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CONTINUOUS MESH ADAPTATION MODELS FOR CFD

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(P): Find the design parameter γ_{opt} for which a functional j is minimum.

γ : aircraft shape parameter

j : performances via CFD approximate model

\mathcal{M} : mesh parameterization

$$j(\mathcal{M}, \gamma) = J(\mathcal{M}, \gamma, U(\mathcal{M}, \gamma)) ,$$

$$\Psi_{State}(\mathcal{M}, \gamma, U(\mathcal{M}, \gamma)) = 0 .$$

- minimize the functional with respect to shape γ ,
- **minimize the error on functional with respect to a mesh parameter, \mathcal{M} .**

Adaptation for a functional

- Best ideal mesh density minimizing a continuous model of the error, see [Babuska-Strouboulis](#)
- Adjoint based:
 - . superconvergence for a functional, [Giles](#)
 - . a posteriori error for a functional, [Becker-Rannacher](#)

Mesh optimization strategy

Minimize an error functional \bar{j} with respect to mesh parameter, \mathcal{M} .

$$\text{with } \Psi_{STATE}(\gamma, U) = 0$$

$$\bar{\mathcal{M}} = \text{ArgMin } \bar{\mathbf{J}}(\mathcal{M}, \gamma, U^{exact} - U)$$

with $\Psi_{ERROR}(\mathcal{M}, \gamma, U^{exact} - U) = 0$.

$$\bar{\mathbf{J}}(U^{exact} - U(\mathcal{M})) = |J_{exacte}(U^{exact})' \cdot (U^{exact} - U(\mathcal{M}))|^2$$

Mesh parameterization: Riemannian metric

We modelize a mesh as a **continuous medium**, with an anisotropic property, the **local metric** (*):

$$\mathcal{M}_{x,y} = \mathcal{R}_{\mathcal{M}}^{-1} \begin{pmatrix} (m_{\zeta})^{-2} & 0 \\ 0 & (m_{\theta})^{-2} \end{pmatrix} \mathcal{R}_{\mathcal{M}},$$

(*)(George, Hecht,..., Fortin, Habashi, Vallet,...)

- interpolation error,
- approximation error: elliptic theory, extension to CFD.

P1-Interpolation error

$$\mathcal{E}_{\mathcal{M}} = \int \left(\left| \frac{\partial^2 u}{\partial \xi^2} \right| \cdot m_{\xi}^2 + \left| \frac{\partial^2 u}{\partial \eta^2} \right| \cdot m_{\eta}^2 \right)^2 dx dy$$

where ξ and η are directions of diagonalization of the Hessian of u .

Discontinuous case:

$(u(\xi + \delta, \eta) - 2u(\xi, \eta) + u(\xi - \delta, \eta)) / \delta^2$ bounded in $L^{\frac{1}{2}}$.

$$\min_{\mathcal{M}} \mathcal{E}_{\mathcal{M}}$$

under the constraint $N_{\mathcal{M}} = N$.

Optimal Metric construction.

$$\mathcal{M} = \mathcal{R} \tilde{\Lambda} \mathcal{R}^{-1}, \text{ where } \tilde{\Lambda} = \text{diag}(\tilde{\lambda}_i)$$

$$\text{and } \tilde{\lambda}_i = \min \left(\max \left(\frac{c |\lambda_i|}{\varepsilon}, \frac{1}{h_{max}^2} \right), \frac{1}{h_{min}^2} \right).$$

$$\mathcal{M}_{opt} = \frac{C}{N} \mathcal{R}^{-1} \begin{pmatrix} \left| \frac{\partial^2 u}{\partial \eta^2} \right|^{-5/6} & \left| \frac{\partial^2 u}{\partial \xi^2} \right|^{1/6} & & 0 \\ & 0 & & \left| \frac{\partial^2 u}{\partial \xi^2} \right|^{-5/6} & \left| \frac{\partial^2 u}{\partial \eta^2} \right|^{1/6} \end{pmatrix} \mathcal{R}.$$

avec:

$$C = \int \left(\left| \frac{\partial^2 u}{\partial \xi^2} \right| \cdot \left| \frac{\partial^2 u}{\partial \eta^2} \right| \right)^{\frac{2}{6}} dx dy.$$

$$\text{Erreur minimale: } \mathcal{E}_{opt} = \frac{4C^2}{N^2} \int \left(\left| \frac{\partial^2 u}{\partial \xi^2} \right|^{1/3} \left| \frac{\partial^2 u}{\partial \eta^2} \right|^{1/3} \right) dx dy$$

Anisotropic mesh reconstruction

Purpose: to generate a mesh such that all edges have a length of (or close to) one in the prescribed metric.

For P a vertex $\mathcal{M}(P)$ the metric at P , an edge PX has a local parametrization $PX = P + t\overrightarrow{PX}$. Its average length is:

$$l_{\mathcal{M}}(\overrightarrow{PX}) = \int_0^1 \sqrt{{}^t\overrightarrow{PX} \mathcal{M}(P + t\overrightarrow{PX}) \overrightarrow{PX}} dt. \quad (1)$$

- add nodes in edges longer than 1+...

