

Fictitious domain methods for finite element methods, application to structural mechanics

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- Motivation
- Introduction
- Existence, Uniqueness and Optimality
- *A posteriori* error estimators
- Conclusion and Perspectives

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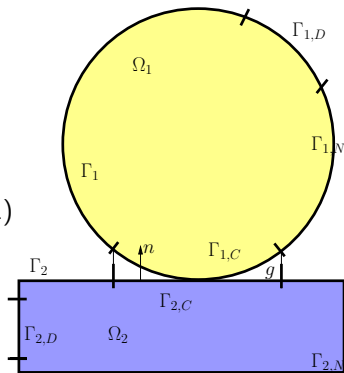
Motivation

Unilateral contact problem between two elastic bodies

We consider the unilateral contact problem between two elastic bodies without friction, as follows:

$$(1) \quad \left\{ \begin{array}{ll} \text{Find } u = (u_1, u_2) \text{ such as:} & \\ -\operatorname{div} \sigma(u_i) = f_i & \text{in } \Omega_i, \\ \sigma(u_i) = A \varepsilon(u_i) & \text{in } \Omega_i, \\ u_i = u_{i,D} & \text{on } \Gamma_{i,D}, \\ \sigma(u_i) n_i = l_i & \text{on } \Gamma_{i,N}, \\ \text{C.C.} & \text{on } \Gamma_{i,C} \end{array} \right.$$

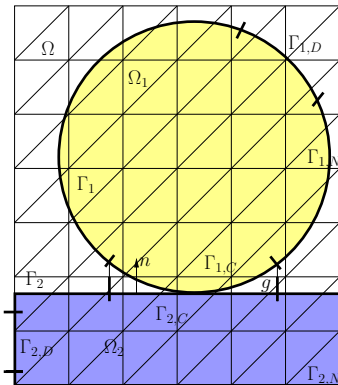
where $\varepsilon(v) = \frac{1}{2}(\nabla v + \nabla v^T)$.



Example of contact problem.

Motivation

Fictitious domain approach



E. Bécache, P.Joly
and G. Scarella, 2001.



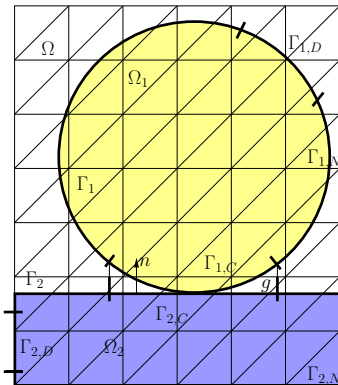
S. Tahir, 2006.

Example of fictitious domain method for an unilateral contact problem between two bodies.

Nitsche's method: consistency and avoid using of lagrangien multiplier

Motivation

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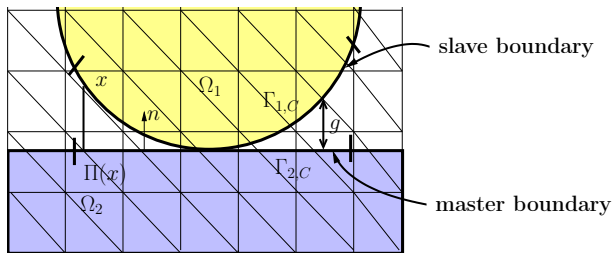
S. Tahir, 2006.

Example of fictitious domain method for an unilateral contact problem between two bodies.

Nitsche's method: consistency and avoid using of lagrangien multiplier

Motivation

Unilateral contact problem without friction



Example of projection, gap and unit normal vector.

Orthogonal projection:

$$\Pi : \begin{array}{l} \Gamma_{1,C} \rightarrow \\ x \mapsto \end{array} \begin{array}{l} \Gamma_{2,C} \\ \Pi(x) \end{array}$$

Initial gap:

$$g : \begin{array}{l} \Gamma_{1,C} \rightarrow \\ x \mapsto \end{array} \begin{array}{l} \mathbb{R} \\ \|x - \Pi(x)\| \end{array}$$

unit normal vector:

$$n : \begin{array}{l} \Gamma_{1,C} \rightarrow \\ x \mapsto \end{array} \begin{array}{l} \mathbb{R}^d \\ n_2(\Pi(x)) \end{array} .$$

Motivation

Unilateral contact conditions without friction

The contact conditions in small deformation can be expressed, as follows:

$$\left\{ \begin{array}{lll} \llbracket u \cdot n \rrbracket \leq g & \text{on } \Gamma_{1,C} & (i), \\ \sigma_n(u_1) \leq 0 & \text{on } \Gamma_{1,C} & (ii), \\ \sigma_n(u_1)(\llbracket u \cdot n \rrbracket - g) = 0 & \text{on } \Gamma_{1,C} & (iii), \\ \llbracket \sigma(u)n \rrbracket = 0 & \text{on } \Gamma_{1,C} & (iv), \\ \sigma_t(u_1) = 0 & \text{on } \Gamma_{1,C} & (v), \end{array} \right. \quad (2)$$

with

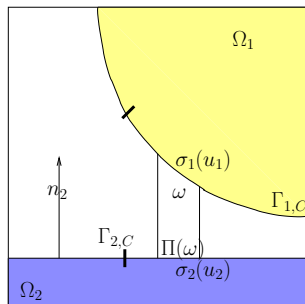
$$\llbracket u \cdot n \rrbracket = (u_2 \circ \Pi - u_1) \cdot n,$$

and

$$\llbracket \sigma(u)n \rrbracket = \sigma(u_1)n_1 + \sigma(u_2 \circ \Pi)n_2 \circ \Pi \quad |\det(J_\Pi)|.$$

