

The fully mixed methods for the Signorini Problem

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PDE model

- **Signorini problem:** simulate the interaction between a linearly elastic body and a rigid foundation without friction.
- **Application:** electropainting, the elastic contact problem, the artificial outflow boundary condition of viscous flow.

Signorini problem (P)

$$-\Delta u = f \quad \text{in } \Omega,$$

$$u = 0 \quad \text{on } \Gamma_D,$$

with Signorini condition

$$u \geq 0, \quad \nabla u \cdot \mathbf{n} \geq 0, \quad u(\nabla u \cdot \mathbf{n}) = 0 \quad \text{on } \Gamma_S.$$

The penalty approach for (P)

$$-\Delta u_\varepsilon = f \quad \text{in } \Omega,$$

$$u_\varepsilon = 0 \quad \text{on } \Gamma_D,$$

$$\nabla u_\varepsilon \cdot \mathbf{n} = \frac{1}{\varepsilon} [u_\varepsilon]_- \quad \text{on } \Gamma_S.$$

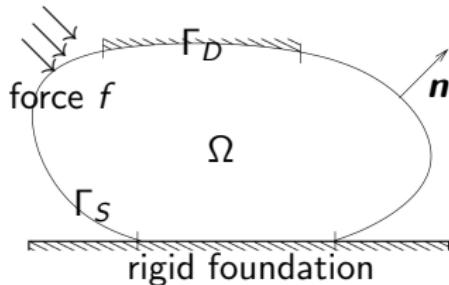


Figure 1: region Ω , Γ_D and Γ_S

Notations

- Ω : smooth and bbd' domain,
- $\partial\Omega := \overline{\Gamma_D \cup \Gamma_S}$, $\Gamma_D \cap \Gamma_S = \emptyset$,
- $f \in L^2(\Omega)$: given external force,
- $u: \Omega \rightarrow \mathbb{R}$,
- \mathbf{n} : unit outer normal vector,
- ε : the penalty parameter ($0 < \varepsilon \ll 1$),
- $[s]_- := \max(0, -s)$.

Variational inequality and Lagrange multiplier form

(VI) Find $u \in \tilde{X}$ such that

$$(\nabla u, \nabla(v - u)) \geq (f, v - u) \quad \forall v \in \tilde{X}.$$

Set **Lagrange multiplier** $\lambda := \nabla u \cdot \mathbf{n} \in H^{-\frac{1}{2}}(\Gamma_S)$. Note that $\lambda \geq 0$.

(VL) Find $(u, \lambda) \in X \times \tilde{\Lambda}_{-1/2}$, such that

$$(\nabla u, \nabla v) - \langle \lambda, v \rangle_{\Gamma_S} = (f, v) \quad \forall v \in X,$$

$$\langle \mu - \lambda, u \rangle_{\Gamma_S} \geq 0 \quad \forall \mu \in \tilde{\Lambda}_{-1/2}.$$

$$X := H_0^1(\Gamma_D), \quad \tilde{X} := \{v \in X : v|_{\Gamma_S} \geq 0\}, \quad \tilde{\Lambda}_{-1/2} := \{\mu \in H^{-\frac{1}{2}}(\Gamma_S) : \mu \geq 0\}.$$

Iteration algorithms for solving (VI) or (VL):

- Uzawa algorithm [Glowinski 84],
- the over-relaxation method [Glowinski 84],
- the penalty method [Haslinger 04],
- the Active/Inactive set method (Primal-dual Active set) [Ito 08].

The variational form of penalty method

(P) $_\varepsilon$ Find $u_\varepsilon \in X$ such that

$$(\nabla u_\varepsilon, \nabla v) - \frac{1}{\varepsilon} \int_{\Gamma_S} [u_\varepsilon]_- v \, ds = (f, v) \quad (\forall v \in X).$$

Introduce $\sigma := \nabla u$ as an additional unknown.

Fully mixed problem and variational inequality

(MP)

$$\sigma - \nabla u = 0 \quad \text{in } \Omega, \quad (2a)$$

$$\nabla \cdot \sigma + f = 0 \quad \text{in } \Omega, \quad (2b)$$

$$u = 0 \quad \text{on } \Gamma_D, \quad (2c)$$

$$u \geq 0, \quad \sigma \cdot n \geq 0, \quad u\sigma \cdot n = 0 \quad \text{on } \Gamma_S. \quad (2d)$$

(MVI) Find $(\sigma, u) \in \tilde{V} \times Q$ s.t.

$$(\sigma, \tau - \sigma) + (u, \nabla \cdot (\tau - \sigma)) \geq 0 \quad \forall \tau \in \tilde{V}, \quad (3a)$$

$$(\nabla \cdot \sigma, v) + (f, v) = 0 \quad \forall v \in Q. \quad (3b)$$

$$\tilde{V} := \{\tau \in H(\text{div}; \Omega) : \tau \cdot n \geq 0 \text{ on } \Gamma_S\}, \quad Q := L^2(\Omega).$$

Compare (MVI) with (VI):

- Unknown σ has a clear physical meaning (**stress**), and is computed directly.
- (MVI) is weaker than (VI), the flexible choice of finite element subspace for u (like **P0-element**).

Perious research

- the theories of abstract mixed variational inequalities [Brezzi 78]
- mixed methods for the elastoplasticity problem [Reddy 92, Han 95]
- unilateral elastic contact problem [Slimane 04] with the penalty-type stabilization term $\frac{1}{\varepsilon}(\nabla \cdot \boldsymbol{\sigma}_\varepsilon + f)$
- Uzawa-type iteration based on the minimization problem [Wang 09]
- a modified Uzawa algorithm, using the **BEM to compute the discrete $H^{\frac{1}{2}}$ -inner product** [Gatica 11]
- mixed (hemi-)VI for Stokes/NS with subgradient type bc [Han 21]

Motivation (**No discrete $H^{\frac{1}{2}}$ -inner product involved**)

- propose two equivalent mixed multiplier formulations for **(MVI)** and revisit the well-posedness
- investigate two penalty approaches and establish the well-posedness
- finite element approximation (RT0/P0 for $\boldsymbol{\sigma}/u$),
- the Active/Inactive set algorithms

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Proposition 1 (MVI) \Rightarrow (MP)

Let $(\sigma, u) \in \tilde{V} \times Q$ be the solution of (MVI). If $u \in H^1(\Omega)$ and $\sigma \cdot \mathbf{n} \in L^2(\Gamma_S)$, then (σ, u) satisfies (2) where the Signorini condition (2d) holds a.e. on Γ_S .

Theorem 1 (The well-posedness of (MVI))

Given $f \in Q$, (MVI) admits a unique solution $(\sigma, u) \in \tilde{V} \times Q$ with the estimate

$$\|\sigma\|_V + \|u\|_Q \leq C\|f\|_Q.$$

Remark: The unique existence of (MVI) has been proved by [Gatica 11]. We present an alternative proof revealing that σ can be seen as a projection onto a closed convex subspace of V .

Two equivalent mixed multiplier formulations

(MLM-I) Find $(\sigma, u, p) \in \tilde{V} \times Q \times \Lambda_{1/2}$ such that

$$(\sigma, \tau) + (u, \nabla \cdot \tau) - \langle \tau \cdot n, p \rangle_{\Gamma_S} = 0 \quad \forall \tau \in V, \quad (4a)$$

$$(\nabla \cdot \sigma, v) + (f, v) = 0 \quad \forall v \in Q, \quad (4b)$$

$$\langle \tau \cdot n - \sigma \cdot n, p \rangle_{\Gamma_S} \geq 0 \quad \forall \tau \in \tilde{V}; \quad (4c)$$

(MLM-II) Find $(\sigma, u, p) \in V \times Q \times \tilde{\Lambda}_{1/2}$ such that

$$(\sigma, \tau) + (u, \nabla \cdot \tau) - \langle \tau \cdot n, p \rangle_{\Gamma_S} = 0 \quad \forall \tau \in V, \quad (5a)$$

$$(\nabla \cdot \sigma, v) + (f, v) = 0 \quad \forall v \in Q, \quad (5b)$$

$$\langle \sigma \cdot n, q - p \rangle_{\Gamma_S} \geq 0 \quad \forall q \in \tilde{\Lambda}_{1/2}. \quad (5c)$$

$$\Lambda_{1/2} := H_{00}^{\frac{1}{2}}(\Gamma_S), \quad \tilde{\Lambda}_{1/2} := \{p \in \Lambda_{1/2} : p \geq 0 \text{ a.e.}\}, \quad \Lambda_{-1/2} := H^{-\frac{1}{2}}(\Gamma_S) = \Lambda_{1/2}^*.$$

Assume that $\sigma \cdot n \in L^2(\Gamma_S)$,

$$\begin{cases} (4c) \rightsquigarrow \sigma \cdot n = \max(0, \sigma \cdot n - \rho p) \text{ to ensure } \sigma \cdot n \geq 0 \text{ on } \Gamma_S, \\ (5c) \rightsquigarrow p = \max(0, p - \rho \sigma \cdot n) \text{ to ensure } u \geq 0 \text{ on } \Gamma_S. \end{cases}$$

$$\text{(MVI)} \Leftrightarrow \text{(MLM-I)} \Leftrightarrow \text{(MLM-II)}$$

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Mesh and the finite element spaces($RT_0/P_0 + P_0(e)$)

- Ω is a polygon ($d = 2$) or polyhedron ($d = 3$),
- \mathcal{T}_h : a quasi-uniform regular triangulation to Ω , $h = \max_{K \in \mathcal{T}_h} \text{diam}(K)$,
- \mathcal{E}_h : the set of elements faces, \mathcal{E}_{h,Γ_S} the mesh on Γ_S inherited from \mathcal{T}_h .

$$V_h := \{\boldsymbol{\tau}_h \in V : \boldsymbol{\tau}_h|_T \in RT_0(T)\}, \quad \tilde{V}_h := \{\boldsymbol{\tau}_h \in V_h : \boldsymbol{\tau}_h \cdot \mathbf{n} \geq 0 \text{ on } \Gamma_S\},$$

$$\Lambda_h := \{p_h \in \Lambda : p_h|_e \in P_0(e)\} = \{\boldsymbol{\tau}_h \cdot \mathbf{n}|_{\Gamma_S} : \boldsymbol{\tau}_h \in V_h\},$$

$$\tilde{\Lambda}_h := \{p_h \in \Lambda : p_h \geq 0\}, \quad Q_h := \{v_h \in Q : v_h|_T \in P_0(T)\}.$$

$P_0(T)$ and $P_0(e)$: the set of piecewise constant functions on T and e ,

$$RT_0(T) := \{(a, b)^\top + c(x, y)^\top : a, b, c \in \mathbb{R}, (x, y)^\top \in T\} \quad (d = 2),$$

$$RT_0(T) := \{(a, b, c)^\top + d(x, y, z)^\top : a, b, c, d \in \mathbb{R}, (x, y, z)^\top \in T\} \quad (d = 3).$$

(MVI)_h Find $(\sigma_h, u_h) \in \tilde{V}_h \times Q_h$ such that

$$(\sigma_h, \tau_h - \sigma_h) + (u_h, \nabla \cdot (\tau_h - \sigma_h)) \geq 0 \quad \forall \tau_h \in \tilde{V}_h, \quad (6a)$$

$$(\nabla \cdot \sigma_h, v_h) + (f, v_h) = 0 \quad \forall v_h \in Q_h. \quad (6b)$$

Theorem 2 (The well-posedness of (MVI)_h)

(MVI)_h admits a unique solution $(\sigma_h, u_h) \in \tilde{V}_h \times Q_h$ with the estimate

$$\|\sigma_h\|_V + \|u_h\|_Q \leq C\|f\|_Q. \quad (7)$$

Key: σ_h can be seen as a projection onto a closed convex subspace of V_h .

