

Scientific Computation of Two-Phase Ferrofluid Flows

Gareth Johnson

Faculty Adviser: Prof. Ricardo Nochetto

University of Maryland
AMSC 664: Advanced Scientific Computing II
Supported by Johns Hopkins University Applied Physics Lab

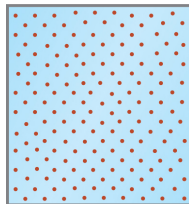
May 15, 2019

What is a Ferrofluid?

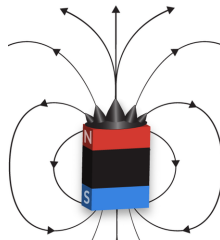
Awesome video of ferrofluids.

What is a Ferrofluid? (Backup)

A ferrofluid is a colloid of nanoscale ferromagnetic particles suspended in a carrier fluid such as oil, water, or an organic solvent.

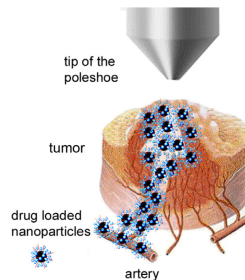


Ferrofluids become magnetized when under the effect of a magnetic field.



Applications

- Initially created to pump rocket fuel once a spacecraft entered a weightless environment.
- Commercial applications:
 - Vibration damping
 - Sensors
 - Acoustics
- Recent research areas:
 - Magnetic drug targeting
 - Adaptive deformable mirrors



PDE Model for Two-Phase Ferrofluid Flow

- Dr. Nochetto and collaborators developed a model for two-phase ferrofluid flows and devised an energy stable numerical scheme [3].
- The model was not derived, but instead was assembled.
- Important results from [3]:
 - Proved an energy law for the PDE model.
 - Proved the numerical scheme was energy stable and the existence of a local solution.
 - For an even simpler model, they proved stability, convergence, and the existence of solutions.

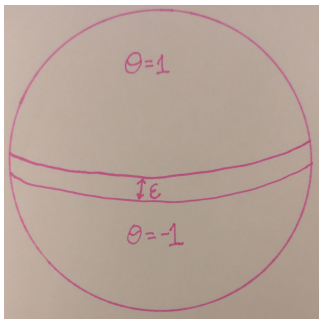
Modeling a Two-Phase Fluid

- In order to track both fluids, a diffuse interface is used.
- The phase variable θ is introduced, which takes values in $[-1, 1]$.
- The evolution of θ is given by a modified Cahn–Hilliard equation:

$$\begin{cases} \theta_t + \operatorname{div}(\mathbf{u}\theta) + \gamma \Delta \psi = 0 & \text{in } \Omega \\ \psi - \epsilon \Delta \theta + \frac{1}{\epsilon} f(\theta) = 0 & \text{in } \Omega \\ \partial_\eta \theta = \partial_\eta \psi = 0 & \text{on } \Gamma, \end{cases}$$

where

- $0 < \epsilon \ll 1$ is related to the interface thickness,
- $\gamma > 0$ is the constant mobility,
- ψ is the chemical potential,
- $f(\theta)$ is the truncated double well potential.



Modeling of the Magnetic Field

- Instead of using the magnetostatics equations, a simplified approach was used.
- Define the magnetic field by

$$\mathbf{h} := \mathbf{h}_a + \mathbf{h}_d,$$

where

- \mathbf{h}_a – smooth harmonic (curl-free and div-free) applied magnetizing field,
- \mathbf{h}_d – demagnetizing field.
- Then the magnetic field is induced via the scalar potential φ by

$$\mathbf{h} = \nabla \varphi,$$

along with,

$$-\Delta \varphi = \operatorname{div}(\mathbf{m} - \mathbf{h}_a) \quad \text{in } \Omega, \quad \partial_\eta \varphi = (\mathbf{h}_a - \mathbf{m}) \cdot \boldsymbol{\eta} \quad \text{on } \Gamma.$$

