

# Méthodes numériques pour des problèmes inverses en mécanique des fluides

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## 1 Introduction

## 2 Numerical reconstruction for the Stokes problem

- Presentation of the data assimilation method
- Theoretical and numerical results

## 3 Application to blood flow

- A simplified framework
- The fluid-structure interaction problem

## A unique continuation problem

We look for  $(u, p) \in H^1(\Omega)^d \times L_0^2(\Omega)$  such that

$$\begin{cases} -\nu \Delta u + \nabla p & = 0 & \text{in } \Omega \\ \nabla \cdot u & = 0 & \text{in } \Omega \end{cases}$$

and

$$u = u_M \quad \text{in } \omega_M$$

where  $\omega_M \subset \Omega$  is an open domain.

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**Unique continuation property:**

$$u_M = 0 \text{ in } \omega_M \Rightarrow (u, p) = (0, 0) \text{ in } H^1(\Omega)^d \times L_0^2(\Omega).$$

[Fabre, Lebeau (1996)]

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**Stability?**

## Ill-posedness of inverse problems

- The solution of inverse problems generally does not depend continuously on the measurements
- Inverse problems are **ill-posed**.
- Continuity is restored in the presence of an a priori bound on the solution.
- We talk about **conditional stability** properties.

## Conditional stability for the continuation problem

Let  $K \subset\subset \Omega$ . There exists  $C > 0$  and  $\tau \in (0, 1)$  such that, for all  $(u, p) \in H^1(\Omega)^d \times H^1(\Omega)$  solution of the homogeneous Stokes equation such that  $\|u\|_{L^2(\Omega)} \leq M$

$$\|u\|_{L^2(K)} \leq CM^{1-\tau} \|u\|_{L^2(\omega_M)}^\tau.$$

[Lin, Uhlmann, Wang (2010)]

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### Remarks

These inequalities are proven thanks to three-balls inequalities

$$\|u\|_{L^2(B(R_2))} \leq C \|u\|_{L^2(B(R_1))}^\tau \|u\|_{L^2(B(R_3))}^{1-\tau} \text{ for } R_1 < R_2 < R_3.$$

[Alessandrini, Rondi, Rosset, Vessella (2009)]



# A classical strategy for the numerical resolution: Tikhonov regularization

We consider the functional:

$$J_\alpha(v) = \frac{1}{2} \|u(v) - u_M\|_{L^2(\omega_M)}^2$$

where  $u(v)$  satisfies

$$\begin{cases} -\nu \Delta u + \nabla p & = f & \text{in } \Omega \\ \nabla \cdot u & = 0 & \text{in } \Omega \\ u & = v & \text{on } \partial\Omega \end{cases}$$

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Choice of  $\alpha$  ? choice of  $\|\cdot\|_{\partial\Omega}$  ?

- Add of an a priori.
- Adaptative choice of  $\alpha$  with respect to the noise or to the mesh size.
  - Morozov criteria: for  $u_M^\delta$  a noisy data, we choose  $\alpha$  such that

$$\|u(v_\alpha^\delta) - u_M^\delta\|_{L^2(\omega_M)} \simeq \delta$$

where  $v_\alpha^\delta$  minimizes  $J_\alpha$ .

- Balance between the discretization error and the regularization error.

[Burman, Hansbo, Larson (2016)]

# The discretize-then-regularize strategy

In what follows, we consider another strategy consisting of discretizing first then regularizing the discrete problem by adding stabilisation terms.

