

Coupling Stokes and Darcy equations: modeling and numerical methods

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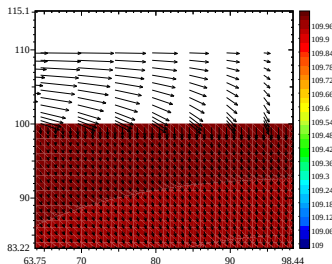
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Motivation

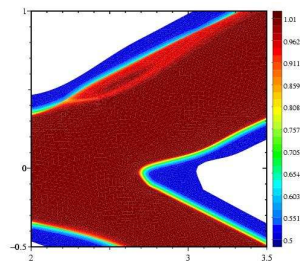
- We want to model the *filtration of a free fluid through a porous medium*:

Environmental application



[simulation by E. Miglio]

Biomedical simulations



[simulation by P. Zunino]

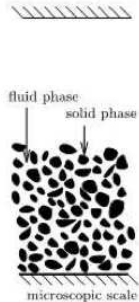
Industrial applications: filters, porous foams, fuel cells...

- We must introduce a suitable modeling framework and set up effective numerical methods.

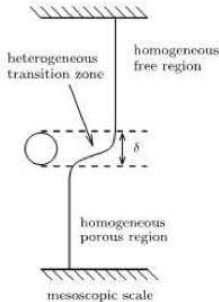
Modeling filtration (i)

To model the filtration process three different approaches are generally considered:

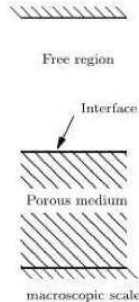
I - Microscale



II - Mesoscale



III - Macroscale

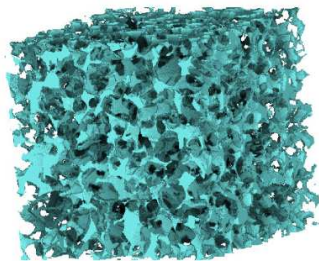


[Chandesris and Jamet, *Transp. Porous Med.*, 2009]

Modeling filtration (ii)

I - **Microscale** approach

The geometry of the porous medium must be taken into account and the Navier-Stokes equations must be solved in the whole domain.



[R. Hilfer, Lecture Notes in Physics 554, p. 203-241 (2000)]

Modeling filtration (iii)

II - Mesoscale approach

A transition region is considered between the porous medium and the free fluid [Whitaker and Ochoa-Tapia (1995); Jackson et al (2012)].

Penalization approach [Iliev et al. (2004,2007); Angot (1999)]:

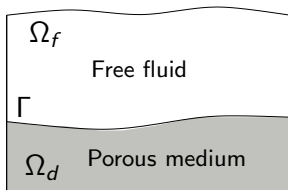
$$\begin{aligned} -\operatorname{div} \sigma(\mathbf{u}, p) + (\mathbf{u} \cdot \nabla) \mathbf{u} + \frac{n\nu}{K} \mathbf{u} + \frac{C_F}{\sqrt{K}} |\mathbf{u}| \mathbf{u} &= \mathbf{f} & \text{in } \Omega_f \cup \Omega_d \\ \operatorname{div} \mathbf{u} &= 0 \end{aligned}$$

The penalization terms are set to zero in Ω_f .

Modeling filtration (iv)

III - Macroscale approach

Sharp interface Γ with continuity conditions across it.



- Free flow: **Navier-Stokes** equations

$$\begin{aligned} -\operatorname{div} \sigma(\mathbf{u}_f, p_f) + (\mathbf{u}_f \cdot \nabla) \mathbf{u}_f &= \mathbf{f}_f \\ \operatorname{div} \mathbf{u}_f &= 0 \end{aligned} \quad \text{in } \Omega_f$$

where $\sigma(\mathbf{u}_f, p_f) = 2\mu \nabla^s \mathbf{u}_f - p_f \mathbf{I}$ is the Cauchy stress tensor.

- Porous media flow: **Darcy** equation

$$\begin{aligned} \mathbf{u}_d + \frac{\mathbf{K}}{\mu} (\nabla p_d - \mathbf{f}_d) &= \mathbf{0} \\ \operatorname{div} \mathbf{u}_d &= 0 \end{aligned} \quad \Leftrightarrow \quad -\operatorname{div} \left(\frac{\mathbf{K}}{\mu} (\nabla p_d - \mathbf{f}_d) \right) = 0 \quad \text{in } \Omega_d$$

Modeling filtration (v)

Continuity conditions (I)

- 1 Continuity of normal velocities (due to incompressibility)

$$\mathbf{u}_f \cdot \mathbf{n} = \mathbf{u}_d \cdot \mathbf{n} \quad \text{on } \Gamma$$

- 2 Continuity of normal stresses

$$-\mathbf{n} \cdot \boldsymbol{\sigma}(\mathbf{u}_f, p_f) \cdot \mathbf{n} = p_d \quad \text{on } \Gamma$$

(pressures may be discontinuous across Γ : $p_f - p_d \neq 0$ on Γ);

