# Reconstruction de volumes à partir de coupes 

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## Motivation

## (8)





Frequent problem in medical imaging (MRI, CT) and computational geometry: how to reconstruct a volume from a few slices (or more generally from partial data)?

## Motivation





## Formulation with inner / outer constraints




Formulation with inner / outer constraints


What is a "good shape" satisfying the constraints?

Geometric optimization in real life


## Modeling

Let $\omega^{i n t}, \omega^{e x t} \subset \mathbb{R}^{N}$

Geometric optimization problem

$$
\inf \left\{J(E) \mid \omega^{i n t} \subset E \subset \mathbb{R}^{N} \backslash \omega^{e x t}\right\}
$$

where $J$ is a geometric energy

- Natural choice: J=perimeter or Willmore energy
- A natural topology is the $L^{1}$ topology of characteristic functions of sets
- The problem is however ill-posed (at least for the perimeter) when $\left|\omega^{i n t}\right|=\left|\omega^{e x t}\right|=0$.


## Perimeter in the BV sense

## Perimeter

$E$ has finite perimeter if its characteristic function $\mathbb{1}_{E} \in B V$ Denote $P(E)=T V\left(\mathbb{1}_{E}\right)$ its perimeter.

The perimeter functional is lower semicontinuous for the $L^{1}$ topology.


$$
P(E)=\operatorname{length}(\partial E)
$$



$$
P(E)=\operatorname{area}(\partial E)
$$

## Perimeter in the BV sense

## Perimeter

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The perimeter functional is lower semicontinuous for the $L^{1}$ topology.


$$
P(E) \neq \operatorname{length}(\partial E)
$$



$$
P(E) \neq \operatorname{length}(\partial E)
$$

Natural formulation of the reconstruction problem for the perimeter

$$
\inf \left\{P(E) \mid \omega^{i n t} \subset E^{1}, \omega^{e x t} \subset E^{0}\right\}
$$



## Bernoulli-Euler elastic energy in $\mathbb{R}^{2}$



## Curvature

Let $\gamma$ be a $C^{2}$ curve in $\mathbb{R}^{2}$,

$$
\kappa=\frac{\operatorname{det}\left(\gamma^{\prime \prime}, \gamma^{\prime}\right)}{\left|\gamma^{\prime}\right|^{3}}
$$

## Bernoulli-Euler energy

Let $E$ be a set with $C^{2}$ boundary,

$$
W(E)=\int_{\partial E} \kappa^{2} \mathrm{~d} \mathcal{H}^{1}
$$

## Willmore energy (in $\mathbb{R}^{3}$ )

## Mean curvature

Let $M=C^{2}$ surface in $\mathbb{R}^{3}$,

$$
H=\frac{1}{2}\left(\kappa_{1}+\kappa_{2}\right)
$$

$\kappa_{1}, \kappa_{2}$ : principal curvatures

## Willmore energy

If $E$ has $C^{2}$ boundary $\partial E$,

$$
W(E)=\int_{\partial E} H^{2} \mathrm{~d} \mathcal{H}^{2} .
$$

Natural formulation of the reconstruction problem for the Willmore energy

The Willmore energy is not lower semicontinuous in $L^{1}$.
For minimization purposes, use its relaxation $\bar{W}$ (i.e. its lower semicontinuous envelope).

We address the following problem:

$$
\inf \left\{\bar{W}(E) \mid \omega^{i n t} \subset E^{1}, \omega^{\text {ext }} \subset E^{0}\right\}
$$

## Approximation of the problem

I: Perimeter approximation


Thus, $\int \varepsilon\left|\nabla u_{\varepsilon}\right|^{2} \mathrm{~d} x \approx \frac{1}{\varepsilon}$ Area $\approx \frac{1}{\varepsilon} \varepsilon P(E)=P(E) \quad$ as $\varepsilon \rightarrow 0$.
However, any constant function has zero energy! How to force $u_{\varepsilon}$ to be close to a characteristic function, i.e. a binary function?

## Perimeter approximation

Use a double-well potential, for instance $G(s)=\frac{1}{2} s^{2}(1-s)^{2}$.


If $\sup _{\varepsilon}\left(\int \frac{1}{\varepsilon} G\left(u_{\varepsilon}\right) \mathrm{d} x\right)<+\infty$ then $u_{\varepsilon} \rightarrow 0$ or 1 a.e. as $\varepsilon \rightarrow 0$.
Therefore, $u_{\varepsilon}$ approximates a characteristic function.