Motivation

Action-reaction principle



Example of action-reaction principle.

$$\forall \omega \subset \Gamma_{1,C}, \quad \int_{\omega} \sigma(u_1) \cdot n_1 \, d\Gamma = - \int_{\omega} \sigma(u_2 \circ \Pi) \, |\det(J_{\Pi})| \cdot n_2 \circ \Pi \, d\Gamma.$$

- Motivation
- **Introduction**
- Existence, Uniqueness and Optimality
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We define the convex cone K_c of admissible displacement

$$K_c = \{v = (v_1, v_2) \in H^1(\Omega_1)^d \times H^1(\Omega_2)^d \mid v_1 = u_{1,D} \text{ on } \Gamma_{1,D} \\ \text{and } v_2 = u_{2,D} \text{ on } \Gamma_{2,D} \mid \llbracket v \cdot n \rrbracket - g \leq 0 \text{ on } \Gamma_{1,C}\}.$$

The weak formulation of the contact problem as variational inequality, reads as:

$$\begin{cases} \text{Find } u \in K_c \text{ such as} \\ a(u, v - u) \geq L(v - u) \end{cases} \quad \forall v \in K_c. \quad (3)$$

Stampacchia's theorem

It exists a unique solution of the weak formulation (3)

Using Green's formula and equilibrium equation, it holds:

$$a(u, v) - \sum_{i=1,2} \int_{\Gamma_{i,D}} \sigma_n(u_i) n_i \cdot v_i \, d\Gamma - \int_{\Gamma_{1,C}} \sigma_n(u_1) \llbracket v \cdot n \rrbracket \, d\Gamma = L(v).$$

Reminds: $a(u, v) = \sum_{i=1,2} \int_{\Omega_i} \sigma(u_i) : \varepsilon(v_i) \, d\Omega,$

$$L(v) = \sum_{i=1,2} \int_{\Omega_i} f_i v_i \, d\Omega + \sum_{i=1,2} \int_{\Gamma_{i,N}} l_i v_i \, d\Gamma.$$

proposition

The classical reformulation of contact conditions of (i) -(iii) for $\gamma > 0$ is:

$$\sigma_n(u) = -\frac{1}{\gamma} \llbracket \llbracket u \cdot n \rrbracket - g - \gamma \sigma_n(u) \rrbracket_+ \quad \text{a.e.}$$

- Remark:

$$[[v \cdot n]] = [[v \cdot n]] - \theta \gamma \sigma_n(v) + \theta \gamma \sigma_n(v) \quad , \forall \theta \in \mathbb{R}$$

- For the contact conditions, we replace $-\int_{\Gamma_{1,C}} \sigma_n(u_1) [[v \cdot n]] \, d\Gamma$ by

$$\int_{\Gamma_{1,C}} \frac{1}{\gamma} [[[u \cdot n]] - g - \gamma \sigma_n(u)]_+ ([[v \cdot n]] - \theta \gamma \sigma_n(v)) \, d\Gamma \\ - \int_{\Gamma_{1,C}} \theta \gamma \sigma_n(u) \sigma_n(v) \, d\Gamma$$

- For the contact conditions, we replace $-\int_{\Gamma_{i,D}} \sigma(u_i) n_i \cdot v_i \, d\Gamma$ by

$$\int_{\Gamma_{i,D}} \frac{1}{\gamma} (u_i - u_{i,D} - \gamma \sigma(u_i) n_i) \cdot (v_i - \gamma \theta \sigma(v_i) n_i) \, d\Gamma - \int_{\Gamma_{i,D}} \theta \gamma \sigma(u_i) \sigma(v_i) \, d\Gamma$$

Introduction

Method based on Nitsche's method

We obtain the following formulation based on Nitsche's method

$$\left\{ \begin{array}{l} a(u, v) \\ + \sum_{i=1,2} \int_{\Gamma_{i,D}} \frac{1}{\gamma} (u_i - u_{i,D} - \gamma \sigma(u_i) n_i) \cdot (v_i - \gamma \theta \sigma(v_i) n_i) \, d\Gamma \\ - \sum_{i=1,2} \int_{\Gamma_{i,D}} \theta \gamma \sigma(u_i) \cdot \sigma(v_i) \, d\Gamma \\ + \int_{\Gamma_{1,C}} \frac{1}{\gamma} [\![u \cdot n]\!] - g - \gamma \sigma_n(u)]_+ ([\![v \cdot n]\!] - \theta \gamma \sigma_n(v)) \, d\Gamma \\ - \int_{\Gamma_{1,C}} \theta \gamma \sigma_n(u) \sigma_n(v) \, d\Gamma = L(v). \end{array} \right. \quad (4)$$

The parameter $\theta \in \mathbb{R}$ determines the choice of the method:

- $\theta = 1$ symmetric method, derived of a potentiel.
- $\theta = 0$ non symmetric method.
- $\theta = -1$ skew symmetric method.

