

An introduction to optimal design

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Foreword (I)

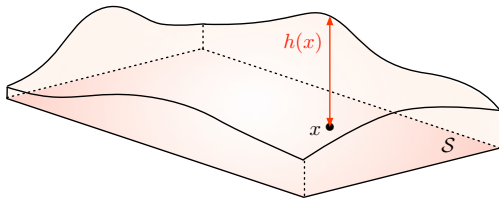
This course focuses on **parametric optimal design** problems, of the form:

$$\min_{h \in H} J(h) \text{ s.t. } G(h) = 0, \quad (\mathcal{P})$$

where

- The **design variable** h belongs to a **fixed vector space** H ;
- The objective and constraint functions $J(h)$ and $G(h)$ depend on h via a **state** $u_h \in V$, solution to an h -dependent boundary value problem:

$$\text{Search for } u_h \in V \text{ s.t. } \forall v \in V, \quad a_h(u_h, v) = \ell_h(v).$$

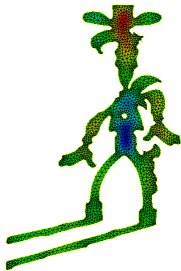


An elastic plate can be described by its height $h : S \rightarrow \mathbb{R}$ over a fixed cross-section S .

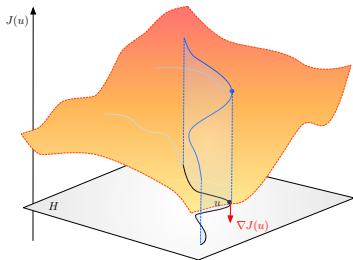
Foreword (II)

In Course 1, we have seen that:

- For a given design $h \in H$, the numerical simulation of u_h can be realized by the **Finite Element Method**, e.g. using **FreeFem**.
- The solution of (\mathcal{P}) can be achieved by a (constrained) **gradient-based** strategy.



Solution of a PDE by the Finite Element Method.



The gradient algorithm in action

⇒ This course is devoted to the calculation of the **derivative** of a function $J(h)$ depending on h via the solution to a PDE depending on h .

Part II

Optimal control and parametric optimization problems

- 1 Generalities about parametric optimization problems
 - Presentation of a model problem
 - Non existence of optimal design
- 2 Calculation of derivatives of functions of the design

A model problem involving the conductivity equation (I)

- We aim to optimize the **thermal conductivity** $h : D \rightarrow \mathbb{R}$ inside a given domain $D \subset \mathbb{R}^d$.
- The **temperature** u_h is the solution in $H^1(D)$ to the “state”, conductivity equation:

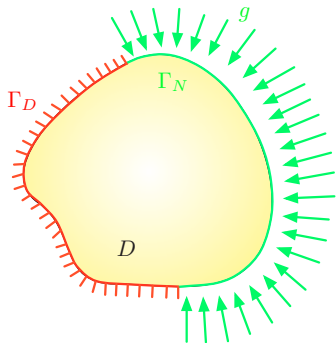
$$\begin{cases} -\operatorname{div}(h\nabla u_h) & = f & \text{in } D, \\ u_h & = 0 & \text{on } \Gamma_D, \\ h \frac{\partial u_h}{\partial n} & = g & \text{on } \Gamma_N, \end{cases}$$

where $f \in L^2(D)$ and $g \in L^2(\Gamma_N)$ are sources.

- The set \mathcal{U}_{ad} of **design variables** is:

$$\mathcal{U}_{\text{ad}} = \left\{ h \in L^\infty(D), \alpha \leq h(x) \leq \beta \text{ a.e. } x \in D \right\} \subset L^\infty(D),$$

where $0 < \alpha < \beta$ are fixed bounds.



A model problem involving the conductivity equation (II)

- To set ideas, we consider an unconstrained problem of the form:

$$\min_{h \in \mathcal{U}_{\text{ad}}} J(h), \text{ where } J(h) = \int_D j(u_h) \, dx,$$

and $j : \mathbb{R} \rightarrow \mathbb{R}$ is a smooth function satisfying **growth conditions**:

$$\forall s \in \mathbb{R}, |j(s)| \leq C(1 + |s|^2), \text{ and } j'(s) \leq C(1 + |s|).$$

- Many variants are possible, e.g. featuring **constraints** on h or u_h .
- In this setting,
 - For any $h \in \mathcal{U}_{\text{ad}}$, the state u_h is evaluated on the **fixed** domain D ;
 - The design variable h acts as a **parameter** in the coefficients of the state equation.
- Before dealing with the (local) resolution of this problem, let us inquire about the existence of **global solutions**.