## The Cahn-Hilliard functional

## (Van der Waals)-Cahn-Hilliard energy

The phase-field approximation of perimeter is given by

$$
P_{\varepsilon}(u)=\int\left(\frac{\varepsilon}{2}|\nabla u|^{2}+\frac{1}{\varepsilon} G(u)\right) \mathrm{d} x
$$



where $G$ is a double-well potential.


$$
\begin{gathered}
\text { e.g., } \\
G(s)=\frac{1}{2} s^{2}(1-s)^{2}
\end{gathered}
$$

## Phase-field approximation of perimeter

Convergence of $P_{\varepsilon}$ (Modica, Mortola - 1977)
$P_{\varepsilon}$ converges to

$$
P(u)= \begin{cases}\lambda P(E) & \text { si } u=\mathbb{1}_{E} \in B V \\ +\infty & \text { otherwise }\end{cases}
$$

in the sense of $\Gamma$-convergence
where $\lambda$ is a constant depending only on potential $G$.

## Property of $\Gamma$-convergence

Let $X$ be a metric space and $\left(F_{\varepsilon}\right)$ a sequence of equicoercive functionals converging to $F$ in the sense of $\Gamma$-convergence in $X$. If $u_{\varepsilon}$ is a minimizer of $F_{\varepsilon}$, then there exists a minimizer $u$ of $F$, s.t. $u_{\varepsilon} \rightarrow u$.

## Optimal profile

One can define the phase-field optimal profile associated with $E$ :

$$
u_{\varepsilon}(x)=q\left(\frac{1}{\varepsilon} d_{s}(x, E)\right) \quad \text { with } \quad q(s)=\frac{1}{2}\left(1-\tanh \left(\frac{s}{2}\right)\right)
$$



## Signed distance <br> $$
d_{s}(x, E)=d(x, E)-d\left(x, \mathbb{R}^{N} \backslash E\right)
$$

## Convergences

For a bounded set $E$

- $u_{\varepsilon} \rightarrow \mathbb{1}_{E}$
- $P_{\varepsilon}\left(u_{\varepsilon}\right) \rightarrow \lambda P(E)$ if $E$ has finite perimeter
as $\varepsilon \rightarrow 0$.

Phase field approximation of the Willmore energy The $L^{2}$-gradient of $P_{\varepsilon}$ satisfies

$$
-\nabla_{L^{2}} P_{\varepsilon}(u)=\varepsilon \Delta u-\frac{1}{\varepsilon} G^{\prime}(u)
$$

The gradient flow of perimeter is the mean curvature flow and $-\nabla_{L^{2}} P_{\varepsilon}\left(u_{\varepsilon}\right)$ approximates the mean curvature of $\partial E$ in the transition zone of $u_{\varepsilon}$ when $u_{\varepsilon} \approx \mathbb{1}_{E}$.

Approximation of the Willmore energy
In $\mathbb{R}^{2}$ and $\mathbb{R}^{3}$, the energy

$$
u \mapsto P_{\varepsilon}(u)+W_{\varepsilon}(u)=P_{\varepsilon}(u)+\int \frac{1}{2 \varepsilon}\left(\varepsilon \Delta u-\frac{1}{\varepsilon} G^{\prime}(u)\right)^{2} \mathrm{~d} x
$$

「-converges to $E \mapsto \lambda(P(E)+W(E))$ if $E$ is $C^{2}$ and compact

- De Giorgi + Bellettini, Paolini (1993) + Röger, Schätzle (2006)


## Optimal profile

With the same phase-field profile associated with $E$

$$
u_{\varepsilon}(x)=q\left(\frac{1}{\varepsilon} d_{s}(x, E)\right)
$$

one has

## Convergences

For a bounded set $E$

- $u_{\varepsilon} \rightarrow \mathbb{1}_{E}$
- $P_{\varepsilon}\left(u_{\varepsilon}\right) \rightarrow \lambda P(E)$ if $E$ has finite perimeter
- $W_{\varepsilon}\left(u_{\varepsilon}\right) \rightarrow \lambda W(E)$ if $\partial E$ is $C^{2}$
as $\varepsilon \rightarrow 0$.


## Inclusion-exclusion constraints

Let $\omega^{i n t}, \omega^{e x t} \subset \mathbb{R}^{N}$
Geometric optimization problem

$$
\inf \left\{J(E) \mid \omega^{\text {int }} \subset E^{1}, \omega^{e x t} \subset E^{0}\right\}
$$

where $J$ is either $P$, or $W$
One defines obstacle constraints:

$$
u_{\varepsilon}^{i n t}(x)=q\left(\frac{1}{\varepsilon} d_{s}\left(x, \omega^{i n t}\right)\right) \quad \text { and } \quad u_{\varepsilon}^{e x t}(x)=1-q\left(\frac{1}{\varepsilon} d_{s}\left(x, \omega^{e x t}\right)\right)
$$

Key property

$$
\omega^{\text {int }} \subset E \subset \mathbb{R}^{N} \backslash \omega^{\text {ext }} \quad \Longleftrightarrow \quad u_{\varepsilon}^{\text {int }} \leqslant u_{\varepsilon} \leqslant u_{\varepsilon}^{\text {ext }}
$$

In the phase field approximation, constraints can be interpreted as a linear obstacle problem!