Introduction

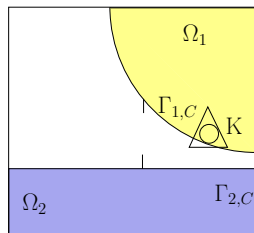
Method based on Nitsche's method

Using standard Galerkin method, we obtain the following discrete formulation based on Nitsche's method

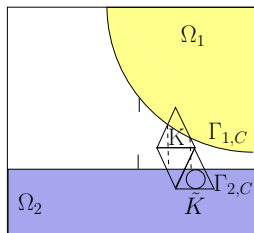
$$\left\{ \begin{array}{l} a(u^h, v^h) \\ + \sum_{i=1,2} \int_{\Gamma_{i,D}} \frac{1}{\gamma} (u_i^h - u_{i,D}^h - \gamma \sigma(u_i^h) n_i) \cdot (v_i^h - \gamma \theta \sigma(v_i^h) n_i) \, d\Gamma \\ - \sum_{i=1,2} \int_{\Gamma_{i,D}} \theta \gamma \sigma(u_i^h) \cdot \sigma(v_i^h) \, d\Gamma \\ + \int_{\Gamma_{1,C}} \frac{1}{\gamma} ([[u^h \cdot n]] - g - \gamma \sigma_n(u^h))_+ ([[v^h \cdot n]] - \theta \gamma \sigma_n(v^h)) \, d\Gamma \\ - \int_{\Gamma_{1,C}} \theta \gamma \sigma_n(u^h) \sigma_n(v^h) \, d\Gamma = L(v^h). \end{array} \right. \quad (5)$$

Introduction

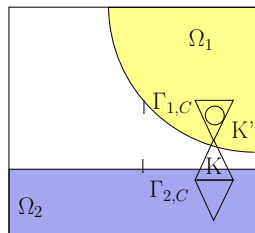
Stabilization of normal constraints on the border



a) If $\Omega_1 \cap K$ is sufficiently large



b) If $\exists \tilde{K} \in \mathcal{S}_K$ such that $\Omega_2 \cap \tilde{K}$ is sufficiently large



c) Otherwise

 J. Haslinger and Y. Renard, 2009.

 E. Burman and P. Hansbo, 2010.

 E. Burman and P. Hansbo, 2012.

Introduction

Method based on Nitsche's method

Contact

$$P_\gamma^h : \begin{array}{l} V_1^h \times V_2^h \\ v \end{array} \rightarrow \begin{array}{l} L^2(\Gamma_{1,C}) \\ \llbracket v \cdot n \rrbracket - \gamma R_{\hat{\rho}}(v) \end{array},$$

Dirichlet

$$\bar{P}_{i,\gamma}^h : \begin{array}{l} V_i^h \\ v_i \end{array} \rightarrow \begin{array}{l} L^2(\Gamma_{i,D})^d \\ v_i - \gamma \bar{R}_{\hat{\rho}}(v_i) \end{array}.$$

Using finite element methods, it holds:

$$\left\{ \begin{array}{l} a(u^h, v^h) \\ + \sum_{i=1,2} \int_{\Gamma_{i,D}} \frac{1}{\gamma} (\bar{P}_{i,\gamma}^h(u_i^h) - u_{i,D}^h) \cdot \bar{P}_{i,\gamma\theta}^h(v_i^h) \, d\Gamma \\ - \sum_{i=1,2} \int_{\Gamma_{i,D}} \theta \gamma \bar{R}_{\hat{\rho}}(u_i^h) \cdot \bar{R}_{\hat{\rho}}(v_i^h) \, d\Gamma \\ + \int_{\Gamma_{1,C}} \frac{1}{\gamma} [P_\gamma^h(u^h) - g]_+ P_{\gamma\theta}^h(v^h) \, d\Gamma \\ - \int_{\Gamma_{1,C}} \theta \gamma R_{\hat{\rho}}(u^h) R_{\hat{\rho}}(v^h) \, d\Gamma = L(v^h) \quad \forall v^h \in V^h. \end{array} \right. \quad (6)$$

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Theorem : Existence, Uniqueness

We take $\gamma = \gamma_0 h_K$ and $\gamma_0 > 0$. If $\theta \neq -1$, then we suppose γ_0 sufficiently small. it exists a unique solution $v^h \in V^h$ of discrete problem (6).

Theorem : Consistence

We define u a sufficiently regular solution of the continuous problem (1) and the contact conditions (2), then u is solution of our discrete problem (6).



M. Fabre, J. Pousin, Y. Renard, 2014.

Theorem : Optimal *a priori* error estimate

We define u a solution of the contact problem in $H^{\frac{3}{2}+\nu}(\Omega_1) \times H^{\frac{3}{2}+\nu}(\Omega_2)$ with $1/2 \geq \nu > 0$ if $k = 1$ and $1 > \nu > 0$ if $k = 2$. If $\theta \neq -1$, we suppose $\gamma_0 > 0$ sufficiently small. The solution u^h of the problem (6) satisfy the following estimation of *a priori* error:

$$\begin{aligned} & \sum_{i=1,2} \left\| u_i - u_i^h \right\|_{1,\Omega_i}^2 + \left\| \gamma^{\frac{1}{2}}(\sigma_n(u) + \frac{1}{\gamma}[P_\gamma^{h,\hat{\rho}}(u^h) - g]_+) \right\|_{0,\Gamma_{1,C}}^2 \\ & + \sum_{i=1,2} \left\| \gamma^{\frac{1}{2}}(\bar{R}_{\hat{\rho}}(u_i^h) - \sigma(u_i)) \right\|_{0,\Gamma_{i,D}}^2 \leq Ch^{1+2\nu} \sum_{i=1,2} \|u\|_{\frac{3}{2}+\nu,\Omega_i}^2 \end{aligned} \quad (7)$$

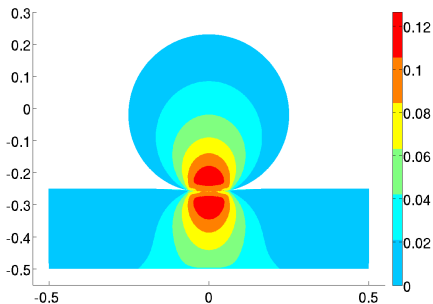
with $C > 0$ a constant independent of h and u .



