

# A Stokes-Biot Stable $H(\text{div})$ -based mixed method for generalized poroelasticity

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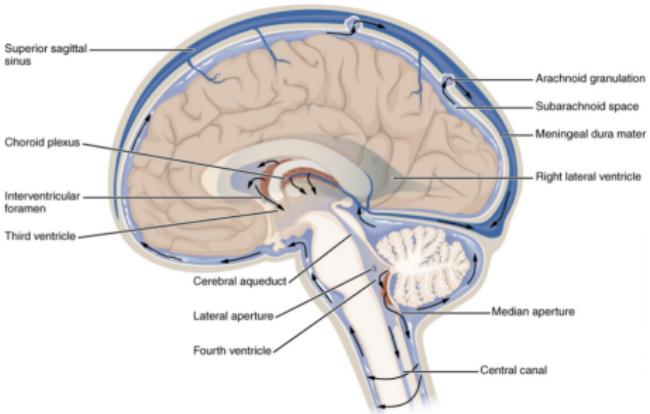
[ **simula**.research laboratory ]

August 9, 2018

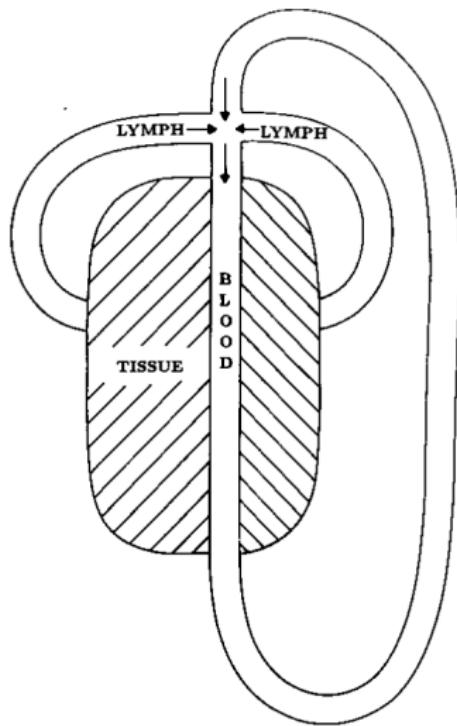
# Generalized Poroelasticity

1. Biomedical Motivation
2. 3-Field Formulation for MPET with Energy Estimates
3. Stokes-Biot Stability, and Semi-Discrete MPET Estimates
4. Current Work: A modified Stokes-Biot stability definition

# The Waterscape Project at Simula



# Black Box Models Are Common In Clinical Theory

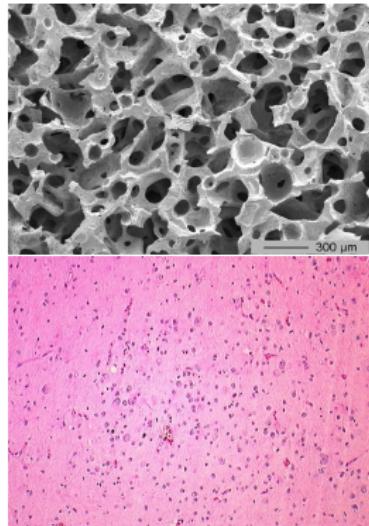


$$\frac{d}{dt}V = J_V - J_L$$

$$J_V = K_f (P_V - p_{hyd} - \sigma(\Pi_V - \Pi_I))$$
$$J_L = R_L^{-1} (p_{hyd} + P_p - P_L)$$

Figure: [Xie et. al. 1995]

# A Mechanical Response: Biot's Equations



**Figure:** Porous Isotropic Media

## Quantities

$\sigma$	effective elastic stress tensor
$\epsilon$	linearized strain tensor
$\mathcal{C}$	linear, isotropic stiffness tensor
$\vec{q}$	fluid flux

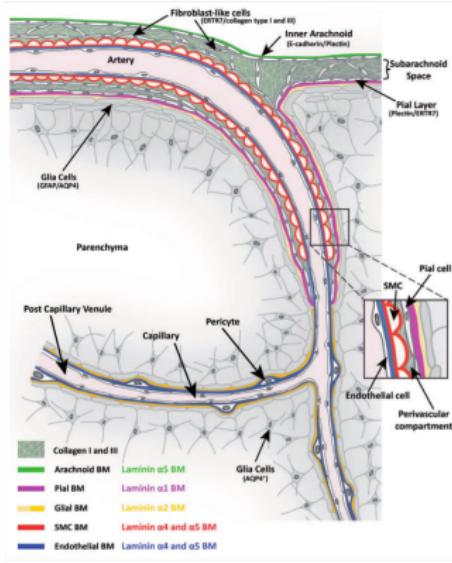
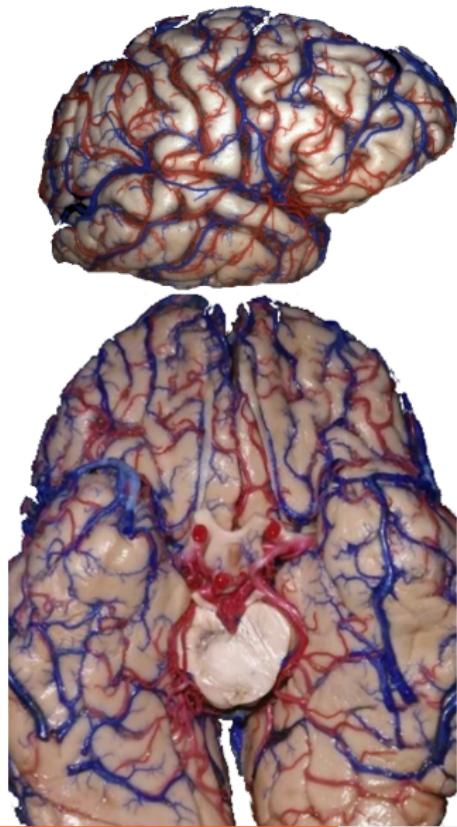
## Relations

