

A BUBBLE-ENRICHED POLYTOPAL METHOD FOR CONTACT MECHANICS

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References for this presentation

- Design of the method, application to poromechanics with Coulomb friction: [Droniou et al., 2024a].
- Analysis for Tresca friction: [Droniou et al., 2024b].

Outline

1 Mixed-dimensional models

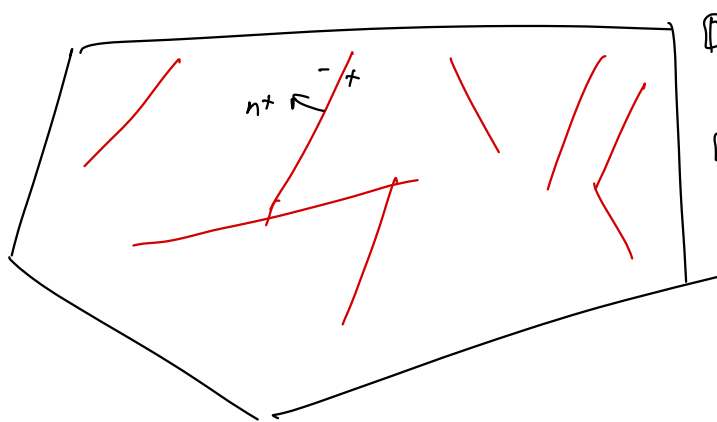
2 Scheme

3 Analysis

4 Numerical results

Matrix and fracture network

Notations: Γ fracture, two sides \pm with outward normals \mathbf{n}^\pm , jump $[[\cdot]]$, tangential $\boldsymbol{\tau}$ and normal \mathbf{n} component of vectors.



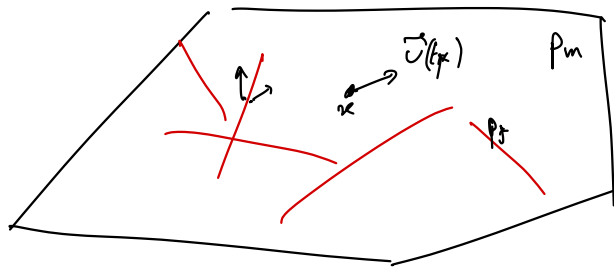
$$\begin{aligned} [[v]] &= v^+ - v^- \\ [[v]]_n &= (v^+ - v^-) \cdot n \end{aligned}$$

Poromechanics

Unknowns: **displacement \mathbf{u}** in matrix (discontinuous at fractures), pressure p_m in matrix, pressure p_f in fracture.

Physics: quasi-static contact-mechanics for \mathbf{u} , Darcy law for p_m , Poiseuille law for p_f .

Contact: without friction, with Tresca friction, with Coulomb friction...



Mechanics with Tresca friction: strong form

$$\begin{cases} -\operatorname{div} \boldsymbol{\sigma}(\mathbf{u}) = \mathbf{f} & \text{on } \Omega \setminus \bar{\Gamma}, \\ \gamma_n^+ \boldsymbol{\sigma}(\mathbf{u}) + \gamma_n^- \boldsymbol{\sigma}(\mathbf{u}) = \mathbf{0} & \text{on } \Gamma, \\ T_n(\mathbf{u}) \leq 0, \llbracket \mathbf{u} \rrbracket_n \leq 0, \llbracket \mathbf{u} \rrbracket_n T_n(\mathbf{u}) = 0 & \text{on } \Gamma, \\ |\mathbf{T}_\tau(\mathbf{u})| \leq g & \text{on } \Gamma, \\ \mathbf{T}_\tau(\mathbf{u}) \cdot \llbracket \mathbf{u} \rrbracket_\tau + g |\llbracket \mathbf{u} \rrbracket_\tau| = 0 & \text{on } \Gamma, \end{cases}$$

$\boldsymbol{\sigma}(\mathbf{u}) \mathbf{n}^+$

→

where:

- $\boldsymbol{\sigma}(\mathbf{u}) = 2\mu\boldsymbol{\epsilon}(\mathbf{u}) + \lambda\operatorname{div}\mathbf{u}$ with λ, μ Lamé coefficients.
- $T_n(\mathbf{u}) = \gamma_n^+ \boldsymbol{\sigma}(\mathbf{u}) \cdot \mathbf{n}^+$ and $\mathbf{T}_\tau(\mathbf{u}) = \gamma_n^+ \boldsymbol{\sigma}(\mathbf{u}) - T_n(\mathbf{u})\mathbf{n}^+$ tangential and normal surface tractions.
- g Tresca threshold.

