

SHAPE AND TOPOLOGY OPTIMIZATION TECHNIQUES IN EIT

Martin Kanitsar

28. August 2009

Theory

Introduction in EIT 2
(Shape) Gradient method
Topological expansion

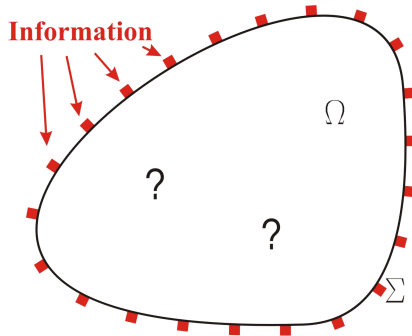
Numerics

Fast sweeping
Finite volume method

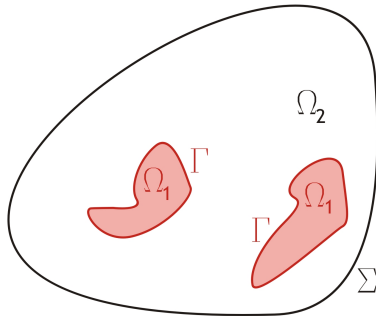
Results

2D
3D

Detect objects in a domain Ω with boundary Σ



Detect objects in a domain Ω with boundary Σ



Concept of Electrical Impedance Tomography

- ▶ background and objects have different physical properties, i.e. conductivity values

Concept of Electrical Impedance Tomography

- ▶ background and objects have different physical properties, i.e. conductivity values
- ▶ apply current through electrodes on Σ
- ▶ take measurements of the electrical potential on Σ

Concept of Electrical Impedance Tomography

- ▶ background and objects have different physical properties, i.e. conductivity values
- ▶ apply current through electrodes on Σ
- ▶ take measurements of the electrical potential on Σ
- ▶ use an appropriate model to describes the physical situation

Applications

- ▶ medicine
 - ▶ blood clots in the lungs
 - ▶ breast cancer
- ▶ geophysics
 - ▶ pollutant plumes
 - ▶ oil inclusions

Mathematical model

$$-\operatorname{div}(q \nabla u) = 0 \text{ in } \Omega, \quad (1.1)$$

$$q \partial_n u = f \text{ on } \Sigma, \quad (1.2)$$

$$\int_{\Sigma} f(x) = 0. \quad (1.3)$$

u ... electrical potential

q ... conductivity

f ... applied current

Relationship to Maxwell's equations

Maxwell's equation in a macroscopic time harmonic field:

$$\nabla \times H = J - i\omega D \quad (1.4)$$

H ... magnetic field	D ... displacement	ω ... angular freq.
E ... electrical field	J ... current density	ϵ ... permittivity

Relationship to Maxwell's equations

Maxwell's equation in a macroscopic time harmonic field:

$$\nabla \times H = J - i\omega D \quad (1.4)$$

H ... magnetic field D ... displacement ω ... angular freq.
E ... electrical field J ... current density ϵ ... permittivity

- ▶ medium reacts linearly : $D = \epsilon E$
- ▶ Ohm's law : $J = qE$
- ▶ relationship : $E = -\nabla u$

$$0 = \operatorname{div}(\nabla \times H) = -\operatorname{div}((q - i\omega\epsilon) \nabla u) . \quad (1.5)$$

Consider real part of (1.5): $-\operatorname{div}(q \nabla u) = 0$.

Mathematical model

The so-called forward problem

$$-\operatorname{div}(q \nabla u) = 0 \text{ in } \Omega, \quad (1.6)$$

$$q \partial_n u = f \text{ on } \Sigma, \quad (1.7)$$

$$\int_{\Sigma} f(x) = 0, \quad (1.8)$$

u ... electrical potential

q ... conductivity

f ... applied current

Mathematical model

The so-called forward problem

$$-\operatorname{div}(q \nabla u) = 0 \text{ in } \Omega, \quad (1.6)$$

$$q \partial_n u = f \text{ on } \Sigma, \quad (1.7)$$

$$\int_{\Sigma} f(x) = 0, \quad (1.8)$$

with a normalization restriction

$$\int_{\Sigma} u - m = 0. \quad (1.9)$$

u ... electrical potential

q ... conductivity

f ... applied current

Regularized output least squares problem

$\{m_i\}_{i=1}^M$...measurements on Σ ,
 M ... number of measurements

We have to minimize

$$J(u, q) = \frac{1}{2} \sum_{i=1}^M \int_{\Sigma} |u_i - m_i|^2 ds + \beta R(q), \quad (2.1)$$

