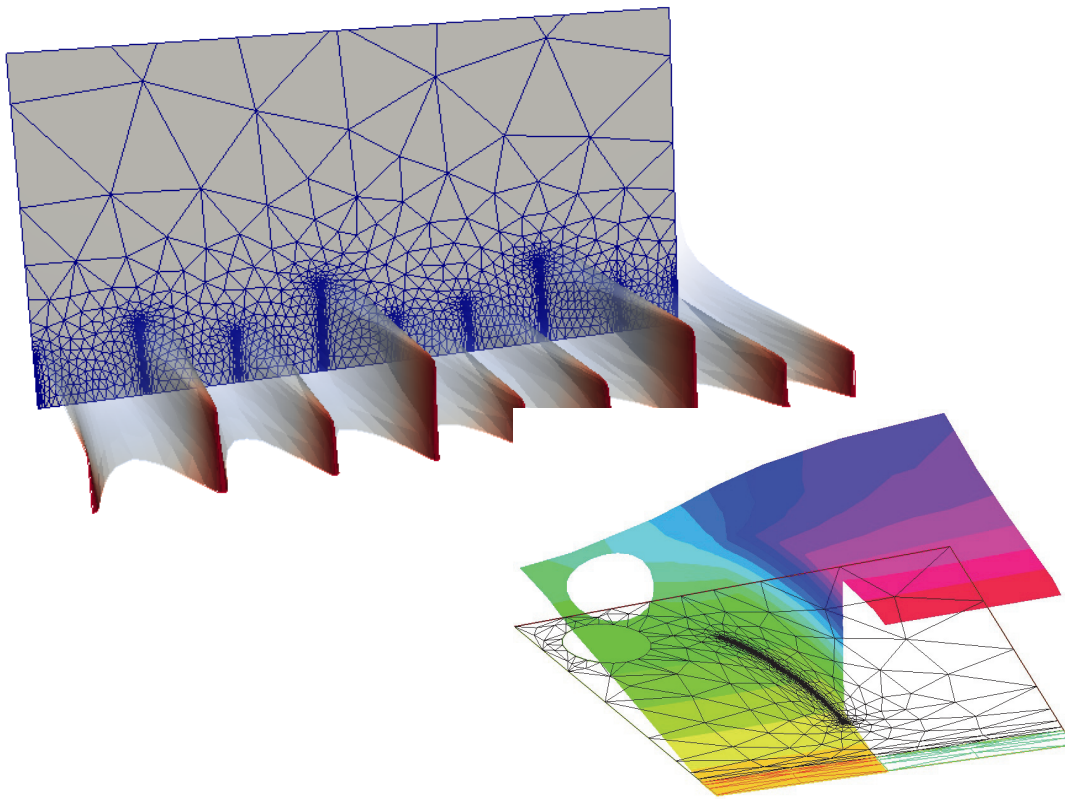
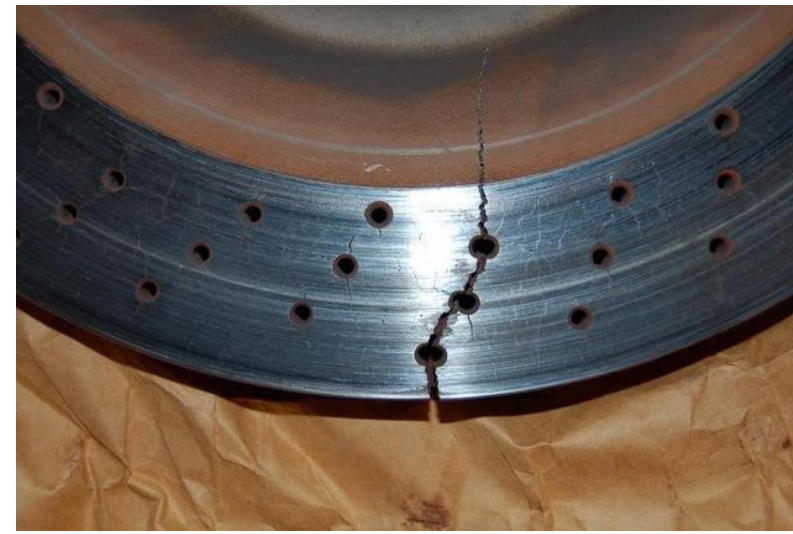
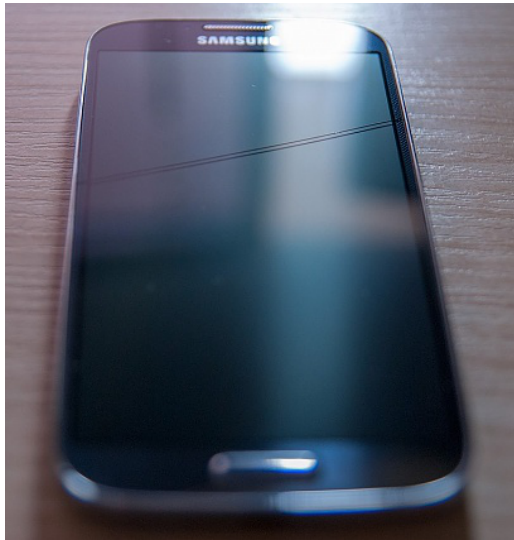


Energy functional minimization combined with an anisotropic mesh adaptation

Simona Perotto

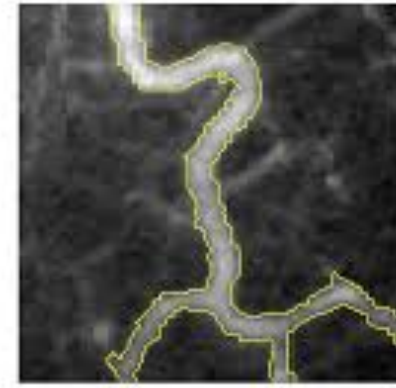
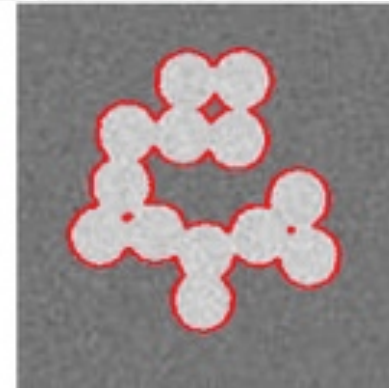
MOX-Modeling and Scientific Computing
Department of Mathematics, Politecnico di Milano

crack propagation in brittle materials



cracks $\cap$ images = energy functional minimization

image segmentation



Crack propagation in brittle materials

Ambrosio-Tortorelli approximation to Francfort-Marigo model:

$$I_\varepsilon : H^1(\Omega) \times H^1(\Omega; [0, 1]) \rightarrow \mathbb{R}$$

$$I_\varepsilon(u, v) = \underbrace{\int_{\Omega} (v^2 + \eta) |\nabla u|^2 d\mathbf{x}}_{\text{elastic energy}} + \underbrace{\kappa \int_{\Omega} \left[\frac{1}{4\varepsilon} (1 - v)^2 + \varepsilon |\nabla v|^2 \right] d\mathbf{x}}_{\text{fictitious crack energy}}$$

★ $0 < \eta \ll \varepsilon \ll 1$

★ $\kappa > 0$

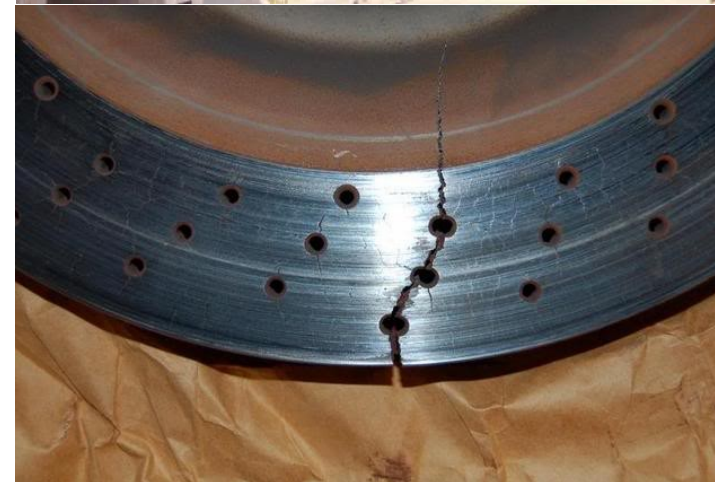
★ $v \in H^1(\Omega; [0, 1])$ (phase field) 1 : sound material
0 : crack

★ $u \in \mathcal{A}(g(t))$ (body displacement)

$$\mathcal{A}(g(t)) = \{u \in H^1(\Omega) : u|_{\Omega_{D^\pm}} = g(t)|_{\Omega_{D^\pm}}\}$$

admissible displacements

$$g : [0, T] \times \Omega \rightarrow \mathbb{R} \quad : \quad g(t) = \begin{cases} t & \text{on } \Omega_{D^+}, \\ -t & \text{on } \Omega_{D^-}, \\ 0 & \text{elsewhere,} \end{cases}$$



Crack propagation in brittle materials

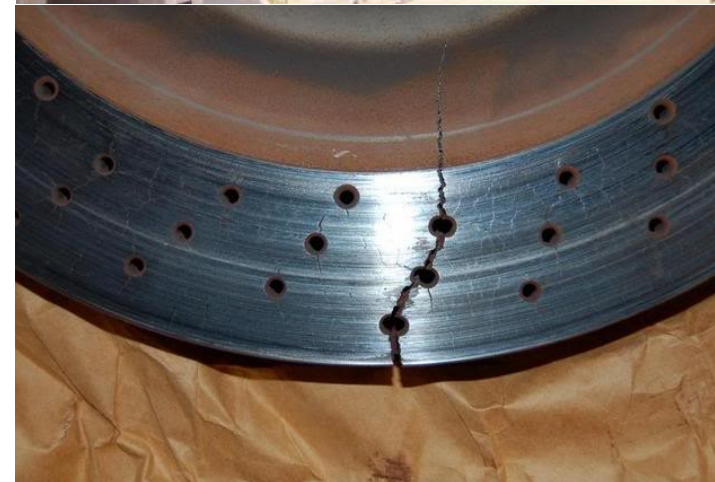
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Theoretical investigation :

- ☀ Γ -convergence to Francfort-Marigo functional;
[B. Bourdin, G.A. Francfort, J.-J. Marigo (2000)]
- ☀ existence of minimizers for $I_\varepsilon(u, v)$ ($\varepsilon, \eta > 0$);
[L. Ambrosio, V.M. Tortorelli (1992)]



Crack propagation in brittle materials

Ambrosio-Tortorelli approximation to Francfort-Marigo model:

$$I_\varepsilon : H^1(\Omega) \times H^1(\Omega; [0, 1]) \rightarrow \mathbb{R}$$

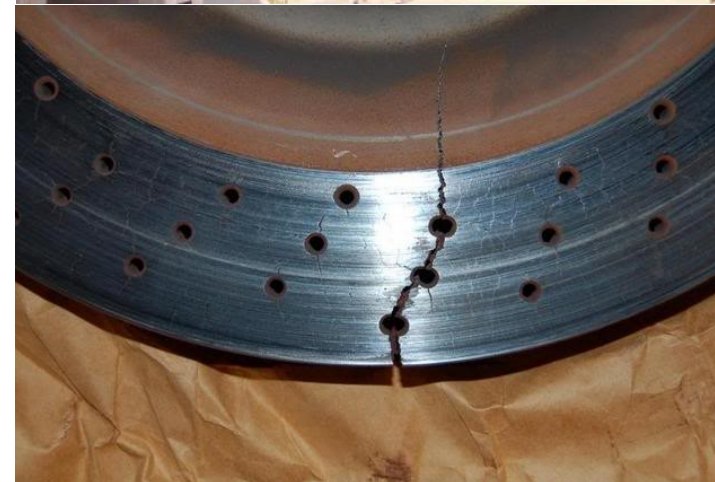
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Crack evolution :

- ☀ time discretization : $0 = t_0 < t_1 < \dots < t_F = T$
- ☀ minimization at each time step: $t = t_0$

$$(u_\varepsilon(t_0), v_\varepsilon(t_0)) \in \arg \min_{\substack{u \in \mathcal{A}(g(t_0)), \\ v \in H^1(\Omega; [0, 1])}} I_\varepsilon(u, v)$$

*quasi-static
approach*



Crack propagation in brittle materials

Ambrosio-Tortorelli approximation to Francfort-Marigo model:

$$I_\varepsilon : H^1(\Omega) \times H^1(\Omega; [0, 1]) \rightarrow \mathbb{R}$$

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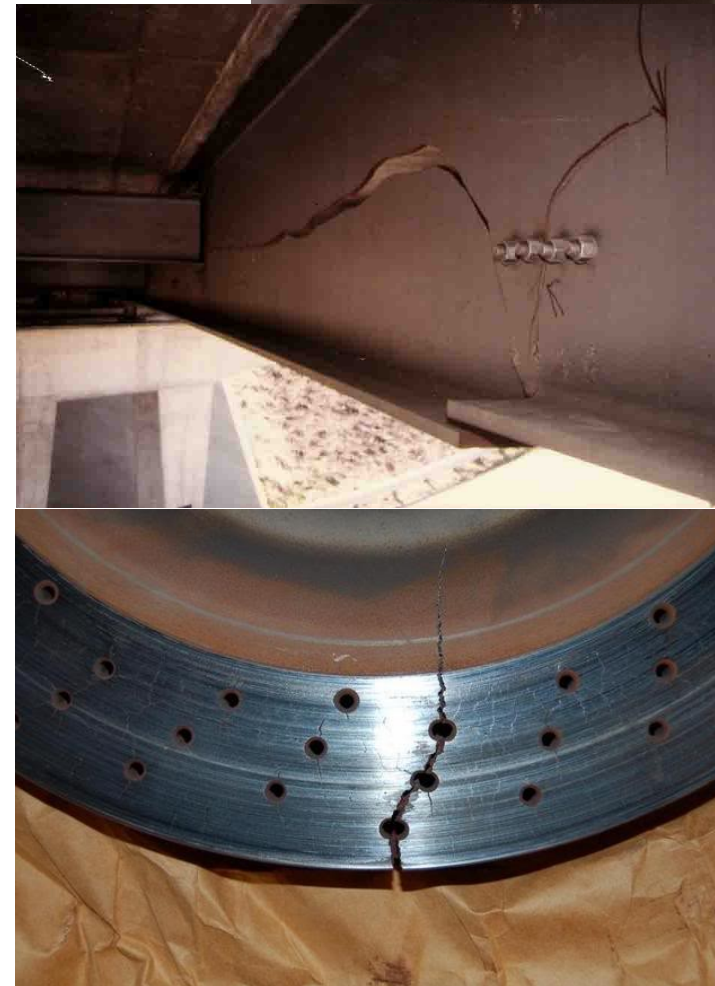
Crack evolution :