[[Burman \(2013, 2014\)](#)]

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[Burman (2013, 2014)]

### Stabilisation methods

- designed to sort out stability issues for discretized problems
- formed by adding terms to the discrete Galerkin formulation
- vanish quickly enough so that optimal error estimates can be obtained
- originally developed for advection-diffusion equations, fluid equations

[Brooks, Hughes (1981)], [Hughes, Franca, Balestra (1986)], [Johnson, Nävert, Pitkäranta (2016)]

## Variational formulation

[M.B., Burman, Fernandez, Voisembert (2021)]

We define:

$$V := [H^1(\Omega)]^d, \quad V_0 := [H_0^1(\Omega)]^d, \quad L_0 := L_0^2(\Omega), \quad \text{and} \quad L := L^2(\Omega)$$

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We look for  $(u, p) \in V \times L_0$  such that

$$a(u, v) - b(p, v) + b(q, u) = (f, v)_{L^2(\Omega)}, \quad \forall (v, q) \in V_0 \times L$$

where

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We define  $A : (V \times L_0) \times (V_0 \times L)$  by

$$A[(u, p), (v, q)] := a(u, v) - b(p, v) + b(q, u)$$



## A mixed formulation of the minimization problem

We want to find  $(u, p) \in V \times L_0$  that minimizes the functional

$$J(u) = \frac{1}{2} \int_{\omega_M} |u - u_M|^2$$

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We introduce the Lagrangian in  $(V \times L_0) \times (V_0 \times L)$

$$\mathcal{L}[(u, p), (z, y)] = J(u) + A[(u, p), (z, y)] - (f, z)_{L^2(\Omega)}$$

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This problem is ill-posed!

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for all  $(v_h, q_h) \in V_h \times Q_h^0$  and  $(w_h, x_h) \in W_h \times Q_h$

## Properties of the stabilising terms

### Choice of the stabilisation terms $S$ and $S^*$

- They allow to get a well-posed problem at the discrete level.
- They are consistent with the continuous formulation in the sense that, for the solution of the problem  $[(u, p), (z, y)]$ , we have

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At the discrete level, if the measurements are exact in the sense that  $u_M = u|_{\omega_M}$ , we have

$$(z, y) = (0, 0).$$

We will assume that  $(u, p)$  belongs to  $[H^2(\Omega)]^d \times H^1(\Omega)$ .

## Design of the stabilising terms

$$S[(u_h, p_h), (v_h, q_h)] = \gamma_u \sum_{F \in \mathcal{F}_i} \int_F h_F [[\nabla u_h]] [[\nabla v_h]] + \gamma_{div} \int_{\Omega} (\nabla \cdot u_h)(\nabla \cdot v_h)$$

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$$S^*[(z_h, y_h), (w_h, x_h)] = \gamma_u^* \int_{\Omega} \nabla z_h : \nabla w_h + \gamma_p^* \int_{\Omega} y_h x_h$$



## Well-posedness of the discrete problem

We set

$$\begin{aligned} & \mathcal{A}([(u_h, p_h), (z_h, y_h)], [(v_h, q_h), (w_h, x_h)]) \\ &= A[(u_h, p_h), (w_h, x_h)] - S^*[(z_h, y_h), (w_h, x_h)] + A[(v_h, q_h), (z_h, y_h)] + (u_h, v_h)_{L^2(\omega_M)} \\ &+ S[(u_h, p_h), (v_h, q_h)] \end{aligned}$$

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We have

$$S^*[(z_h, y_h), (z_h, y_h)] \geq C(\|z_h\|_{V_0}^2 + \|y_h\|_L^2)$$

and

$$(u_h, u_h)_{L^2(\omega_M)} + S[(u_h, p_h), (u_h, p_h)] \geq Ch^2(\|u_h\|_V^2 + \|p_h\|_L^2)$$

# Error estimate

## Error estimate

Let  $f \in L^2(\Omega)$  be given. We assume that  $u_M := u|_{\omega_M}$ . We assume that the solution  $(u, p)$  belongs to  $[H^2(\Omega)]^d \times H^1(\Omega)$ . Then, for all  $K \subset\subset \Omega$ , there exists  $\tau \in (0, 1)$  such that

$$\|u - u_h\|_{L^2(K)} \leq Ch^\tau (\|u\|_{[H^2(\Omega)]^d} + \|p\|_{H^1(\Omega)}) + h\|f\|_{L^2(\Omega)}.$$

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## Error estimate in presence of noise

Let  $f \in L^2(\Omega)$  and  $\delta u \in L^2(\omega_M)$  be given. We assume that  $u_M := u|_{\omega_M} + \delta u$ . We assume that the solution  $(u, p)$  belongs to  $[H^2(\Omega)]^d \times H^1(\Omega)$ . Then, for all  $K \subset\subset \Omega$ , there exists  $\tau \in (0, 1)$  such that

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# Error estimate

## Remarks

- No error estimate for the pressure.
- No global estimates.

