

# AI-aided, incremental numerical approach for finite strain poroelasticity: On the brain tissue deformation

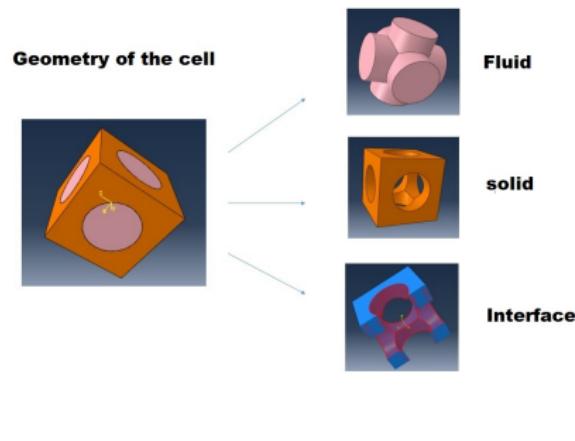
H. Dehghani and A. Zilian

Institute of Computational Engineering and Sciences, Department of Engineering, Faculty of Science, Technology and Medicine, University of Luxembourg

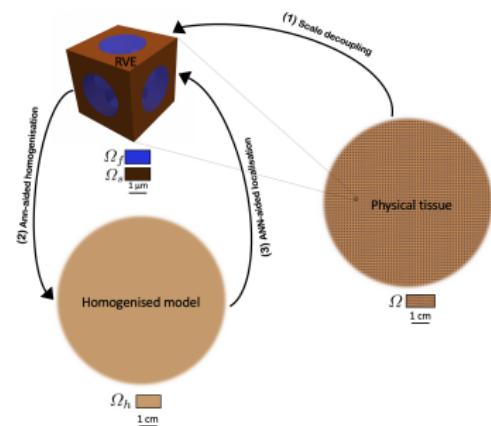
May 21, 2021



# Poroelasticity



**Figure:** Phase Decoupling.



**Figure:** Phases and Scales.

# Poroelastic response

why incremental analysis?

$$\mathbf{u}_{t+\Delta t} = \mathbf{u}_t + \Delta \mathbf{u}$$

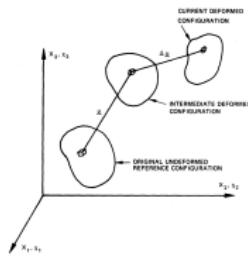
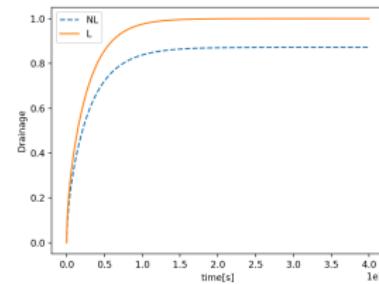
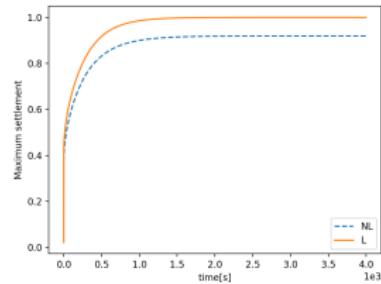
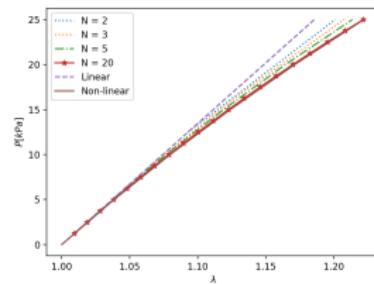


Figure 2.1 – Original and Deformed Configuration



# Poroelasticity

$$\nabla \cdot \sigma = 0$$

$$\sigma = -p\mathbf{I} + \mu \left( \nabla \mathbf{v} + (\nabla \mathbf{v})^\top \right)$$

$$\nabla \cdot \mathbf{v} = 0$$

$$\nabla \cdot \tau = 0$$

$$\tau = \mathbb{C}\mathbf{e}$$

$$\tau \mathbf{n} = \sigma \mathbf{n}$$

$$\dot{\mathbf{u}} = \mathbf{v}$$

in  $\Omega_f$

in  $\Omega_f$

in  $\Omega_f$

in  $\Omega_s$

in  $\Omega_s$

on  $\Gamma$

on  $\Gamma$

$$\nabla_{\mathbf{x}} \cdot \tau_E = 0,$$

$$\tau_E := \tilde{\mathbb{C}} \nabla_{\mathbf{x}} \mathbf{u}^{(0)} - \tilde{\alpha} p^{(0)}$$