- 3 Beavers–Joseph–Saffman condition

$$-\boldsymbol{\tau} \cdot \boldsymbol{\sigma}(\mathbf{u}_f, p_f) \cdot \mathbf{n} = \frac{\alpha_d \mu}{\sqrt{\boldsymbol{\tau} \cdot \mathbf{K} \cdot \boldsymbol{\tau}}} \mathbf{u}_f \cdot \boldsymbol{\tau}$$

[Beavers and Joseph (1967); Saffman (1971);
Jäger and Mikelić (1996,2000,2001) Miglio et al (2002); Layton et al (2003)]

Modeling filtration (vi)

Continuity conditions (II)

- ① Continuity of normal velocities (due to incompressibility)

$$\mathbf{u}_f \cdot \mathbf{n} = \mathbf{u}_d \cdot \mathbf{n} \quad \text{on } \Gamma$$

- ② Continuity of pressures

$$p_f = p_d \quad \text{on } \Gamma$$

[Ene and Sanchez-Palencia (1975); Levy and Sanchez-Palencia (1975)]

Modeling filtration (vii)

- The **microscale** approach is very demanding: careful reconstruction of the geometry and solution of the Navier-Stokes equations.
- The **macroscale** approach allows using ad-hoc models in each subdomain.
Once the coupling conditions are set, a global well-posed problem may be set up and **non-overlapping domain decomposition methods** can be used to solve the corresponding interface problem (*substructuring techniques*).
- The **mesoscale** approach requires solving a more expensive problem in $\Omega_f \cup \Omega_d$ than the **macroscale** approach.

Coupling through a sharp interface:
domain decomposition methods without overlap
(macroscale setting)

Weak form of the Stokes-Darcy problem

Find $\mathbf{u}_f \in H^1(\Omega_f)$, $p_f \in L^2(\Omega_f)$, $p_d \in H^1(\Omega_d)$:

$$\int_{\Omega_f} 2\mu \nabla^s \mathbf{u}_f : \nabla^s \mathbf{v} + \int_{\Gamma} \frac{\alpha_d \mu}{\sqrt{\boldsymbol{\tau} \cdot \mathbf{K} \cdot \boldsymbol{\tau}}} (\mathbf{u}_f)_{\boldsymbol{\tau}} \mathbf{v}_{\boldsymbol{\tau}} - \int_{\Omega_f} p_f \operatorname{div} \mathbf{v} + \int_{\Gamma} p_d (\mathbf{v} \cdot \mathbf{n}) = \int_{\Omega_f} \mathbf{f}_f \cdot \mathbf{v}$$

$$\int_{\Omega_f} q_f \operatorname{div} \mathbf{u}_f = 0$$

$$\int_{\Omega_d} \nabla q_d \cdot \frac{\mathbf{K}}{\mu} \nabla p_d - \int_{\Gamma} (\mathbf{u}_f \cdot \mathbf{n}) q_d = \int_{\Omega_d} \frac{\mathbf{K}}{\mu} \mathbf{f}_d \cdot \nabla q_d$$

Algebraic form

A suitable conforming finite element approximation of the coupled problem leads to the linear system:

$$\left(\begin{array}{ccc|c} (A_f + \bar{\alpha} M_\Gamma)_{II} & (A_f)_{I\Gamma} & D_I^T & 0 \\ (A_f)_{\Gamma I} & (A_f)_{\Gamma\Gamma} & D_\Gamma^T & C_\Gamma^T \\ D_I & D_\Gamma & 0 & 0 \\ \hline 0 & -C_\Gamma & 0 & A_d \end{array} \right) \begin{pmatrix} (\mathbf{u}_f)_I \\ (\mathbf{u}_f)_n \\ p_f \\ p_d \end{pmatrix} = \mathbf{b}$$

with $(\mathbf{u}_f)_n$ vector of the nodal values of $\mathbf{u}_f \cdot \mathbf{n}$ on Γ

An equivalent interface equation

The Stokes-Darcy problem can be formulated in terms of the solution λ (normal velocity across Γ) of the interface problem

$$S_f \lambda + S_d \lambda = \chi \quad \text{on } \Gamma$$

- S_f “fluid” operator:

$$S_f : \lambda \text{ (normal velocities on } \Gamma) \xrightarrow[\text{Stokes}]{\text{solve}} -\mathbf{n} \cdot \boldsymbol{\sigma}(\mathbf{u}_f, p_f) \cdot \mathbf{n} \text{ (normal stresses on } \Gamma).$$

- S_d “porous medium” operator:

$$S_d : \lambda \text{ (normal velocities on } \Gamma) \xrightarrow[\text{Darcy}]{\text{solve}} p_d|_{\Gamma} \text{ (pressure } p_d \text{ on } \Gamma) .$$

Discrete interface equation for the normal velocity:

$$(\boldsymbol{\Sigma}_f + \boldsymbol{\Sigma}_d)(\mathbf{u}_f)_n = \chi_s + \chi_d$$