- ★ time discretization : $0 = t_0 < t_1 < \dots < t_F = T$
- ★ minimization at each time step: $t = t_k$

$$(u_\varepsilon(t_k), v_\varepsilon(t_k)) \in \arg \min_{\substack{u \in \mathcal{A}(g(t_k)), \\ v \in H^1(\Omega; [0, 1]), v \leq v_\varepsilon(t_{k-1})}} I_\varepsilon(u, v)$$

quasi-static
approach

irreversibility of the crack
[A. Giacomini (2005)]



The modified Ambrosio-Tortorelli model

[M. Artina, M. Fornasier, S. Micheletti, S.P. (2015a)]

Idea : relax the constraints on the displacement and on the phase field via penalization:

$$u, v \in H^1(\Omega)$$

[B. Bourdin (2007)]

★ $v \leq v_\varepsilon(t_{k-1})$  **equality** constraint. If at $t = t_{k-1}$

$$CR_{k-1} = \{\mathbf{x} \in \bar{\Omega} \mid v_\varepsilon(t_{k-1}) < \text{CRTOL}\} \neq \emptyset$$

the **irreversibility** is enforced by

$$v_\varepsilon(\mathbf{x}, t_i) = 0 \quad \forall \mathbf{x} \in CR_{k-1} \quad \text{and} \quad \forall i : k \leq i \leq F.$$



$$(u_\varepsilon(t_k), v_\varepsilon(t_k)) \in \arg \min_{\substack{u \in H^1(\Omega), \\ v \in H^1(\Omega; [0, 1])}} I_{\varepsilon, k}^{\text{penalty}}(u, v) \quad k = 1, \dots, F$$

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$$\begin{aligned} I_{\varepsilon, k}^{\text{penalty}}(u, v) &= \int_{\Omega} (v^2 + \eta) |\nabla u|^2 d\mathbf{x} + \kappa \int_{\Omega} \left[\frac{1}{4\varepsilon} (1 - v)^2 + \varepsilon |\nabla v|^2 \right] d\mathbf{x} \\ &+ \underbrace{\frac{1}{\gamma_A} \int_{\Omega_{D^\pm}} (g(t_k) - u)^2 d\mathbf{x}}_{\text{admissible displacement}} + \underbrace{\frac{1}{\gamma_B} \int_{CR_{k-1}} v^2 d\mathbf{x}}_{\text{irreversibility}} \end{aligned}$$

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$$(u_\varepsilon(t_k), v_\varepsilon(t_k)) \in \arg \min_{\substack{u \in H^1(\Omega), \\ v \in H^1(\Omega; [0, 1])}} I_{\varepsilon, k}^{\text{penalty}}(u, v) \quad k = 1, \dots, F$$

$$v \in H^1(\Omega; [0, 1])$$

automatically fulfilled

The modified Ambrosio-Tortorelli model

[M. Artina, M. Fornasier, S. Micheletti, S.P. (2015a)]

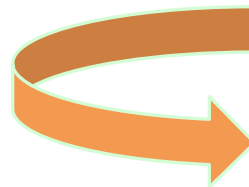
$$\begin{aligned} I(u, v) &= \int_{\Omega} [(v^2 + \eta)|\nabla u|^2 d\mathbf{x} + \alpha(1 - v)^2 + \varepsilon|\nabla v|^2] d\mathbf{x} \\ &+ \frac{1}{\gamma_A} \int_{\Omega_{D^\pm}} (g(t_k) - u)^2 d\mathbf{x} + \frac{1}{\gamma_B} \int_{CR_{k-1}} v^2 d\mathbf{x} \end{aligned}$$

[S. Burke, Ch. Ortner, E. Süli (2010)]

Looking for critical points : Fréchet differentiable

$$\begin{aligned} I'(w, z; \varphi, \psi) &= 2 \left(\int_{\Omega} (z^2 + \eta) \nabla w \cdot \nabla \varphi d\mathbf{x} + \frac{1}{\gamma_A} \int_{\Omega_{D^\pm}} (w - g(t_k)) \varphi d\mathbf{x} \right) \\ &+ 2 \left(\int_{\Omega} [z\psi |\nabla w|^2 + \alpha(z - 1)\psi + \varepsilon \nabla z \cdot \nabla \psi] d\mathbf{x} + \frac{1}{\gamma_B} \int_{CR_{k-1}} z\psi d\mathbf{x} \right) \\ &=: 2a_{\gamma_A}(z; w, \varphi) + 2b_{\gamma_B}(w; z, \psi) \end{aligned}$$

If $(u, v) \in H^1(\Omega) \times (H^1(\Omega) \cap L^\infty(\Omega))$ **is a critical point,**
then $0 \leq v(\mathbf{x}) \leq 1$ **a.e. in Ω .**



discrete counterpart

The discrete modified Ambrosio-Tortorelli model

[M. Artina, M. Fornasier, S. Micheletti, S.P. (2015a)]

$$I_h(u_h, v_h) = \int_{\Omega} \left[(P_h(v_h^2) + \eta) |\nabla u_h|^2 d\mathbf{x} + \alpha P_h((1 - v_h)^2) + \varepsilon |\nabla v_h|^2 \right] d\mathbf{x} \\ + \frac{1}{\gamma_A} \int_{\Omega_{D^\pm}} P_h((g_h(t_k) - u_h)^2) d\mathbf{x} + \frac{1}{\gamma_B} \int_{CR_{k-1}} P_h(v_h^2) d\mathbf{x}.$$

$P_h : C^0(\overline{\Omega}) \rightarrow X_h$ (Lagrangian interpolant) + hypotheses on g_h and on the stiffness matrix

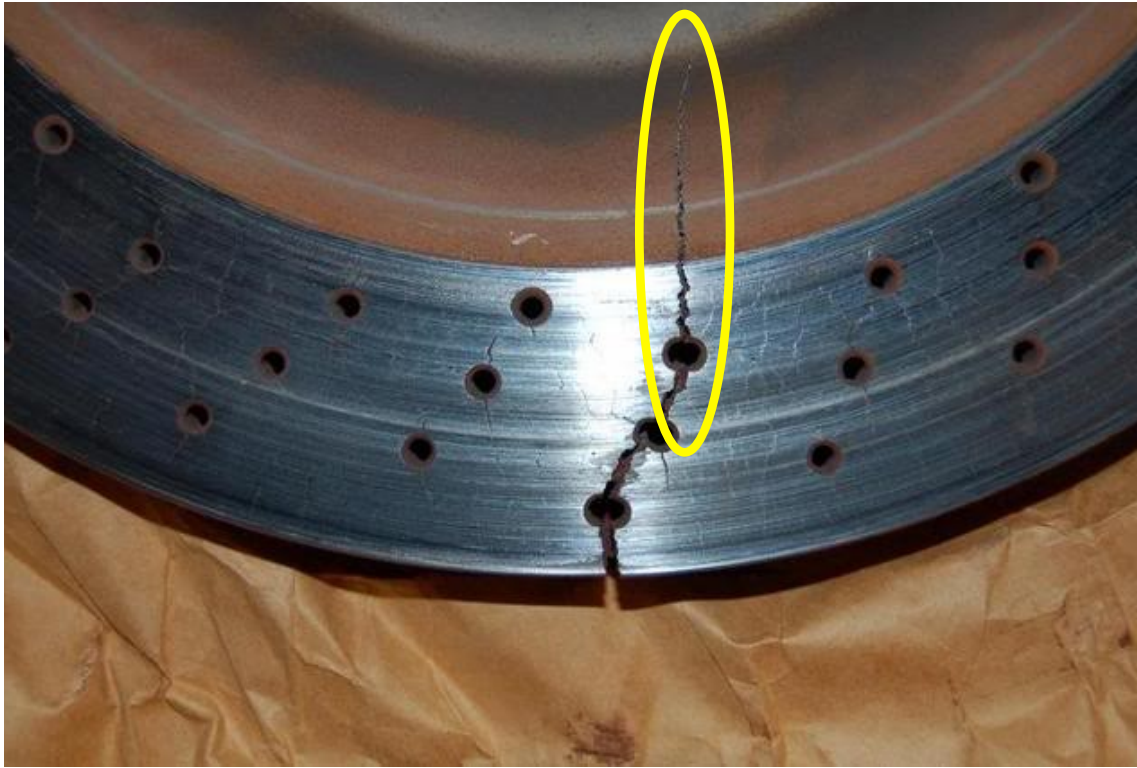
Looking for critical points : Fréchet differentiable

$$I'_h(u_h, v_h; \varphi_h, \psi_h) = 2 \left(\int_{\Omega} (P_h(v_h^2) + \eta) \nabla u_h \cdot \nabla \varphi_h d\mathbf{x} + \frac{1}{\gamma_A} \int_{\Omega_{D^\pm}} P_h((u_h - g_h(t_k)) \varphi_h) d\mathbf{x} \right) \\ + 2 \left(\int_{\Omega} \left[P_h(v_h \psi_h) |\nabla u_h|^2 + \alpha P_h((v_h - 1) \psi_h) + \varepsilon \nabla v_h \cdot \nabla \psi_h \right] d\mathbf{x} \right. \\ \left. + \frac{1}{\gamma_B} \int_{CR_{k-1}^h} P_h(v_h \psi_h) d\mathbf{x} \right) =: 2a_{\gamma_A}^h(v_h; u_h, \varphi_h) + 2b_{\gamma_B}^h(u_h; v_h, \psi_h).$$

If $(u_h, v_h) \in X_h \times X_h$, with X_h the P1-finite element space, is a critical point, then $0 \leq v_h \leq 1$ in Ω .



How to select the mesh?

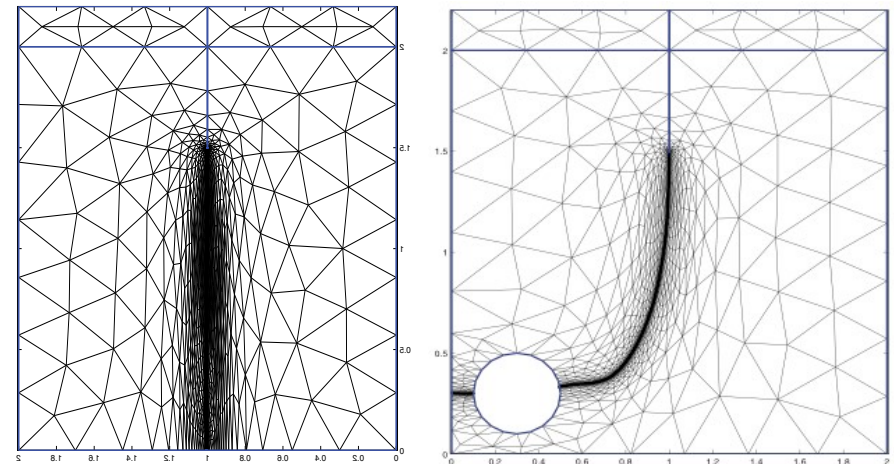


sharp features to be captured

anisotropic adapted meshes
sharply capture these features

reduce the dofs (or increase the accuracy) by tuning size, shape and orientation of mesh elements

[e.g., M. Fortin, J. Dompierre, W.G. Habashi, M.G. Vallet et al. (2000,2002)]



Anisotropic mesh adaptation

looking for anisotropic information

[L. Formaggia, S.P. (2001)]

interpolation error estimates

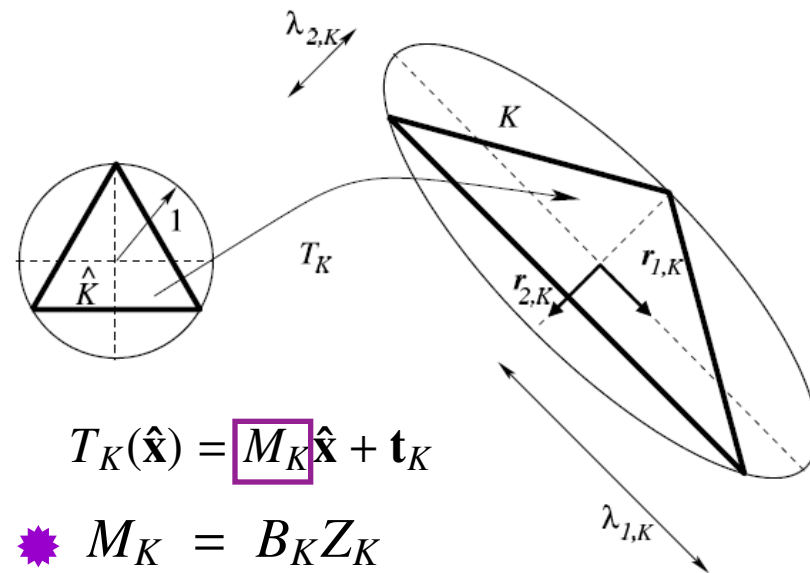
[L. Formaggia, S.P. (2001,2003)]

error indicators/estimators

[L. Formaggia, S.P. (2003); S. Micheletti,
S.P. (2008,2010,2011); P.E. Farrell, S.
Micheletti, S.P.(2010); M. Artina, M.
Fornasier, S. Micheletti, S.P. (2015)]

adaptation procedure

[L. Formaggia, S.P. (2001);
L. Formaggia, S. Micheletti, S.P. (2004)]



$$T_K(\hat{\mathbf{x}}) = \boxed{M_K} \hat{\mathbf{x}} + \mathbf{t}_K$$

$$\star M_K = B_K Z_K$$

$$\star B_K = R_K^T \Lambda_K R_K,$$

$$R_K^T = [\mathbf{r}_{1,K}, \mathbf{r}_{2,K}]$$

$$\Lambda_K = \text{diag}(\lambda_{1,K}, \lambda_{2,K})$$

$$\boxed{s_K = \lambda_{1,K} / \lambda_{2,K}}$$

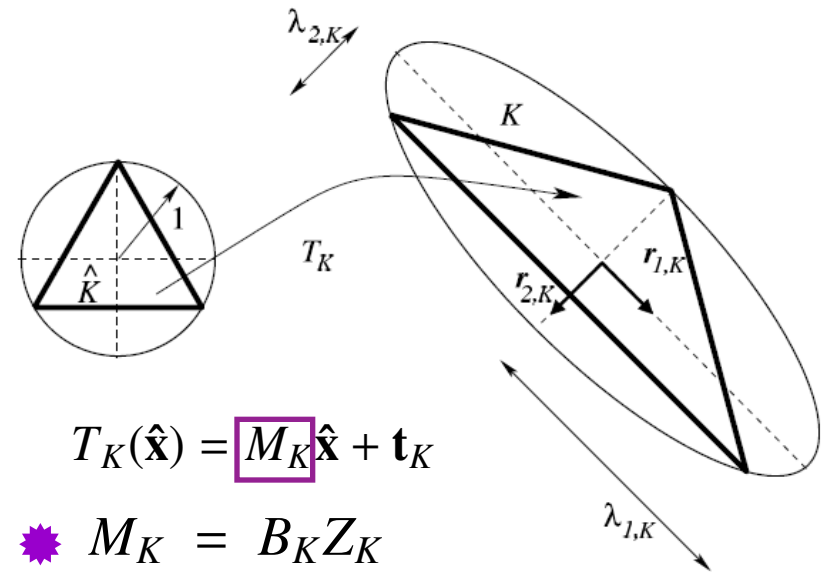
Anisotropic mesh adaptation

looking for anisotropic information

[L. Formaggia, S.P. (2001)]

