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Shape optimization and optimal control for transient heat conduction problems using an isogeometric approach

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Abstract

This work is concerned with the development of a framework to solve shape optimization problems for transient heat conduction problems within the context of isogeometric analysis (IGA). A general objective functional is used to accommodate both shape optimization and passive control problems under transient conditions. An adjoint sensitivity analysis, which accounts for possible discontinuities in the objective functional, is performed analytically and subsequently discretized within the context of IGA. The gradient of the objective functional is used in a descent algorithm to solve optimization problems. Numerical examples are presented to validate and demonstrate the capacity to manage thermal fields under transient conditions.

Keywords: Shape optimization, Isogeometric analysis, Transient heat conduction, Passive optimal control, Adjoint method

1. introduction

Engineering devices used for thermal management in applications where the temperature and/or the heat exchanged play an important role, such as heat exchangers, cooling components, heat sinks or thermal protection layers, are typically designed to control the maximum or minimum temperature that a system should be exposed to or to guarantee a certain heat exchange rate. Given some performance requirements, a judicious choice of materials and shapes is made and the original design is subsequently optimized to improve its performance. This optimization process is typically based on steady-state conditions. Nevertheless, in many technologically-relevant applications, thermal conditions fluctuate during operation. In these cases, an active control system is sometimes used for thermal management. The drawback of an active control system is its additional cost, which may hinder its usefulness. An attractive alternative is passive control, which relies on a suitably-designed system that takes a priori into account fluctuations in the thermal fields. In particular, passive control may be achieved through a shape optimization procedure that finds an optimal shape under transient heat conduction conditions.

Shape optimization has received renewed attention due to development of Isogeometric Analysis (IGA) as an alternative to the (classical) finite-element method (FEM) [1]. The main advantage of IGA is that the numerical analysis is carried out with the same shape functions used in most commercially-available Computer-Aided Design (CAD) programs. Consequently, the bottleneck of converting a CAD-generated design into an FEM-ready mesh is avoided [2]. CAD

software typically uses non-uniform rational B-splines (NURBS) [3] to represent the design geometry. In fact, prior to the development of IGA, shape optimization has also relied on NURBS for design purposes (see, e.g., Braibant and Fleury [4], Espath et al. [5] and Wang and Zhang [6]). More recently, the integration of IGA and NURBS-based shape optimization has been exploited to improve the overall efficiency and accuracy of the method, particularly in terms of an unified design and analysis parametrization that provides an enhanced sensitivity analysis (see, e.g., Wall et al. [7], Qian [8]).

Isogeometric shape optimization has been used for curved beam structures in Nagy et al. [9] and Nagy et al. [10], vibrating membranes in Manh et al. [11], fluid mechanics in Nørtoft and Gravesen [12], pulsatile ventricular assist devices in Long et al. [13], shells in Nagy et al. [14], Kiendl et al. [15] and Ha [16], photonic crystals in Qian and Sigmund [17], composite fiber orientation in [18, 19], heat conduction problems in Yoon et al. [20], Stokes flow problems in Park et al. [21]. Isogeometric shape optimization using boundary element method is presented in Li and Qian [22] and Lian et al. [23]. The aforementioned contributions illustrate the wide range of applications of the isogeometric shape optimization approach. However, the method has been limited to static or steady-state conditions. Recently, the isogeometric shape optimization method was extended in Wang and Turteltaub [24] to analyze quasi-static processes with slowly-varying external loads.

In the present work, one goal is to extend the isogeometric approach to carry out shape design and passive control of problems governed by a transient behavior, in particular transient heat conduction. Previous contributions related to inverse shape design work for transient heat conduction, which have relied mostly on a classical FEM analysis, can be found in Dulikravich [25], Jarny et al. [26], Huang and Chaing [27] and Korycki [28]. Shape sensitivity analysis for the linear or nonlinear transient heat conduction problems can be found in Haftka [29], Tortorelli and Haber [30], Tortorelli et al. [31], Słuzalec and Kleiber [32], Kleiber and Słuzalec [33], Dems and Rousselet [34, 35], Gu et al. [36] and Korycki [37]. Shape optimization work of thermoelastic problems with transient thermal field can be found in Kane et al. [38], Gao and Grandhi [39] and Song et al. [40]. Other related design sensitivity analysis may be found in Haftka and Grandhi [41], Van Keulen et al. [42], Choi and Kim [43] and Nanthakumar et al. [44, 45].

A second goal in the present work is to extend the design sensitivity for situations where the objective functional is measured only in selected areas of the design region and during selected intervals of the design time period. This modeling feature is useful in cases where only portions of the design region and/or specific time intervals of the analysis period are relevant for the overall shape design. In that context, the sensitivity analysis requires an extended version of the transport theorems to account for discontinuities in the objective functional.

The structure of the present work is as follows: the general formulation of the shape optimal design and passive control considering the jump conditions in the objective functional is presented in a continuous setting in Sec. 2 (i.e., continuous description of the design and analysis spaces). By assigning a Lagrange multiplier for the strong form of the governing equations at every material point during the transient time interval, the continuous adjoint sensitivity analysis is developed in Sec. 3. Necessary details are included to treat the discontinuities of the characteristic functions that represent the regions (in space and/or time) where the objective functional is defined. The isogeometric analysis and design discretization is discussed briefly in Sec. 4. The design problem and its sensitivity are subsequently discretized in Sec. 5 within the context of IGA. This section also includes the algorithm used to numerically solve shape optimization problems. To validate the methodology, two benchmark problems, namely a minimum

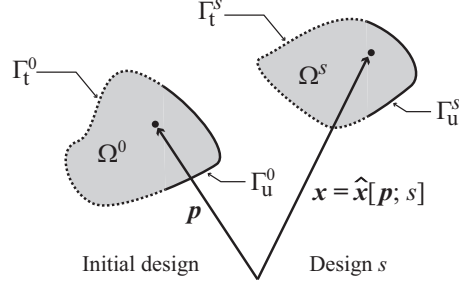


Figure 1: Family of design domains Ω^s generated through design functions $\hat{\mathbf{x}}[\cdot; s]$.

surface problem and a passive temperature control problem at pre-determined time are presented in Sec. 6.1. To further illustrate the range of possible applications, two examples are presented in Sec. 6.2 and Sec. 6.3, namely the shape design of a plunger for a molten glass forming die and a thermal protection system (TPS) for a re-entry ballistic vehicle nose. Finally, some concluding remarks are given in Sec. 7.

2. Problem statement

2.1. Design function and heat conduction problem

Consider an isotropic, homogeneous and linearly thermally-conducting material that occupies a region Ω^s with boundary Γ^s as shown in Fig. 1. The superscript s is a continuous scalar parameter that represents different configurations (or states) of the region. Each configuration corresponds to a design of the structure. The state of the region with $s = 0$, i.e. Ω^0 , is called the referential or initial design. A material design point $\mathbf{p} \in \Omega^0$ is mapped to a position $\mathbf{x} = \hat{\mathbf{x}}[\mathbf{p}, s] \in \Omega^s$ by a design function $\hat{\mathbf{x}}$ as indicated in Fig. 1. For simplicity, the mapping $\hat{\mathbf{x}}$ is henceforth also denoted as \mathbf{x} and the meaning of the symbol may be inferred from the context (i.e., position or design mapping).

In the domain Ω^s and during a time interval $\mathcal{T} = [0, T]$, with T a given final analysis time, the governing equation for a transient heat conduction problem can be expressed as

$$l[\theta[\mathbf{x}, t]] := \rho c \frac{\partial \theta[\mathbf{x}, t]}{\partial t} - k \nabla^2 \theta[\mathbf{x}, t] - Q[\mathbf{x}, t] = 0, \quad (\mathbf{x}, t) \in \Omega^s \times \mathcal{T}, \quad (1)$$

where $\theta = \theta[\mathbf{x}, t]$ is the temperature at point \mathbf{x} and time t , $c > 0$ is the heat capacity, $\rho > 0$ is the mass density, $k > 0$ is the thermal conductivity, $Q = Q[\mathbf{x}, t]$ is the inner heat generation rate per unit volume (volumetric heat supply) and ∇^2 is the Laplacian operator. For subsequent use, the governing equation is written in terms of an operator l as defined in (1). It is assumed that there are only three types of boundary conditions on the boundary Γ^s , which is divided as follows:

$$\Gamma^s = \Gamma_\theta^s \cup \Gamma_q^s \cup \Gamma_e^s$$

where Γ_θ^s represents portions where the temperature is specified, Γ_q^s corresponds to portions where the contact heat supply is given and Γ_e^s is the part of the boundary where heat is exchanged with

the environment through convection. The boundary and initial conditions are as follows:

$$\begin{aligned}
\theta &= \hat{\theta} & \text{on } \Gamma_{\theta}^s \times \mathcal{T} \\
\mathbf{q} \cdot \mathbf{n} &= -\hat{q} & \text{on } \Gamma_{\mathbf{q}}^s \times \mathcal{T} \\
\mathbf{q} \cdot \mathbf{n} &= -q_e = h(\theta - \theta_e) & \text{on } \Gamma_e^s \times \mathcal{T} \\
\theta[\mathbf{x}, 0] &= \theta_0[\mathbf{x}] & \text{in } \Omega^s
\end{aligned} \tag{2}$$

with the heat flux vector \mathbf{q} given by Fourier's model, i.e.,

$$\mathbf{q} = -k\nabla\theta.$$

In (2), $\hat{\theta} = \hat{\theta}[\mathbf{x}, t]$ and $\hat{q} = \hat{q}[\mathbf{x}, t]$ are the specified temperature and contact heat supply on $\Gamma_{\theta}^s \times \mathcal{T}$ and $\Gamma_{\mathbf{q}}^s \times \mathcal{T}$, respectively, $\mathbf{n} = \mathbf{n}[\mathbf{x}]$ is the unit outward normal vector to the boundary, h is the convection coefficient, $\theta_e = \theta_e[t]$ is the convective exchange temperature (ambient temperature) and $\theta_0[\mathbf{x}]$ is the initial temperature field in the domain. For notational convenience, the contact heat supply on Γ_e^s is denoted as q_e . The contact heat supplies \hat{q} and q_e are given as the negative of the normal heat flux, hence they take positive values if heat flows from the external environment into the system and negative values otherwise.