Dimensional analysis of the Stokes-Darcy problem

The Stokes-Darcy problem can be characterized in terms of three dimensionless numbers:

- the Reynolds number

$$\text{Re} = \frac{U_f X_f \rho}{\mu}$$

- the Euler number

$$\text{Eu} = \frac{P_f}{\rho U_f^2}$$

- the Darcy number [Bear (1991)]

$$\text{Da} = \frac{K}{X_f^2}$$

[MD (in preparation)]

Dimensionless Schur complement system

The Schur complement system with respect to $(\mathbf{u}_f)_n$ becomes:

$$(\boldsymbol{\Sigma}_s + \boldsymbol{\Sigma}_d)(\mathbf{u}_f)_n = \chi_s + \chi_d$$

with

$$\boldsymbol{\Sigma}_f = (\text{Re Eu})^{-1} \hat{\boldsymbol{\Sigma}}_f$$

$$\boldsymbol{\Sigma}_d = (\text{Re Eu Da})^{-1} \hat{\boldsymbol{\Sigma}}_d$$

and

$$\begin{aligned} \hat{\boldsymbol{\Sigma}}_f = & (\mathbf{A}_f)_{\Gamma\Gamma} - (\mathbf{A}_f)_{\Gamma I} (\mathbf{A}_f + \alpha \text{Da}^{-1/2} \mathbf{M}_\Gamma)_{II}^{-1} (\mathbf{A}_f)_{I\Gamma} \\ & + \left((\mathbf{A}_f)_{\Gamma I} (\mathbf{A}_f + \alpha \text{Da}^{-1/2} \mathbf{M}_\Gamma)_{II}^{-1} \mathbf{D}_I^T - \mathbf{D}_\Gamma^T \right) \\ & \cdot \left(\mathbf{D}_I (\mathbf{A}_f + \alpha \text{Da}^{-1/2} \mathbf{M}_\Gamma)_{II}^{-1} \mathbf{D}_I^T \right)^{-1} \\ & \cdot \left(\mathbf{D}_I (\mathbf{A}_f + \alpha \text{Da}^{-1/2} \mathbf{M}_\Gamma)_{II}^{-1} (\mathbf{A}_f)_{I\Gamma} - \mathbf{D}_\Gamma \right) \end{aligned}$$

$$\hat{\boldsymbol{\Sigma}}_d = \mathbf{C}_\Gamma^T \mathbf{A}_d^{-1} \mathbf{C}_\Gamma$$

If we multiply by (Re Eu Da), we obtain:

$$(\text{Da } \hat{\Sigma}_f + \hat{\Sigma}_d)(\mathbf{u}_f)_n = \hat{\chi}_f + \hat{\chi}_d$$

Remark that we have independence of Re.

We can prove that

$$\text{cond}((\text{Da } \hat{\Sigma}_f + \hat{\Sigma}_d)) \leq c_1 \frac{k_+}{k_-} \cdot \frac{\text{Da } k_- + h}{\text{Da } k_+ + \frac{k_-}{k_+} h} \cdot \frac{1}{h}$$

	Da :	1	10^{-3}	10^{-6}	10^{-12}
Re = 1	$h = 1/4$	8.1586	55.4938	303.7483	305.7232
	$h = 1/8$	12.9894	43.9005	617.5967	634.4864
	$h = 1/16$	23.4144	32.1862	1.1558e+03	1.2809e+03
	$h = 1/32$	44.8932	23.3303	1.8151e+03	2.5736e+03
Re = 10^2	$h = 1/4$	8.1586	55.4938	303.7483	305.7232
	$h = 1/8$	12.9894	43.9005	617.5967	634.4864
	$h = 1/16$	23.4144	32.1862	1.1558e+03	1.2809e+03
	$h = 1/32$	44.8932	23.3303	1.8151e+03	2.5736e+03

Dirichlet-Neumann type preconditioners (i)

It can be proved that

$$\text{cond}((\text{Da } \hat{\Sigma}_f)^{-1}(\hat{\Sigma}_f + \hat{\Sigma}_d)) \leq C \frac{1 + \frac{1}{k_- \text{Da}}}{1 + \frac{k_-}{k_+} \frac{h^2}{\text{Da}}}$$

from which we can see that

- if $\text{Da} \gg h^2$, then $\text{cond}((\text{Da } \hat{\Sigma}_f)^{-1}(\hat{\Sigma}_f + \hat{\Sigma}_d)) \sim 1$
- if $\text{Da} \ll h^2$, then $\text{cond}((\text{Da } \hat{\Sigma}_f)^{-1}(\hat{\Sigma}_f + \hat{\Sigma}_d)) \sim h^{-2}$

