

# *A posteriori* error analysis of finite element approximation of quasi-Newtonian flows

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supported by INDAM: Istituto Nazionale di Alta Matematica *Francesco Severi*

Joint work with Endre Süli  
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Multiscale Problems: Modeling, Adaptive Discretization,  
Stabilization, Solvers,  
Cortona, September 18-22, 2006



# Targets

Our target is to obtain a **reliable** finite element solution of quasi-Newtonian steady incompressible flows in an **efficient** way. To reach this target we need:

- computable quantities that tell us where the error in our numerical solution is large: **upper bound of the error**;
- computable quantities that tell us where the error in our numerical solution is small: **lower bound of the error**;
- an adaptive method that automatically refine the mesh where the error in the numerical solution is large and coarsen the mesh where the error is small.

## Remark

- *refining*  $\leadsto$  *reliability of the numerical solution*;
- *coarsening*  $\leadsto$  *efficiency on the numerical method*.



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# Outline

- Model problem;
- Weak formulation of the continuous problem and discrete weak formulation;
- Quasi-norm, **uniform monotonicity** and **local Hölder continuity**;
- Upper bounds of the error:
  - Upper bound in terms of residual functionals (weak residuals);
  - Upper bound in terms of computable residuals (strong residuals);
- Lower bounds of the error:
  - Lower bound in terms of residual functionals (weak residuals);
  - Lower bound in terms of computable residuals (strong residuals);
- Numerical experiments:
  - Adaptive method based on our *a posteriori* error estimates;
  - Power-law problems;
  - Carreau law problems.



# Model problem

## Continuous strong form

$$\begin{aligned} -\nabla \cdot \boldsymbol{\sigma} &= \mathbf{f}, & \text{in } \Omega, \\ \nabla \cdot \mathbf{u} &= \mathbf{0}, & \text{in } \Omega, \\ \mathbf{u} &= \mathbf{0}, & \text{on } \partial\Omega = \Gamma, \end{aligned}$$

where

$$\boldsymbol{\sigma}(\mathbf{x}, \mathbf{u}) = k(\mathbf{x}, |\mathbf{e}(\mathbf{u})|) \mathbf{e}(\mathbf{u}) - p \mathbf{I}$$

is the stress tensor and

$$e(\mathbf{u})_{ij} = \frac{1}{2} \left( \frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right), \quad i, j = 1, \dots, d.$$

are the components of the strain tensor  $\mathbf{e}(\mathbf{u}) \in \mathbb{R}_{\text{symm}}^{d \times d}$ .

- $\Omega$  bounded Lipschitz domain,  $|\Omega| = 1$ ;
- $k(\mathbf{x}, |\mathbf{e}(\mathbf{u})|)$  is the viscosity coefficient.



# Viscosity coefficient $k(x, |e(\mathbf{u})|)$

We want to deal with:

- *pseudoplastic fluids: increasing viscosity when  $|e(\mathbf{u})| \rightarrow \mathbf{0}$ ;*
- *dilatant fluids: decreasing viscosity when  $|e(\mathbf{u})| \rightarrow \mathbf{0}$ ;*
- *degenerate constitutive laws: the viscosity function can tend to  $\mathbf{0}$  or to  $\infty$ ;*
- *non-degenerate constitutive laws: the viscosity function is bounded away from  $\mathbf{0}$  and from  $\infty$ .*







# Examples

- (a) the classical linear Stokes equations which govern the stationary flow of a viscous incompressible Newtonian fluid:

$$k(x, |e(\mathbf{u})|) = 2\mu,$$

corresponds to  $r = 2$ ;

- (b) the power-law model with

$$k(x, |e(\mathbf{u})|) = 2\mu |e(\mathbf{u})|^{r-2}$$

corresponds to  $\alpha = 1$ ;

- (c) the Carreau law

$$k(x, |e(\mathbf{u})|) = k_\infty + (k_0 - k_\infty)(1 + \lambda |e(\mathbf{u})|^2)^{(\theta-2)/2},$$

with  $k_0 > k_\infty \geq 0$ ,  $\lambda > 0$ ,  $\theta \in (1, \infty)$ ;

- if  $k_\infty = 0$ , then  $\alpha = 0$  with  $r = \theta$ ;
- if  $k_\infty > 0$ , then  $r = 2$  and  $\theta \in (1, 2]$ , (pseudoplastic behaviour).



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# Continuous weak formulation

Find  $\mathbf{u} \in \mathbf{V}$  and  $p \in \mathbf{Q}$  such that

$$\begin{aligned} a(\mathbf{u}, \mathbf{v}) + b(p, \mathbf{v}) &= (\mathbf{f}, \mathbf{v}) & \forall \mathbf{v} \in \mathbf{V}, \\ b(q, \mathbf{u}) &= 0 & \forall q \in \mathbf{Q}, \end{aligned}$$

where

$$a(\mathbf{u}, \mathbf{v}) = \int_{\Omega} k(x, |e(\mathbf{u})|) e(\mathbf{u}) : e(\mathbf{v}) \, d\Omega, \quad b(q, \mathbf{v}) = - \int_{\Omega} (\nabla \cdot \mathbf{v}) q \, d\Omega,$$

$\mathbf{V} = [\mathbf{W}_0^{1,r}(\Omega)]^d$ , with the norm  $\|\mathbf{v}\|_{\mathbf{V}} = \|e(\mathbf{v})\|_{L^r(\Omega)}$ ,

$\mathbf{Q} = L_0^{r'}(\Omega) = L^{r'}(\Omega)/\mathbb{R}$ , with the norm  $\|q\|_{\mathbf{Q}} = \inf_{c \in \mathbb{R}} \|q + c\|_{L^{r'}(\Omega)}$ ,

and the bilinear form  $b(\cdot, \cdot)$  satisfies the following **inf-sup condition**:

$$\inf_{q \in \mathbf{Q}} \sup_{\mathbf{v} \in \mathbf{V}} \frac{b(q, \mathbf{v})}{\|q\|_{\mathbf{Q}} \|\mathbf{v}\|_{\mathbf{V}}} \geq c_0 \quad \forall q \in \mathbf{Q}.$$

Let  $\mathbf{V}'$  and  $\mathbf{Q}'$  denote the dual spaces of  $\mathbf{V}$  and  $\mathbf{Q}$ .



# Discrete weak formulation

Find  $\mathbf{u}_h \in \mathbf{V}_h$  and  $p_h \in \mathbf{Q}_h$  such that

$$\begin{aligned} a(\mathbf{u}_h, \mathbf{v}_h) + b(p_h, \mathbf{v}_h) &= (\mathbf{f}, \mathbf{v}_h) & \forall \mathbf{v}_h \in \mathbf{V}_h, \\ b(q_h, \mathbf{u}_h) &= \mathbf{0} & \forall q_h \in \mathbf{Q}_h. \end{aligned}$$

- $\mathcal{T}_h$  shape-regular partition of  $\Omega$ ;
- $\mathbf{V}_h \subset \mathbf{V}$  continuous velocity F.E. space defined on  $\mathcal{T}_h$ ;
- $\mathbf{Q}_h \subset \mathbf{Q}$  continuous/discontinuous pressure F.E. space on  $\mathcal{T}_h$ ;
- the pair  $(\mathbf{V}_h, \mathbf{Q}_h)$  satisfies the following inf-sup condition:

$$\inf_{q_h \in \mathbf{Q}_h} \sup_{\mathbf{v}_h \in \mathbf{V}_h} \frac{b(q_h, \mathbf{v}_h)}{\|q_h\|_{\mathbf{Q}} \|\mathbf{v}_h\|_{\mathbf{V}}} \geq c'_0, \quad c'_0 \text{ independent of } h.$$



# Quasi-norm in Sobolev space

To derive upper and lower bounds in the general case of a viscosity coefficient satisfying only the general assumptions described before we need to introduce a new family of tools.