$$\begin{aligned}\sigma &= \mathcal{C}\epsilon \\ \epsilon &= \frac{1}{2}(\nabla u + \nabla u^T) \\ \sigma(u) &= 2\mu\epsilon + \lambda \text{tr}(\epsilon)\mathbb{I} \\ \vec{q} &= -\kappa\nabla p\end{aligned}$$

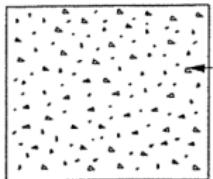
## Biot's Equations: Single Pore Fluid

$$\begin{aligned}-\operatorname{div} \sigma(u) + \alpha \nabla p &= 0, \\ c \dot{p} + \alpha \operatorname{div} \dot{u} - \operatorname{div}(\kappa \nabla p) &= g.\end{aligned}$$

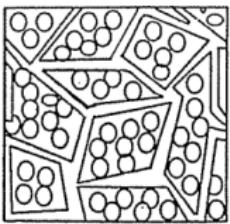
# The Brain Has Multiple Interacting Fluid Networks



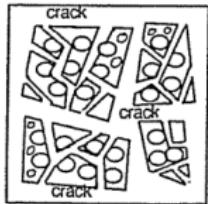
# Generalized Quasi-Static Linear Poroelasticity (MPET)



One Network



Two Networks



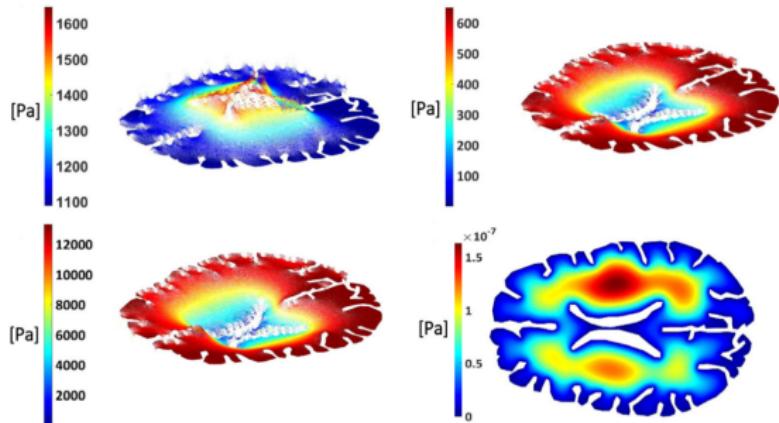
Three Networks

MPET Equations\*: 'A' networks

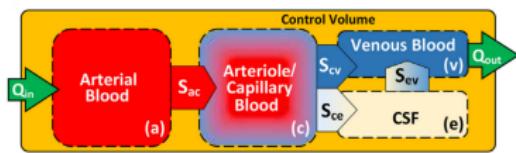
$$\begin{aligned}
 -\operatorname{div} \sigma(u) + \sum_{n=1}^A \alpha_n \nabla p_n &= 0, \\
 c_n \dot{p}_n + \alpha_n \operatorname{div} \dot{u} - \operatorname{div} (\kappa_n \nabla p_n) \\
 - \sum_{m=1}^A T_{m \rightarrow n} &= g_n. \\
 n &= 1, 2, \dots, A
 \end{aligned}$$

\* Barenblatt [1960] and Aifantiss [1980]

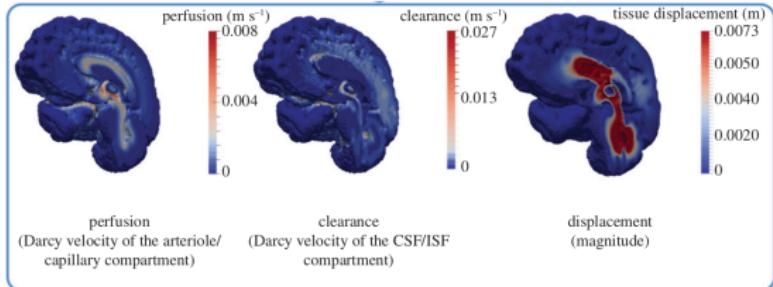
# Application Of MPET To The Brain: 4 Networks



[Vardakakis et. al. 2016]



[Vardakakis et. al. 2016]