Theorem 3 (The error estimates)

Let (σ, u) and (σ_h, u_h) be the unique solutions of **(MVI)** and **(MVI)_h**, respectively. Let p and p_h satisfy (4a) and (9a), respectively. Assume that $\sigma \in H^1(\Omega)^d$, $\nabla \cdot \sigma \in H^1(\Omega)$, $\sigma \cdot n|_{\Gamma_S} \in H^1(\Gamma_S)$, $u \in H^1(\Omega)$ and $p \in H^1(\Gamma_S)$. Then, we have

$$\|\sigma - \sigma_h\|_V + \|u - u_h\|_Q + \|p - p_h\|_\Lambda \leq Ch. \quad (8)$$

Discretization of (MLM-I) and (MLM-II)

(MLM-I)_h Find $(\sigma_h, u_h, p_h) \in \tilde{V}_h \times Q_h \times \Lambda_h$ such that

$$(\sigma_h, \tau_h) + (u_h, \nabla \cdot \tau_h) - (p_h, \tau_h \cdot \mathbf{n})_{\Gamma_S} = 0 \quad \forall \tau_h \in V_h, \quad (9a)$$

$$(\nabla \cdot \sigma_h, v_h) + (f, v_h) = 0 \quad \forall v_h \in Q_h, \quad (9b)$$

$$(\tau_h \cdot \mathbf{n} - \sigma_h \cdot \mathbf{n}, p_h)_{\Gamma_S} \geq 0 \quad \forall \tau_h \in \tilde{V}_h. \quad (9c)$$

(MLM-II)_h Find $(\sigma_h, u_h, p_h) \in V_h \times Q_h \times \tilde{\Lambda}_h$ such that

$$(\sigma_h, \tau_h) + (u_h, \nabla \cdot \tau_h) - (p_h, \tau_h \cdot \mathbf{n})_{\Gamma_S} = 0 \quad \forall \tau_h \in V_h, \quad (10a)$$

$$(\nabla \cdot \sigma_h, v_h) + (f, v_h) = 0 \quad \forall v_h \in Q_h, \quad (10b)$$

$$(q_h - p_h, \sigma_h \cdot \mathbf{n})_{\Gamma_S} \geq 0 \quad \forall q_h \in \tilde{\Lambda}_h. \quad (10c)$$

Note that $\sigma_h \cdot \mathbf{n} \in L^2(\Gamma_S)$.

$$\left\{ \begin{array}{l} (9c) \rightsquigarrow \sigma_h \cdot \mathbf{n} = P_{\tilde{\Lambda}_h}(\sigma_h \cdot \mathbf{n} - \rho p_h) = \max(0, \sigma_h \cdot \mathbf{n} - \rho p_h) \\ \text{iteration } \sigma_h^{(n)} \cdot \mathbf{n} = P_{\tilde{\Lambda}_h}(\sigma_h^{(n-1)} \cdot \mathbf{n} - \rho p_h^{(n-1)}). \\ (10c) \rightsquigarrow p_h = P_{\tilde{\Lambda}_h}(p_h - \rho \sigma_h \cdot \mathbf{n}) = \max(0, p_h - \rho \sigma_h \cdot \mathbf{n}) \\ \text{iteration } p_h^{(n)} = P_{\tilde{\Lambda}_h}(p_h^{(n-1)} - \rho \sigma_h^{(n-1)} \cdot \mathbf{n}). \end{array} \right.$$

Remark (Two ways to define the projection: $H^{\frac{1}{2}}$ V.S. L^2)

1. (cf. [Gatica 11]) **Riesz representation** $R : (H_{00}^{\frac{1}{2}}(\Gamma_S))^* \rightarrow H_{00}^{\frac{1}{2}}(\Gamma_S)$.
 $\langle \tau \cdot n - \sigma \cdot n, p \rangle_{\Gamma_S} \geq 0 \rightsquigarrow (R\tau \cdot n - R\sigma \cdot n, p)_{H^{\frac{1}{2}}(\Gamma_S)} \geq 0$
 $\rightsquigarrow (R\tau \cdot n - R\sigma \cdot n, -\rho p + R\sigma \cdot n - R\sigma \cdot n)_{H^{\frac{1}{2}}(\Gamma_S)} \leq 0 \quad (\forall \rho > 0)$
 $\rightsquigarrow R\sigma \cdot n = P_{\tilde{\Lambda}_{1/2}}(-\rho p + R\sigma \cdot n) \quad (\text{Projection in } H^{\frac{1}{2}}(\Gamma_S)),$
 $\langle \sigma \cdot n, p - q \rangle_{\Gamma_S} \geq 0 \rightsquigarrow (R\sigma \cdot n, p - q)_{H^{\frac{1}{2}}(\Gamma_S)} \geq 0$
 $\rightsquigarrow (-\rho R\sigma \cdot n + p - p, p - q)_{H^{\frac{1}{2}}(\Gamma_S)} \leq 0 \quad (\forall \rho > 0)$
 $\rightsquigarrow p = P_{\tilde{\Lambda}_{1/2}}(-\rho R\sigma \cdot n + p).$

Discrete $H^{\frac{1}{2}}(\Gamma_S)$ inner product. Discretization of Riesz projection, and the discrete version of the $H^{\frac{1}{2}}(\Gamma_S)$ inner product, the discrete version of projection of $P_{\tilde{\Lambda}_{1/2}}$!

2. (Our proposal): No discrete Riesz projection and discrete $H^{\frac{1}{2}}(\Gamma_S)$ inner product. Just L^2 inner product in the discrete problem. But we can still prove the **well-posedness, error analysis (similar to [Gatica 11]) and efficient iteration scheme using projection.**

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Algorithm 1 The Active/Inactive set method based on **(MLM-I)_h**

(Step 1) Choose a $\rho > 0$. Find $(\boldsymbol{\sigma}_h^{(0)}, \mathbf{u}_h^{(0)}) \in V_h \times Q_h$ such that

$$\boldsymbol{\sigma}_h^{(0)} \cdot \mathbf{n} = 0 \quad \text{on } \Gamma_S,$$

$$(\boldsymbol{\sigma}_h^{(0)}, \boldsymbol{\tau}_h) + (\mathbf{u}_h^{(0)}, \nabla \cdot \boldsymbol{\tau}_h) = 0 \quad \forall \boldsymbol{\tau}_h \in V_h,$$

$$(\nabla \cdot \boldsymbol{\sigma}_h^{(0)}, v_h) + (f_h, v_h) = 0 \quad \forall v_h \in Q_h.$$

Set $p_h^{(0)} = 0$ and $n = 1$.

(Step 2) Decompose Γ_S into two subsets:

$$\Gamma_S^{1,(n)} := \{e \in \Gamma_S : \boldsymbol{\sigma}_h^{(n-1)} \cdot \mathbf{n} - \rho p_h^{(n-1)} < 0 \text{ on } e\} \quad (\text{The active set}),$$

$$\Gamma_S^{2,(n)} := \Gamma_S \setminus \Gamma_S^{1,(n)} \quad (\text{The inactive set}).$$

(Step 3) Find $(\boldsymbol{\sigma}_h^{(n)}, \mathbf{u}_h^{(n)}) \in V_h \times Q_h$ such that

$$\boldsymbol{\sigma}_h^{(n)} \cdot \mathbf{n} = 0 \quad \text{on } \Gamma_S^{1,(n)}, \quad (11a)$$

$$(\boldsymbol{\sigma}_h^{(n)}, \boldsymbol{\tau}_h) + (\mathbf{u}_h^{(n)}, \nabla \cdot \boldsymbol{\tau}_h) = 0 \quad \forall \boldsymbol{\tau}_h \in V_{h, \Gamma_S^{1,(n)}}, \quad (11b)$$

$$(\nabla \cdot \boldsymbol{\sigma}_h^{(n)}, v_h) + (f_h, v_h) = 0 \quad \forall v_h \in Q_h. \quad (11c)$$

(Step 4) Find $p_h^{(n)} \in \Lambda_h$ such that

$$(p_h^{(n)}, \boldsymbol{\tau}_h \cdot \mathbf{n})_{\Gamma_S} = (\boldsymbol{\sigma}_h^{(n)}, \boldsymbol{\tau}_h) + (\mathbf{u}_h^{(n)}, \nabla \cdot \boldsymbol{\tau}_h) \quad \forall \boldsymbol{\tau}_h \in V_h. \quad (12)$$

(Step 5) $n \rightarrow n + 1$. Iterate (Step2)-(Step4) until convergence.

Algorithm 2 The Active/Inactive set method based on **(MLM-II)_h**

(Step 1) Same to Algorithm 1.

(Step 2) Decompose Γ_S into two subsets:

$$\Gamma_S^{1,(n)} := \{e \in \Gamma_S : p_h^{(n-1)} - \rho \sigma_h^{(n-1)} \cdot \mathbf{n} < 0 \text{ on } e\} \quad (\text{The active set}),$$

$$\Gamma_S^{2,(n)} := \Gamma_S \setminus \Gamma_S^{1,(n)} \quad (\text{The inactive set}).$$

(Step 3) Find $(\sigma_h^{(n)}, u_h^{(n)}) \in V_h \times Q_h$ such that

$$\sigma_h^{(n)} \cdot \mathbf{n} = 0 \quad \text{on } \Gamma_S^{2,(n)}, \quad (13a)$$

$$(\sigma_h^{(n)}, \tau_h) + (u_h^{(n)}, \nabla \cdot \tau_h) = 0 \quad \forall \tau_h \in V_{h, \Gamma_S^{2,(n)}}, \quad (13b)$$

$$(\nabla \cdot \sigma_h^{(n)}, v_h) + (f_h, v_h) = 0 \quad \forall v_h \in Q_h. \quad (13c)$$

(Step 4) Find $p_h^{(n)} \in \Lambda_h$ such that

$$(p_h^{(n)}, \tau_h \cdot \mathbf{n})_{\Gamma_S} = (\sigma_h^{(n)}, \tau_h) + (u_h^{(n)}, \nabla \cdot \tau_h) \quad \forall \tau_h \in V_h. \quad (14)$$

(Step 5) $n \rightarrow n + 1$. Iterate (Step2)-(Step4) until convergence.