Part II

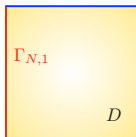
Optimal control and parametric optimization problems

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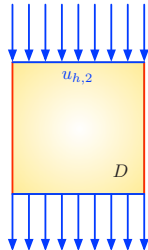
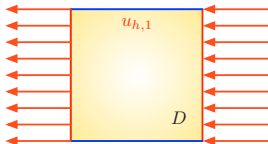
Non existence of optimal design (I)

- This counter-example is discussed in details in [All] §5.2.
- The defining domain is the unit square $D = (0, 1)^2$.
- We consider a variant of the model problem, made of two physical scenarios:

$$\left\{ \begin{array}{ll} -\operatorname{div}(h\nabla u_{h,1}) = 0 & \text{in } D, \\ h \frac{\partial u_{h,1}}{\partial n} = e_1 \cdot n & \text{in } \Gamma_{N,1}, \\ h \frac{\partial u_{h,1}}{\partial n} = 0 & \text{in } \Gamma_{N,2}, \end{array} \right. \quad \text{and} \quad \left\{ \begin{array}{ll} -\operatorname{div}(h\nabla u_{h,2}) = 0 & \text{in } D, \\ h \frac{\partial u_{h,2}}{\partial n} = 0 & \text{in } \Gamma_{N,1}, \\ h \frac{\partial u_{h,2}}{\partial n} = e_2 \cdot n & \text{in } \Gamma_{N,2}. \end{array} \right.$$



$\Gamma_{N,2}$



(Left) Boundary conditions, (middle) boundary data for $u_{h,1}$; (right) boundary data for $u_{h,2}$.

Non existence of optimal design (II)

The optimization problem of interest in this context is:

$$\min_{h \in \mathcal{U}_{\text{ad}}} J(h),$$

where the objective function is:

$$J(h) = \underbrace{\int_{\Gamma_{N,1}} \mathbf{e}_1 \cdot \mathbf{n} u_{h,1} \, ds}_{= \int_{\text{right}} u_{h,1} \, ds - \int_{\text{left}} u_{h,1} \, ds > 0} - \underbrace{\int_{\Gamma_{N,2}} \mathbf{e}_2 \cdot \mathbf{n} u_{h,2} \, ds}_{= \int_{\text{top}} u_{h,1} \, ds - \int_{\text{bottom}} u_{h,1} \, ds > 0},$$

and the set \mathcal{U}_{ad} of admissible designs encompasses a **volume constraint**:

$$\mathcal{U}_{\text{ad}} = \left\{ h \in L^\infty(D), \quad \alpha < h(x) < \beta \text{ a.e. } x \in D, \quad \int_D h \, dx = V_T \right\}.$$

In other terms, one aims to

- **Minimize** the temperature difference between the left and right sides in **Case 1**.
- **Maximize** the temperature difference between the top and bottom sides in **Case 2**.

Non existence of optimal design (III)

Theorem (Non existence of global solution).

The parametric optimization problem $\min_{h \in \mathcal{U}_{\text{ad}}} J(h)$ does not have a global solution.

Hint of the proof: The proof comprises three stages:

Step 1: One calculates a **lower bound** m on the values of $J(h)$ for $h \in \mathcal{U}_{\text{ad}}$:

$$\forall h \in \mathcal{U}_{\text{ad}}, J(h) \geq m.$$

Step 2: One proves that the value m cannot be attained by an element in \mathcal{U}_{ad} :

$$\forall h \in \mathcal{U}_{\text{ad}}, J(h) > m.$$

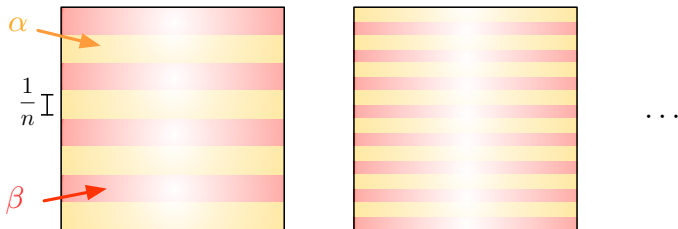
Step 3: One constructs a **minimizing sequence** of designs $h^n \in \mathcal{U}_{\text{ad}}$:

$$J(h^n) \xrightarrow{n \rightarrow \infty} m.$$

Hence, m is the infimum of $J(h)$ over \mathcal{U}_{ad} but it is not attained by any $h \in \mathcal{U}_{\text{ad}}$.