Mechanics with Tresca friction: spaces

Spaces:

- For the displacement:

$$\mathbf{U}_0 = H_0^1(\Omega \setminus \bar{\Gamma})^d$$



$u=0$ on \mathcal{D}

Mechanics with Tresca friction: spaces

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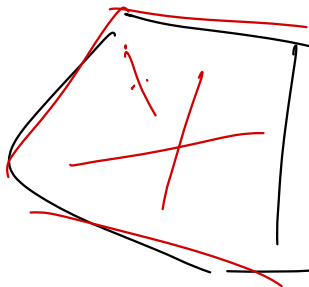
- For the displacement:

$$\mathbf{U}_0 = H_0^1(\Omega \setminus \bar{\Gamma})^d$$

- Space of jumps:

$$H_{0,j}^{1/2}(\Gamma) = \{ \llbracket \mathbf{v} \rrbracket : \mathbf{v} \in \mathbf{U}_0 \}$$

with $H_{0,j}^{-1/2}(\Gamma)$ dual space, and $\langle \cdot, \cdot \rangle_\Gamma$ duality pairing.



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with $H_{0,j}^{-1/2}(\Gamma)$ dual space, and $\langle \cdot, \cdot \rangle_\Gamma$ duality pairing.

- Lagrange multiplier (for contact inequalities):

$$\mathbf{C}_f = \left\{ \underline{\boldsymbol{\mu}} \in H_{0,j}^{-1/2}(\Gamma) : \langle \underline{\boldsymbol{\mu}}, \mathbf{v} \rangle_\Gamma \leq \langle g, |\mathbf{v}_\tau| \rangle_\Gamma \quad \forall \mathbf{v} \in H_{0,j}^{1/2}(\Gamma) \text{ with } \underline{\mathbf{v}} \cdot \mathbf{n}^+ \leq 0 \right\}.$$

Formally equivalent to $\underline{\boldsymbol{\mu}}_n \geq 0$ and $|\underline{\boldsymbol{\mu}}_\tau| \leq g$.

$\mu_n \geq 0$

Mechanics with Tresca friction: mixed-variational formulation

Weak formulation: find $\mathbf{u} \in \mathbf{U}_0$ and $\lambda \in \mathbf{C}_f$ s.t.

$$\int_{\Omega} \sigma(\mathbf{u}) : \epsilon(\mathbf{v}) + \langle \lambda, [[\mathbf{v}]] \rangle_{\Gamma} = \int_{\Omega} \mathbf{f} \cdot \mathbf{v} \quad \forall \mathbf{v} \in \mathbf{U}_0,$$

$$\langle \mu - \lambda, [[\mathbf{u}]] \rangle_{\Gamma} \leq 0 \quad \forall \mu \in \mathbf{C}_f$$

Note: $\lambda = -\gamma_n^+ \sigma(\mathbf{u}) = \gamma_n^- \sigma(\mathbf{u})$.

Handwritten notes and annotations:

- $H_{0,1}^{1/2}$ (circled)
- $\langle \lambda, [[\sigma]] \rangle = \ell(\sigma)$ (highlighted in yellow)
- $H_{0,1}^{-1/2}$ (with arrows pointing to the yellow highlight and the norm expression below)
- $\forall \sigma$ (with arrow pointing to the yellow highlight)
- $\| \lambda \|$ (circled)
- $H_{0,1}^{-1/2}$ (circled)