Supersonic business jet

Final mesh (iteration 9) contains

798,756 vertices,

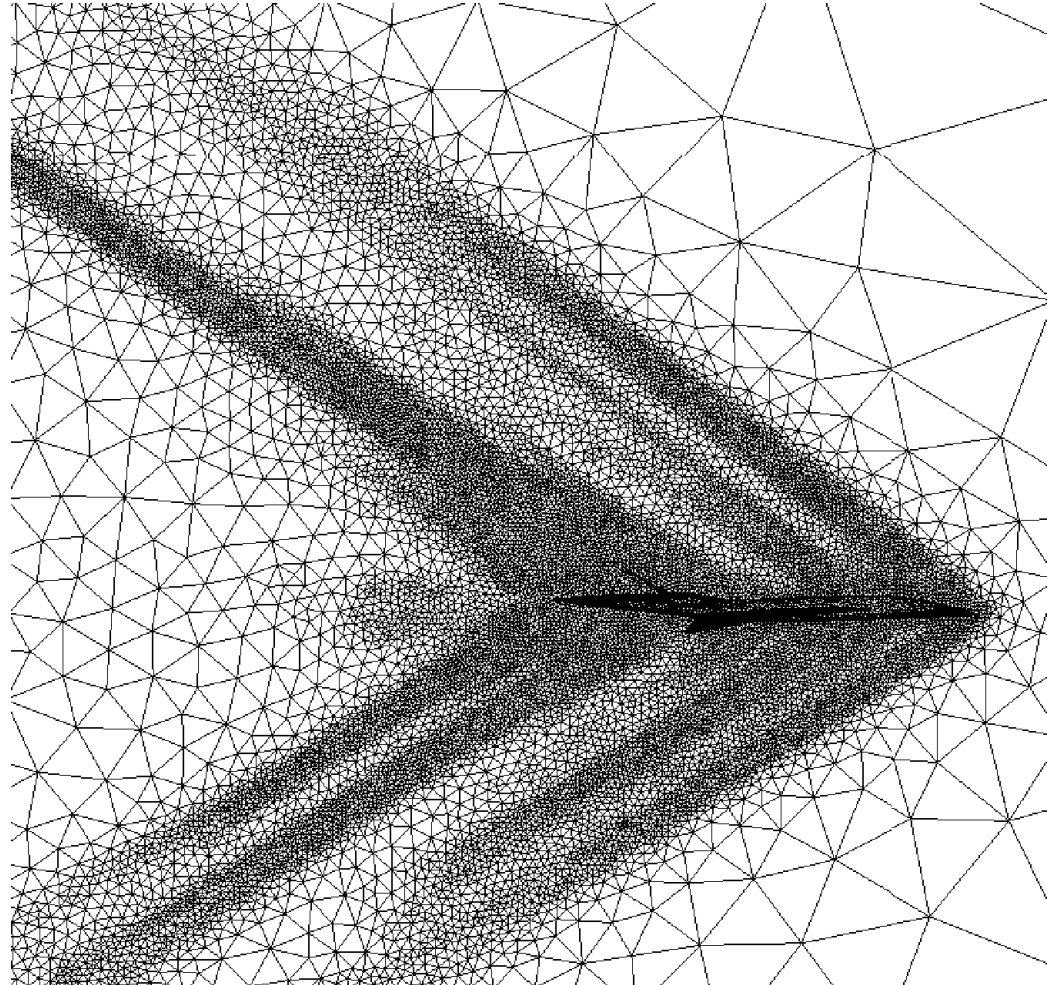
38,492 boundary triangles and

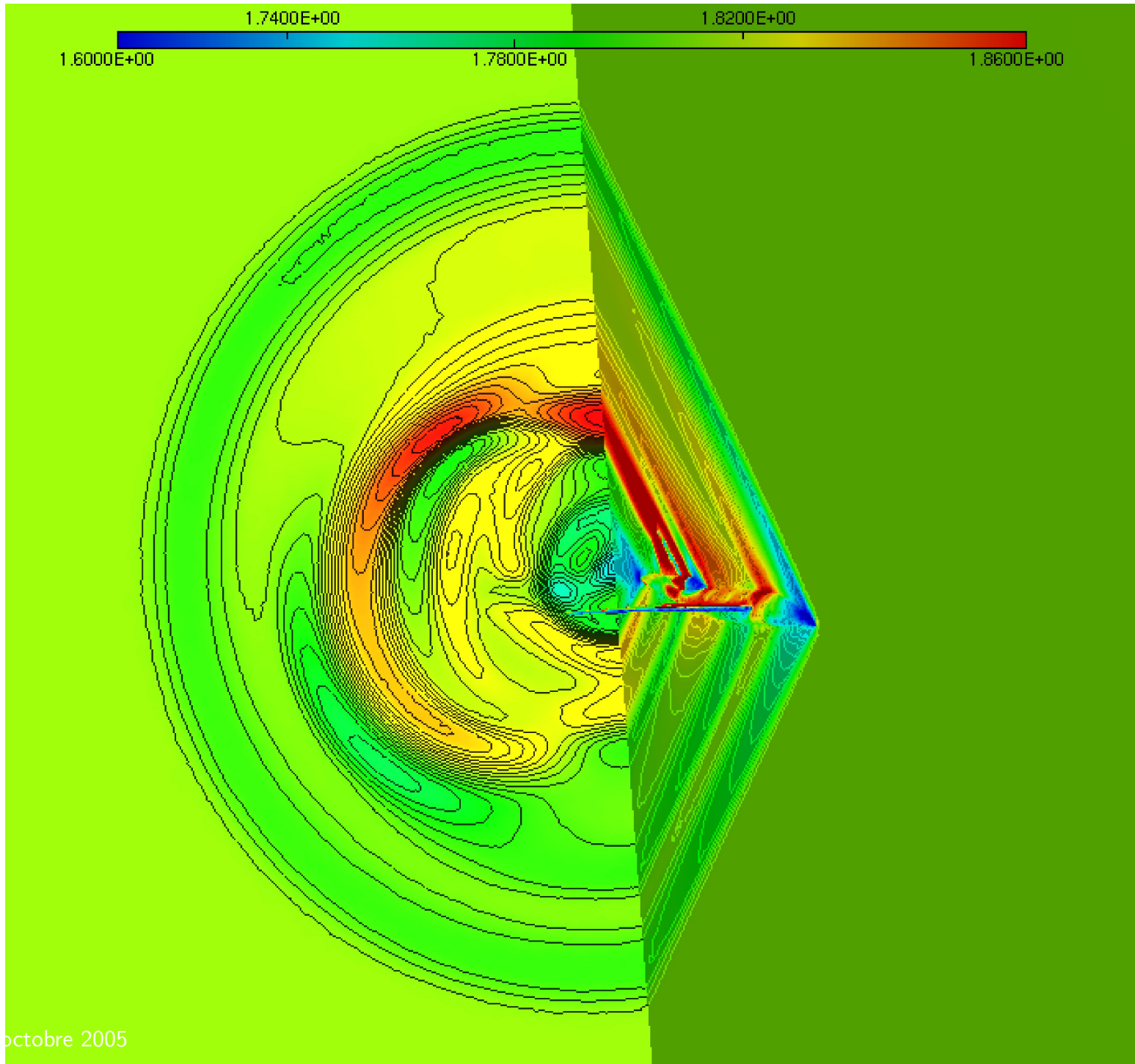
4,714,162 tetrahedra.