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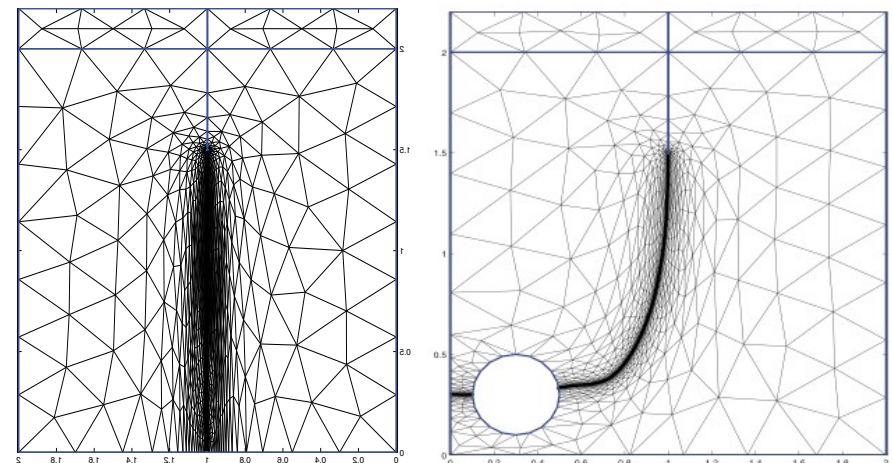
$$\boxed{s_K = \lambda_{1,K} / \lambda_{2,K}}$$

[L. Formaggia, S.P. (2003); S. Micheletti, S.P. (2008,2010,2011); P.E. Farrell, S. Micheletti, S.P.(2010); M. Artina, M. Fornasier, S. Micheletti, S.P. (2015)]

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[L. Formaggia, S.P. (2001);

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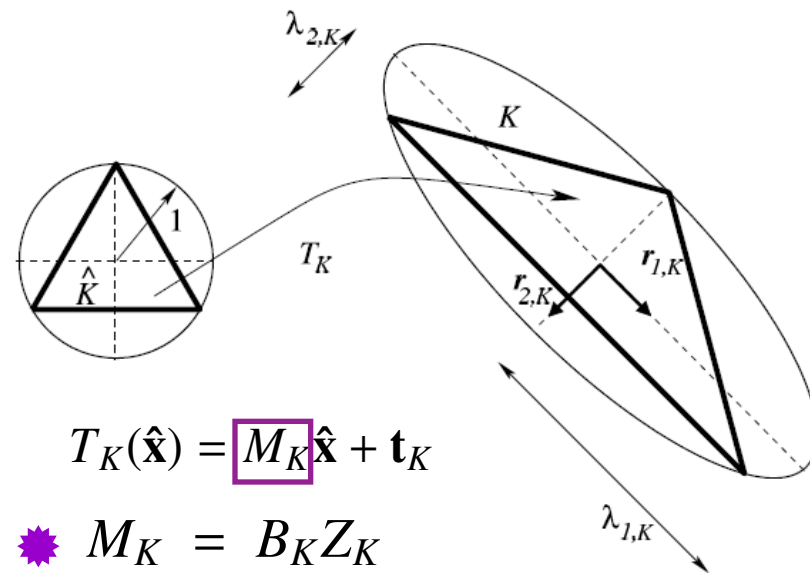
Anisotropic mesh adaptation

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$$R_K^T = [\mathbf{r}_{1,K}, \mathbf{r}_{2,K}]$$

$$\Lambda_K = \text{diag}(\lambda_{1,K}, \lambda_{2,K})$$

$$[G_K(\mathbf{w})]_{i,j} = \sum_{T \in \Delta_K} \int_T w_i w_j d\mathbf{x}, \quad \text{with } i, j = 1, 2$$

$$\boxed{s_K = \lambda_{1,K} / \lambda_{2,K}}$$

A

$$\|v - I_K^1(v)\|_{L^2(K)} \leq C_1^* \left(\sum_{i=1}^2 \lambda_{i,K}^2 (\mathbf{r}_{i,K}^T \boxed{G_K(\nabla v)} \mathbf{r}_{i,K}) \right)^{1/2}$$

I

$$\|v - I_K^1(v)\|_{L^2(K)} \leq C_1 \boxed{h_K} \boxed{|v|_{H^1(\Delta_K)}}$$

An anisotropic a posteriori error estimator

[M. Artina, M. Fornasier, S. Micheletti, S.P. (2015a)]

Goal : set an optimization procedure to minimize $I(u, v)$ by successive minimizations of $I_h(u_h, v_h)$ on anisotropic meshes.

$$|I'(u_h, v_h; \varphi, \psi)|$$

[S. Burke, Ch.. Ortner, E. Süli. (2010)]

$$|I'(u_h, v_h; \varphi, \psi)| \leq C \sum_{K \in \mathcal{T}_h} \left\{ \underbrace{\rho_K^A(v_h, u_h)}_{\text{residual}} \omega_K(\varphi) + \underbrace{\rho_K^B(u_h, v_h)}_{\text{residual}} \omega_K(\psi) \right\}$$

critical point of I_h
residual
residual

$$\forall \varphi, \psi \in H^1(\Omega)$$

$$\begin{aligned} \rho_K^A(v_h, u_h) = & \frac{1}{2} \| [\![\nabla u_h]\!] \|_{L^\infty(\partial K)} \| v_h^2 + \eta \|_{L^2(\partial K)} \left(\frac{h_K}{\lambda_{1,K} \lambda_{2,K}} \right)^{\frac{1}{2}} \\ & + \| 2v_h (\nabla v_h \cdot \nabla u_h) \|_{L^2(K)} + \frac{\delta_{K, \Omega_D^\pm}}{\gamma_A} \left(\| u_h - g_h(t_k) \|_{L^2(K)} \right. \\ & + \| g_h(t_k) - g(t_k) \|_{L^2(K)} \Big) + \frac{1}{\lambda_{2,K}} \left[\| v_h^2 - P_h(v_h^2) \|_{L^\infty(K)} \| \nabla u_h \|_{L^2(K)} \right. \\ & + \left. \frac{|K|^{1/2} h_K^2}{\gamma_A} |u_h - g_h(t_k)|_{W^{1,\infty}(K)} \right]. \end{aligned}$$

$$\begin{aligned} \rho_K^B(u_h, v_h) = & \| (|\nabla u_h|^2 + \alpha) v_h - \alpha \|_{L^2(K)} + \frac{\varepsilon}{2} \| [\![\nabla v_h]\!] \|_{L^2(\partial K)} \left(\frac{h_K}{\lambda_{1,K} \lambda_{2,K}} \right)^{1/2} \\ & + \frac{\delta_{K, CR_{k-1}}}{\gamma_B} \| v_h \|_{L^2(K)} + \frac{h_K^2}{\lambda_{2,K}} \left[\| |\nabla u_h|^2 + \alpha \|_{L^2(K)} \right. \\ & + \left. \frac{|K|^{1/2} \delta_{K, CR_{k-1}}}{\gamma_B} \right] |v_h|_{W^{1,\infty}(K)} \end{aligned}$$

An anisotropic a posteriori error estimator

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$$|I'(u_h, v_h; \varphi, \psi)| \leq C \sum_{K \in \mathcal{T}_h} \{ \rho_K^A(v_h, u_h) \underbrace{\omega_K(\varphi)}_{\text{weight}} + \rho_K^B(u_h, v_h) \underbrace{\omega_K(\psi)}_{\text{weight}} \}$$

critical point of I_h

$\forall \varphi, \psi \in H^1(\Omega)$

$$\omega_K(w) = \left[\sum_{i=1}^2 \lambda_{i,K}^2 (\mathbf{r}_{i,K}^T G_{\Delta_K}(w) \mathbf{r}_{i,K}) \right]^{1/2}$$

$$\left. \begin{array}{l} \varphi = u - u_h \\ \psi = v - v_h \end{array} \right\} I(u, v) - I(u_h, v_h) = \frac{1}{2} I'(u_h, v_h; u - u_h, v - v_h) + \underbrace{R}_{\text{third order remainder}}$$

+ gradient recovery procedures

[O.C. Zienkiewicz, J.Z. Zhu (1987,1992)]

$$\begin{aligned}
 I(u, v) &= \int_{\Omega} [(v^2 + \eta)|\nabla u|^2] d\mathbf{x} + \alpha(1 - v)^2 + \varepsilon|\nabla v|^2 \, d\mathbf{x} \\
 &+ \frac{1}{\gamma_A} \int_{\Omega_{D^\pm}} (g(t_k) - u)^2 d\mathbf{x} + \frac{1}{\gamma_B} \int_{CR_{k-1}} v^2 d\mathbf{x}
 \end{aligned}$$

non-convex
functional

The (alternate) minimization algorithm :

1. Set $k = 0$;
2. If $k = 0$, set $v^1 = 1$; else $v^1 = v(t_{k-1})$.

3. Set $i = 1$; $err = 1$;

while $err \geq VTOL$ **do**

4. $u^i = \arg \min_{z \in H^1(\Omega)} I(z, v^i)$;

5. $v^{i+1} = \arg \min_{z \in H^1(\Omega)} I(u^i, z)$;

6. $err = \|v^{i+1} - v^i\|_{L^\infty(\Omega)}$;

7. $i \leftarrow i + 1$;

end while

8. $u(t_k) = u^{i-1}$; $v(t_k) = v^i$;

9. $k \leftarrow k + 1$;

10. if $k > F$, stop; else goto 2.

two separate
convex
minimizations

How to make effective an error estimator: a metric-based approach

[L. Formaggia, S.P. (2001); L. Formaggia, S. Micheletti, S. P. (2004)]

CRITERION : minimize the number of the mesh elements for a fixed accuracy
+ error equidistribution

estimator $\longrightarrow$ *metric* $\longrightarrow$ *adapted mesh*

predictive procedure:

"Mesh is not a data but an unknown of the problem" (M. Fortin)

At each step of the adaptive process:

- ★ the actual mesh $\mathcal{T}_h^{(j)}$; *approximation of the differential problem*
- ★ the predicted (piecewise constant) metric $\widetilde{M}^{(j+1)}$ computed on $\mathcal{T}_h^{(j)}$;
- ★ the updated mesh $\mathcal{T}_h^{(j+1)}$.

How to make effective an error estimator: a metric-based approach

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"Mesh is not a data but an unknown of the problem" (M. Fortin)

At each step of the adaptive process:

- the actual mesh $\mathcal{T}_h^{(j)}$; construction of the error estimator local optimization problems
- the predicted (piecewise constant) metric $\widetilde{M}^{(j+1)}$ computed on $\mathcal{T}_h^{(j)}$;
- the updated mesh $\mathcal{T}_h^{(j+1)}$. BAMG

+ stopping criterion

Merging minimization with anisotropic mesh adaptation

[M. Artina, M. Fornasier, S. Micheletti, S.P. (2015a)]

1. Set $k = 0$, $\mathcal{T}_h^{(1)} = \mathcal{T}_h$;
2. If $k = 0$, set $v_h^1 = 1$; else $v_h^1 = v_h(t_{k-1})$;
3. Set $i = 1$; $err_{mesh} = 1$; $err = 1$;
- while** $err_{mesh} > \text{MESHTOL}$ & $err > \text{VTOL}$ **do**

$$4. u_h^i = \arg \min_{z_h \in X_h^{(i)}} I(z_h, v_h^i);$$

$$5. v_h^{i+1} = \arg \min_{z_h \in X_h^{(i)}} I(u_h^i, z_h);$$

minimization

6. Compute the new metric $\mathcal{M}^{(i+1)}$ based on u_h^i and v_h^{i+1} with $\text{TOL} = \text{REFTOL}$;

7. Build the adapted mesh $\mathcal{T}_h^{(i+1)}$;

$$8. err = \|v_h^{i+1} - v_h^i\|_{L^\infty(\Omega)};$$

$$9. err_{mesh} = |\#\mathcal{T}_h^{(i+1)} - \#\mathcal{T}_h^{(i)}| / \#\mathcal{T}_h^{(i)};$$

$$10. \text{Set } v_h^1 = \Pi_{i \rightarrow i+1}(v_h^{i+1});$$

$$11. i \leftarrow i + 1;$$

end while

mesh adaptation

$$12. u_h(t_k) = \Pi_{i-1 \rightarrow i}(u_h^{i-1}); v_h(t_k) = v_h^1; \mathcal{T}_h^k = \mathcal{T}_h^{(i)};$$

$$13. \text{Set } \mathcal{T}_h^{(1)} = \mathcal{T}_h^k;$$

$$14. k \leftarrow k + 1;$$

15. if $k > F$, stop; else goto 2.