$$\dot{p}^{(0)} = -M[\tilde{\alpha} : \nabla_{\mathbf{x}} \dot{\mathbf{u}}^{(0)} + \nabla_{\mathbf{x}} \cdot \langle \mathbf{w} \rangle_f]$$

$$\langle \mathbf{w} \rangle_f = -K \nabla_{\mathbf{x}} p^{(0)}$$

$$\langle \mathbf{w} \rangle_f = \left\langle \mathbf{v}^{(0)} \right\rangle_f - \phi \dot{\mathbf{u}}^{(0)}$$

$$K := \langle W \rangle_f, \quad \tilde{\alpha} := \phi I - \text{Tr} \langle M \rangle_s$$

$$\tilde{\mathbb{C}} := \langle \mathbb{C} + \mathbb{C}M \rangle_s, \quad M := -\frac{1}{\langle \text{Tr} Q \rangle_s}$$

$$\langle \psi \rangle_k = \frac{1}{|\Omega|} \int_{\Omega_k} \psi(\mathbf{x}, \mathbf{y}) d\mathbf{y} \quad k = f, s,$$

Implicit interdependency  
between effective coefficients

# Cell/RVE Problems

$$\nabla_y^2 W^T - \nabla_y P + I = 0 \quad \text{in } \Omega_f$$

$$\nabla_y \cdot W^T = 0 \quad \text{in } \Omega_f$$

$$W = 0 \quad \text{on } \Gamma$$

$$\langle P \rangle_f = 0$$

$$\nabla_y \cdot (\mathbb{C}Q) = 0 \quad \text{in } \Omega_s$$

$$(\mathbb{C}Q)\mathbf{n} + \mathbf{n} = 0 \quad \text{on } \Gamma$$

$$\nabla_y \cdot (\mathbb{C}\mathbb{M}) = 0 \quad \text{in } \Omega_s$$

$$(\mathbb{C}\mathbb{M})\mathbf{n} + \mathbb{C}\mathbf{n} = 0 \quad \text{on } \Gamma$$

$$\langle \mathcal{A} \rangle_s = 0$$

Efficient?

# Localisation and Microscopic properties update

$$\langle \varepsilon^{(1)} \rangle = \langle \mathbb{M} \rangle \varepsilon^{(0)} + \langle \mathbf{Q} \rangle p^{(0)}$$

Back to the incremental formulation

$$\langle \varepsilon_{t+\Delta t}^{(1)} \rangle = \langle \varepsilon_t^{(1)} \rangle + \langle \Delta \varepsilon^{(1)} \rangle$$

$$\mathbf{F} = \langle \varepsilon_{t+\Delta t}^{(1)} \rangle / (1 - \phi) + \mathbf{I}$$

$$\mathbf{E} = \frac{1}{2} (\mathbf{F}^T \cdot \mathbf{F} - \mathbf{I})$$

$$\phi = 1 - (1 - \phi_i) J$$

$$W(I_1, J) = C_{10}(I_1 - 3) - 2C_{10} \ln(J)$$

$$+ \frac{1}{D_1} \ln(J)^2$$

$$C_{10} = \frac{E_i}{4(1 + \nu_i)}$$

$$D_1 = \frac{6(1 - 2\nu_i)}{E_i}$$

$$C_{ijkl}^t = \frac{\partial W}{\partial E_{ij} \partial E_{kl}}$$

Sufficiently small increment

where,

$$J := \det(\mathbf{F})$$

$$C_{ijkl}^t \approx C_{ijkl}$$

$$E = \frac{C_{11}(C_{11} + C_{12}) - 2C_{12}^2}{C_{11} + C_{12}}$$

$$\nu = \frac{C_{12}}{C_{11} + C_{12}}$$

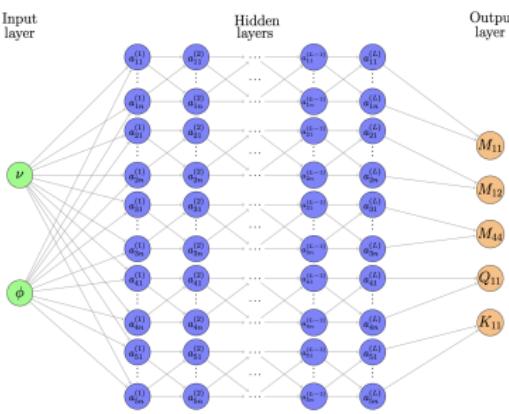
# Application of ANN

$$(\mathbb{M}, \mathbf{Q}, \mathbf{K}) = ANN(\mathbb{C}, \mu, \phi)$$

$$(M_{11}, M_{12}, M_{44}, Q_{11}, K_{11}) = ANN(E, \nu, \mu, \phi)$$

$$(M_{11}, M_{12}, M_{44}, Q_{11}, K_{11}) = ANN(\nu, \phi)$$

No extrapolation risk!