Total CPU time 44 hours on a 600MHz workstation with 1Gb of memory.

The remeshing time is 2% of total time time.

Supersonic business jet





Isotropic mesh and asymptotics

How does behave the approximation error?

$m_\zeta^{\frac{1}{2}} = m_\theta^{\frac{1}{2}} = d$: local mesh density

$$\mathcal{M} = d \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}.$$

$$d_h = h^{-1} d \quad \forall h > 0.$$

A First model: Dirichlet model, Galerkin

Babuska-Rheinbolt: local problems + pollution error

Mesh quality and element location effects

A priori: Projection/Aubin-Nitsche

Our option: extract a global smooth behavior from second-order a priori estimates.

Approximation-error splitting

- \mathcal{P}_1 -continuous, vertex-centered approximations,

$$\Pi_{\mathcal{M}} : H^k(\Omega) \rightarrow V_h = \{v, \text{continue, } P_1 \text{ by element } \}$$

$$U^{exact} - U(\mathcal{M}) = U^{exact} - \Pi_{\mathcal{M}}U^{exact} + \Pi_{\mathcal{M}}U^{exact} - U(\mathcal{M})$$

$U^{exact} - \Pi_{\mathcal{M}}U^{exact}$ is interpolation error

$\Pi_{\mathcal{M}}U^{exact} - U(\mathcal{M})$ is the implicit error, solution of a discrete system.

Implicit error : elliptic case

$$\langle \nabla(u), \nabla\phi \rangle = \int f\phi \, dx .$$

$$\langle \nabla(u_h), \nabla\Pi_h\phi \rangle = \int f\phi \, dx .$$

$$\langle \nabla(u_h - \Pi_h u), \nabla\Pi_h\phi \rangle = \langle \nabla(u - \Pi_h u), \nabla\Pi_h\phi \rangle$$

Lemma : We assume that the continuous solution u is in $\mathcal{C}^3(\bar{\Omega})$ and that the continuous mesh size m is in $\mathcal{C}^2(\bar{\Omega})$. Then, for any function ϕ of $\mathcal{D}_3(\Omega)$, we have

$$\begin{aligned} \int_{\Omega} \frac{\partial(u - \Pi_h u)}{\partial x} \frac{\partial \Pi_h \phi}{\partial x} dM &+ \int_{\Omega} \frac{\partial(u - \Pi_h u)}{\partial y} \frac{\partial \Pi_h \phi}{\partial y} dM \\ &= h^2 \int_{\Omega} g'(m) \phi dM + O_{\phi}(h^3) \end{aligned}$$

(2)

$$g'(m) = g'_1(m) + g'_2(m) :$$

$$\begin{aligned} (g'_1(m), \phi) &= -\frac{3}{48} \int_{\Omega} \phi \frac{\partial}{\partial y} \left(m^2 \frac{\partial^3 u}{\partial x \partial y^2} \right) dM \\ &\quad + \frac{1}{48} \int_{\Omega} \phi \frac{\partial}{\partial x} \left(m^2 \frac{\partial^3 u}{\partial x^3} \right) dM \end{aligned}$$

$$\begin{aligned} (g'_2(m), \phi) &= -\frac{1}{4} \int_{\Omega} \phi \frac{\partial}{\partial y} \left(\frac{m^2}{6} \frac{\partial^3 u}{\partial x^2 \partial y} \right) + \phi \frac{\partial^2}{\partial y^2} \left(m^2 \frac{\partial^2 u}{\partial x^2} \right) dM \\ &\quad + \frac{3}{24} \int_{\Omega} \phi \frac{\partial}{\partial y} \left(m^2 \frac{\partial^3 u}{\partial y^3} \right) + \phi \frac{\partial^2}{\partial y^2} \left(m^2 \frac{\partial^2 u}{\partial y^2} \right) dM . \end{aligned}$$

Remark: H^2 pivot space for gradient with respect to m .

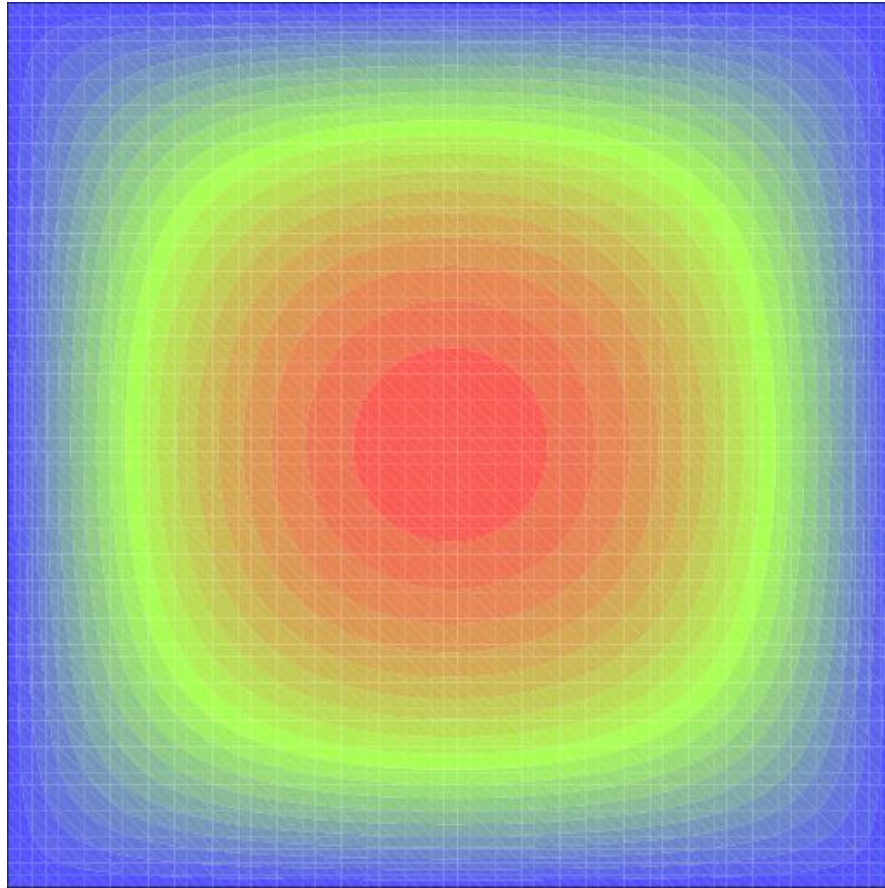
Numerical implementation

Given u and its derivatives, the continuous optimization problem is discretized on a background mesh.