s.t.(1.6) – (1.9)

with $u = (u_1, u_2, \dots, u_M)$ and $R(q) = \int_{\Omega} (|\nabla q|^2 + \epsilon)^{\frac{1}{2}} dx$

Reduction Process

Instead of the constraint minimization problem

$$\begin{aligned} & \min_{(u,q)} J(u, q) \\ & \text{s.t. (1.6) -- (1.9)} \end{aligned}$$

we consider the unconstraint minimization problem

$$\min_q \hat{J}(q)$$

with

$$\hat{J}(q) = J(u(q), q)$$

Representation of q

Assume conductivity is piecewise constant:

$$q(x) = \begin{cases} q_1 & \forall x \in \Omega_1, \\ q_2 & \forall x \in \Omega_2. \end{cases}$$

$$q = q(q_1, q_2, \Omega_1, \Omega_2) \stackrel{\Omega_1 = \Omega \setminus \Omega_2}{=} q(q_1, q_2, \Omega_2)$$

$$\phi(x) = \begin{cases} > 0 & \forall x \in \Omega_1, \\ < 0 & \forall x \in \Omega_2. \end{cases}$$

$$q(q_1, q_2, \phi) = q_1 \mathcal{H}^\epsilon(\phi(x)) + q_2 (1 - \mathcal{H}^\epsilon(\phi(x))). \quad (2.2)$$

Shape gradient

$$\hat{J}'(q)[\tilde{q}] = \langle -\nabla p \nabla u - \beta r(q), \tilde{q} \rangle_{\Omega}. \quad (2.3)$$

with the solution p of the adjoint problem

$$\begin{aligned} -\operatorname{div}(q \nabla p) &= 0 && \text{in } \Omega, \\ q \partial_n p &= u - m && \text{on } \Sigma, \\ \int_{\Omega} p &= 0. \end{aligned}$$

Shape gradient

$$\hat{J}'(q)[\tilde{q}] = \langle -\nabla p \nabla u - \beta r(q), \tilde{q} \rangle_{\Omega}. \quad (2.3)$$

with the solution p of the adjoint problem

$$\begin{aligned} -\operatorname{div}(q \nabla p) &= 0 && \text{in } \Omega, \\ q \partial_n p &= u - m && \text{on } \Sigma, \\ \int_{\Omega} p &= 0. \end{aligned}$$

$$\begin{aligned} q(q_1, q_2, \phi) &= q_1 \mathcal{H}^{\epsilon}(\phi) + q_2 (1 - \mathcal{H}^{\epsilon}(\phi)). \\ q_{\phi}[\tilde{\phi}] &= (q_1 - q_2) \delta^{\epsilon}(\tilde{\phi}), \quad q_{q_1} = \mathcal{H}^{\epsilon}(\phi). \end{aligned} \quad (2.4)$$

Shape gradient

Use the chain rule:

$$\begin{aligned}\hat{J}_\phi(q) [\tilde{\phi}] &= \hat{J}'(q) [q_\phi(\tilde{\phi})] \\ &= \int_{\Omega} -(\nabla p \nabla u + \beta r(q)) (q_1 - q_2) \delta^\epsilon(\tilde{\phi}) \\ &=: \int_{\Omega} \hat{\mathcal{J}}_\phi(\tilde{\phi}).\end{aligned}$$

$$\hat{J}_{q_1}(q) = \int_{\Omega} -(\nabla p \nabla u + \beta r(q)) (\mathcal{H}^\epsilon(\phi)).$$

Algorithm

1. Choose ϕ^0, q_1^0, q_2^0 ; $k_{\max} \in \mathbb{N}$. Set $k = 0$.
2. Compute q^k, u^k, p^k .
3. Compute $\hat{\mathcal{J}}_\phi(\phi^k, q_1^k, q_2^k)$ and set $\phi^{k+1} = \phi^k - \alpha_\phi^k \hat{\mathcal{J}}_\phi$.