2.2. Transient shape optimization problem

In order to develop a versatile framework for shape optimization problems, which can accommodate various situations within a unified formulation, a general objective functional is defined such that it can measure the performance on pre-selected parts where the physical phenomenon occurs, both in space and time. This is achieved through the combination of two local performance functions, denoted as ψ_{ω} and ψ_{γ} and defined, respectively, in Ω^s and Γ^s , and three characteristic functions, denoted as ω , γ and ζ , that specify, respectively, the (sub)-parts of the region Ω^s , boundary Γ^s and time interval \mathcal{T} where the performance is monitored. In particular, the overall objective functional \mathcal{J} is defined as

$$\mathcal{J}[s] := \int_0^T \zeta[t] \Psi[t; s] dt \quad \text{with} \quad \Psi[t; s] := \Psi_{\omega}[t; s] + \Psi_{\gamma}[t; s] \tag{3}$$

where Ψ_{ω} is a functional defined over the region Ω^s as

$$\Psi_{\omega}[t; s] := \int_{\Omega^s} \omega[\mathbf{p}, t] \psi_{\omega}[\theta[\mathbf{x}, t; s]] d\Omega, \tag{4}$$

and Ψ_{γ} is a functional defined on the boundary Γ^s as

$$\Psi_{\gamma}[t; s] := \int_{\Gamma^s} \gamma[\mathbf{p}, t] \psi_{\gamma}[\theta[\mathbf{x}, t; s], q[\mathbf{x}, t; s]] d\Gamma. \tag{5}$$

The functions ψ_{ω} and ψ_{γ} , appearing in (4) and (5), are the local performance functions. The function ψ_{ω} measures the performance in Ω^s based on the temperature field θ while the function ψ_{γ} measures the performance on Γ^s based on the temperature θ and the contact heat supply $q = -\mathbf{q} \cdot \mathbf{n}$, where \mathbf{n} is the outward unit vector on the boundary. The parameter s in the argument of the functions and functionals shown in the equations above indicate that these are evaluated for a given design function $\hat{\mathbf{x}}[\cdot, s]$. The material design point \mathbf{p} used in the argument of the

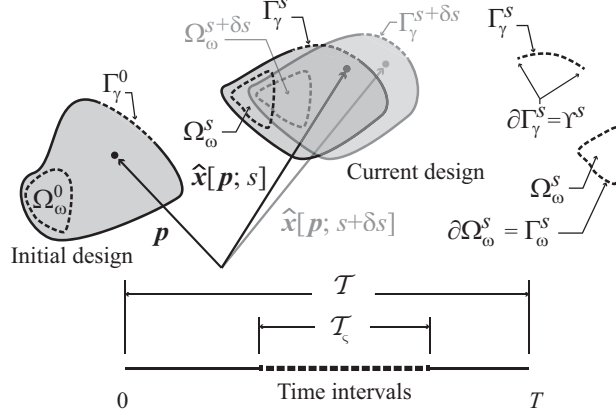


Figure 2: Illustration of subdomain Ω_ω^s , boundary portion Γ_γ^s and time subinterval \mathcal{T}_ζ where the objective functional \mathcal{J} is measured.

characteristic functions ω and γ should be interpreted as $\mathbf{p} = \hat{\mathbf{p}}[\mathbf{x}, s]$, where $\hat{\mathbf{p}}$ is the inverse of the design function $\hat{\mathbf{x}}$. Following the approach proposed in [43] and [24], the functions ω and γ are associated with, respectively a (sub)domain $\Omega_\omega^s \subseteq \Omega^s$ and a portion of the boundary $\Gamma_\gamma^s \subseteq \Gamma^s$ in which the performance function is measured.

The functions ω and γ correspond to fixed design points in the reference design domain. Correspondingly, the subdomain $\Omega_\omega^s \subseteq \Omega^s$ and the portion $\Gamma_\gamma^s \subseteq \Gamma^s$ are assumed to be obtained through a mapping of the referential subdomain $\Omega_\omega^0 \subseteq \Omega^0$ and the referential portion $\Gamma_\gamma^0 \subseteq \Gamma^0$. For increased design flexibility, the subdomain Ω_ω^0 and the portion Γ_γ^0 may themselves be prescribed as functions of time. Hence, the characteristic functions are defined as

$$\omega[\mathbf{p}, t] = \begin{cases} 1, & \text{if } \mathbf{p} \in \Omega_\omega^0(t) \\ 0, & \text{otherwise} \end{cases}, \quad \gamma[\mathbf{p}, t] = \begin{cases} 1, & \text{if } \mathbf{p} \in \Gamma_\gamma^0(t) \\ 0, & \text{otherwise} \end{cases}. \quad (6)$$

Similarly, the characteristic function ζ that appears in (3) is associated with a time (sub)-interval $\mathcal{T}_\zeta \subseteq \mathcal{T}$ such that

$$\zeta[t] = \begin{cases} 1, & \text{if } t \in \mathcal{T}_\zeta \\ 0, & \text{otherwise} \end{cases}. \quad (7)$$

In a typical optimization problem, the design may be subjected to a resource constraint that limits the total volume of the design region, i.e.,

$$\Sigma[s] := \int_{\Omega^s} d\Omega \leq \Sigma_0, \quad (8)$$

where Σ_0 is a given upper bound. In addition, it is often required to consider constraints of the following form:

$$\hat{d}_c[\mathbf{x}, s] \leq 0 \quad c = 1, \dots, N_c, \quad (9)$$

where the scalar functions \hat{d}_c represent pointwise constraints (such as upper or lower bounds) and N_c is the total number of constraints. The optimization problem consists on finding an admissible design function $\hat{\mathbf{x}}[\cdot, s^*]$, which satisfies the constraints (8) and (9), such that $\mathcal{J}[s^*] = \min_s \mathcal{J}[s]$.

3. Continuous adjoint sensitivity analysis

3.1. Material and spatial design derivatives

Continuous shape modifications in a shape optimization problem can be understood as a smooth sequence of deformations from an initial (referential) design as shown in Fig. 1. The dependence of functions and functionals on the design function $\hat{\mathbf{x}}[\cdot, s]$ is indicated with a dependence on the design parameter s . Similar to the notion of material and spatial time derivatives used in continuum mechanics, it is possible to define a material design derivative and a spatial design derivative. The material design derivative (at constant \mathbf{p}) is denoted here with either a superimposed open ring, $\dot{(\cdot)}$, or with the operator $D(\cdot)/D_s$ (also known as the total design derivative), while the spatial design derivative (at constant \mathbf{x}) is indicated using an apostrophe, $(\cdot)'$, or using the operator $\partial(\cdot)/\partial s$ (partial design derivative). For functions that only depend on s , the derivative with respect to s is indicated as $d(\cdot)/ds$. The material and spatial design derivatives of a function \mathbf{h} are related as follows:

$$\dot{\mathbf{h}} = \frac{D\mathbf{h}}{D_s} = \mathbf{h}' + (\nabla\mathbf{h})\mathbf{v}, \quad (10)$$

where ∇ refers to the gradient with respect to \mathbf{x} and \mathbf{v} is the so-called design velocity. The design velocity of a material design point \mathbf{p} is the material design derivative of the design function $\mathbf{x} = \hat{\mathbf{x}}[\mathbf{p}, s]$, i.e.

$$\mathbf{v} := \dot{\mathbf{x}} = \frac{D\mathbf{x}}{D_s}. \quad (11)$$

The spatial design derivative and the spatial gradient commute, i.e.,

$$(\nabla\mathbf{h})' = \nabla(\mathbf{h}') = \nabla\mathbf{h}'. \quad (12)$$

3.2. Transport relations

For a volume integral of a generic smooth function f defined in Ω^s and a characteristic function ω as given in (6), the volume transport theorem is extended as

$$\frac{d}{ds} \int_{\Omega^s} \omega f d\Omega = \int_{\Omega^s} \omega f' d\Omega + \int_{\Gamma^s} \omega f v_n d\Gamma + \int_{S_\omega^s} \llbracket \omega \rrbracket f v_n d\Gamma, \quad (13)$$

where $v_n = \mathbf{v} \cdot \mathbf{n}$ is the normal design velocity with \mathbf{n} the outward normal vector on the boundary, and $S_\omega^s = \partial\Omega_\omega^s - \Gamma^s$ is the interior boundary of the sub-region (see [46]). The notation $\llbracket \cdot \rrbracket = (\cdot)^+ - (\cdot)^-$ indicates the jump of a quantity across a surface/line/point of discontinuity, with $(\cdot)^\pm$ representing the values on the \pm sides according to the normal vector (pointing towards the +side).

For a boundary integral of a generic smooth function f defined on Γ^s and a characteristic function γ as given in (6), the boundary transport theorem is extended as

$$\frac{d}{ds} \int_{\Gamma^s} \gamma f d\Gamma = \int_{\Gamma^s} (\gamma f' + \gamma (\nabla f \cdot \mathbf{n}) v_n - \kappa f v_n) d\Gamma + \int_{\Upsilon^s} \llbracket \gamma \rrbracket f v_m d\Upsilon, \quad (14)$$

where κ is the total curvature of Γ^s (twice the mean curvature of a two-dimensional surface embedded in three space dimensions), $\Upsilon^s = \partial\Gamma_\gamma^s$ is the edge of the boundary Γ_γ^s , $v_m = \mathbf{v} \cdot \mathbf{m}$, with \mathbf{m} being the normal vector pointing outward of $\partial\Gamma_\gamma^s$ and div_Γ refers to the surface divergence (see [47], [48] and [49]).

3.3. Augmented functionals

The optimal design process is to find a design such that the objective functional is minimized (or maximized) while simultaneously satisfying all the design constraints. The governing equation of the transient heat conduction problem shown in (1) and the associated load and boundary conditions in (2) can be treated as an equality constraint in variational form.

In order to include the constraints in the sensitivity analysis, it is convenient to introduce an augmented objective functional (Lagrangian) defined as

$$L[s, \Lambda_c, \Lambda_\Sigma] := \tilde{\mathcal{J}}[s] + \sum_{c=1}^{N_c} \langle \Lambda_c, \hat{d}_c[s] \rangle_{\Omega^s} + \Lambda_\Sigma (\Sigma[s] - \Sigma_0) , \quad (15)$$

where

$$\tilde{\mathcal{J}}[s] = \mathcal{J}[s] + \int_0^T \langle l[\theta], \vartheta \rangle_{\Omega^s} dt. \quad (16)$$

In the preceding equations, the inner product $\langle \cdot, \cdot \rangle_{\Omega^s}$ refers to an integral over the region Ω^s of the product of the arguments. In (15), the inequality constraints given in (8) and (9) are multiplied by the corresponding Lagrange multipliers $\Lambda_c = \Lambda_c[\mathbf{x}]$ and Λ_Σ , respectively. Moreover, using integration by parts, the transient heat conduction equation (1) is included in (16) as a constraint in the following form:

$$\langle l[\theta], \vartheta \rangle_{\Omega^s} = \int_{\Omega^s} \left(\rho c \frac{\partial \theta}{\partial t} \vartheta + k \nabla \theta \cdot \nabla \vartheta - Q \vartheta \right) d\Omega - \int_{\Gamma^s} q \vartheta d\Gamma = 0 .$$

The Lagrange multiplier field $\vartheta = \vartheta[\mathbf{x}, t]$ corresponds to the so-called adjoint temperature field. Observe that since the constraint must be satisfied, then the values of \mathcal{J} and $\tilde{\mathcal{J}}$, as well as the values of their shape derivatives, essentially coincide. This feature is used in the adjoint method to compute the shape derivative of the functional.

3.4. Shape derivatives of $\tilde{\mathcal{J}}$

In view of (16) and since the variables t and s are independent of each other, the design derivative of $\tilde{\mathcal{J}}$ can be written as

$$\frac{d\tilde{\mathcal{J}}}{ds} = \frac{d\mathcal{J}}{ds} + \int_0^T \frac{d}{ds} \langle l[\theta], \vartheta \rangle_{\Omega^s} dt, \quad (17)$$

which consists of two parts: one is the design derivative of the original objective functional and the other one is the design derivative of the equality constraint $\langle l[\theta], \vartheta \rangle_{\Omega^s} = 0$. However, in order to directly compute $d\mathcal{J}/ds$, it is required to determine the shape derivatives of the temperature, θ' , and the contact heat supply, q' , which typically are computationally expensive to obtain (using, e.g., a direct perturbation method). Instead, following the adjoint method, the strategy is to manipulate the right hand side of (17) in order to express it in a convenient format (i.e., eliminating θ' and q'), thus allowing an efficient computation of $d\tilde{\mathcal{J}}/ds$. In turn, $d\tilde{\mathcal{J}}/ds$ delivers the desired expression of $d\mathcal{J}/ds$ without the terms θ' and q' . The steps to convert the right hand side of (17) into a convenient format are indicated in the subsequent sections.