Modeling of Ferrohydrodynamics

- A simplified version of Shliomis model is used, which couples an advection–reaction equation for the magnetization \mathbf{m} :

$$\mathbf{m}_t + (\mathbf{u} \cdot \nabla) \mathbf{m} = -\frac{1}{\mathcal{T}}(\mathbf{m} - \kappa_\theta \mathbf{h}),$$

with the Navier–Stokes equations of incompressible fluids for the velocity–pressure pair (\mathbf{u}, p) :

$$\begin{aligned} \mathbf{u}_t + (\mathbf{u} \cdot \nabla) \mathbf{u} - \operatorname{div}(\nu_\theta \mathbf{T}(\mathbf{u})) + \nabla p &= \mu_0(\mathbf{m} \cdot \nabla) \mathbf{h} + \frac{\lambda}{\epsilon} \theta \nabla \psi, \\ \operatorname{div} \mathbf{u} &= 0, \end{aligned}$$

where

- \mathcal{T} is the relaxation time of the ferrofluid,
 - κ_θ is the magnetic susceptibility of the phase variable,
 - ν_θ is the viscosity of the phase variable,
 - μ_0 is the constitutive parameter related to the Kelvin force,
 - $\frac{\lambda}{\epsilon} \theta \nabla \psi$ is the capillary force.
- This is supplemented with a no-slip condition on the boundary:

$$\mathbf{u} = 0 \quad \text{on } \Gamma.$$

- The model reads: Consider a bounded convex polygon/polyhedron domain $\Omega \subset \mathbb{R}^d$ ($d = 2$ or 3) with boundary Γ . The evolution of the system is given by the following set of equations in strong form in Ω

$$\theta_t + \operatorname{div}(\mathbf{u}\theta) + \gamma\Delta\psi = 0, \quad (1a)$$

$$\psi - \epsilon\Delta\theta + \frac{1}{\epsilon}f(\theta) = 0, \quad (1b)$$

$$\mathbf{m}_t + (\mathbf{u} \cdot \nabla)\mathbf{m} = -\frac{1}{\mathcal{J}}(\mathbf{m} - \kappa_\theta\mathbf{h}), \quad (1c)$$

$$-\Delta\varphi = \operatorname{div}(\mathbf{m} - \mathbf{h}_a), \quad (1d)$$

$$\mathbf{u}_t + (\mathbf{u} \cdot \nabla)\mathbf{u} - \operatorname{div}(\nu_\theta\mathbf{T}(\mathbf{u})) + \nabla p = \mu_0(\mathbf{m} \cdot \nabla)\mathbf{h} + \frac{\lambda}{\epsilon}\theta\nabla\psi, \quad (1e)$$

$$\operatorname{div}\mathbf{u} = 0, \quad (1f)$$

for every $t \in [0, t_F]$, where $\mathbf{T}(\mathbf{u}) = \frac{1}{2}(\nabla\mathbf{u} + \nabla\mathbf{u}^T)$ denotes the symmetric gradient and $\mathbf{h} = \nabla\varphi$. The system (1) is supplemented with the boundary conditions

$$\partial_\eta\theta = \partial_\eta\psi = 0, \quad \mathbf{u} = 0, \quad \text{and} \quad \partial_\eta\varphi = (\mathbf{h}_a - \mathbf{m}) \cdot \boldsymbol{\eta} \quad \text{on } \Gamma. \quad (2)$$

Define the backward difference operator $\delta f^k = f^k - f^{k-1}$.

For given smooth initial data $\{\Theta^0, \mathbf{M}^0, \mathbf{U}^0\}$ and timestep τ , compute $\{\Theta^k, \Psi^k, \mathbf{M}^k, \Phi^k, \mathbf{U}^k, P^k\} \in \mathbb{G}_h \times \mathbb{Y}_h \times \mathbb{M}_h \times \mathbb{X}_h \times \mathbb{U}_h \times \mathbb{P}_h$ for every $k \in \{1, \dots, K\}$ that solves

$$\left(\frac{\delta \Theta^k}{\tau}, \Lambda \right) - (\mathbf{U}^k \Theta^{k-1}, \nabla \Lambda) - \gamma (\nabla \Psi^k, \nabla \Lambda) = 0, \quad (3a)$$

$$(\Psi^k, \Upsilon) + \epsilon (\nabla \Theta^k, \nabla \Upsilon) + \frac{1}{\epsilon} (f(\Theta^{k-1}), \Upsilon) + \frac{1}{\eta} (\delta \Theta^k, \Upsilon) = 0, \quad (3b)$$