Algorithm

1. Choose ϕ^0, q_1^0, q_2^0 ; $k_{\max} \in \mathbb{N}$. Set $k = 0$.
2. Compute q^k, u^k, p^k .
3. Compute $\hat{\mathcal{J}}_\phi(\phi^k, q_1^k, q_2^k)$ and set $\phi^{k+1} = \phi^k - \alpha_\phi^k \hat{\mathcal{J}}_\phi$.
4. Repeat step 2, but with ϕ^{k+1} .
5. Compute $\hat{\mathcal{J}}_{q_i}(\phi^{k+1}, q_1^k, q_2^k)$ and set $q_i^{k+1} = q_i^{k+1} - \alpha_i^k \hat{\mathcal{J}}_{q_i}$.

Algorithm

1. Choose ϕ^0, q_1^0, q_2^0 ; $k_{\max} \in \mathbb{N}$. Set $k = 0$.
2. Compute q^k, u^k, p^k .
3. Compute $\hat{\mathcal{J}}_\phi(\phi^k, q_1^k, q_2^k)$ and set $\phi^{k+1} = \phi^k - \alpha_\phi^k \hat{\mathcal{J}}_\phi$.
4. Repeat step 2, but with ϕ^{k+1} .
5. Compute $\hat{\mathcal{J}}_{q_i}(\phi^{k+1}, q_1^k, q_2^k)$ and set $q_i^{k+1} = q_i^{k+1} - \alpha_i^k \hat{\mathcal{J}}_{q_i}$.
6. Perform a distance re-initialization of ϕ^{k+1} , set $k = k + 1$.
7. Repeat 2-6 until $k = k_{\max}$.

Definition Topological Expansion

In the following we fix q_1 and q_2

$$q(q_1, q_2, \Omega_2) = q(\Omega_2) \quad (3.1)$$

and use $\beta = 0$. Therefore the cost functional becomes

$$J(u(\Omega_2), q(\Omega_2)) = J(\Omega_2).$$

Definition Topological Expansion

In the following we fix q_1 and q_2

$$q(q_1, q_2, \Omega_2) = q(\Omega_2) \quad (3.1)$$

and use $\beta = 0$. Therefore the cost functional becomes

$$J(u(\Omega_2), q(\Omega_2)) = J(\Omega_2).$$

The topological expansion $\mathcal{T}^\varepsilon$ of J at Ω is defined as

$$\mathcal{T}^\varepsilon(\hat{x}) = \frac{J(\Omega_2^\varepsilon) - J(\Omega)}{|\Omega_1^\varepsilon|}.$$

with $\Omega_1^\varepsilon = B(\hat{x}, \varepsilon)$, $\Omega_2^\varepsilon = \Omega \setminus \Omega_1^\varepsilon$. The topological derivative $\mathcal{T}$ is defined by $\mathcal{T} = \lim_{\varepsilon \rightarrow 0} \mathcal{T}^\varepsilon$, if the limit exists.

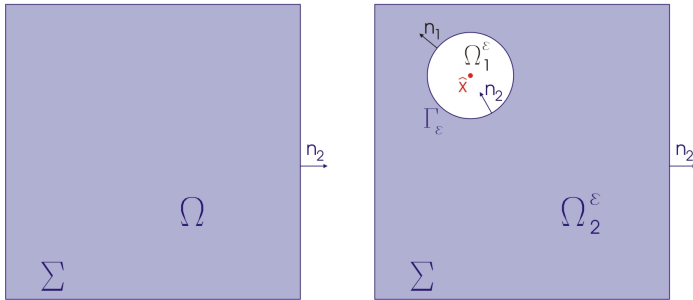


Figure: topological expansion domain

Coupled System

The solution u^ε of

$$\begin{aligned} -\operatorname{div}(q\nabla u^\varepsilon) &= 0 \text{ in } \Omega, \\ q\partial_n u^\varepsilon &= f \text{ on } \Sigma. \end{aligned}$$