3.4.1. Shape derivatives of the objective functional

Bearing in mind that the contact heat supply on Γ_e^s is $q_e = -h(\theta - \theta_e)$, it follows that

$$q'_e = -h\theta' \quad \text{and} \quad \nabla q_e = -h\nabla\theta \quad \text{on } \Gamma_e^s. \quad (18)$$

Using the transport relations shown in (13) and (14) and in view of (12) and (18), the shape derivative of the objective functional \mathcal{J} can be obtained as

$$\begin{aligned} \frac{d\mathcal{J}}{ds} &= \int_0^T \mathcal{S} \int_{\Omega^s} \omega \psi_{\omega, \theta} \theta' \, d\Omega dt + \int_0^T \mathcal{S} \left(\int_{\Gamma^s} \omega \psi_{\omega} \nu_n \, d\Gamma + \int_{S_{\omega}^s} \llbracket \omega \rrbracket \psi_{\omega} \nu_n \, d\Gamma \right) dt \\ &+ \int_0^T \mathcal{S} \left(\int_{\Gamma_q^s} \gamma \psi_{\gamma, \theta} \theta' \, d\Gamma + \int_{\Gamma_{\theta}^s} \gamma \psi_{\gamma, \theta} \hat{\theta}' \, d\Gamma + \int_{\Gamma_e^s} \gamma \psi_{\gamma, \theta} \theta' \, d\Gamma \right) dt \\ &+ \int_0^T \mathcal{S} \left(\int_{\Gamma_q^s} \gamma \psi_{\gamma, q} \hat{q}' \, d\Gamma + \int_{\Gamma_{\theta}^s} \gamma \psi_{\gamma, q} q' \, d\Gamma - \int_{\Gamma_e^s} \gamma h \psi_{\gamma, q} \theta' \, d\Gamma \right) dt \\ &+ \int_0^T \mathcal{S} \int_{\Gamma^s} \gamma \left((\nabla \psi_{\gamma} \cdot \mathbf{n}) \nu_n - \psi_{\gamma} \kappa \nu_n \right) d\Gamma dt + \int_0^T \mathcal{S} \int_{\Gamma^s} \llbracket \gamma \rrbracket \psi_{\gamma} \nu_m \, d\Upsilon dt. \end{aligned} \quad (19)$$

3.4.2. Shape derivative of the governing equation.

The second term on the right hand side of (16) involves the heat conduction problem, treated as an equality constraint integrated in space and time. In view of obtaining a useful expression for this term, consider first the shape derivative of $\langle l[\theta], \vartheta \rangle_{\Omega^s}$, which, using the transport relations (13) and (14) and in view of (12), can be expressed as

$$\begin{aligned} \frac{d}{ds} \langle l[\theta], \vartheta \rangle_{\Omega^s} &= \int_{\Omega^s} \left(\rho c \frac{\partial \theta'}{\partial t} \vartheta + \rho c \frac{\partial \theta}{\partial t} \vartheta' - Q' \vartheta - Q \vartheta' \right) d\Omega \\ &+ \int_{\Omega^s} \left(k \nabla \theta' \cdot \nabla \vartheta + k \nabla \theta \cdot \nabla \vartheta' \right) d\Omega + \int_{\Gamma^s} \left(\rho c \frac{\partial \theta}{\partial t} \vartheta + k \nabla \theta \cdot \nabla \vartheta - Q \vartheta \right) \nu_n \, d\Gamma \\ &- \int_{\Gamma_q^s} \left(\hat{q}' \vartheta + \hat{q} \vartheta' \right) d\Gamma - \int_{\Gamma_{\theta}^s} \left(q' \vartheta + q \vartheta' \right) d\Gamma - \int_{\Gamma_e^s} \left(q' \vartheta + q \vartheta' \right) d\Gamma \\ &- \int_{\Gamma^s} \left((\nabla (q \vartheta) \cdot \mathbf{n}) \nu_n - (q \vartheta) \kappa \nu_n \right) d\Gamma. \end{aligned} \quad (20)$$

Noting that $\langle l[\theta], \vartheta' \rangle_{\Omega^s} = 0$ and using (18), it follows from (20) that

$$\begin{aligned} \frac{d}{ds} \langle l[\theta], \vartheta \rangle_{\Omega^s} &= \int_{\Omega^s} \left(\rho c \frac{\partial \theta'}{\partial t} \vartheta - Q' \vartheta + k \nabla \theta' \cdot \nabla \vartheta \right) d\Omega + \int_{\Gamma^s} \left(\rho c \frac{\partial \theta}{\partial t} \vartheta + k \nabla \theta \cdot \nabla \vartheta - Q \vartheta \right) \nu_n \, d\Gamma \\ &- \int_{\Gamma_q^s} \hat{q}' \vartheta \, d\Gamma - \int_{\Gamma_{\theta}^s} q' \vartheta \, d\Gamma + \int_{\Gamma_e^s} h \theta' \vartheta \, d\Gamma - \int_{\Gamma^s} \left((\nabla (q \vartheta) \cdot \mathbf{n}) \nu_n - (q \vartheta) \kappa \nu_n \right) d\Gamma. \end{aligned} \quad (21)$$

Since the transient heat conduction equation (1), expressed in strong form as $l[\theta] = 0$, is satisfied for every $\mathbf{x} \in \Omega^s$ and every $t \in \mathcal{T}$, then the relation (21) holds for every $t \in \mathcal{T}$. Integrating the first term in the first integral on the right side of (21) over the time interval $\mathcal{T} = [0, T]$, exchanging the order of integration and subsequently integrating by parts in time provides the following relation:

$$\int_0^T \int_{\Omega^s} \rho c \frac{\partial \theta'}{\partial t} \vartheta \, d\Omega dt = \int_{\Omega^s} [\rho c \theta' \vartheta]_{t=0}^{t=T} \, d\Omega - \int_0^T \int_{\Omega^s} \rho c \theta' \frac{\partial \vartheta}{\partial t} \, d\Omega dt. \quad (22)$$

It is convenient to introduce the adjoint time variable $\tau := T - t$ and the convention that an arbitrary function f may refer to either a function of t or τ , i.e., $f = f(t) = f(\tau)$ with τ related to t as indicated (i.e., the proper argument may be inferred from the context). In that case, it holds that

$$\int_0^T f dt = \int_0^T f d\tau \quad \text{and} \quad \int_0^T \frac{\partial f}{\partial t} dt = - \int_0^T \frac{\partial f}{\partial \tau} d\tau .$$

Following this, (22) becomes

$$\int_0^T \int_{\Omega^s} \rho c \frac{\partial \theta'}{\partial t} \vartheta d\Omega dt = \int_{\Omega^s} [\rho c \theta' \vartheta]_{t=0}^{t=T} d\Omega + \int_0^T \int_{\Omega^s} \rho c \theta' \frac{\partial \vartheta}{\partial \tau} d\Omega d\tau . \quad (23)$$

Integrating by parts over Ω the third term in the first integral on the right side of (21) (in order to transfer the gradient from θ to ϑ), one has

$$\int_{\Omega^s} k \nabla \theta' \cdot \nabla \vartheta d\Omega = - \int_{\Omega^s} (k \nabla^2 \vartheta) \theta' d\Omega + \int_{\Gamma^s} q^* \theta' d\Gamma \quad (24)$$

where q^* is the adjoint contact heat supply on the boundary Γ^s , which is related to the adjoint heat flux \mathbf{q}^* as follows:

$$\mathbf{q}^* := -\mathbf{q}^* \cdot \mathbf{n} \quad \mathbf{q}^* := -k \nabla \vartheta . \quad (25)$$

3.5. Transient adjoint system

Integrating (19) in time and in view of (17), (21), (23) and (24), the design derivative of the augmented functional defined in (16) can be expressed as

$$\frac{d\tilde{\mathcal{J}}}{ds} = \Phi_1 + \Phi_2, \quad (26)$$

where

$$\begin{aligned} \Phi_1 := & \int_0^T \int_{\Omega^s} \left(\rho c \frac{\partial \vartheta}{\partial \tau} - k \nabla^2 \vartheta + \varsigma \omega \psi_{\omega, \theta} \right) \theta' d\Omega d\tau + \int_{\Omega^s} [\rho c \theta' \vartheta]_{t=0}^{t=T} d\Omega \\ & + \int_0^T \left(\int_{\Gamma_q^s} (\varsigma \gamma \psi_{\gamma, \theta} + q^*) \theta' d\Gamma + \int_{\Gamma_\theta^s} (\varsigma \gamma \psi_{\gamma, q} - \vartheta) q' d\Gamma \right) d\tau \\ & + \int_0^T \left(\int_{\Gamma_h^s} \left(h \left(\vartheta + \frac{\varsigma \gamma \psi_{\gamma, \theta}}{h} - \varsigma \gamma \psi_{\gamma, q} \right) + q^* \right) \theta' d\Gamma \right) d\tau \end{aligned} \quad (27)$$

and

$$\begin{aligned} \Phi_2 := & \int_0^T \varsigma \left(\int_{\Gamma^s} \omega \psi_{\omega} \nu_n d\Gamma + \int_{S_\omega^s} \llbracket \omega \rrbracket \psi_{\omega} \nu_n d\Gamma + \int_{\Gamma^s} \llbracket \gamma \rrbracket \psi_{\gamma} \nu_m d\Upsilon \right) dt \\ & + \int_0^T \left(\int_{\Gamma^s} \left(\rho c \frac{\partial \vartheta}{\partial t} \vartheta + k \nabla \theta \cdot \nabla \vartheta - \mathcal{Q} \vartheta \right) \nu_n d\Gamma \right) dt \\ & + \int_0^T \left(\int_{\Omega^s} -\mathcal{Q}' \vartheta d\Omega + \int_{\Gamma_\theta^s} (\varsigma \gamma \psi_{\gamma, \theta} + q^*) \hat{\vartheta}' d\Gamma + \int_{\Gamma_q^s} (\varsigma \gamma \psi_{\gamma, q} - \vartheta) \hat{q}' d\Gamma \right) dt \\ & + \int_0^T \int_{\Gamma^s} (\varsigma \gamma ((\nabla \psi_{\gamma} \cdot \mathbf{n}) \nu_n - \psi_{\gamma} \kappa \nu_n) - (\nabla (q \vartheta) \cdot \mathbf{n}) \nu_n + (q \vartheta) \kappa \nu_n) d\Gamma dt. \end{aligned} \quad (28)$$

All the terms that contain either θ' or q' have been grouped in Φ_1 (except $\hat{\theta}'$ and \hat{q}' which correspond to derivatives of prescribed boundary conditions) while the remaining terms have been collected in Φ_2 . The initial temperature $\theta[\mathbf{x}, 0] = \theta_0[\mathbf{x}]$ is henceforth assumed to be the same for all designs s (in an Eulerian sense). Correspondingly, the spatial shape derivative of the initial condition is zero, i.e., $\theta'[\mathbf{x}, 0] = 0$ for all s . As a consequence of this assumption, the term $\rho c \theta' \vartheta$ appearing in Φ_1 is zero at $t = 0$.