For  $\alpha \in [0, 1]$  and  $t \in (0, \infty)$ , let be

$$\Xi_\alpha(r) = r^\alpha(1+r)^{1-\alpha},$$

for any  $\alpha \in [0, 1]$ ,  $r \mapsto \Xi_\alpha(r)$  is a strictly *monotonic increasing* and *concave* function of  $r \in (0, \infty)$ .

▶ Assumptions on  $k(x, t)$



# Quasi-norm in Sobolev space

## Assumption

We assume that  $k \in C(\bar{\Omega} \times (0, \infty))$  and that, given  $r \in (1, \infty)$ , there exist constants  $\alpha \in [0, 1]$  and  $\varepsilon, K_1, K_2 > 0$  such that, for all  $x \in \bar{\Omega}$ ,

$$(A1) \quad K_1[\Xi_\alpha(t+s)]^{r-2}(t-s) \leq k(x,t)t - k(x,s)s, \quad \forall t \geq s > 0,$$

$$(A2) \quad k(x,t) \leq K_2[\Xi_\alpha(t)]^{r-2} \quad \forall t > 0, \text{ and}$$

$$|k(x,t)t - k(x,s)s| \leq K_2[\Xi_\alpha(t+s)]^{r-2}|t-s|$$

for all  $s, t > 0$  satisfying  $|(s/t) - 1| \leq \varepsilon$ .

► Monot&Cont

## Definition

For all  $\mathbf{v}, \mathbf{w} \in [\mathbf{W}^{1,r}(\Omega)]^d$  with  $1 < r < \infty$ , let us define the following quasi-norm:

$$|\mathbf{v}|_{(\mathbf{w}, r, \alpha)}^2 = \int_{\Omega} [\Xi_\alpha(|e(\mathbf{v})| + |e(\mathbf{w})|)]^{r-2} |e(\mathbf{v})|^2 \, d\Omega.$$

[Barret-Liu 1993, Barrett-Liu 1994]





# Quasi-norms in Sobolev space

## Proposition (part I)

Suppose that  $r \in (1, \infty)$ ,  $\alpha \in [0, 1]$  and  $\mathbf{w} \in [\mathbf{W}^{1,r}(\Omega)]^d$ ;

- (i)  $\forall \mathbf{v} \in [\mathbf{W}^{1,r}(\Omega)]^d$ ,  $|\mathbf{v}|_{(\mathbf{w},r,\alpha)} \geq 0$ ;
- (ii)  $\forall \mathbf{v} \in [\mathbf{W}^{1,r}(\Omega)]^d$ ,  $|\mathbf{v}|_{(\mathbf{w},r,\alpha)} = 0$  iff  $\mathbf{v} = \mathbf{0}$ ;
- (iii) (Quasi-triangle-inequality): there exists a constant  $C = C(r)$  such that

$$|\mathbf{v}_1 + \mathbf{v}_2|_{(\mathbf{w},r,\alpha)} \leq C (|\mathbf{v}_1|_{(\mathbf{w},r,\alpha)} + |\mathbf{v}_2|_{(\mathbf{w},r,\alpha)})$$

for all  $\mathbf{v}_1, \mathbf{v}_2 \in [\mathbf{W}^{1,r}(\Omega)]^d$ ;



# Quasi-norms in Sobolev space

## Proposition (Part II)

(iv) For  $1 < r \leq 2$ ,

$$|\mathbf{v}|_{(\mathbf{w}, r, \alpha)}^{2/r} \leq \|e(\mathbf{v})\|_{L^r(\Omega)}$$

$$\|e(\mathbf{v})\|_{L^r(\Omega)} \leq [\Xi_\alpha(\|e(\mathbf{v})\|_{L^r(\Omega)} + \|e(\mathbf{w})\|_{L^r(\Omega))}]^{(2-r)/2} |\mathbf{v}|_{(\mathbf{w}, r, \alpha)}$$

for all  $\mathbf{v} \in [\mathbf{W}^{1,r}(\Omega)]^d$ .

For  $2 \leq r < \infty$ ,

$$\|e(\mathbf{v})\|_{L^r(\Omega)}^{r/2} \leq |\mathbf{v}|_{(\mathbf{w}, r, \alpha)}$$

$$|\mathbf{v}|_{(\mathbf{w}, r, \alpha)} \leq [\Xi_\alpha(\|e(\mathbf{v})\|_{L^r(\Omega)} + \|e(\mathbf{w})\|_{L^r(\Omega))}]^{(r-2)/2} \|e(\mathbf{v})\|_{L^r(\Omega)}$$

for all  $\mathbf{v} \in [\mathbf{W}^{1,r}(\Omega)]^d$ .



# Quasi-norms in Sobolev space

## Lemma

Suppose that  $r \in (1, \infty)$  and define the constants  $C_2 = 2^{-|r-2|}K_1$  and  $C_3 = 2^{|r-2|/\max\{2, r'\}}K_2$ ; then, for  $i = 1, 2$ , and all  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{w}$  in  $\mathbf{V}$ ,

$$C_2 |\mathbf{v}_1 - \mathbf{v}_2|_{(\mathbf{v}_i, r, \alpha)}^2 \leq a(\mathbf{v}_1, \mathbf{v}_1 - \mathbf{v}_2) - a(\mathbf{v}_2, \mathbf{v}_1 - \mathbf{v}_2),$$

$$|a(\mathbf{v}_1, \mathbf{w}) - a(\mathbf{v}_2, \mathbf{w})| \leq C_3 |\mathbf{v}_1 - \mathbf{v}_2|_{(\mathbf{v}_i, r, \alpha)}^{\min\{1, \frac{2}{r'}\}} \|\mathbf{w}\|_{\mathbf{V}} \\ \times [\Xi_\alpha (\|\mathbf{v}_1\|_{\mathbf{V}} + \|\mathbf{v}_2\|_{\mathbf{V}})]^{\max\{0, \frac{r-2}{2}\}}.$$

► Assumptions on  $k(x, t)$

We are going to use these inequalities with  $\mathbf{v}_1 = \mathbf{u}$  and  $\mathbf{v}_2 = \mathbf{u}_h$ .