Remark: In Alg. 1, the projection $\sigma_h^{(n)} \cdot \mathbf{n} = \max(0, \sigma_h^{(n-1)} \cdot \mathbf{n} - \rho p_h^{(n-1)})$ is carried out only on $\Gamma_S^{1,(n)}$. From (11b) and (12), $p_h^{(n)} = 0$ on $\Gamma_S^{2,(n)}$.

In Alg. 2, according to (13b) and (14), we have $p_h^{(n)} = 0$ on $\Gamma_S^{1,(n)}$, same as the projection $p_h^{(n)} = \max(0, p_h^{(n-1)} - \rho \sigma_h^{(n-1)} \cdot \mathbf{n})$ on $\Gamma_S^{1,(n)}$.

Proposition 2

Assume that after n_0 iteration steps, the active(also inactive) sets becomes invariant, i.e., $\Gamma_S^{i,(n_0+1)} = \Gamma_S^{i,(n_0)}$ ($i = 1, 2$). Then Alg. 1 and Alg. 2 are convergent,

$$(\sigma_h^{(n+1)}, u_h^{(n+1)}, p_h^{(n+1)}) = (\sigma_h^{(n)}, u_h^{(n)}, p_h^{(n)}) \quad (\forall n \geq n_0).$$

Moreover, we have $\sigma_h^{(n)} \cdot \mathbf{n} \geq 0$, $p_h^{(n)} \geq 0$, $\sigma_h^{(n)} \cdot \mathbf{n} p_h^{(n)} = 0$ on Γ_S .

Remark: In Proposition 2, we assume the active (also inactive) sets are invariant after a couple of iterations, which is the same assumption referred to [Kashiwabara 13, Zhou 23]. In fact, this assumption cannot be prescribed. Nevertheless, in our numerical experiments, the algorithm performed well, and the iteration stopped after only a few steps.

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Example 1 (for Alg. 1 and 2)

Setting: Square region $\Omega = (0, 1)^2$, $f = 50(\frac{1}{2} - x)(\frac{1}{2} - y)$,
 $\Gamma_S = \Gamma_{S1} \cup \Gamma_{S2} = (0, 1) \times \{1\} \cup (0, 1) \times \{0\}$, $\Gamma_D = \partial\Omega \setminus \Gamma_S$.

The simulation results for Alg. 1 and 2 are almost identical. So we only display the result of Alg. 1.

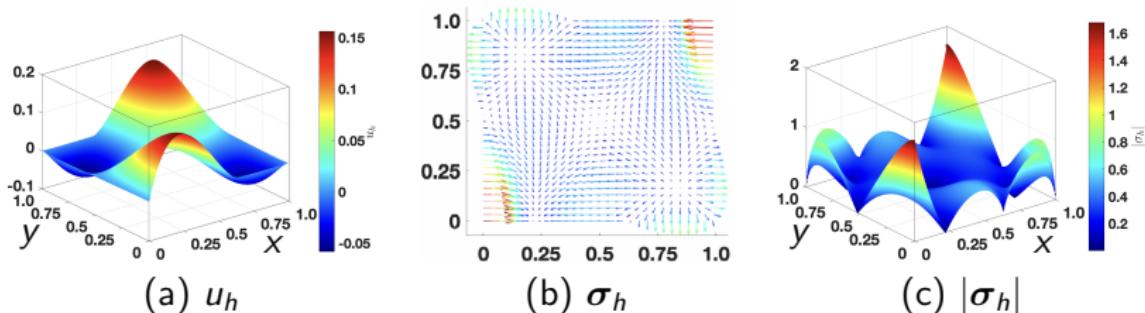
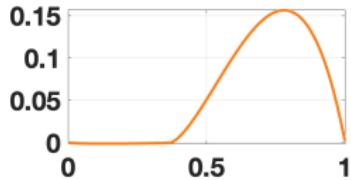
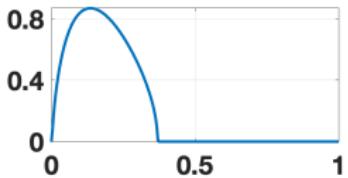


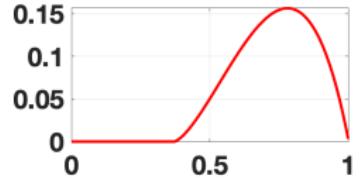
Figure 2: Example 1: The numerical solution (σ_h, u_h) .



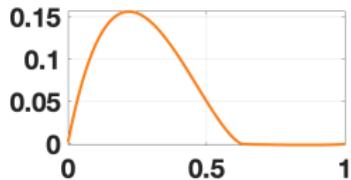
(a) u_h



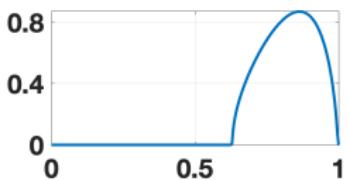
(b) $\sigma_h \cdot \mathbf{n}$



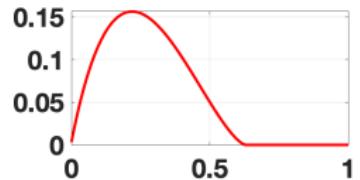
(c) p_h



(d) u_h



(e) $\sigma_h \cdot \mathbf{n}$



(f) p_h

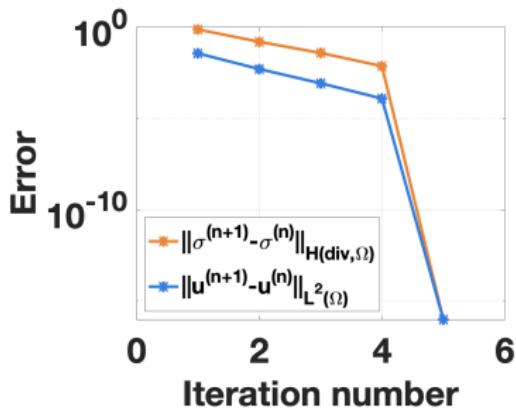
Figure 3: Example 1: $(u_h, \sigma_h \cdot \mathbf{n}, p_h)$ on Γ_{S1} (top) and Γ_{S2} (bottom) of $(\text{MLM-I})_h$ by using Alg. 1.

Table 1: Example 1: The errors $\|\boldsymbol{\sigma}_{ref} - \boldsymbol{\sigma}_h\|_V$ and $\|u_{ref} - u_h\|_Q$ for Alg. 1

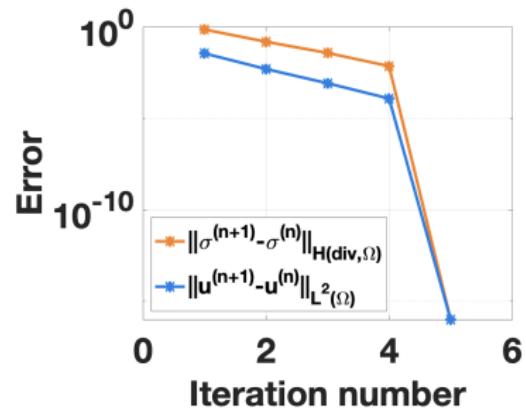
h	2^{-2}	2^{-3}	2^{-4}	2^{-5}	2^{-6}
$\ \boldsymbol{\sigma}_{ref} - \boldsymbol{\sigma}_h\ _V$	1.65	0.795	0.383	0.178	0.0764
Order	-	1.05	1.06	1.10	1.22
$\ u_{ref} - u_h\ _Q$	0.0465	0.0247	0.0121	0.00568	0.00244
Order	-	0.91	1.03	1.10	1.22

Table 2: Example 1: The errors $\|\boldsymbol{\sigma}_{ref} - \boldsymbol{\sigma}_h\|_V$ and $\|u_{ref} - u_h\|_Q$ for Alg. 2

h	2^{-2}	2^{-3}	2^{-4}	2^{-5}	2^{-6}
$\ \boldsymbol{\sigma}_{ref} - \boldsymbol{\sigma}_h\ _V$	1.65	0.795	0.383	0.178	0.0764
Order	-	1.05	1.06	1.10	1.22
$\ u_{ref} - u_h\ _Q$	0.0465	0.0247	0.0121	0.00568	0.00244
Order	-	0.91	1.03	1.10	1.22



(a) Alg. 1



(b) Alg. 2

Figure 4: Example 1: The iteration errors $\|\sigma_h^{(n+1)} - \sigma_h^{(n)}\|_V$ and $\|u_h^{(n+1)} - u_h^{(n)}\|_Q$ by Alg. 1 and 2 with $h = 2^{-6}$.

Example 2 (for Alg. 1 and 2)

Setting: L-shaped region $\Omega = (0, 2)^2 \setminus [1, 2]^2$, $f = 25(\frac{1}{2} - x)(\frac{1}{2} - y)$.
 $\Gamma_S = \Gamma_{S_1} \cup \Gamma_{S_2} = (0, 2) \times \{0\} \cup \{0\} \times (0, 2)$, $\Gamma_D = \partial\Omega \setminus \Gamma_S$.

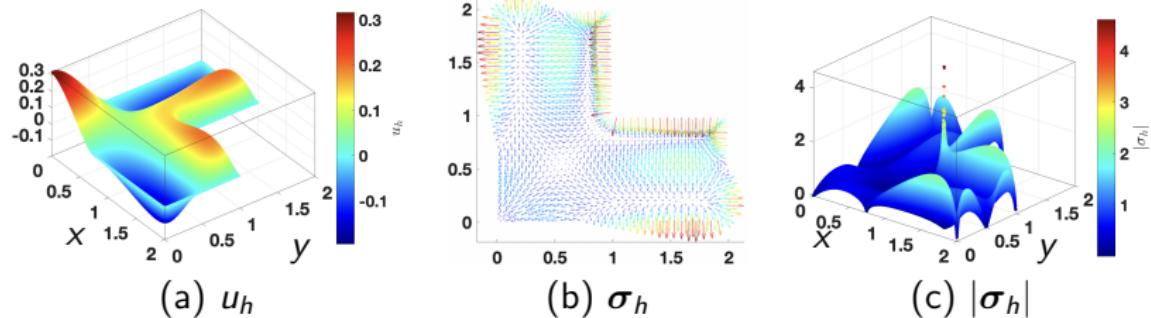
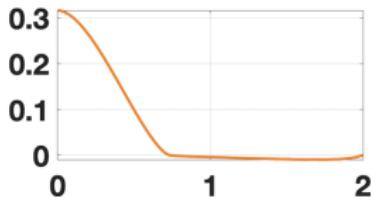
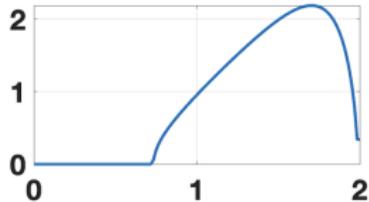


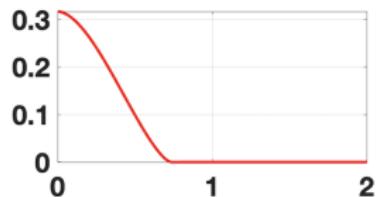
Figure 5: Example 2: The numerical solution (σ_h, u_h) .