can be written as

$$u^\varepsilon = u_1^\varepsilon \mathbb{1}_{\Omega_1^\varepsilon} + u_2^\varepsilon \mathbb{1}_{\Omega_2^\varepsilon},$$

with u_1^ε and u_2^ε the solutions of the coupled system

$$\begin{aligned} -\Delta u_2^\varepsilon &= 0 \text{ in } \Omega_2^\varepsilon, & -\Delta u_1^\varepsilon &= 0 \text{ in } \Omega_1^\varepsilon, \\ q_2 \partial_{n_2} u_2^\varepsilon &= f \text{ on } \Sigma, & u_1^\varepsilon &= u_2^\varepsilon \text{ in } \Gamma_\varepsilon, \\ q_2 \partial_{n_1} u_2^\varepsilon &= q_1 \partial_{n_1} u_1^\varepsilon \text{ on } \Gamma_\varepsilon, \end{aligned}$$

Expansion of u_2^ε

$$u_2^\varepsilon = u|_{\Omega_2^\varepsilon} + \mathcal{R}_2^\varepsilon$$

$$\begin{aligned} -\Delta u &= 0 \text{ in } \Omega, & -\Delta u_2^\varepsilon &= 0 \text{ in } \Omega_2^\varepsilon, \\ q_2 \partial_{n_2} u &= f \text{ on } \Sigma, & q_2 \partial_{n_2} u_2^\varepsilon &= f \text{ on } \Sigma, \\ & & q_2 \partial_{n_1} u_2^\varepsilon &= q_1 \partial_{n_1} u_1^\varepsilon \text{ on } \Gamma_\varepsilon. \end{aligned}$$

$$-\Delta \mathcal{R}_2^\varepsilon = 0 \text{ in } \Omega_2^\varepsilon, \tag{3.2}$$

$$q_2 \partial_{n_2} \mathcal{R}_2^\varepsilon = 0 \text{ on } \Sigma, \tag{3.3}$$

$$q_2 \partial_{n_1} \mathcal{R}_2^\varepsilon = q_1 \partial_{n_1} u_1^\varepsilon - q_2 \partial_{n_1} u \text{ on } \Gamma_\varepsilon. \tag{3.4}$$

Expansion of u_2^ε

$$u_2^\varepsilon = u|_{\Omega_2^\varepsilon} + \mathcal{R}_2^\varepsilon$$

$$\begin{aligned} -\Delta u &= 0 \text{ in } \Omega, & -\Delta u_2^\varepsilon &= 0 \text{ in } \Omega_2^\varepsilon, \\ q_2 \partial_{n_2} u &= f \text{ on } \Sigma. & q_2 \partial_{n_2} u_2^\varepsilon &= f \text{ on } \Sigma, \\ & & q_2 \partial_{n_1} u_2^\varepsilon &= q_1 \partial_{n_1} u_1^\varepsilon \text{ on } \Gamma_\varepsilon. \end{aligned}$$

$$-\Delta \mathcal{R}_2^\varepsilon = 0 \text{ in } \Omega_2^\varepsilon, \quad (3.2)$$

$$q_2 \partial_{n_2} \mathcal{R}_2^\varepsilon = 0 \text{ on } \Sigma, \quad (3.3)$$

$$q_2 \partial_{n_1} \mathcal{R}_2^\varepsilon = q_1 \partial_{n_1} u_1^\varepsilon - q_2 \partial_{n_1} u \text{ on } \Gamma_\varepsilon. \quad (3.4)$$

Main Idea to approximate $\mathcal{R}_2^\varepsilon$ by $\tilde{\mathcal{R}}_2^\varepsilon$

- compute solution of (3.2) and (3.4)
- compensate error on Σ by considering (3.2) and (3.3)

