

Plasma boundary reconstruction using topological asymptotic expansion

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joint work with

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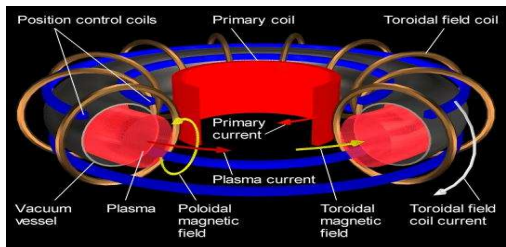
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**Conference honouring Jean Roberts and Jérôme Jaffré,
Paris, December 2014**

Plan

- 1 The plasma problem
- 2 Inverse problem
- 3 Topological sensitivity method
- 4 Algorithm and numerical results
- 5 Conclusion

The Tokamak: The Tokamak is an experimental machine which aims to confine the plasma in a magnetic field to control the nuclear fusion of atoms of mass law. The real-time reconstruction of the plasma magnetic equilibrium in a Tokamak is a key point to access high performance regimes.



The plasma equilibrium:

We denote by (r, φ, z) the three-dimensional cylindrical coordinates system. Since the tokamak is an axisymmetric toroidal device, we may assume that all magnetic quantities do not depend on the toroidal angle φ .

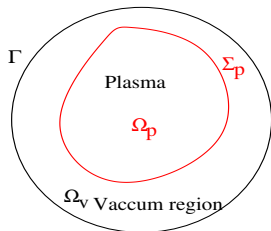
The plasma equilibrium may be studied in any cross section (r, z) , named poloidal section.

It is described by the equation

$$L\psi = 0 \text{ in } \Omega_v$$

- Ω_v is the domain included between the tokamak boundary Γ and the plasma boundary Σ , called the vacuum region.
- ψ is the poloidal magnetic flux.
- \mathcal{L} is the Grad-Shafranov operator

$$\mathcal{L} = -\frac{\partial}{\partial r} \left(\frac{1}{\mu r} \frac{\partial}{\partial r} \right) - \frac{\partial}{\partial z} \left(\frac{1}{\mu r} \frac{\partial}{\partial z} \right)$$

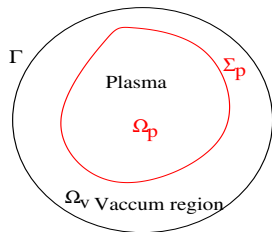


The Tokamak problem

The Tokamak problem : We consider here the inverse problem of determining plasma boundary Σ_p location from over-specified boundary data on Γ .

Knowing a complete set of Cauchy data, the poloidal flux ψ satisfies the system

$$\begin{cases} L\psi & = 0 & \text{in } \Omega \setminus \overline{\Omega_p}, \\ \frac{1}{r} \frac{\partial \psi}{\partial n} & = \Phi & \text{on } \Gamma, \\ \psi & = \psi_m & \text{on } \Gamma, \\ \psi & = 0 & \text{on } \Sigma_p. \end{cases}$$



- Ω is the domain limited by the boundary Γ ,
- Φ is the magnetic field and ψ_m is the measured poloidal flux on Γ .

In this formulation the domain Ω is unknown since the free plasma boundary Σ_p is unknown. This problem is ill posed in the sense of Hadamard.

The Tokamak problem

Formulation of the problem: In order to determine the unknown plasma boundary Σ_p location we propose two formulations for the considered inverse problem.