Its adjoint and gradient are obtained by reverse-mode Automated Differentiation with TAPENADE.

Optimization : One-Shot SQP and multi-level preconditioner.

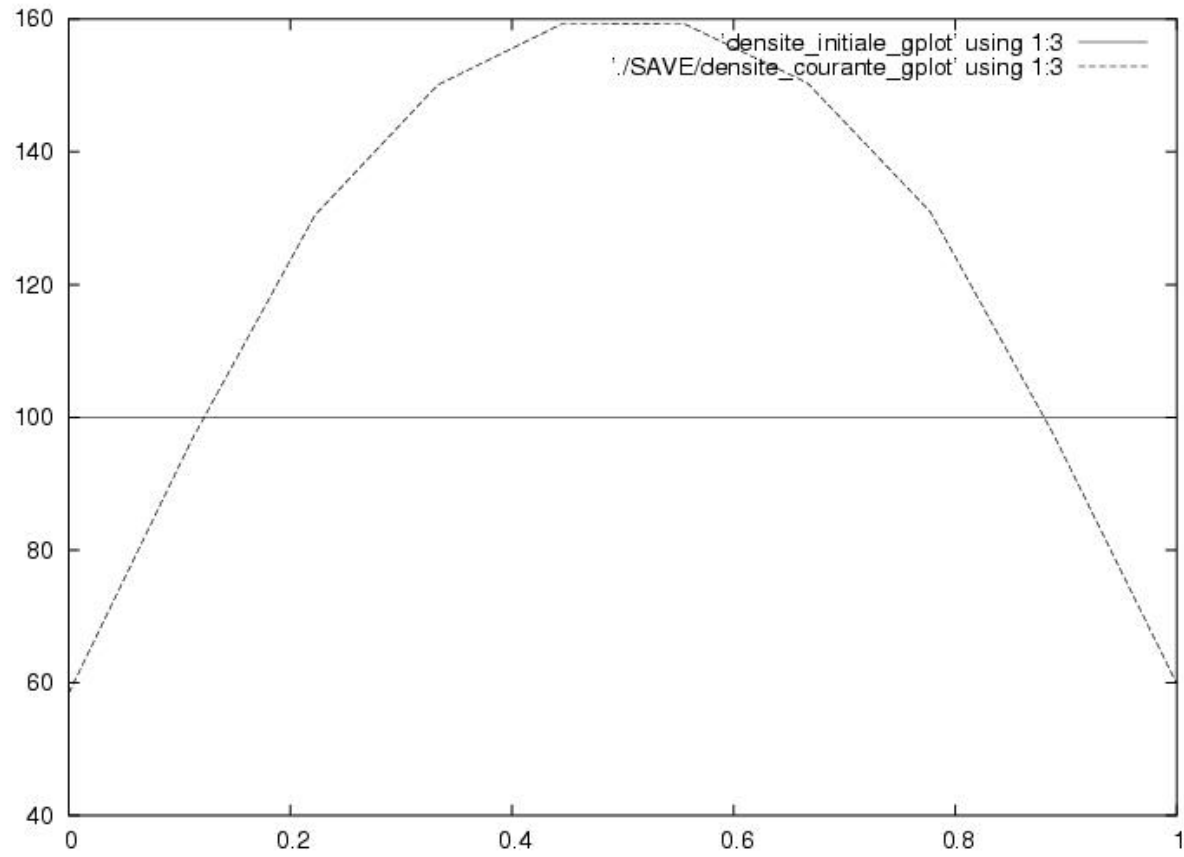
Example: Dirichlet problem, Minimization of L^2 error on whole domain, on left half domain.