Expansion of $J(\Omega_2^\varepsilon)$

$$u_2^\varepsilon(x) = u|_{\Omega_2^\varepsilon} + \mathcal{R}_2^\varepsilon = u|_{\Omega_2^\varepsilon} + \tilde{\mathcal{R}}_2^\varepsilon + \text{Rest} \quad (3.5)$$

$$\begin{aligned} J(\Omega_2^\varepsilon) &= \frac{1}{2} \int_{\Sigma} (u_2^\varepsilon - m)^2 \stackrel{(3.5)}{=} \frac{1}{2} \int_{\Sigma} ((u - m) + (\tilde{\mathcal{R}}_2^\varepsilon + \text{Rest}))^2 \\ &= \frac{1}{2} \int_{\Sigma} (u - m)^2 \\ &\quad + \frac{1}{2} \int_{\Sigma} 2(u - m) (\tilde{\mathcal{R}}_2^\varepsilon + \text{Rest}) + (\tilde{\mathcal{R}}_2^\varepsilon + \text{Rest})^2 \\ &\approx J(\Omega) + |\Omega_1^\varepsilon| (\mathcal{I}_0 + \mathcal{I}_1^\varepsilon + \mathcal{I}_2^\varepsilon + \mathcal{I}_3^\varepsilon + \mathcal{I}_4^\varepsilon) \end{aligned}$$

$$\mathcal{T}^\varepsilon = \frac{J(\Omega_2^\varepsilon) - J(\Omega)}{|\Omega_1^\varepsilon|} \approx \mathcal{T}_0 + \mathcal{T}_1^\varepsilon + \mathcal{T}_2^\varepsilon + \mathcal{T}_3^\varepsilon + \mathcal{T}_4^\varepsilon$$

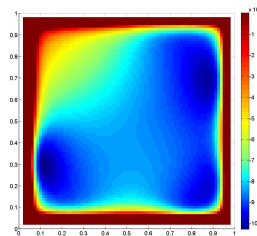
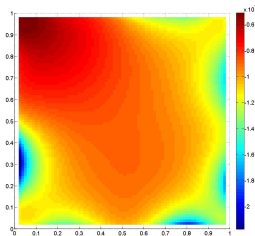


Figure: top view of $\mathcal{T}$ and $\mathcal{T}^\varepsilon$

Discretization scheme

Necessary conditions for a distance function

$$|\nabla\phi(x)| = 1 \quad x \in \Omega, \quad (1.1)$$

$$\phi(x) = 0 \quad x \in \Gamma \quad (1.2)$$

Discretization scheme

Necessary conditions for a distance function

$$|\nabla \phi(x)| = 1 \quad x \in \Omega, \quad (1.1)$$

$$\phi(x) = 0 \quad x \in \Gamma \quad (1.2)$$

Discretized equation of (1.1)

$$\max(\phi_{i,j} - \phi_{i_{\min},j}, 0)^2 + \max(\phi_{i,j} - \phi_{i,j_{\min}}, 0)^2 = h^2 \quad (1.3)$$

$$\phi_{i_{\min},j} = \min(\phi_{i-1,j}, \phi_{i+1,j}) \quad \text{and} \quad \phi_{i,j_{\min}} = \min(\phi_{i,j-1}, \phi_{i,j+1})$$

Fast sweeping algorithm

► Initialization

$$\phi_{i,j} = \begin{cases} 0 & \text{if } x_{i,j} \in \Gamma \\ c & \text{otherwise} \end{cases}$$

with c being an upper boundary of the true solution

Fast sweeping algorithm

► Initialization

$$\phi_{i,j} = \begin{cases} 0 & \text{if } x_{i,j} \in \Gamma \\ c & \text{otherwise} \end{cases}$$

with c being an upper boundary of the true solution

► Update

- Compute solution $\bar{\phi}$

$$\max(\bar{\phi} - \phi_{i_{\min},j}, 0)^2 + \max(\bar{\phi} - \phi_{i,j_{\min}}, 0)^2 = h^2$$