# Quasi-norms in Sobolev space

## Lemma

Let  $(\mathbf{u}, p) \in \mathbf{V} \times \mathbf{Q}$  and  $(\mathbf{u}_h, p_h) \in \mathbf{V}_h \times \mathbf{Q}_h$  denote the solutions to the continuous and discrete problems, respectively, and let  $r \in (1, \infty)$ ; then

$$\begin{aligned} \|\mathbf{u}\|_{\mathbf{V}} &\leq G^{-1}\left(\frac{1}{K_1} \|\mathbf{f}\|_{\mathbf{V}'}\right), \\ \|\mathbf{u}_h\|_{\mathbf{V}} &\leq G^{-1}\left(\frac{1}{K_1} \|\mathbf{f}\|_{\mathbf{V}'}\right), \\ \|p\|_{\mathbf{Q}} &\leq \frac{1}{c_0} \left( \|\mathbf{f}\|_{\mathbf{V}'} + K_2(H \circ G^{-1})\left(\frac{1}{K_1} \|\mathbf{f}\|_{\mathbf{V}'}\right) \right), \\ \|p_h\|_{\mathbf{Q}} &\leq \frac{1}{c'_0} \left( \|\mathbf{f}\|_{\mathbf{V}'} + K_2(H \circ G^{-1})\left(\frac{1}{K_1} \|\mathbf{f}\|_{\mathbf{V}'}\right) \right), \end{aligned}$$

where  $c_0$  and  $c'_0$  are the continuous and discrete inf-sup constants, respectively, and  $G$  and  $H$  are continuous strictly monotonic increasing functions defined on  $[0, \infty)$ .



# Quasi-norms in Sobolev space

## Lemma

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$$C_2 |\mathbf{u} - \mathbf{u}_h|_{(\mathbf{u}, r, \alpha)}^2 \leq a(\mathbf{u}, \mathbf{u} - \mathbf{u}_h) - a(\mathbf{u}_h, \mathbf{u} - \mathbf{u}_h),$$

$$|a(\mathbf{u}, \mathbf{w}) - a(\mathbf{u}_h, \mathbf{w})| \leq C_4 |\mathbf{u} - \mathbf{u}_h|_{(\mathbf{u}, r, \alpha)}^{\min\{1, \frac{2}{r}\}} \|\mathbf{w}\|_{\mathbf{V}}$$

where  $C_4 = C_3 2^{\max\{0, \frac{r-2}{2}\}} [(\Xi_\alpha \circ \mathbf{G}^{-1}) (\frac{1}{K_1} \|\mathbf{f}\|_{\mathbf{V}'})]^{\max\{0, \frac{r-2}{2}\}}$ .



# Upper bound in terms of residual functionals

We define  $S_1 \in V'$  by

$$\langle S_1, w \rangle = (f, w) - a(u_h, w) - b(p_h, w), \quad \forall w \in V.$$

Similarly, we define  $S_2 \in Q'$  by

$$\langle S_2, q \rangle = -b(q, u_h), \quad \forall q \in Q.$$

## Theorem

Let  $(u, p) \in V \times Q$  and  $(u_h, p_h) \in V_h \times Q_h$  denote the solutions to continuous and discrete problems, respectively. Then, there exists a positive constant  $C$  depending on  $K_1, K_2, c_0, c'_0, r, \|f\|_{V'}$  such that

$$\|u - u_h\|_V^{R_U} + \|p - p_h\|_Q^R \leq C \left( \|S_1\|_{V'}^{R'_U} + \|S_2\|_{Q'}^{R'_Q} \right),$$

where  $R_U = \max\{r, 2\}$ ,  $R = \max\{r', 2\}$ ,  $1/R_U + 1/R'_U = 1$ ,  
 $1/R + 1/R' = 1$ .



# Computable residuals

## Definition

$$\mathbf{R}_T([\mathbf{u}_h, p_h]) = \nabla \cdot (k(x, |e(\mathbf{u}_h)|) e(\mathbf{u}_h)) - \nabla p + \mathbf{f} \Big|_T, \quad \forall T \in \mathcal{T}_h,$$

$$\mathbf{J}_E([\mathbf{u}_h, p_h]) = \llbracket \hat{\mathbf{n}}_E \cdot (k(x, |e(\mathbf{u}_h)|) e(\mathbf{u}_h) - p_h \mathbf{I}) \rrbracket_E, \quad \forall E \in \mathcal{E}_{h,\Omega},$$

$$\nabla \cdot \mathbf{u}_h$$

- $\Pi_T \mathbf{R}_T([\mathbf{u}_h, p_h])$  polynomial approx. of  $\mathbf{R}_T([\mathbf{u}_h, p_h])$  on  $T \in \mathcal{T}_h$ ;
- $\Pi_E \mathbf{J}_E([\mathbf{u}_h, p_h])$  polynomial approx. of  $\mathbf{J}_E([\mathbf{u}_h, p_h])$  on  $E \in \mathcal{E}_{h,\Omega}$ .
- $\mathcal{T}_h$  shape-regular partition of  $\Omega$ ;
- $\forall T \in \mathcal{T}_h$ ,  $\mathcal{E}(T)$  set of its  $(d - 1)$ -dimensional faces;
- $\mathcal{E}_h = \bigcup_{T \in \mathcal{T}_h} \mathcal{E}(T)$  be the set of all faces of  $\mathcal{T}_h$ ;
- $\mathcal{E}_{h,\Omega} = \{E \in \mathcal{E}_h : E \not\subset \partial\Omega\}$ .



# Upper bound in terms of computable residuals

## Theorem

Let  $(\mathbf{u}, p) \in \mathbf{V} \times \mathbf{Q}$  and  $(\mathbf{u}_h, p_h) \in \mathbf{V}_h \times \mathbf{Q}_h$  denote the solution to the continuous and discrete problems, respectively. Then, there exists a positive constant  $\mathbf{C}_U$  depending on  $\mathbf{K}_1, \mathbf{K}_2, c_0, c'_0, r, \|\mathbf{f}\|_{\mathbf{V}'}$ , and on the minimal angle of the triangulation such that

$$\begin{aligned} \|\mathbf{u} - \mathbf{u}_h\|_{\mathbf{V}}^{\mathbf{R}_U} + \|p - p_h\|_{\mathbf{Q}}^{\mathbf{R}} &\leq \mathbf{C}_U \\ &\times \left[ \left( \sum_{T \in \mathcal{T}_h} h_T^{r'_T} \|\Pi_T \mathbf{R}_T([\mathbf{u}_h, p_h])\|_{L^{r'}(T)}^{r'_T} \right)^{\frac{\mathbf{R}'_U}{r'}} \right. \\ &+ \left. \left( \sum_{E \in \mathcal{E}_{h,\Omega}} h_E \|\Pi_E \mathbf{J}_E([\mathbf{u}_h, p_h])\|_{L^{r'}(E)}^{r'_E} \right)^{\frac{\mathbf{R}'_U}{r'}} + \|\nabla \cdot \mathbf{u}_h\|_{L^{r'}(\Omega)}^{\mathbf{R}'_U} \dots \end{aligned}$$





# Upper bound in terms of computable residuals

## Theorem

$$\dots + \left( \sum_{T \in \mathcal{T}_h} h_T^{r'} \| \mathbf{R}_T([\mathbf{u}_h, p_h]) - \Pi_T \mathbf{R}_T([\mathbf{u}_h, p_h]) \|_{L^{r'}(T)}^{r'} \right)^{\frac{R'_U}{r'}} + \left( \sum_{E \in \mathcal{E}_{h,\Omega}} h_E \| \mathbf{J}_E([\mathbf{u}_h, p_h]) - \Pi_E \mathbf{J}_E([\mathbf{u}_h, p_h]) \|_{L^{r'}(E)}^{r'} \right)^{\frac{R'_U}{r'}} \Bigg],$$

where  $R_U = \max\{r, 2\}$ ,  $R = \max\{r', 2\}$ ,  $1/R_U + 1/R'_U = 1$ ,  
 $1/R + 1/R' = 1$ .