The first formulation : consists in finding the optimal location of the plasma boundary Σ_p minimizing the cost function

$$T(\psi, \Sigma_p) := \int_{\Gamma} |\psi - \psi_m|^2 ds$$

where ψ is the solution to

$$\left\{ \begin{array}{ll} L\psi = 0 & \text{in } \Omega \setminus \overline{\Omega_p}, \\ \frac{1}{r} \frac{\partial \psi}{\partial n} = \Phi & \text{on } \Gamma, \\ \psi = 0 & \text{on } \Sigma_p. \end{array} \right.$$

The Tokamak problem

The Kohn-Vogelius cost function : For any plasma domain Ω_p , we define two forward problems:

the first one is associated to the magnetic field Φ (Newmann datum):

$$(\mathcal{P}_N) \begin{cases} L\psi_N = 0 & \text{in } \Omega \setminus \overline{\Omega_p} \\ \frac{1}{r} \frac{\partial \psi_N}{\partial n} = \Phi & \text{on } \Gamma \\ \psi_N = 0 & \text{on } \Sigma_p. \end{cases}$$

the second one is associated to the measured poloidal flux ψ_m (Dirichlet datum)

$$(\mathcal{P}_D) \begin{cases} L\psi_D = 0 & \text{in } \Omega \setminus \overline{\Omega_p} \\ \psi_D = \psi_m & \text{on } \Gamma \\ \psi_D = 0 & \text{on } \Sigma_p. \end{cases}$$

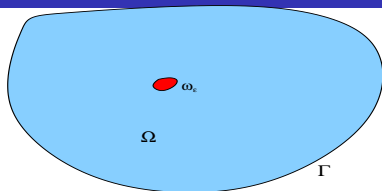
The identification process is based on the minimization of the following energy function

$$\mathcal{K}(\Omega \setminus \overline{\Omega_p}) = \int_{\Omega \setminus \overline{\Omega_p}} \frac{1}{r} |\nabla \psi_D - \nabla \psi_N|^2 dx.$$

Topological Sensitivity analysis

Topological gradient method:

Main idea : studying the variation of the design function \mathcal{J} with respect to the creation of a small hole ω_ε in Ω .



It leads to an asymptotic expansion of the form

$$\mathcal{J}(\Omega \setminus \overline{\omega_\varepsilon}) - \mathcal{J}(\Omega) = f(\varepsilon)\delta\mathcal{J}(z) + o(f(\varepsilon)).$$

where

- $f(\varepsilon)$: is a scalar function known explicitly and goes to zero with ε
 $\lim_{\varepsilon \rightarrow 0} f(\varepsilon) = 0$.
- $\delta\mathcal{J}$: topological gradient, called also topological sensitivity.

In order to minimize the cost function, the **best location** to insert a small hole in Ω is where $\delta\mathcal{J}$ is **most negative**.

In fact if $\delta\mathcal{J}(z) < 0$, we have $\mathcal{J}(\Omega \setminus \overline{\omega_\varepsilon}) < \mathcal{J}(\Omega)$ for small ε .

History :

- ✘ It has been introduced by Schumacher [1995] as “numerical approach” in structural mechanics using circular holes and Neumann b.c.
- ✘ Sokolowski [1999]: extended this idea to more general function using the adjoint method (case circular holes and Neumann b.c.).
- ✘ Masmoudi [2001]: introduced the Dirichlet condition case and given a more general approach to compute the topological gradient.

More recently, it has been generalized for different PDE: Elasticity, Laplace, Stokes, Helmholtz, Maxwell, Navier-Stokes,

The Topological Sensitivity

Example: The Laplace operator admits an asymptotic expansion on the form

$$\mathcal{J}(\Omega \setminus \overline{\omega_\varepsilon}) - \mathcal{J}(\Omega) = f(\varepsilon)\delta\mathcal{J}(z) + o(f(\varepsilon)).$$

The topological gradient $\delta\mathcal{J}$ and the scalar function $f(\varepsilon)$ are described by the following table

B.C. on $\partial\omega_\varepsilon$	Topo. gradient $\delta\mathcal{J}$	function $f(\varepsilon)$
2D Dirichlet	$4\pi u_0(z) v_0(z) + \delta J$	$-1/\log(\varepsilon)$
3D Dirichlet	$6\pi u_0(z) v_0(z) + \delta J$	ε
2D Neumann	$-2\pi \nabla u_0(z) \cdot \nabla v_0(z) + \delta J$	ε^2
3D Neumann	$-2\pi \nabla u_0(z) \cdot \nabla v_0(z) + \delta J$	ε^3

- u_0 : solution to the Laplace operator, computed in the non perturbed domain.
- v_0 : solution to the adjoint problem, computed in the non perturbed domain.
- the term δJ depends on the considered cost function.