- Set

$$\phi_{i,j} = \min(\phi_{i,j}, \bar{\phi}).$$

We sweep the whole domain with four alternating orderings:

$$(1) \ i = 1 : I, \ j = 1 : J,$$

$$(2) \ i = I : 1, \ j = 1 : J,$$

$$(3) \ i = I : 1, \ j = J : 1,$$

$$(4) \ i = 1 : I, \ j = J : 1.$$

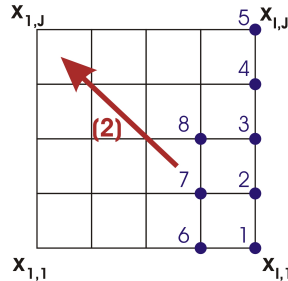
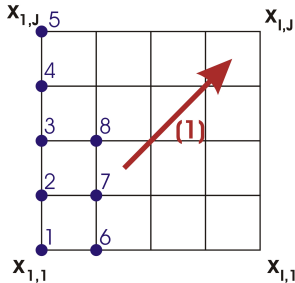


Illustration of 4 sweeps

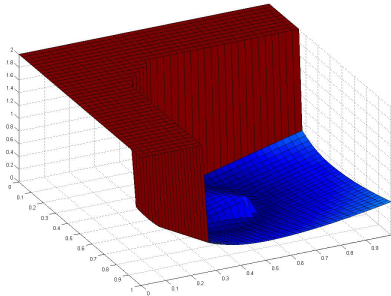


Figure: after 1 sweep

Illustration of 4 sweeps

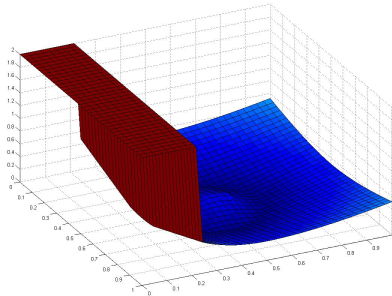


Figure: after 2 sweeps

Illustration of 4 sweeps

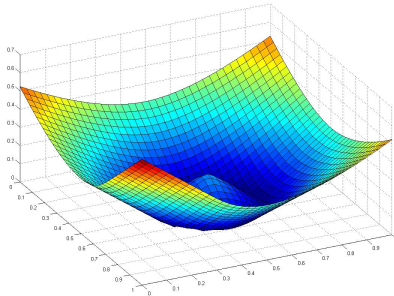


Figure: after 3 sweeps

Illustration of 4 sweeps

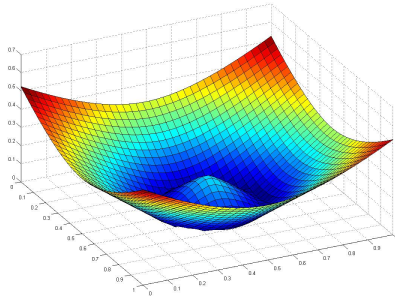


Figure: after 4 sweeps

Basic theoretical results

Theorem

The iterative solution by the fast sweeping algorithm converges monotonically to the solution of the discretized system.

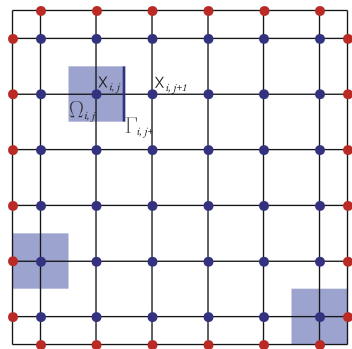
Theorem

Let $\Gamma = \{\mathbf{x}_0\}$ and $\phi_{i,j}$ be the solution of the discretized system. The fast sweeping algorithm converges after 4 sweeps to the solution $\phi_{i,j}$ which satisfies

$$d(\mathbf{x}_{i,j}) \leq \phi_{i,j} \leq d(\mathbf{x}_{i,j}) + O(|h \log h|).$$

Elliptic equation with Neumann boundary constraints

$$\begin{aligned} -\operatorname{div}(q\nabla u) &= 0 \text{ in } \Omega, \\ q\partial_n u &= f \text{ on } \Sigma, \\ \int_{\Sigma} f(x) &= 0. \end{aligned}$$