# Lower bound in terms of residual functionals

## Theorem

Let  $(\mathbf{u}, p) \in \mathbf{V} \times \mathbf{Q}$  and  $(\mathbf{u}_h, p_h) \in \mathbf{V}_h \times \mathbf{Q}_h$  denote the solution to the continuous and discrete problems, respectively. Then, there exists a positive constant  $c$  depending on  $\mathbf{K}_1, \mathbf{K}_2, r, \|\mathbf{f}\|_{\mathbf{V}'}$  such that

$$c \left[ \|S_1\|_{\mathbf{V}'}^{R'_L} + \|S_2\|_{\mathbf{Q}'}^{R'_U} \right] \leq \| \mathbf{u} - \mathbf{u}_h \|_{\mathbf{V}}^{R_L} + \| p - p_h \|_{\mathbf{Q}}^{R_U}$$

where  $R_L = \min\{r, 2\}$ ,  $R_U = \max\{r', 2\}$ ,  $1/R_L + 1/R'_L = 1$ ,  $1/R_U + 1/R'_U = 1$ , and  $S_1$  and  $S_2$  are residual functionals which are computably bounded.

- $R_L = \min\{r, 2\}$ ;
- $R_U = \max\{r, 2\}$ .

▶ upper



# Lower bound in terms of computable residuals

## Theorem

Let  $(\mathbf{u}, p) \in \mathbf{V} \times \mathbf{Q}$  and  $(\mathbf{u}_h, p_h) \in \mathbf{V}_h \times \mathbf{Q}_h$  denote the solution to the continuous and discrete problems, respectively. Then, there exists a positive constant  $c_L$  depending on  $\mathbf{K}_1, \mathbf{K}_2, r, \|\mathbf{f}\|_{\mathbf{V}'}$  and on the minimal angle of the triangulation such that

$$\begin{aligned}
 & c_L \left[ \left( \sum_{T \in \mathcal{T}_h} h_T^{r'} \|\Pi_T \mathbf{R}_T([\mathbf{u}_h, p_h])\|_{L^{r'}(T)}^{r'} \right)^{\frac{R'_L}{r'}} \right. \\
 & \left. + \left( \sum_{E \in \mathcal{E}_{h, \Omega}} h_E \|\Pi_E \mathbf{J}_E([\mathbf{u}_h, p_h])\|_{L^{r'}(E)}^{r'} \right)^{\frac{R'_L}{r'}} + \|\nabla \cdot \mathbf{u}_h\|_{L^{r'}(\Omega)}^{R'_L} \right] \\
 & \leq \|\mathbf{u} - \mathbf{u}_h\|_{\mathbf{V}}^{R_L} + \|p - p_h\|_{\mathbf{Q}}^{R_Q} \dots
 \end{aligned}$$



# Lower bound in terms of computable residuals

## Theorem

$$\begin{aligned}
 & + \left( \sum_{T \in \mathcal{T}_h} h_T^{r'} \left\| \mathbf{R}_T([\mathbf{u}_h, p_h]) - \Pi_T \mathbf{R}_T([\mathbf{u}_h, p_h]) \right\|_{L^{r'}(T)} \right)^{\frac{R'_L}{r'}} \\
 & + \left( \sum_{E \in \mathcal{E}_{h,\Omega}} h_E \left\| \mathbf{J}_E([\mathbf{u}_h, p_h]) - \Pi_E \mathbf{J}_E([\mathbf{u}_h, p_h]) \right\|_{L^{r'}(E)} \right)^{\frac{R'_L}{r'}}
 \end{aligned}$$

where  $R_L = \min\{r, 2\}$ ,  $R = \max\{r', 2\}$ ,  $1/R_L + 1/R'_L = 1$ ,  
 $1/R + 1/R' = 1$ .



# From two-sides bounds to the adaptive method

We have proved

- upper and lower bounds in terms of residual functionals;
- upper and lower bounds in terms of computable residuals;
- (remarks....)

now we are ready to use them in our adaptive algorithm that is based on the local information given by:



# Adaptive algorithm

## Definition

For each  $T \in \mathcal{T}_h$  let us define the following elemental error indicators:

$$\begin{aligned} \mathbf{R}_{\text{mom},T} &= h_T^{r'} \|\Pi_T \mathbf{R}_T([\mathbf{u}_h, \mathbf{p}_h])\|_{L^{r'}(T)} \\ &\quad + \frac{1}{2} \sum_{E \in \mathcal{E}(T) \cap \mathcal{E}_{h,\Omega}} h_E \|\Pi_E \mathbf{J}_E([\mathbf{u}_h, \mathbf{p}_h])\|_{L^{r'}(E)} \\ &\quad + \sum_{E \in \mathcal{E}(T) \cap \mathcal{E}_{h,N}} h_E \|\Pi_E \mathbf{J}_E([\mathbf{u}_h, \mathbf{p}_h])\|_{L^{r'}(E)}, \end{aligned}$$

$$\mathbf{R}_{\text{cont},T} = \|\nabla \cdot \mathbf{u}_h\|_{L^r(T)}.$$



# Adaptive algorithm

## Algorithm

- 1 *Solve the problem on the given mesh;*



# Adaptive algorithm

## Algorithm

- 1 *Solve the problem on the given mesh;*
- 2 *compute the elemental error indicators  $\mathbf{R}_{\text{mom},T}$  and  $\mathbf{R}_{\text{cont},T}$*





# Adaptive algorithm

## Algorithm

- 1 *Solve the problem on the given mesh;*
- 2 *compute the elemental error indicators  $\mathbf{R}_{\text{mom},T}$  and  $\mathbf{R}_{\text{cont},T}$*
- 3 *sort the vectors  $\mathbf{R}.\text{mom}[T]$ ,  $\mathbf{R}.\text{cont}[T]$ ,  $\forall T \in \mathcal{T}_h$ ;*



# Adaptive algorithm

## Algorithm

- 1 Solve the problem on the given mesh;
- 2 compute the elemental error indicators  $\mathbf{R}_{\text{mom},T}$  and  $\mathbf{R}_{\text{cont},T}$
- 3 sort the vectors  $\mathbf{R}.\text{mom}[T]$ ,  $\mathbf{R}.\text{cont}[T]$ ,  $\forall T \in \mathcal{T}_h$ ;
- 4 mark for refinement the  $\rho\%$  of the elements with the larger values of  $\mathbf{R}.\text{mom}$ ;



# Adaptive algorithm

## Algorithm

- 1 *Solve the problem on the given mesh;*
- 2 *compute the elemental error indicators  $\mathbf{R}_{\text{mom},T}$  and  $\mathbf{R}_{\text{cont},T}$*
- 3 *sort the vectors  $\mathbf{R}.\text{mom}[T]$ ,  $\mathbf{R}.\text{cont}[T]$ ,  $\forall T \in \mathcal{T}_h$ ;*
- 4 *mark for refinement the  $\rho\%$  of the elements with the larger values of  $\mathbf{R}.\text{mom}$ ;*
- 5 *mark for refinement the  $\rho\%$  of the elements with the larger values of  $\mathbf{R}.\text{cont}$ ;*