$$\left(\frac{\delta \mathbf{M}^k}{\tau}, \mathbf{Z} \right) - \mathcal{B}_h^m(\mathbf{U}^k, \mathbf{Z}, \mathbf{M}^k) + \frac{1}{\mathcal{J}}(\mathbf{M}^k, \mathbf{Z}) = \frac{1}{\mathcal{J}}(\varkappa_\theta \mathbf{H}^k, \mathbf{Z}), \quad (3c)$$

$$(\nabla \Phi^k, \nabla X) = (\mathbf{h}_a^k - \mathbf{M}^k, \nabla X), \quad (3d)$$

$$\begin{aligned} \left(\frac{\delta \mathbf{U}^k}{\tau}, \mathbf{V} \right) + \mathcal{B}_h(\mathbf{U}^{k-1}, \mathbf{U}^k, \mathbf{V}) + (\nu_\theta \mathbf{T}(\mathbf{U}^k), \mathbf{T}(\mathbf{V})) - (P^k, \operatorname{div} \mathbf{V}) &= \mu_0 \mathcal{B}_h^m(\mathbf{V}, \mathbf{H}^k, \mathbf{M}^k) \\ &+ \frac{\lambda}{\epsilon} (\Theta^{k-1} \nabla \Psi^k, \mathbf{V}), \end{aligned} \quad (3e)$$

$$(Q, \operatorname{div} \mathbf{U}^k) = 0. \quad (3f)$$

Numerical Implementation Details

Discretization of the Numerical Scheme:

- Time Discretization: Backward Euler is used.
- Space Discretization: A mix of Continuous and Discontinuous Galerkin is used, approximating the spaces with polynomials of degree 2 in each variable (i.e. Q_2 elements).
 - Continuous: Cahn–Hilliard, Magnetic potential, and Navier Stokes equations.
 - Discontinuous: Magnetization equations.

Fixed Point Solver:

- A Picard–like iteration is used.
- Utilizes the "lagging" of the velocity \mathbf{U} to solve each subsystem.
- Iterates until a fixed point for \mathbf{U}^k is reached.
- Given \mathbf{U}^{k-1}
 - 1) Compute Θ^k and Ψ^k substituting \mathbf{U}^{k-1} for \mathbf{U}^k .
 - 2) Next compute \mathbf{M}^k and Φ^k using (Θ^k, Ψ^k) from the previous iteration and substituting \mathbf{U}^{k-1} for \mathbf{U}^k .
 - 3) Finally, compute \mathbf{U}^k and P^k using $(\Theta^k, \Psi^k, \mathbf{M}^k, \Phi^k)$ from the previous two iterations.
 - 4) Repeat steps 1-3 using \mathbf{U}^k from the previous iteration as input until \mathbf{U}^k does not change between iterations.

Semester Plan

- Implement and unit test solvers for each subsystem in "isolation".
 - Cahn–Hilliard (3a)–(3b)
 - Navier–Stokes (3e)–(3f)
 - Magnetization (3c)–(3d)
- Combine the solvers using a Picard iteration to implement the full two phase ferrofluid flow.

Adaptive Mesh Refinement

For adaptive refinement, we used a method of marking cells for refinement known as Dofler marking. Our adaptive refinement procedure is:

- Compute the element indicator η_T on each element, where

$$\eta_T^2 = h_T \int_{\partial T} \left| \left[\left[\frac{\partial \Theta}{\partial \eta} \right] \right] \right|^2 dS \quad \forall T \in \mathcal{T}_h,$$

and $[\![\cdot]\!]$ is the jump across the face.

- Order the elements from highest to lowest based on their element indicator value.
- Compute the set of elements whose error makes up 55 percent of the total error for refinement, starting with the higher valued elements.
- Compute the set of elements whose error makes up 5 percent of the total error for coarsening, starting with the lowered valued elements.
- Refine/coarsen the mesh and transfer the solution from the old mesh to the new mesh.

We enforce a limit on the number of times an element can be refined. This procedure is performed every 5 iterations and on the initial condition.

