Two-scale PDE-based numerical modelling of gradient-driven transport in heterogeneous materials

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Fine-scale model

▶ Gradient-driven transport:

$$\frac{\partial u}{\partial t} + \nabla \cdot (-K\nabla u) = 0 \quad \text{in } \Omega$$

• Ω comprised of two sub-domains Ω_a (connected) and Ω_b (inclusions):

$$K = \begin{cases} K_a & \text{in } \Omega_a \\ K_b & \text{in } \Omega_b \end{cases}$$

$$\frac{\partial u_a}{\partial t} + \nabla \cdot (-K_a \nabla u_a) = 0 \quad \text{in } \Omega_a$$
$$\frac{\partial u_b}{\partial t} + \nabla \cdot (-K_b \nabla u_b) = 0 \quad \text{in } \Omega_b$$



Heterogeneous domain $\Omega_a \blacksquare \Omega_b \blacksquare$

Computational cost of direct numerical simulation is prohibitively expensive when the medium exhibits small-scale heterogeneity.

Macroscopic averaging



Assumptions:

- 1. At each point $x \in \Omega$, there exists a micro-cell C_x .
- 2. We assume that $C_{x,b}$ is entirely located in the interior of the micro-cell C_x

Classical Macroscopic Model

Rendard and de Marsily (1997); Szymkiewicz and Lewandowska (2006); Davit et al. (2013)

► Macroscopic Model:

$$\frac{\partial U}{\partial t} + \nabla_x \cdot \left(-\boldsymbol{K}_{\text{eff}} \nabla_x U \right) = 0; \quad U = \frac{1}{|\mathcal{C}_x|} \int_{\mathcal{C}_x} u \, dV$$

► Effective conductivity:

$$\left(\boldsymbol{K}_{\text{eff}}\right)_{:,j} = \frac{1}{\left|\mathcal{C}_{x}\right|} \int_{\mathcal{C}_{x}} K e_{j} \, dV + \frac{1}{\left|\mathcal{C}_{x}\right|} \int_{\mathcal{C}_{x}} \nabla_{y} \chi_{j} \, dV$$



$$\nabla_y \cdot (K \nabla_y (\chi_j + y_j)) = 0 \quad y \in \mathcal{C}_x$$

subject to:

$$\frac{1}{|\mathcal{C}_x|} \int_{\mathcal{C}_x} \chi_j \, dV = 0$$



Macroscopic averaging

Whitaker (1998); Davit et al. (2013)

Average over the micro-cell:

$$\frac{1}{|\mathcal{C}_x|} \int_{\mathcal{C}_{x,a}} \frac{\partial u_a}{\partial t} \, dV + \frac{1}{|\mathcal{C}_x|} \int_{\mathcal{C}_{x,a}} \left(\nabla \cdot \boldsymbol{q}_a \right) \, dV = 0$$

Temporal averaging theorem: ►

$$\frac{1}{|\mathcal{C}_x|} \int_{\mathcal{C}_{x,a}} \frac{\partial u_a}{\partial t} \, dV = \underbrace{\frac{|\mathcal{C}_{x,a}|}{|\mathcal{C}_x|}}_{\varepsilon_a} \frac{\partial}{\partial t} \underbrace{\frac{1}{|\mathcal{C}_{x,a}|}}_{U_a} \int_{\mathcal{C}_{x,a}} u_a \, dV$$



$$\frac{1}{|\mathcal{C}_x|} \int_{\mathcal{C}_{x,a}} \left(\nabla \cdot \boldsymbol{q}_a \right) \, dV = \nabla_x \cdot \frac{1}{|\mathcal{C}_x|} \int_{\mathcal{C}_{x,a}} \boldsymbol{q}_a \, dV - \frac{1}{|\mathcal{C}_x|} \int_{\Gamma_x} \boldsymbol{q}_a \cdot \boldsymbol{n} \, ds$$

Macroscopic (averaged) equations: ►

$$\varepsilon_a \frac{\partial U_a}{\partial t} + \nabla_x \cdot \boldsymbol{Q}_a = S \qquad \varepsilon_b \frac{\partial U_b}{\partial t} = -S$$