MESHTOL : mesh stagnation

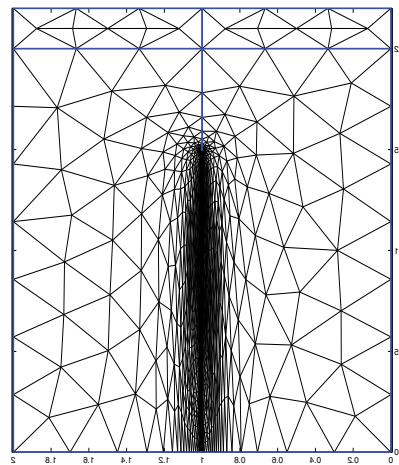
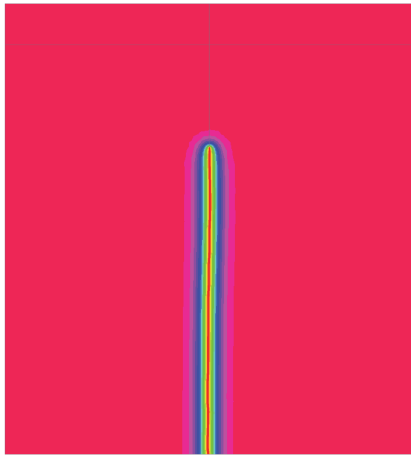
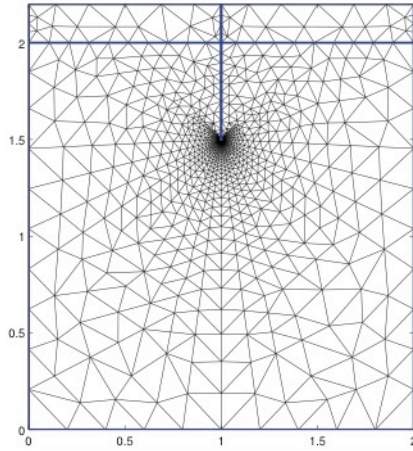
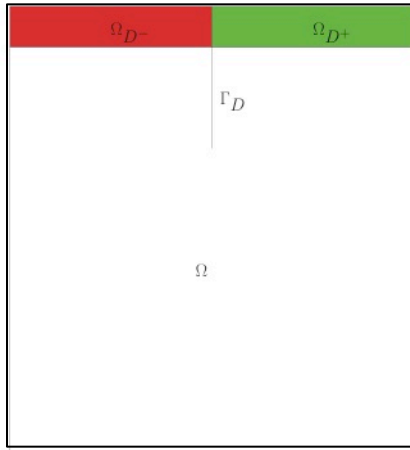
VTOL : minimization check

REFTOL : accuracy on the functional

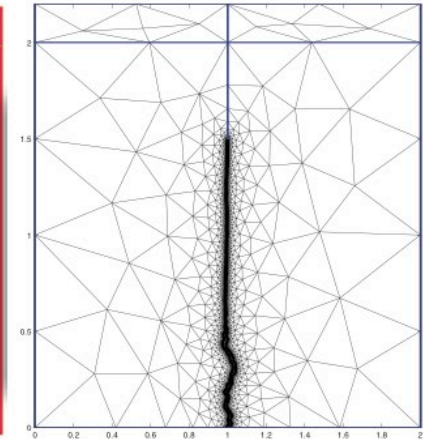
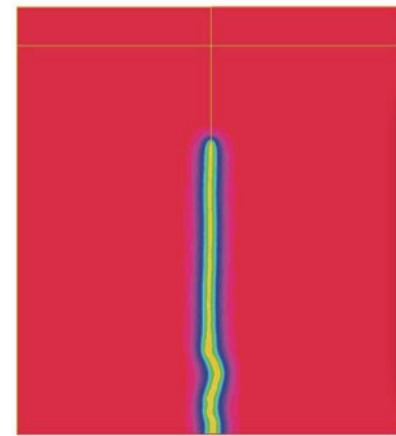


Numerical assessment: straight crack

[S. Burke, Ch. Ortner, E. Süli (2010)]



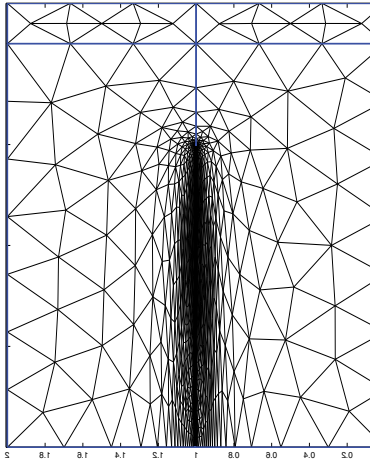
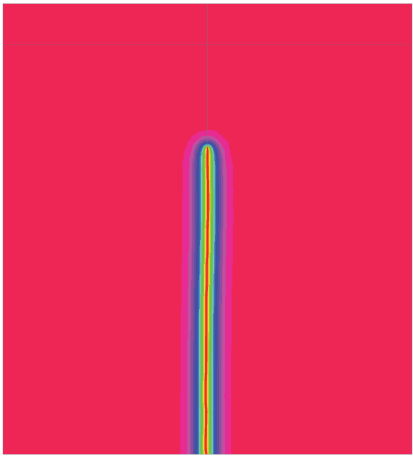
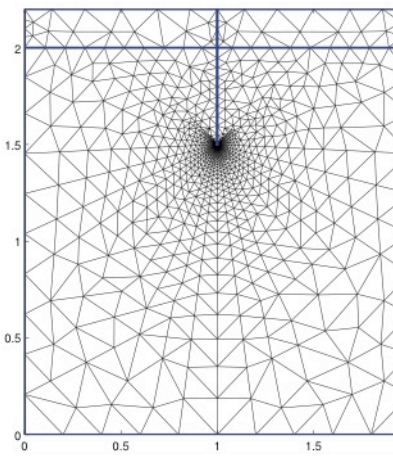
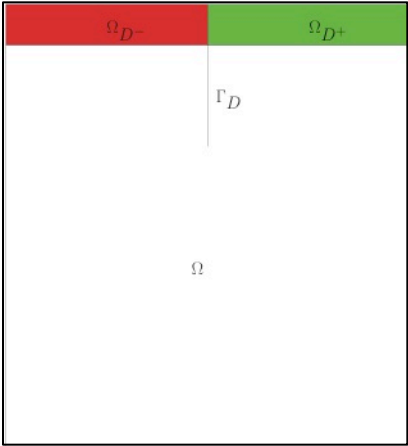
33927 elements
max aspect ratio : 1891.5



78025 elements

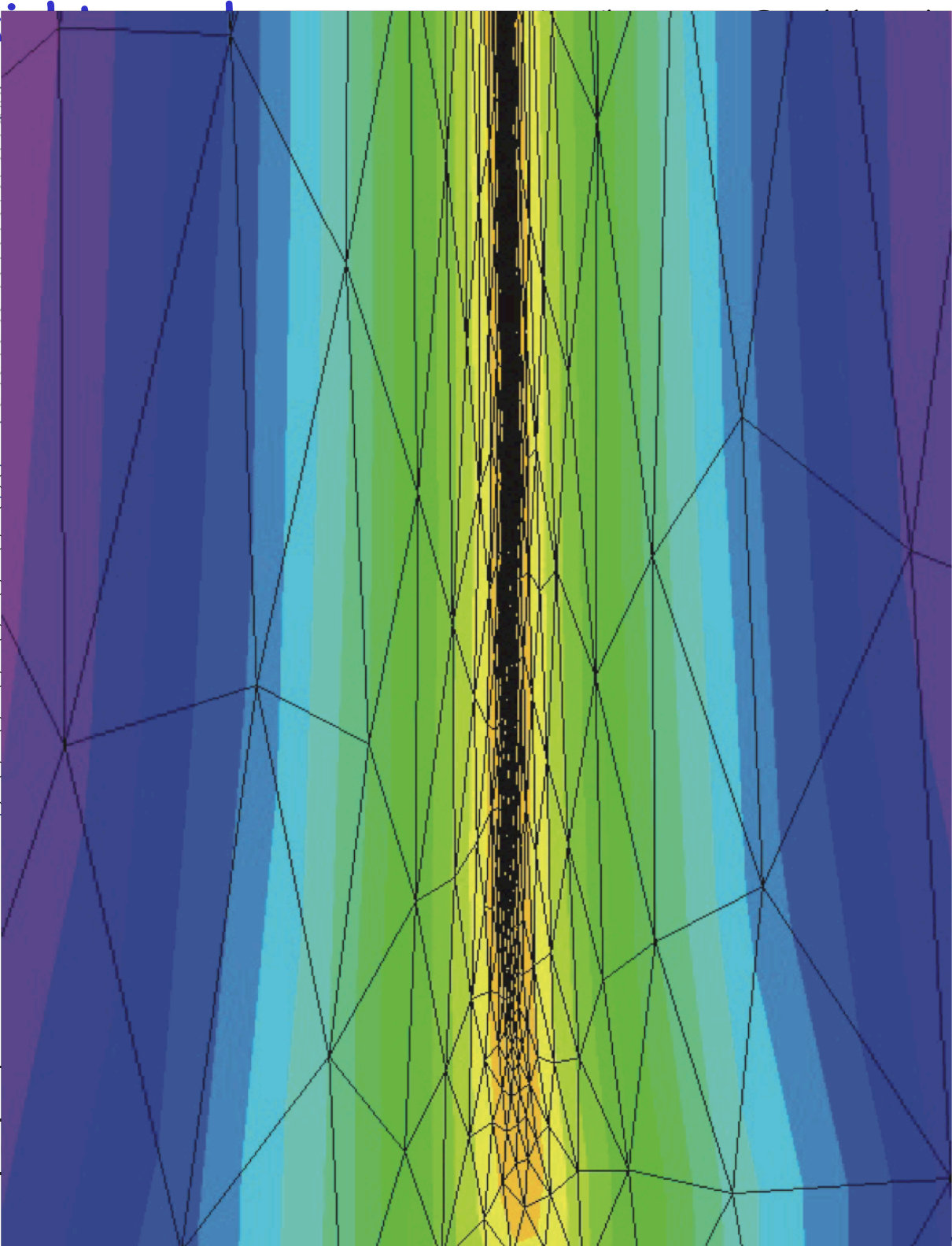
$\varepsilon = 2 \cdot 10^{-2}$	$\eta = 10^{-5}$	$\gamma_A = \gamma_B = 10^{-5}$	$\Delta t = 10^{-2}$
CRTOL= $3 \cdot 10^{-4}$	VTOL= $2 \cdot 10^{-3}$	MESHTOL= 10^{-2}	REFTOL= 10^{-2}

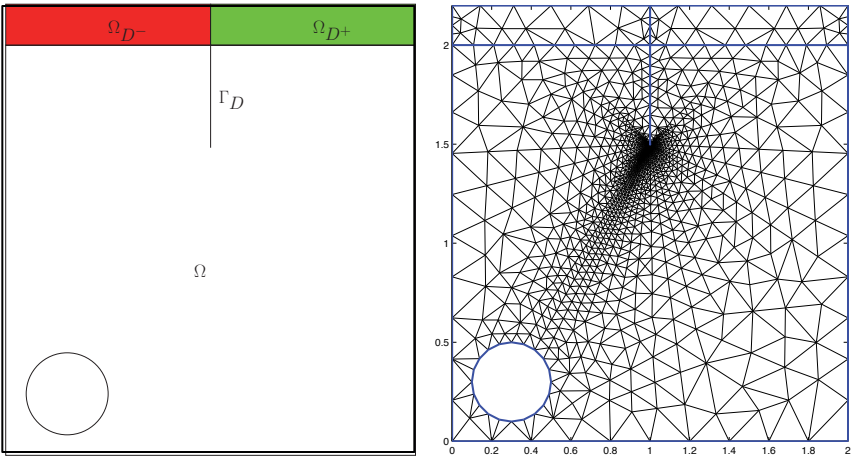
Numerical assessment: strain



33927 elements
max aspect ratio : 1891.5

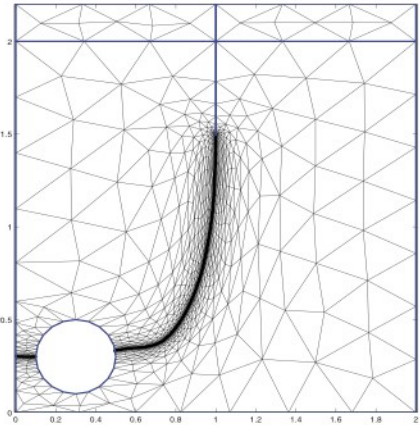
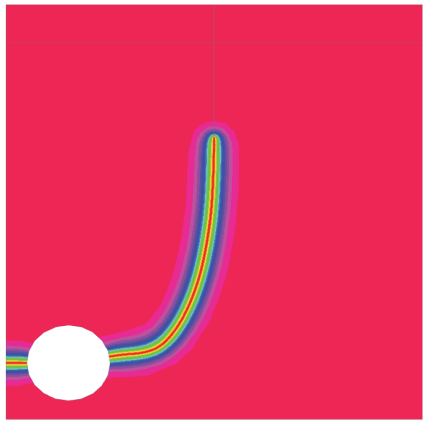
$\varepsilon = 2 \cdot 10^{-2}$	
CRTOL= $3 \cdot 10^{-4}$	VT



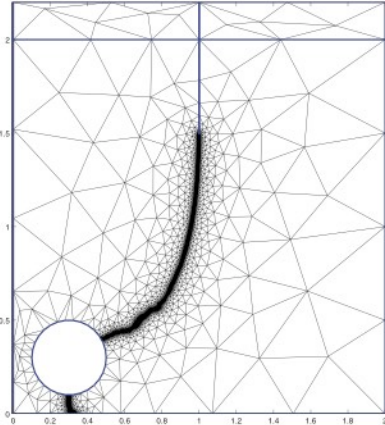
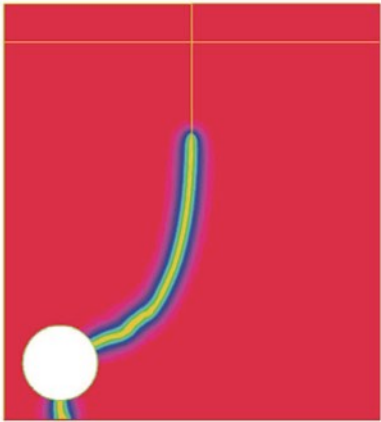
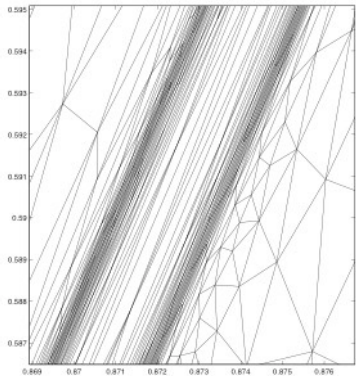


Minimization	Adaptation	Total
192	344	536
452	765	1217

aniso
iso



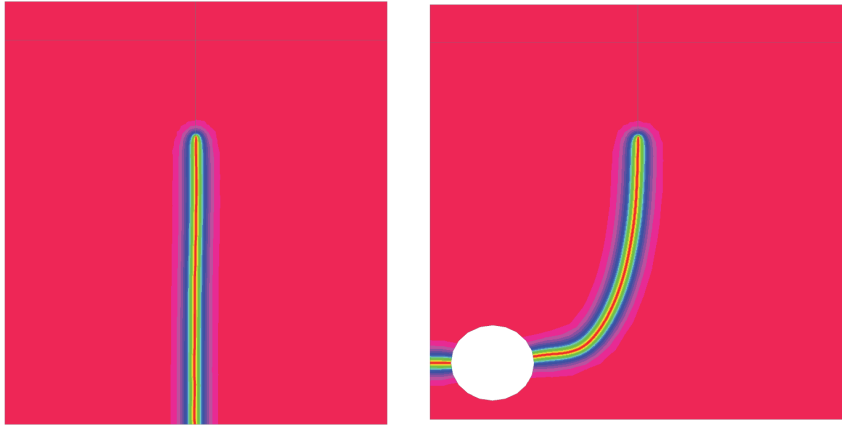
15987 elements
max aspect ratio : 1469.9