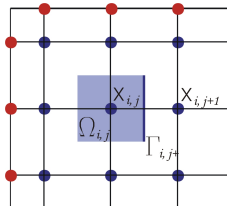


$$\Omega_{i,j} = \left(x_{i-\frac{1}{2},j}, x_{i+\frac{1}{2},j}\right) \times \left(x_{i,j-\frac{1}{2}}, x_{i,j+\frac{1}{2}}\right)$$

Discretization scheme

Gauss' Theorem

$$0 = \int_{\Omega_{i,j}} \operatorname{div}(q \nabla u) = \int_{\Gamma_{i+,j} \cup \Gamma_{i-,j} \cup \Gamma_{i,j+} \cup \Gamma_{i,j-}} q \partial_n u. \quad (2.1)$$



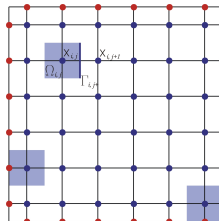
Numerical approximation:

$$\int_{\Gamma_{i,j+}} q \partial_n u \approx (u_{i,j+1} - u_{i,j}) q_{i,j+}$$

Resulting equations

$$u_{i-1,j} q_{i-,j} + u_{i+1,j} q_{i+,j} + u_{i,j-1} q_{i,j-} + u_{i,j+1} q_{i,j+} - u_{i,j} (q_{i-,j} + q_{i+,j} + q_{i,j-} + q_{i,j+}) = 0.$$

$$u_{i,3} q_{i,3-} + u_{i-1,2} q_{i-,2} + u_{i+1,2} q_{i+,2} - u_{i,2} (q_{i,3-} + q_{i-,2} + q_{i+,2}) = - \int_{x_i - \frac{h}{2}}^{x_i + \frac{h}{2}} f(x, 0) dx$$



Structure of the Matrix

$$L_h \cdot u_h = d_h. \quad (2.2)$$

$$L_h(1 : 8, 1 : 8) =$$

$$\begin{pmatrix} * & q_{2+,2} & & & & & & \\ q_{3-,2} & * & q_{3+,2} & & & & & \\ & q_{4-,2} & * & q_{4+,2} & & & & \\ & & q_{5-,2} & * & & & & \\ \hline q_{2,3-} & & & & & & & \\ & q_{3,3-} & & & & & & \\ & & q_{4,3-} & & & & & \\ & & & q_{5,3-} & & & & \end{pmatrix} \begin{pmatrix} q_{2,2+} & & & & & & & \\ & q_{3,2+} & & & & & & \\ & & q_{4,2+} & & & & & \\ & & & q_{5,2+} & & & & \\ \hline & * & q_{2+,3} & & & & & \\ q_{3-,3} & & * & q_{3+,3} & & & & \\ & q_{4-,3} & & * & q_{4+,3} & & & \\ & & q_{5-,3} & & * & & & \end{pmatrix}$$

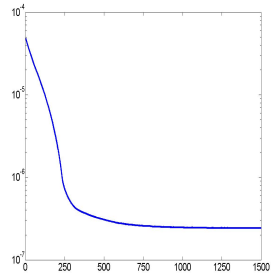
Basic theoretical result

Theorem

The Matrix $\begin{pmatrix} L_h & \mathbb{1} \\ \mathbb{1}^T & 0 \end{pmatrix}$ is nonsingular.