Outline

1 Mixed-dimensional models

2 Scheme

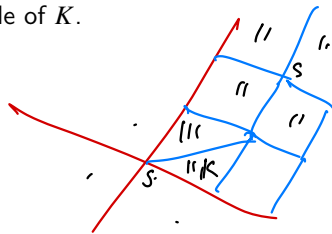
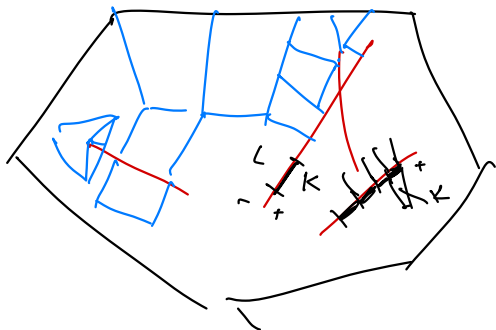
3 Analysis

4 Numerical results

Mesh

Polytopal mesh, compatible with fractures

- $\mathcal{M}, \mathcal{F}, \mathcal{V}$ cells, faces and vertices. \mathcal{X}_z entities \mathcal{X} on z .
- $\mathcal{F}_{\Gamma, K}^+$ faces of K on positive side of fracture.
- For $s \in \mathcal{V}$, \mathcal{K}_s : set of cells in \mathcal{M}_s on the same side of K .
- If $\sigma \in \mathcal{F}_{\Gamma}$: K on positive side, L on negative side.

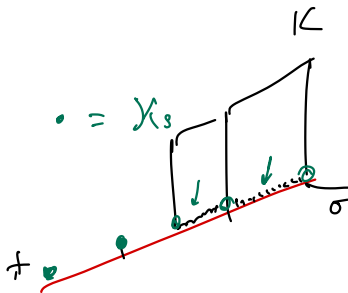
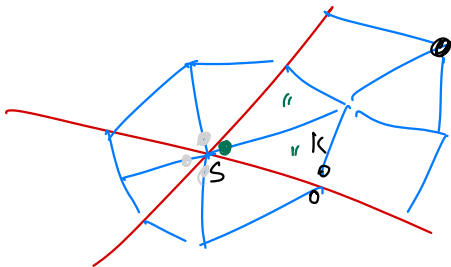


Discrete spaces

Displacement: nodal unknowns (discontinuous across fracture) and one bubble on each fracture face (positive side).

$$\underline{U}_{0,\mathcal{D}} = \left\{ \mathbf{v}_{\mathcal{D}} = ((\mathbf{v}_{\mathcal{K}s})_{K \in \mathcal{M}, s \in \mathcal{V}_K}, (\mathbf{v}_{K\sigma})_{K \in \mathcal{M}, \sigma \in \mathcal{F}_{\Gamma,K}^+}) : \right.$$

$$\left. \begin{aligned} \mathbf{v}_{\mathcal{K}s} \in \mathbb{R}^d, \mathbf{v}_{K\sigma} \in \mathbb{R}^d, \mathbf{v}_{\mathcal{K}s} = 0 \text{ if } s \in \mathcal{V}^{\text{ext}} \\ \mathbf{v}_{\mathcal{K}s} = \mathbf{v}_{\mathcal{L}s} \text{ if } K, L \in \mathcal{M}_s \text{ are on the same side of } \Gamma \end{aligned} \right\}.$$



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Lagrange multipliers: piecewise constant on fracture faces.

$$\mathbf{M}_{\mathcal{D}} = \{ \lambda_{\mathcal{D}} \in L^2(\Gamma)^d : \lambda_{\sigma} := (\lambda_{\mathcal{D}})|_{\sigma} \text{ is constant for all } \sigma \in \mathcal{F}_{\Gamma} \}.$$

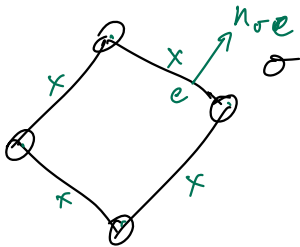
Discrete dual cone:

$$\mathbf{C}_{\mathcal{D}} = \{ \lambda_{\mathcal{D}} \in \mathbf{M}_{\mathcal{D}} : \lambda_{\mathcal{D},\mathbf{n}} \geq 0, |\lambda_{\mathcal{D},\boldsymbol{\tau}}| \leq g \} \subset \mathbf{C}_f.$$