Topological Sensitivity analysis

It has been successfully used for various applications:

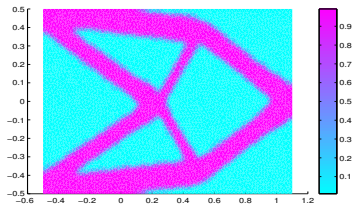
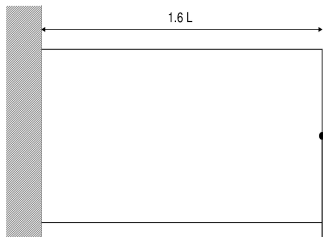
Structural mechanics:

 Guarreau, Guillaume, Masmoudi (2001)

Maximization of the compliance for a 2D cantilever beam: The initial domain is a plain rectangle with one edge clamped and a pointwise load is applied to the middle of the opposite edge.

Aim : find the optimal domain with a volume less than 40% of the initial one.

Idea : sensitivity analysis w.r.t. remove small part of the domain



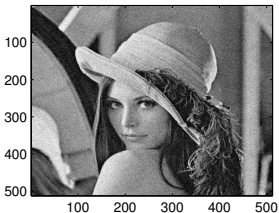
Topological Sensitivity analysis

Image restoration: based on the edges detection.

Idea : edges are considered as a set of small craks.



Jaafar, Jaoua, Masmoudi, Siala (2006)



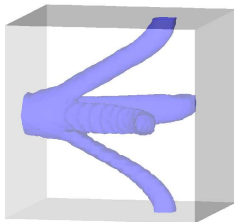
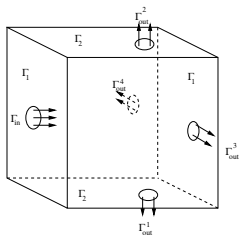
Lena example: Original image, Noisy image(20% g. noise) Restorated image

Fluid mechanics:

 Hassine, Abdelwahed, Masmoudi (2008)

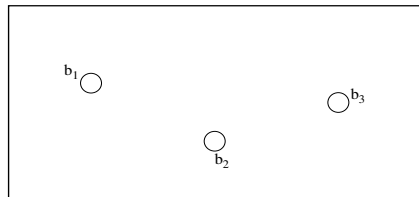
Aim : Find the optimal shape design of the tubes that connect the inlet to the outlets of the cavity minimizing the dissipated power in the fluid.

Idea : sensitivity analysis w.r.t. inserting a small obstacle in the fluid flow

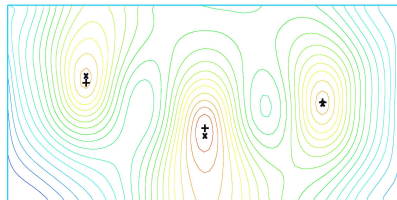


Main features of the topological gradient method:

- 1 No initial guess, with a priori poor information on the optimal shape.
- 2 It depends on solutions computed on the safe body.
- 3 It is fast and easy to be implemented.
- 4 The flaws location are characterized as the local minima of the topological gradient δj .



Exact locations



Isovalues of δj

Topological Sensitivity analysis for the Grad-Shafranov equation

Sensitivity analysis for the tracking function: Let ω_ε be a small geometry perturbation inside the domain Ω with a Dirichlet boundary condition on $\partial\omega_\varepsilon$. The tracking function is defined by

$$T(\Omega \setminus \overline{\omega_\varepsilon}) = \int_{\Gamma} |\psi_\varepsilon - \psi_m|^2$$

where ψ_ε satisfies the problem

$$\begin{cases} L\psi_\varepsilon & = 0 & \text{in } \Omega \setminus \overline{\omega_\varepsilon}, \\ \frac{1}{r} \frac{\partial \psi_\varepsilon}{\partial n} & = \Phi & \text{on } \Gamma, \\ \psi_\varepsilon & = 0 & \text{on } \partial\omega_\varepsilon. \end{cases}$$