PCG iterations, $\text{tol} = 10^{-10}$, $\text{Re} = 1$							
Da :	1	10^{-2}	10^{-3}	10^{-5}	10^{-9}	10^{-12}	10^{-14}
$h = 1/4$	8	8	9	9	9	9	9
$h = 1/8$	16	15	15	21	21	21	20
$h = 1/16$	27	23	16	41	46	47	46
$h = 1/32$	40	33	20	45	80	80	81
<i>no prec.</i>							
Da :	1	10^{-2}	10^{-3}	10^{-5}	10^{-9}	10^{-12}	10^{-14}
$h = 1/4$	5	9	9	10	10	10	10
$h = 1/8$	5	11	17	24	24	24	24
$h = 1/16$	5	11	19	56	64	65	65
$h = 1/32$	5	11	19	74	131	135	135

$$P = (\text{Da } \hat{\Sigma}_f)^{-1}$$

Dirichlet-Neumann type preconditioners (ii)

It can be proved that

$$\text{cond}(\hat{\Sigma}_d^{-1}(\hat{\Sigma}_f + \hat{\Sigma}_d)) \leq C \frac{1 + \frac{k_+}{k_-} \frac{k_+ \text{Da}}{h^2}}{1 + k_- \text{Da}}$$

from which we can see that

- if $\text{Da} \gg h^2$, then $\text{cond}(\hat{\Sigma}_d^{-1}(\hat{\Sigma}_f + \hat{\Sigma}_d)) \sim h^{-2}$
- if $\text{Da} \ll h^2$, then $\text{cond}(\hat{\Sigma}_d^{-1}(\hat{\Sigma}_f + \hat{\Sigma}_d)) \sim 1$

PCG iterations, $\text{tol} = 10^{-10}$, $\text{Re} = 1$							
Da :	1	10^{-2}	10^{-3}	10^{-5}	10^{-9}	10^{-12}	10^{-14}
$h = 1/4$	8	8	9	9	9	9	9
$h = 1/8$	16	15	15	21	21	21	20
$h = 1/16$	27	23	16	41	46	47	46
$h = 1/32$	40	33	20	45	80	80	81
<i>no prec.</i>							
Da :	1	10^{-2}	10^{-3}	10^{-5}	10^{-9}	10^{-12}	10^{-14}
$h = 1/4$	10	8	8	5	2	1	1
$h = 1/8$	24	20	18	6	2	1	1
$h = 1/16$	60	52	44	10	2	1	1
$h = 1/32$	149	135	114	17	2	1	1

$P = (\hat{\Sigma}_d)^{-1}$

Comments

- Using $(\hat{\Sigma}_f)^{-1}$ requires solving a **Stokes problem**:

$$\mathbf{y} = \hat{\Sigma}_f^{-1} \mathbf{w} \Leftrightarrow y = \mathbf{u}_f \cdot \mathbf{n} \text{ on } \Gamma \quad \text{where}$$

$$\begin{aligned} \text{Stokes}(\mathbf{u}_f, p_f) &= 0 && \text{in } \Omega_f \\ -\mathbf{n} \cdot \boldsymbol{\sigma}(\mathbf{u}_f, p_f) \cdot \mathbf{n} &= \mathbf{w} && \text{on } \Gamma \end{aligned}$$

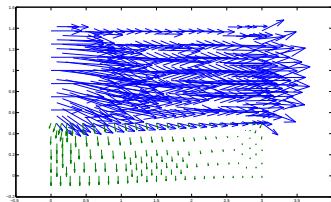
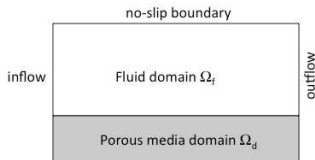
- Using $(\hat{\Sigma}_d)^{-1}$ requires solving a **Darcy problem**:

$$\mathbf{y} = \hat{\Sigma}_d^{-1} \mathbf{w} \Leftrightarrow y = -\frac{\mathbf{K}}{\mu} \frac{\partial p_d}{\partial n} \text{ on } \Gamma \quad \text{where}$$

$$\begin{aligned} -\operatorname{div} \left(\frac{\mathbf{K}}{\mu} \nabla p_d \right) &= 0 && \text{in } \Omega_d \\ p_d &= w && \text{on } \Gamma \end{aligned}$$

Numerical results (i)

We consider the following setting proposed in [Hanspal et al, 2009]:



- $\Omega_f = 0.015 \times 0.005$ m, $\Omega_d = 0.015 \times 0.0025$ m
- Inflow with max horizontal velocity 0.1 m/s, $Re = 500$

K:	10^{-10} m^2	10^{-12} m^2	10^{-14} m^2	10^{-10} m^2	10^{-12} m^2	10^{-14} m^2
Da:	$1.3 \cdot 10^{-3}$	$1.3 \cdot 10^{-7}$	$1.3 \cdot 10^{-9}$	$1.3 \cdot 10^{-3}$	$1.3 \cdot 10^{-7}$	$1.3 \cdot 10^{-9}$
$h = 1/4$	31	31	32	36	37	36
$h = 1/8$	68	76	76	98	106	107
$h = 1/16$	93	154	159	168	>250	>250
no prec.			$P = (\text{Da} \hat{\Sigma}_f)^{-1}$			
	10^{-10} m^2	10^{-12} m^2	10^{-14} m^2			
	$1.3 \cdot 10^{-3}$	$1.3 \cdot 10^{-7}$	$1.3 \cdot 10^{-9}$			
	6	3	2			
	10	4	3			
	17	5	3			
$P = (\hat{\Sigma}_d)^{-1}$						