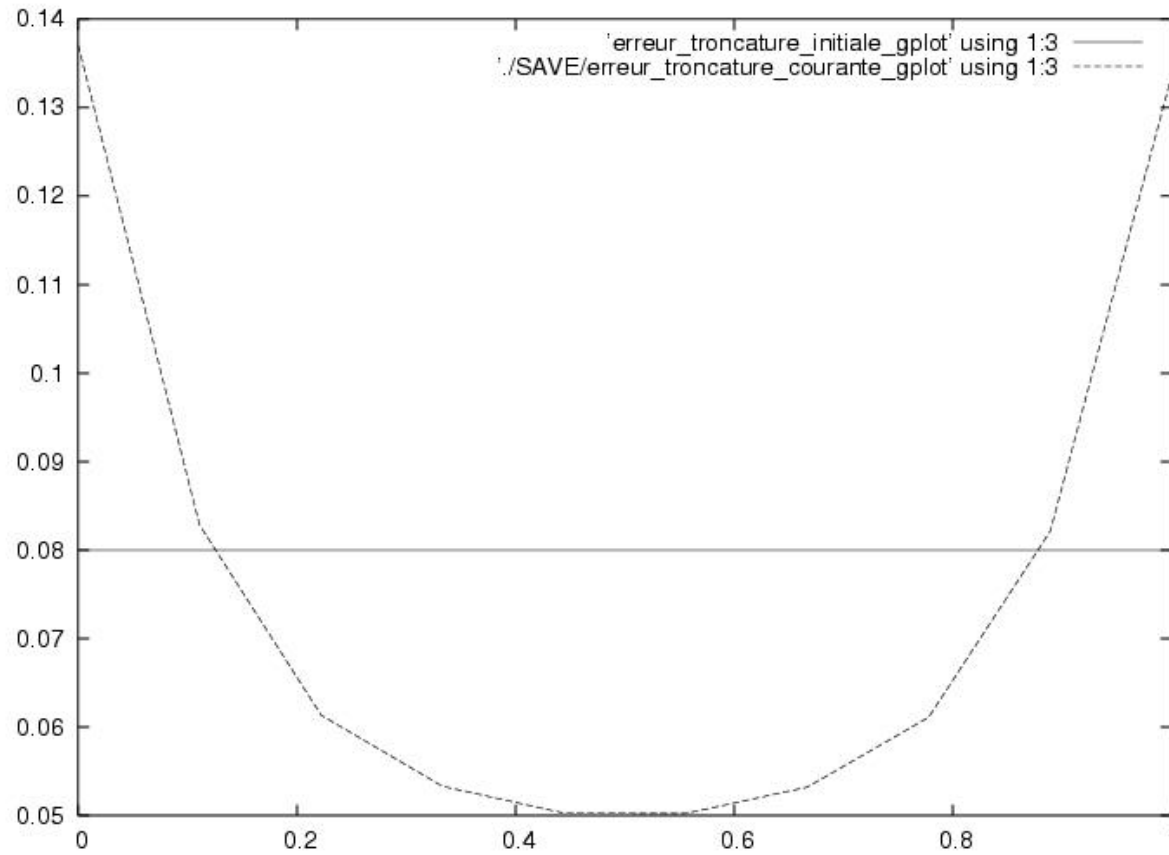
pb_2d

Optimization of mesh for a Dirichlet problem

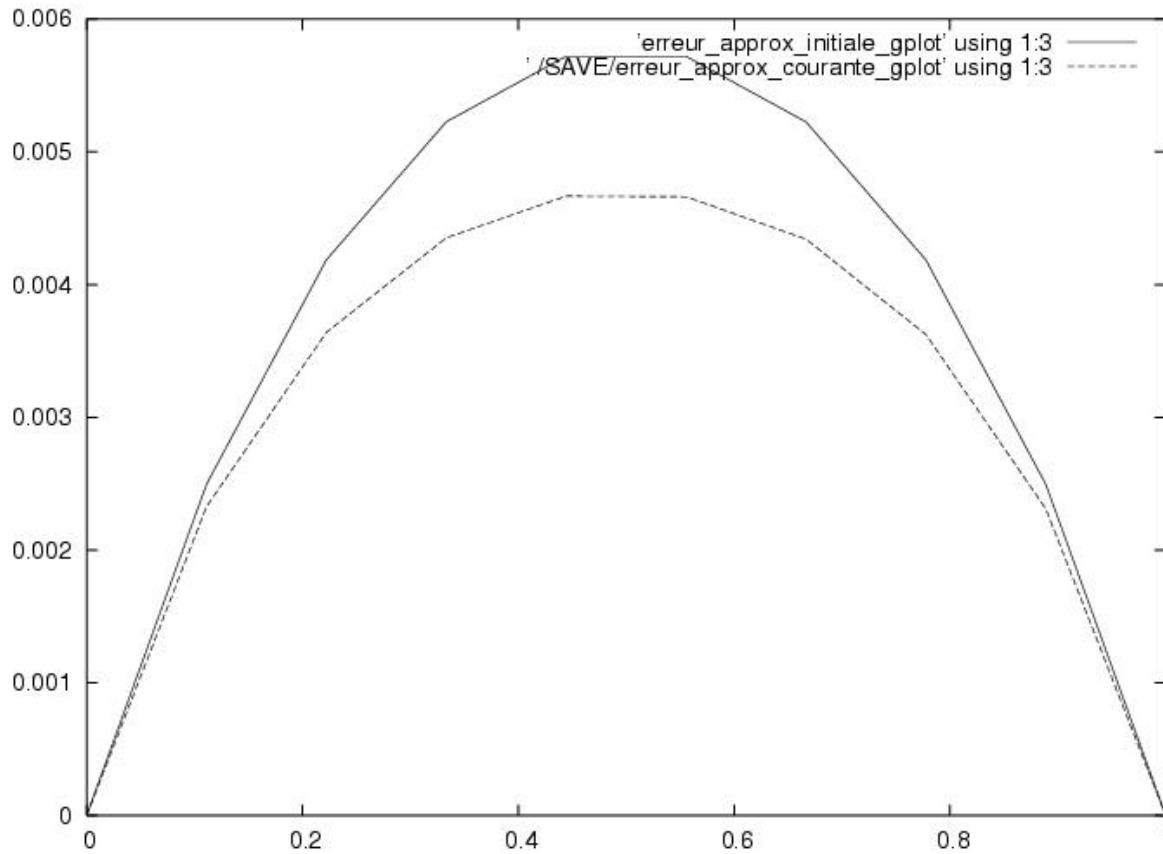
case 1: $\text{Min} \int_0^1 \int_0^1 |u - u_h|^2 dx dy$, case 2: $\text{Min} \int_0^{\frac{1}{2}} (\int_0^1 |u - u_h|^2 dy) dx$,