Reconstructions in $\mathbf{U}_{0,\mathcal{D}}$: faces

- From nodes, reconstruct edge values and use them to define the face gradient:

$$\nabla^{K\sigma} \mathbf{v}_{\mathcal{D}} = \frac{1}{|\sigma|} \sum_{e=s_1s_2 \in \mathcal{E}_{\sigma}} |e| \frac{\mathbf{v}_{\mathcal{K}S_1} + \mathbf{v}_{\mathcal{K}S_2}}{2} \otimes \mathbf{n}_{\sigma e}.$$



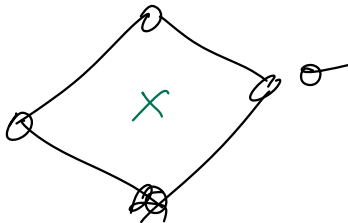
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- Reconstruct face averaged value from nodes, and face displacement:

$$\bar{\mathbf{v}}_{K\sigma} = \sum_{s \in \mathcal{V}_\sigma} \omega_s^\sigma \mathbf{v}_{\mathcal{K}S} \quad \text{and} \quad \underbrace{\Pi^{K\sigma} \mathbf{v}_{\mathcal{D}}(\mathbf{x})}_{\text{green underline}} = \underbrace{\nabla^{K\sigma} \mathbf{v}_{\mathcal{D}}(\mathbf{x} - \bar{\mathbf{x}}_\sigma)}_{\text{green underline}} + \bar{\mathbf{v}}_{K\sigma} \quad \forall \mathbf{x} \in \sigma.$$



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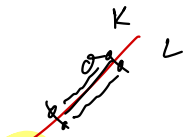
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- Jump reconstructions:

$$\llbracket \mathbf{v}_{\mathcal{D}} \rrbracket_\sigma = \frac{1}{|\sigma|} \int_\sigma (\underbrace{\Pi^{K\sigma} \mathbf{v}_{\mathcal{D}} - \Pi^{L\sigma} \mathbf{v}_{\mathcal{D}}}_{\text{jump}}) + \bar{\mathbf{v}}_{K\sigma} = \bar{\mathbf{v}}_{K\sigma} - \bar{\mathbf{v}}_{L\sigma} + \mathbf{v}_{K\sigma}.$$



Reconstructions in $\mathbf{U}_{0,\mathcal{D}}$: cells

Same principles...

- Using reconstructed face values **and bubble**, define the cell gradient:

$$\nabla^K \mathbf{v}_{\mathcal{D}} = \frac{1}{|K|} \sum_{\sigma \in \mathcal{F}_K} |\sigma| \underbrace{(\bar{\mathbf{v}}_{K\sigma})}_{\text{bubble}} \otimes \mathbf{n}_{K\sigma} + \sum_{\sigma \in \mathcal{F}_{\Gamma,K}^+} \frac{|\sigma|}{|K|} \mathbf{v}_{K\sigma} \otimes \mathbf{n}_{K\sigma}.$$

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Global reconstructions: $[[\cdot]]_{\mathcal{D}}, \nabla^{\mathcal{D}}, \Pi_{\mathcal{D}}, \epsilon_{\mathcal{D}}, \text{div}_{\mathcal{D}}, \sigma_{\mathcal{D}}.$

Interpolator

Displacement: $\mathcal{I}_{\mathbf{U}_{0,\mathcal{D}}} : C_0^0(\Omega \setminus \Gamma) \rightarrow \mathbf{U}_{0,\mathcal{D}}$ defined by:

$$(\mathcal{I}_{\mathbf{U}_{0,\mathcal{D}}} \mathbf{v})_{\mathcal{K}s} = \mathbf{v}|_{\mathcal{K}}(\mathbf{x}_s) \quad \forall \mathcal{K} \in \mathcal{M}, \forall s \in \mathcal{V}_{\mathcal{K}},$$

$$(\mathcal{I}_{\mathbf{U}_{0,\mathcal{D}}} \mathbf{v})_{\mathcal{K}\sigma} = \frac{1}{|\sigma|} \int_{\sigma} (\gamma^{K\sigma} \mathbf{v} - \Pi^{K\sigma}(\mathcal{I}_{\mathbf{U}_{0,\mathcal{D}}} \mathbf{v})) \quad \forall \mathcal{K} \in \mathcal{M}, \forall \sigma \in \mathcal{F}_{\Gamma, \mathcal{K}}^+.$$