Theorem: If $\omega_\varepsilon = X_0 + \varepsilon\omega \subset \Omega$ where $X_0 \in \Omega$, $\varepsilon > 0$ and $\omega \subset \mathbb{R}^2$ is a given, regular and bounded domain containing the origin, the tracking function T admits the following asymptotic expansion

$$T(\Omega \setminus \overline{\omega_\varepsilon}) = T(\Omega) - \frac{1}{\log(\varepsilon)} \frac{2\pi}{x_0} \psi_0(X_0) \phi_0(X_0) + o\left(-\frac{1}{\log(\varepsilon)}\right)$$

where ϕ_0 is the solution to the associated adjoint problem.

The topological sensitivity analysis for the Kohn-Vogelius function:

the Kohn-Vogelius function \mathcal{K} is defined by

$$\mathcal{K}(\Omega \setminus \overline{\omega_\varepsilon}) = \int_{\Omega \setminus \overline{\omega_\varepsilon}} \frac{1}{r} |\nabla \psi_N^\varepsilon - \nabla \psi_D^\varepsilon|^2 dx,$$

with ψ_N^ε and ψ_D^ε are the solutions to the Neumann and Dirichlet perturbed problems

$$(\mathcal{P}_N^\varepsilon) \begin{cases} L\psi_N^\varepsilon = 0 & \text{in } \Omega \setminus \overline{\omega_\varepsilon} \\ \frac{1}{r} \nabla \psi_N^\varepsilon \mathbf{n} = \Phi & \text{on } \Gamma \\ \psi_N^\varepsilon = 0 & \text{on } \partial\omega_\varepsilon, \end{cases} \quad (\mathcal{P}_D^\varepsilon) \begin{cases} L\psi_D^\varepsilon = 0 & \text{in } \Omega \setminus \overline{\omega_\varepsilon} \\ \psi_D^\varepsilon = \psi_m & \text{on } \Gamma \\ \psi_D^\varepsilon = 0 & \text{on } \partial\omega_\varepsilon. \end{cases}$$

Theorem: The function \mathcal{K} admits the following asymptotic expansion

$$\mathcal{K}(\Omega \setminus \overline{\omega_\varepsilon}) = \mathcal{K}(\Omega) + \frac{-2\pi}{\log(\varepsilon)} \frac{1}{x_0} \left[\left| \psi_N^0(X_0) \right|^2 - \left| \psi_D^0(X_0) \right|^2 \right] + o\left(\frac{-1}{\log(\varepsilon)}\right). \quad (1)$$

Numerical validation of the theoretical asymptotic expansion

For a given small perturbation $\omega_z = z + \varepsilon\omega \subset \Omega$, we will study the variation of the function $\Delta_z(\varepsilon)$

$$\Delta_z(\varepsilon) = j(\Omega_{z,\varepsilon}) - j(\Omega) - \delta j(z), \quad z \in \Omega$$

with respect to ε .

We expect to prove numerically that the function Δ_z satisfies the estimate

$$\Delta_z(\varepsilon) = o\left(\frac{-1}{\log(\varepsilon)}\right).$$

Denoting by β the parameter describing the behaviour of $\Delta_z(\varepsilon)$ with respect to $\frac{-1}{\log(\varepsilon)}$, i.e.

$$|\Delta_z(\varepsilon)| = O\left(\left|\frac{-1}{\log(\varepsilon)}\right|^\beta\right).$$

Then, β can be characterized as the slope of the line approximating the variation of the function $\varepsilon \mapsto \log(|\Delta_z(\varepsilon)|)$ with respect to $\log(-\log(\varepsilon))$.