131367 elements

$\varepsilon = 2 \cdot 10^{-2}$	$\eta = 10^{-5}$	$\gamma_A = \gamma_B = 10^{-5}$	$\Delta t = 10^{-2}$
CRTOL= $3 \cdot 10^{-4}$	VTOL= $2 \cdot 10^{-3}$	MESHTOL= 10^{-2}	REFTOL= 10^{-2}

Variants of the basic crack model

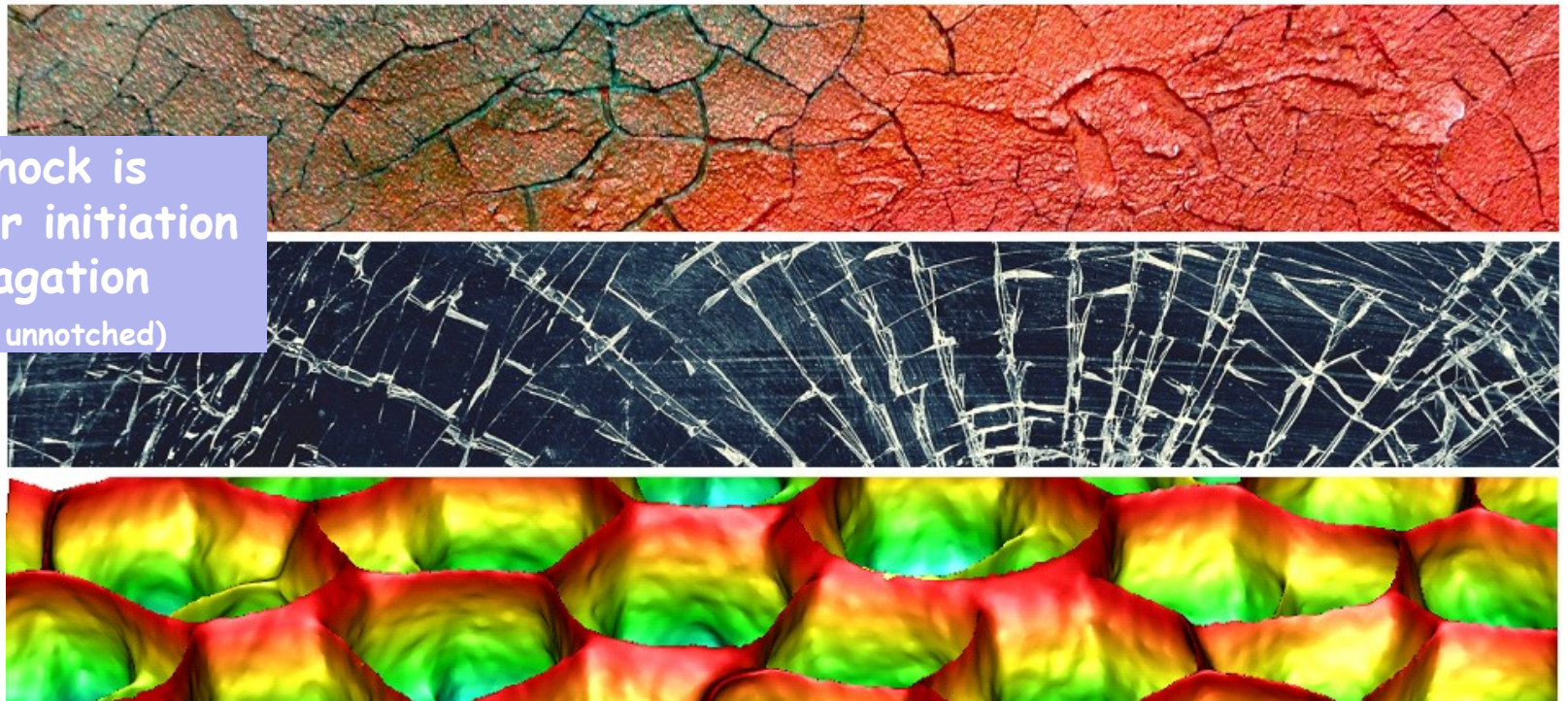


anti-plane case:
the force applied is
orthogonal to the domain

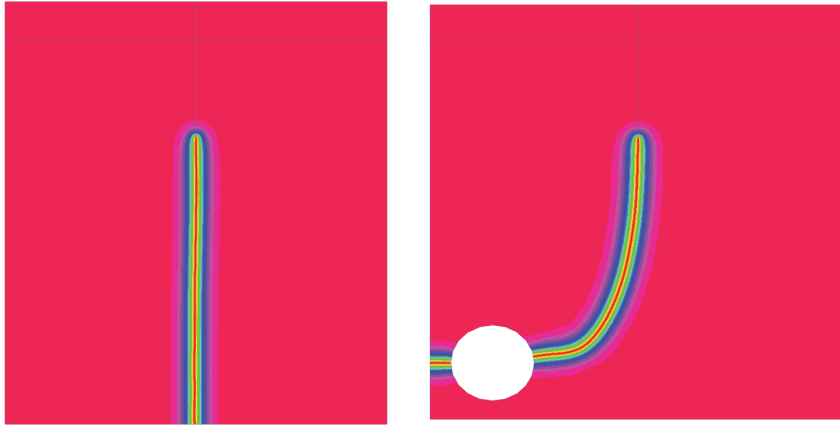
★ Brittle fractures induced by a thermal shock.

[J.-J. Marigo, C. Maurini, P. Sicsic (2013), B. Bourdin, J.-J. Marigo, C. Maurini, P. Sicsic (2014)]

thermal shock is
responsible for initiation
and propagation
(the material is unnotched)



Variants of the basic crack model



anti-plane case:
the force applied is
orthogonal to the domain

★ Brittle fractures induced by a thermal shock.

[J.-J. Marigo, C. Maurini, P. Sicsic (2013), B. Bourdin, J.-J. Marigo, C. Maurini, P. Sicsic (2014)]

temperature field

$$\begin{cases} \frac{\partial T}{\partial t} - \nabla \cdot (k_c \nabla T) = 0 & \text{in } \Omega \times (0, t_F] \\ T = T_0 - \Delta T & \text{on } \gamma_{\text{shock}} \times (0, t_F] \\ T(\mathbf{x}, 0) = T_0 & \text{on } \Omega, \end{cases}$$

thermal diffusivity

temperature drop

thermo-mechanical model

$$\mathcal{J}(\mathbf{u}, \alpha) = \frac{1}{2} \int_{\Omega} (1 - \alpha)^2 A(\varepsilon(\mathbf{u}) - \varepsilon_{th}) : (\varepsilon(\mathbf{u}) - \varepsilon_{th}) d\mathbf{x} + \int_{\Omega} \frac{G_c}{4c_w} \left(\frac{\alpha}{l} + l |\nabla \alpha|^2 \right) d\mathbf{x}$$

★ $\varepsilon(\mathbf{u}) = 1/2(\nabla \mathbf{u} + (\nabla \mathbf{u})^T)$ (total strain tensor)

★ G_c (body toughness)

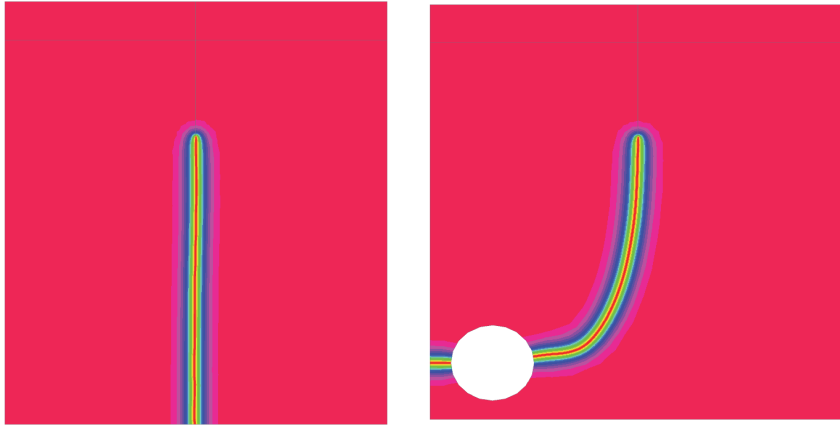
★ $\varepsilon_{th}(T) = \mu(T_0 - T)\mathbf{I}$ (inelastic strain)

★ A (stiffness tensor)

★ μ (thermal volumetric expansion coefficient)

★ l (internal length)

Variants of the basic crack model



anti-plane case:
the force applied is
orthogonal to the domain

★ Brittle fractures induced by a *thermal shock effect*.

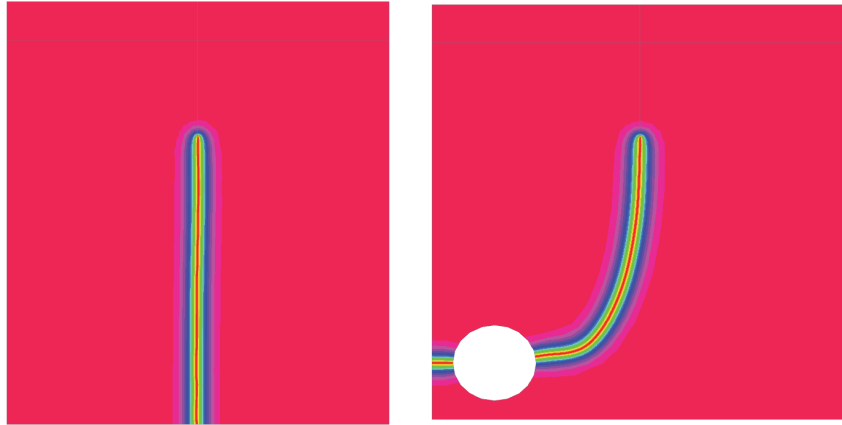
[J.-J. Marigo, C. Maurini, P. Sicsic (2013), B. Bourdin, J.-J. Marigo, C. Maurini, P. Sicsic (2014)]

Homogeneous uniformly heated specimen, free at the boundary, no pre-stress

Three phases :

- ★ **the very early stage** : a uniform strip propagates orthogonally to the immersed surface;
- ★ **at some critical time** : a bifurcation occurs and the damage exhibits periodically distributed bands of equal length and grows at the center of these bands;
- ★ **later**: a selective crack arrest takes place, some bands stop the others keep on propagating.

Variants of the basic crack model



anti-plane case:
the force applied is
orthogonal to the domain

★ Brittle fractures induced by a *thermal shock effect*.

[J.-J. Marigo, C. Maurini, P. Sicsic (2013), B. Bourdin, J.-J. Marigo, C. Maurini, P. Sicsic (2014)]

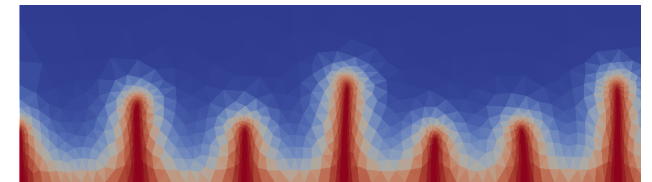
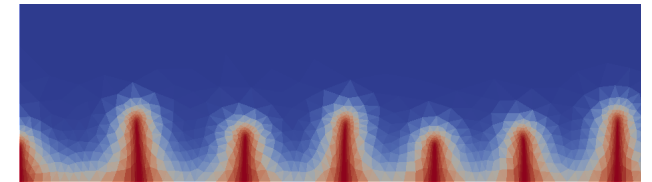
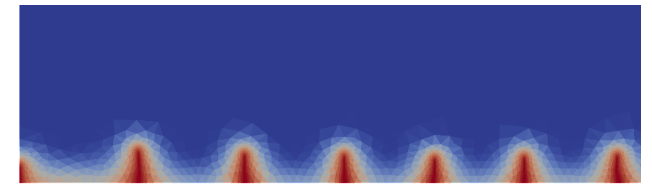
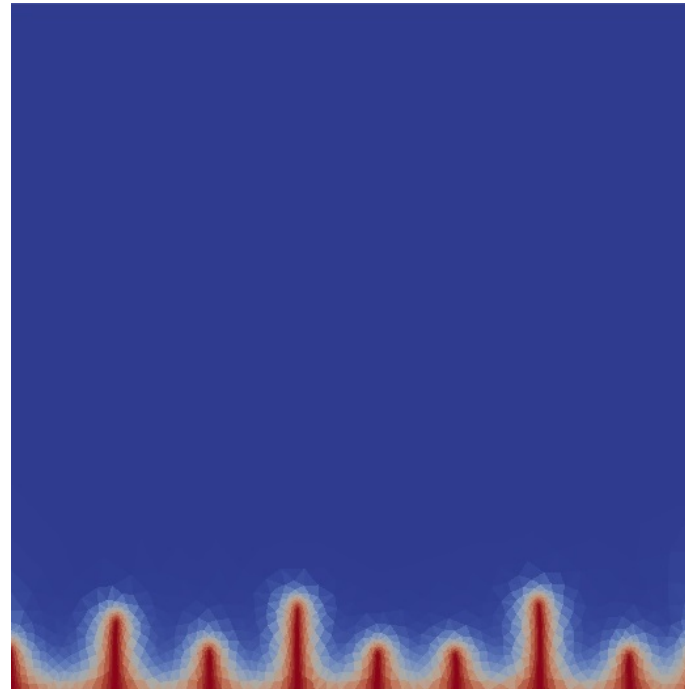
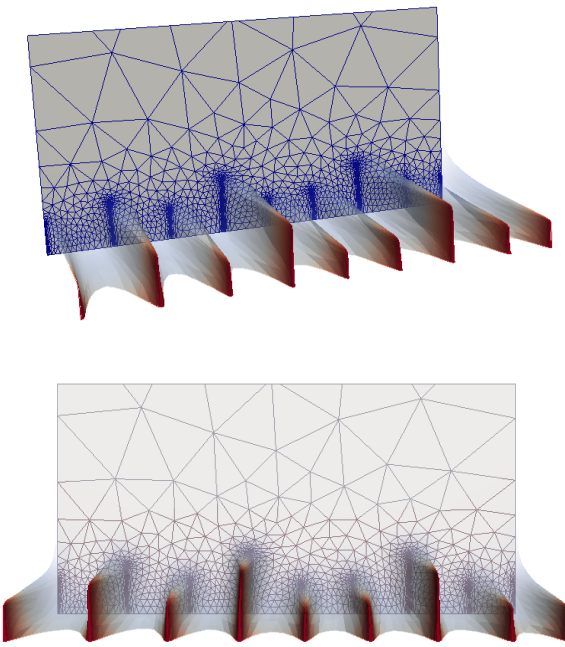
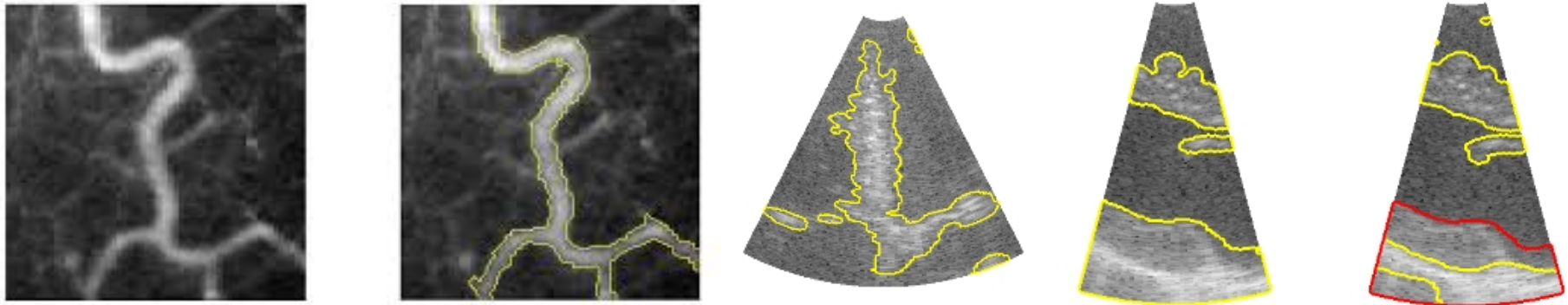
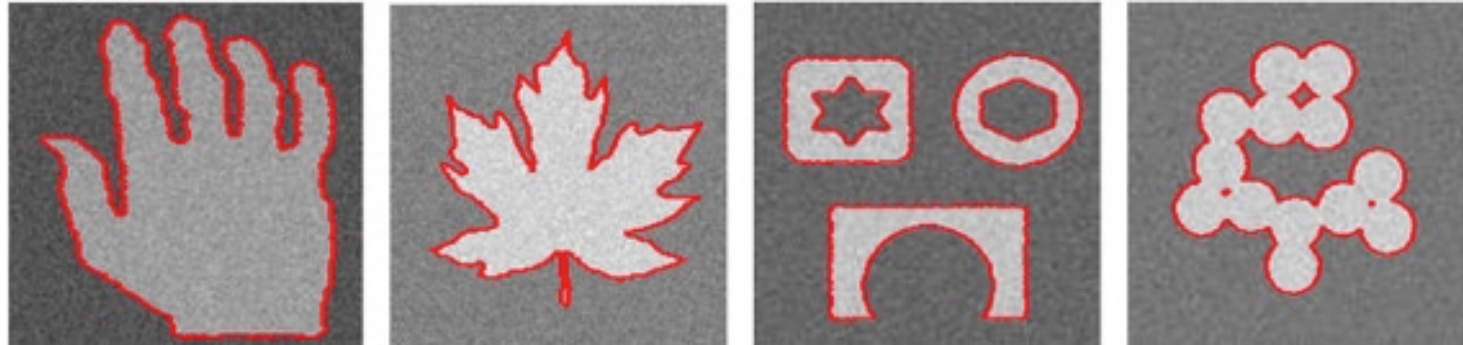