In order to eliminate the terms appearing in Φ_1 that involve the implicitly-dependent shape derivatives θ' and q' , the following adjoint system is required:

$$\begin{aligned} \rho c \frac{\partial \vartheta}{\partial \tau} - k \nabla^2 \vartheta - Q^* &= 0 \quad \text{with} \quad Q^* = -\zeta \omega \psi_{\omega, \theta} & (\mathbf{x}, \tau) \in \Omega^s \times \mathcal{T} \\ \vartheta &= \hat{\vartheta} \quad \text{with} \quad \hat{\vartheta} = \zeta \gamma \psi_{\gamma, q} & (\mathbf{x}, \tau) \in \Gamma_\theta^s \times \mathcal{T} \\ \mathbf{q}^* \cdot \mathbf{n} &= -\hat{q}^* \quad \text{with} \quad \hat{q}^* = -\zeta \gamma \psi_{\gamma, \theta} & (\mathbf{x}, \tau) \in \Gamma_q^s \times \mathcal{T} \\ \mathbf{q}^* \cdot \mathbf{n} &= -q_e^* = h(\vartheta - \vartheta_e) \quad \text{with} \quad \vartheta_e = -\frac{\zeta \gamma \psi_{\gamma, \theta}}{h} + \zeta \gamma \psi_{\gamma, q} & (\mathbf{x}, \tau) \in \Gamma_e^s \times \mathcal{T} \\ \vartheta[\mathbf{x}, 0] &= 0 & \mathbf{x} \in \Omega^s. \end{aligned} \quad (29)$$

In (29), the adjoint volumetric heat supply Q^* in $\Omega^s \times \mathcal{T}$, the adjoint temperature $\hat{\vartheta}$ in $\Gamma_\theta^s \times \mathcal{T}$, the adjoint contact heat supply \hat{q}^* in $\Gamma_q^s \times \mathcal{T}$ and the adjoint ambient temperature ϑ_e in $\Gamma_e^s \times \mathcal{T}$ are known once the primary field equation for θ has been solved. Observe that the characteristic function ζ needs to be evaluated at $\tau = T - t$. Moreover, the ‘‘initial’’ adjoint temperature corresponds to $\tau = 0$ (i.e., for the final time $t = T$ of the primary problem). The adjoint problem (29) is therefore formally a transient heat conduction problem from $\tau = 0$ to $\tau = T$. In the numerical implementation, the boundary conditions shown in (29) are evaluated based on the solved primary system temperature field $\theta[\mathbf{x}, t]$ at each time step. Imposing these time- and temperature-dependent boundary conditions, especially the essential boundary conditions, needs to be careful in isogeometric analysis due to the non-interpolatory of the NURBS basis. Numerical implementation aspects can be found in [50].

Choosing an adjoint temperature ϑ that satisfies (29), the term Φ_1 vanishes, i.e.,

$$\Phi_1 = 0, \quad (30)$$

hence the shape derivative of the objective functional is provided by the term Φ_2 given in (28).

3.6. Continuous adjoint sensitivity

In order to obtain the final expression for the shape derivative, further simplifications for the term Φ_2 can be carried out. First, the spatial gradient of ψ_γ can be expressed as

$$\nabla \psi_\gamma = \psi_{\gamma, \theta} \nabla \theta + \psi_{\gamma, q} \nabla q. \quad (31)$$

In view of (31) and (29)₂, it follows that the terms containing gradients in the fourth integral on the right hand side of (28) can be written as

$$\begin{aligned} & \int_{\Gamma^s} (\zeta \gamma (\nabla \psi_\gamma \cdot \mathbf{n}) v_n - (\nabla (q \vartheta) \cdot \mathbf{n}) v_n) \, d\Gamma \\ &= \int_{\Gamma^s} (\zeta \gamma \psi_{\gamma, \theta} \nabla \theta - q \nabla \vartheta) \cdot \mathbf{n} v_n \, d\Gamma + \int_{\Gamma_q^s} (\zeta \gamma \psi_{\gamma, q} - \vartheta) (\nabla \hat{q} \cdot \mathbf{n}) v_n \, d\Gamma \\ & - \int_{\Gamma_e^s} h (\zeta \gamma \psi_{\gamma, q} - \vartheta) (\nabla \theta \cdot \mathbf{n}) v_n \, d\Gamma. \end{aligned} \quad (32)$$

Substituting (32) in (28), in view of (30) and using the fact that $d\tilde{\mathcal{J}}/ds = d\mathcal{J}/ds$, it follows from (26) that the shape derivative of the objective functional is

$$\begin{aligned}
\frac{d\mathcal{J}}{ds} = & \int_0^T \left(\int_{\Omega^s} -Q' \vartheta d\Omega + \int_{\Gamma_q^s} (\varsigma\gamma\psi_{\gamma,q} - \vartheta) (\hat{q}' + (\nabla\hat{q} \cdot \mathbf{n}) v_n) d\Gamma \right) dt \\
& + \int_0^T \left(\int_{\Gamma_\theta^s} (\varsigma\gamma\psi_{\gamma,\theta} + q^*) \hat{\theta}' d\Gamma - \int_{\Gamma_e^s} h (\varsigma\gamma\psi_{\gamma,q} - \vartheta) (\nabla\theta \cdot \mathbf{n}) v_n d\Gamma \right) dt \\
& + \int_0^T \varsigma \left(\int_{S_\omega^s} \llbracket \omega \rrbracket \psi_\omega v_n d\Gamma + \int_{\Upsilon^s} \llbracket \gamma \rrbracket \psi_\gamma v_m d\Upsilon \right) dt \\
& + \int_0^T \int_{\Gamma^s} \left(\varsigma\omega\psi_\omega + \rho c \frac{\partial\theta}{\partial t} \vartheta + k\nabla\theta \cdot \nabla\vartheta - Q\vartheta \right) v_n d\Gamma dt \\
& + \int_0^T \int_{\Gamma^s} \left(\varsigma\gamma (\psi_{\gamma,\theta} (\nabla\theta \cdot \mathbf{n}) - \psi_{\gamma,\kappa}) - q\nabla\vartheta \cdot \mathbf{n} + (q\vartheta)\kappa \right) v_n d\Gamma dt.
\end{aligned} \tag{33}$$

The above equation is a general expression of the continuous adjoint gradient for the optimization problem with an objective functional given by (3) and a temperature field that satisfies the transient heat conduction problem given by (1) and (2). This general expression can be applied for problem with design-dependent temperature, contact heat supply and/or volumetric heat supply. For simplicity, however, it is henceforth assumed that the externally-prescribed heat supplies and temperature are such that $\hat{\theta}' = 0$, $\hat{q}' = 0$, $\nabla\hat{q} \cdot \mathbf{n} = 0$ and $Q' = 0$. Under the above-mentioned assumptions, the shape derivative of the objective functional \mathcal{J} for a transient heat conduction problem given in (33) may be expressed as

$$\begin{aligned}
\frac{d\mathcal{J}}{ds} \Big|_{\hat{\theta}', \hat{q}', \nabla\hat{q} \cdot \mathbf{n}, Q'=0} = & \int_{\Gamma^s} \left(\int_0^T \mathbf{g} dt \right) \cdot \mathbf{v} d\Gamma + \int_{S_\omega^s} \left(\int_0^T \llbracket \omega \rrbracket \psi_\omega \mathbf{n} dt \right) \cdot \mathbf{v} d\Gamma \\
& + \int_{\Upsilon^s} \left(\int_0^T \llbracket \gamma \rrbracket \psi_\gamma \mathbf{m} dt \right) \cdot \mathbf{v} d\Upsilon,
\end{aligned} \tag{34}$$

where

$$\begin{aligned}
\mathbf{g} = & \left(\varsigma\omega\psi_\omega + \rho c \frac{\partial\theta}{\partial t} \vartheta + k\nabla\theta \cdot \nabla\vartheta - Q\vartheta - \gamma_e h (\varsigma\gamma\psi_{\gamma,q} - \vartheta) (\nabla\theta \cdot \mathbf{n}) \right. \\
& \left. + \varsigma\gamma (\psi_{\gamma,\theta} (\nabla\theta \cdot \mathbf{n}) - \psi_{\gamma,\kappa}) - q\nabla\vartheta \cdot \mathbf{n} + (q\vartheta)\kappa \right) \mathbf{n}
\end{aligned} \tag{35}$$

and γ_e in (35) is the characteristic function that is equal to 1 on the boundary Γ_e and 0 otherwise. For regular design points on the boundary (i.e., points where the fields are continuous), the unconstrained local shape gradient is given by the time integral of \mathbf{g} . For singular points (including edges), the local gradient contains additional terms that provide information in the tangential direction, as indicated in (34).

4. Isogeometric analysis and design discretization

In general, the NURBS-based geometrical model can be expressed as

$$\mathbf{x} = \sum_I R^I \mathbf{x}^I \tag{36}$$

where R^I is a NURBS function associated with the control point of coordinates \mathbf{x}^I and the index I runs over all control points that characterize the geometry.

The basic idea behind isogeometric analysis is to employ the NURBS basis functions used in CAD to define the geometry also as shape functions for analysis. This approach provides a unified environment between computer added design and finite element analysis. Under this framework, the temperature field can be discretized using NURBS shape functions as follows:

$$\theta = \sum_I \theta^I R^I, \quad I = 1, 2, \dots, a, \quad (37)$$

where θ^I is the temperature coefficient associated to the I th control point and a is the number of the control points used for analysis.

The representation (37), together with the corresponding approximations for the test functions, the volumetric and contact heat supplies and the heat flux fields, are substituted in a weak formulation of problem (1) in order to obtain a system of equations for the unknowns θ^I , i.e.,

$$\mathbf{C} \frac{\partial \boldsymbol{\theta}}{\partial t} + \mathbf{K} \boldsymbol{\theta} = \mathbf{f} \quad (38)$$

where $\boldsymbol{\theta} = \{\theta^1, \theta^2, \dots\}$ is the vector of temperature coefficients, \mathbf{C} is the global capacitance matrix, \mathbf{K} is the global conductance matrix and \mathbf{f} is the global heat supply vector.

Introducing a parameter $\beta \in [0, 1]$ and a time step Δt in the finite-difference approximation in time $\partial(\cdot)/\partial t \approx (1/\Delta t)(\beta(\cdot)^{t+\Delta t} + (1-\beta)(\cdot)^t)$, the fully-discrete system of equations can be expressed from (38) as

$$\mathbf{A} \boldsymbol{\theta}^{t+\Delta t} = \mathbf{b} \quad (39)$$

with

$$\mathbf{A} := \mathbf{C} + \beta \Delta t \mathbf{K} \quad \mathbf{b} := \Delta t \mathbf{f} + (\mathbf{C} - (1-\beta) \Delta t \mathbf{K}) \boldsymbol{\theta}^t$$

and where $\boldsymbol{\theta}^{t+\Delta t}$ corresponds to an approximation of $\boldsymbol{\theta}$ at time $t + \Delta t$, which is obtained from the approximation $\boldsymbol{\theta}^t$ by solving (39). The cases $\beta = 0, 0.5, 1$ correspond, respectively, to the forward Euler, Crank-Nicolson and backward Euler methods. The isogeometric approach can be used to solve the primary and adjoint problems in order to determine the fields required to compute shape derivatives.

The discretization of the analysis and design can be carried out in two different levels (see Fig. 3). Typically, in the analysis space, it is required to first carry out a refinement in order to have sufficient accuracy. This can be achieved using knot refinement. In the design space, using a coarse mesh can reduce the design parameters and simplify the process of updating the locations of the interior control points. The links between the design and analysis discretization spaces can be summarized as follows (also see [9, 24, 23]):

1. The design model is updated in the design discretization space
2. The analysis discretization space is a refined space from the design discretization space through NURBS knot refinement
3. The sensitivity analysis is carried out in the design discretization space but using the field variables from the analysis discretization space

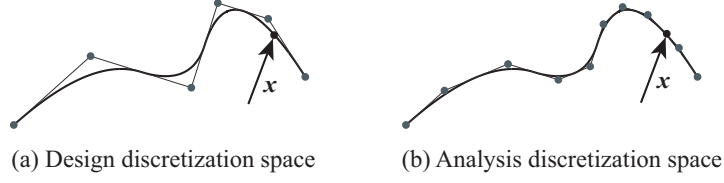


Figure 3: Schematic of geometrical representation using NURBS: (a) Coarse representation to define the design geometry (design control points), (b) Refinement used for analysis (analysis control points) that preserve the design geometry.