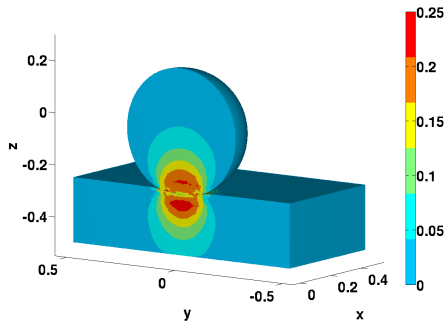
M. Fabre, J. Pousin, Y. Renard, 2014.

Numerical experiments: getfem++

Referential solution $\theta = -1$, $\gamma_0 = \frac{1}{200}$ and Lagrange's elements P_2



2D and $h = \frac{1}{400}$.



3D and $h = \frac{1}{30}$.

Table: rate of convergence with $\gamma_0 = \frac{1}{200}$, Lagrange's elements P_1 in 2D

méthode	$\theta = 1$	$\theta = 0$	$\theta = -1$	optimal rate
$\ u - u^h\ _{0,\Omega_1}$	2.25	2.10	2.00	2 ?
$\ u - u^h\ _{0,\Omega_2}$	1.72	1.73	1.73	2 ?
$\ u - u^h\ _{1,\Omega_1}$	1.11	1.12	1.12	1
$\ u - u^h\ _{1,\Omega_2}$	0.97	0.97	0.97	1

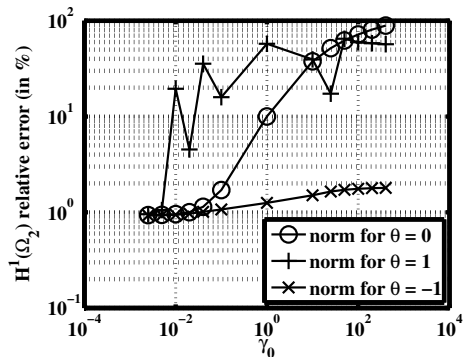
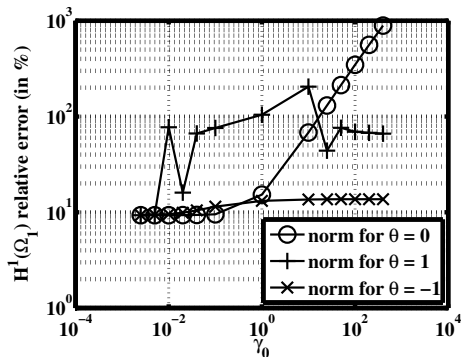
Table: rate of convergence with $\gamma_0 = \frac{1}{200}$, Lagrange's elements P_2 in 2D

méthode	$\theta = 1$	$\theta = 0$	$\theta = -1$	optimal rate
$\ u - u^h\ _{0,\Omega_1}$	1.87	2.23	2.90	2.5 ?
$\ u - u^h\ _{0,\Omega_2}$	2.03	2.14	2.21	2.5 ?
$\ u - u^h\ _{1,\Omega_1}$	1.58	1.80	1.70	1.5
$\ u - u^h\ _{1,\Omega_2}$	1.30	1.31	1.31	1.5

Numerical experiments: getfem++

Influence of γ_0 on the relative H^1 -norm

$h = \frac{1}{90}$, elements P_2 in 2D



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A *posteriori* error estimators

Properties of a *posteriori* error estimators

① $\|u - u^h\| \lesssim C_1 \eta + \zeta \quad \text{and} \quad \eta \lesssim C_2 \|u - u^h\| + \zeta.$

- ② A local error estimator allow to locale the error and adopt a method of adaptive refinement.

 M. Ainsworth and J.T. Oden, 1997.

 M. Ainsworth and J. T. Oden, 2000.

 A. Ern and J.-L. Guermond, 2004. ...

 P. Hild and S. Nicaise, 2007.

A *posteriori* error estimators

Without fictitious domain and without stabilization

We define the unilateral contact problem Without fictitious domain approach between an elastic body and a rigid body:

$$\left\{ \begin{array}{l} \text{Find } u^h \in V^h \text{ such as :} \\ a(u^h, v^h) - \int_{\Gamma_C} \theta \gamma \sigma_n(u^h) \sigma_n(v^h) \, d\Gamma \\ + \int_{\Gamma_C} \frac{1}{\gamma} [P_\gamma^h(u^h)]_+ P_{\theta\gamma}^h(v^h) \, d\Gamma = L(v^h), \quad \forall v^h \in V^h, \end{array} \right. \quad (8)$$

with $a(u, v) = \int_{\Omega} \sigma(u) : \varepsilon(v) \, d\Omega$, $L(v) = \int_{\Omega} f v \, d\Omega + \int_{\Gamma_N} \ell v \, d\Gamma$.