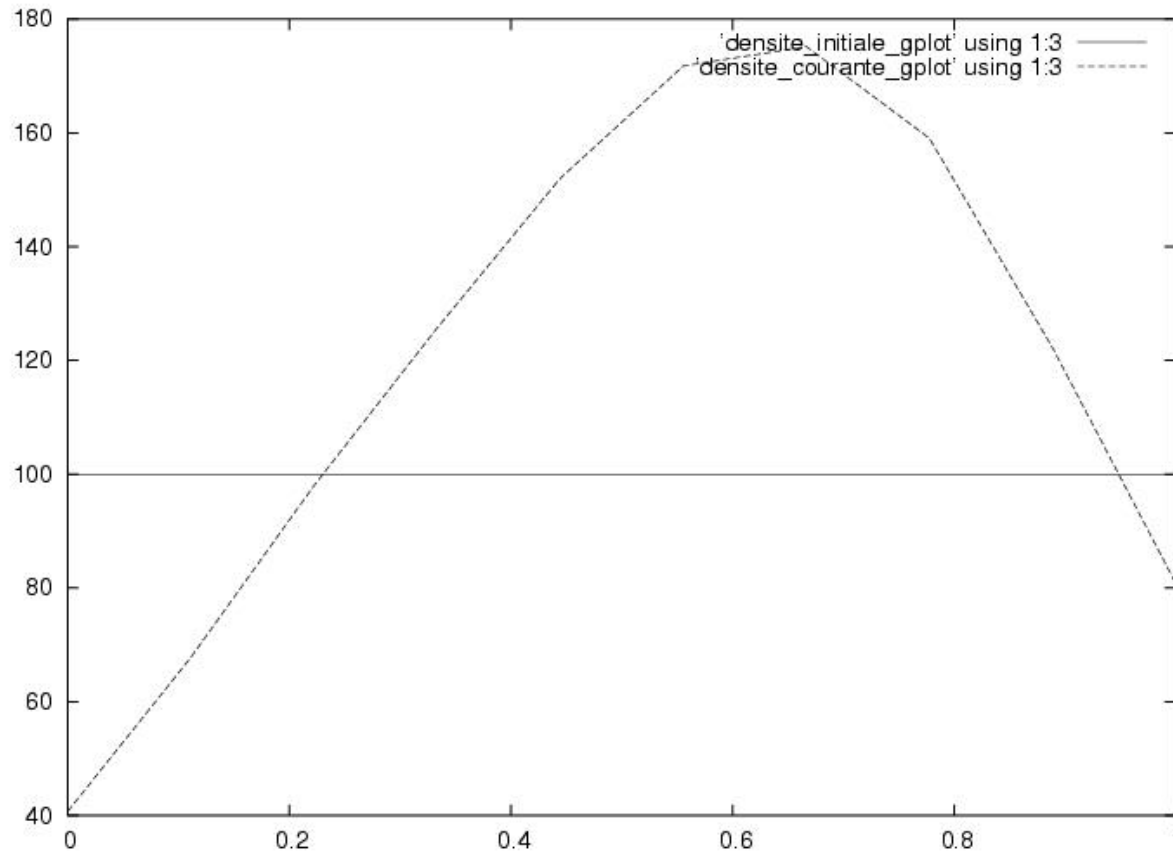
Case 1: Node-density initial(line)/final(dashes)



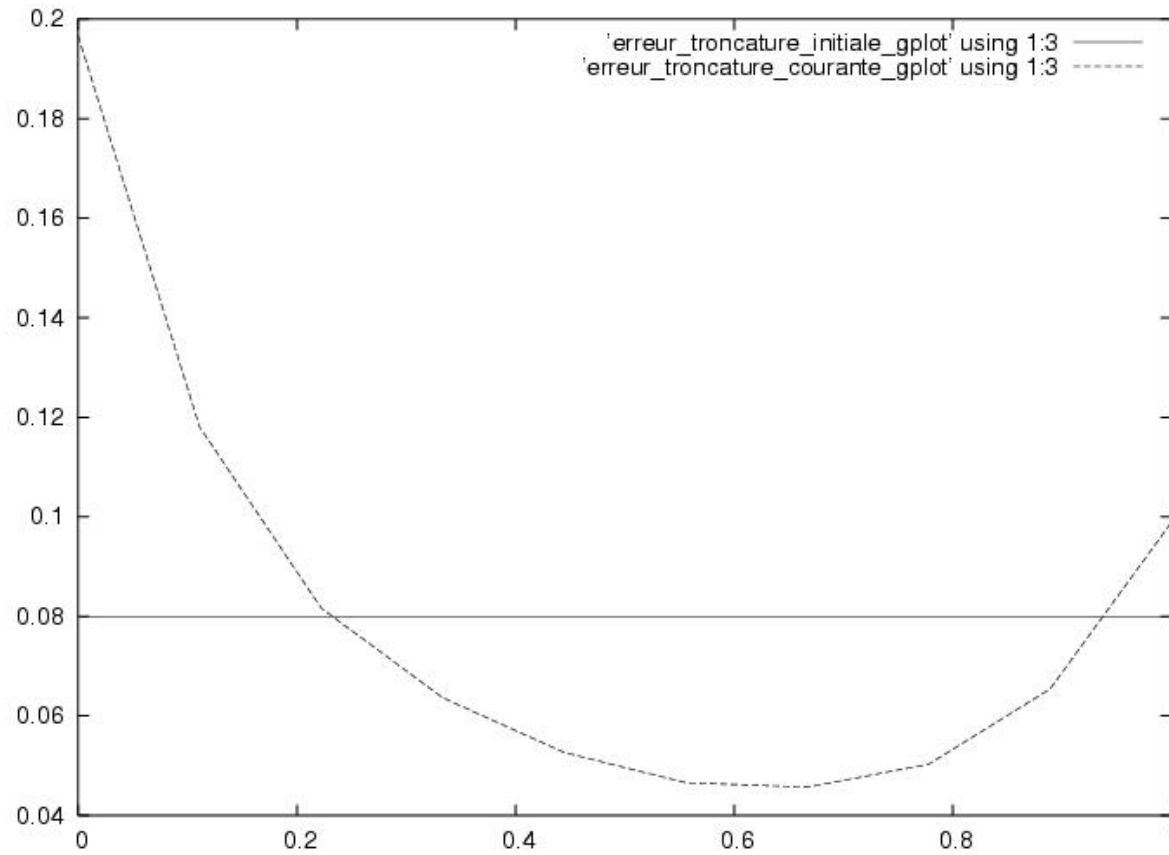
Case 1: Truncation error initial(line)/final(dashes)