Numerical results (ii)

Following [Zunino (2002)] we consider an artery with

- radius = 0.3 cm
- wall thickness = 0.03 cm
- blood density $\rho_b = 1.04 \text{ g/cm}^3$
- blood (kinematic) viscosity $\nu_b = 0.033 \text{ cm}^2/\text{s}$
- wall permeability $\mathbf{K} = 2 \cdot 10^{-14} \text{ cm}^2$

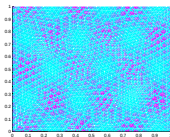
This corresponds to

- $\text{Re} = 272.73$
- $\text{Da} = 2.78 \cdot 10^{-14}$

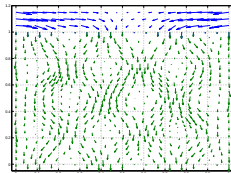
Dofs ($\mathbf{u}_f + p_f + p_d$)					h	No prec.	PCG with $\hat{\Sigma}_d$
259	+	73	+	138	1/4	32	1
973	+	259	+	495	1/8	76	1
3769	+	973	+	1869	1/16	147	2
14833	+	3769	+	7257	1/32	226	2

Numerical results (iii)

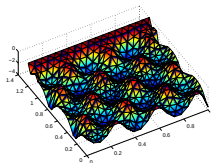
- $\Omega_f = (0, 10) \times (10, 12)$, $\Omega_d = (0, 10) \times (0, 10)$
- $\text{Re} = 1000$
- Variable permeability $10^{-8} \leq \mathbf{K} \leq 10^{-6} \text{ m}^2$
($\text{Da} = 3.3 \cdot 10^{-10}$)



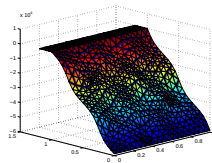
Velocity field



Vertical velocity



Hydrodynamic pressure

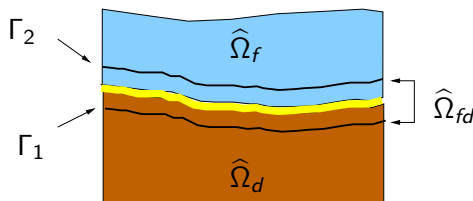


Coupling through a transition region: interface control domain decomposition methods (mesoscale setting)

Joint work with: **Paola Gervasio** (Università di Brescia)
Alfio Quarteroni (EPF Lausanne)

The ICDD method for the Stokes-Darcy problem (i)

Similarly to the mesoscale approach, we introduce a transition region overlapping Ω_f and Ω_d :



Stokes

$$\begin{aligned} -\mathbf{div} \boldsymbol{\sigma}(\mathbf{u}_f, p_f) &= \mathbf{f}_f & \text{in } \hat{\Omega}_f \\ \mathbf{div} \mathbf{u}_f &= 0 & \text{in } \hat{\Omega}_f \\ \Phi_f(\mathbf{u}_f, p_f) &= \lambda_1 & \text{on } \Gamma_1 \end{aligned}$$

Darcy

$$\begin{aligned} -\mathbf{div} \left(\frac{\mathbf{K}}{\mu} (\nabla p_d - \mathbf{f}_d) \right) &= 0 & \text{in } \hat{\Omega}_d \\ \Phi_d(p_d) &= \lambda_2 & \text{on } \Gamma_2 \end{aligned}$$

λ_1, λ_2 control interface functions

The ICDD method for the Stokes-Darcy problem (ii)

Consider controls corresponding to the traces of the Stokes velocity and the Darcy pressure on the interfaces:

$$\mathbf{u}_f = \boldsymbol{\lambda}_1 \text{ on } \Gamma_1 \quad \text{and} \quad p_d = \lambda_2 \text{ on } \Gamma_2.$$

We set the minimization problem:

$$\inf_{\boldsymbol{\lambda}_1, \lambda_2} J(\boldsymbol{\lambda}_1, \lambda_2)$$

with

$$J(\boldsymbol{\lambda}_1, \lambda_2) = \frac{1}{2} \|\mathbf{u}_f - \mathbf{u}_d\|_{L^2(\Gamma_1)}^2 + \frac{1}{2} \|p_f - p_d\|_{L^2(\Gamma_2)}^2$$

[MD, Gervasio, Quarteroni (submitted 2014)]

The ICDD method for the Stokes-Darcy problem (iii)

The corresponding optimality system becomes:

- state problems:

$$\begin{aligned} -\mathbf{div} \boldsymbol{\sigma}(\mathbf{u}_f, p_f) &= \mathbf{f}_f & \text{in } \hat{\Omega}_f \\ \mathbf{div} \mathbf{u}_f &= 0 & \text{in } \hat{\Omega}_f \\ \mathbf{u}_f &= \boldsymbol{\lambda}_1 & \text{on } \Gamma_1 \end{aligned} \qquad \begin{aligned} -\mathbf{div} \left(\frac{\mathbf{K}}{\mu} (\nabla p_d - \mathbf{f}_d) \right) &= 0 & \text{in } \hat{\Omega}_d \\ p_d &= \lambda_2 & \text{on } \Gamma_2 \end{aligned}$$