Non existence of optimal design (IV)

- The minimizing sequence h^n is constructed as a **laminate**, i.e. a succession of layers with maximum and minimum conductivities.



Two elements in a minimizing sequence h^n of conductivities: the horizontal stripes allow to reduce the temperature difference between the left and right sides, while insulating the top side from the bottom one.

- This is an expression of the **homogenization effect**:

To get more optimized, designs tend to create very thin structures, at the microscopic level.

Non existence of optimal design (V)

- In general, optimal design problems do not have global solutions, for a deep physical reason:

It is generally possible to achieve better performance by using smaller patterns.

- See [Mu] for many such examples of non existence of optimal design.
- To ensure existence of an optimal design, two techniques are usually employed:
 - **Relaxation**: the set \mathcal{U}_{ad} of admissible designs is **enlarged** to enclose “microscopic designs”. This is the essence of the **Homogenization method** [All2].
 - **Restriction**: the set \mathcal{U}_{ad} is restricted to, e.g. **more regular** designs, with fixed scale patterns.
- In practice, one is often interested in the search of **local minimizers**, which are e.g. “close” to an initial design inspired by intuition.

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 - The rigorous practice of the adjoint method
 - A higher view of the adjoint method
 - The formal method of C ea

Derivative of the objective function (I)

Let us return to our (further simplified) toy problem:

$$\min_{h \in \mathcal{U}_{\text{ad}}} J(h),$$

where

$$J(h) = \int_D j(u_h) \, dx,$$

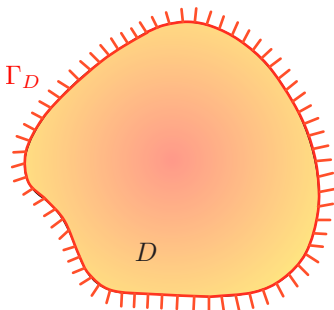
the set of admissible designs is:

$$\mathcal{U}_{\text{ad}} = \left\{ h \in L^\infty(D), \alpha \leq h(x) \leq \beta \text{ a.e. } x \in D \right\},$$

and the **temperature** u_h is the solution in $H_0^1(D)$ to:

$$\begin{cases} -\operatorname{div}(h \nabla u_h) & = f & \text{in } D, \\ u_h & = 0 & \text{on } \partial D. \end{cases}$$

We aim to prove that $J(h)$ is differentiable, and to calculate its **derivative**.



Derivative of the objective function (II)

For a given design $h \in \mathcal{U}_{\text{ad}}$,

- One **variational formulation** characterizing u_h is:

$$\text{Search for } u_h \in H_0^1(D) \text{ s.t. } \forall v \in H_0^1(D), \quad \int_D h \nabla u_h \cdot \nabla v \, dx = \int_D f v \, dx.$$

- This problem has a unique solution $u_h \in H_0^1(D)$, which satisfies:

$$\|u_h\|_{H_0^1(D)} \leq C \|f\|_{L^2(D)},$$

for some constant $C > 0$, owing to the **Lax-Milgram theorem**.

Derivative of the objective function (III)

Theorem.

The objective function

$$J(h) = \int_D j(u_h) \, dx$$

is Fréchet differentiable at any $h \in \mathcal{U}_{\text{ad}}$, and its derivative reads

$$\forall \hat{h} \in L^\infty(D), \quad J'(h)(\hat{h}) = \int_D (\nabla u_h \cdot \nabla p_h) \hat{h} \, dx,$$

where the **adjoint state** $p_h \in H_0^1(D)$ is the unique solution to the problem:

$$\begin{cases} -\operatorname{div}(h \nabla p_h) = -j'(u_h) & \text{in } D, \\ p_h = 0 & \text{on } \partial D. \end{cases}$$

The adjoint method (I)

Proof: The proof is divided into three steps:

- 1 Using the **implicit function theorem**, we prove that the design-to-state mapping

$$\mathcal{U}_{\text{ad}} \ni h \mapsto u_h \in H_0^1(D)$$

is **Fréchet differentiable**, with derivative $\hat{h} \mapsto u'_h(\hat{h})$.