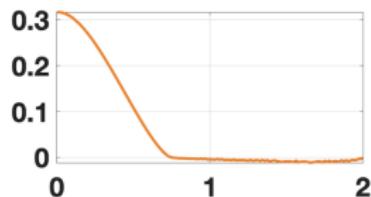
(a) u_h on Γ_{S1}



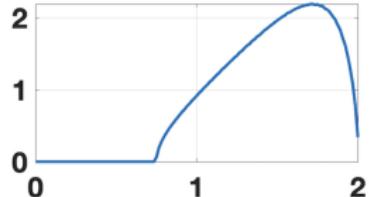
(b) $\sigma_h \cdot \mathbf{n}$ on Γ_{S1}



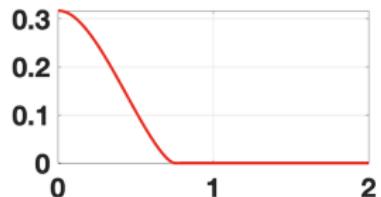
(c) p_h on Γ_{S1}



(d) u_h on Γ_{S2}



(e) $\sigma_h \cdot \mathbf{n}$ on Γ_{S2}



(f) p_h on Γ_{S2}

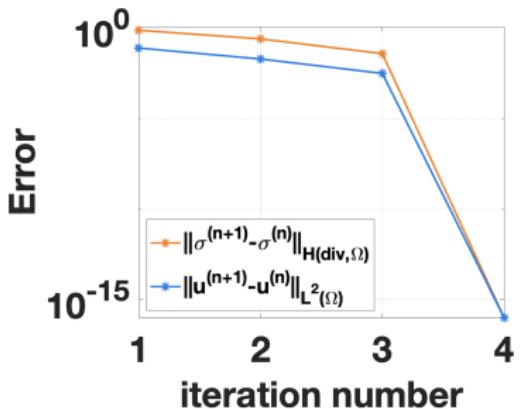
Figure 6: Example 2: $(u_h, \sigma_h \cdot \mathbf{n}, p_h)$ on Γ_{S1} (top) and Γ_{S2} (bottom) by using Alg. 1.

Table 3: Example 2: The errors $\|\boldsymbol{\sigma}_{ref} - \boldsymbol{\sigma}_h\|_V$ and $\|u_{ref} - u_h\|_Q$ for Alg. 1

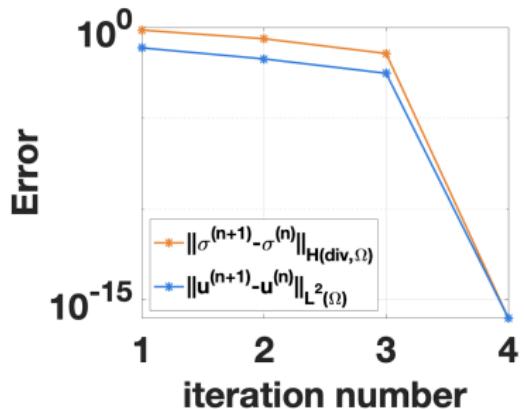
h	2^{-2}	2^{-3}	2^{-4}	2^{-5}	2^{-6}
$\ \boldsymbol{\sigma}_{ref} - \boldsymbol{\sigma}_h\ _V$	2.17	1.02	0.479	0.252	0.139
Order	-	1.14	1.30	0.86	0.86
$\ u_{ref} - u_h\ _Q$	0.114	0.0539	0.0229	0.0103	0.00511
Order	-	1.13	1.48	1.06	1.02

Table 4: Example 2: The errors $\|\boldsymbol{\sigma}_{ref} - \boldsymbol{\sigma}_h\|_V$ and $\|u_{ref} - u_h\|_Q$ for Alg. 2

h	2^{-2}	2^{-3}	2^{-4}	2^{-5}	2^{-6}
$\ \boldsymbol{\sigma}_{ref} - \boldsymbol{\sigma}_h\ _V$	2.17	1.02	0.479	0.252	0.139
Order	-	1.14	1.30	0.86	0.86
$\ u_{ref} - u_h\ _Q$	0.114	0.0539	0.0229	0.0103	0.00511
Order	-	1.13	1.48	1.06	1.02



(a) Alg. 1



(b) Alg. 2

Figure 7: Example 2: The iteration errors $\|\sigma_h^{(n+1)} - \sigma_h^{(n)}\|_V$ and $\|u_h^{(n+1)} - u_h^{(n)}\|_Q$ by Alg. 1 and 2 with $h = 2^{-3}$.

Example 3 (for Alg. 1 and 2)

Setting: Half-disk region $\Omega = \{(x, y) : x^2 + y^2 < 1, y < 0\}$, $f = 5x$.
 $\Gamma_S = \{(x, -\sqrt{1-x^2}) : x \in (-1, 1)\}$ and $\Gamma_D = \partial\Omega \setminus \Gamma_S$.

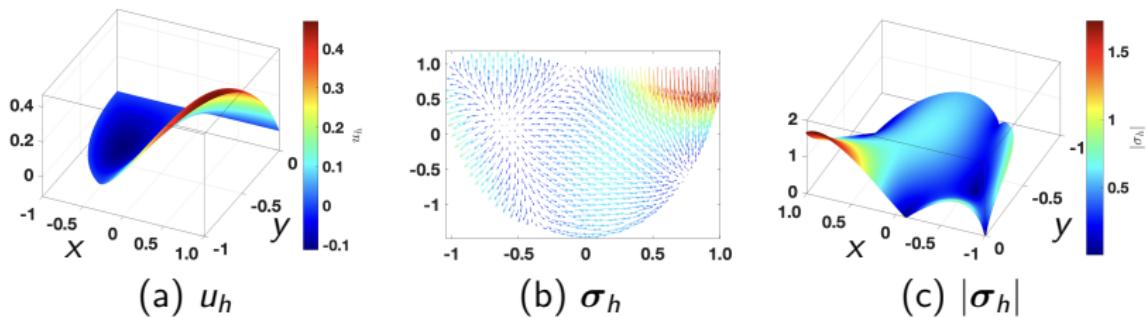


Figure 8: Example 3: The numerical solution (σ_h, u_h) .

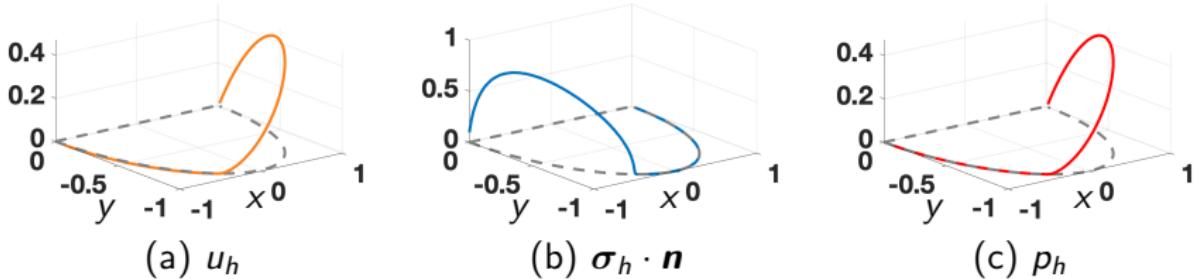


Figure 9: Example 3: $(u_h, \sigma_h \cdot \mathbf{n}, p_h)$ on Γ_s by using Alg. 1.

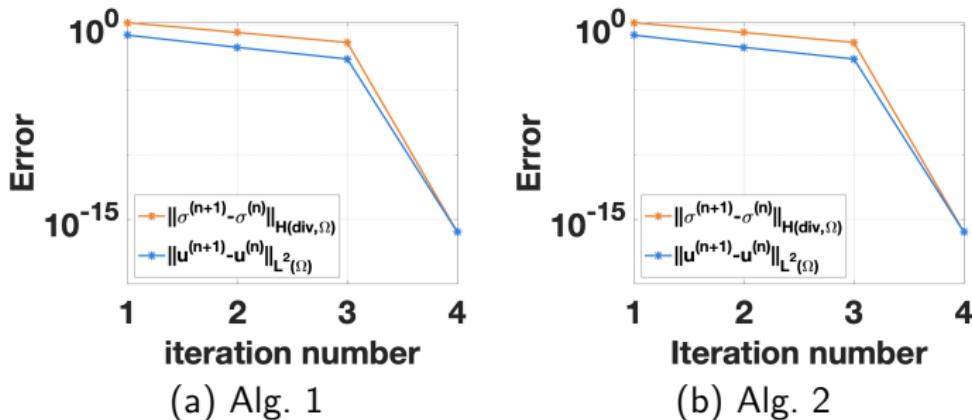


Figure 10: Example 3: The iteration errors $\|\sigma_h^{(n+1)} - \sigma_h^{(n)}\|_V$ and $\|u_h^{(n+1)} - u_h^{(n)}\|_Q$ by Alg. 1 and 2 with $h = 2^{-3}$.