# Adaptive algorithm

## Algorithm

- 1 Solve the problem on the given mesh;
- 2 compute the elemental error indicators  $\mathbf{R}_{\text{mom},T}$  and  $\mathbf{R}_{\text{cont},T}$
- 3 sort the vectors  $\mathbf{R}.\text{mom}[T]$ ,  $\mathbf{R}.\text{cont}[T]$ ,  $\forall T \in \mathcal{T}_h$ ;
- 4 mark for refinement the  $\rho\%$  of the elements with the larger values of  $\mathbf{R}.\text{mom}$ ;
- 5 mark for refinement the  $\rho\%$  of the elements with the larger values of  $\mathbf{R}.\text{cont}$ ;
- 6 pre-mark for coarsening the  $\gamma\%$  of the elements with the smaller values of  $\mathbf{R}.\text{mom}$ ;



# Adaptive algorithm

## Algorithm

- 1 *Solve the problem on the given mesh;*
- 2 *compute the elemental error indicators  $\mathbf{R}_{\text{mom},T}$  and  $\mathbf{R}_{\text{cont},T}$*
- 3 *sort the vectors  $\mathbf{R}.\text{mom}[T]$ ,  $\mathbf{R}.\text{cont}[T]$ ,  $\forall T \in \mathcal{T}_h$ ;*
- 4 *mark for refinement the  $\rho\%$  of the elements with the larger values of  $\mathbf{R}.\text{mom}$ ;*
- 5 *mark for refinement the  $\rho\%$  of the elements with the larger values of  $\mathbf{R}.\text{cont}$ ;*
- 6 *pre-mark for coarsening the  $\gamma\%$  of the elements with the smaller values of  $\mathbf{R}.\text{mom}$ ;*
- 7 *pre-mark for coarsening the  $\gamma\%$  of the elements with the smaller values of  $\mathbf{R}.\text{cont}$ ;*



# Adaptive algorithm

## Algorithm

- 1 *Solve the problem on the given mesh;*
- 2 *compute the elemental error indicators  $\mathbf{R}_{\text{mom},T}$  and  $\mathbf{R}_{\text{cont},T}$*
- 3 *sort the vectors  $\mathbf{R}.\text{mom}[T]$ ,  $\mathbf{R}.\text{cont}[T]$ ,  $\forall T \in \mathcal{T}_h$ ;*
- 4 *mark for refinement the  $\rho\%$  of the elements with the larger values of  $\mathbf{R}.\text{mom}$ ;*
- 5 *mark for refinement the  $\rho\%$  of the elements with the larger values of  $\mathbf{R}.\text{cont}$ ;*
- 6 *pre-mark for coarsening the  $\gamma\%$  of the elements with the smaller values of  $\mathbf{R}.\text{mom}$ ;*
- 7 *pre-mark for coarsening the  $\gamma\%$  of the elements with the smaller values of  $\mathbf{R}.\text{cont}$ ;*
- 8 *mark for coarsening the element (compatibility rule);*



# Adaptive algorithm

## Algorithm

- 1 Solve the problem on the given mesh;
- 2 compute the elemental error indicators  $\mathbf{R}_{\text{mom},T}$  and  $\mathbf{R}_{\text{cont},T}$
- 3 sort the vectors  $\mathbf{R}.\text{mom}[T]$ ,  $\mathbf{R}.\text{cont}[T]$ ,  $\forall T \in \mathcal{T}_h$ ;
- 4 mark for refinement the  $\rho\%$  of the elements with the larger values of  $\mathbf{R}.\text{mom}$ ;
- 5 mark for refinement the  $\rho\%$  of the elements with the larger values of  $\mathbf{R}.\text{cont}$ ;
- 6 pre-mark for coarsening the  $\gamma\%$  of the elements with the smaller values of  $\mathbf{R}.\text{mom}$ ;
- 7 pre-mark for coarsening the  $\gamma\%$  of the elements with the smaller values of  $\mathbf{R}.\text{cont}$ ;
- 8 mark for coarsening the element (compatibility rule);
- 9  $\rho = \rho/f$ ,  $\gamma = \gamma/f$ ;



# Adaptive algorithm

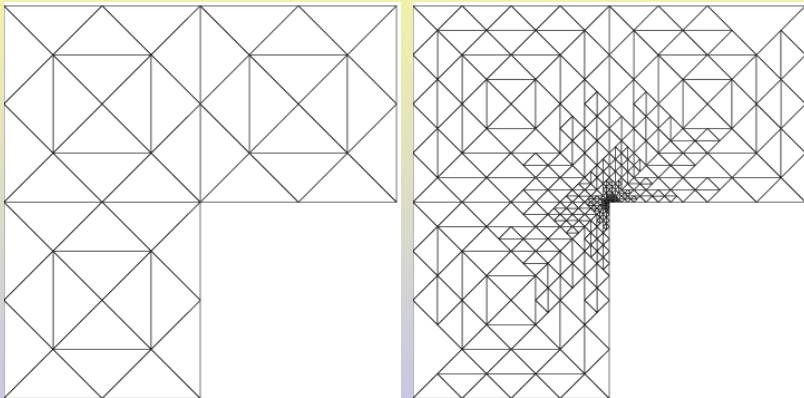
## Algorithm

- 1 Solve the problem on the given mesh;
- 2 compute the elemental error indicators  $\mathbf{R}_{\text{mom},T}$  and  $\mathbf{R}_{\text{cont},T}$
- 3 sort the vectors  $\mathbf{R}.\text{mom}[T]$ ,  $\mathbf{R}.\text{cont}[T]$ ,  $\forall T \in \mathcal{T}_h$ ;
- 4 mark for refinement the  $\rho\%$  of the elements with the larger values of  $\mathbf{R}.\text{mom}$ ;
- 5 mark for refinement the  $\rho\%$  of the elements with the larger values of  $\mathbf{R}.\text{cont}$ ;
- 6 pre-mark for coarsening the  $\gamma\%$  of the elements with the smaller values of  $\mathbf{R}.\text{mom}$ ;
- 7 pre-mark for coarsening the  $\gamma\%$  of the elements with the smaller values of  $\mathbf{R}.\text{cont}$ ;
- 8 mark for coarsening the element (compatibility rule);
- 9  $\rho = \rho/f$ ,  $\gamma = \gamma/f$ ;
- 10 **adapt the mesh and back to point 1;**





# Linear test problem



**Figure:** Starting and ending meshes for a linear singular problem, we use P2-P1 elements  
[Verfürth 1989, Carstensen-Funken 2000]



# Linear test problem

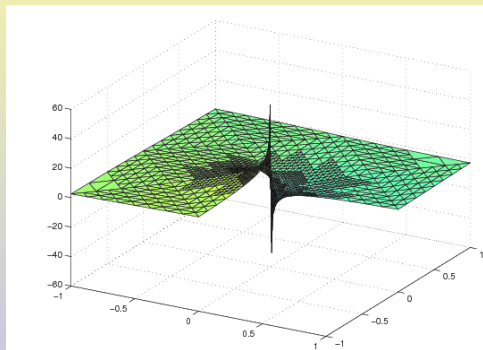
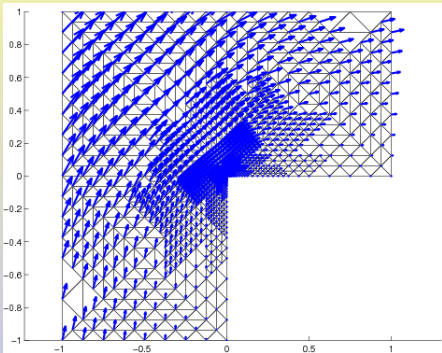


Figure: Velocity and pressure for a linear singular problem  
[Verfürth 1989, Carstensen-Funken 2000]



# Numerical results on adapted meshes

In the following we consider:

- two different geometries  $\Omega_1$  and  $\Omega_2$ ;
- for each of them, a **power-law** model and a **Carreau law** model;
- for each of them, a problem with  $r = 1.3$  and one with  $r = 3.3$ .