Numerical validation of the theoretical asymptotic expansion

The tracking function : three locations $\omega^i = X_0^i + \varepsilon B(0, 1) \subset \Omega$, $i = 1, 2, 3$ have been tested, where $\Omega = B(X_0, 1)$ with $X_0 = (2, 0)$.

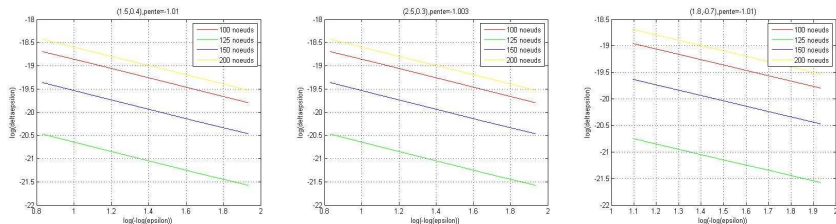


Figure: Variation of $\log(|\Delta z_i(\varepsilon)|)$, $i = 1, \dots, 3$ with respect to $\log(-\log(\varepsilon))$.

The perturbation	ω^1	ω^2	ω^3
Location X_0^i	$X_0^1 = (1.5, 0.4)$	$X_0^2 = (2.5, 0.3)$	$X_0^3 = (1.8, -0.7)$
The slope β_i	$\beta_1 = -1.01$	$\beta_2 = -1.01$	$\beta_3 = -1.03$

Numerical simulations **confirm** the asymptotic expansion.

Numerical validation of the theoretical asymptotic expansion

The Kohn-Vogelius function :

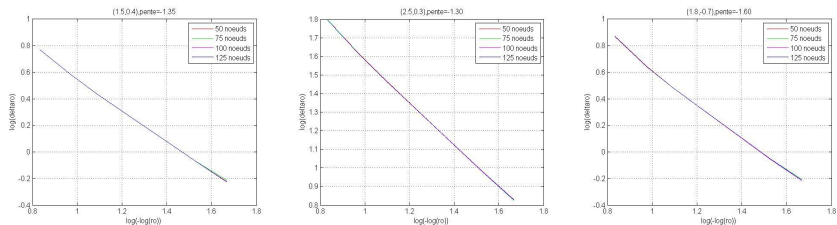


Figure: Variation of $\log(|\Delta_{z_i}(\varepsilon)|)$, $i = 1, \dots, 3$ with respect to $\log(-\log(\varepsilon))$.

The perturbation	ω^1	ω^2	ω^3
The slope β_i	$\beta_1 = -1.35$	$\beta_2 = -1.30$	$\beta_3 = -1.60$

Numerical simulations **confirm** the asymptotic expansion. A slight superconvergence is observed.

One-shot algorithm:

- Compute the topological sensitivity $\delta j(x, y)$, $(x, y) \in \Omega$,
- determine the plasma location by

$$\Omega_\rho = \{(x, y) \in \Omega; \delta j(x, y) \leq (1 - \rho) g_{min}\}$$

where $g_{min} = \min_{(x,y) \in \Omega} \delta j(x, y)$ and $\rho \in]0, 1[$ is a heuristically determined small parameter.

Location of the plasma region from analytic data

Using the Tracking function :

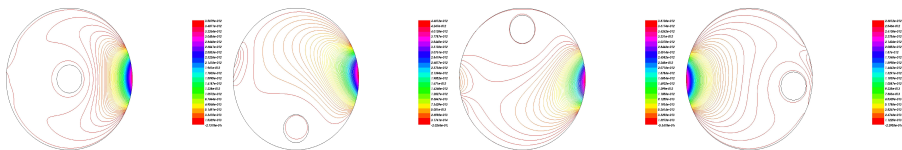


Figure: Various locations of the plasma region

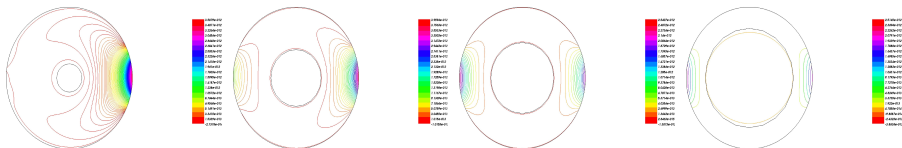


Figure: Various sizes of the plasma region

Flat isovalues \implies instability w.r.t. ε .