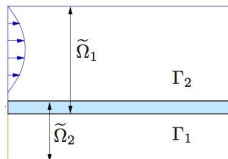
- pseudo-adjoint problems:

$$\begin{aligned} -\mathbf{div} \boldsymbol{\sigma}(\mathbf{w}_f, q_f) &= \mathbf{0} & \text{in } \hat{\Omega}_f \\ \mathbf{div} \mathbf{w}_f &= 0 & \text{in } \hat{\Omega}_f \\ \mathbf{w}_f &= \mathbf{u}_f - \mathbf{u}_d & \text{on } \Gamma_1 \end{aligned} \qquad \begin{aligned} -\mathbf{div} \left(\frac{\mathbf{K}}{\mu} \nabla \psi_d \right) &= 0 & \text{in } \hat{\Omega}_d \\ \psi_d &= p_d - p_f & \text{on } \Gamma_2 \end{aligned}$$

- pseudo-Euler-Lagrange equations:

$$(\mathbf{u}_f - \mathbf{u}_d) + \psi_d = \mathbf{0} \text{ on } \Gamma_1 \qquad (p_d - p_f) + q_f = 0 \text{ on } \Gamma_2.$$

Numerical results (i)



BiCGStab iterations:

- Dependence on K

K	10^{-7}	10^{-8}	10^{-9}	10^{-10}	10^{-11}	10^{-12}	10^{-14}
iter	30	11	5	3	2	2	1
$\inf J$	9.6e-21	6.8e-22	2.9e-25	1.3e-25	7.0e-27	8.2e-31	1.5e-24
$\ \mathbf{u}_f - \mathbf{u}_d\ _{L^2(\Omega_{fd})}$	3.9e-6	1.7e-6	1.9e-6	1.9e-6	1.9e-6	1.9e-6	1.9e-6
$\ p_f - p_d\ _{L^2(\Omega_{fd})}$	4.5e-10	2.7e-9	4.7e-9	5.2e-9	5.3e-9	5.3e-9	5.3e-9

- Dependence on δ

δ	$7.5 \cdot 10^{-4}$	$7.5 \cdot 10^{-5}$	$7.5 \cdot 10^{-6}$
iter ($K = 10^{-8}$)	4	10	14
iter ($K = 10^{-12}$)	1	2	2

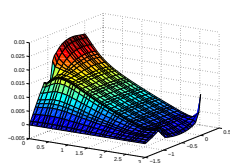
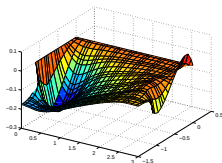
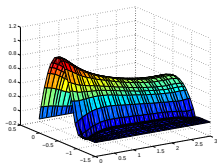
Physical considerations lead to choose $\delta \approx \sqrt{|K|}$.

- Dependence on h

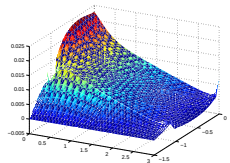
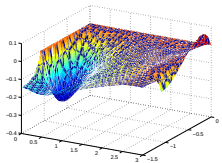
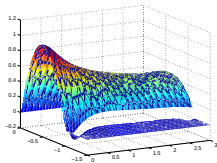
h	0.1	0.05	0.025
iter ($K = 10^{-8}$)	16	30	41
iter ($K = 10^{-12}$)	2	2	2

Comparison with the “sharp-interface” solution ($K = 10^{-7} \text{ m}^2$)

ICDD:



Sharp interface (Beavers-Joseph-Saffman):



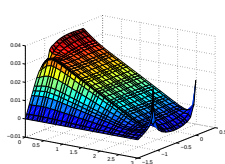
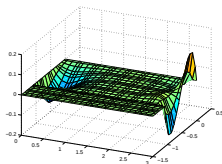
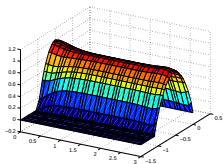
Horizontal velocity

Vertical velocity

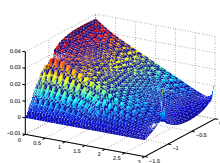
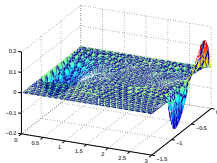
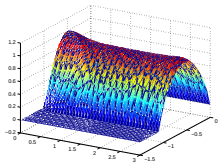
Hydrodynamic pressure

Comparison with the “sharp-interface” solution ($K = 10^{-12} \text{ m}^2$)

ICDD:



Sharp interface (Beavers-Joseph-Saffman):