We take

$$f_K = \int_K f(x) \, dx / |K|.$$

A *posteriori* error estimators

Without fictitious domain and without stabilization

We introduce the local error estimator η_K and the global η defined by

definition of a *posteriori* error estimators

$$\begin{aligned}\eta_K &= \left(\sum_{i=1}^4 \eta_{iK}^2 \right)^{1/2} \quad \text{and} \quad \eta = \left(\sum_{K \in \mathcal{T}_h} \eta_K^2 \right)^{1/2}, \\ \eta_{1K} &= h_K \| \operatorname{div} \sigma(u^h) + f_K \|_{0,K}, \\ \eta_{2K} &= h_K^{1/2} \left(\sum_{E \in E_K^{\text{int}} \cup E_K^N} \| J_{E,n}(u^h) \|_{0,E}^2 \right)^{1/2}, \\ \eta_{3K} &= h_K^{1/2} \left(\sum_{E \in E_K^C} \| \sigma_t(u^h) \|_{0,E}^2 \right)^{1/2}, \\ \eta_{4K} &= h_K^{1/2} \left(\sum_{E \in E_K^C} \left\| \frac{1}{\gamma} [P_\gamma^h(u^h) - g]_+ + \sigma_n(u^h) \right\|_{0,E}^2 \right)^{1/2}.\end{aligned}\tag{9}$$

A posteriori error estimators

Without fictitious domain and without stabilization

$J_{E,n}(u^h)$ means the constraint jump of u^h in the normal direction, i.e.

$$J_{E,n}(u^h) = \begin{cases} \llbracket \sigma(u^h) \nu_E \rrbracket, & \forall E \in E_h^{int}, \\ \sigma(u_i^h) \nu_E - \ell_{i,E}, & \forall E \in E_h^N. \end{cases} \quad (10)$$

The local and global approximation terms are given by

$$\zeta_K = \left(h_K^2 \sum_{K' \subset \omega_K} \|f_i - f_{i,K'}\|_{0,K'}^2 + h_E \sum_{E \in E_K^N} \|\ell_i - \ell_{i,E}\|_{0,E}^2 \right)^{1/2},$$
$$\zeta = \left(\sum_{K \in T_h} \zeta_K^2 \right)^{1/2}.$$

A *posteriori* error estimators

Without fictitious domain and without stabilization

saturation assumption:

The solution u of the weak problem and the solution u^h of discrete problem (8) are such as:

$$\left\| \sigma_n(u - u^h) \right\|_{0, \Gamma_C} \lesssim h^{-1/2} \|u - u^h\|_{1, \Omega}. \quad (11)$$

 R. Becker, P. Hansbo and R. Stenberg, 2003.

 B.I. Wohlmuth, 1999.

A *posteriori* error estimators

Without fictitious domain and without stabilization

Theorem : Upper error bound (Reliability) :

We take u the solution of variational problem with $u \in (H^{\frac{3}{2}+\nu}(\Omega))^d$ ($\nu > 0$ and $d = 2, 3$) and u^h the solution of the discrete problem (8). We suppose that if $\theta \neq -1$ then $\gamma_0 > 0$ is sufficiently small. Assume that the saturation assumption holds as well. Then we obtain:

$$\begin{aligned} & \|u - u^h\|_{1,\Omega} + h^{1/2} \left\| \sigma_n(u) + \frac{1}{\gamma} [P_\gamma(u^h)]_+ \right\|_{0,\Gamma_C} \\ & + h^{1/2} \|\sigma_n(u) - \sigma_n(u^h)\|_{0,\Gamma_C} \lesssim (1 + \gamma_0)\eta + \zeta. \end{aligned} \tag{12}$$

A posteriori error estimators

Without fictitious domain and without stabilization

Theorem : Lower error bound (Efficacy) :

For all elements $K \in \mathcal{T}_h$, the following local lower error bounds hold:

$$\eta_{1K} \lesssim \|u - u^h\|_{1,K} + \zeta_K, \quad (13)$$

$$\eta_{2K} \lesssim \|u - u^h\|_{1,\omega_K} + \zeta_K. \quad (14)$$

For all elements K such as $K \cap E_K^C \neq \emptyset$, the following local lower error bounds hold:

$$\eta_{3K} \lesssim \|u - u^h\|_{1,K} + \zeta_K, \quad (15)$$

$$\eta_{4K} \lesssim \sum_{E \in E_K^C} h_K^{1/2} \left(\left\| \sigma_n(u) + \frac{1}{\gamma} [P_\gamma(u^h)]_+ \right\|_{0,E} + \left\| \sigma_n(u - u^h) \right\|_{0,E} \right) \quad (16)$$

A posteriori error estimators

Numerical experiments: a square with slip and separation

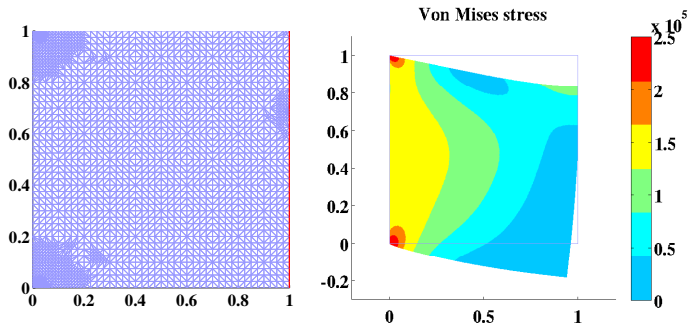


Figure: Left panel: mesh with adaptive refinement and contact boundary. Right panel: plot of Von Mises stress. Parameters $\gamma_0 = 1/E$, $\theta = -1$ and elements P_2 .