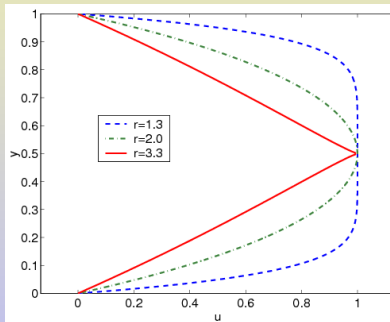


Figure: Velocity profiles for the power-law fully developed flow



# Numerical results on adapted meshes

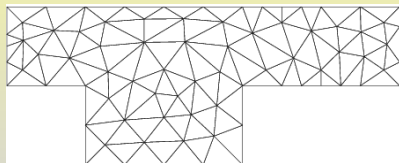


Figure: Starting mesh on the domain  $\Omega_1$

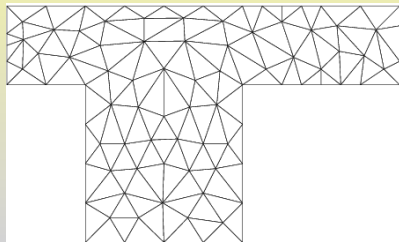


Figure: Starting mesh on the domain  $\Omega_2$

▶ Carreau meshes  $r = 1.3$



# Power-law: $r = 1.3$ , $\mu = 10^{-2}$

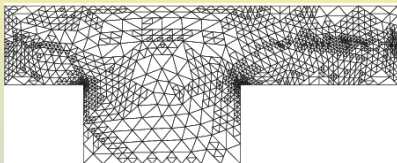


Figure: Power-law,  $r = 1.3$ : final mesh on the domain  $\Omega_1$

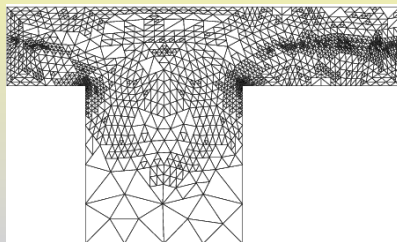


Figure: Power-law,  $r = 1.3$ : final mesh on the domain  $\Omega_2$

▶ profiles

▶  $ur = 1.3$



# Power-law: $r = 1.3$ , $\mu = 10^{-2}$

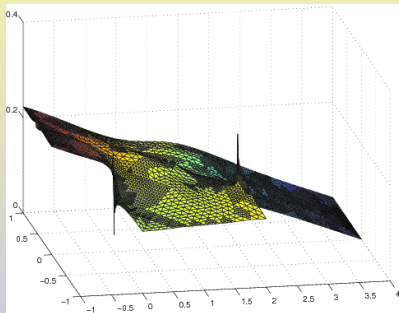


Figure: Power-law,  $r = 1.3$ : pressure on the final mesh on  $\Omega_1$

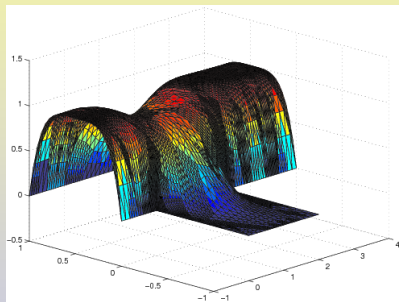


Figure: Power-law,  $r = 1.3$ :  $u$  component of the velocity on  $\Omega_1$



# Power-law: $r = 1.3$ , $\mu = 10^{-2}$

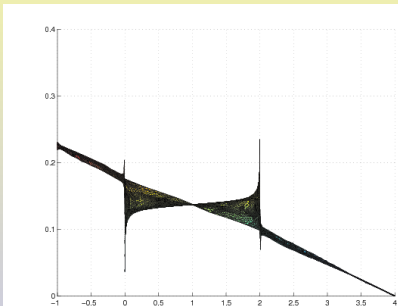


Figure: Power-law,  $r = 1.3$ : pressure on the final mesh on  $\Omega_1$

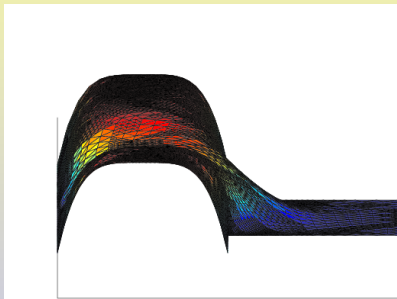


Figure: Power-law,  $r = 1.3$ :  $u$  component of the velocity on  $\Omega_1$



# Power-law: $r = 1.3$ , $\mu = 10^{-2}$

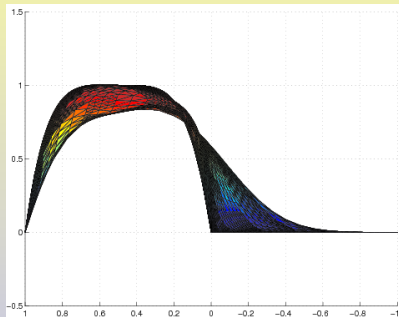


Figure: Power-law,  $r = 1.3$ :  $u$  profile on the final mesh on  $\Omega_1$

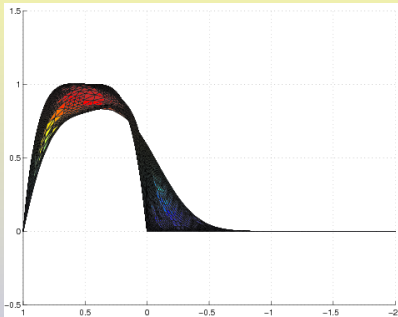


Figure: Power-law,  $r = 1.3$ :  $u$  profile of the velocity on  $\Omega_2$

▶ side vortex view 33





# Power-law: $r = 1.3$ , $\mu = 10^{-2}$

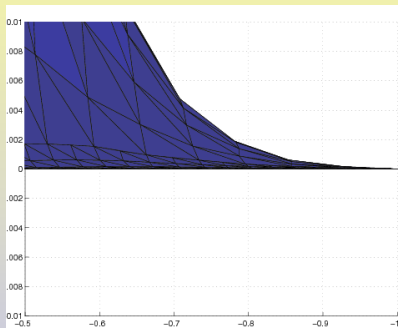


Figure: Power-law,  $r = 1.3$ :  $u$  profile on the final mesh on  $\Omega_1$ , detail in the cavity

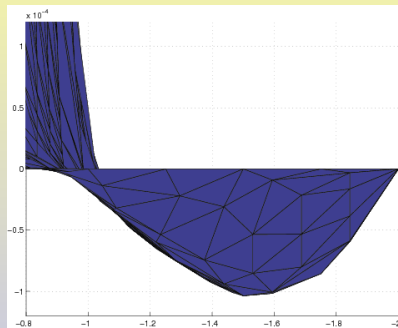


Figure: Power-law,  $r = 1.3$ :  $u$  profile of the velocity on  $\Omega_2$ , detail in the cavity