Table 5: Example 3: The errors $\|\boldsymbol{\sigma}_{ref} - \boldsymbol{\sigma}_h\|_V$ and $\|u_{ref} - u_h\|_Q$ for Alg. 1

h	2^{-2}	2^{-3}	2^{-4}	2^{-5}	2^{-6}
$\ \boldsymbol{\sigma}_{ref} - \boldsymbol{\sigma}_h\ _V$	0.311	0.161	0.0720	0.0331	0.0160
Order	-	0.97	1.27	1.03	1.17
$\ u_{ref} - u_h\ _Q$	0.0528	0.0254	0.0111	0.00509	0.00252
Order	-	1.07	1.32	1.03	1.12

Table 6: Example 3: The errors $\|\boldsymbol{\sigma}_{ref} - \boldsymbol{\sigma}_h\|_V$ and $\|u_{ref} - u_h\|_Q$ for Alg. 2

h	2^{-3}	2^{-4}	2^{-5}	2^{-6}	2^{-7}
$\ \boldsymbol{\sigma}_{ref} - \boldsymbol{\sigma}_h\ _V$	0.311	0.161	0.0720	0.0331	0.0160
Order	-	0.97	1.27	1.03	1.17
$\ u_{ref} - u_h\ _Q$	0.0528	0.0254	0.0111	0.00509	0.00252
Order	-	1.07	1.32	1.03	1.12

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(MP_ε-I) Find $(\sigma_\varepsilon, u_\varepsilon, p_\varepsilon) \in V \times Q \times \Lambda_{1/2}$ such that:

$$(\sigma_\varepsilon, \tau) + (u_\varepsilon, \nabla \cdot \tau) - \langle \tau \cdot \mathbf{n}, p_\varepsilon \rangle_{\Gamma_S} = 0 \quad (\forall \tau \in V), \quad (15a)$$

$$(\nabla \cdot \sigma_\varepsilon, v) + (f, v) = 0 \quad (\forall v \in Q), \quad (15b)$$

$$\langle \sigma_\varepsilon \cdot \mathbf{n}, q \rangle_{\Gamma_S} - \frac{1}{\varepsilon} ([p_\varepsilon]_-, q)_{\Gamma_S} = 0 \quad (\forall q \in \Lambda_{1/2}); \quad (15c)$$

(MP_ε-II) Find $(\sigma_\varepsilon, u_\varepsilon) \in \hat{V} \times Q$ such that:

$$(\sigma_\varepsilon, \tau) + (u_\varepsilon, \nabla \cdot \tau) - \frac{1}{\varepsilon} ([\sigma_\varepsilon \cdot \mathbf{n}]_-, \tau \cdot \mathbf{n})_{\Gamma_S} = 0 \quad (\forall \tau \in \hat{V}), \quad (16a)$$

$$(\nabla \cdot \sigma_\varepsilon, v) + (f, v) = 0 \quad (\forall v \in Q). \quad (16b)$$

$$\hat{V} := \{\tau \in V : \tau \cdot \mathbf{n}|_{\Gamma_S} \in \Lambda = L^2(\Gamma_S)\} \subset V.$$

The strong forms of (MP_ε-I) and (MP_ε-II)

$$(\text{MP}_\varepsilon\text{-I}'): -\nabla \cdot \sigma_\varepsilon = f, \quad \sigma_\varepsilon - \nabla u_\varepsilon = 0 \quad \text{in } \Omega, \quad (17a)$$

$$u_\varepsilon = 0 \quad \text{on } \Gamma_D, \quad (17b)$$

$$\sigma_\varepsilon \cdot \mathbf{n} = \frac{1}{\varepsilon} [u_\varepsilon]_- \quad \text{on } \Gamma_S; \quad (17c)$$

$$(\text{MP}_\varepsilon\text{-II}'): -\nabla \cdot \sigma_\varepsilon = f, \quad \sigma_\varepsilon - \nabla u_\varepsilon = 0 \quad \text{in } \Omega, \quad (18a)$$

$$u_\varepsilon = 0 \quad \text{on } \Gamma_D, \quad (18b)$$

$$u_\varepsilon = \frac{1}{\varepsilon} [\sigma_\varepsilon \cdot \mathbf{n}]_- \quad \text{on } \Gamma_S. \quad (18c)$$

Well-posedness of $(\mathbf{MP}_\varepsilon\text{-I})$ and $(\mathbf{MP}_\varepsilon\text{-II})$

Theorem 4 (The well-posedness of $(\mathbf{MP}_\varepsilon\text{-I})$)

There exists a unique solution $(\sigma_\varepsilon, u_\varepsilon, p_\varepsilon) \in V \times Q \times \Lambda_{1/2}$ of $(\mathbf{MP}_\varepsilon\text{-I})$ satisfying

$$\|\sigma_\varepsilon\|_V^2 + \|u_\varepsilon\|_Q^2 + \frac{1}{\varepsilon} \|[p_\varepsilon]_-\|_{L^2(\Gamma_S)}^2 + \|p_\varepsilon\|_{\Lambda_{1/2}}^2 \leq C \|f\|_Q^2. \quad (19)$$

Theorem 5 (The well-posedness of $(\mathbf{MP}_\varepsilon\text{-II})$)

There exists a unique solution $(\sigma_\varepsilon, u_\varepsilon) \in \hat{V} \times Q$ of $(\mathbf{MP}_\varepsilon\text{-II})$ with the estimate

$$\|\sigma_\varepsilon\|_V^2 + \|u_\varepsilon\|_Q^2 + \frac{1}{\varepsilon} \|[\sigma_\varepsilon \cdot \mathbf{n}]_-\|_{\Lambda}^2 \leq C \|f\|_Q^2. \quad (20)$$

Remark

$$(\mathbf{MP}_\varepsilon\text{-I}) \Leftrightarrow (\mathbf{MP}_\varepsilon\text{-II})$$

To investigate the penalty error, we set

$$\mathbf{e}_{\sigma,\varepsilon} := \sigma - \sigma_\varepsilon, \quad e_{p,\varepsilon} := p - p_\varepsilon, \quad e_{u,\varepsilon} := u - u_\varepsilon.$$

Theorem 6 (The penalty error of $(\mathbf{MP}_\varepsilon\text{-I})$)

Let $(\sigma_\varepsilon, u_\varepsilon, p_\varepsilon)$ and (σ, u, p) be the unique solution of $(\mathbf{MP}_\varepsilon\text{-I})$ and $(\mathbf{MLM-I})$ respectively. Assume that $\sigma \cdot n \in \Lambda$. Then we have

$$\|\mathbf{e}_{\sigma,\varepsilon}\|_V + \|e_{u,\varepsilon}\|_Q + \|e_{p,\varepsilon}\|_{\Lambda_{1/2}} + \sqrt{\varepsilon} \|\mathbf{e}_{\sigma,\varepsilon} \cdot n\|_\Lambda \leq C\sqrt{\varepsilon} \|\sigma \cdot n\|_\Lambda. \quad (21)$$

Additionally, if $\sigma \cdot n \in \Lambda_{1/2}$, then we have

$$\|\mathbf{e}_{\sigma,\varepsilon}\|_V + \|e_{u,\varepsilon}\|_Q + \|e_{p,\varepsilon}\|_{\Lambda_{1/2}} + \sqrt{\varepsilon} \|\mathbf{e}_{\sigma,\varepsilon} \cdot n\|_\Lambda \leq C\varepsilon \|\sigma \cdot n\|_{\Lambda_{1/2}}. \quad (22)$$

Theorem 7 (The penalty error of $(\mathbf{MP}_\varepsilon\text{-II})$)

Assume that $\sigma \cdot n \in \Lambda$. Let $(\sigma_\varepsilon, u_\varepsilon)$ and (σ, u) be the unique solution of $(\mathbf{MP}_\varepsilon\text{-II})$ and (\mathbf{MVI}) . We have the error estimate

$$\|\mathbf{e}_{\sigma,\varepsilon}\|_V + \|e_{u,\varepsilon}\|_Q + \sqrt{\varepsilon} \|e_{p,\varepsilon}\|_\Lambda \leq C\sqrt{\varepsilon} \|p\|_\Lambda. \quad (23)$$

In addition, if $[\sigma_\varepsilon \cdot n]_- \in \Lambda_{1/2}$ (or equivalently, $p_\varepsilon \in \Lambda_{1/2}$), then we have

$$\|\mathbf{e}_{\sigma,\varepsilon}\|_V + \|e_{u,\varepsilon}\|_Q + \sqrt{\varepsilon} \|e_{p,\varepsilon}\|_\Lambda + \|e_{p,\varepsilon}\|_{\Lambda_{1/2}} \leq C\varepsilon \|p\|_{\Lambda_{-1/2}}. \quad (24)$$

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Mesh and the finite element spaces($RT_0/P_0 + P_0(e)$)

- Ω is a polygon ($d = 2$) or polyhedron ($d = 3$),
- \mathcal{T}_h : a quasi-uniform regular triangulation to Ω , $h = \max_{K \in \mathcal{T}_h} \text{diam}(K)$,
- \mathcal{E}_h : the set of elements faces, \mathcal{E}_{h,Γ_S} the mesh on Γ_S inherited from \mathcal{T}_h .

$$V_h := \{\boldsymbol{\tau}_h \in V : \boldsymbol{\tau}_h|_E \in RT_0(T) \ (\forall T \in \mathcal{T}_h)\},$$

$$Q_h := \{v_h \in Q : v_h|_T \in P_0(T) \ (\forall T \in \mathcal{T}_h)\},$$

$$\Lambda_h := \{p_h \in \Lambda : p_h|_e \in P_0(e) \ (\forall e \in \mathcal{E}_{h,\Gamma_S})\} = \{\boldsymbol{\tau}_h \cdot \mathbf{n}|_{\Gamma_S} : \boldsymbol{\tau}_h \in V_h\},$$

$P_0(T)$ and $P_0(e)$: the set of piecewise constant functions on T and e ,

$$RT_0(T) := \{(a, b)^\top + c(x, y)^\top : a, b, c \in \mathbb{R}, (x, y)^\top \in T\} \quad (d = 2),$$

$$RT_0(T) := \{(a, b, c)^\top + d(x, y, z)^\top : a, b, c, d \in \mathbb{R}, (x, y, z)^\top \in T\} \quad (d = 3).$$

Discretization of $(\mathbf{MP}_\varepsilon\text{-I})$ and $(\mathbf{MP}_\varepsilon\text{-II})$

$(\mathbf{MP}_\varepsilon\text{-I})_h$ Find $(\sigma_{\varepsilon h}, u_{\varepsilon h}, p_{\varepsilon h}) \in V_h \times Q_h \times \Lambda_h$ such that

$$(\sigma_{\varepsilon h}, \tau_h) + (u_{\varepsilon h}, \nabla \cdot \tau_h) - (\tau_h \cdot \mathbf{n}, p_{\varepsilon h})_{\Gamma_S} = 0 \quad (\forall \tau_h \in V_h), \quad (25a)$$

$$(\nabla \cdot \sigma_{\varepsilon h}, v_h) + (f_h, v_h) = 0 \quad (\forall v_h \in Q_h), \quad (25b)$$

$$\left(\sigma_{\varepsilon h} \cdot \mathbf{n} - \frac{1}{\varepsilon} [p_{\varepsilon h}]_-, q_h \right)_{\Gamma_S} = 0 \quad (\forall q_h \in \Lambda_h). \quad (25c)$$

$(\mathbf{MP}_\varepsilon\text{-II})_h$ Find $(\sigma_{\varepsilon h}, u_{\varepsilon h}) \in V_h \times Q_h$ such that

$$(\sigma_{\varepsilon h}, \tau_h) + (u_{\varepsilon h}, \nabla \cdot \tau_h) - \frac{1}{\varepsilon} ([\sigma_{\varepsilon h} \cdot \mathbf{n}]_-, \tau_h \cdot \mathbf{n})_{\Gamma_S} = 0 \quad (\forall \tau_h \in V_h), \quad (26a)$$

$$(\nabla \cdot \sigma_{\varepsilon h}, v_h) + (f_h, v_h) = 0 \quad (\forall v_h \in Q_h). \quad (26b)$$