- 2 We characterize $u'_h(\hat{h})$ as the solution to an h, \hat{h} dependent variational problem.
- 3 We calculate the derivative of $J(h)$ in terms of $u'_h(\hat{h})$, thanks to the **chain rule**.
- 4 We give a more convenient structure to $J'(h)(\hat{h})$, introducing an **adjoint state** ρ_h to eliminate the occurrence of $u'_h(\hat{h})$.

Reminder: the implicit function theorem

The **implicit function theorem** is a key result, ensuring the **existence** and **smoothness** of a solution $u = u_\theta$ to a parametrized, non linear equation of the form:

$$\mathcal{F}(\theta, u) = 0,$$

where u is the unknown and θ is a “parameter”; see [La], Chap. I, Th. 5.9.

Theorem (Implicit function theorem).

Let Θ, E, F be Banach spaces, $\mathcal{V} \subset \Theta$, $U \subset E$ be open sets. and $\mathcal{F} : \mathcal{V} \times U \rightarrow F$ be a function of class C^p for $p \geq 1$. Let $(\theta_0, u_0) \in \mathcal{V} \times U$ be such that $\mathcal{F}(\theta_0, u_0) = 0$ and assume that:

The derivative $\frac{\partial \mathcal{F}}{\partial u}(\theta_0, u_0) : E \rightarrow F$ is a linear **isomorphism**.

Then there exist open subsets $\mathcal{V}' \subset \mathcal{V}$ of θ_0 in Θ and $U' \subset U$ of u_0 in E , and a mapping $g : \mathcal{V}' \rightarrow U'$ of class C^p satisfying the properties:

- 1 $g(\theta_0) = u_0$,
- 2 For all $\theta \in \mathcal{V}'$, the equation $\mathcal{F}(\theta, u) = 0$ has a unique solution $u \in U'$, given by $u = g(\theta)$.

The adjoint method (II)

Step 1: Differentiability of $h \mapsto u_h$:

- For any $h \in \mathcal{U}_{\text{ad}}$, u_h is the unique solution in $H_0^1(D)$ to the variational problem:

$$\forall v \in H_0^1(D), \quad \int_D h \nabla u_h \cdot \nabla v \, dx = \int_D f v \, dx.$$

- Let

$$\mathcal{F} : \mathcal{U}_{\text{ad}} \times H_0^1(D) \rightarrow H^{-1}(D)$$

be the mapping defined by:

For all $h \in \mathcal{U}_{\text{ad}}$, $u \in H_0^1(D)$, the linear form $\mathcal{F}(u, h) \in H^{-1}(D)$ on $H_0^1(D)$ is:

$$\mathcal{F}(h, u) : v \mapsto \int_D h \nabla u \cdot \nabla v \, dx - \int_D f v \, dx.$$

The adjoint method (III)

One verifies that

- \mathcal{F} is a mapping of class \mathcal{C}^1 ;
- For given $h \in \mathcal{U}_{\text{ad}}$, u_h is the unique solution u to the equation

$$\mathcal{F}(h, u) = 0.$$

- The differential of the partial mapping $u \mapsto \mathcal{F}(h, u)$ reads:

$$H_0^1(D) \ni \hat{u} \mapsto \left[v \mapsto \int_D h \nabla \hat{u} \cdot \nabla v \, dx \right] \in H^{-1}(D).$$

It is an isomorphism, owing to the **Lax-Milgram theorem**:

For all $g \in H^{-1}(D)$, there exists a unique $u \in H_0^1(D)$ s.t.

$$\forall v \in H_0^1(D), \int_D h \nabla u \cdot \nabla v \, dx = \langle g, v \rangle_{H^{-1}(D), H_0^1(D)}.$$

\Rightarrow The **implicit function theorem** guarantees that the mapping $h \mapsto u_h$ is of class \mathcal{C}^1 .