Horizontal velocity

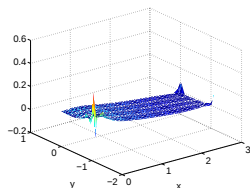
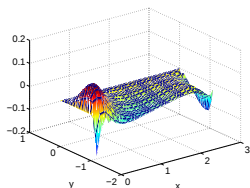
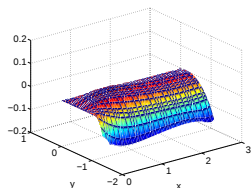
Vertical velocity

Hydrodynamic pressure

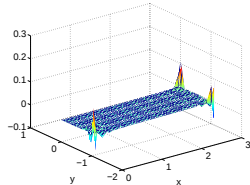
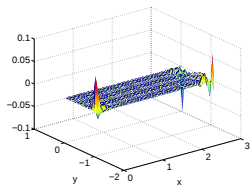
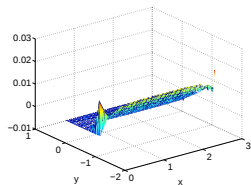
Comparison with the “sharp-interface” solution

Stokes domain

$$K = 10^{-7} \text{ m}^2$$



$$K = 10^{-12} \text{ m}^2$$



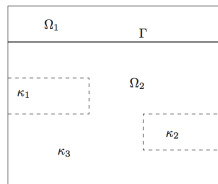
$$\frac{u_{ICDD} - u_{BJS}}{\|u_{BJS}\|_{\infty}}$$

$$\frac{v_{ICDD} - v_{BJS}}{\|v_{BJS}\|_{\infty}}$$

$$\frac{p_{ICDD} - p_{BJS}}{\|p_{BJS}\|_{\infty}}$$

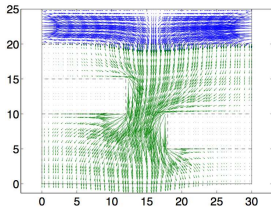
Numerical results (ii)

$K_1 = 10^{-10}$, $K_2 = 10^{-14}$, $K_3 = 10^{-6} \text{ m}^2$
 $\delta = 0.05\%$, convergence in 4 iterations
($\text{tol} = 10^{-9}$)

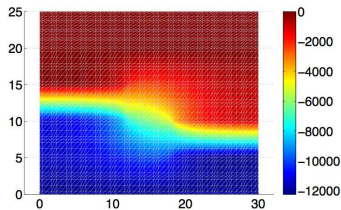


ICDD:

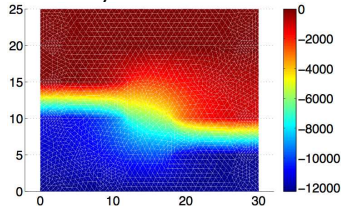
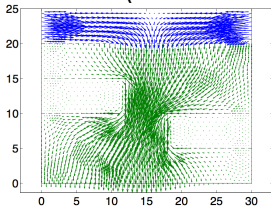
Velocity field



Hydrodynamic pressure

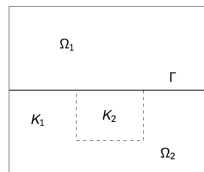


Sharp interface (Beavers-Joseph-Saffman):



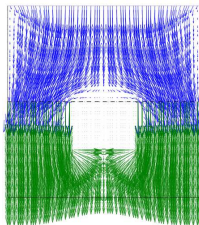
Numerical results (iii)

$K_1 = 10^{-7}$, $K_2 = 10^{-10} \text{ m}^2$, $\text{Re} = 100$.
 $\delta = 0.05\%$, convergence in 2 iterations
($\text{tol} = 10^{-9}$)

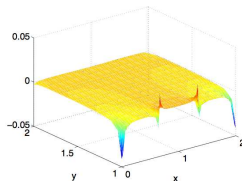


ICDD:

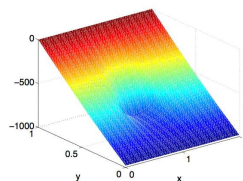
Velocity field



Hydrodynamic pressure Stokes



Hydrodynamic pressure Darcy



This case is not meaningful using Beavers-Joseph-Saffman:
ICDD can treat more general fluid regimes.

Conclusions

- “Sharp-interface” approach:
An interface problem and ad-hoc preconditioners for the Schur complement system (Steklov-Poincaré equation).
- “Transition-region” approach:
A minimization problem which takes into account the physical properties of the problem without relying on any specific interface conditions.

Optimized Schwarz (Robin-Robin type) preconditioner (i)

How to deal with intermediate values of Da ?

Considering the Schur complement of the linear system associated to the Stokes-Darcy problem with respect to both $(\mathbf{u}_f)_n$ and $p_{d|\Gamma}$, we obtain the following *augmented interface system*:

$$\begin{pmatrix} \boldsymbol{\Sigma}_s & \mathbf{C}_\Gamma^T \\ -\mathbf{C}_\Gamma & \boldsymbol{\Sigma}_p \end{pmatrix} \begin{pmatrix} (\mathbf{u}_f)_n \\ p_{d|\Gamma} \end{pmatrix} = \begin{pmatrix} \chi_s \\ \chi_p \end{pmatrix}$$

where $\boldsymbol{\Sigma}_p \approx \boldsymbol{\Sigma}_d^{-1}$.