$$z_k^{(i)} = w_{kj}^{(i)} a_j^{(i-1)} + b_k^{(i)}$$
$$a_k^{(i)} = \text{ReLU}(z_k^{(i)})$$

$$n = 50 \ L = 3$$

$$n = 50 \ L = 3$$

$$n = 20 \ L = 3$$

$$n = 50 \ L = 3$$

$$n = 10 \ L = 3$$

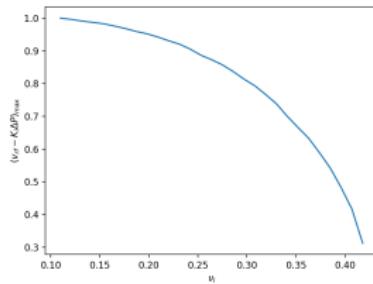
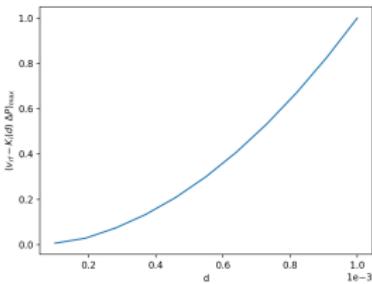
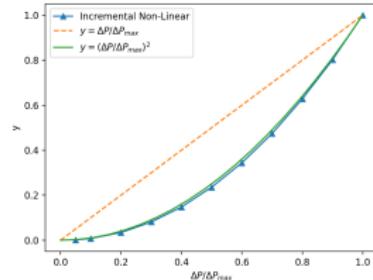
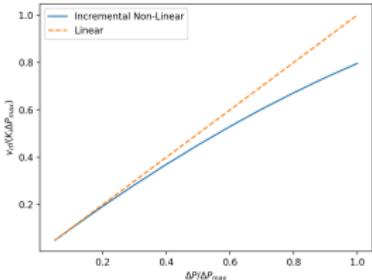
# Incremental nonlinear algorithm

$$\begin{aligned}\boldsymbol{u}_{t+\Delta t}^{(0)} &= \boldsymbol{u}_t^{(0)} + \Delta \boldsymbol{u}^{(0)} \\ p_{t+\Delta t}^{(0)} &= p_t^{(0)} + \Delta p^{(0)}\end{aligned}$$

$$\begin{aligned}0 = & \int_{\partial\mathcal{B}} (\boldsymbol{t}_s^t + \Delta \boldsymbol{t}_s) \cdot \delta \boldsymbol{u}^{(0)} \, dS - \int_{\mathcal{B}} \left( \boldsymbol{\tau}_E^t + (\tilde{\mathbb{C}}_t : \Delta \boldsymbol{\varepsilon}^{(0)} - \tilde{\boldsymbol{\alpha}}_t \Delta p^{(0)}) \right) : \nabla_{\boldsymbol{x}} \delta \boldsymbol{u}^{(0)} \, dV \\ & + \int_{\mathcal{B}} \left( \frac{1}{M_{t-\Delta t}} \dot{p}_t^{(0)} + \frac{1}{M_t} \Delta \dot{p}^{(0)} \right) \delta p^{(0)} \, dV + \int_{\partial\mathcal{B}} (\boldsymbol{v}_{rf}^t - \boldsymbol{\kappa}_t \Delta \nabla_{\boldsymbol{x}} p^{(0)}) \cdot \boldsymbol{n} \delta p^{(0)} \, dS \\ & - \int_{\mathcal{B}} (\boldsymbol{v}_{rf}^t - \boldsymbol{\kappa}_t \Delta \nabla_{\boldsymbol{x}} p^{(0)}) \cdot \nabla_{\boldsymbol{x}} \delta p^{(0)} \, dV + \int_{\mathcal{B}} (\tilde{\boldsymbol{\alpha}}_{t-\Delta t} : \dot{\boldsymbol{\varepsilon}}_t + \tilde{\boldsymbol{\alpha}}_t : \Delta \dot{\boldsymbol{\varepsilon}}) \delta p^{(0)} \, dV \\ & \forall \delta \boldsymbol{u}^{(0)}, \delta p^{(0)}.\end{aligned}$$