The adjoint method (IV)

Step 2: Characterization of the derivative $u'_h(\hat{h})$ by a variational problem:

- Let us return to the variational formulation for u_h :

$$\forall v \in H_0^1(D), \quad \int_D h \nabla u_h \cdot \nabla v \, dx = \int_D f v \, dx.$$

- Thanks to **Step 1**, we may take derivatives with respect to h in a direction $\hat{h} \in L^\infty(D)$, for an arbitrary test function $v \in h_0^1(D)$:

$$\int_D \hat{h} \nabla u_h \cdot \nabla v \, dx + \int_D h \nabla u'_h(\hat{h}) \cdot \nabla v \, dx = 0.$$

- Hence, for all $\hat{h} \in L^\infty(D)$, $u'_h(\hat{h})$ is the unique solution in $H_0^1(D)$ to:

Search for $u'_h(\hat{h}) \in H_0^1(D)$ s.t. $\forall v \in H_0^1(D)$,

$$\int_D h \nabla u'_h(\hat{h}) \cdot \nabla v \, dx = - \int_D \hat{h} \nabla u_h \cdot \nabla v \, dx.$$

Reminder: the “chain rule”

The **chain rule** is a *fundamental result*, which supplies the **Fréchet derivative** of the **composite** $G \circ F$ of two functions

$$F : U \rightarrow V \text{ and } G : V \rightarrow W$$

between three normed vector spaces $(U, \|\cdot\|_U)$, $(V, \|\cdot\|_V)$ and $(W, \|\cdot\|_W)$.

Theorem (Chain rule).

Let $x \in U$ be a point such that:

- F is Fréchet differentiable at x ;
- G is Fréchet differentiable at $F(x) \in V$.

Then, the composite function $G \circ F : U \rightarrow W$ is **Fréchet differentiable** at x , and its Fréchet derivative $v \mapsto (G \circ F)'(x)(v)$ is the linear mapping defined by:

$$\forall v \in U, (G \circ F)'(x)(v) = G'(F(x))(F'(x)(v)).$$

Step 3: Calculation of the derivative of $J(h)$:

Since $h \mapsto u_h$ is of class C^1 , the **chain rule** yields immediately:

$$\forall \hat{h} \in L^\infty(D), \quad J'(h)(\hat{h}) = \int_D j'(u_h) u'_h(\hat{h}) \, dx.$$

- This expression is **awkward**: the dependence $\hat{h} \mapsto J'(h)(\hat{h})$ is not explicit and it is difficult to find a **descent direction**, i.e. a vector $\hat{h} \in L^\infty(D)$ such that:

$$J'(h)(\hat{h}) < 0.$$

\Rightarrow One would have to try multiple $\hat{h} \in L^\infty(D)$, calculate $u'_h(\hat{h})$, ... until finding one s.t. $J'(h)(\hat{h}) < 0$.

The adjoint method (VI)

Step 4: Reformulation of $J'(h)(\hat{h})$ using an adjoint state:

The **adjoint state** p_h is the unique solution in $H_0^1(D)$ to the variational problem:

$$\forall v \in H_0^1(D), \int_D h \nabla p_h \cdot \nabla v \, dx = - \int_D j'(u_h) v \, dx,$$

to be compared with the variational formulation for $u'_h(\hat{h}) \in H_0^1(D)$:

$$\forall v \in H_0^1(D), \int_D h \nabla u'_h(\hat{h}) \cdot \nabla v \, dx = - \int_D \hat{h} \nabla u_h \cdot \nabla v \, dx.$$

Then, we calculate:

$$\begin{aligned} J'(h)(\hat{h}) &= \int_D j'(u_h) u'_h(\hat{h}) \, dx, \\ &= - \int_D h \nabla p_h \cdot \nabla u'_h(\hat{h}) \, dx, \\ &= - \int_D h \nabla u'_h(\hat{h}) \cdot \nabla p_h \, dx, \\ &= \int_D \hat{h} \nabla u_h \cdot \nabla p_h \, dx. \end{aligned}$$

where the last line uses the variational formulation of $u'_h(\hat{h})$ with p_h as test function.

About the adjoint state (I)

- The above introduction of the **adjoint state** p_h , as a solution to the problem:

$$\forall v \in H_0^1(D), \int_D h \nabla p_h \cdot \nabla v \, dx = - \int_D j'(u_h) v \, dx,$$

and the allowed simplifications may look “miraculous” at first glance.