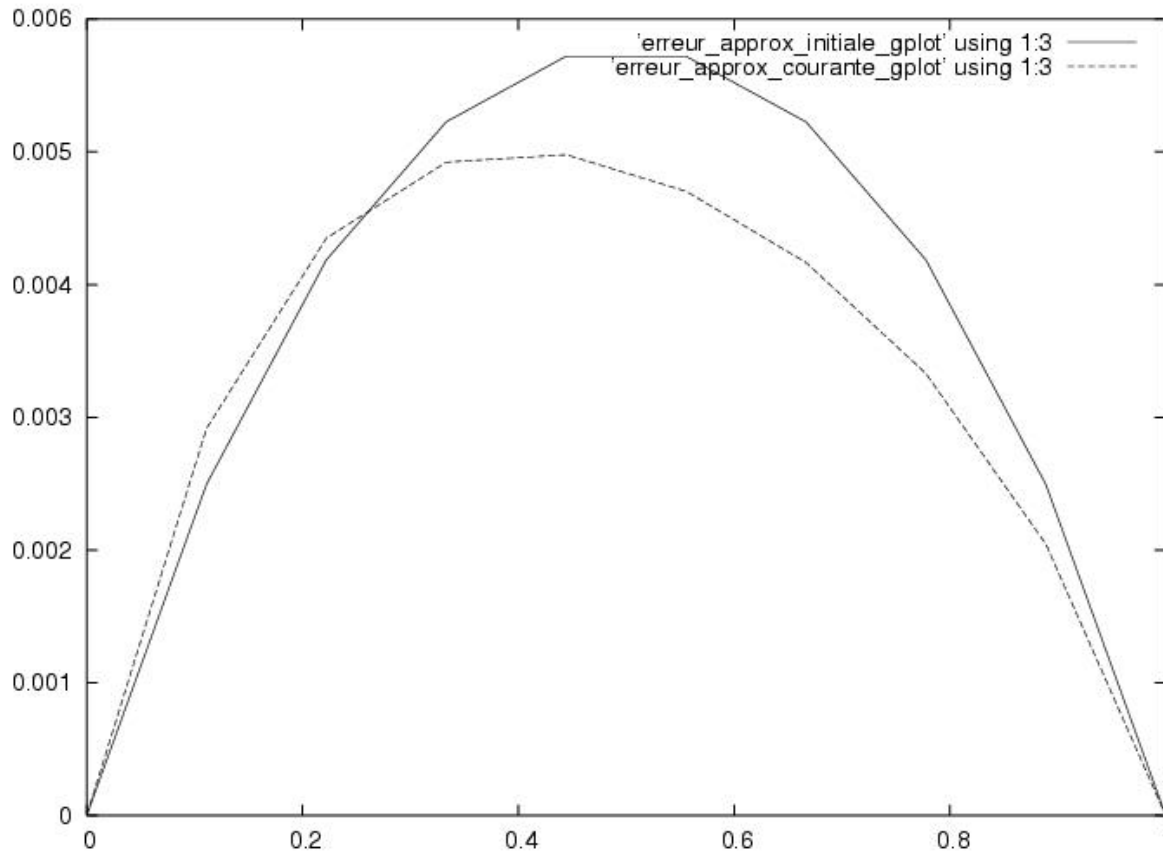
Case 1: Approximation error initial(line)/final(dashes)



Case 2: Node-density initial(line)/final(dashes)



Case 2: truncation error initial(line)/final(dashes)



Case 2: approximation error initial(line)/final(dashes)

Towards the anisotropic case:

Step 1: Pointwise optimization

Given at a point a mesh density $d = (m_\zeta m_\theta)^{-1}$, find **numerically** at each node the:

- optimal stretching direction,
- optimal stretching strength.

Build the reduced error model based on these outputs.

Step 2: constrained global optimization

Minimize the reduced error model with respect to mesh density.

A VARIATIONAL MODEL FOR EULER EQUATIONS

A simplifying assumption: pure Galerkin formulation W and $\phi \in (H^1(\Omega))^5$, V_h is included in V .

$$\int_{\Omega} \phi \nabla \cdot \mathcal{F}(W) d\Omega - \int_{\partial\Omega} \phi \bar{\mathcal{F}}(W) \cdot n d\partial\Omega = 0.$$

$$\int_{\Omega} \phi_h \nabla \cdot \mathcal{F}_h(W_h) d\Omega - \int_{\partial\Omega} \phi_h \bar{\mathcal{F}}_h(W_h) \cdot n d\partial\Omega = 0$$

where $\mathcal{F}_h(W_h)$ is the interpolate of \mathcal{F} , i.e. $\mathcal{F}_h(W_h) = \Pi_h \mathcal{F}(W_h)$, and same for $\bar{\mathcal{F}}_h(W_h)$.

Replace ϕ by $\phi_h = \Pi_h \phi$:

$$\int_{\Omega} \phi_h \nabla \cdot \mathcal{F}_h(W_h) d\Omega - \int_{\partial\Omega} \phi_h \bar{\mathcal{F}}_h(W_h) \cdot n d\partial\Omega =$$

$$\int_{\Omega} \phi_h \nabla \cdot \mathcal{F}(W) d\Omega - \int_{\partial\Omega} \phi_h \bar{\mathcal{F}}(W) \cdot n d\partial\Omega$$

Boundary data are involved inside $\bar{\mathcal{F}}(W)$ and $\bar{\mathcal{F}}_h(W)$ Assumption: these terms can be split in W -dependant terms, denoted respectively by $\bar{\mathcal{F}}^{out}(W)$ and $\bar{\mathcal{F}}_h^{out}(W_h)$, and constant terms, denoted $\bar{\mathcal{F}}^{in}$ and $\bar{\mathcal{F}}_h^{in}$.

Therefore:

$$\int_{\Omega} \phi_h \nabla \cdot \mathcal{F}_h(W_h) d\Omega - \int_{\partial\Omega} \phi_h \bar{\mathcal{F}}_h^{out}(W_h) \cdot n d\partial\Omega =$$
$$\int_{\Omega} \phi_h \nabla \cdot \mathcal{F}(W) d\Omega - \int_{\partial\Omega} \phi_h \bar{\mathcal{F}}^{out}(W) \cdot n d\partial\Omega$$

Error estimate

Assumption: both W and ϕ are several time continuously differentiable.

$$\begin{aligned} & \int_{\Omega} \phi_h \nabla \cdot (\mathcal{F}_h(W_h) - \Pi_h \mathcal{F}(W)) d\Omega - \\ & \int_{\partial\Omega} \phi_h (\bar{\mathcal{F}}_h^{out}(W_h) - \Pi_h \bar{\mathcal{F}}^{out}(W)) \cdot n d\partial\Omega = \\ & \int_{\Omega} \phi_h \nabla \cdot (\mathcal{F}(W) - \Pi_h \mathcal{F}(W)) d\Omega - \\ & \int_{\partial\Omega} \phi_h (\bar{\mathcal{F}}^{out}(W) - \Pi_h \bar{\mathcal{F}}^{out}(W)) \cdot n d\partial\Omega. \end{aligned}$$

The left-hand side will be inverted and the right-hand side will be expanded to get the error estimate.