A posteriori error estimators

Numerical experiments: a square with slip and separation

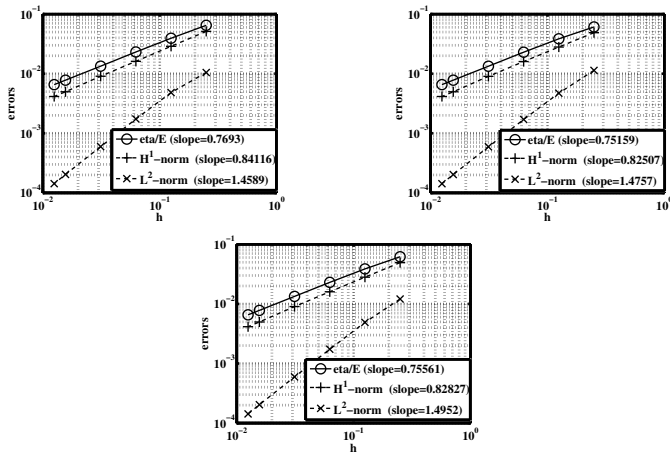


Figure: Convergence curves of the error estimator η , the L^2 and H^1 -norms of the error $u - u^h$, for $\gamma_0 = 1/E$ ($\theta = 1, 0$ or -1).

A posteriori error estimators

Numerical experiments: a square with slip and separation

Table: $\theta = -1$ and $\gamma_0 = 1/E$.

size of mesh h	1/8	1/16	1/32	1/64	1/80	slope
degrees of freedom	128	512	2048	8192	12800	
$\ u - u^h\ _{0,\Omega} (\times 10^{-4})$	48.9718	17.3613	5.9619	2.0360	1.4255	1.4952
$\ u - u^h\ _{1,\Omega} (\times 10^{-3})$	28.1269	16.0087	9.0385	4.9714	4.1467	0.8283
η_1	8359.9	4179.95	2089.97	1044.99	835.99	1
η_2	37649.9	22607.7	13213.2	7723.58	6506.99	0.7428
η_3	1464.81	558.637	192.194	70.7559	53.7733	1.3544
η_4	2854.93	832.228	229.683	62.842	44.0949	1.8004
η	38700.1	23012.7	13380.8	7794.52	6560.84	0,7779
Eff_E	1.3759	1.4375	1.4804	1.5677	1.5820	

A *posteriori* error estimators

With fictitious domain

We introduce the local error estimator η_K and the global η defined by

definition of a *posteriori* error estimators

$$\begin{aligned}\eta_K &= \left(\sum_{i=1}^4 \eta_{iK}^2 \right)^{1/2}, \quad \eta = \left(\sum_{K \in \mathcal{T}_h} \eta_K^2 \right)^{1/2}, \\ \eta_{1K} &= h_K \|\operatorname{div} \sigma(u_i^h) + f_{i,K}\|_{0,K}, \quad \eta_{3K} = h_K^{1/2} \left(\sum_{E \in E_K^C} \|\sigma_t(u^h)\|_{0,E}^2 \right)^{1/2}, \\ \eta_{2K} &= h_K^{1/2} \left(\sum_{E \in E_K^{\text{int}} \cup E_K^N} \|J_{E,n}(u^h)\|_{0,E}^2 \right)^{1/2}, \\ \eta_{4K} &= h_K^{1/2} \left(\sum_{E \in E_K^C} \left\| \frac{1}{\gamma} [P_\gamma^{h,\hat{\rho}}(u^h) - g]_+ + \sigma_n(u^h) \right\|_{0,E}^2 \right)^{1/2}, \\ \eta_{5K} &= h_K^{1/2} \left(\sum_{E \in E_K^D} \left\| \frac{1}{\gamma} \bar{P}_{i,\gamma}^{h,\hat{\rho}}(u_i^h) - u_{i,D} + \sigma(u_i^h) \right\|_{0,E}^2 \right)^{1/2},\end{aligned}$$

A *posteriori* error estimators

With fictitious domain

saturations assumptions:

The solution u of the weak problem and the solution u^h of discrete problem (8) are such as:

$$\begin{aligned} \|\sigma_n(u - u^h)\|_{0,\Gamma_{1,C}}^2 &\lesssim h^{-1} \sum_{i=1,2} \|u_i - u_i^h\|_{1,\Omega_i}^2 \\ \|\sigma_n(u) - R_{\hat{\rho}}(u^h)\|_{0,\Gamma_{1,C}}^2 &\lesssim h^{-1} \sum_{i=1,2} \|u_i - u_i^h\|_{1,\Omega_i}^2, \end{aligned} \quad (17)$$

$$\begin{aligned} \sum_{i=1,2} \|\sigma(u_i - u_i^h)\|_{0,\Gamma_{i,D}}^2 &\lesssim h^{-1} \sum_{i=1,2} \|u_i - u_i^h\|_{1,\Omega_i}^2 \\ \sum_{i=1,2} \|\sigma(u_i) - \bar{R}_{\hat{\rho}}(u_i^h)\|_{0,\Gamma_{i,D}}^2 &\lesssim h^{-1} \sum_{i=1,2} \|u_i - u_i^h\|_{1,\Omega_i}^2, \end{aligned} \quad (18)$$

$$\sum_{i=1,2} \|\sigma(u_i)n_i - \sigma(u_i^h)n_i\|_{0,\Gamma_{i,N}}^2 \lesssim h^{-1} \sum_{i=1,2} \|u_i - u_i^h\|_{1,\Omega_i}^2, \quad (19)$$

$$\sum_{i=1,2} \|\sigma(u_i) - \sigma(u_i^h)\|_{0,\Omega_i}^2 \lesssim h^{-2} \sum_{i=1,2} \|u_i - u_i^h\|_{1,\Omega_i}^2. \quad (20)$$