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## Theoretical study

- **Error estimate for the pressure:** for  $(u, p) \in H^1(\Omega) \times L^2(\Omega)$  such that  $\|u\|_{H^1(\Omega)} + \|p\|_{L^2(\Omega)} \leq M$

$$\|u\|_{H^1(K)} + \|p\|_{L^2(K)} \leq CM^{1-\tau} \left( \|u\|_{H^1(\omega_M)} + \|p\|_{L^2(\omega_M)} \right)^\tau.$$

[M.B., Egloffé, Grandmont (2013)], [Badra, Caubet, Dardé (2016)]

- **Global estimates:** for  $(u, p) \in H^2(\Omega) \times H^1(\Omega)$  such that  $\|u\|_{H^2(\Omega)} + \|p\|_{H^1(\Omega)} \leq M$

$$\|u\|_{L^2(\Omega)} \leq C \frac{M}{\log \left( 1 + \frac{M}{\|u\|_{L^2(\omega_M)}} \right)}$$

[Badra, Caubet, Dardé (2016)]



## Error curves

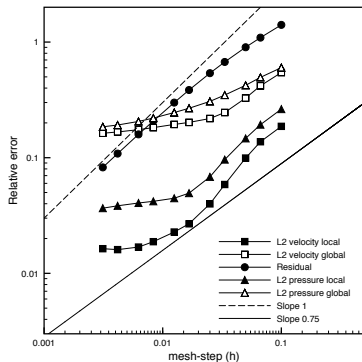


Figure: Error without noise

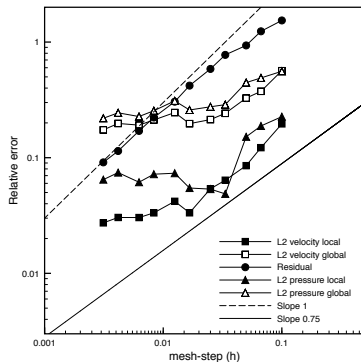
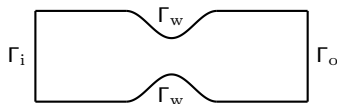


Figure: Error with 10% noise

## A non-stationary problem

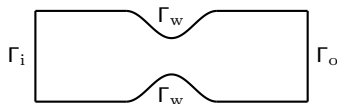
- Blood flow in a stenotic blood vessel



- Measurements of the velocity on the whole domain (corresponding to 4D-MRI).  
We denote by  $u_M^n$  the measurement of the velocity at time  $t_n = n\Delta t$  in  $\Omega$ .

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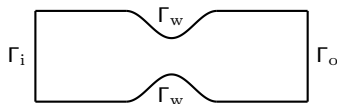


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$$\delta p = \frac{1}{|\Gamma_i|} \int_{\Gamma_i} p - \frac{1}{|\Gamma_o|} \int_{\Gamma_o} p.$$

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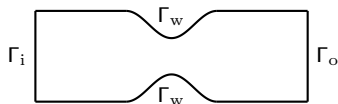
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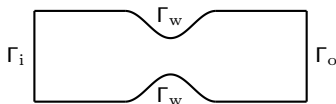
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[\[Bertoglio, Nunez, Galarce, Nordsletten, Osses \(2018\)\]](#)
- Resolution of the inverse problem directly on the nonstationary problem  
[\[Bellassoued, Imanuvilov, Yamamoto \(2016\), M.B. \(2016\)\]](#)