Interpolation errors

We recall that $\Pi_h \mathcal{F}(W) = \mathcal{F}_h(W)$. The left-hand side writes:

$$\begin{aligned} LHS &= \int_{\Omega} \phi_h \nabla \cdot (\mathcal{F}_h(W_h) - \mathcal{F}_h(W)) d\Omega - \\ &\int_{\partial\Omega} \phi_h (\bar{\mathcal{F}}_h^{out}(W_h) - \bar{\mathcal{F}}_h^{out}(W)) \cdot n d\partial\Omega. \end{aligned}$$

We linearize it as follows:

$$LHS = \int_{\Omega} \phi_h \nabla \cdot (\Pi_h \frac{\partial \mathcal{F}}{\partial W} (W_h - W)) d\Omega - \int_{\partial\Omega} \phi_h (\Pi_h \frac{\partial \bar{\mathcal{F}}_h^{out}}{\partial W} (W_h - W)) \cdot n d\partial\Omega.$$

Where the derivatives $\frac{\partial \mathcal{F}}{\partial W}$ and $\frac{\partial \bar{\mathcal{F}}_h^{out}}{\partial W}$ are evaluated from vertex values of W .

$$LHS = \mathcal{A}_h(W)(W_h - \Pi_h W)$$

$$LHS = \mathcal{A}_h(W)(W_h - \Pi_h W)$$

Assumption: the Jacobian $\mathcal{A}_h(W)$ of the discretized Euler system is invertible. The **implicit error** $W_h - \Pi_h W$ is the unique solution of:

$$W_h - \Pi_h W = (\mathcal{A}_h(W))^{-1} RHS .$$

In the right-hand side:

$$\begin{aligned}
 RHS &= \int_{\Omega} \phi_h \nabla \cdot (\mathcal{F}(W) - \Pi_h \mathcal{F}(W)) d\Omega - \\
 &\int_{\partial\Omega} \phi_h (\bar{\mathcal{F}}^{out}(W) - \Pi_h \bar{\mathcal{F}}^{out}(W)) \cdot n d\partial\Omega
 \end{aligned}$$

we recall that $\phi_h = \Pi_h \phi$ and we add and subtract a ϕ term:

$$RHS = RHS_1 + RHS_2$$

with:

$$\begin{aligned}
 RHS_1 &= \int_{\Omega} (\Pi_h \phi - \phi) \nabla \cdot (\mathcal{F}(W) - \Pi_h \mathcal{F}(W)) d\Omega - \\
 &\int_{\partial\Omega} (\Pi_h \phi - \phi) (\bar{\mathcal{F}}^{out}(W) - \Pi_h \bar{\mathcal{F}}^{out}(W)) \cdot n d\partial\Omega
 \end{aligned}$$

Assuming smoothness of ϕ and $\mathcal{F}(W)$, we deduce that on Ω , interpolation errors are of order two and their gradients are of order one, same on boundary, and RHS_1 is thus of order three.

$$RHS_1 \leq \text{const.}h^3$$

The second term writes:

$$RHS_2 = \int_{\Omega} \phi \nabla \cdot (\mathcal{F}(W) - \Pi_h \mathcal{F}(W)) d\Omega - \int_{\partial\Omega} \phi (\bar{\mathcal{F}}^{out}(W) - \Pi_h \bar{\mathcal{F}}^{out}(W)) \cdot n d\partial\Omega$$

and we transform it as follows:

$$\begin{aligned} RHS_2 &= - \int_{\Omega} (\nabla \phi) \cdot (\mathcal{F}(W) - \Pi_h \mathcal{F}(W)) d\Omega \\ &\quad + \int_{\partial\Omega} \phi (\mathcal{F}(W) - \Pi_h \mathcal{F}(W)) \cdot n d\partial\Omega - \\ &\quad \int_{\partial\Omega} \phi (\bar{\mathcal{F}}^{out}(W) - \Pi_h \bar{\mathcal{F}}^{out}(W)) \cdot n d\partial\Omega \end{aligned}$$

Same as for elliptic (easier):

$$RHS_2 \leq \text{const.} h^2.$$

$$RHS_2 = h^2 (G(W,d), \phi) + R, \quad R = o(h^2).$$

Provisional conclusion

The above study shows that the implicit error $W_h - \Pi_h W$ is a linear function of the interpolation error $W - \Pi_h W$.

A first option consists in reducing the interpolation error. This option is studied in Sections 4 and 5.

In Section 6 we shall define how to minimize with the implicit error by introducing an adjoint.

Implicit error reduction, general case

$$\frac{\partial \Psi}{\partial W}(W, d) Y = G(W, d),$$

W is the unknown, d mesh density, Y implicit error.

$$\begin{aligned} \frac{\partial \Psi}{\partial W}(W, d) Y - G(W, d) &= 0 \\ \left(\frac{\partial \Psi}{\partial W}(W, d) \right)^* \Pi - \frac{\partial \bar{J}}{\partial Y}(Y, d) &= 0 \\ \bar{j}' &= - \left(\frac{\partial \left(\frac{\partial \Psi}{\partial W} Y - G \right)}{\partial d}(W, d) \right)^* \Pi + \frac{\partial \bar{J}}{\partial d}(Y, d) = 0 \end{aligned}$$

which can be solved as for elliptic case.

CONCLUSION

This talk proposes rather sophisticated algorithms involving:
flow solver,
local error evaluation,
sensitivity analysis by Automated Differentiation,
adjoint and gradient based optimization,
controlled generation of next mesh,
interpolation of solution,
till a convergence of a global loop is attained.

These sophisticated algorithms will not bring only robustness, but also efficiency.