal{L}}{\partial \rho}, \tilde{\rho} \right\rangle + \left\langle \frac{\partial \mathcal{L}}{\partial \mathbf{u}}, \mathbf{u}'(\rho)(\tilde{\rho}) \right\rangle.$$

Using the KKT conditions (the derivatives of \mathcal{L} with regard to \mathbf{u} and \mathbf{p} equal zero), we have for all $\mathbf{v} \in V$

$$\begin{aligned} \left\langle \frac{\partial \mathcal{L}}{\partial \mathbf{u}}, \mathbf{v} \right\rangle = & J'\left(\int_{\Omega_V} j(\rho, \mathbf{u}) \, d\mathbf{x}\right) \int_{\Omega_V} \frac{\partial j}{\partial \mathbf{u}}(\rho, \mathbf{u}) \cdot \mathbf{v} \, d\mathbf{x} \\ & + L'\left(\int_{\Gamma_L} l(\mathbf{u}) \, d\mathbf{s}\right) \int_{\Gamma_L} \frac{\partial l}{\partial \mathbf{u}}(\mathbf{u}) \cdot \mathbf{v} \, d\mathbf{s} + \int_{\Omega_V} A^*(\rho)\bar{\mathbf{e}}(\mathbf{v}) : \bar{\mathbf{e}}(\mathbf{p}) \, d\mathbf{x} = 0. \end{aligned} \tag{A4}$$

We have established the adjoint equation, whose solution is the adjoint variable \mathbf{p} . On the other hand, Equation (A4) is valid for any $\mathbf{v} \in V$, in particular for $\mathbf{v} = \mathbf{u}'(\rho)(\tilde{\rho})$; thus,

$$\langle F'(\rho), \tilde{\rho} \rangle = \left\langle \frac{\partial \mathcal{L}}{\partial \rho}, \tilde{\rho} \right\rangle$$

and we have

$$\left\langle \frac{\partial \mathcal{L}}{\partial \rho}, \tilde{\rho} \right\rangle = \left\langle \frac{\partial \mathcal{L}_1}{\partial \rho}, \tilde{\rho} \right\rangle + \left\langle \frac{\partial \mathcal{L}_2}{\partial \rho}, \tilde{\rho} \right\rangle + \left\langle \frac{\partial \mathcal{L}_3}{\partial \rho}, \tilde{\rho} \right\rangle$$

where

$$\begin{aligned} \mathcal{L}_1 &= J\left(\int_{\Omega_V} j(\rho, \mathbf{u}) \, d\mathbf{x}\right) \\ \mathcal{L}_2 &= \int_{\Omega_V} A^*(\rho)\bar{\mathbf{e}}(\mathbf{u}) : \bar{\mathbf{e}}(\mathbf{p}) \, d\mathbf{x} \\ \mathcal{L}_3 &= - \int_{\Omega_V} \mathbf{f}(\rho) \cdot \mathbf{p} \, d\mathbf{x}. \end{aligned}$$

By developing each of the three terms

$$\begin{aligned} \left\langle \frac{\partial \mathcal{L}_1}{\partial \rho}, \tilde{\rho} \right\rangle &= J'\left(\int_{\Omega_V} j(\rho, \mathbf{u}) \, d\mathbf{x}\right) \int_{\Omega_V} \frac{\partial j}{\partial \rho}(\rho, \mathbf{u})\tilde{\rho} \, d\mathbf{x} \\ \left\langle \frac{\partial \mathcal{L}_2}{\partial \rho}, \tilde{\rho} \right\rangle &= \int_{\Omega_V} A^{*'}(\rho)\bar{\mathbf{e}}(\mathbf{u}) : \bar{\mathbf{e}}(\mathbf{p})\tilde{\rho} \, d\mathbf{x} \\ \left\langle \frac{\partial \mathcal{L}_3}{\partial \rho}, \tilde{\rho} \right\rangle &= - \int_{\Omega_V} \mathbf{f}'(\rho) \cdot \mathbf{p}\tilde{\rho} \, d\mathbf{x}. \end{aligned}$$

Finally, the sensitivity of the objective function of problem (A1) is

$$\begin{aligned} \langle F'(\rho), \tilde{\rho} \rangle = & J'\left(\int_{\Omega_V} j(\rho, \mathbf{u}) \, d\mathbf{x}\right) \int_{\Omega_V} \frac{\partial j}{\partial \rho}(\rho, \mathbf{u})\tilde{\rho} \, d\mathbf{x} \\ & + \int_{\Omega_V} A^{*'}(\rho)\bar{\mathbf{e}}(\mathbf{u}) : \bar{\mathbf{e}}(\mathbf{p})\tilde{\rho} \, d\mathbf{x} - \int_{\Omega_V} \mathbf{f}'(\rho) \cdot \mathbf{p}\tilde{\rho} \, d\mathbf{x} \end{aligned} \tag{A5}$$

with \mathbf{p} being the solution of the adjoint problem (A4).