Image segmentation



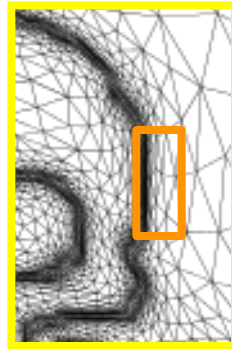
- thresholding/clustering/edge-detection/region-growing methods;
- PDE-based methods (e.g., level set method);
- variational methods.

Goal: identification of subregions of interest in a greyscale image.

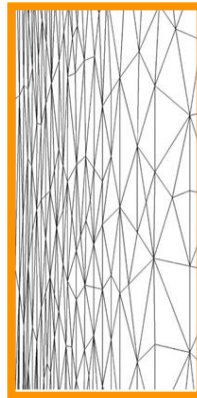
Mesh adaptation-aided image segmentation



strong gradients across the image edges



anisotropic adapted meshes sharply capture these features



reduce the dofs (or increase the accuracy) by tuning size, shape and orientation of mesh elements

[e.g., M. Fortin, J. Dompierre, W.G. Habashi, M.G. Vallet et al. (2000,2002)]

Goal: identification of subregions of interest in a greyscale image.

Image segmentation: a variational model

[D. Mumford, J. Shah (1989);
B. Bourdin (1999); A. Chambolle (1995)]

$$I(u, E) = \int_{\Omega} (u - f)^2 d\Omega + \beta \int_{\Omega \setminus E} |\nabla u|^2 d\Omega + \gamma \mathcal{H}(E) \quad (\text{Mumford-Shah})$$

variational approximation
↓
original image
↑
separating edge
↑

- ★ $f \in L^{\infty}(\Omega)$
- ★ E closed one-dimensional subset of Ω
- ★ $u \in H^1(\Omega \setminus E)$
- ★ β, γ positive parameters

Image segmentation: a variational model

[D. Mumford, J. Shah (1989);
B. Bourdin (1999); A. Chambolle (1995)]

$$I(u, E) = \underbrace{\int_{\Omega} (u - f)^2 d\Omega}_{\text{mismatch}} + \beta \underbrace{\int_{\Omega \setminus E} |\nabla u|^2 d\Omega}_{\text{smoothness}} + \gamma \underbrace{\mathcal{H}(E)}_{\text{Hausdorff measure}} \quad (\text{Mumford-Shah})$$

- ★ $f \in L^{\infty}(\Omega)$
- ★ E closed one-dimensional subset of Ω
- ★ $u \in H^1(\Omega \setminus E)$
- ★ β, γ positive parameters

numerically hard



Ambrosio-Tortorelli approximation

$$I_{\epsilon}(u, v) = \underbrace{\int_{\Omega} (u - f)^2 d\Omega}_{\text{mismatch}} + \beta \underbrace{\int_{\Omega} (v^2 + \eta) |\nabla u|^2 d\Omega}_{\text{smoothness}} + \gamma \underbrace{\int_{\Omega} \left(\epsilon |\nabla v|^2 + \frac{1}{4\epsilon} (v - 1)^2 \right) d\Omega}_{\text{Hausdorff measure}}$$

- ★ v with values in $[0, 1]$: approximate indicator of the set E with thickness ϵ



object
boundary

object and background

[L. Ambrosio, V.M. Tortorelli (1990,1992)]
[P. D'Ambra, G. Tartaglione (2015)]

Image segmentation: a variational model

[D. Mumford, J. Shah (1989);
B. Bourdin (1999); A. Chambolle (1995)]

$$I(u, E) = \underbrace{\int_{\Omega} (u - f)^2 d\Omega}_{\text{mismatch}} + \beta \underbrace{\int_{\Omega \setminus E} |\nabla u|^2 d\Omega}_{\text{smoothness}} + \gamma \underbrace{\mathcal{H}(E)}_{\text{Hausdorff measure}} \quad (\text{Mumford-Shah})$$

- ★ $f \in L^{\infty}(\Omega)$
- ★ E closed one-dimensional subset of Ω
- ★ $u \in H^1(\Omega \setminus E)$
- ★ β, γ positive parameters

numerically hard



Ambrosio-Tortorelli approximation

$$I_{\epsilon}(u, v) = \int_{\Omega} (u - f)^2 d\Omega + \beta \int_{\Omega} (v^2 + \eta) |\nabla u|^2 d\Omega + \gamma \int_{\Omega} \left(\epsilon |\nabla v|^2 + \frac{1}{4\epsilon} (v - 1)^2 \right) d\Omega$$

u and v defined on the whole Ω

Image segmentation: a variational model

[D. Mumford, J. Shah (1989);
B. Bourdin (1999); A. Chambolle (1995)]

$$I(u, E) = \underbrace{\int_{\Omega} (u - f)^2 d\Omega}_{\text{mismatch}} + \beta \underbrace{\int_{\Omega \setminus E} |\nabla u|^2 d\Omega}_{\text{smoothness}} + \gamma \underbrace{\mathcal{H}(E)}_{\text{Hausdorff measure}} \quad (\text{Mumford-Shah})$$

- ★ $f \in L^{\infty}(\Omega)$
- ★ E closed one-dimensional subset of Ω
- ★ $u \in H^1(\Omega \setminus E)$
- ★ β, γ positive parameters

numerically hard



Ambrosio-Tortorelli approximation

$$I_{\epsilon}(u, v) = \int_{\Omega} (u - f)^2 d\Omega + \beta \int_{\Omega} (v^2 + \eta) |\nabla u|^2 d\Omega + \gamma \int_{\Omega} \left(\epsilon |\nabla v|^2 + \frac{1}{4\epsilon} (v - 1)^2 \right) d\Omega$$

$\{I_{\epsilon}\}_{\epsilon}$ is Γ -convergent to $I(u, E)$ for $\epsilon \rightarrow 0$

Synthetic images

(sensitivity to model parameters)

[A.S. Chiappa, S. Micheletti, R. Peli, S.P., submitted]

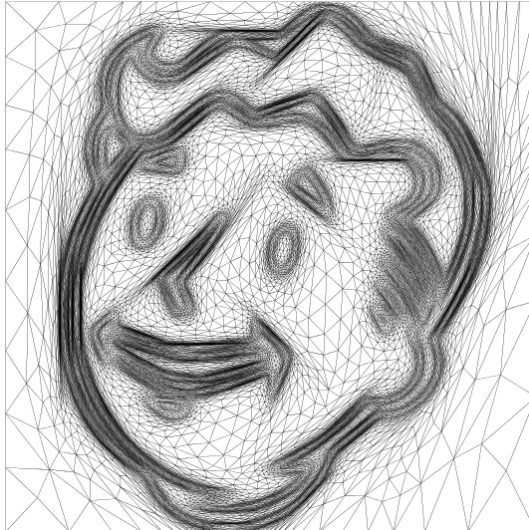
sensitivity to ε



(a) true image



(b) approximate image u_h



16472
triangles

(c) anisotropic adapted mesh



(d) approximate indicator v_h

$$\epsilon = 5 \cdot 10^{-2}, \eta = 10^{-4}, \beta = 10^{-2}, \gamma = 4,$$

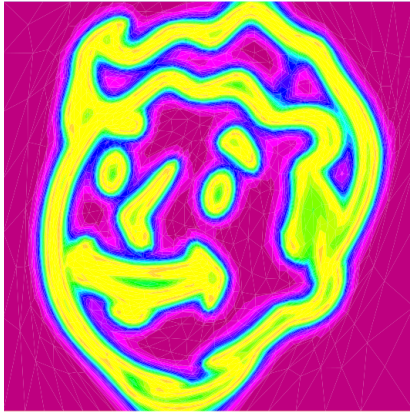
$$\text{TOL} = 50, \text{TOLs} = 10^{-2}, \text{TOLfp} = 5 \cdot 10^{-3}, \text{Ns} = 10, \text{Nfp} = 30$$

Synthetic images

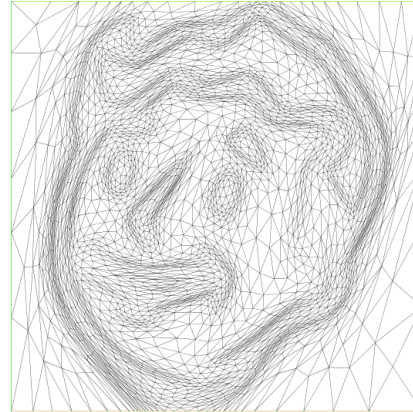
[A.S. Chiappa, S. Micheletti, R. Peli, S.P., submitted]

(sensitivity to model parameters)

sensitivity to ϵ



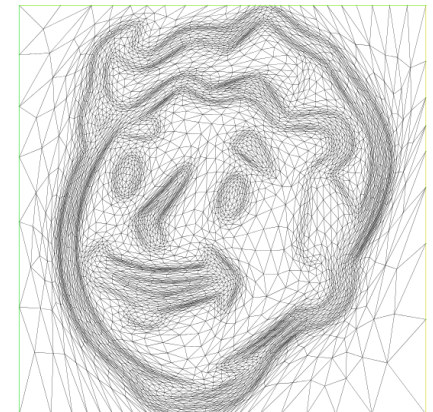
(a) $\epsilon = 1$: indicator v_h



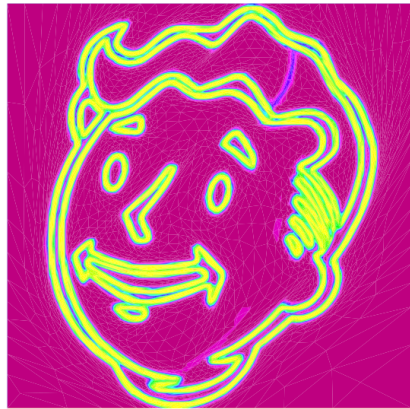
(b) $\epsilon = 1$: adapted mesh



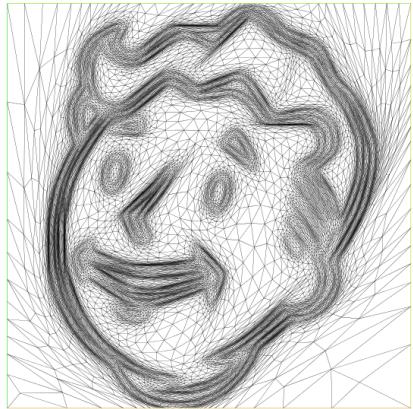
(c) $\epsilon = 5 \cdot 10^{-1}$: indicator v_h



(d) $\epsilon = 5 \cdot 10^{-1}$: adapted mesh



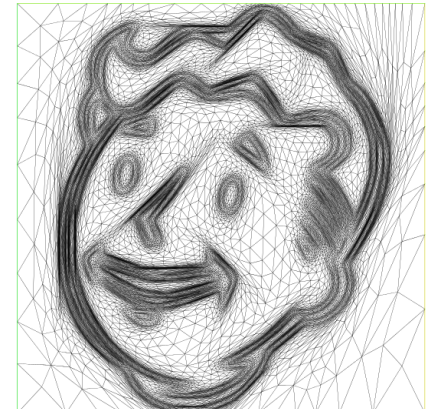
(e) $\epsilon = 10^{-1}$: indicator v_h



(f) $\epsilon = 10^{-1}$: adapted mesh



(g) $\epsilon = 5 \cdot 10^{-2}$: indicator v_h



(h) $\epsilon = 5 \cdot 10^{-2}$: adapted mesh

ϵ	$\max_K s_K$	# vertices	# elements
1	38.71	2654	5236
$5 \cdot 10^{-1}$	34.43	4455	8834
10^{-1}	140.91	11426	22752
$5 \cdot 10^{-2}$	381.39	16472	32855