5. Isogeometric shape design optimization

5.1. Discrete shape gradients with respect to design control points

At the design level, the discretization typically uses a coarser mesh compared with the discretization used for analysis. The design shape is expressed using (36). Following the definition in (11), the design velocity can be discretized as follows:

$$\mathbf{v}[\mathbf{p}; s] = \dot{\hat{\mathbf{x}}}[\mathbf{p}; s] = \sum_I R^I[\mathbf{p}] \frac{d\mathbf{x}^I[s]}{ds}. \quad (40)$$

Substituting (40) in (34) gives

$$\left. \frac{d\mathcal{J}}{ds} \right|_{\hat{\theta}^l, \hat{q}^l, \nabla \hat{q} \cdot \mathbf{n}, Q^l=0} = \sum_I \frac{\partial \mathcal{J}}{\partial \mathbf{x}^I} \cdot \frac{d\mathbf{x}^I}{ds}. \quad (41)$$

where

$$\mathcal{J}_{,x^I} = \frac{\partial \mathcal{J}}{\partial \mathbf{x}^I} = \int_{\Gamma^s} R^I \left(\int_0^T \mathbf{g} dt \right) d\Gamma + \int_{S_\omega^s} R^I \left(\int_0^T \llbracket \omega \rrbracket \psi_\omega \mathbf{n} dt \right) d\Gamma + \int_{\Upsilon^s} R^I \left(\int_0^T \llbracket \gamma \rrbracket \psi_\gamma \mathbf{m} dt \right) d\Upsilon. \quad (42)$$

Similarly, for the volume constraint, it can be deduced that

$$\Sigma_{,x^I} = \frac{\partial \Sigma}{\partial \mathbf{x}^I} = \int_{\Gamma^s} R^I \mathbf{n} d\Gamma. \quad (43)$$

After solving the primary and adjoint system, all the unknown state variables involved in (42) and (43) can be evaluated. The normal vector and curvature terms can be calculated directly using relations from differential geometry.

5.2. Normalization of the search direction

In isogeometric shape optimization, the descent direction predicted from the discrete shape gradient is strongly dependent on the discretization. This discretization-dependency can slow down the convergence speed and may lead the process into a sub-optimal solution. The source of this discretization-dependency can be traced back to the lack of consistency with the local steepest descent search direction in the continuous formulation. This inconsistency can be alleviated using a normalization approach as presented in the work of Kiendl et al. [15] and Wang et al.

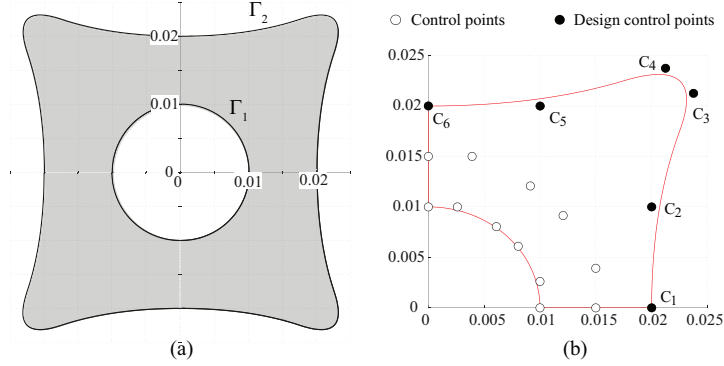


Figure 4: (a) Initial shape of a 2D plate with a circular orifice and (b) the NURBS parameterization for a quarter of the plate (values in m).

[51]. The normalization scheme employed in the present work is a diagonally-lumped matrix mapping-based approach [51] (also referred to as “sensitivity weighting” in [15]), which results in

$$\bar{\mathcal{J}}_{,x^l} = \frac{\mathcal{J}_{,x^l}}{W_{x^l}}, \quad (44)$$

where W_{x^l} is a factor that accounts for the support of NURBS function associated to the control point x^l , i.e.,

$$W_{x^l} = \int_{\Gamma^s} R^l d\Gamma. \quad (45)$$

5.3. Iterative descent method.

With the sensitivity analysis given in (44), the shape optimization problem presented above can be solved using different iterative algorithms (e.g., classical steepest descent or sequential quadratic programming). For simplicity, a descent algorithm that incorporates the global constraint through a penalty-like formulation is summarized in Algorithm 1, where the design index s in the continuous formulation is formally replaced by an iteration number n and the increment δs is formally replaced by a step size α .

6. Numerical examples

6.1. Verification examples

6.1.1. Minimum surface problem

Consider a 2D plate with a circular orifice in it as shown in Fig. 4(a). The plate has an initial temperature of $\theta_0 = 100^\circ\text{C}$ and is placed in an environment with an ambient temperature of $\theta_e = 0^\circ\text{C}$. The plate is made of a material with a mass density of $\rho = 7800 \text{ kg/m}^3$, a heat capacity of $c = 420 \text{ J/(kg}\cdot^\circ\text{C)}$ and a thermal conductivity coefficient of $k = 20 \text{ W/(m}\cdot^\circ\text{C)}$. The convection coefficient on the external boundary Γ_2 is $h = 50 \text{ W/(m}^2\cdot^\circ\text{C)}$ and on the internal boundary Γ_1 it is assumed that no heat is exchanged, i.e., the contact heat supply is zero ($\hat{q} = 0$). In this case the steady state, in which $\partial\theta/\partial t$ vanishes, is reached only for the uniform temperature limit case as $t \rightarrow \infty$ that corresponds to the situation when no more heat is exchanged between the plate and

Algorithm 1 Descent algorithm with global constraint

Initialize ($n = 0$)

Choose a initial design domain $\Omega^{(n=0)}$ with design control points for its boundary, a step size $\alpha > 0$, a penalty factor $\beta_p > 0$ and a tolerance ϵ and set $\Lambda_\Sigma^{(0)} = 0$

Main loop ($n \geq 0$)

while $|\mathcal{J}^{(n+1)} - \mathcal{J}^{(n)}|/\mathcal{J}^{(n)} \geq \epsilon$ **do**

 Solve the transient problem (39), compute adjoint boundary conditions and solve the adjoint problem (29)

for All control points I on the design boundary **do**

 Compute the gradient of the objective and the volume constraint with respect to discrete variables from (42) and (43), respectively

 Compute the normalized search direction from (44)

end for

Constrained minimization

Initialize ($m = 0$), $\Lambda_\Sigma^{(n+1,0)} = \Lambda_\Sigma^{(n)}$

while $|\Lambda_\Sigma^{(n+1,m+1)} - \Lambda_\Sigma^{(n+1,m)}|/\Lambda_\Sigma^{(n+1,m)} > \epsilon$ for $\Lambda_\Sigma^{(n+1,m)} \neq 0$ **do**

for All control points I on the design boundary **do**

 Update the location of the design control points for sub-iteration m :

$$\left(\mathbf{x}^I\right)^{(n+1,m)} = \left(\mathbf{x}^I\right)^{(n)} - \alpha \left(\mathcal{J}_{,\mathbf{x}^I}^{(n)} + \Lambda_\Sigma^{(n+1,m)} \Sigma_{,\mathbf{x}^I}^{(n)}\right)$$

 and denote new location also as $\left(\mathbf{x}^I\right)^{(n+1,m)}$

end for

 Update the volume of the design region $\Sigma^{(n+1,m)}$ and $\Lambda_\Sigma^{(n+1,m+1)}$

$$\Lambda_\Sigma^{(n+1,m+1)} = \max\left(0, \Lambda_\Sigma^{(n+1,m)} + \beta_p \left(\Sigma^{(n+1,m)} - \Sigma_0\right)\right)$$

 Check convergence for case $\Lambda_\Sigma^{(n+1,m+1)} = 0$ separately

$m \leftarrow m + 1$

end while

 Set $\Lambda_\Sigma^{(n+1)} = \Lambda_\Sigma^{(n+1,m+1)}$ and $\left(\mathbf{x}^I\right)^{(n+1)} = \left(\mathbf{x}^I\right)^{(n+1,m)}$

 Update internal control points based on boundary control points

$n \leftarrow n + 1$

end while

its surroundings. In practice, with the initial design and the given data, the plate cools down and reaches a state of near-uniform temperature of 0°C , to within a tolerance of 10^{-3}°C , after 7200 seconds.

Suppose that one would like to find the optimal shape of the external boundary Γ_2 such that the heat exchange rate with the external environment is minimized while using the same amount of material as the original design. This benchmark problem can be alternatively thought of as a minimum surface problem that admits a simple solution, namely a circular external shape (which corresponds to the minimum surface for Γ_2). The verification consists of formulating the problem as a minimization of the heat exchange rate and finding the circular shape starting from the initial shape shown in Fig. 4(a). It should be noted that this problem, formulated in terms of the heat exchanged, cannot be solved based on the steady-state problem since the steady-state solution for *any shape* corresponds to zero temperature and zero heat exchange rate. Instead, this problem can be solved using a transient-state design approach. To achieve this, the problem can be formulated as a minimization of the total heat exchanged during a given time interval with an objective function defined over the external boundary Γ_2 as follows:

$$\mathcal{J} = \int_0^T \Psi_\gamma dt \quad \text{with} \quad \Psi_\gamma = \int_{\Gamma_2} h(\theta - \theta_e) d\Gamma \quad (46)$$

subjected to a volume constraint

$$\Sigma = \Sigma_0,$$

where Σ_0 is the volume of the initial design. In this problem, $\Psi_\gamma \geq 0$ since the surface temperature satisfies $\theta \geq \theta_e$.

For the optimization problem, it is assumed that the model has horizontal and vertical symmetries, so it is sufficient to consider only a quarter of the domain as shown in Fig. 4(b). Correspondingly, zero contact heat is assumed on the lateral boundaries $x_1 = 0$ and $x_2 = 0$. The locations of six control points, denoted as C_I , $I = 1, \dots, 6$, are chosen as the discrete design variables. The shape optimization is carried out with a design time $T = 300$ s. The transient solutions, for both the primary and adjoint systems, are obtained using the isogeometric analysis framework while the integration over time for the computation of the gradient is done using the trapezoidal rule. The adjoint equations corresponding to this problem can be derived from (29) as

$$\begin{aligned} \rho c \frac{\partial \vartheta}{\partial \tau} - k \nabla^2 \vartheta - Q^* &= 0 \quad \text{with} \quad Q^* = 0 & (\mathbf{x}, \tau) \in \Omega^s \times \mathcal{T} \\ \mathbf{q}^* \cdot \mathbf{n} = -q_e^* &= h(\vartheta - \vartheta_e) \quad \text{with} \quad \vartheta_e = -1 & (\mathbf{x}, \tau) \in \Gamma_e^s \times \mathcal{T} \\ \vartheta[\mathbf{x}, 0] &= 0 & \mathbf{x} \in \Omega^s. \end{aligned} \quad (47)$$

The characteristic function $\gamma[\mathbf{p}, t]$ is equal to 1 for all times if $\mathbf{p} \in \Gamma_2^0 = \Gamma_2$ and zero otherwise. It is worth mentioning that the numerical sensitivity obtained from the method presented here agrees well with those calculated using finite differences, as is shown in Table 1. Due to symmetry, the control point C_1 is only allowed to move horizontally while C_6 is only allowed to move vertically. The corresponding sensitivity analysis regarding to the restricted directions are omitted.

Using the iterative optimization approach summarized in Algorithm 1, the optimization process converges within 5 iterations to within a small tolerance of 10^{-4} as shown in Fig. 5(a). The corresponding optimal shape is shown in Fig. 5(b) in comparison to the initial and the exact

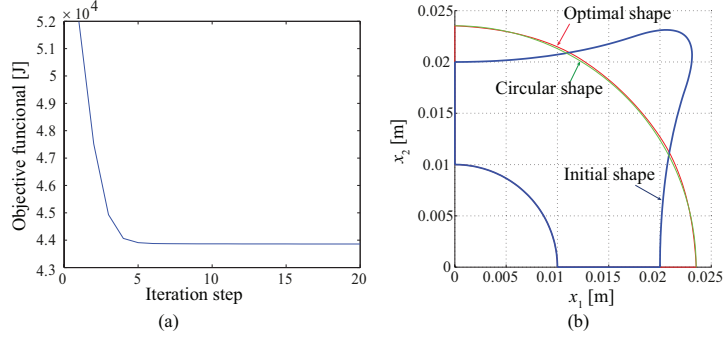


Figure 5: (a) Iteration history and (b) the optimal shape.