## A direct method combined with a data assimilation method

At each time  $t_n$ , we look for  $(u^n, p^n)$  which minimizes

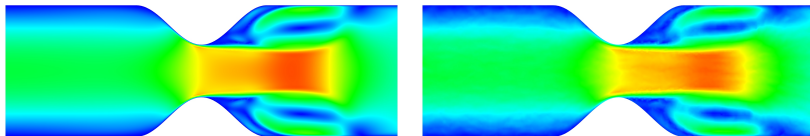
$$\int_{\Omega} |u^n - u_M^n|^2 dx$$

under the constraint that it satisfies the Oseen equations

$$\begin{cases} (u_M^n \cdot \nabla)u^n - \nu \Delta u^n + \nabla p^n = -\frac{u_M^{n+1} - u_M^n}{\Delta t} & \text{in } \Omega, \\ \nabla \cdot u^n = 0 & \text{in } \Omega. \end{cases}$$

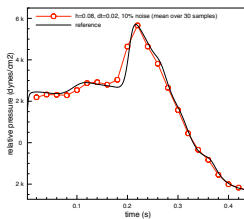
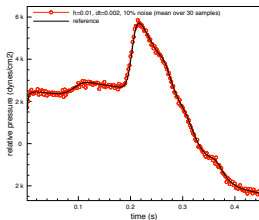
## Numerical results

[M.B., Burman, Fernandez, Voiseibert (2021)]



Velocity magnitude.

Left: reference, right: reconstruction with space-time subsampling and 10% of noise



RPD (black line: reference, red dotted line: reconstruction)

Left: with 10% of noise, right: with space-time subsampling and 10% of noise



## Taking into account the motion of the vessel

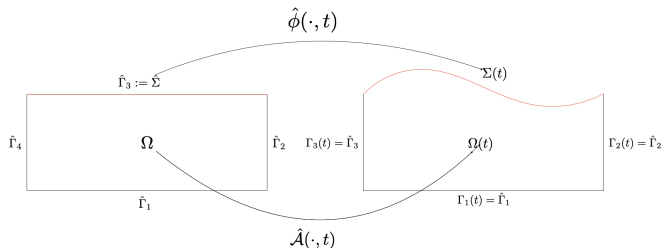
Work in progress with M. Abgalessi, M.A. Fernandez, D. Lombardi and M. Nechita

- For realistic data, it is important to take into account the wall motion.
- Moreover, the images give measurements on the velocity and the displacement of the structure.

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Work in progress with M. Abgalessi, M.A. Fernandez, D. Lombardi and M. Nechita

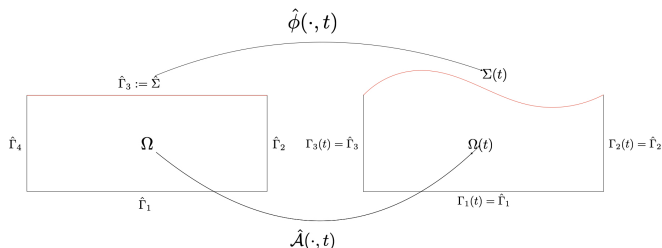
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- We do not have boundary conditions on the inlet and outlet.

## The fluid-structure interaction model

$$\left\{ \begin{array}{ll} \rho^f (\partial_t \mathbf{u}_f + \mathbf{u}_f \cdot \nabla \mathbf{u}_f) - \nabla \cdot \sigma(\mathbf{u}_f, p) = 0 & \text{in } \Omega(t), \\ \nabla \cdot \mathbf{u}_f = 0 & \text{in } \Omega(t), \\ \sigma(\mathbf{u}_f, p) = \nu \epsilon(\mathbf{u}_f) - p \text{Id} & \end{array} \right.$$

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### Remarks

Now, the pressure constant is fixed under the condition that the external pressure is known.