Cahn–Hilliard Matrix Structure

- Rearranging the system we have that

$$\begin{aligned} \left(\Theta^k, \Lambda \right) - \tau \gamma (\nabla \Psi^k, \nabla \Lambda) &= \left(\Theta^{k-1}, \Lambda \right) + \tau (\mathbf{U}^k \Theta^{k-1}, \nabla \Lambda), \\ (\Psi^k, \Upsilon) + \epsilon (\nabla \Theta^k, \nabla \Upsilon) + \frac{1}{\eta} (\Theta^k, \Upsilon) &= \frac{1}{\eta} (\Theta^{k-1}, \Upsilon) - \frac{1}{\epsilon} (f(\Theta^{k-1}), \Upsilon). \end{aligned}$$

- This leads to the following block matrix form

$$\begin{pmatrix} M & -\tau \gamma K \\ \epsilon K + \frac{1}{\eta} M & M \end{pmatrix} \begin{pmatrix} \Theta^k \\ \Psi^k \end{pmatrix} = \begin{pmatrix} f \\ g \end{pmatrix},$$

where M and K are the usual mass and stiffness matrices and f and g are the respective right hand sides.

Reduced Phase System

- We eliminate the chemical potential by adding $\tau\gamma KM^{-1}$ the second row to the first.
- Doing so yields the following reduced phase system

$$\left(M + \tau\epsilon\gamma KM^{-1}K + \frac{\tau\gamma}{\eta}K\right)\Theta^k = f + \tau\gamma KM^{-1}g.$$

- The chemical potential is then found by solving

$$M\Psi^k = g - \left(\epsilon K + \frac{1}{\eta}M\right)\Theta^k.$$

Solving the Reduced Phase System

- In order to precondition the reduced phase system, we consider the following preconditioner

$$P = M + \tau\gamma\epsilon K \text{diag}(M)^{-1} K + \tau \frac{\gamma}{\eta} K.$$

- When solving with GMRES, we then use an incomplete LU preconditioner of the preconditioner matrix P .
- Note: We never explicitly compute the matrix $M + \tau\epsilon\gamma KM^{-1}K + \frac{\tau\gamma}{\eta} K$. When solving with GMRES, you only need the action of the matrix. This allows us to not compute the inverse of M as we can compute its action on a vector x ,

$$y = M^{-1}x,$$

by solving

$$My = x$$

using CG.

- Finally, we solve for the chemical potential using CG by using the action of the inverse of M .

Cahn–Hilliard Solver Verification

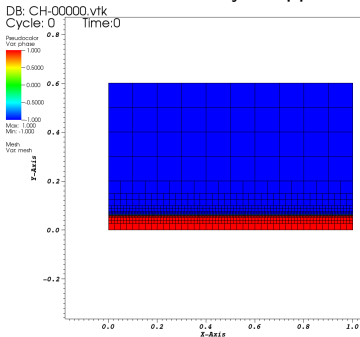
We verify the solver using three unit tests.

- 1) A flat initial profile with no forcing and adaptive refinement.
- 2) A circular profile with no forcing and adaptive refinement.
- 3) A forced solution on a series of globally refined grids. This is done to compute the convergence rate.

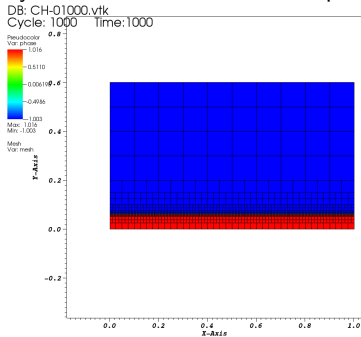
The initial condition for the flat profile was

$$f(x, y) = \begin{cases} -1 & y > .06 \\ 1 & \text{otherwise} \end{cases}.$$

The maximum level of refinement was 5 and the initial data was refined 20 times. Model parameters were chosen as $\epsilon = .001$, $\gamma = .2$, $\eta = .000001$. The error of the solution stayed approximately constant over 1000 time steps.