Lagrange multiplier: $\mathcal{I}_{\mathbf{M}_{\mathcal{D}}} : L^2(\Gamma) \rightarrow \mathbf{M}_{\mathcal{D}}$ defined by

$$(\mathcal{I}_{\mathbf{M}_{\mathcal{D}}} \lambda)_{\sigma} = \frac{1}{|\sigma|} \int_{\sigma} \lambda \quad \forall \sigma \in \mathcal{F}_{\Gamma}.$$

Scheme

Find $\mathbf{u}_D \in \mathbf{U}_{0,D}$ and $\lambda_D \in \mathbf{M}_D$ s.t.

$$\int_{\Omega} \sigma_D(\mathbf{u}_D) : \epsilon_D(\mathbf{v}_D) + \sum_{K \in \mathcal{M}} (2\mu_K + \lambda_K) S_K(\mathbf{u}_D, \mathbf{v}_D) + \int_{\Gamma} \lambda_D \cdot \llbracket \mathbf{v}_D \rrbracket_D = \int_{\Omega} \mathbf{f} \cdot \tilde{\Pi}^D \mathbf{v}_D \quad \forall \mathbf{v}_D \in \mathbf{U}_{0,D}$$

$$\int_{\Gamma} (\mu_D - \lambda_D) \cdot \llbracket \mathbf{u}_D \rrbracket_D \leq 0 \quad \forall \mu_D \in \mathbf{M}_D,$$

where

$$S_K(\mathbf{u}_D, \mathbf{v}_D) = h_K^{d-2} \sum_{s \in \mathcal{V}_K} \left(\mathbf{u}_{Ks} - \Pi^K \mathbf{u}_D(\mathbf{x}_s) \right) \cdot \left(\mathbf{v}_{Ks} - \Pi^K \mathbf{v}_D(\mathbf{x}_s) \right) + h_K^{d-2} \sum_{\sigma \in \mathcal{F}_{\Gamma,K}^+} \mathbf{u}_{K\sigma} \cdot \mathbf{v}_{K\sigma}.$$

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Theorem (Error estimate)

If $\mathbf{u} \in H^2(\mathcal{M})$ and $\lambda \in H^1(\mathcal{F}_\Gamma)$, then

$$\|\nabla^{\mathcal{D}} \mathbf{u}_{\mathcal{D}} - \nabla \mathbf{u}\|_{L^2(\Omega \setminus \bar{\Gamma})} + \|\lambda_{\mathcal{D}} - \lambda\|_{-1/2, \Gamma} \lesssim C_{\mathbf{u}, \widehat{h}_{\mathcal{D}}}$$

- $\|\cdot\|_{-1/2, \Gamma}$ discrete $H^{-1/2}$ -like seminorm, defined later.

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- Error estimate comes from a more abstract version that only requires $\lambda \in L^2(\Gamma)$.

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- $\|\cdot\|_{-1/2, \Gamma}$ discrete $H^{-1/2}$ -like seminorm, defined later.
- Error estimate comes from a more abstract version that only requires $\lambda \in L^2(\Gamma)$.
- Error analysis based on **consistency** and **stability**.