Appendix A.1. Particular Case of Compliance F_2

In this case, we have $\mathbf{f} = \mathbf{0}$ and $j(\rho, \mathbf{u}(\rho)) = 0$. Thus, from Equation (A5), the sensitivity of the corresponding optimization problem is

$$\langle F'_2(\rho), \tilde{\rho} \rangle = \int_{\Omega_V} A^{*'}(\rho) \bar{\mathbf{e}}(\mathbf{u}) : \bar{\mathbf{e}}(\mathbf{p}) \tilde{\rho} \, dx$$

where \mathbf{p} is the adjoint variable solution of the adjoint problem. From Equation (A4), the adjoint system is written for all $\mathbf{v} \in V$ as follows:

$$\int_{\Gamma_L} \mathbf{g} \cdot \mathbf{v} \, ds + \int_{\Omega_V} A^*(\rho) \bar{\mathbf{e}}(\mathbf{v}) : \bar{\mathbf{e}}(\mathbf{p}) \, dx = 0. \tag{A6}$$

Clearly, $\mathbf{p} = -\mathbf{u}$ is the solution of Equation (A6). Thus, the compliance minimization problem is self-adjoint. Finally, the sensitivity of function F_2 is

$$\langle F'_2(\rho), \tilde{\rho} \rangle = - \int_{\Omega_V} A^{*'}(\rho) \bar{\mathbf{e}}(\mathbf{u}) : \bar{\mathbf{e}}(\mathbf{u}) \tilde{\rho} \, dx \tag{A7}$$

where $A^{*'}(\rho) = \left[\frac{E_{\text{air}}}{E_{\text{iron}}} + (1 - \frac{E_{\text{air}}}{E_{\text{iron}}}) g'(\rho) \right] A(1)$.

Appendix A.2. Particular Case of Global Stress Measure $\hat{F}_{3,p}$:

In this case, we have

$$L \left(\int_{\Gamma_L} l(\mathbf{u}(\rho)) \, ds \right) = 0 \tag{A8}$$

$$\text{and } J \left(\int_{\Omega_V} j(\rho, \mathbf{u}(\rho)) \, dx \right) = \left(\int_{\Omega_V} [\rho^m \sigma_{VM}(1, \mathbf{u}(\rho))]^p \, dx \right)^{\frac{1}{p}} \tag{A9}$$

where $\mathbf{f} = \mathbf{0}$ as well as $j(\rho, \mathbf{u}(\rho)) = [\rho^m \sigma_{VM}(1, \mathbf{u}(\rho))]^p$. From (A4), the adjoint problem is given for all $\mathbf{v} \in V$ as follows:

$$J' \left(\int_{\Omega_V} j(\rho, \mathbf{u}) \, dx \right) \int_{\Omega_V} \frac{\partial j}{\partial \mathbf{u}}(\rho, \mathbf{u}) \cdot \mathbf{v} \, dx + \int_{\Omega_V} A^*(\rho) \bar{\mathbf{e}}(\mathbf{v}) : \bar{\mathbf{e}}(\mathbf{p}) \, dx = 0 \tag{A10}$$

where

$$J' \left(\int_{\Omega_V} j(\rho, \mathbf{u}) \, dx \right) \int_{\Omega_V} \frac{\partial j}{\partial \mathbf{u}}(\rho, \mathbf{u}) \cdot \mathbf{v} \, dx = \frac{1}{p} \left(\int_{\Omega_V} j(\rho, \mathbf{u}) \, dx \right)^{\frac{1}{p}-1} \int_{\Omega_V} \frac{\partial j}{\partial \mathbf{u}}(\rho, \mathbf{u}) \cdot \mathbf{v} \, dx.$$

All that remains now is to develop the term $\int_{\Omega_V} \frac{\partial j}{\partial \mathbf{u}}(\rho, \mathbf{u}) \cdot \mathbf{v} \, dx$:

$$\int_{\Omega_V} \frac{\partial j}{\partial \mathbf{u}}(\rho, \mathbf{u}) \cdot \mathbf{v} \, dx = \int_{\Omega_V} \rho^{mp} p \sigma_{VM}^{p-1}(1, \mathbf{u}) \left\langle \frac{\partial \sigma_{VM}}{\partial \mathbf{u}}, \mathbf{v} \right\rangle \, dx$$