[Vardakakis et. al. 2017]

# Numerical Robustness: Limiting Parameter Values

## Recent Numerical Work: Robust-In-Limit Methods

### ■ $\lambda$ : Youngs Modulus

- ① Brain Models:  $\nu \approx 1/2$
- ② Implies:  $\lambda \rightarrow \infty$
- ③ Lee, Piersanti, Mardal, and Rognes [2018, in prep]  
Total Pressure Mixed Formulation, MPET

### ■ $\kappa$ : Hydraulic Conductivity

- ①  $\approx 10^{-7} - \approx 10^{-17}$ , for general tissue matrix varieties
- ② Rodrigo, et. al. [2017]  
3-Field Mixed Formulation For Biot

### ■ $C$ : Storativity

- ①  $C \approx 0$ : Nearly Incompressible solid phase, and liquid phase
- ② Associated with spurious pressure oscillations
- ③ Lee [2017]  
3-Field Mixed Formulation For Biot

# Search: Robust Numerical Method For MPET

- Goal: Robust Method in the  $\kappa_n \rightarrow 0$  limit case

- Goal: Robust Method in the  $c_n \rightarrow 0$  limit case - e.g. 'locking free'

- Assume

$$\alpha_n \neq 0,$$

$$c_n > 0 \text{ or } \kappa_n > 0,$$

$$T_{m \rightarrow n} = \gamma_{mn}(p_n - p_m).$$

$$\gamma_{mm} = 0$$

$$\gamma_{nm} = \gamma_{mn}$$

MPET Equations\*: 'A' networks

$$-\operatorname{div} \sigma(u) + \sum_{n=1}^A \alpha_n \nabla p_n = 0,$$

$$c_n \dot{p}_n + \alpha_n \operatorname{div} \dot{u} - \operatorname{div}(\kappa_n \nabla p_n)$$

$$- \sum_{m=1}^A \gamma_{mn} (p_n - p_m) = g_n.$$

$$n = 1, 2, \dots, A$$

# 3-Field Formulation

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## 3-Field Mixed MPET Formulation

$$-\operatorname{div} \sigma(u) + \sum_{n=1}^A \alpha_n \nabla p_n = 0$$

$$z_n + \kappa_n \nabla p_n = 0$$

$$c_n \dot{p}_n + \alpha_n \operatorname{div} \dot{u} + \operatorname{div} z_n + \sum_{m=1}^A \gamma_{mn} (p_n - p_m) = g_n.$$

$$n = 1, 2, \dots, A$$

$$\begin{array}{lll} p_n = 0, & \sigma \cdot \mathbf{n} = 0, & x \in \overline{\Gamma_t}, \\ u = 0, & z_n \cdot \mathbf{n} = 0, & x \in \overline{\Gamma_c}. \end{array}$$

$$u \in H^1(0, T; \mathbf{U}) \quad z_n \in L^2(0, T; \mathbf{W}) \quad p_n \in H^1(0, T; \mathbf{Q}) \\ g_n \in \mathbf{Q}$$

Where  $\mathbf{U} = H_{\Gamma_c}^1(\Omega)^d$ ,  $\mathbf{W} = H_{\Gamma_c}(\operatorname{div}; \Omega)$ ,  $\mathbf{Q} = L^2(\Omega)$ .

## 3-Field Mixed MPET Variational Formulation

Find  $u \in H^1(0, T; \mathbf{U})$ ,  $z_n \in L^2(0, T; \mathbf{W})$ , and  $p_n \in L^2(0, T; \mathbf{Q})$   
 $\forall (v, w_n, p_n) \in \mathbf{U} \times \mathbf{W} \times \mathbf{Q}$

$$\langle \mathcal{C}\epsilon(u), \epsilon(v) \rangle - \sum_{n=1}^A \alpha_n \langle p_n, \operatorname{div} v \rangle = 0,$$

$$\kappa_n^{-1} \langle z_n, w_n \rangle - \langle p_n, \operatorname{div} w_n \rangle = 0,$$

$$c_n \langle \dot{p}_n, q_n \rangle + \alpha_n \langle \operatorname{div} \dot{u}, q_n \rangle + \langle \operatorname{div} z_n, q_n \rangle + \sum_{m=1}^A \gamma_{mn} \langle p_n - p_m, q_n \rangle = \langle g_n, q_n \rangle.$$

\* We also define a new norm for  $H(\operatorname{div})$  for later use; let  $\kappa > 0$ :

$$H_{\Gamma_c}(\operatorname{div}; \kappa^{-1}) = \{z \in H_{\Gamma_c}(\operatorname{div}) \mid \kappa^{-1} \langle z, z \rangle + \langle \operatorname{div} z, \operatorname{div} z \rangle < \infty\}$$

# Developing Tools

**Lemma 1** (Thompson & Rognes): Let  $B$  be a banach space with dual  $B'$ , and  $\langle \cdot, v \rangle_{B',B}$  denote the duality pairing. Suppose  $X \in C^0(0, T; B)$ ,  $G : [0, T] \rightarrow \mathbb{R}^+ \in L^1([0, T])$ ,  $\mathcal{C} \geq 0$  is a constant, and that  $F \in L^1(0, T; B')$ . Suppose  $\forall t \in [0, T]$

$$\|X(t)\|_B^2 \leq \|X(0)\|_B^2 + \int_0^t |\langle F(s), X(s) \rangle_{B',B}| + G(s) ds + \mathcal{C}$$