Discrete Korn inequality

Discrete H^1 -norm:

$$\|\mathbf{v}_{\mathcal{D}}\|_{1,\mathcal{D}}^2 := \sum_{K \in \mathcal{M}} \left(\|\nabla^K \mathbf{v}_{\mathcal{D}}\|_{L^2(K)}^2 + S_K(\mathbf{v}_{\mathcal{D}}, \mathbf{v}_{\mathcal{D}}) \right).$$

Theorem (Discrete Korn inequality)

For all $\mathbf{v} \in \mathbf{U}_{0,\mathcal{D}}$,

$$\|\mathbf{v}_{\mathcal{D}}\|_{1,\mathcal{D}}^2 \lesssim \|\mathbb{E}_{\mathcal{D}}(\mathbf{v}_{\mathcal{D}})\|_{L^2(\Omega \setminus \bar{\Gamma})}^2 + \sum_{K \in \mathcal{M}} S_K(\mathbf{v}_{\mathcal{D}}, \mathbf{v}_{\mathcal{D}}).$$

Discrete Korn inequality: idea of proof

- Start from the \mathbb{P}^1 -conforming reconstruction $\mathcal{R}_{av,h}\mathbf{v}_h$ on a triangular submesh (averaging cell values at each node) and

$$\|\nabla_h(\mathbf{v}_h - \mathcal{R}_{av,h}\mathbf{v}_h)\|_{L^2}^2 \lesssim \sum_{\sigma \in \mathcal{F}} h_\sigma^{-1} \|[[\mathbf{v}_h]]_\sigma\|_{L^2(\sigma)}^2.$$



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- Using Korn for $\mathcal{R}_{av,h}\mathbf{v}_h$ gives

$$\|\nabla_h \mathbf{v}_h\|_{L^2}^2 \lesssim \|\epsilon_h(\mathbf{v}_h)\|_{L^2}^2 + \sum_{\sigma \in \mathcal{F}} h_\sigma^{-1} \|[[\mathbf{v}_h]]_\sigma\|_{L^2(\sigma)}^2.$$

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- Do the same procedure but building $\mathcal{R}_{av,h}\mathbf{v}_h$ from **averages over cell values the same side of Γ** :

$$\|\nabla_h \mathbf{v}_h\|_{L^2}^2 \lesssim \|\mathfrak{C}_h(\mathbf{v}_h)\|_{L^2}^2 + \sum_{\sigma \in \mathcal{F} \setminus \mathcal{F}_\Gamma} h_\sigma^{-1} \|[[\mathbf{v}_h]]_\sigma\|_{L^2(\sigma)}^2.$$

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- Conclude by proving that, for $\sigma = K|L$ not in \mathcal{F}_Γ ,

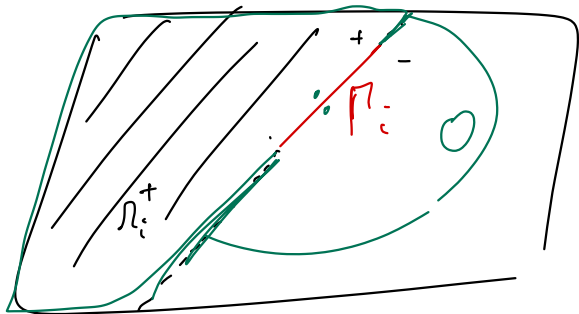
$$h_\sigma^{-1} \|[[\mathbf{v}_h]]_\sigma\|_{L^2(\sigma)}^2 \lesssim S_K(\mathbf{v}_\mathcal{D}, \mathbf{v}_\mathcal{D}) + S_L(\mathbf{v}_\mathcal{D}, \mathbf{v}_\mathcal{D}).$$

Discrete inf-sup property I

Definition (Discrete $H^{-1/2}(\Gamma)$ -norm)

For $\lambda_{\mathcal{D}} \in \mathbf{M}_{\mathcal{D}}$:

$$\|\lambda_{\mathcal{D}}\|_{-1/2, \Gamma} = \sum_{i \in I} \|\lambda_{\mathcal{D}}\|_{-1/2, \Gamma_i} \quad \text{with} \quad \|\lambda_{\mathcal{D}}\|_{-1/2, \Gamma_i} = \sup_{\mathbf{v}_i \in H^1(\Omega_i^+; \Gamma_i)^d \setminus \{0\}} \frac{\int_{\Gamma_i} \lambda_{\mathcal{D}} \mathbf{v}_i}{\|\mathbf{v}_i\|_{H^1(\Omega_i^+)}}$$