- After closer inspection, its definition is **exactly what is needed** to simplify $J'(\hat{h})$:
 - The right-hand side is what is needed to write:

$$J'(h)(\hat{h}) = \int_D j'(u_h) u'_h(\hat{h}) \, dx = - \int_D h \nabla u'_h(\hat{h}) \cdot \nabla p_h \, dx.$$

- This last expression, featuring the action of $v \mapsto \int_D h \nabla u'_h(\hat{h}) \cdot \nabla v \, dx$ is the only available information about $u'_h(\hat{h})$:

$$- \int_D h \nabla u'_h(\hat{h}) \cdot \nabla p_h \, dx = \int_D \hat{h} \nabla u_h \cdot \nabla p_h \, dx.$$

About the adjoint state (II)

- The adjoint state p_h satisfies

$$\begin{cases} -\operatorname{div}(h\nabla p_h) = -j'(u_h) & \text{in } D, \\ p_h = 0 & \text{on } \partial D. \end{cases}$$

Loosely speaking, it is a “**virtual temperature**” driven by a source (or sink) equal to the rate of change of the integrand of $J(h)$ at the state described by u_h .

- From the last expression, one obviously obtains a **descent direction**:

$$\hat{h} = -\nabla u_h \cdot \nabla p_h \Rightarrow J'(h)(\hat{h}) < 0,$$

which can be interpreted as the power induced by the “virtual temperature” p_h .

- We shall see soon a second interpretation of p_h as the **Lagrange multiplier** associated to the PDE constraint if we formulate our optimization problem as:

$$\min_{(h,u)} \int_D j(u) \, dx \quad \text{s.t.} \quad \begin{cases} -\operatorname{div}(h\nabla u) = f & \text{in } D, \\ u = 0 & \text{on } \partial D. \end{cases}$$

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The adjoint method in an abstract framework (I)

- Let us consider an abstract optimal design functional, of the form:

$$J(h) = j(u_h),$$

where

- The **design** h belongs to a **Hilbert space** $(H, \langle \cdot, \cdot \rangle_H)$.
- The **state** u_h belongs to another Hilbert space $(V, \langle \cdot, \cdot \rangle_V)$.
- It is the solution to a **boundary value problem**

$$\text{Search for } u_h \in V \text{ s.t. } \mathcal{F}(h, u_h) = 0, \quad (BVP)$$

where $\mathcal{F} : H \times V \rightarrow V$ is a suitable operator.

- The **observable** $j : V \rightarrow \mathbb{R}$ is smooth enough.
- We aim to calculate the derivative $J'(h)(\hat{h})$ and to find a **descent direction** for $J(h)$.

The adjoint method in an abstract framework (II)

The first 3 steps of the previous derivation unfold exactly as in there:

- 1 The **implicit function theorem** ensures that the mapping $h \mapsto u_h$ is differentiable.
- 2 Differentiation in (BVP) yields a characterization of the **derivative** $u'_h(\hat{h}) \in V$:

$$\left[\frac{\partial \mathcal{F}}{\partial h}(h, u_h) \right] (\hat{h}) + \left[\frac{\partial \mathcal{F}}{\partial u}(h, u_h) \right] u'_h(\hat{h}) = 0.$$

- 3 Likewise, by applying the chain rule, we calculate the derivative of $J(h)$:

$$J'(h)(\hat{h}) = \left\langle j'(u_h), u'_h(\hat{h}) \right\rangle_V.$$

Unfortunately, this formula does not easily provide a **descent direction**.

\Rightarrow One would have to try multiple $\hat{h} \in H$, calculate $u'_h(\hat{h})$, ... until finding one such that $J'(h)(\hat{h}) < 0$.

The adjoint method in an abstract framework (III)

- To overcome this issue, we introduce the **adjoint state** $p_h \in V$ as the solution to:

$$\left[\frac{\partial \mathcal{F}}{\partial u}(h, u_h) \right]^* p_h = -j'(u_h),$$

where $\left[\frac{\partial \mathcal{F}}{\partial u}(h, u_h) \right]^* : V \rightarrow V$ is the adjoint operator of $\left[\frac{\partial \mathcal{F}}{\partial u}(h, u_h) \right] : V \rightarrow V$:

$$\forall v, w \in V, \quad \left\langle \left[\frac{\partial \mathcal{F}}{\partial u}(h, u_h) \right]^* v, w \right\rangle_V = \left\langle v, \left[\frac{\partial \mathcal{F}}{\partial u}(h, u_h) \right] w \right\rangle_V.$$