# Power-law: $r = 1.3$ , $\mu = 10^{-2}$

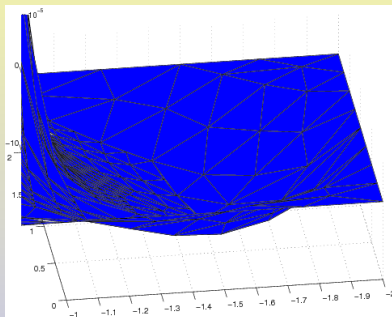


Figure: Power-law,  $r = 1.3$ :  $u$  component on the final mesh on  $\Omega_2$ , detail in the cavity

▶ meshes  $r = 1.3$

▶  $u r = 3.3$



# Monitoring the adaptive process

## Definitions (Total upper residual)

$$\begin{aligned}
 \text{tot.R.U} &= \left( \sum_{T \in \mathcal{T}_h} h_T^{r'} \|\Pi_T \mathbf{R}_T([\mathbf{u}_h, p_h])\|_{L^{r'}(T)} \right)^{\frac{R'_U}{r'}} \\
 &+ \left( \sum_{E \in \mathcal{E}_h, \Omega, \Gamma_N} h_E \|\Pi_E \mathbf{J}_E([\mathbf{u}_h, p_h])\|_{L^{r'}(E)} \right)^{\frac{R'_U}{r'}} \\
 &+ \|\nabla \cdot \mathbf{u}_h\|_{L^r(\Omega)}^{R'_U}
 \end{aligned}$$



# Monitoring the adaptive process

## Definitions (Total lower residual)

$$\begin{aligned}
 \text{tot.R.L} &= \left( \sum_{T \in \mathcal{T}_h} h_T^{r'} \|\Pi_T \mathbf{R}_T([\mathbf{u}_h, p_h])\|_{L^{r'}(T)} \right)^{\frac{r'}{r'}} \\
 &+ \left( \sum_{E \in \mathcal{E}_{h, \Omega, \Gamma_N}} h_E \|\Pi_E \mathbf{J}_E([\mathbf{u}_h, p_h])\|_{L^{r'}(E)} \right)^{\frac{r'}{r'}} \\
 &+ \|\nabla \cdot \mathbf{u}_h\|_{L^r(\Omega)}^{r'}
 \end{aligned}$$



# Power-law: $r = 1.3$ , $\mu = 10^{-2}$

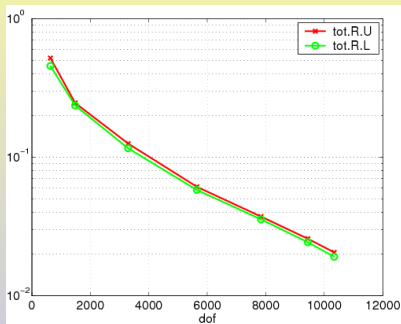


Figure: Power-law,  $r = 1.3$ : total residuals during the adaptive process on  $\Omega_1$

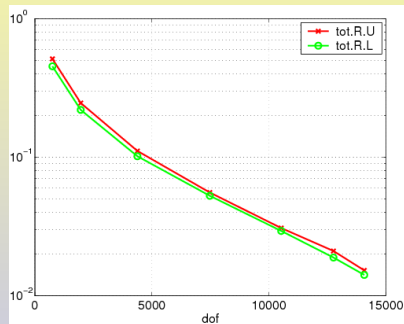


Figure: Power-law,  $r = 1.3$ : total residuals during the adaptive process on  $\Omega_2$



# Power-law: $r = 3.3$ , $\mu = 10^{-2}$

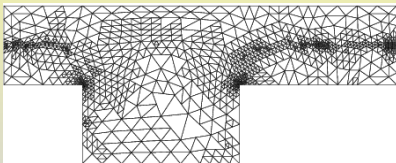


Figure: Power-law,  $r = 3.3$ : final mesh on the domain  $\Omega_1$

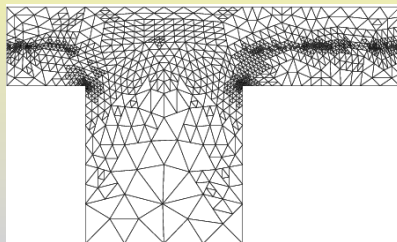


Figure: Power-law,  $r = 3.3$ : final mesh on the domain  $\Omega_2$

▶ profiles



# Power-law: $r = 3.3$ , $\mu = 10^{-2}$

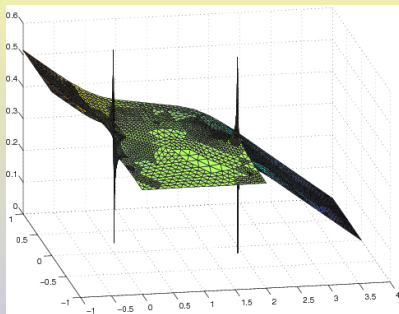


Figure: Power-law,  $r = 3.3$ : pressure on the final mesh on  $\Omega_1$

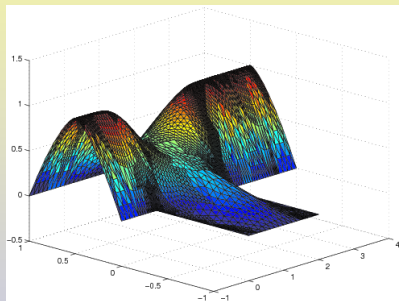


Figure: Power-law,  $r = 3.3$ :  $u$  component of the velocity on  $\Omega_1$



# Power-law: $r = 3.3$ , $\mu = 10^{-2}$

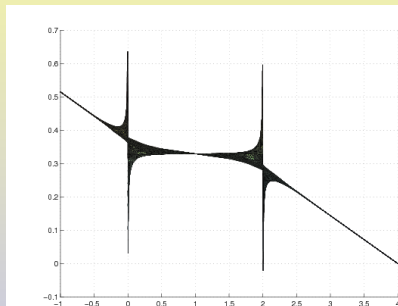


Figure: Power-law,  $r = 3.3$ : pressure on the final mesh on  $\Omega_1$

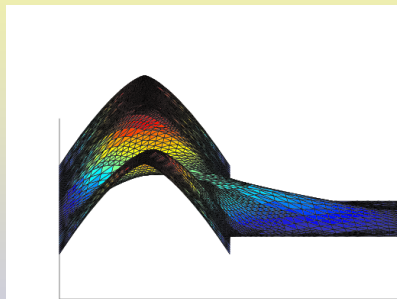


Figure: Power-law,  $r = 3.3$ :  $u$  component of the velocity on  $\Omega_1$





# Power-law: $r = 3.3$ , $\mu = 10^{-2}$

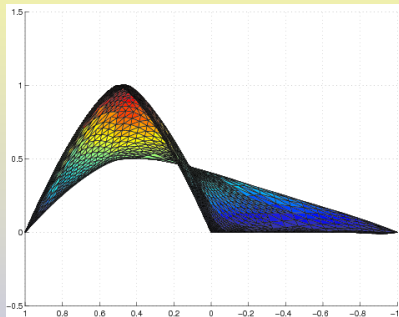


Figure: Power-law,  $r = 3.3$ :  $u$  profile on the final mesh on  $\Omega_1$

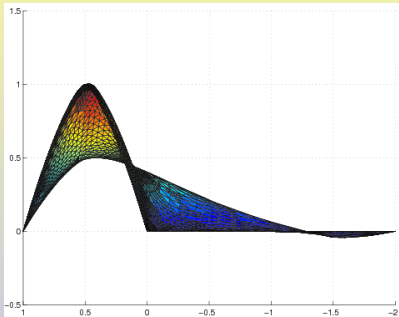


Figure: Power-law,  $r = 3.3$ :  $u$  profile of the velocity on  $\Omega_2$

▶ side vortex view 13



# Power-law: $r = 3.3$ , $\mu = 10^{-2}$

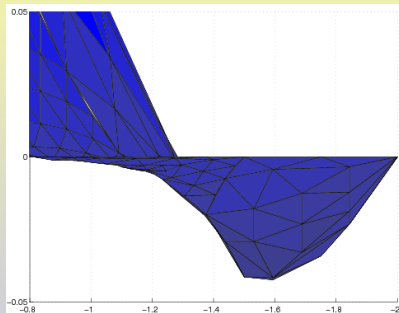


Figure: Power-law,  $r = 3.3$ :  $u$  profile of the velocity on  $\Omega_2$ , detail in the cavity