Discrete inf-sup property II

Theorem (Discrete inf-sup condition)

$$\sup_{\mathbf{v}_{\mathcal{D}} \in \mathbf{U}_{0,\mathcal{D}} \setminus \{0\}} \frac{\int_{\Gamma} \boldsymbol{\lambda}_{\mathcal{D}} \cdot \llbracket \mathbf{v}_{\mathcal{D}} \rrbracket_{\mathcal{D}}}{\|\mathbf{v}_{\mathcal{D}}\|_{1,\mathcal{D}}} \gtrsim \|\boldsymbol{\lambda}_{\mathcal{D}}\|_{-1/2,\Gamma} \quad \forall \boldsymbol{\lambda}_{\mathcal{D}} \in \mathbf{M}_{\mathcal{D}}.$$

Discrete inf-sup property III

Tools:

- Clément-like H^1 -stable interpolator $\mathcal{I}_{U_0, \mathcal{D}}^{i, a}$, adapted to fractures.

Discrete inf-sup property IV

Tools:

- Clément-like H^1 -stable interpolator $\mathcal{I}_{U_{0,D}}^{i,a}$, adapted to fractures.
- Fortin property for jump:

$$\int_{\sigma} \llbracket \mathcal{I}_{U_{0,D}}^{i,a} \mathbf{v} \rrbracket_{\sigma} = \begin{cases} \int_{\sigma} (\gamma^{K\sigma} \mathbf{v} - \llbracket \mathcal{I}_{U_{0,D}}^{i,a} \mathbf{y} \rrbracket_{\sigma}) & \text{if } \sigma \in \mathcal{F}_{\Gamma_i, K}^+, \\ 0 & \text{if } \sigma \in \mathcal{F}_{\Gamma_i, K}^+ \setminus \mathcal{F}_{\Gamma_i, K}^+. \end{cases}$$

Consequence: if $\mathbf{v}_i \in H^1(\Omega_i^+; \Gamma_i)$, $\tilde{\mathbf{v}}_i$ extension of \mathbf{v}_i by 0 and $\mathbf{v}_D = \mathcal{I}_{U_{0,D}}^{i,a} \tilde{\mathbf{v}}_i$,

$$\int_{\Gamma} \lambda_D \cdot \llbracket \mathbf{v}_D \rrbracket_D = \sum_{\sigma \in \mathcal{F}_{\Gamma}} \lambda_{\sigma} \cdot \int_{\sigma} \llbracket \mathcal{I}_{U_{0,D}}^{i,a} \tilde{\mathbf{v}}_i \rrbracket_{\sigma} = \int_{\Gamma_i} \lambda_D \cdot \mathbf{v}_i.$$

Outline

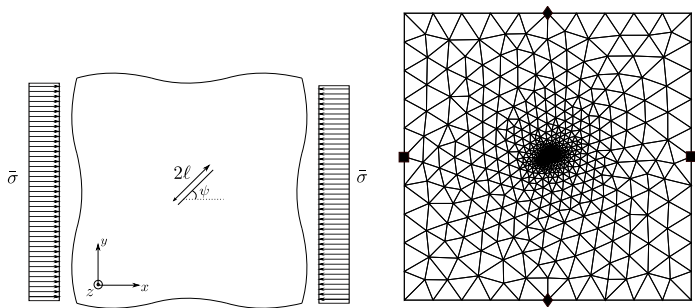
1 Mixed-dimensional models

2 Scheme

3 Analysis

4 Numerical results

2D domain with fracture under compression I



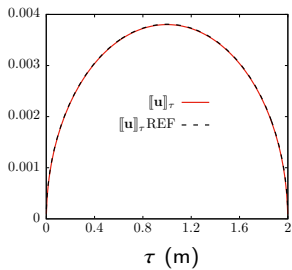
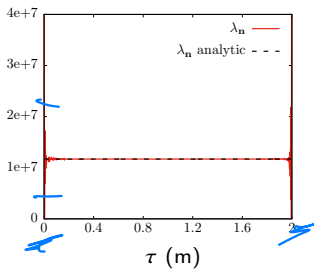
Analytical solution (τ coordinate along fracture):