- An elementary calculation yields:

$$\begin{aligned} J'(h)(\hat{h}) &= \left\langle j'(u_h), u'_h(\hat{h}) \right\rangle_V \\ &= - \left\langle \left[\frac{\partial \mathcal{F}}{\partial u}(h, u_h) \right]^* p_h, u'_h(\hat{h}) \right\rangle_V \quad (\text{Def. of } p_h) \\ &= - \left\langle \left[\frac{\partial \mathcal{F}}{\partial u}(h, u_h) \right] u'_h(\hat{h}), p_h \right\rangle_V. \quad (\text{Def. of adjoint operator}) \end{aligned}$$

- Invoking now the problem satisfied by $u'_h(\hat{h})$, we obtain:

$$J'(h)(\hat{h}) = \left\langle \left[\frac{\partial \mathcal{F}}{\partial h}(h, u(h)) \right] \hat{h}, p_h \right\rangle_V.$$

The adjoint method in an abstract framework (IV)

- Let us introduce the **adjoint** $\left[\frac{\partial \mathcal{F}}{\partial h}(h, u_h)\right]^* : V \rightarrow H$ of $\left[\frac{\partial \mathcal{F}}{\partial h}(h, u_h)\right] : H \rightarrow V$:

$$\forall v \in V, h \in H, \quad \left\langle \left[\frac{\partial \mathcal{F}}{\partial h}(h, u_h)\right]^* v, h \right\rangle_H = \left\langle v, \left[\frac{\partial \mathcal{F}}{\partial h}(h, u_h)\right] h \right\rangle_V.$$

We end up with:

$$J'(h)(\hat{h}) = \left\langle \left[\frac{\partial \mathcal{F}}{\partial h}(h, u_h)\right]^* p_h, \hat{h} \right\rangle_H.$$

- Now, a **descent direction** \hat{h} for $J(h)$ is immediately revealed:

$$\hat{h} = - \left[\frac{\partial \mathcal{F}}{\partial h}(h, u_h)\right]^* p_h \Rightarrow J'(h)(\hat{h}) = - \left\| \left[\frac{\partial \mathcal{F}}{\partial h}(h, u_h)\right]^* p_h \right\|_H^2.$$

- The evaluation of this descent direction \hat{h} demands:
 - The calculation of u_h (Finite Element solution);
 - The calculation of p_h (Finite Element solution);
 - The calculation of the operator $\left[\frac{\partial \mathcal{F}}{\partial h}(h, u_h)\right]^*$ (derivative of the explicit dependence of the boundary value problem w.r.t the design).

The adjoint method: final comments

- The adjoint state p_h can be interpreted as the **Lagrange multiplier** associated to the PDE constraint if we formulate the minimization problem of $J(h)$ as:

$$\min_{(h,u)} \int_D j(u) \, dx \text{ s.t. } \begin{cases} -\operatorname{div}(h\nabla u) = f & \text{in } D, \\ u = 0 & \text{on } \partial D. \end{cases}$$

- This general methodology allows to deal with
 - Different physical situations,
 - Other optimal design frameworks, including “true” **geometric optimization**.
- Another (formal and sometimes dangerous) method allows to calculate shape derivatives: **Céa's method**.

Part II

Optimal control and parametric optimization problems

- 1 Generalities about parametric optimization problems
- 2 Calculation of derivatives of functions of the design
 - The rigorous practice of the adjoint method
 - A higher view of the adjoint method
 - The formal method of C ea

The formal method of C ea (I)

- The method of C ea is a **formal way**, inspired from **constrained optimization** theory, to calculate the derivative of $J(h)$.
- It **assumes** that the mapping $h \mapsto u_h$ is differentiable.
- The original, **unconstrained** problem

$$\min_{h \in \mathcal{U}_{\text{ad}}} J(h), \text{ where } J(h) = \int_D j(u_h) \, dx,$$

and u_h is the solution to the conductivity equation:

$$\begin{cases} -\operatorname{div}(h \nabla u_h) = f & \text{in } D, \\ u_h = 0 & \text{on } \partial D, \end{cases}$$

can be rewritten as a **constrained** optimization problem:

$$\min_{\substack{u \in H_0^1(D), \\ h \in \mathcal{U}_{\text{ad}}}} \int_D j(u) \, dx, \text{ s.t. } \begin{cases} -\operatorname{div}(h \nabla u) = f & \text{in } D, \\ u = 0 & \text{on } \partial D, \end{cases}$$

where the PDE relating h to u is accounted for by a **constraint**.