Theorem 8 (The well-posedness of $(\mathbf{MP}_\varepsilon\text{-I})_h$)

There exists a unique solution $(\sigma_{\varepsilon h}, u_{\varepsilon h}, p_{\varepsilon h}) \in V_h \times Q_h \times \Lambda_h$ to $(\mathbf{MP}_\varepsilon\text{-I})_h$ with the estimate

$$\|\sigma_{\varepsilon h}\|_V + \|u_{\varepsilon h}\|_Q + \frac{1}{\sqrt{\varepsilon}} \| [p_{\varepsilon h}]_- \|_\Lambda + \|p_{\varepsilon h}\|_\Lambda \leq C \|f_h\|_Q. \quad (27)$$

Theorem 9 (The well-posedness of $(\mathbf{MP}_\varepsilon\text{-II})_h$)

There exists a unique solution $(\sigma_{\varepsilon h}, u_{\varepsilon h}) \in V_h \times Q_h$ to $(\mathbf{MP}_\varepsilon\text{-II})_h$ with the boundedness

$$\|\sigma_{\varepsilon h}\|_V + \|u_{\varepsilon h}\|_Q + \frac{1}{\sqrt{\varepsilon}} \| [\sigma_{\varepsilon h} \cdot \mathbf{n}]_- \|_\Lambda \leq C \|f_h\|_Q. \quad (28)$$

The error estimates

Set the notations:

$$\mathbf{e}_{\sigma,\varepsilon h} := \boldsymbol{\sigma} - \boldsymbol{\sigma}_{\varepsilon h} = (\boldsymbol{\sigma} - \Pi_h \boldsymbol{\sigma}) + (\Pi_h \boldsymbol{\sigma} - \boldsymbol{\sigma}_{\varepsilon h}) =: \boldsymbol{\rho}_h + \boldsymbol{\theta}_h,$$

$$e_{u,\varepsilon h} := u - u_{\varepsilon h}, \quad e_{p,\varepsilon h} := p - p_{\varepsilon h}.$$

Theorem 10 ((MP_ε-I)_h and (MLM-I))

Let $(\boldsymbol{\sigma}_{\varepsilon h}, u_{\varepsilon h}, p_{\varepsilon h})$ and $(\boldsymbol{\sigma}, u, p)$ be the solutions of **(MP_ε-I)_h** and **(MLM-I)** respectively. Assume that $\boldsymbol{\sigma} \in H^1(\Omega)^d$, $\nabla \cdot \boldsymbol{\sigma} \in H^1(\Omega)$, $u \in H^1(\Omega)$, $p \in H^1(\Gamma_S)$ and $\boldsymbol{\sigma} \cdot \mathbf{n} \in H^1(\Gamma_S)$. Then we have:

$$\|\mathbf{e}_{\sigma,\varepsilon h}\|_V + \|e_{u,\varepsilon h}\|_Q + \sqrt{\varepsilon} \|\mathbf{e}_{\sigma,\varepsilon h} \cdot \mathbf{n}\|_\Lambda \leq C(h + \sqrt{\varepsilon}), \quad (29a)$$

$$\|e_{p,\varepsilon h}\|_\Lambda \leq C(\sqrt{\varepsilon} + \sqrt{h}). \quad (29b)$$

Theorem 11 ((MP_ε-I)_h and (MP_ε-I))

Let $(\boldsymbol{\sigma}_{\varepsilon h}, u_{\varepsilon h}, p_{\varepsilon h})$ and $(\boldsymbol{\sigma}_\varepsilon, u_\varepsilon, p_\varepsilon)$ be the solutions of **(MP_ε-I)_h** and **(MP_ε-I)**, respectively. Assume that $\boldsymbol{\sigma}_\varepsilon \in H^1(\Omega)^d$, $\nabla \cdot \boldsymbol{\sigma}_\varepsilon \in H^1(\Omega)$, $u_\varepsilon \in H^1(\Omega)$, $p_\varepsilon \in H^1(\Gamma_S)$, $\boldsymbol{\sigma}_\varepsilon \cdot \mathbf{n} \in H^1(\Gamma_S)$. We have the error estimates:

$$\|\mathbf{e}_{\sigma,h}\|_V + \|e_{u,h}\|_Q + \sqrt{\varepsilon} \|\mathbf{e}_{\sigma,h} \cdot \mathbf{n}\|_\Lambda \leq C_\varepsilon h, \quad (30a)$$

$$\|e_{p,h}\|_\Lambda \leq C_\varepsilon \sqrt{h}. \quad (30b)$$

Theorem 12 ((MP_ε-II)_h and (MLM-II))

Let $(\sigma_{\varepsilon h}, u_{\varepsilon h})$ and (σ, u, p) be the solutions of (MP_ε-II)_h and (MLM-II) respectively. Assume that

$\sigma \in H^1(\Omega)^d$, $\nabla \cdot \sigma \in H^1(\Omega)$, $u \in H^1(\Omega)$, $p \in H^1(\Gamma_S)$ and $\sigma \cdot n \in H^1(\Gamma_S)$. Then we have the error estimate

$$\|\mathbf{e}_{\sigma, \varepsilon h}\|_V + \|e_{u, \varepsilon h}\|_Q \leq C(h + \sqrt{\varepsilon}), \quad (31a)$$

$$\|e_{p, \varepsilon h}\|_\Lambda \leq C(\sqrt{\varepsilon} + \sqrt{h}). \quad (31b)$$

Theorem 13 ((MP_ε-II)_h and (MP_ε-II))

Let $(\sigma_{\varepsilon h}, u_{\varepsilon h})$ and $(\sigma_\varepsilon, u_\varepsilon)$ be the solutions of (MP_ε-II)_h and (MP_ε-II), respectively. Under the same assumption of Theorem 11, we have the error estimates:

$$\|\mathbf{e}_{\sigma, h}\|_V + \|e_{u, h}\|_Q \leq C_\varepsilon h, \quad (32a)$$

$$\|e_{p, h}\|_\Lambda \leq C_\varepsilon \sqrt{h}. \quad (32b)$$

Remark

- ▶ Convergence rates are similar to the non-mixed method
- ▶ The idea of **prime-dual method** can be used to design the iteration algorithms.

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Algorithm 3 An iteration algorithm for $(\mathbf{MP}_{\varepsilon}-\mathbf{I})_h$

(Step 1) Find $(\boldsymbol{\sigma}_{\varepsilon h}^{(0)}, u_{\varepsilon h}^{(0)}) \in V_h \times Q_h$ such that

$$\boldsymbol{\sigma}_{\varepsilon h}^{(0)} \cdot \mathbf{n} = 0 \quad \text{on } \Gamma_S, \quad (33a)$$

$$(\boldsymbol{\sigma}_{\varepsilon h}^{(0)}, \boldsymbol{\tau}_h) + (u_{\varepsilon h}^{(0)}, \nabla \cdot \boldsymbol{\tau}_h) = 0 \quad (\forall \boldsymbol{\tau}_h \in V_{h, \Gamma_S}), \quad (33b)$$

$$(\nabla \cdot \boldsymbol{\sigma}_{\varepsilon h}^{(0)}, v_h) + (f_h, v_h) = 0 \quad (\forall v_h \in Q_h). \quad (33c)$$

Find $p_{\varepsilon h}^{(0)} \in \Lambda_h$ such that

$$(p_{\varepsilon h}^{(0)}, \boldsymbol{\tau}_h \cdot \mathbf{n})_{\Gamma_S} = (\boldsymbol{\sigma}_{\varepsilon h}^{(0)}, \boldsymbol{\tau}_h) + (u_{\varepsilon h}^{(0)}, \nabla \cdot \boldsymbol{\tau}_h) \quad (\forall \boldsymbol{\tau}_h \in V_h). \quad (34)$$

Set $n = 1$.

(Step 2) Divide Γ_S into two parts:

$$\Gamma_S^{1,(n)} := \{e \in \Gamma_S : p_{\varepsilon h}^{(n-1)} > 0\} \text{ (The active set),} \quad \Gamma_S^{2,(n)} := \Gamma_S \setminus \Gamma_S^{1,(n)} \text{ (The inactive set).}$$

Find $(\boldsymbol{\sigma}_{\varepsilon h}^{(n)}, u_{\varepsilon h}^{(n)}) \in V_h \times Q_h$ such that

$$\boldsymbol{\sigma}_{\varepsilon h}^{(n)} \cdot \mathbf{n} = 0 \quad \text{on } \Gamma_S^{1,(n)}, \quad (35a)$$

$$(\boldsymbol{\sigma}_{\varepsilon h}^{(n)}, \boldsymbol{\tau}_h) + (u_{\varepsilon h}^{(n)}, \nabla \cdot \boldsymbol{\tau}_h) + \varepsilon(\boldsymbol{\tau}_h \cdot \mathbf{n}, \boldsymbol{\sigma}_{\varepsilon h}^{(n)} \cdot \mathbf{n})_{\Gamma_S^{2,(n)}} = 0 \quad (\forall \boldsymbol{\tau}_h \in V_{h, \Gamma_S^{1,(n)}}), \quad (35b)$$

$$(\nabla \cdot \boldsymbol{\sigma}_{\varepsilon h}^{(n)}, v_h) + (f_h, v_h) = 0 \quad (\forall v_h \in Q_h). \quad (35c)$$

(Step 3) Modify the value of $\boldsymbol{\sigma}_{\varepsilon h}^{(n)}$ such that

$$\boldsymbol{\sigma}_{\varepsilon h}^{(n)} \cdot \mathbf{n} \geq 0 \quad \text{on } \Gamma_S^{2,(n)}. \quad (36)$$

Find $p_{\varepsilon h}^{(n)} \in \Lambda_h$ such that

$$(p_{\varepsilon h}^{(n)}, \boldsymbol{\tau}_h \cdot \mathbf{n})_{\Gamma_S} = (\boldsymbol{\sigma}_{\varepsilon h}^{(n)}, \boldsymbol{\tau}_h) + (u_{\varepsilon h}^{(n)}, \nabla \cdot \boldsymbol{\tau}_h) \quad (\forall \boldsymbol{\tau}_h \in V_h). \quad (37)$$

(Step 4) If $\Gamma_S^{1,(n)} = \Gamma_S^{1,(n-1)}$ ($n > 1$), then the algorithm is finished. Otherwise, $n \rightarrow n + 1$ and then iterate (Step2) and (Step3).