Table 1: Comparison between adjoint gradient and finite difference gradient for the transient heat dissipation problem.

C_I	Component	Adjoint gradient	Finite difference gradient	Relative difference
C_1	1	1.3528×10^5	1.3665×10^5	1.01%
	2	~	~	~
C_2	1	2.4255×10^5	2.4427×10^5	0.71%
	2	-0.3706×10^5	-0.3733×10^5	0.75%
C_3	1	8.9649×10^5	8.9665×10^5	0.018%
	2	3.0640×10^5	3.0542×10^5	0.32%
C_4	1	3.0640×10^5	3.0542×10^5	0.32%
	2	8.9649×10^5	8.9665×10^5	0.018%
C_5	1	-0.3706×10^5	-0.3733×10^5	0.75%
	2	2.4255×10^5	2.4427×10^5	0.71%
C_6	1	~	~	~
	2	1.3528×10^5	1.3665×10^5	1.01%

circular shape. As can be seen from Fig. 5(b), the obtained optimal shape agrees closely with the circular shape, which provides a validation of the design methodology. The control points and the corresponding weights of the initial and optimal designs are presented in Tabel 2. The knot vectors used for the two directions in the index space are $\xi = [0 \ 0 \ 0 \ 1/3 \ 1/2 \ 2/3 \ 1 \ 1 \ 1]$ and $\eta = [0 \ 0 \ 0 \ 1 \ 1 \ 1]$, respectively.

6.1.2. Passive temperature control problem

In this example, a heat flux $q = 10\text{kW/m}^2$ applied to the left side (Γ_1) of a 2D plate, as shown in Fig. 6. The rest of the boundary is isolated such that the contact heat flux on $\Gamma_{2,3}$ can be neglected. The material properties of the plate are assumed to be the same as in the previous example. The initial temperature of the plate is assumed to be 0°C . With the heat flux on the left side, the temperature on the right side of the plate (Γ_2) reaches a (nearly) uniform temperature of about 36°C at $t = 400$ s. By modifying the shape of the upper boundary (Γ_3) of the plate, it is possible to control the temperature on the right side (Γ_2) such that it reaches a

Table 2: Geometry information of the heat dissipation problem

Initial design				Optimal design			
$I(i, j)$	Locations		Weights	$I(i, j)$	Locations		Weights
	x_1^I	x_2^I			x_1^I	x_2^I	
(1, 1)	0.0100	0.0000	1.00	(1, 1)	0.0100	0.0000	1.00
(2, 1)	0.0100	0.0026	0.90	(2, 1)	0.0100	0.0026	0.90
(3, 1)	0.0080	0.0061	0.85	(3, 1)	0.0080	0.0061	0.85
(4, 1)	0.0061	0.0080	0.85	(4, 1)	0.0061	0.0080	0.85
(5, 1)	0.0026	0.0100	0.90	(5, 1)	0.0026	0.0100	0.90
(6, 1)	0.0000	0.0100	1.00	(6, 1)	0.0000	0.0100	1.00
(1, 2)	0.0150	0.0000	1.00	(1, 2)	0.0150	0.0000	1.00
(2, 2)	0.0150	0.0039	0.90	(2, 2)	0.0150	0.0039	0.90
(3, 2)	0.0121	0.0091	0.85	(3, 2)	0.0121	0.0091	0.85
(4, 2)	0.0091	0.0121	0.85	(4, 2)	0.0091	0.0121	0.85
(5, 2)	0.0039	0.0150	0.90	(5, 2)	0.0039	0.0150	0.90
(6, 2)	0.0000	0.0150	1.00	(6, 2)	0.0000	0.0150	1.00
(1, 3)	0.0200	0.0000	1.00	(1, 3)	0.0235	0.0000	1.00
(2, 3)	0.0200	0.0100	1.00	(2, 3)	0.0230	0.0092	1.00
(3, 3)	0.0238	0.0213	1.00	(3, 3)	0.0172	0.0163	1.00
(4, 3)	0.0213	0.0238	1.00	(4, 3)	0.0163	0.0172	1.00
(5, 3)	0.0100	0.0200	1.00	(5, 3)	0.0092	0.0230	1.00
(6, 3)	0.0000	0.0200	1.00	(6, 3)	0.0000	0.0235	1.00

target distribution at a desired time. In particular, suppose that the target at $t = T = 400$ s is a uniformly distributed temperature of 40°C on the right side. The objective of this problem can be formulated as follows:

$$\mathcal{J} = \int_0^T \zeta \Psi_\gamma dt \quad \text{with} \quad \Psi_\gamma = \int_{\Gamma_2} (\theta - \check{\theta})^2 d\Gamma \quad (48)$$

where $\zeta = \delta[t - t_0]$ is the characteristic function for the time interval, $\delta[t - t_0]$ is the Dirac delta function with $t_0 = T = 400$ s and $\check{\theta} = 40^\circ\text{C}$ is the target temperature. The adjoint equations corresponding to this problem can be derived from (29) as

$$\begin{aligned} \rho c \frac{\partial \vartheta}{\partial \tau} - k \nabla^2 \vartheta - Q^* &= 0 \quad \text{with} \quad Q^* = 0 & (\mathbf{x}, \tau) \in \Omega^s \times \mathcal{T} \\ \mathbf{q}^* \cdot \mathbf{n} &= -\hat{q}^* \quad \text{with} \quad \hat{q}^* = -2\zeta(\theta - \check{\theta}) & (\mathbf{x}, \tau) \in \Gamma_q^s \times \mathcal{T} \\ \vartheta[\mathbf{x}, 0] &= 0 & \mathbf{x} \in \Omega^s. \end{aligned} \quad (49)$$

The numerical sensitivity obtained from the method presented here agrees well with those calculated using finite differences, as is shown in Table 3. The optimization process converges within 6 iterations to within a small tolerance of 10^{-4} as shown in Fig. 7(a). The corresponding optimal shape and the temperature contours at $t = 400$ s are plotted in Fig. 7(b), from which it can be seen that the temperature at the right side matches the target temperature very well. To

Table 3: Comparison between adjoint gradient and finite difference gradient for temperature control problem.

$I(i, j)$	Component	Adjoint gradient	Finite difference gradient	Relative difference
(3,2)	1	13.9820	14.3893	2.91%
	2	55.9280	57.5575	2.91%
(3,3)	1	13.9571	14.3640	2.91%
	2	55.8286	57.4564	2.91%

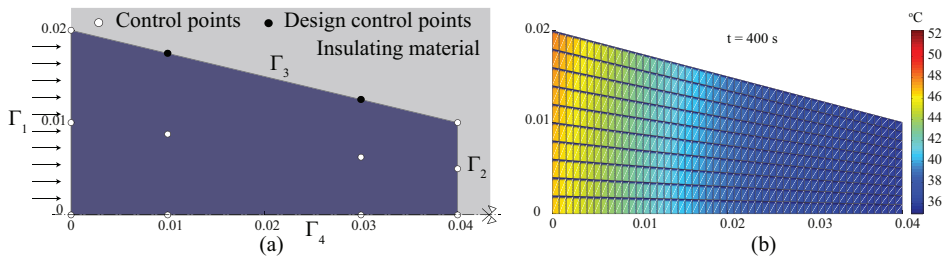


Figure 6: (a)The plate immersed in the isolation material and (b) the temperature contour at $t = 400$ s.

a nearly uniform temperature distribution, the shape optimization procedure effectively creates a uni-directional heat flow pattern on the right side of the domain while the target temperature can be reached using less material compared to the initial design (i.e., in this case the resource constraint is not active). The control points and the corresponding weights of the initial and optimal designs are presented in Table 4. The knot vectors used for the two directions in the index space are $\xi = [0\ 0\ 0\ 1\ 1\ 1]$ and $\eta = [0\ 0\ 0\ 1/2\ 1\ 1\ 1]$, respectively.

6.2. Shape optimization of a plunger

The shape optimization methodology is tested in this section with an example drawn from the literature, namely a two-dimensional plunger that is designed to form a television bulb panel

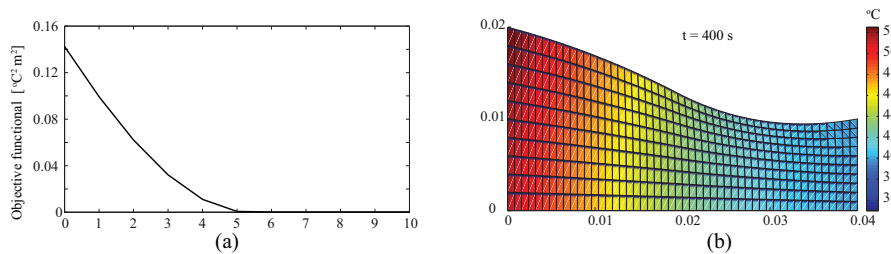


Figure 7: (a) Iteration history and (b) the optimal shape and the temperature contour at $t = 400$ s.

Table 4: Geometry information of the temperature control problem

$I(i, j)$	Initial design			$I(i, j)$	Optimal design		
	Locations		Weights		Locations		Weights
	x_1^I	x_2^I			x_1^I	x_2^I	
(1, 1)	0.0000	0.0000	1.00	(1, 1)	0.0000	0.0000	1.00
(2, 1)	0.0000	0.0100	1.00	(2, 1)	0.0000	0.0100	1.00
(3, 1)	0.0000	0.0200	1.00	(3, 1)	0.0000	0.0200	1.00
(1, 2)	0.0100	0.0000	1.00	(1, 2)	0.0100	0.0000	1.00
(2, 2)	0.0100	0.0088	1.00	(2, 2)	0.0100	0.0088	1.00
(3, 2)	0.0100	0.0175	1.00	(3, 2)	0.0100	0.0174	1.00
(1, 3)	0.0300	0.0000	1.00	(1, 3)	0.0300	0.0000	1.00
(2, 3)	0.0300	0.0063	1.00	(2, 3)	0.0300	0.0063	1.00
(3, 3)	0.0300	0.0125	1.00	(3, 3)	0.0289	0.0081	1.00
(1, 4)	0.0400	0.0000	1.00	(1, 4)	0.0400	0.0000	1.00
(2, 4)	0.0400	0.0050	1.00	(2, 4)	0.0400	0.0050	1.00
(3, 4)	0.0400	0.0100	1.00	(3, 4)	0.0400	0.0100	1.00

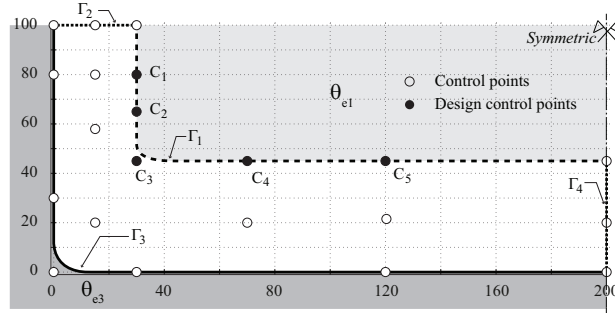


Figure 8: A plunger model and its NURBS control points (distances in mm).

from molten glass [52], as is shown in Fig. 8. Although this example pertains to a technology that is not currently widely used, it is an existing reference that serves to illustrate the main features of the method and may be applicable to similar situations. As shown in the figure, the workpiece is brought into contact with a surface at a temperature of $\theta_{e3} = 1000$ °C on the boundary Γ_3 . The boundary Γ_1 is a cooling surface that exchanges heat with a cooling fluid at a temperature of $\theta_{e1} = 0$ °C. The quality of the product surface is affected by temperature fluctuations on the contact surface Γ_3 [52]. An optimization is formulated where the shape of the surface Γ_1 is modified in order to reduce the temperature variance on the contact surface Γ_3 . To this end, denote as $\check{\theta} = \check{\theta}[t]$ the average temperature along Γ_3 at time t , i.e.,

$$\check{\theta} = \frac{1}{|\Gamma_3|} \int_{\Gamma_3} \theta d\Gamma \quad \text{with} \quad |\Gamma_3| = \int_{\Gamma_3} d\Gamma$$