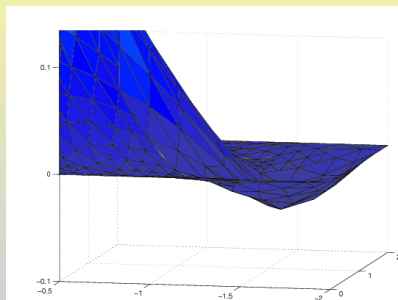


Figure: Power-law,  $r = 3.3$ :  $u$  component of the velocity on  $\Omega_2$ , detail in the cavity

▶  $ur = 1.3$



# Power-law: $r = 3.3$ , $\mu = 10^{-2}$

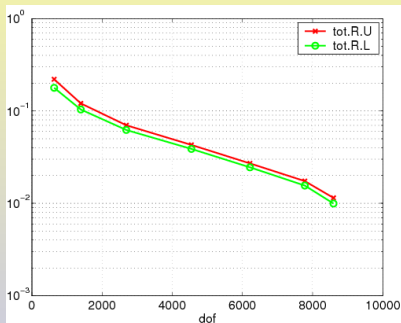


Figure: Power-law,  $r = 3.3$ : total residuals during the adaptive process on  $\Omega_1$

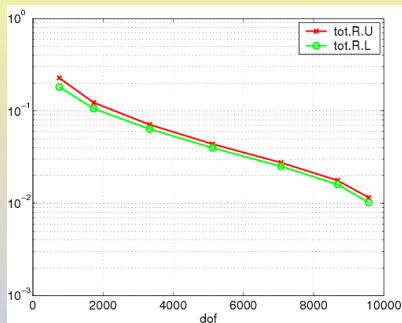


Figure: Power-law,  $r = 3.3$ : total residuals during the adaptive process on  $\Omega_2$



# Carreau law: $\theta = r = 1.3$ , $k_0 = 3 \cdot 10^{-2}$ , $k_\infty = 0$ , $\lambda = 10^{-4}$

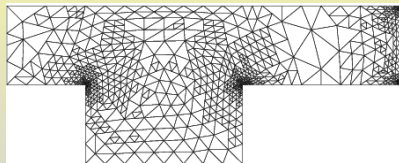


Figure: Carreau law,  $r = 1.3$ : final mesh on the domain  $\Omega_1$

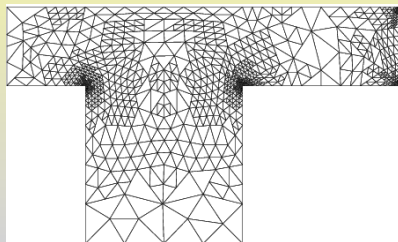


Figure: Carreau law,  $r = 1.3$ : final mesh on the domain  $\Omega_2$

▶ Starting meshes



# Carreau law: $\theta = r = 1.3$ , $k_0 = 3 \cdot 10^{-2}$ , $k_\infty = 0$ , $\lambda = 10^{-4}$

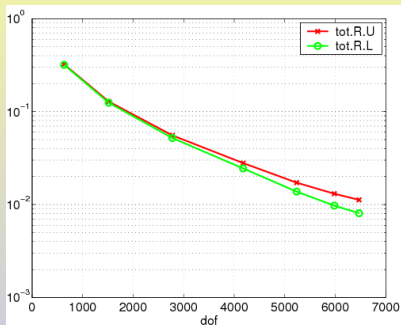


Figure: Carreau law,  $r = 1.3$ : total residuals during the adaptive process on  $\Omega_1$

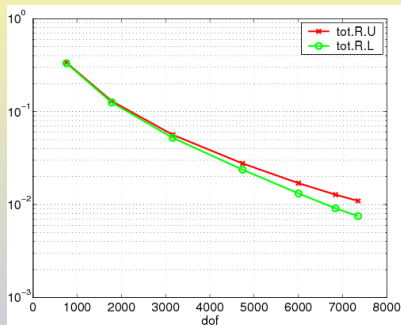


Figure: Carreau law,  $r = 1.3$ : total residuals during the adaptive process on  $\Omega_2$



# Carreau law: $\theta = r = 3.3$ , $k_0 = 3 \cdot 10^{-2}$ , $k_\infty = 0$ , $\lambda = 10^{-4}$

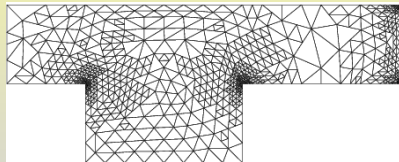


Figure: Carreau law,  $r = 3.3$ : final mesh on the domain  $\Omega_1$

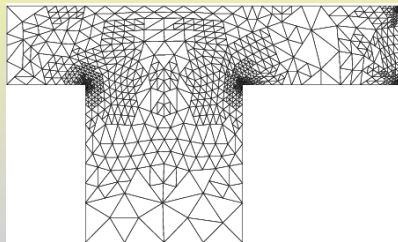


Figure: Carreau law,  $r = 3.3$ : final mesh on the domain  $\Omega_2$



# Carreau law: $\theta = r = 3.3$ , $k_0 = 3 \cdot 10^{-2}$ , $k_\infty = 0$ , $\lambda = 10^{-4}$

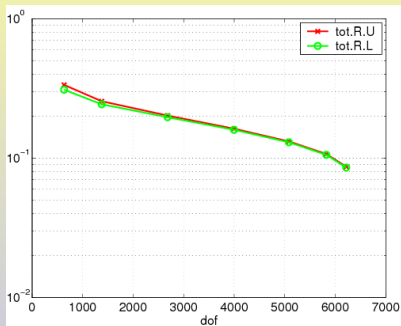


Figure: Carreau law,  $r = 3.3$ : total residuals during the adaptive process on  $\Omega_1$

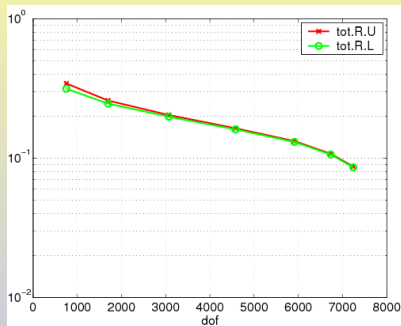


Figure: Carreau law,  $r = 3.3$ : total residuals during the adaptive process on  $\Omega_2$



# Thank you

Thank you for your attention