Two-scale Model (Model 1)

Showalter (1997); Szymkiewicz and Lewandowska (2008); Carr and Turner (2014)



Macroscopic equation: Microscopic equation:

Microscopic BC:

Source term:

$$\begin{split} \varepsilon_a \frac{\partial U_a}{\partial t} + \nabla_x \cdot (-\mathbf{K}_{\text{eff}} \nabla_x U_a) &= S \,, \quad x \in \Omega \\ \frac{\partial u_b}{\partial t} + \nabla \cdot (-K_b \nabla u_b) &= 0 \,, \quad y \in \mathcal{C}_{x,b} \\ u_b &= U_a \,, \quad y \in \Gamma_x \\ S &= -\frac{1}{|\mathcal{C}_x|} \int_{\Gamma_x} q_b \cdot \boldsymbol{n} \, ds \end{split}$$

Two-scale Model (Model 2)



Spatial Discretisation (Model 1)

Carr and Turner (2014); Carr et al. (2015)



Spatial Discretisation (Model 2)



Time discretisation (Model 1 and Model 2)

Carr et al. (2011, 2015); Carr and Turner (2014); Hochbruck et al. (1998)

Spatial discretisation can be expressed in the form:

$$\frac{d\boldsymbol{u}}{dt} = \boldsymbol{g}(\boldsymbol{u}), \quad \boldsymbol{u}(0) = \boldsymbol{u}_0$$

where number of unknowns is very large.

Exponential Euler method:

$$\boldsymbol{u}_{n+1} = \boldsymbol{u}_n + \tau_n \boldsymbol{J}_n^{-1} (e^{\tau_n \boldsymbol{J}_n} - \boldsymbol{I}) \boldsymbol{g}_n$$

Explicit scheme



Jacobian structure J_n (Zoomed in)

► Krylov subspace methods for computing J⁻¹_n(e^{τ_nJ_n − I)g_n converge rapidly without preconditioning, and require only matrix-vector products with J_n:}

$$\boldsymbol{J}_{n}\boldsymbol{v}\approx\frac{\boldsymbol{g}(\boldsymbol{u}_{n}+\varepsilon\|\boldsymbol{v}\|_{2}\boldsymbol{v})-\boldsymbol{g}(\boldsymbol{u}_{n})}{\varepsilon\|\boldsymbol{v}\|_{2}}\,,\quad\varepsilon\approx10^{-8}$$



$$\frac{\partial u_a}{\partial t} + \nabla \cdot (-K_a \nabla u_a) = 0 \quad \text{in } \Omega_a$$
$$\frac{\partial u_b}{\partial t} + \nabla \cdot (-K_b \nabla u_b) = 0 \quad \text{in } \Omega_b$$
$$K_b/K_a = 10^{-p}$$

Carr et al. (2015)

 $K_b/K_a = 10^{-2}$



Carr et al. (2015)

 $K_b/K_a = 10^{-4}$



Carr et al. (2015)

 $K_b/K_a = 10^{-4}$





$$\frac{\partial \theta_a}{\partial t} + \nabla \cdot (-K_a \nabla \psi_a - K_a \mathbf{e}_2) = 0 \quad \text{in } \Omega_a$$
$$\frac{\partial \theta_b}{\partial t} + \nabla \cdot (-K_b \nabla \psi_b - K_b \mathbf{e}_2) = 0 \quad \text{in } \Omega_b$$
$$K_b / K_a = 10^{-3}$$









Summary and Conclusions

- Presented a new two-scale model for gradient-driven transport/flow problems in heterogeneous materials (Model 2)
- ▶ The novel approach avoids the need for an effective parameter in the macroscopic equation by computing the macroscopic flux as the average of the microscopic fluxes over the micro-cell.
- ▶ Numerical experiments demonstrated that both two-scale models (Model 1 and Model 2) produce numerical solutions that are in excellent agreement with the fine-scale model at a reduced computational cost.
- Model 1 requires less computational time
- Model 2 is more accurate and able to capture additional fine-scale features in the solution.

Thank you!

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