[MD (2013)]

Optimized Schwarz (Robin-Robin type) preconditioner (ii)

Consider the **Robin-Robin method**:

- Solve the Stokes problem in Ω_f with interface conditions on Γ :

$$\begin{aligned} -\boldsymbol{\tau} \cdot \boldsymbol{\sigma}(\mathbf{u}_f^{(k)}, p_f^{(k)}) \cdot \mathbf{n} &= \frac{\alpha_d \mu}{\sqrt{\boldsymbol{\tau} \cdot \mathbf{K} \cdot \boldsymbol{\tau}}} \mathbf{u}_f^{(k)} \cdot \boldsymbol{\tau} \\ -\mathbf{n} \cdot \boldsymbol{\sigma}(\mathbf{u}_f^{(k)}, p_f^{(k)}) \cdot \mathbf{n} - \alpha_f \mathbf{u}_f^{(k)} \cdot \mathbf{n} &= p_d^{(k-1)} - \alpha_f \mathbf{u}_d^{(k-1)} \cdot \mathbf{n} \end{aligned}$$

- Solve the Darcy problem in Ω_d with interface condition on Γ :

$$p_d^{(k)} + \alpha_p \mathbf{u}_d^{(k)} \cdot \mathbf{n} = -\mathbf{n} \cdot \boldsymbol{\sigma}(\mathbf{u}_f^{(k)}, p_f^{(k)}) \cdot \mathbf{n} + \alpha_p \mathbf{u}_f^{(k)} \cdot \mathbf{n}$$

where the parameters α_f and α_p must be chosen in a suitable way to ensure (optimal) convergence.

Optimized Schwarz (Robin-Robin type) preconditioner (iii)

The Robin-Robin method can be proved to be equivalent to a Gauss-Seidel scheme for the augmented interface Schur complement system multiplied by the preconditioner:

$$P_{OS}^{-1} = \begin{pmatrix} I & -\alpha_f I \\ \alpha_p^{-1} I & I \end{pmatrix}$$

which can be used in the framework of Krylov methods like GMRES or BiCGStab.

The theory of **Optimized Schwarz methods** allows to characterize α_f and α_p to maximize the convergence rate of the method.

[MD, Gerardo-Giorda (in preparation)]

Numerical results (i)

Domain $\Omega_f = (0, 1) \times (1, 2)$ and $\Omega_d = (0, 1) \times (0, 1)$. Solution

$$\mathbf{u}_f = ((y-1)^2 + (y-1) + \sqrt{\text{Da}}, x(x-1))$$

$$p_f = \frac{2}{\text{Re Eu}}(x + y - 1)$$

$$p_d = \frac{1}{\text{Re Eu}} \left(\frac{1}{\text{Da}}(x(1-x)(y-1) + (y-1)^3/3) + 2x \right)$$

We set $\text{Re} = 10^3$ and $\text{Da} = 10^{-6}, 10^{-8}, 10^{-10}, 10^{-12} \text{ m}^2$.

Using optimized Schwarz methods we can estimate

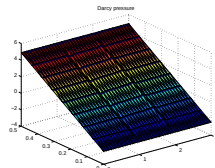
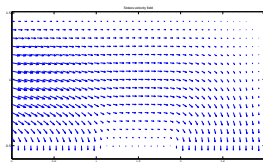
$$\alpha_f = 5.7873, \alpha_p = 0.0528.$$

Da:	10^{-6}	10^{-8}	10^{-10}	10^{-12}
$h = 1/8$	18	18	18	18
$h = 1/16$	20	20	18	18
$h = 1/32$	17	20	20	20

BiCGStab iterations with $\text{tol} = 10^{-10}$

Numerical results (iii)

Setting as in [Hanspal et al, 2009] but with discontinuous permeability (reduced by one order of magnitude) and inflow from the left (no-slip conditions on the remaining fluid boundary):



Here $Re = 500$ and Da between $1.34 \cdot 10^{-5}$ and $1.34 \cdot 10^{-6}$.
Optimal coefficients: $\alpha_f = 13.24$ and $\alpha_p = 0.024$.

h :	3/10	3/20	3/40
elements	120	480	1920
iterations	19	20	20

BiCGStab iterations with $tol = 10^{-10}$

The ICDD method

ICDD is inspired to both the [Virtual Control Method](#) by J.L. Lions, O. Pironneau (1998) and the [Least Squares Method](#) by R. Glowinski, Q. Dinh, J. Periaux (1983).

VCM/LSM	ICDD
$J(\lambda) = \frac{1}{2} \ u_1 - u_2\ _{L^2(\Omega_{12})}$ distributed observation classical Optimality System (integration by parts)	$J(\lambda) = \frac{1}{2} \ u_1 - u_2\ _{L^2(\Gamma_1 \cup \Gamma_2)}$ interface observation non-standard OS

[Interface observation](#) makes ICDD [more suitable for heterogeneous problems](#) and it leads to more efficient algorithms than distributed observation.