## Numerical scheme for perimeter

## Approximating a solution to

$$
\min \left\{P_{\varepsilon}(u) \mid u_{\varepsilon}^{\text {int }} \leqslant u \leqslant u_{\varepsilon}^{e x t}\right\}
$$

- Initialize $u^{0}$;
- At step $n$, given $u^{n}$, use a splitting method:
- $u^{n+1 / 2}$ is obtained by one step of an implicit discrete $L^{2}$ gradient flow for $P_{\varepsilon}$, i.e.
$u^{n+1 / 2}-u^{n}=\delta_{t}\left(\varepsilon \Delta u^{n+1 / 2}-\frac{1}{\varepsilon} G^{\prime}\left(u^{n+1 / 2}\right) \quad\right.$ (discrete Allen-Cahn equation)
- Get $u^{n+1}$ from $u^{n+1 / 2}$ by projecting onto the constraints

$$
u_{\varepsilon}^{i n t} \leqslant u \leqslant u_{\varepsilon}^{e x t}
$$

## Implicit discrete gradient flow

Finding $u^{n+1 / 2}$ is equivalent to finding a fixed point of the map:

$$
v \mapsto\left(I_{d}-\delta_{t} \varepsilon \Delta\right)^{-1}\left[\left(u^{n}+\frac{\delta_{t}}{\varepsilon} G^{\prime}(v)\right)\right]
$$

Picard iterations give a stable scheme, and solving in Fourier domain provides an excellent spatial accuracy

## Matlab code (projection is embedded into the fixed point scheme)

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```
```

```
%%%%%%%%%%%%%%%%% Parameters %%%%%%%%%%%%%%%%%%%
```

```
%%%%%%%%%%%%%%%%% Parameters %%%%%%%%%%%%%%%%%%%
    epsiton = 2/N;
    epsiton = 2/N;
    T =1;
    T =1;
delta_t = 1/N^2;
delta_t = 1/N^2;
%%%%%%%%%%%%%%%%%% Heat Kerne1 %%%%%%%%%%%%%%%%%%%%%%
%%%%%%%%%%%%%%%%%% Heat Kerne1 %%%%%%%%%%%%%%%%%%%%%%
    K1 = ones(N,1)*[0:N/2,-N/2+1:-1];
    K1 = ones(N,1)*[0:N/2,-N/2+1:-1];
        M = 1./(1+4* pi^2* de7ta_t* (K1.^2 + K1'.^2));
        M = 1./(1+4* pi^2* de7ta_t* (K1.^2 + K1'.^2));
    %%%%%%%%%%%%%%%% Minimization scheme %%%%%%%%%%%%%%
    %%%%%%%%%%%%%%%% Minimization scheme %%%%%%%%%%%%%%
    for n=1:T/delta_t,
    for n=1:T/delta_t,
    U = U1_0;
    U = U1_0;
    U1_0_fourier = fft2(U1_0);
    U1_0_fourier = fft2(U1_0);
    res = 1;
    res = 1;
    96%%%%%%%%% fixed point iteration %9%%%%%%%%%%%%%%
    96%%%%%%%%% fixed point iteration %9%%%%%%%%%%%%%%
    while res > 10^(-4),
    while res > 10^(-4),
    U_plus = ifft2(M.*(U1_0_fourier - delta_t/epsilon^2*fft2(U.*(U-1).*(2*U-1))));
    U_plus = ifft2(M.*(U1_0_fourier - delta_t/epsilon^2*fft2(U.*(U-1).*(2*U-1))));
    U_plus = max(min(1-U2,U_plus),U1);
    U_plus = max(min(1-U2,U_plus),U1);
    res = norm((U_plus-U));
    res = norm((U_plus-U));
    U = U_plus;
    U = U_plus;
    end
    end
    U1_0 = U;
    U1_0 = U;
```

end

```
```

end

```

\section*{Numerical scheme for Willmore}

Same principle, but now the \(L^{2}\) flow is:
\[
\left\{\begin{array}{l}
\partial_{t} v=\Delta \mu-\frac{1}{\varepsilon^{2}} G^{\prime \prime}(v) \mu \\
\mu=\frac{1}{\varepsilon} G^{\prime}(v)-\varepsilon \Delta v
\end{array}\right.
\]