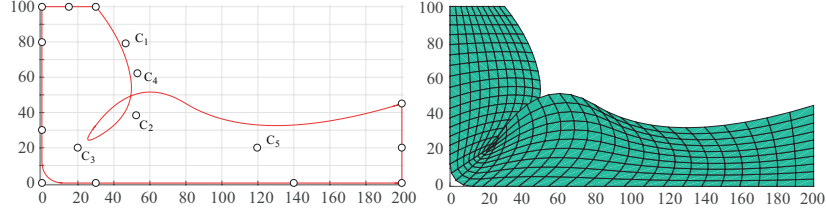


Figure 9: The irregular geometry and mesh during the iteration without restricting the movements of the control points.

and define a performance functional as

$$\mathcal{J} = \int_0^T \Psi_\gamma dt \quad \text{with} \quad \Psi_\gamma = \int_{\Gamma_3} (\theta - \check{\theta})^2 d\Gamma, \quad (50)$$

s.t. $x_1^I \geq 20 \text{ mm}, \quad x_2^I \geq 20 \text{ mm} \quad I = 1, \dots, 5$

where $x_i^I, i = 1, 2$ are the coordinates of the control points $C_I, I = 1, \dots, 5$ as shown in the figure. The locations of the control points C_I that characterize the cooling boundary Γ_1 are used as design variables. For simplicity, the design control points C_1 and C_2 are only allowed to move horizontally, C_4 and C_5 are only allowed to move vertically, and C_3 is allowed to move both horizontally and vertically. The restrictions of these movements is to avoid the mesh and geometry irregularity. Without these restrictions, the geometry and mesh may become irregular, as is shown in Fig. 9. The convection coefficients are $h = 3.15 \times 10^{-4} \text{W}/(\text{mm}^2 \cdot ^\circ\text{C})$ on Γ_1 and $h = 2.88 \times 10^{-4} \text{W}/(\text{mm}^2 \cdot ^\circ\text{C})$ on Γ_3 , respectively. The thermal conductivity coefficient is $k = 27.52 \times 10^{-3} \text{W}/(\text{mm} \cdot ^\circ\text{C})$ and the effective heat capacity is $(\rho c) = 2.288 \times 10^{-3} \text{J}/(\text{mm}^3 \cdot ^\circ\text{C})$. On the boundaries Γ_2 and Γ_4 it is assumed that there is no heat exchange and the initial temperature on the whole domain is taken as 0°C .

In order to study the influence of the design time T , the optimization problem was solved with two distinct values, namely $T = 500 \text{ s}$ and $T = 1000 \text{ s}$. The sensitivity obtained using the continuous adjoint method was verified with the finite difference method, which are shown in Table 5 and Table 6, respectively. The optimal shapes obtained are plotted in Fig. 10 for the two design times. As may be seen in the figure, the design for $T = 500 \text{ s}$ has a more complex shape than the design for $T = 1000 \text{ s}$, which may be ascribed to the influence of the details of the transient states. In contrast, the design for $T = 1000 \text{ s}$ is closer to the steady state limit case, which is more heavily influenced by stationary conditions. The corresponding convergence histories are plotted in Fig. 11(a) for $T = 500 \text{ s}$ and Fig. 11(b) for $T = 1000 \text{ s}$. These convergence histories also reflect the complexity of the designs in the sense that generally more iterations are required to converge for designs that are more influenced by the transient states (in this case more iterations are required to converge for the design for $T = 500 \text{ s}$ than for $T = 1000 \text{ s}$). The control points and the corresponding weights of the initial and optimal designs are presented in Table 7. The knot vectors used for the two directions in the index space are $\xi = [0 \ 0 \ 0.2 \ 0.4 \ 0.5 \ 0.7 \ 1 \ 1 \ 1]$ and $\eta = [0 \ 0 \ 0 \ 1 \ 1 \ 1]$, respectively.

The space-averaged temperature fluctuations on the boundary Γ_3 of the initial and the two optimal design shapes during the first 1000 seconds are plotted in Fig. 12(a) in terms of the

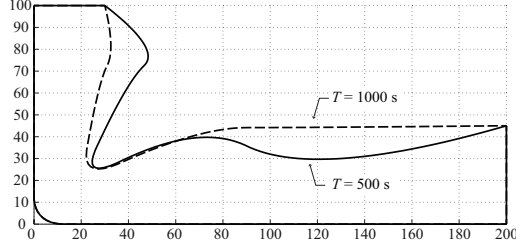


Figure 10: Optimal shapes for design times $T = 500\text{s}$ and $T = 1000\text{s}$ (distances in mm)

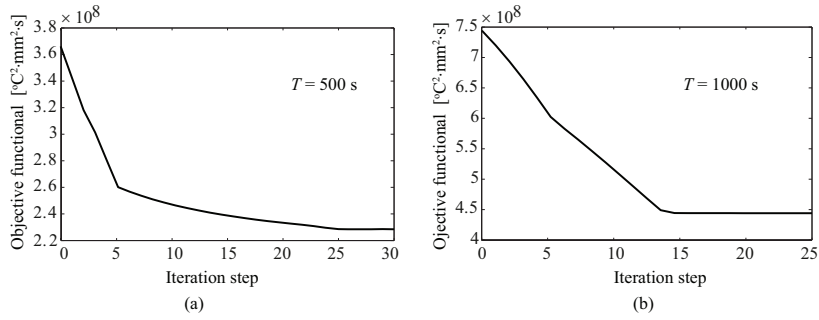


Figure 11: Iteration history of the optimization for design times (a) $T = 500\text{ s}$ and (b) $T = 1000\text{ s}$.

corresponding transient temperature variance, which is evaluated as

$$\tilde{\theta}[t] = \sqrt{\Psi_\gamma[t]/|\Gamma_3|},$$

where, as before, $|\Gamma_3|$ denotes the length of the boundary Γ_3 . The quantity $\tilde{\theta}[t]$ measures, at each time t , the average deviation of the local temperature from the average temperature (a larger value indicating an undesirable fluctuation).

From the plots in Fig. 12(a), it can be seen that the optimal shapes for both $T = 500\text{ s}$ and $T = 1000\text{ s}$ have a lower variance than the initial shape throughout the time interval $0 \leq t \leq 1000\text{ s}$ except during the first 150 s for the case $T = 1000\text{ s}$, in which the variance is slightly larger. However, the time-averaged variance for both designs is lower than the original design. The optimal shape for the case $T = 500\text{ s}$ provides the lowest temperature variance during the first 410 seconds, but after that the temperature variance of the case $T = 1000\text{ s}$ becomes the lowest. This result is consistent with the expected influence of the design parameter T on the design, namely that a design performs better during its own design interval $[0, T]$ but not necessarily during other times intervals (either larger or shorter than the design interval).

In terms of local temperature fluctuations, as shown in Fig. 12(b), the optimization procedure reduces the ratio of the maximum temperature θ_{\max} on Γ_3 to the average temperature $\check{\theta}$ and, simultaneously, increases the ratio of the minimum temperature θ_{\min} on Γ_3 to the average temperature $\check{\theta}$, as compared with the corresponding ratios for the original design. The evolution of the average temperature $\check{\theta}$ on Γ_3 is shown in Fig. 12(c), which indicates that both optimal designs generally decrease the average temperature compared with the average temperature of the initial

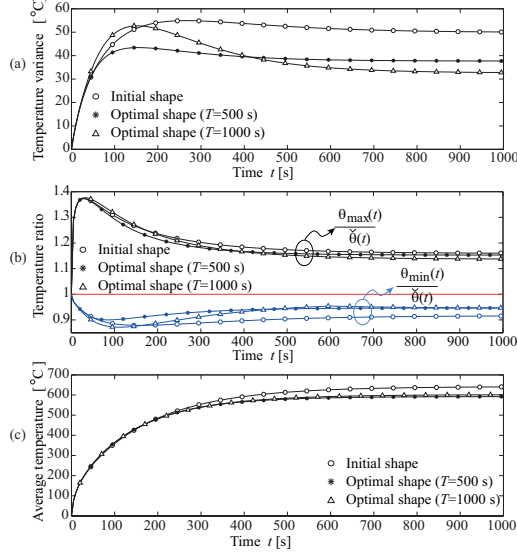


Figure 12: (a) The temperature variance $\tilde{\theta}$, (b) the ratios of the maximum and minimum temperatures to the average temperature and (c) the transient average temperature on the boundary Γ_3 for $0 \leq t \leq 1000$ s for the initial and optimal ($T = 500$ s and $T = 1000$ s) designs.

design.

The example presented in this section illustrates the significant effect that the transient response has on the optimal design and highlights the importance of making a judicious choice of the design parameters depending on the desired performance of the design.

6.3. Shape optimization of a thermal protection panel

In this example, the structural shape and the heating rate of a ballistic re-entry vehicle are considered, as shown in Fig. 13(a) and Fig. 14, which is adapted from [53]. The vehicle has two parts: the recovery tip (forward part) and the destructible frustum (aft part, see Fig. 13(a)). The outer boundary Γ_1 has a heavy heating rate during re-entry, as shown in Fig. 14 for $0 \leq t \leq 50$ s, which is used as (time-dependent) boundary condition. For times greater than 50 s, it is assumed that the heating rate is negligible in comparison to the rate for $0 \leq t \leq 50$ s. In order to protect the internal equipments inside the tip, a layer of low thermal conductivity material is used to insulate the heat from outside the tip. The cross section of the recovery tip and its NURBS parameterization are shown in Fig. 13(b). The thermal conductivity coefficient of the thermal insulation material is $k = 8.0 \times 10^{-5} \text{ W}/(\text{mm} \cdot ^\circ\text{C})$, the heat capacity is $c = 514 \text{ J}/(\text{kg} \cdot ^\circ\text{C})$ and the material density is $\rho = 0.22 \times 10^{-6} \text{ kg}/\text{mm}^3$. The initial temperature is assumed to be 0°C , the convection coefficient on Γ_2 is taken $h = 6.0 \times 10^{-3} \text{ W}/(\text{mm}^2 \cdot ^\circ\text{C})$, the ambient temperature on Γ_2 is assumed to be 0°C and on the remaining boundaries the heat exchanged is taken as zero.

With the goal of minimizing the heat transferred through the boundary Γ_2 , an objective functional is formulated as

$$\mathcal{J} = \int_0^T \int_{\Gamma_2} h(\theta - \theta_e) d\Gamma dt \quad (51)$$

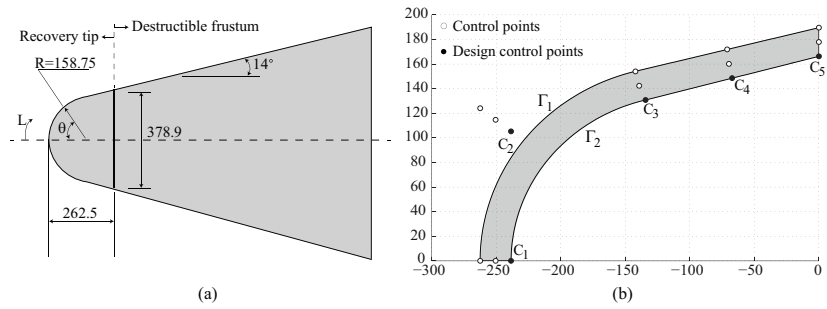


Figure 13: (a) The sketch of a re-entry vehicle (adapted from [53]) and (b) the cross section of the recovery tip skin and its NURBS parameterization (distances in mm).