Then we have the following estimate

$$\begin{aligned} \|X(t)\|_B &\leq \|X(0)\|_B + \max\left\{2\|F\|_{L^1(0,T;B')}, \left(2 \int_0^T G(s) ds + 2\mathcal{C}\right)^{1/2}\right\} \\ &\leq \|X(0)\|_B + \max\{2\|F\|_{L^1(0,T;B')}, \sqrt{2}\|G\|_{L^1([0,T])}^{1/2} + \sqrt{2\mathcal{C}}\}. \end{aligned}$$

# Developing Tools

**Corollary 1** (Thompson & Rognes)\* Let  $X_1, X_2, \dots, X_A$  and  $F_1, F_2, \dots, F_A$  and  $J$  be continuous, non-negative functions, and  $C \geq 0$ . Suppose that

$$\sum_{n=1}^A X_n^2(t) \leq \sum_{n=1}^A X_n^2(0) + \sum_{n=1}^A \int_0^T F_n(s) X_n(s) ds + \int_0^T J(s) ds + C \quad (1)$$

Then the following inequality holds with proportionality constant no greater than  $2A$ :

$$\sum_{n=1}^A X_n(t) \lesssim \sum_{n=1}^A X_n(0) + \left( \sum_{n=1}^A \int_0^T F_n(s) ds + \left( \int_0^T J(s) ds + C \right)^{1/2} \right). \quad (2)$$

\* The case of  $A = 1$  first appears in Lee [2016].

# Energy Estimates

**Lemma 2** (Thompson & Rognes): Suppose the exact solutions satisfy  $u \in H^1(0, T; \mathbf{U})$ ,  $z_n \in L^2(0, T; \mathbf{W})$ , and  $p_n \in H^1(0, T; \mathbf{Q})$ , with sources  $g_n \in L^1(0, T; \mathbf{Q})$ . Suppose there exists  $c > 0$  such that  $\kappa_n, c_n, \alpha_n > c$   $\forall n = 1, 2, \dots, A$  then

$$\begin{aligned} & \int_0^T \left( \sum_{m,n=1}^A \gamma_{mn} \|p_n - p_m\|_{L^2}^2 + \sum_{n=1}^A \kappa_n^{-1} \|z_n\|_{L^2}^2 \right)^{\frac{1}{2}} \mathrm{d}s \\ & + \|u(t)\|_1 + \sum_{n=1}^A \sqrt{c_n} \|p_n(t)\|_{L^2} \lesssim \|u(0)\|_1 + \sum_{n=1}^A \|p_n(0)\|_{L^2} \\ & + \sum_{n=1}^A \frac{1}{\sqrt{c_n}} \|g_n\|_{L^1(0,T;L^2)} \end{aligned}$$

We Control:

$$\sup_{t \in [0, T]} \|u(t)\|_1, \quad \sup_{t \in [0, T]} \|p_n(t)\|_{L^2}, \quad \|p_n - p_m\|_{L^2(0, T; L^2)}, \quad \kappa_n^{-1} \|z_n\|_{L^2(0, T; L^2)}.$$

# Energy Estimates

## Proof Outline.

Select  $v = \dot{u}$ ,  $w_j = z_j$ , and  $q_j = p_j$ . Use  $\gamma_{mn}$  symmetric:

$$\begin{aligned} \langle \mathcal{C}\epsilon(u), \epsilon(\dot{u}) \rangle + \sum_{n=1}^A \kappa_n^{-1} \|z_n\|_{L^2}^2 + \sum_{n=1}^A \frac{d}{dt} c_n \|p_n\|_{L^2}^2 \\ + \frac{1}{2} \sum_{n,m=1}^A \gamma_{mn} \|p_n - p_m\|_{L^2}^2 = \sum_{n=1}^A \langle g_n, p_n \rangle \end{aligned}$$

Integrating from 0 to T and using standard inequalities gives

$$\begin{aligned} \|u(t)\|_{H^1}^2 + \sum_{n=1}^A c_n \|p(t)\|_{L^2}^2 + \int_0^T \sum_{n,m=1}^A \gamma_{mn} \|p_n - p_m\|_{L^2}^2 ds + \\ \int_0^T \sum_{n=1}^A \kappa_n^{-1} \|z_n\|_{L^2}^2 ds \lesssim \|u(0)\|_{H^1}^2 + \sum_{n=1}^A c_n \|p_n(0)\|_{L^2}^2 + \\ \sum_{n=1}^A \int_0^T (1/\sqrt{c_n}) \|g_n(s)\| \sqrt{c_n} \|p_n(s)\| ds \end{aligned}$$