It can be discretized at step \(n\) as
\[
\left\{\begin{array}{l}
u^{n+1 / 2}=u^{n}+\delta_{t}\left[\Delta \mu^{n+1 / 2}-\frac{1}{\varepsilon^{2}} G^{\prime \prime}\left(u^{n+1 / 2}\right) \mu^{n+1 / 2}\right] \\
\mu^{n+1 / 2}=\frac{1}{\varepsilon} G^{\prime}\left(u^{n+1 / 2}\right)-\varepsilon \Delta u^{n+1 / 2} .
\end{array}\right.
\]
whose solution \(\left(u^{n+1 / 2}, \mu^{n+1 / 2}\right)\) is a fixed point of the map:
\[
v \mapsto\left(\begin{array}{cc}
I_{d} & -\delta_{t} \Delta \\
\varepsilon \Delta & I_{d}
\end{array}\right)^{-1}\binom{u^{n}-\frac{\delta_{t}}{\varepsilon} G^{\prime \prime}(u) \mu}{\frac{1}{\varepsilon^{2}} G^{\prime}(u)}
\]

Again, an efficient and accurate scheme can be designed using Fourier transform.

\section*{First experiments}


Perimeter


Willmore energy

\section*{Interpretation}

In some cases, the energy
\[
P_{1, \varepsilon}\left(u_{\varepsilon}\right)= \begin{cases}P_{\varepsilon}(u) & \text { if } u_{\varepsilon}^{i n t} \leqslant u_{\varepsilon} \leqslant u_{\varepsilon}^{e x t} \\ +\infty & \text { otherwise }\end{cases}
\]
converges to
\[
F_{1}(u)=\lambda(P(E)+\mathcal{H}(E))
\]
if \(u=\mathbb{1}_{E}\), and \(u_{\varepsilon} \rightarrow u\) as \(\varepsilon \rightarrow 0\)

- \(\mathcal{H}(E)=\) length (in 2 D ) or area (in 3D) of the set \(\left(E^{0} \cap \omega^{\text {int }}\right) \bigcup\left(E^{1} \cap \omega^{\text {out }}\right)\)

However, the term \(\lambda \mathcal{H}(E)\) may favor constraints violation

\section*{Constraints violation}

A situation where violating the outer constraint is more favorable for \(F_{1}\) :

\(F_{1}\) (left configuration) \(<F_{1}\) (right configuration)
In contrast, defining \(F_{2}(E)=\lambda(P(E)+2 \mathcal{H}(E))\)
one has
\(F_{2}(\) left configuration \()>F_{2}(\) right configuration \()\)

\section*{Remedy: use fat constraints}

Thicken the constraints to give them volume


The function \(U_{\varepsilon}^{\text {int }}(x)=q\left(\frac{1}{\varepsilon} d_{s}\left(x, \Omega_{\varepsilon}^{\text {int }}\right)\right)\) takes values in \([0,1]\).


\section*{Convergence result}

Theorem (Bretin, Dayrens, M.)
The energy
\[
P_{2, \varepsilon}(u)= \begin{cases}P_{\varepsilon}(u) & \text { si } U_{\varepsilon}^{\text {int }} \leqslant u \leqslant U_{\varepsilon}^{\text {ext }} \\ +\infty & \text { sinon }\end{cases}
\]

「-converges to
\[
F_{2}(u)=\lambda(P(E)+2 \mathcal{H}(E))
\]
if \(u=\mathbb{1}_{E}\), as \(\varepsilon \rightarrow 0\)
- Minimal sets for \(F_{2}\) satisfy inclusion-exclusion constraints in reasonable cases.
- Characterizing the \(\Gamma\)-limit for the Willmore energy is delicate due to the non locality and ghost parts.

\section*{Numerical experiments}

"Thin" perimeter

"Fat" perimeter

\section*{3D reconstruction with Willmore energy}


Reconstruction of a 3D brain image from real MRI slices

\section*{Confined elastica (i.e. a minimizer of the constrained Bernoulli-Euler energy)}


An elastica in a fox head

\section*{Can be used for smoothing pixellized surfaces}

cf Bretin, Lachaud, Oudet, 2011 where was used a penalization of the constraints violation set volume (acting as a repulsion force)

\section*{Other "slices"}



\section*{Alternative partial data}



\section*{Joint reconstruction of several domains}

The method is applied jointly to several phases \(u_{1}, u_{2}, \ldots u_{n}\) in two cases:
- Disjoint phases: prescribe \(\sum_{i} u_{i} \leq 1\)
- Nested phases: prescribe \(u_{1} \leq u_{2} \cdots \leq u_{n}\)

\section*{Joint reconstruction of several domains}



Two disjoint domains

\section*{Joint reconstruction of several domains}



Nested domains

\section*{Joint reconstruction of several domains}


Two disjoint domains (two phases segmentation of MRI data)

\section*{Joint reconstruction of several domains}


Several disjoint domains

\section*{Conclusion}
- Our model has no topological prior;
- Can be adapted to many situations;
- But limited to volume reconstruction; what about surfaces with boundary?
- Stable, fast, accurate numerical schemes can be designed;
- Extension to the anisotropic case is possible;
- Theoretical characterization of the constrained relaxed Willmore energy is an open problem.```