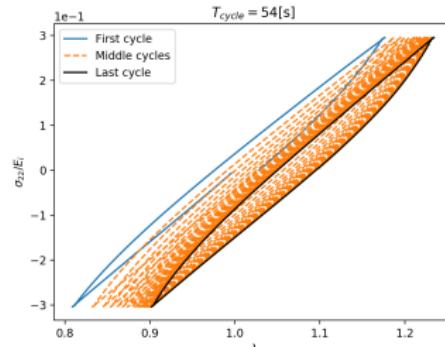
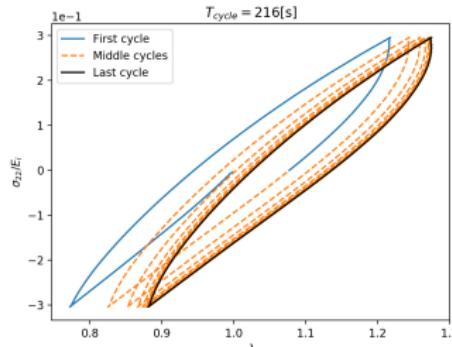
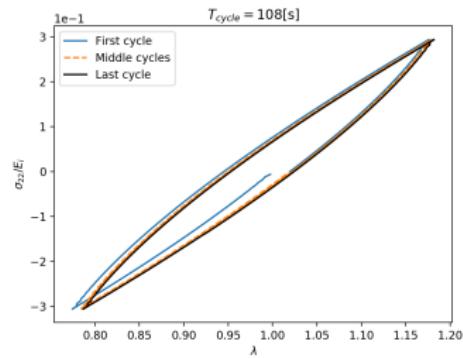
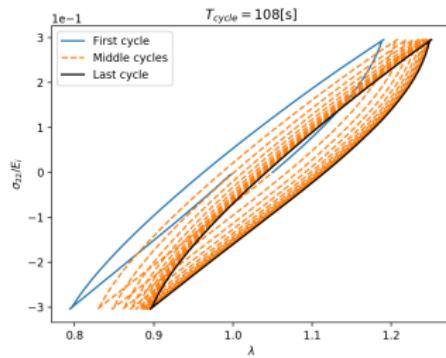
# Deviation from Darcy's law

$$y = \frac{v_{rf} - (K_i \Delta P)}{v_{rf,max} - (K_i \Delta P_{max})} \quad \Delta P_{max} = E_i / 4$$

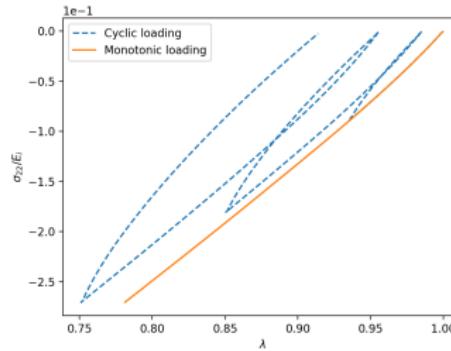
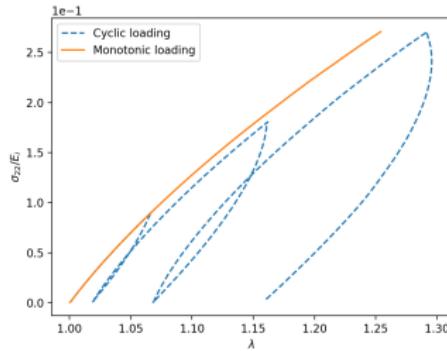
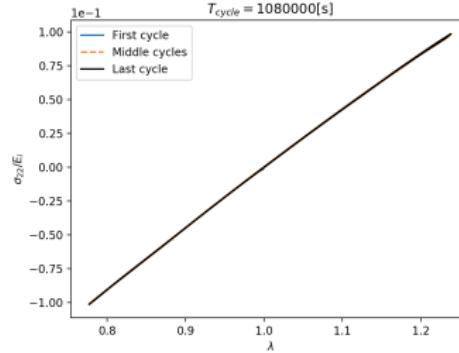
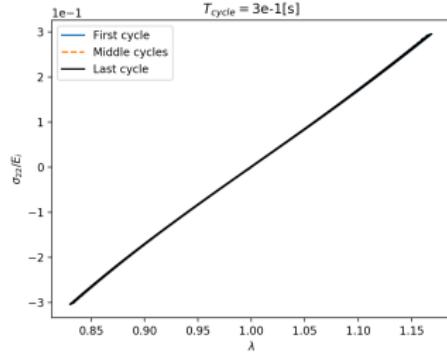


# Uniaxial cyclic test simulating brain tissue

$$L_c = 1[\text{mm}] \quad \mu_c = 10^{-3}[\text{Pa.s}] \quad d = 20^{-6}[\text{m}] \quad f_c = 10^{-3}[\text{N}]$$
$$\nu_i = 0.3[-] \quad E_i = 13.5 * 10^3[\text{Pa}] \quad T_{cycle} = 108[\text{s}] \quad width = 9.5L_c \quad length = 14L_c$$



# Uniaxial cyclic test simulating brain tissue



Thank you for your attention