In order to develop the term $\left\langle \frac{\partial \sigma_{VM}}{\partial \mathbf{u}}, \mathbf{v} \right\rangle$, we introduce Voigt notation for $\bar{\sigma}$ constraints, denoted as $\vec{\sigma} = (\sigma_{xx}, \sigma_{yy}, \sigma_{zz}, \sigma_{xy}, \sigma_{yz}, \sigma_{zx})^T$. This notation allows us to express the stress tensor $\bar{\sigma}$ using a 6-element vector sufficient to characterize the stresses (by symmetry of the stress tensor $\bar{\sigma}$). Thus, we can see from the definition of the von Mises constraints that:

$$\sigma_{VM} = \sigma_{VM}(1, \vec{\sigma}(\mathbf{u})).$$

Then, by chain derivation,

$$\int_{\Omega_V} \frac{\partial j}{\partial \mathbf{u}}(\rho, \mathbf{u}) \cdot \mathbf{v} \, d\mathbf{x} = \int_{\Omega_V} \rho^{mp} p \sigma_{VM}^{p-1}(1, \mathbf{u}) \frac{\partial \sigma_{VM}}{\partial \vec{\sigma}}^T \left\langle \frac{\partial \vec{\sigma}}{\partial \mathbf{u}}, \mathbf{v} \right\rangle \, d\mathbf{x}$$

Using the definition of the directional derivative and the linearity of Hooke’s law in Voigt notation, we have

$$\left\langle \frac{\partial \vec{\sigma}}{\partial \mathbf{u}}, \mathbf{v} \right\rangle = \vec{\sigma}(\mathbf{v}).$$

Finally,

$$\int_{\Omega_V} \frac{\partial j}{\partial \mathbf{u}}(\rho, \mathbf{u}) \cdot \mathbf{v} \, d\mathbf{x} = \int_{\Omega_V} \rho^{mp} p \sigma_{VM}^{p-1}(1, \mathbf{u}) \frac{\partial \sigma_{VM}}{\partial \vec{\sigma}}^T \vec{\sigma}(\mathbf{v}) \, d\mathbf{x}$$

where $\frac{\partial \sigma_{VM}}{\partial \vec{\sigma}}$ is given by (21). Thus, the final expression of the adjoint problem is given for all $\mathbf{v} \in V$ as follows:

$$\left(\int_{\Omega_V} \rho^{mp} \sigma_{VM}^p(1, \mathbf{u}) \, d\mathbf{x} \right)^{\frac{1}{p}-1} \int_{\Omega_V} \rho^{mp} \sigma_{VM}^{p-1}(1, \mathbf{u}) \frac{\partial \sigma_{VM}}{\partial \vec{\sigma}}^T \vec{\sigma}(\mathbf{v}) \, d\mathbf{x} + \int_{\Omega_V} A^*(\rho) \bar{\mathbf{e}}(\mathbf{v}) : \bar{\mathbf{e}}(\mathbf{p}) \, d\mathbf{x} = 0. \quad (\text{A11})$$

From the expression of the derivative due to Equation (A5), and by using the following development, we obtain:

$$\begin{aligned} J' \left(\int_{\Omega_V} j(\rho, \mathbf{u}) \, d\mathbf{x} \right) \int_{\Omega_V} \frac{\partial j}{\partial \rho}(\rho, \mathbf{u}) \tilde{\rho} \, d\mathbf{x} &= \frac{1}{p} \left(\int_{\Omega_V} j(\rho, \mathbf{u}) \, d\mathbf{x} \right)^{\frac{1}{p}-1} \int_{\Omega_V} mp \sigma_{VM}^p(1, \mathbf{u}) \rho^{mp-1} \tilde{\rho} \, d\mathbf{x} \\ &= \left(\int_{\Omega_V} \rho^{mp} \sigma_{VM}^p(1, \mathbf{u}) \, d\mathbf{x} \right)^{\frac{1}{p}-1} \int_{\Omega_V} m \sigma_{VM}^p(1, \mathbf{u}) \rho^{mp-1} \tilde{\rho} \, d\mathbf{x}. \end{aligned}$$

The sensitivity of the $\hat{F}_{3,p}$ function is:

$$\begin{aligned} \left\langle \hat{F}'_{3,p}(\rho), \tilde{\rho} \right\rangle &= \int_{\Omega_V} A^{*'}(\rho) \bar{\mathbf{e}}(\mathbf{u}) : \bar{\mathbf{e}}(\mathbf{p}) \tilde{\rho} \, d\mathbf{x} \\ &+ \left(\int_{\Omega_V} \rho^{mp} \sigma_{VM}^p(1, \mathbf{u}) \, d\mathbf{x} \right)^{\frac{1}{p}-1} \int_{\Omega_V} m \sigma_{VM}^p(1, \mathbf{u}) \rho^{mp-1} \tilde{\rho} \, d\mathbf{x}. \quad (\text{A12}) \end{aligned}$$

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