The formal method of C ea (II)

- This alternative viewpoint suggests the definition of the **Lagrangian**:

$$\mathcal{L} : \mathcal{U}_{\text{ad}} \times H_0^1(D) \times H_0^1(D) \rightarrow \mathbb{R},$$

by the formula:

$$\mathcal{L}(h, u, p) = \underbrace{\int_D j(u) \, dx}_{\text{Objective function at stake}} + \underbrace{\int_D h \nabla u \cdot \nabla p \, dx - \int_D f p \, dx}_{\text{Enforcement of the PDE constraint } -\text{div}(h \nabla u) = f \text{ with a Lagrange multiplier } p}.$$

- By construction, it holds, for any $\hat{p} \in H_0^1(D)$:

$$J(h) = \mathcal{L}(h, u_h, \hat{p}).$$

The formal method of Céa (III)

In order to solve the **constrained** optimal design problem, we search for the **saddle points** (u, p) of $\mathcal{L}(h, \cdot, \cdot)$ for each, given $h \in \mathcal{U}_{\text{ad}}$.

- Imposing the partial derivative of \mathcal{L} **with respect to p** to vanish amounts to

$$\forall \hat{p} \in H_0^1(D), \int_D h \nabla u \cdot \nabla \hat{p} \, dx - \int_D f \hat{p} \, dx = 0;$$

this is the variational formulation for $u = u_h$.

- Imposing the partial derivative of \mathcal{L} **with respect to u** to vanish amounts to

$$\forall \hat{u} \in H_0^1(D), \int_D h \nabla p \cdot \nabla \hat{u} \, dx = - \int_D j'(u) \hat{u} \, dx;$$

since $u = u_h$, we recognize the variational formulation for $p = p_h$.

The formal method of Céa (IV)

- Recall that, for arbitrary $\hat{p} \in H_0^1(D)$,

$$J(h) = \mathcal{L}(h, u_h, \hat{p}).$$

- Since we have assumed that $h \mapsto u_h$ is differentiable, the **chain rule** yields:

$$J'(h)(\hat{h}) = \frac{\partial \mathcal{L}}{\partial h}(h, u_h, \hat{p})(\hat{h}) + \frac{\partial \mathcal{L}}{\partial u}(h, u_h, \hat{p})(u'_h(\hat{h})).$$

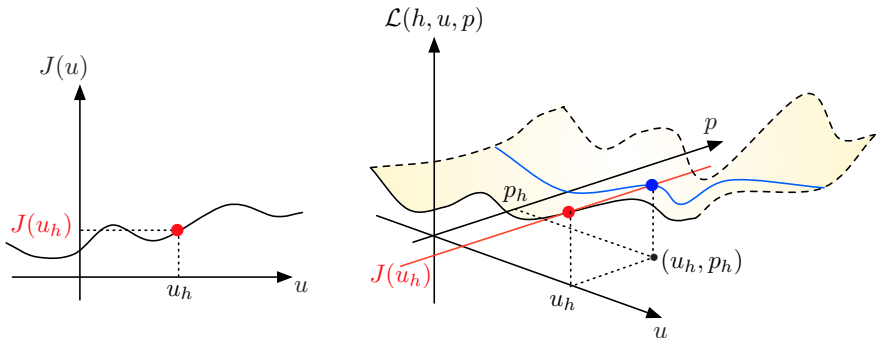
- Now taking $\hat{p} = p_h$, the last term in the above right-hand side vanishes:

$$J'(h)(\hat{h}) = \frac{\partial \mathcal{L}}{\partial h}(h, u_h, p_h)(\hat{h}).$$

- The above partial derivative is the derivative of the mapping $h \mapsto \int_D h \nabla u \cdot \nabla p \, dx$ evaluated at $u = u_h$ and $p = p_h$:

$$J'(h)(\hat{h}) = \int_D \hat{h} \nabla u_h \cdot \nabla p_h \, dx.$$

The formal method of Céa: intuition



intuition: The function $J(h)$ is “twisted” into the value $\mathcal{L}(h, u_h, p_h)$ at the parametrized saddle point (u_h, p_h) , which is “easy” to differentiate with respect to h .

Overview of the next lectures

- **Course 3** Numerical methods for parametric optimization.
 - Numerical implementation recipes;
 - Density-based topology optimization.







- **Practical session** A FreeFem implementation of the SIMP method.

Thank you!

Thank you for your attention!

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