Apply the generalized Gronwall-type inequality to the above. □

# Energy Estimates

**Remark** If in addition  $z_n \in H^1(0, T; W)$  and  $g_n \in L^1(0, T; Q) \cap L^2(0, T; Q)$

$$\begin{aligned}
& \|u(t)\|_1 + \sum_{n=1}^A \sqrt{c_n} \|p_n(t)\|_{L^2} + \sum_{n=1}^A \sqrt{\kappa_n^{-1}} \|z_n(t)\|_{L^2} + \sum_{m,n=1}^A \sqrt{\gamma_{mn}} \|p_n(t) - p_m(t)\|_{L^2} \\
& + \left( \int_0^T \|\dot{u}\|_1^2 + \sum_{n=1}^A c_n \|\dot{p}_n\|_{L^2}^2 + \sum_{n=1}^A \|z_n\|_{H(\text{div}; \kappa_n^{-1})}^2 + \sum_{m,n=1}^A \gamma_{mn} \|p_n - p_m\|_{L^2}^2 \, ds \right)^{1/2} \\
& \lesssim \|u(0)\|_1 + \sum_{n=1}^A \sqrt{c_n} \|p_n(0)\|_{L^2} + \sum_{n=1}^A \|z_n(0)\|_{H(\text{div}; \kappa_n^{-1})} \\
& + \sum_{m,n=1}^A \sqrt{\gamma_{mn}} \|p_n(0) - p_m(0)\|_{L^2} + \sum_{n=1}^A \frac{1}{\sqrt{c_n}} \|g_n\|_{L^1(0, T; L^2)} \\
& + \left( \sum_{n=1}^A \frac{1}{c_n} \|g_n\|_{L^2(0, T; L^2)}^2 \right)^{1/2}
\end{aligned}$$

We Control:

$$\|\dot{u}\|_{L^2(0, T; H^1)} \quad \|\dot{p}_n\|_{L^2(0, T; L^2)} \quad \|z_n\|_{L^2(0, T; H(\text{div}; \kappa_n^{-1}))} \quad \|p_n - p_m\|_{L^2(0, T; L^2)}$$

$$\sup_{t \in [0, T]} \|u\|_1, \quad \sup_{t \in [0, T]} \|p_n\|_{L^2}, \quad \sup_{t \in [0, T]} \|p_n - p_m\|_{L^2}, \quad \sup_{t \in [0, T]} \kappa_n^{-1} \|z_n\|_{L^2}$$

# Semi-discrete formulation

Let  $\mathbf{U}_h \subset \mathbf{U}$ ,  $\mathbf{W}_h \subset \mathbf{W}$ , and  $\mathbf{Q}_h \subset \mathbf{Q}$  be finite dimensional.

suppose the existence and uniqueness of semi-discrete solutions  $\mathbf{u}_h(t, x) \in C^2(0, T; \mathcal{X}_h)$ ,  $\mathbf{z}_{h,n}(t, x) \in C^1(0, T; W_h)$ , and  $p_{h,n}(t, x) \in C^1(0, T; Q_h)$ , for each  $n \in \{1, 2, \dots, N\}$ , such that for every  $(\mathbf{v}_h, \mathbf{w}_h, q_h) \in \mathcal{X}_h \times W_h \times Q_h$ :

$$\langle \mathcal{C}\epsilon(u_h), \epsilon(v_h) \rangle + \sum_{n=1}^A \alpha_n \mathbf{b}(v_h, p_{h,n}) = 0$$

$$\kappa_n^{-1} (z_{h,n}, w_h) + \mathbf{b}(w_h, p_{h,n}) = 0$$

$$c_n (\dot{p}_{h,n}, q_h) - \alpha_n \mathbf{b}(\dot{u}_h, q_h) - \mathbf{b}(z_{h,n}, q_h) + \sum_{m=1}^N \gamma_{mn} (p_{h,n} - p_{h,m}, q_h) = (g_n, q_h)$$

where  $\mathbf{b}(w, q) = -\langle \nabla \cdot w, q \rangle$

# Stokes-Biot Stability

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# Stokes-Biot Stability

**Definition** The triple  $U_h \times W_h \times Q_h \subset U \times W \times Q$  is called *Stokes-Biot stable* if

- ① The bilinear form  $\langle \mathcal{C}\epsilon(u), \epsilon(v) \rangle$  is bounded on  $U$  and coercive on  $U_h$ ,
- ②  $(W_h, Q_h)$  is a Poisson-Stable pair of discrete spaces,
- ③  $(U_h, Q_h)$  is a Stokes-Stable pair of discrete spaces.