A posteriori error estimators

With fictitious domain

Upper error bound :

We suppose u the solution of variational problem with $u \in (H^{\frac{3}{2}+\nu}(\Omega_1))^d \times (H^{\frac{3}{2}+\nu}(\Omega_2))^d$ ($\nu > 0$ and $d = 2, 3$) and u^h the solution of the discrete problem. We suppose that if $\theta \neq -1$ then $\gamma_0 > 0$ is sufficiently small. Assume that the saturation assumptions (17)-(20) hold as well. Then we obtain:

$$\begin{aligned} & \sum_{i=1,2} \|u_i - u_i^h\|_{1,\Omega_i} + h^{1/2} \left\| \sigma_n(u) + \frac{1}{\gamma} [P_\gamma^{h,\hat{\rho}}(u^h) - g]_+ \right\|_{0,\Gamma_{1,C}} \\ & + h^{1/2} \|\sigma_n(u) - R_{\hat{\rho}}(u^h)\|_{0,\Gamma_{1,C}} + h^{1/2} \sum_{i=1,2} \left\| \sigma(u_i) + \frac{1}{\gamma} \bar{P}_{i,\gamma}^{h,\hat{\rho}}(u_i^h) - u_{i,D} \right\|_{0,\Gamma_{i,D}} \\ & + h^{1/2} \sum_{i=1,2} \|\sigma(u_i) - \bar{R}_{\hat{\rho}}(u_i^h)\|_{0,\Gamma_{i,D}} \lesssim (1 + \gamma_0)\eta + \zeta. \end{aligned}$$

A posteriori error estimators

With fictitious domain

Lower error bound :

For all elements $K \in \mathcal{T}_h$, the following local lower error bounds hold:

$$\eta_{1K} \lesssim \|u_i - u_i^h\|_{1,K} + \zeta_K, \eta_{2K} \lesssim \|u_i - u_i^h\|_{1,\omega_K} + \zeta_K. \quad (22)$$

For all elements K such as $K \cap E_K^C \neq \emptyset$ then $K \cap E_K^D \neq \emptyset$, the following local lower error bounds hold:

$$\eta_{3K} \lesssim \|u_i - u_i^h\|_{1,K} + \zeta_K, \quad (23)$$

$$\eta_{4K} \lesssim \sum_{E \in E_K^C} h_K^{1/2} \left(\left\| \sigma_n(u) + \frac{1}{\gamma} [P_\gamma^h(u^h) - g]_+ \right\|_{0,E} + \left\| \sigma_n(u - u^h) \right\|_{0,E} \right) \quad (24)$$

$$\eta_{5K} \lesssim \sum_{E \in E_K^D} h_K^{1/2} \left(\left\| \frac{1}{\gamma} \overline{P}_{i,\gamma}^{h,\hat{p}}(u_i^h) - u_{i,D} + \sigma(u_i) \right\|_{0,E} + \left\| \sigma(u_i - u_i^h) \right\|_{0,E} \right) \quad (25)$$

A *posteriori* error estimators

With fictitious domain

We define a new stabilized operator R on all domains.

a new stabilized problem

We can define a new stabilized problem :

$$\left\{ \begin{array}{l} \text{Find } u^h \in V^h \text{ such as} \\ \sum_{i=1,2} \int_{\Omega_i} R(u_i^h) : \varepsilon(v_i^h) \, d\Omega - \int_{\Gamma_{1,C}} \theta \gamma R_{\hat{\rho}}(u^h) R_{\hat{\rho}}(v^h) \, d\Gamma \\ - \sum_{i=1,2} \int_{\Gamma_{i,D}} \theta \gamma \bar{R}_{\hat{\rho}}(u_i^h) \cdot \bar{R}_{\hat{\rho}}(v_i^h) \, d\Gamma + \int_{\Gamma_{1,C}} \frac{1}{\gamma} [P_{\gamma}^{h,\hat{\rho}}(u^h) - g]_+ P_{\theta\gamma}^{h,\hat{\rho}}(v^h) \, d\Gamma \\ + \sum_{i=1,2} \int_{\Gamma_{i,D}} \frac{1}{\gamma} (\bar{P}_{i,\gamma}^{h,\hat{\rho}}(u_i^h) - u_{i,D}) \cdot \bar{P}_{i,\gamma\theta}^{h,\hat{\rho}}(v_i^h) \, d\Gamma = L(v^h) \quad \forall v^h \in V^h. \end{array} \right.$$

A posteriori error estimators

With fictitious domain

We introduce the local error estimator η_K and the global η defined by

definition of *a posteriori* error estimators

$$\begin{aligned}\eta_K &= \left(\sum_{i=1}^4 \eta_{iK}^2 \right)^{1/2}, \quad \eta = \left(\sum_{K \in T_h} \eta_K^2 \right)^{1/2}, \quad \eta_{1K} = h_K \|\operatorname{div} R(u_i^h) + f_{i,K}\|_{0,K}, \\ \eta_{2K} &= h_K^{1/2} \left(\sum_{E \in E_K^{\text{int}} \cup E_K^N} \|J_{E,n}(u^h)\|_{0,E}^2 \right)^{1/2}, \quad \eta_{3K} = h_K^{1/2} \left(\sum_{E \in E_K^C} \|R_t(u^h)\|_{0,E}^2 \right)^{1/2}, \\ \eta_{4K} &= h_K^{1/2} \left(\sum_{E \in E_K^C} \left\| \frac{1}{\gamma} [P_\gamma^{h,\hat{\rho}}(u^h) - g]_+ + R_{\hat{\rho}}(u^h) \right\|_{0,E}^2 \right)^{1/2}, \\ \eta_{5K} &= h_K^{1/2} \left(\sum_{E \in E_K^D} \left\| \frac{1}{\gamma} \bar{P}_{i,\gamma}^{h,\hat{\rho}}(u_i^h) - u_{i,D} + \bar{R}_{\hat{\rho}}(u_i^h) \right\|_{0,E}^2 \right)^{1/2}.\end{aligned}$$