$$\lambda_n = \sigma \sin^2(\psi),$$

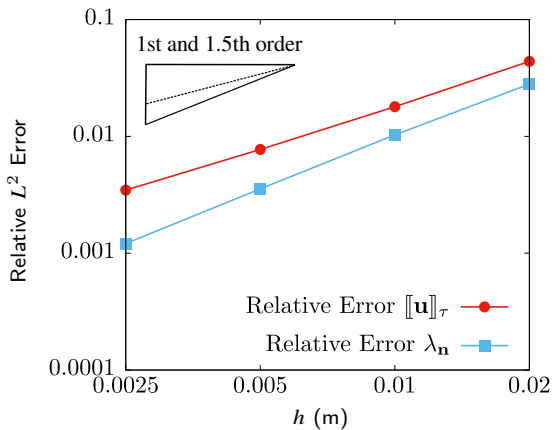
$$\|[\mathbf{u}]_{\tau}\| = \frac{4(1-\nu)}{E} \sigma \sin(\psi) \left(\cos(\psi) - \frac{g}{\lambda_n} \sin(\psi) \right) \sqrt{\ell^2 - (\ell^2 - \tau^2)}.$$

$\psi = \pi/9$, $2\ell = 2$ m, $F = 1/\sqrt{3}$ (so $g = \lambda_n/F$), $E = 25$ GPa and $\nu = 0.25$.

2D domain with fracture under compression II



2D domain with fracture under compression III



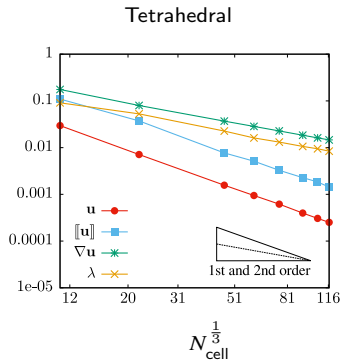
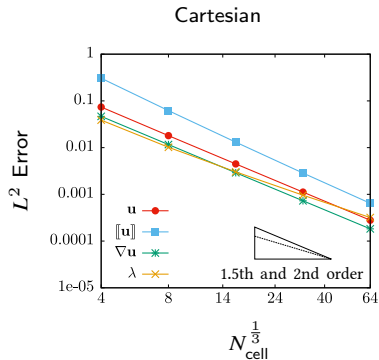
3D manufactured solution I

Setting:

- $\Omega = (-1, 1)^3$, $\Gamma = \{0\} \times (-1, 1)^2$.
- $g = 1$, $\mu = \lambda = 1$.
- Explicit analytical solution such that:
 - sticky-contact for $z < 0$ ($[[u]]_{\mathbf{n}} = 0$, $[[u]]_{\boldsymbol{\tau}} = 0$)
 - slippery-contact for $z > 0$ ($[[u]]_{\mathbf{n}} = 0$, $|[[u]]_{\boldsymbol{\tau}}| > 0$)
- Cartesian, tetrahedral and hexahedral (randomly perturbed Cartesian) meshes.

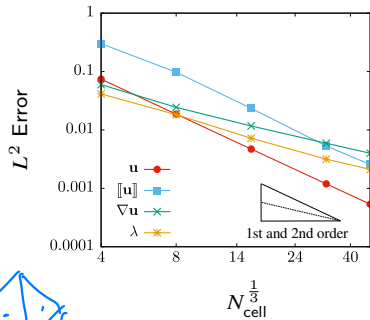


3D manufactured solution II

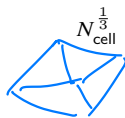
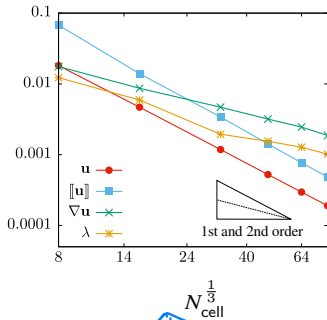


3D manufactured solution III

Hexahedral (cut)



Hexahedral (bary)



Thank you!

References I



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