Define the Poisson Galerkin projection for  $(z_n, p_n)$  by

$$\begin{aligned} \kappa_n^{-1}(\Pi_{W_h} \mathbf{z}^n, \mathbf{w}_h) + \mathbf{b}(\mathbf{w}_h, \Pi_{Q_h} p^n) &= \kappa^{-1}(\mathbf{z}^n, \mathbf{w}_h) + \mathbf{b}(\mathbf{w}_h, p^n), \quad \forall \mathbf{w}_h \in W_h \\ \mathbf{b}(\Pi_{W_h} \mathbf{z}^n, q_h) &= \mathbf{b}(\mathbf{z}^n, q_h), \quad \forall q_h \in Q_h \end{aligned}$$

Define the augmented elastic projection for  $u$  by

$$\langle \mathcal{C}\epsilon(\Pi_{U_h} u), \epsilon(v_h) \rangle = \langle \mathcal{C}\epsilon(u), \epsilon(v_h) \rangle - \sum_{n=1}^N \alpha_n \mathbf{b}(v_h, I_{Q_h}^{p_n}), \quad \forall v_h \in v_h$$

As a first step: assume the material parameters are **uniformly bounded below** and derive interpolation error estimates. c.f. [Rodrigo et. al 2016]

# Interpolation Error Estimate

Claim: Define  $I_{U_h}^u = \Pi_{U_h} u - u$ , and likewise for  $p_n$ , and  $z_n$ . Then

$$\|I_h^{z_n}\|_W + \|I_h^{p_n}\|_Q \lesssim \inf_{w_h \in W_h} \|z_n - w_h\|_W + \inf_{q_h \in Q_h} \|p_n - q_h\|_Q \quad (3)$$

$$\begin{aligned} \|I_h^u\|_U &\lesssim \inf_{v_h \in U_h} \|u - v_h\|_U + \sum_{n=1}^A \inf_{w_h \in W_h} \|z_n - w_h\|_W \\ &\quad + \sum_{n=1}^A \inf_{q_h \in Q_h} \|p_n - q_h\|_Q \end{aligned} \quad (4)$$

## Proof Outline.

The estimate for the Galerkin projector of the Poisson problem, e.g. (3), is well known; for  $\kappa_n$  bounded below, the right-hand side is a fixed constant. We now establish the estimate of (4) □

# Interpolation Error Estimate

## Proof Outline (cont.)

Define  $\hat{u}_h$  by:  $\forall v_h \in U_h \langle \mathcal{C}\epsilon(\hat{u}_h), \epsilon(v_h) \rangle = \langle \mathcal{C}\epsilon(u), \epsilon(v_h) \rangle$ . It is well known that  $\|u - \hat{u}_h\|_1 \lesssim \inf_{v_h \in U_h} \|u - v_h\|_{U_h}$ . Coercivity of  $\langle \mathcal{C}\epsilon(\cdot), \epsilon(\cdot) \rangle$  on  $U_h$  gives

$$\|\hat{u}_h - \Pi_{U_h} u\|_U^2 \lesssim \langle \mathcal{C}\epsilon(\hat{u}_h - \Pi_{U_h} u), \epsilon(\hat{u}_h - \Pi_{U_h} u) \rangle$$

Using the definition of the augmented elastic projector:

$$\begin{aligned} \langle \mathcal{C}\epsilon(\hat{u}_h - \Pi_{U_h} u), \epsilon(\hat{u}_h - \Pi_{U_h} u) \rangle &= \langle \mathcal{C}\epsilon(\hat{u}_h - \textcolor{blue}{u}), \epsilon(\hat{u}_h - \Pi_{U_h} u) \rangle + \\ &\quad \sum_{n=1}^A \alpha_n \mathbf{b}(\hat{u}_h - \Pi_{U_h} u, I_{Q_h}^{p_n}) \end{aligned}$$

Cauchy Schwarz and boundedness gives

$$\|\hat{u}_h - \Pi_{U_h} u\|_U \lesssim \sum_{n=1}^A \|I_h^{p_n}\|_Q$$

The result follows from  $\|u - \Pi_{U_h} u\| \leq \|u - \hat{u}_h\| + \|\hat{u}_h - \Pi_{U_h} u\|$  and the interpolation estimates established for  $I_h^{p_n}$ . □

# Interpolation Error Estimate

Recall:  $I_{\mathbf{U}_h}^u = \Pi_{\mathbf{U}_h} u - u$ , and  $E_{\mathbf{U}_h}^u = \Pi_{\mathbf{U}_h} u - u_h$  and likewise for  $p_n$ , and  $z_n$

**Proposition** (Thompson & Rognes)

$$\begin{aligned}
 & \|E_{\mathbf{U}_h}^u(t)\|_1 + \sum_{n=1}^A \sqrt{c_n} \|E_{\mathbf{Q}_h}^{p_n}(t)\|_{L^2} + \sum_{n=1}^A \sqrt{\kappa_n^{-1}} \|E_{\mathbf{W}_h}^{z_n}\|_{L^2(0,T;L^2)} \\
 & + \sum_{m,n=1}^A \sqrt{\gamma_{m,n}} \|E_{\mathbf{Q}_h}^{p_n,p_m}\|_{L^2(0,T;L^2)} \lesssim \|E_{\mathbf{U}_h}^u(0)\|_1 + \sum_{n=1}^A \sqrt{c_n} \|E_{\mathbf{Q}_h}^{p_n}(0)\|_{L^2} \\
 & + \inf_{v_h \in \mathbf{U}_h} \|\dot{u} - v_h\|_1 + \sum_{n=1}^A \inf_{w_h \in \mathbf{W}_h} (\|\dot{z}_n - w_h\|_W + \|z_n - w_h\|_W) \\
 & + \sum_{n=1}^A \inf_{q_h \in \mathbf{Q}_h} (\|\dot{p}_n - q_h\|_{L^2} + \|p_n - q_h\|_{L^2})
 \end{aligned}$$