Location of the plasma region from analytic data

Using the Kohn-Vogelius function :

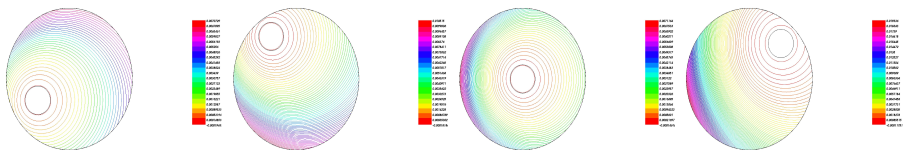


Figure: Various locations of the plasma region

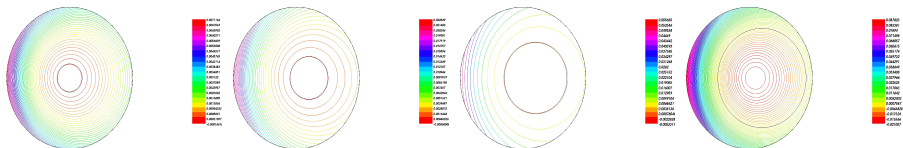


Figure: Various sizes of the plasma region

Sharp isovalues \implies stability w.r.t. ε .

Location of the plasma region from non noisy synthetic data

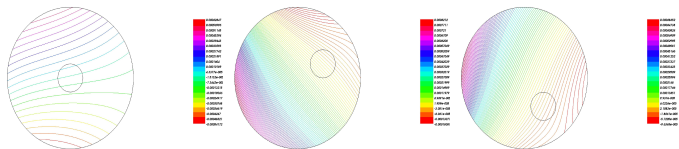


Figure: Reconstruction using the tracking function

Drop the tracking function and use the Kohn-Vogelius one

Reconstruction of smooth shapes using the KV function

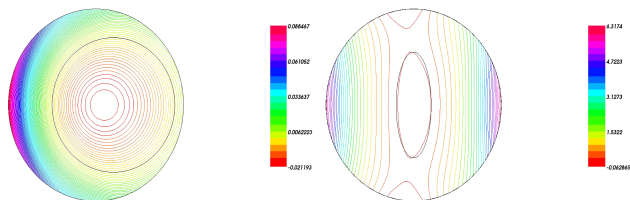


Figure: Reconstruction of a circular and an elliptic shape

the Kohn-Vogelius function efficiently detects both the shape and the location of smooth geometries.

Reconstruction of corner shaped domains using the KV function

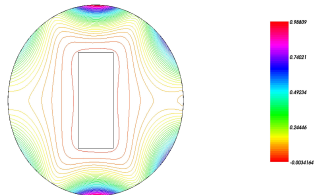


Figure: Reconstruction of shapes with corners

We detect efficiently the location and the shape of a rectangular domain.

Reconstruction of non convex geometries

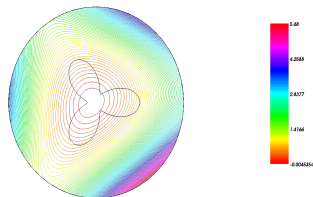


Figure: Reconstruction of a complex geometry

The algorithm locates the region, but fails in reconstructing its shape.

The topological asymptotics provides us with

- A fast algorithm to locate and reconstruct plasma regions in a tokamak
- Its accuracy is good for simple geometries, and quite poor for complex ones
- Might be enough however for real time applications, where

Features :

- No need of a priori information on the location
- One shot algorithm, no iterations
- Once we get the measured data, computations are run on the safe domain only

Prospects for a better accuracy:

- Build an iterative algorithm using the topological asymptotics
- Use the present algorithm as a first guess provider for some shape optimization one ... however more expensive

Thank you for your attention