A posteriori error estimators

With fictitious domain

saturations assumptions

The solution u of the weak problem and the solution u^h of discrete problem (26) are such as:

$$\left\| \sigma_n(u) - R_{\hat{\rho}}(u^h) \right\|_{0,\Gamma_{1,C}}^2 \lesssim h^{-1} \sum_{i=1,2} \|u_i - u_i^h\|_{1,\Omega_i}^2, \quad (28)$$

$$\sum_{i=1,2} \left\| \sigma(u_i) - \bar{R}_{\hat{\rho}}(u_i^h) \right\|_{0,\Gamma_{i,D}}^2 \lesssim h^{-1} \sum_{i=1,2} \|u_i - u_i^h\|_{1,\Omega_i}^2, \quad (29)$$

$$\sum_{i=1,2} \left\| \sigma(u_i)n_i - R(u_i^h)n_i \right\|_{0,\Gamma_{i,N}}^2 \lesssim h^{-1} \sum_{i=1,2} \|u_i - u_i^h\|_{1,\Omega_i}^2, \quad (30)$$

$$\sum_{i=1,2} \left\| \sigma(u_i) - R(u_i^h) \right\|_{0,\Omega_i}^2 \lesssim h^{-2} \sum_{i=1,2} \|u_i - u_i^h\|_{1,\Omega_i}^2. \quad (31)$$

A posteriori error estimators

With fictitious domain

Upper error bound:

We suppose u the solution of variational problem with $u \in (H^{\frac{3}{2}+\nu}(\Omega_1))^d \times (H^{\frac{3}{2}+\nu}(\Omega_2))^d$ ($\nu > 0$ and $d = 2, 3$) and u^h the solution of the discrete problem. We suppose that if $\theta \neq -1$ then $\gamma_0 > 0$ is sufficiently small. Assume that the saturation assumptions (28)-(29) hold as well. Then we obtain:

$$\begin{aligned} & \sum_{i=1,2} \|u_i - u_i^h\|_{1,\Omega_i} + \left\| \sigma_n(u) + h^{1/2} \frac{1}{\gamma} [P_\gamma^{h,\hat{\rho}}(u^h) - g]_+ \right\|_{0,\Gamma_{1,C}} \\ & + h^{1/2} \|\sigma_n(u) - R_{\hat{\rho}}(u^h)\|_{0,\Gamma_{1,C}} + h^{1/2} \sum_{i=1,2} \left\| \sigma(u_i) + \frac{1}{\gamma} \bar{P}_{i,\gamma}^{h,\hat{\rho}}(u_i^h) - u_{i,D} \right\|_{0,\Gamma_{i,D}} \\ & + h^{1/2} \sum_{i=1,2} \|\sigma(u_i) - \bar{R}_{\hat{\rho}}(u_i^h)\|_{0,\Gamma_{i,D}} \lesssim (1 + \gamma_0)\eta + \zeta. \end{aligned}$$

A posteriori error estimators

With fictitious domain

Lower error bound:

For all elements $K \in \mathcal{T}_h$, the following local lower error bounds hold:

$$\eta_{1K} \lesssim \|u_i - u_i^h\|_{1,K} + \zeta_K, \eta_{2K} \lesssim \|u_i - u_i^h\|_{1,\omega_K} + \zeta_K. \quad (32)$$

For all elements K such as $K \cap E_K^C \neq \emptyset$ then $K \cap E_K^D \neq \emptyset$, the following local lower error bounds hold:

$$\eta_{3K} \lesssim \|u_i - u_i^h\|_{1,K} + \zeta_K, \quad (33)$$

$$\eta_{4K} \lesssim \sum_{E \in E_K^C} h_K^{1/2} \left(\left\| \sigma_n(u) + \frac{1}{\gamma} [P_\gamma^h(u^h) - g]_+ \right\|_{0,E} + \left\| \sigma_n(u) - R_{\hat{\rho}}(u^h) \right\|_{0,E} \right) \quad (34)$$

$$\eta_{5K} \lesssim \sum_{E \in E_K^D} h_K^{1/2} \left(\left\| \frac{1}{\gamma} \bar{P}_{i,\gamma}^{h,\hat{\rho}}(u_i^h) - u_{i,D} + \sigma(u_i) \right\|_{0,E} + \left\| \sigma(u_i) - \bar{R}_{\hat{\rho}}(u_i^h) \right\|_{0,E} \right) \quad (35)$$

- Motivation
- Introduction
- Existence, Uniqueness and Optimality
- *A posteriori* error estimators
- Conclusion and Perspectives

Conclusion:

- Existence, Uniqueness and Consistence.
- Optimal *a priori* error estimate and theoretical and numerical results.
- *A posteriori* error estimators.

Perspectives:

- Numerical validation of the *a posteriori* error estimators in the Fictitious domain approach.
- Extension of the model (dynamic, non linear elasticity, great deformation).
- To delate the saturations assumptions.



A. Hansbo, P. Hansbo and M.G. Larson, 2003.

Thank you for your attention.