Algorithm 4 An iteration algorithm for $(\mathbf{MP}_\varepsilon\text{-II})_h$

(Step 1) Find $(\boldsymbol{\sigma}_{\varepsilon h}^{(0)}, \mathbf{u}_{\varepsilon h}^{(0)}) \in V_h \times Q_h$ such that

$$(\boldsymbol{\sigma}_{\varepsilon h}^{(0)}, \boldsymbol{\tau}_h) + (\mathbf{u}_{\varepsilon h}^{(0)}, \nabla \cdot \boldsymbol{\tau}_h) = 0 \quad (\forall \boldsymbol{\tau}_h \in V_h), \quad (38a)$$

$$(\nabla \cdot \boldsymbol{\sigma}_{\varepsilon h}^{(0)}, v_h) + (f_h, v_h) = 0 \quad (\forall v_h \in Q_h). \quad (38b)$$

Set $n = 1$.

(Step 2) Divide Γ_S into two parts:

$$\Gamma_S^{1,(n)} := \{e \in \Gamma_S : \boldsymbol{\sigma}_{\varepsilon h}^{(n-1)} \cdot \mathbf{n} > 0\} \text{ (The active set),}$$

$$\Gamma_S^{2,(n)} := \Gamma_S \setminus \Gamma_S^{1,(n)} \text{ (The inactive set).}$$

Find $(\boldsymbol{\sigma}_{\varepsilon h}^{(n)}, \mathbf{u}_{\varepsilon h}^{(n)}) \in V_h \times Q_h$ such that

$$(\boldsymbol{\sigma}_{\varepsilon h}^{(n)}, \boldsymbol{\tau}_h) + (\mathbf{u}_{\varepsilon h}^{(n)}, \nabla \cdot \boldsymbol{\tau}_h) + \frac{1}{\varepsilon} (\boldsymbol{\sigma}_{\varepsilon h}^{(n)} \cdot \mathbf{n}, \boldsymbol{\tau}_h \cdot \mathbf{n})_{\Gamma_S^{2,(n)}} = 0 \quad (\forall \boldsymbol{\tau}_h \in V_h),$$

$$(\nabla \cdot \boldsymbol{\sigma}_{\varepsilon h}^{(n)}, v_h) + (f_h, v_h) = 0 \quad (\forall v_h \in Q_h).$$

(Step 3) If $\Gamma_S^{1,(n)} = \Gamma_S^{1,(n-1)}$ ($n > 1$), then the algorithm is finished. Otherwise, $n \rightarrow n + 1$ and then go to (Step2).

Proposition 3

Assume that after n_0 iteration steps, the active(also inactive) sets becomes invariant, i.e., $\Gamma_S^{i,(n_0+1)} = \Gamma_S^{i,(n_0)}$ ($i = 1, 2$). Then Alg. 3 and Alg. 4 are convergent,

$$(\boldsymbol{\sigma}_{\varepsilon h}^{(n+1)}, \mathbf{u}_{\varepsilon h}^{(n+1)}) = (\boldsymbol{\sigma}_{\varepsilon h}^{(n)}, \mathbf{u}_{\varepsilon h}^{(n)}) = (\boldsymbol{\sigma}_{\varepsilon h}, \mathbf{u}_{\varepsilon h}).$$

Moreover, we have $\boldsymbol{\sigma}_{\varepsilon h}^{(n)} \cdot \mathbf{n} \geq 0$, $\mathbf{u}_{\varepsilon h}^{(n)} \geq 0$, $\boldsymbol{\sigma}_{\varepsilon h}^{(n)} \cdot \mathbf{n} \mathbf{u}_{\varepsilon h}^{(n)} = 0$ on Γ_S .

Outline

1. PDE model & Background

2. The fully mixed method

Lagrange multiplier formulations

Finite element approximate

The Active/Inactive set algorithms

Numerical experiments

3. The mixed penalty method

The mixed penalty forms

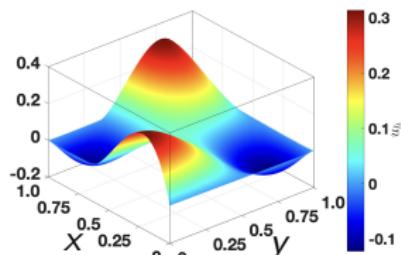
Finite element approximate

The Active/Inactive set algorithms

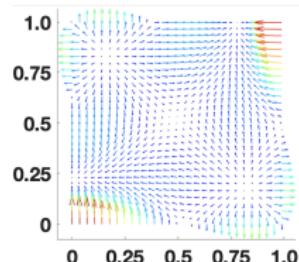
Numerical experiments

Example 4 (for Alg. 3 and 4)

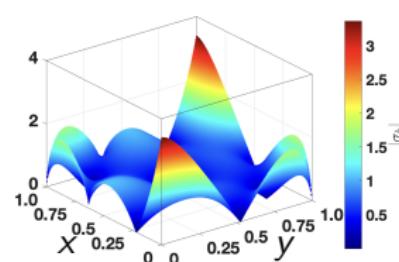
Setting: Square region $\Omega = (0, 1)^2$, $f = 100(\frac{1}{2} - x)(\frac{1}{2} - y)$,
 $\Gamma_S = \Gamma_{S1} \cup \Gamma_{S2}$, $\Gamma_{S1} = (0, 1) \times \{1\}$, $\Gamma_{S2} = \{0\} \times (0, 1)$, $\Gamma_D = \partial\Omega \setminus \Gamma_S$.



(a) u_h



(b) σ_h



(c) $|\sigma_h|$

Figure 11: Example 4: The numerical solution (σ_h, u_h) .

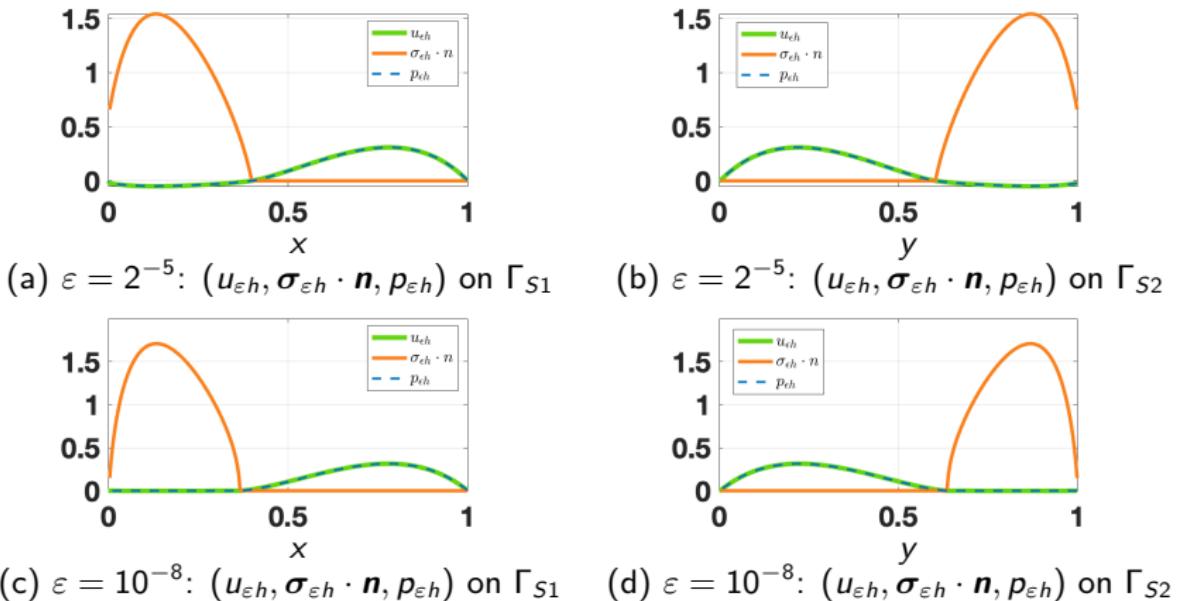
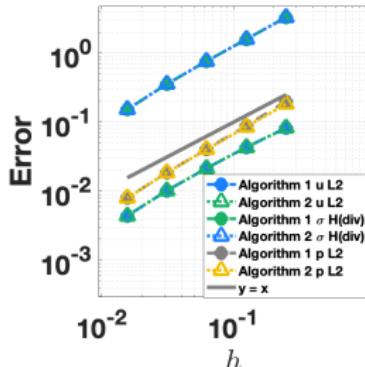
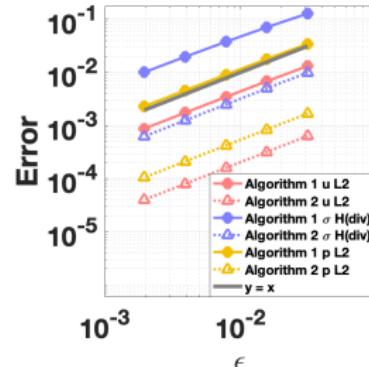


Figure 12: Example 4: $(u_{\varepsilon h}, \sigma_{\varepsilon h} \cdot n, p_{\varepsilon h})$ on Γ_S with $(h, \varepsilon) = (2^{-8}, 2^{-5})$ (see (a)–(b)) and $(h, \varepsilon) = (2^{-8}, 10^{-8})$ (see (c)–(d)) of $(\mathbf{MP}_{\varepsilon}-\mathbf{I})_h$ by using Algorithm 3.



(a) Fixed $\varepsilon = 10^{-8}$



(b) Fixed $h = 2^{-8}$

Figure 13: Example 4: The errors $\|\sigma_{\varepsilon \text{ref}} - \sigma_{\varepsilon h}\|_V$, $\|u_{\varepsilon \text{ref}} - u_{\varepsilon h}\|_Q$ and $\|p_{\varepsilon \text{ref}} - p_{\varepsilon h}\|_\Lambda$ computed by Algorithms 3 and 4.

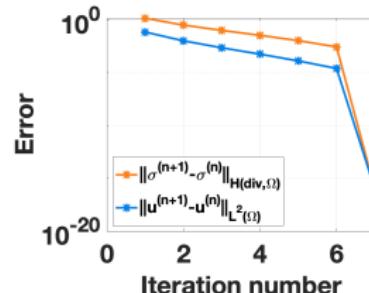
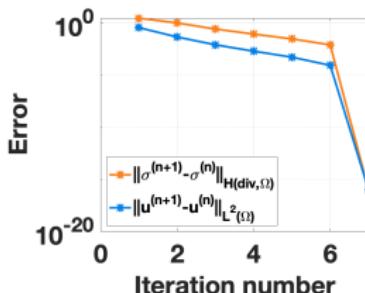


Figure 14: Example 4: The iteration errors $\|\sigma_{\varepsilon h}^{(n+1)} - \sigma_{\varepsilon h}^{(n)}\|_V$ and $\|u_{\varepsilon h}^{(n+1)} - u_{\varepsilon h}^{(n)}\|_Q$ of Algorithms 3 (left) and 4 (right) with $(h, \varepsilon) = (2^{-8}, 10^{-8})$.