# Current Work: Improved Estimates

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## $\kappa$ -dependence In The Galerkin Projector

- The previous estimates work for material coefficients bounded below; namely  $\kappa_n \geq \bar{\kappa} > 0 \forall n = 1, 2, \dots, A$
- The projection strategy discussed is based on the Galerkin Projection for the Poisson problem:

$$\begin{aligned} a(\mathbf{z}^n, \mathbf{w}) + b(\mathbf{w}, p^n) &= \mathcal{F}(\mathbf{w}), \quad \forall \mathbf{w}_h \in W_h \\ b(\mathbf{z}^n, q) &= \mathcal{G}(q), \quad \forall q \in Q \end{aligned}$$

Where  $a(\mathbf{z}, \mathbf{q}) = \kappa^{-1} \langle \mathbf{z}, \mathbf{q} \rangle$ ,  $b(\mathbf{w}, q) = -\langle \operatorname{div} \mathbf{w}, q \rangle$

- The Galerkin Projection estimates are:

$$\|\mathbf{z}^n - \mathbf{z}_h^n\|_W \leq c_1 \inf_{\mathbf{w}_h \in W_h} \|\mathbf{z}^n - \mathbf{w}_h\|_W + c_2 \inf_{q_h \in Q_h} \|p^n - q_h\|_Q$$

$$\|p^n - p_h^n\|_W \leq c_3 \inf_{\mathbf{w}_h \in W_h} \|\mathbf{z}^n - \mathbf{w}_h\|_W + c_4 \inf_{q_h \in Q_h} \|p^n - q_h\|_Q$$

$$c_1 = 2(1 + \beta_h^{-1}) \quad c_2 = \kappa \quad c_3 = 2\kappa^{-1}\beta_h^{-1}(1 + \beta_h^{-1}) \quad (1 + 2\beta_h^{-1})$$

\* $\beta_h$  is a constant depending only on  $c(\Omega) > 0$  via Poincare

# Numerical Tests: Stokes-Biot Stability

$$\begin{aligned}\mathbf{u}(x, y, t) &= \operatorname{curl} \varphi = \begin{pmatrix} \partial_y \varphi \\ -\partial_x \varphi \end{pmatrix}, \quad \varphi(x, y) = [xy(1-x)(1-y)]^2, \\ p(x, y, t) &= 1.\end{aligned}$$

$$P_1 \times RT_1 \times P_0$$

$\kappa \backslash N$	8	16	32	64	128	
$10^{-4}$	0.0187	0.0040	0.0009	0.0002	$5.66 \times 10^{-5}$	$\ \Pi_1 \mathbf{u} - \mathbf{u}_h\ _A$
	0.0590	0.0090	0.0016	0.0003	$8.33 \times 10^{-5}$	$\ \Pi_0 p - p_h\ _{L^2}$
$10^{-6}$	0.0547	0.0302	0.0050	0.0005	$8.93 \times 10^{-5}$	$\ \Pi_1 \mathbf{u} - \mathbf{u}_h\ _A$
	0.3187	0.3098	0.0741	0.0097	0.0012	$\ \Pi_0 p - p_h\ _{L^2}$
$10^{-8}$	0.0578	0.0567	0.0476	0.0165	0.0018	$\ \Pi_1 \mathbf{u} - \mathbf{u}_h\ _A$
	0.3388	0.7067	1.1418	0.6450	0.1142	$\ \Pi_0 p - p_h\ _{L^2}$
$10^{-10}$	0.0578	0.0574	0.0570	0.0550	0.0502	$\ \Pi_1 \mathbf{u} - \mathbf{u}_h\ _A$
	0.3372	0.7176	1.4527	2.7790	3.4403	$\ \Pi_0 p - p_h\ _{L^2}$

\* MMS test problem from [Rodrigo et. al 2016]

# Numerical Tests: $\kappa = 1 \times 10^{-8}$

$P_2 \times RT_0 \times P_0$

$N$	4	8	16	32	64
$\ u - u_h\ _{L^2}$	$5.251e - 04$	$5.734e - 05$	$6.001e - 06$	$6.879e - 07$	$8.366e - 08$
$\ u - u_h\ _1$	$1.002e - 02$	$2.650e - 03$	$6.629e - 04$	$1.651e - 04$	$4.120e - 05$
$\ p - p_h\ _{L^2}$	$4.462e - 03$	$5.834e - 04$	$7.058e - 05$	$8.634e - 06$	$1.074e - 06$
$\ u - u_h\ _{L^2} \approx O(h^3)$		$\ u - u_h\ _1 \approx O(h^2)$		$\ p - p_h\ _{L^2} \approx O(h^3)^*$	