Synthetic images

[A.S. Chiappa, S. Micheletti, R. Peli, S.P., submitted]

(sensitivity to model parameters)

$$\epsilon = 10^{-1}, \eta = 10^{-4}, \gamma = 4,$$

$$\text{TOL} = 80, \text{TOLs} = 10^{-2},$$

$$\text{TOLfp} = 5 \cdot 10^{-3},$$

$$\text{Ns} = 10, \text{Nfp} = 50$$



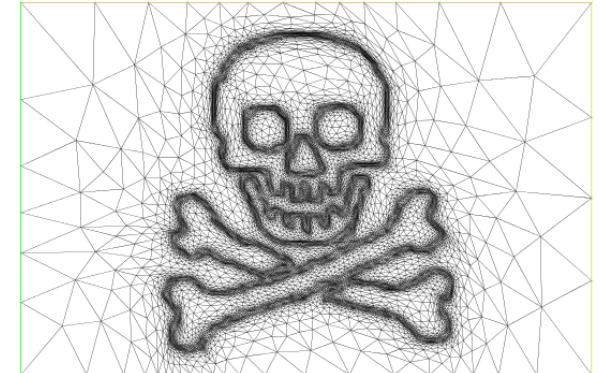
(b) $\beta = 1.5$: indicator v_h



(c) $\beta = 1.5$: adapted mesh



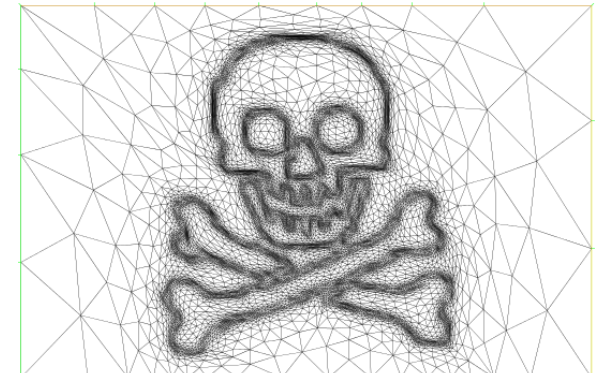
(d) $\beta = 1$: indicator v_h



(e) $\beta = 1$: adapted mesh



(f) $\beta = 5 \cdot 10^{-1}$: indicator v_h



(g) $\beta = 5 \cdot 10^{-1}$: adapted mesh

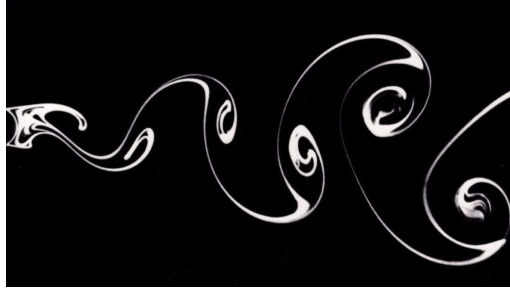
β	$\max_K s_K$	# vertices	# elements
1.5	97.78	19560	39077
1	66.01	16572	33097
$5 \cdot 10^{-1}$	107.31	12294	24553

Synthetic images

[A.S. Chiappa, S. Micheletti, R. Peli, S.P., submitted]

(sensitivity to model parameters)

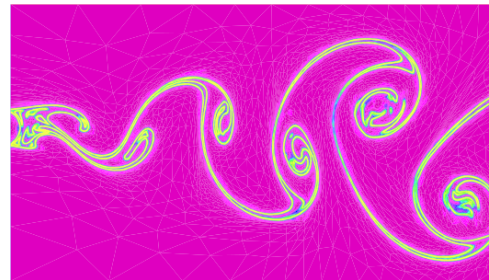
sensitivity to γ



(a) true image

$\epsilon = 10^{-1}, \eta = 10^{-4},$
 $\beta = 5 \cdot 10^{-2}, \text{TOL} = 80,$
 $\text{TOLs} = 10^{-2},$
 $\text{TOLfp} = 5 \cdot 10^{-3},$
 $\text{Ns} = 10, \text{Nfp} = 50$

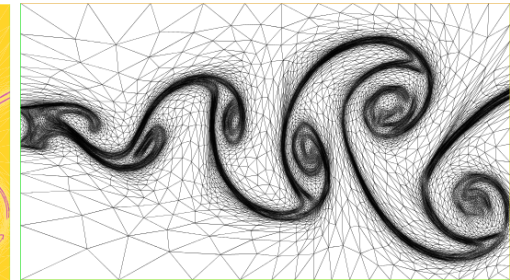
γ	$\max_K s_K$	# vertices	# elements
4	145.09	23455	46727
1	149.50	16246	32299
0.25	218.44	11109	22033



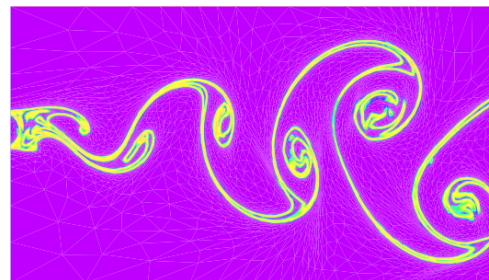
(b) $\gamma = 4$: indicator v_h



(c) $\gamma = 4$: thresholded indicator



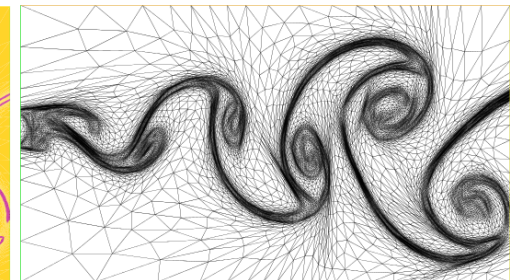
(d) $\gamma = 4$: adapted mesh



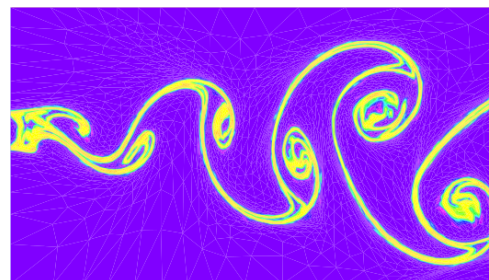
(e) $\gamma = 1$: indicator v_h



(f) $\gamma = 1$: thresholded indicator



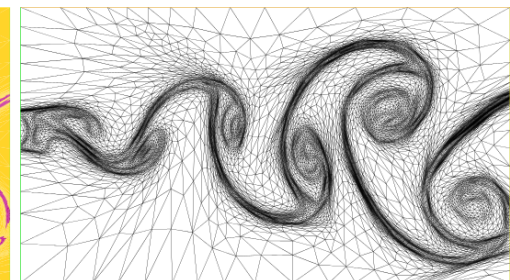
(g) $\gamma = 1$: adapted mesh



(h) $\gamma = 0.25$: indicator v_h



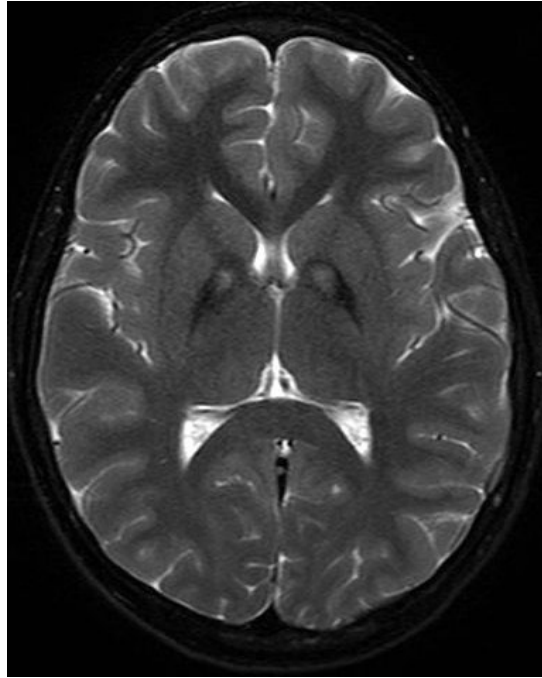
(i) $\gamma = 0.25$: thresholded indicator



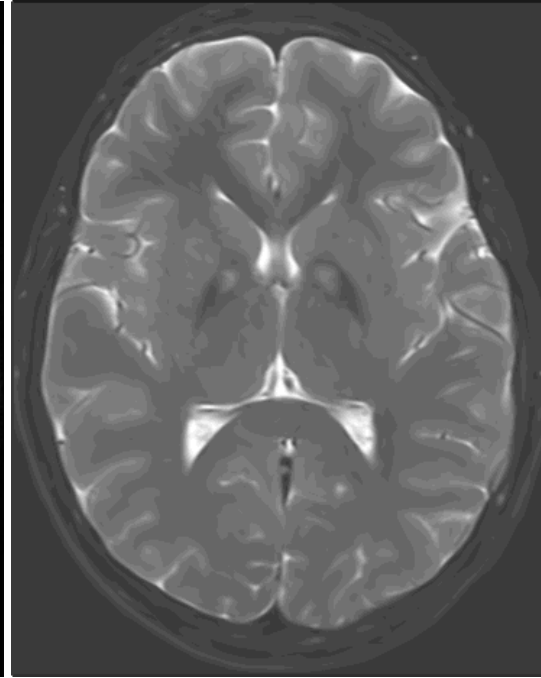
(j) $\gamma = 0.25$: adapted mesh

Real images

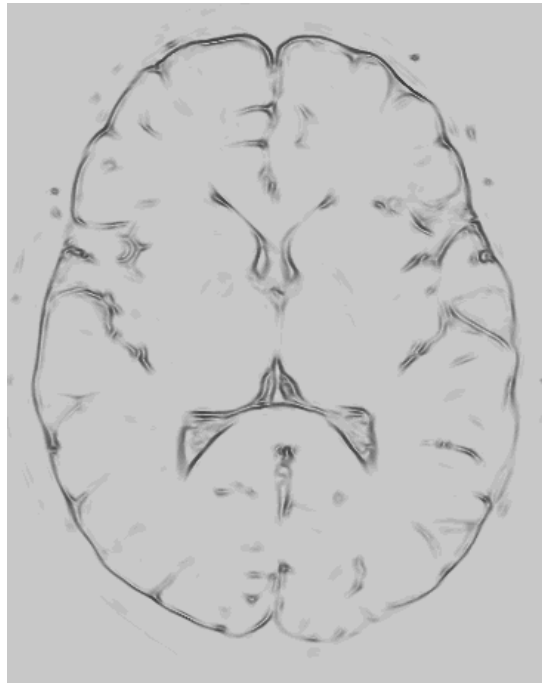
[A.S. Chiappa, S. Micheletti, R. Peli, S.P., submitted]



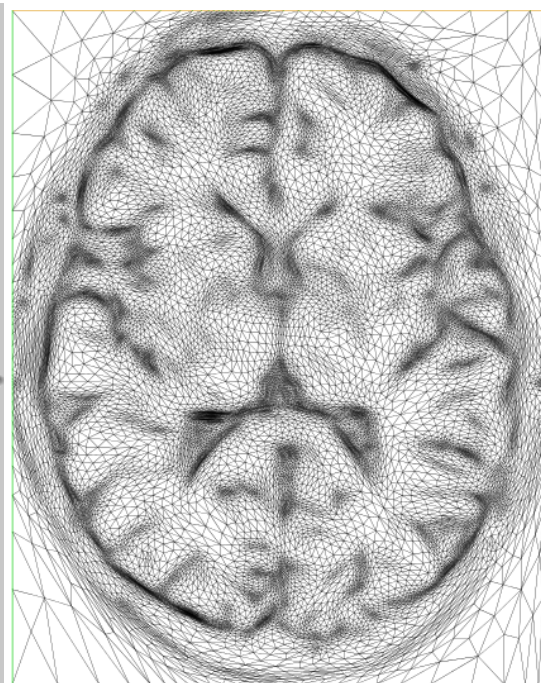
(a) real image (385×479 px)