Example 5 (for Alg. 3 and 4)

Setting: L-shaped region $\Omega = (0, 2)^2 \setminus [1, 2]^2$, $f = 6 \sin(\pi(x - \frac{1}{3})) \cos(\frac{y}{3})$.
 $\Gamma_S = \Gamma_{S1} \cup \Gamma_{S2}$, $\Gamma_{S1} = (0, 2) \times \{0\}$, $\Gamma_{S2} = \{0\} \times (0, 2)$.

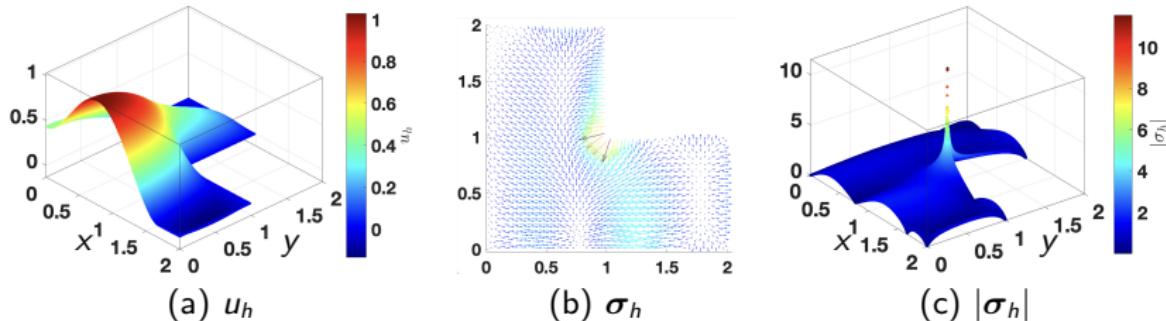


Figure 15: Example 5: The numerical solution (σ_h, u_h) .

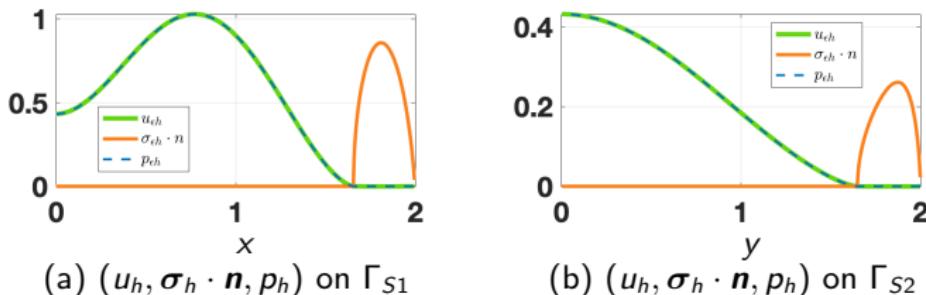
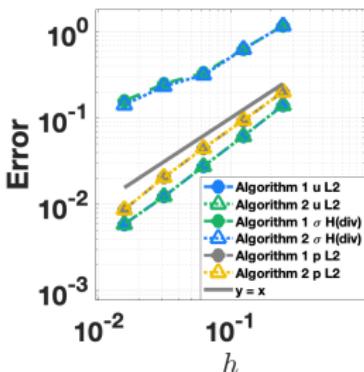
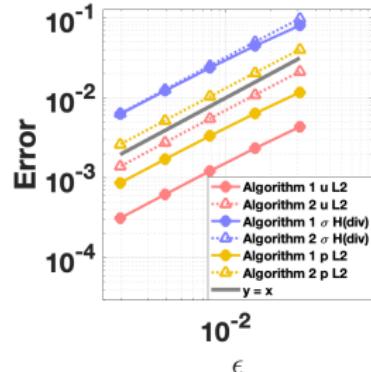


Figure 16: Example 5: $(u_h, \sigma_h \cdot n, p_h)$ on Γ_S .

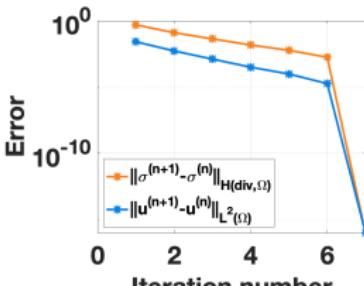


(a) Fixed $\varepsilon = 10^{-8}$

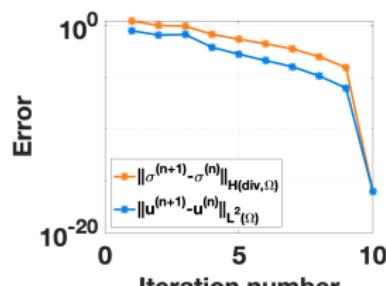


(b) Fixed $h = 2^{-8}$

Figure 17: Example 5: The errors $\|\sigma_{\varepsilon \text{ref}} - \sigma_{\varepsilon h}\|_V$, $\|u_{\varepsilon \text{ref}} - u_{\varepsilon h}\|_Q$ and $\|p_{\varepsilon \text{ref}} - p_{\varepsilon h}\|_\Lambda$ computed by Algorithms 3 and 4.



(a) Algorithm 3



(b) Algorithm 4

Figure 18: Example 5: The iteration errors $\|\sigma_{\varepsilon h}^{(n+1)} - \sigma_{\varepsilon h}^{(n)}\|_V$ and $\|u_{\varepsilon h}^{(n+1)} - u_{\varepsilon h}^{(n)}\|_Q$ of Algorithms 3 and 4 with $(h, \varepsilon) \equiv (2^{-8}, 10^{-8})$.

Example 6 (for Alg. 3 and 4)

Setting: $\Omega = \{(x, y) : \frac{1}{2} < \sqrt{x^2 + y^2} < 1\}$, $f = 5 \sin(\frac{1}{2}\pi x)$.

$\Gamma_S = \{(x, y) : \sqrt{x^2 + y^2} = \frac{1}{2}\}$ and $\Gamma_D = \partial\Omega \setminus \Gamma_S$.

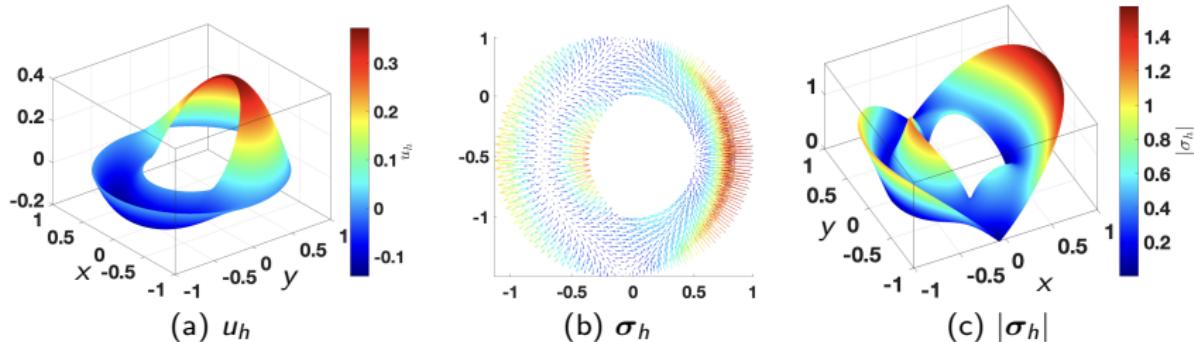


Figure 19: Example 6: The numerical solution (σ_h, u_h) .

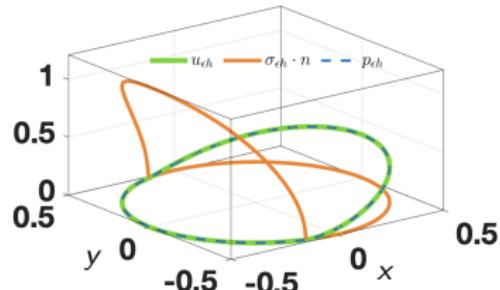
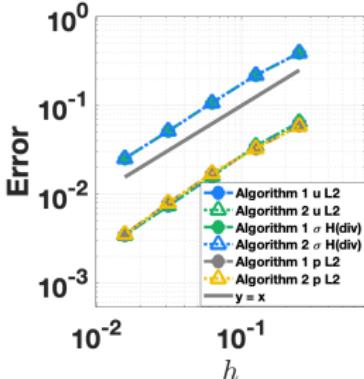
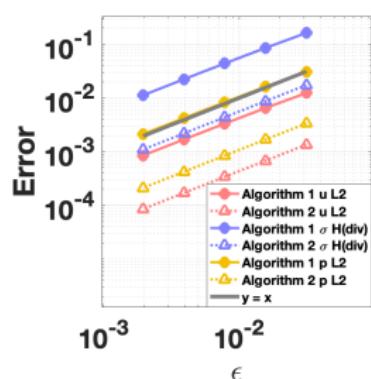


Figure 20: Example 6: $(u_h, \sigma_h \cdot n, p_h)$ on Γ_S .

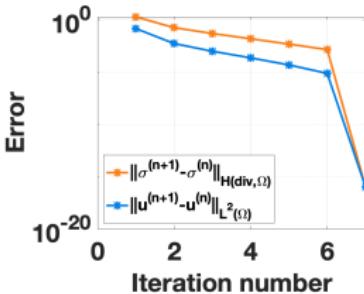


(a) Fixed $\varepsilon = 10^{-8}$

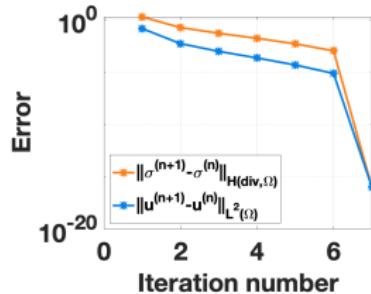


(b) Fixed $h = 2^{-8}$

Figure 21: Example 6: The errors $\|\sigma_{\varepsilon \text{ref}} - \sigma_{\varepsilon h}\|_V$, $\|u_{\varepsilon \text{ref}} - u_{\varepsilon h}\|_Q$ and $\|p_{\varepsilon \text{ref}} - p_{\varepsilon h}\|_\Lambda$ computed by Algorithms 3 and 4.



(a) Algorithm 3



(b) Algorithm 4

Figure 22: Example 6: The iteration errors $\|\sigma_{\varepsilon h}^{(n+1)} - \sigma_{\varepsilon h}^{(n)}\|_V$ and $\|u_{\varepsilon h}^{(n+1)} - u_{\varepsilon h}^{(n)}\|_Q$ of Algorithms 3 and 4 with $(h, \varepsilon) \equiv (2^{-8}, 10^{-8})$.

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THANKS!