Function decrease

(M8B18N72.avi)



Benefit of the topological expansion

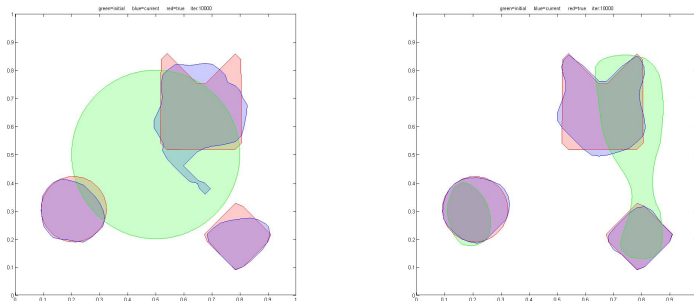


Figure: $M=24$, $\beta = 1e-8$

Dependence on regularization

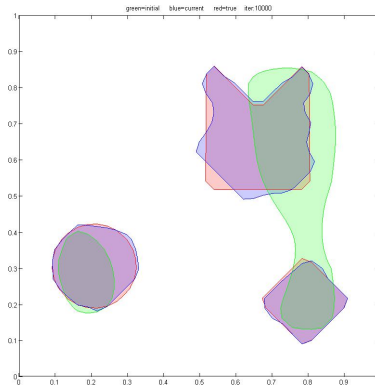


Figure: $M=24$, $\beta = 1e-9$

Dependence on regularization

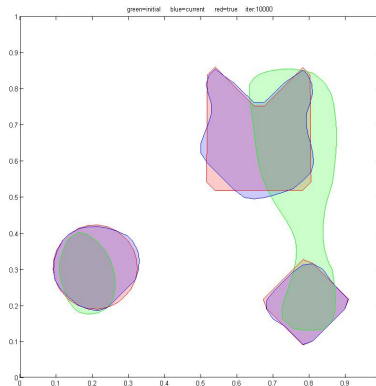


Figure: $M=24$, $\beta = 1e-8$

Dependence on regularization

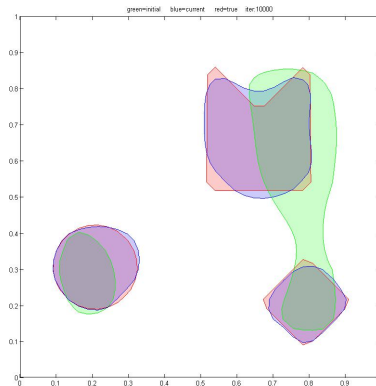


Figure: $M=24$, $\beta = 1e-7$

Dependence on regularization

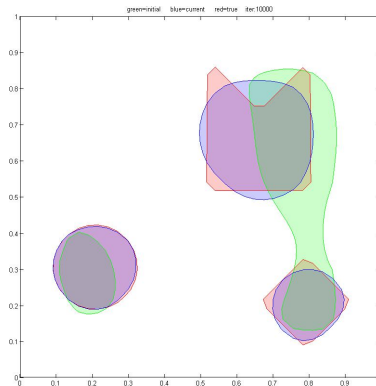
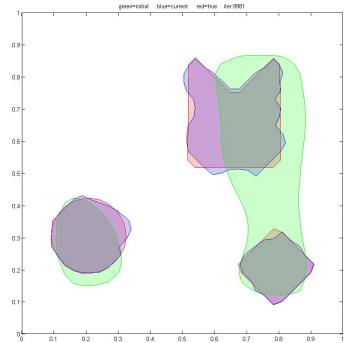
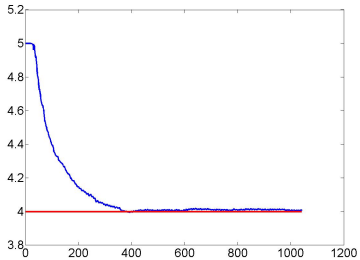
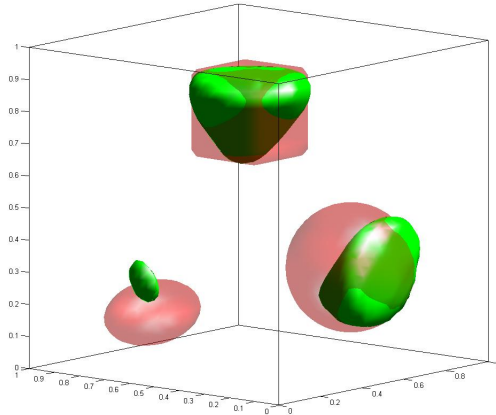


Figure: $M=24$, $\beta = 1e-6$

Different q_1



Topological expansion



Gradient method