(b) approximate image u_h

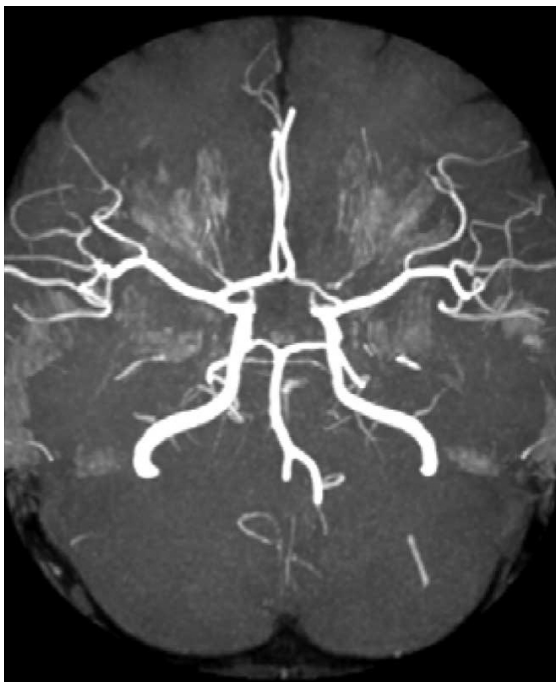


(c) indicator v_h

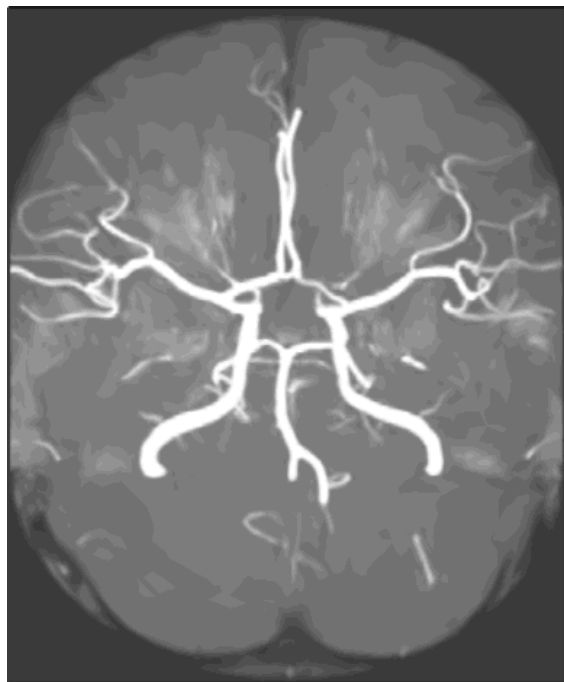


(d) adapted mesh

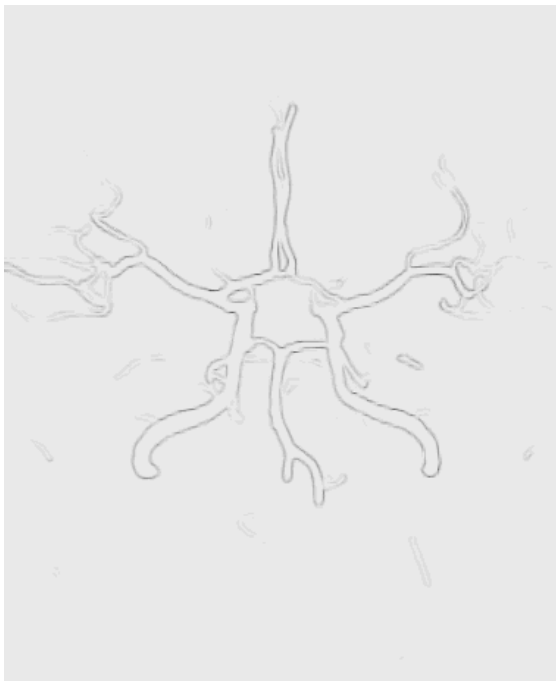
39334
triangles



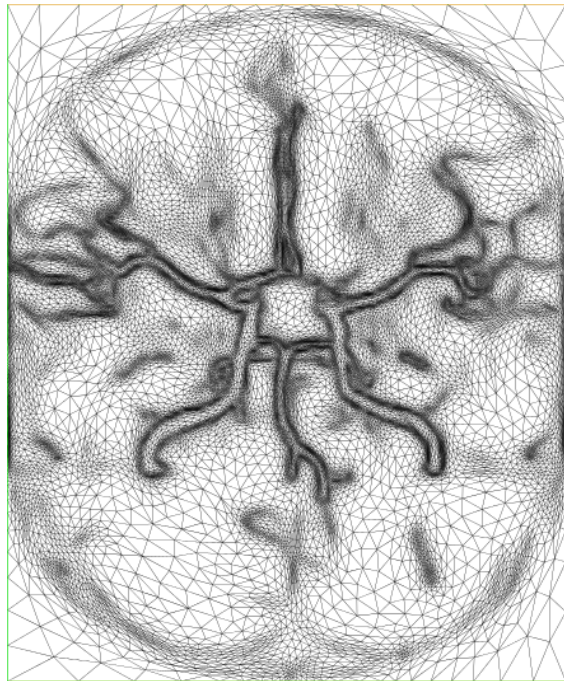
(a) real image (663×804 px)



(b) approximate image u_h



(c) indicator v_h



(d) adapted mesh

49544
triangles

Real images

[A.S. Chiappa, S. Micheletti, R. Peli, S.P., submitted]



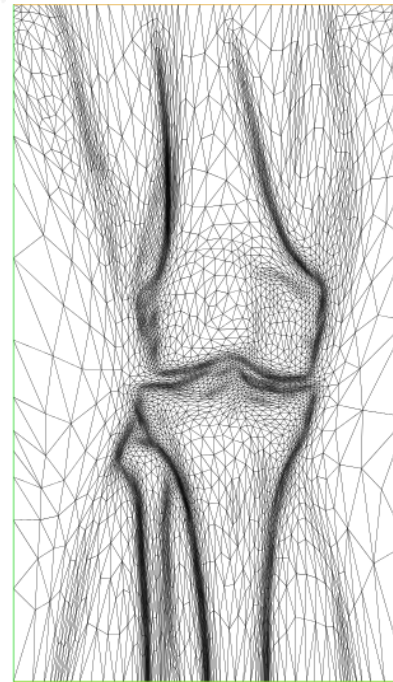
(a) real image (261×448 px)



(b) approximate image u_h



(c) indicator v_h



(d) adapted mesh

11175
triangles

Real images

[A.S. Chiappa, S. Micheletti, R. Peli, S.P., submitted]



(a) real image (312×454 px)



(b) approximate image u_h



(c) indicator v_h



(d) adapted mesh

17056
triangles

input parameters

image	ϵ	β	γ	η	TOL	TOLs	TOLfp	Ns	Nfp
brain	10^{-3}	10^{-1}	$5 \cdot 10^{-1}$	10^{-6}	70	10^{-2}	$5 \cdot 10^{-3}$	10	40
Willis	10^{-3}	10^{-1}	5	10^{-6}	100	10^{-2}	$5 \cdot 10^{-3}$	10	40
knee	10^{-3}	$2 \cdot 10^{-1}$	10	10^{-6}	60	10^{-2}	$5 \cdot 10^{-3}$	10	40
hand	10^{-3}	10^{-1}	10	10^{-6}	60	10^{-2}	$5 \cdot 10^{-3}$	10	40

output data

image	$\max_K s_K$	# vertices	# elements	# ITs	# ITfp
brain	71.94	19726	39334	4	4 – 8 – 9 – 25
Willis	56.51	24845	49544	3	2 – 4 – 4
knee	82.01	5683	11175	4	3 – 5 – 7 – 6
hand	38.01	8592	17056	5	2 – 3 – 2 – 2 – 2

output data

image	E-time	AD-time	FP-time
brain	170.74	43.77 – 24.95 – 19.45 – 17.58	11.26 – 12.57 – 10.62 – 25.51
Willis	102.78	29.76 – 24.12 – 22.93	3.63 – 5.78 – 5.3
knee	56.29	23.56 – 7.66 – 5.56 – 4.64	4.87 – 2.31 – 2.24 – 1.54
hand	79.90	29.83 – 11.32 – 8.53 – 7.48 – 7.3	4.14 – 2.01 – 0.97 – 0.87 – 0.81

Comparison with a pixel mesh

[B. Bourdin (1999)]

$$\epsilon = 10^{-1}, \eta = 10^{-2},$$

$$\beta = 2, \gamma = 10^{-2}$$

$$\text{TOL} = 50, \text{TOLs} = 10^{-2},$$

$$\text{TOLfp} = 5 \cdot 10^{-3},$$

$$\text{Ns} = 10, \text{Nfp} = 50$$



(a) real image (512×512 px)



(b) approximate image u_h



(c) thresholded indicator v_h

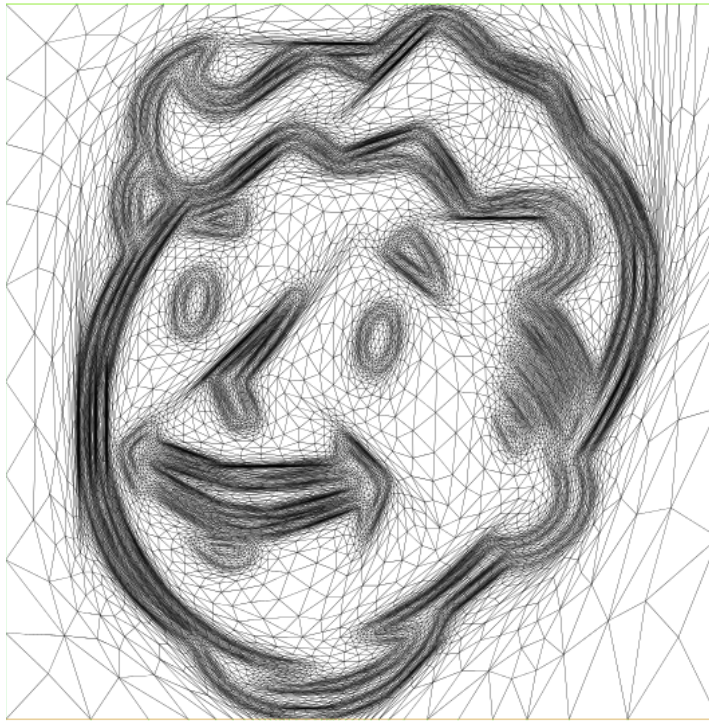


(d) adapted mesh

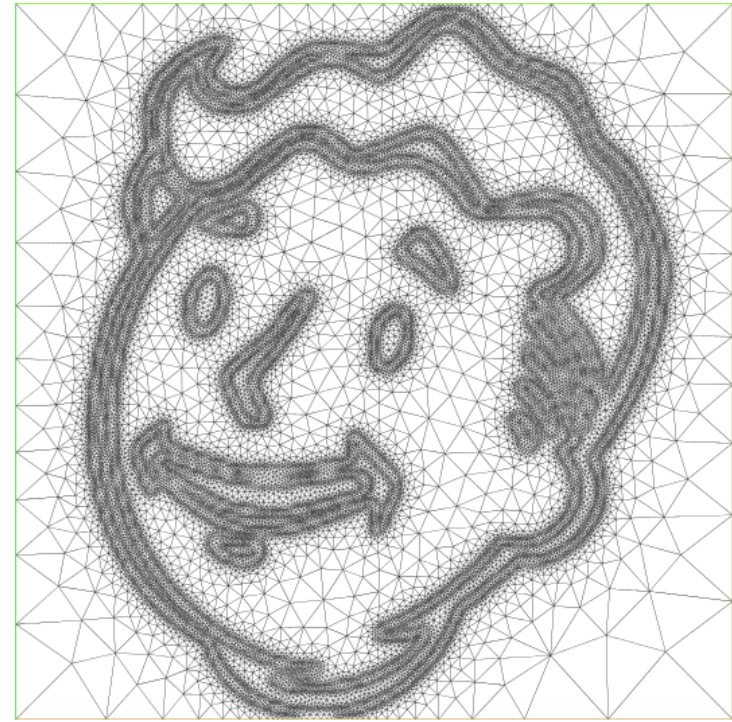
67679
triangles

Comparison with isotropic mesh adaptation

32855
triangles

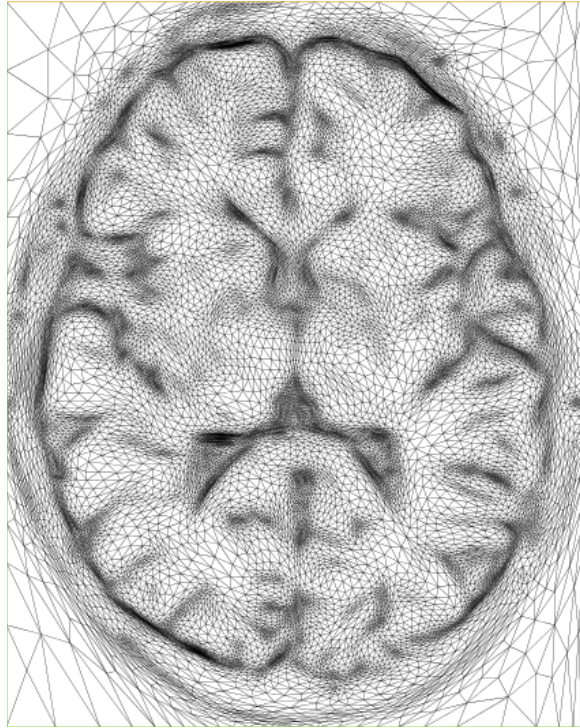


62023
triangles

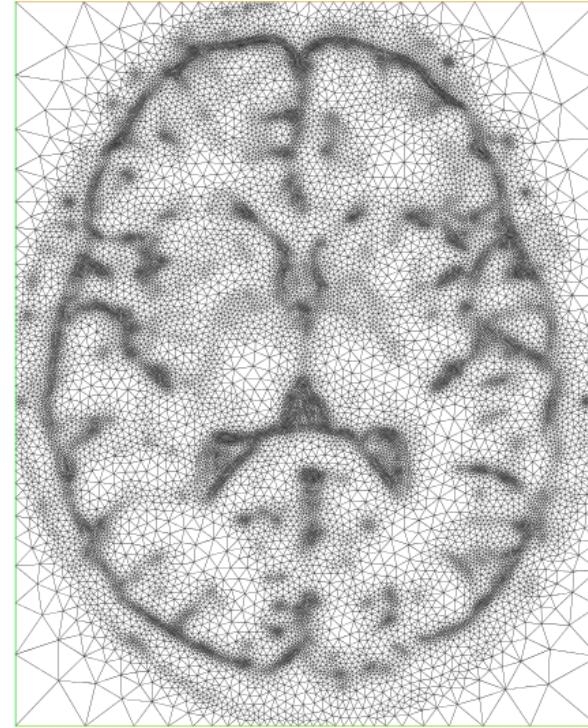


Comparison with isotropic mesh adaptation

39334
triangles

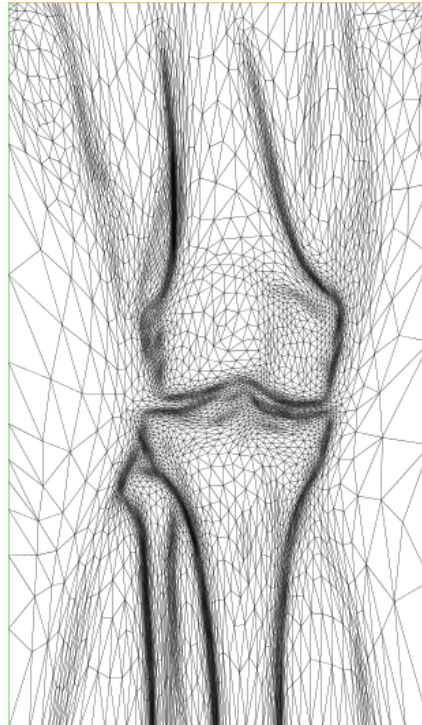


48829
triangles

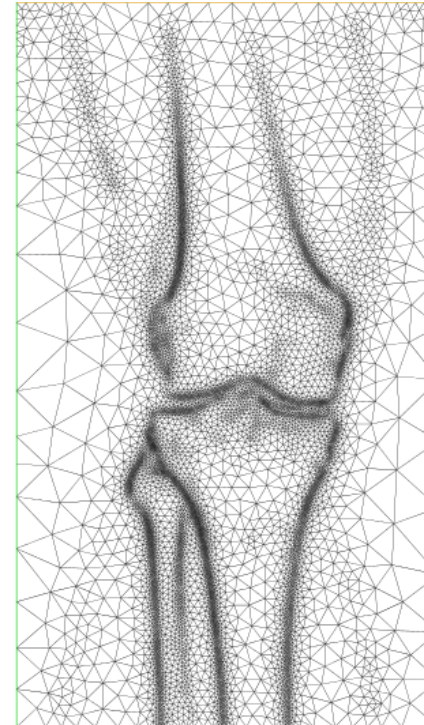


TOL = 70

11175
triangles



19760
triangles



TOL = 60