## Luenberger observer for FSI

In the model, we add filters that involve the measurements.

$$\begin{cases} \rho^f (\partial_t \mathbf{u}_f + \mathbf{u}_f \cdot \nabla \mathbf{u}_f) - \nabla \cdot \sigma(\mathbf{u}_f, \rho) & = 0 & \text{in } \Omega(t), \\ \nabla \cdot \mathbf{u}_f = 0 & & \text{in } \Omega(t), \\ \rho^s \partial_t \mathbf{u}_s + \mathcal{L}_d^e(\mathbf{d}_s) & = (\Pi_{\sigma_{\text{ext}}} - \Pi_{\sigma(\mathbf{u}_f, \rho)}) \hat{\mathbf{n}} & \text{on } \hat{\Sigma}, \\ \partial_t \mathbf{d}_s & = \mathbf{u}_s & \text{on } \hat{\Sigma}, \\ \mathbf{u}_f \circ (I + \mathbf{d}_s) = \mathbf{u}_s & & \text{on } \hat{\Sigma}, \\ \mathbf{u}_f = 0 & & \text{on } \hat{\Gamma}_1 \end{cases}$$



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**Boundary conditions in the outlet and inlet?**

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[Bertoglio, Chapelle, Fernandez, Gerbeau, Moireau (2013)]

Boundary conditions in the outlet and inlet? We set:

$$\mathbf{u}_f = \bar{\mathbf{u}}_f \text{ on } \hat{\Gamma}_2 \cup \hat{\Gamma}_4$$

## The continuation step

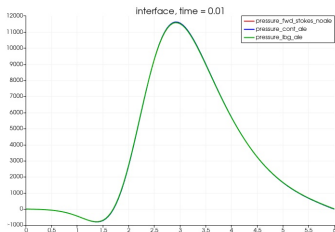
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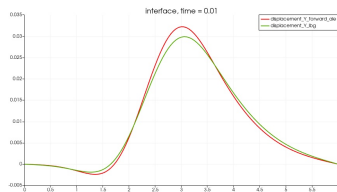
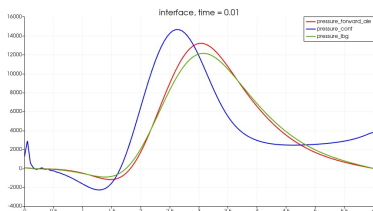
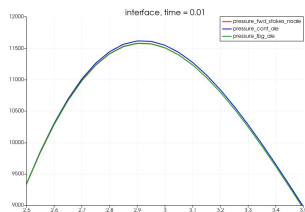
under the constraint that it satisfies the Oseen equations

$$\begin{cases} \frac{1}{\Delta t} \bar{\mathbf{u}}_f^n + (\mathbf{u}_{f,M}^n \cdot \nabla) \bar{\mathbf{u}}_f^n - \nu \Delta \bar{\mathbf{u}}_f^n + \nabla \bar{p}^n = \frac{1}{\Delta t} \mathbf{u}_f^{n-1} & \text{in } \Omega^{n-1}, \\ \nabla \cdot \bar{\mathbf{u}}_f^n = 0 & \text{in } \Omega^{n-1}, \\ \bar{\mathbf{u}}_f^n = 0 & \text{on } \hat{\Gamma}_1, \\ \bar{\mathbf{u}}_f^n \circ (\mathbf{I} + \mathbf{d}_s^{n-1}) = \bar{\mathbf{u}}_s^n & \text{on } \hat{\Sigma}, \\ \rho^s \bar{\mathbf{u}}_s^n + \Delta t \mathcal{L}_d^e(\mathbf{d}_s^{n-1}) = \rho^s \mathbf{u}_s^{n-1} + \Delta t (\Pi_{\sigma_{\text{ext}}} - \Pi_{\sigma_{n-1}}) \hat{\mathbf{n}} & \text{on } \hat{\Sigma} \end{cases}$$

# Numerical results: Stokes and Navier-Stokes equation



Stokes equation: reconstruction of the pressure on the interface at time 0.1s.



Navier-Stokes equation: reconstruction of the pressure and the displacement on the interface at time 0.1s.

## Work in progress

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Thank you for your attention!