$P_4 \times RT_3 \times DG_3^{**}$

$N$	4	8	16	32	
$\ u - u_h\ _{L^2}$	$4.853e - 06$	$1.580e - 07$	$4.950e - 09$	$1.543e - 10$	
$\ u - u_h\ _1$	$3.117e - 04$	$2.057e - 05$	$1.301e - 06$	$8.155e - 08$	
$\ p - p_h\ _{L^2}$	$1.180e - 03$	$7.747e - 05$	$4.865e - 06$	$3.037e - 07$	
$\ u - u_h\ _{L^2} \approx O(h^5)$		$\ u - u_h\ _1 \approx O(h^4)$		$\ p - p_h\ _{L^4} \approx O(h^3)$	

\*\*  $P_4 \times DG_3$  is the Scott-Vogelius (Stokes-Stable) element

# An Alternative Projection: 1 Network (Biot)

- Norms yielding estimates free of  $\kappa$ -dependent constants (Weakly Robust)
- Utilize Stokes Stability

Consider the following model problem:

Find  $(\mathbf{u}, \mathbf{z}, p) \in [H^1]^d \times H(\text{div}; \kappa^{-1}) \times L^2$  such that for every  $(\mathbf{v}, \mathbf{w}, q)$

$$\begin{bmatrix} A_S & 0 & b^T \\ 0 & \kappa^{-1} A_P & b^T \\ b & b & 0 \end{bmatrix} \begin{bmatrix} \mathbf{u} \\ \mathbf{z} \\ p \end{bmatrix} = \begin{bmatrix} \mathcal{W}(\mathbf{v}) \\ \mathcal{X}(\mathbf{w}) \\ \mathcal{Y}(q) \end{bmatrix}$$

$$\langle A_S \mathbf{u}, \mathbf{v} \rangle = \langle \mathcal{C}\epsilon(\mathbf{u}), \epsilon(\mathbf{v}) \rangle \quad \langle A_P \mathbf{z}, \mathbf{w} \rangle = \langle \mathbf{z}, \mathbf{w} \rangle \quad \langle b \mathbf{u}, \mathbf{v} \rangle = -\langle \text{div } u, \mathbf{v} \rangle$$

This problem can be re-cast into a saddle point problem as:

$$\begin{bmatrix} \mathcal{A} & \mathcal{B}^T \\ \mathcal{B} & 0 \end{bmatrix} \begin{bmatrix} \sigma \\ p \end{bmatrix} \begin{bmatrix} \mathcal{F} \\ \mathcal{G} \end{bmatrix}$$

where  $\sigma = (\mathbf{u}, \mathbf{z}) \in [H_{\Gamma_c}^1]^d \times H(\text{div}; \kappa^{-1})$ ,  $p \in L^2$

# An Alternative Projection: Well Posedness

The Following results hold

- $\mathcal{A}$  is bounded
- $\mathcal{A}$  is coercive on  
 $\text{Ker}(\mathcal{B}) = \{\sigma = (\mathbf{u}, \mathbf{z}) \in [H_{\Gamma_c}^1]^d \times H(\text{div}; \kappa^{-1}) \mid \text{div } \mathbf{u} + \text{div } \mathbf{z} = 0\}$  provided  
 $\text{div}(W_h) \subset Q_h$
- $\mathcal{B}$  is bounded
- The inf-sup condition holds for  $\mathcal{B}$  and  $\beta$  coincides with the **Stokes** stability constant

The associated Galerkin projection based on this problem provides:

- Exact cancellation of 'problematic' terms in the semi-discrete error equation for the MPET equation
- Error estimates for  $\|\mathbf{u} - \mathbf{u}_h\|_{H_1}$ ,  $\|\mathbf{z}^n - \mathbf{z}_h^n\|_{H(\text{div}; \kappa^{-1})}$ , and  $\|p^n - p_h^n\|_{L^2}$  with  $\kappa$ -independent constants

# An Alternative Projection: Well Posedness

## Redefining Stokes-Biot stability

**Definition** The triple  $U_h \times W_h \times Q_h \subset U \times W \times Q$  is called *Stokes-Biot stable* if

- 1 The bilinear form  $\langle C\epsilon(u), \epsilon(v) \rangle$  is bounded on  $U$  and coercive on  $U_h$ ,
- 2  $(U_h, Q_h)$  is a Stokes-Stable pair of discrete spaces.
- 3  $\text{div}(W_h) \subset Q_h$

$$\begin{aligned} & \|I_h^u\|_{H^1} + \sum_{n=1}^A \|I_h^{zn}\|_{H(\text{div}; \kappa-1)} + \|I_h^{pn}\|_{L^2} \lesssim c_1 \inf_{v_h \in U_h} \|u - v_h\|_{H^1} \\ & + c_2 \sum_{n=1}^A \inf_{w_h \in W_h} \|z_{h,n} - q_h\|_{H(\text{div}; \kappa-1)} + \inf_{q_h \in Q_h} \|p_n - q_h\|_{L^2} \end{aligned} \quad (5)$$

- $c_1$  and  $c_2$  free of all material parameters
- Only stability constant is Stokes
- Same error equation as previous approach, same discrete error estimates

### Ongoing Work:

- Generalizing to multiple fluid networks (Complete)
- Fully discrete a-priori error estimates
- Redefining Stokes-Biot stability, and locking study (Gaspar, Rodrigo, Mardal, Rognes, Thompson)

# Further Reading

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Thank You For Your Attention.



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