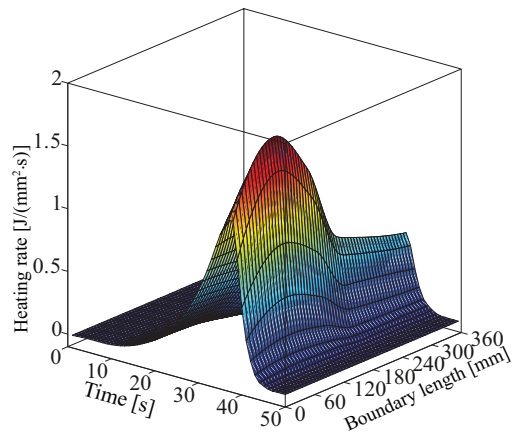


Figure 14: The time-dependent heating rate along the boundary Γ_1 (data from [53]).

Table 5: Comparison between adjoint gradient and finite difference gradient for the plunger design with $T = 500s$.

I	Component	Adjoint gradient	Finite difference gradient	Relative difference
C_1	1	-6.0855×10^5	-6.1078×10^5	0.36%
	2	0	2.6388	~
C_2	1	-3.1106×10^5	-3.0308×10^5	2.57%
	2	3.3747×10^5	3.4454×10^5	2.10%
C_3	1	2.0409×10^6	2.0891×10^6	2.36%
	2	1.1000×10^6	1.1451×10^6	4.10%
C_4	1	0.8443×10^5	0.8658×10^5	2.54%
	2	-6.2055×10^5	-6.2875×10^5	1.32%
C_5	1	0	6.3077	~
	2	-1.5463×10^6	-1.5391×10^6	0.47%

subjected to a volume constraint

$$\Sigma \leq \Sigma_0,$$

where Σ_0 is the volume of the initial design. The design time is chosen as $T = 180$ s, which includes an initial time interval with a large heating rate followed by a time interval with a lower rate in order to take both loading scenarios into account. The locations of five design control points, C_I , $I = 1, \dots, 5$, which characterize the internal boundary Γ_2 , are chosen as design variables. Control point C_1 is only allowed to move horizontally while control point C_5 is only allowed to move vertically. The external boundary Γ_1 is kept fixed since its optimization depends mostly on aerodynamic requirements. Further, it is assumed that the heating rate on the external boundary, as shown in Fig. 14, is the same regardless of the shape of the internal boundary. As in previous examples, the sensitivity obtained using the continuous adjoint method was verified with the finite difference method, which is shown in Table 8. It should be noted that for the sensitivity of the design control point C_5 , extra terms need to be included into the sensitivity formulation due to the geometry discontinuity involved. The optimization is performed using the Algorithm 1 and the convergence history is shown Fig. 15(a), which indicates that after about six iterations the process has converged. The optimal shape obtained is plotted in Fig. 15(b). From the figure, it can be seen that the optimal shape requires a thicker layer close to the tip, which comes at the expense of the thickness in the rear part in order to preserve the total volume. With this change in shape, the heat conducted inside the front section can be reduced about 11.5% compared with the original design, as shown in Fig. 15(a). The control points and the corresponding weights of the initial and optimal designs are presented in 9. The knot vectors used for the two directions in the index space are $\xi = [0 \ 0 \ 0 \ 1/2 \ 1/2 \ 1 \ 1 \ 1]$ and $\eta = [0 \ 0 \ 0 \ 1 \ 1 \ 1]$, respectively.

7. Conclusions

In this work, the continuous adjoint shape sensitivity analysis for transient heat conduction problems is reformulated taking into consideration the discontinuities involved in the objective

Table 6: Comparison between adjoint gradient and finite difference gradient for the plunger design with $T = 1000s$.

I	Component	Adjoint gradient	Finite difference gradient	Relative difference
C_1	1	-6.6243×10^5	-6.6298×10^5	0.08%
	2	0	2.6388	~
C_2	1	-0.6969×10^6	-7.1838×10^6	3.09%
	2	1.0752×10^6	1.0903×10^6	1.40%
C_3	1	7.0695×10^6	7.1734×10^6	1.47%
	2	6.3920×10^6	6.5046×10^6	1.76%
C_4	1	0.3165×10^6	0.3213×10^6	1.52%
	2	-1.2643×10^6	-1.2761×10^6	0.93%
C_5	1	0	3.7868	~
	2	-5.4290×10^5	-5.4290×10^5	2.94%

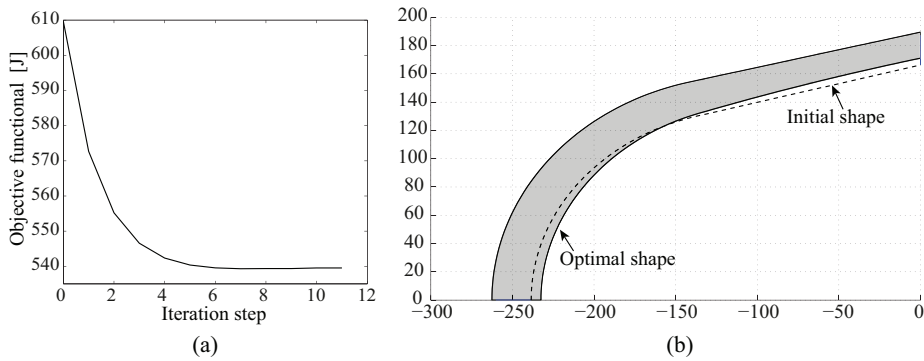


Figure 15: (a) Iteration history of the optimization and (b) the optimal shape.

functionals. The continuous sensitivity analysis, which is applicable to general shape optimization problems, is discretized in the context of isogeometric analysis. The control points of a NURBS description of the shape are used as design variables, which allows a seamless integration between optimization and analysis. The optimization and the analysis are performed at two levels of discretization (coarse and fine, respectively), but no loss of geometrical information occurs in this process. The methodology was tested with benchmark problems, which also illustrate the flexibility provided by the characteristic functions to measure the design performance in selected places and times. The transient isogeometric shape optimization was subsequently applied to cases where thermal conditions fluctuate during operation such as the thermal protection system design for a reentry ballistic vehicle. In these examples, it is shown that shape optimization with accountability of transient states is an attractive approach in applications where active control is not economically feasible. Furthermore, it is possible to combine shape optimization (as passive control) with an active control approach in view of increasing the efficiency of the thermal management. Following a similar approach, it is also possible to further extend the methodology and framework to include other transient situations, such as mechanical problems

Table 7: Geometry information of the plunger design problem (W^I : the weight of I th control point)

$I(i, j)$	Initial design			Optimal design ($T = 500s$)			Optimal design ($T = 1000s$)		
	Locations		W^I	Locations		W^I	Locations		W^I
	x_1^I	x_2^I		x_1^I	x_2^I		x_1^I	x_2^I	
(1, 1)	0.00	100.00	1.00	0.00	100.00	1.00	0.00	100.00	1.00
(2, 1)	0.00	80.00	1.00	0.00	80.00	1.00	0.00	80.00	1.00
(3, 1)	0.00	30.00	1.00	0.00	30.00	1.00	0.00	30.00	1.00
(4, 1)	0.00	0.00	0.71	0.00	0.00	0.71	0.00	0.00	0.71
(5, 1)	30.00	0.00	1.00	30.00	0.00	1.00	30.00	0.00	1.00
(6, 1)	140.00	0.00	1.00	140.00	0.00	1.00	140.00	0.00	1.00
(7, 1)	200.00	0.00	1.00	200.00	0.00	1.00	200.00	0.00	1.00
(1, 2)	15.00	100.00	1.00	15.00	100.00	1.00	15.00	100.00	1.00
(2, 2)	15.00	80.00	1.00	31.7794	80.00	1.00	18.1916	80.00	1.00
(3, 2)	15.00	65.00	1.00	21.4882	65.00	1.00	13.1518	65.00	1.00
(4, 2)	15.00	20.00	1.00	10.00	10.00	1.00	10.00	10.00	1.00
(5, 2)	90.00	20.00	1.00	90.00	20.9044	1.00	90.00	19.4040	1.00
(6, 2)	145.00	20.00	1.00	145.00	10.00	1.00	145.00	19.4109	1.00
(7, 2)	200.00	20.00	1.00	200.00	20.00	1.00	200.00	20.00	1.00
(1, 3)	30.00	100.00	1.00	30.00	100.00	1.00	30.00	100.00	1.00
(2, 3)	30.00	80.00	1.00	53.9705	80.00	1.00	34.5594	80.00	1.00
(3, 3)	30.00	65.00	1.00	39.2689	65.00	1.00	27.3598	65.00	1.00
(4, 3)	30.00	45.00	1.00	20.00	20.00	1.00	20.00	20.00	1.00
(5, 3)	70.00	45.00	1.00	70.00	46.2920	1.00	70.00	44.1486	1.00
(6, 3)	120.00	45.00	1.00	120.00	20.00	1.00	120.00	44.1584	1.00
(7, 3)	200.00	45.00	1.00	200.00	45.00	1.00	200.00	45.00	1.00

involving stress wave propagation.

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Table 8: Comparison between adjoint gradient and finite difference gradient for the design of the TPS panel.

I(i,j)	Component	Adjoint gradient	Finite difference gradient	Relative difference
C_1	1	-1.2554×10^2	-1.1977	4.60%
	2	~	~	~
C_2	1	-0.8187×10^2	-0.8598×10^2	4.78%
	2	0.5957×10^2	0.5768×10^2	3.17%
C_3	1	-0.6797×10^2	-0.7065×10^2	3.94%
	2	1.1446×10^2	1.0911×10^2	4.90%
C_4	1	-0.09265×10^2	-0.09031×10^2	2.58%
	2	0.3558×10^2	0.3710×10^2	4.27%
C_5	1	-0.09137×10^2	-0.09426×10^2	3.07%
	2	0.3460×10^2	0.3612×10^2	4.22%

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Table 9: Geometry information of the TPS design problem

Initial design				Optimal design			
$I(i, j)$	Locations		Weights	$I(i, j)$	Locations		Weights
	x_1^I	x_2^I			x_1^I	x_2^I	
(1, 1)	-262.5000	0.0000	1.00	(1, 1)	-262.5000	0.0000	1.00
(2, 1)	-262.5036	124.0308	0.79	(2, 1)	-262.5036	124.0308	0.79
(3, 1)	-142.1551	154.0344	1.00	(3, 1)	-142.1551	154.0344	1.00
(4, 1)	-71.1000	171.8000	1.00	(4, 1)	-71.1000	171.8000	1.00
(5, 1)	0.0000	189.5000	1.00	(5, 1)	0.0000	189.5000	1.00
(1, 2)	-250.5000	0.0000	1.00	(1, 2)	-247.4999	0.0000	1.00
(2, 2)	-250.5034	114.6552	0.79	(2, 2)	-247.7293	112.8708	0.79
(3, 2)	-139.2520	142.3909	1.00	(3, 2)	-138.9856	143.3841	1.00
(4, 2)	-69.6260	160.1275	1.00	(4, 2)	-70.4480	163.1073	1.00
(5, 2)	0.0000	177.8635	1.00	(5, 2)	0.0000	180.2997	1.00
(1, 3)	-238.5000	0.0000	1.00	(1, 3)	-232.4998	0.0000	1.00
(2, 3)	-238.5033	105.2797	0.79	(2, 3)	-232.9551	101.7108	0.79
(3, 3)	-134.3490	130.7473	1.00	(3, 3)	-133.8162	132.7337	1.00
(4, 3)	-67.1745	148.4900	1.00	(4, 3)	-68.8184	154.4496	1.00
(5, 3)	0.0000	166.2270	1.00	(5, 3)	0.0000	171.0993	1.00

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