Time step 0

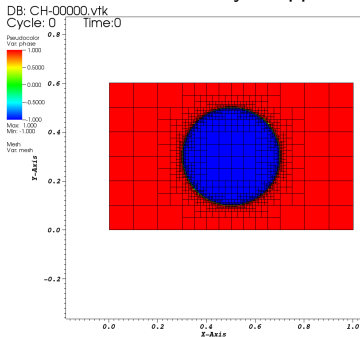


Time step 1000

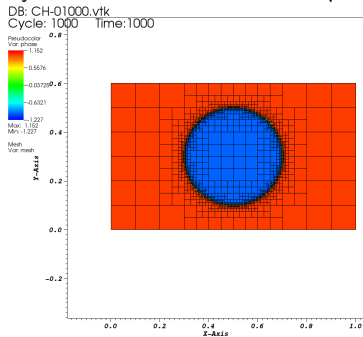
The initial condition for the circular profile was

$$f(x, y) = \begin{cases} -1 & (x, y) \in B((.5, .3), .2) \\ 1 & \text{otherwise} \end{cases}$$

The maximum level of refinement was 5 and the initial data was refined 20 times. Model parameters were chosen as $\epsilon = .001$, $\gamma = .2$, $\eta = .000001$. The error of the solution stayed approximately constant over 1000 time steps.



Time step 0



Time step 1000

The forcing function chosen was $f(x, y) = \cos(2\pi x)\cos(2\pi y)$ and $\Omega = [-1, 1]^2$. Model parameters were chosen as $\epsilon = .2$, $\gamma = .2$, $\eta = .000001$. The error was computed after 1000 iterations.

n cells		H^1 -error			L^2 -error		
2	16	4.264e+00	-	-	3.109e-01	-	-
3	64	8.079e-01	5.28	2.40	2.658e-02	11.70	3.55
4	256	2.038e-01	3.96	1.99	3.310e-03	8.03	3.01
5	1024	5.104e-02	3.99	2.00	4.125e-04	8.02	3.00
6	4096	1.277e-02	4.00	2.00	5.152e-05	8.01	3.00

This is the convergence rate that we expect for both the H^1 and L^2 error for Q_2 elements.

Navier–Stokes Block Matrix Structure

The system has following block matrix form

$$\begin{pmatrix} F & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{U}^k \\ p^k \end{pmatrix} = \begin{pmatrix} f \\ 0 \end{pmatrix},$$

where

$$F = (\mathbf{U}^k, \mathbf{V}) + \tau \mathcal{B}_h(\mathbf{U}^{k-1}, \mathbf{U}^k, \mathbf{V}) + \tau(\nu_\theta \mathbf{T}(\mathbf{U}^k), \mathbf{T}(\mathbf{V})),$$

$$B = (Q, \operatorname{div} \mathbf{U}^k),$$

and

$$f = (\mathbf{U}^{k-1}, \mathbf{V}) + \tau \mu_0 \mathcal{B}_h^m(\mathbf{V}, \mathbf{H}^k, \mathbf{M}^k) + \tau \frac{\lambda}{\epsilon} (\Theta^{k-1} \nabla \Psi^k, \mathbf{V}).$$

We would like to use the block preconditioner

$$P = \begin{pmatrix} F & 0 \\ B & -S \end{pmatrix}$$

where S is the the Schur Complement $S = B^T F^{-1} B$. This preconditioner is desirable as it has the property that

$$P^{-1} \begin{pmatrix} F & B^T \\ B & 0 \end{pmatrix} = \begin{pmatrix} I & F^{-1} B^T \\ 0 & I \end{pmatrix}.$$

Since we don't want to compute inverse matrices, we instead consider the following approximation for P^{-1}

$$P^{-1} \approx \begin{pmatrix} \tilde{F}^{-1} & 0 \\ \tilde{S}^{-1} B F^{-1} & -\tilde{S}^{-1} \end{pmatrix}$$

where \tilde{F}^{-1} is computed as an inverse action and \tilde{S}^{-1} is the Least Squares Commutator [1] defined as

$$\tilde{S}^{-1} = (B \text{diag}(M)^{-1} B^T)^{-1} (B \text{diag}(M)^{-1} F \text{diag}(M)^{-1} B^T) (B \text{diag}(M)^{-1} B^T)^{-1},$$

and M is the mass matrix for the velocity.

The preconditioner was implemented to compute $Y = P^{-1}X$, where X, Y are block vectors. It does so in three steps.

- Compute

$$Y_0 = \tilde{F}^{-1} X_0.$$

- Then compute in a temporary vector N

$$N = X_1 - BX_0 = X_1 - B\tilde{F}^{-1} X_0.$$

- Finally compute

$$Y_1 = S^{-1}N = S^{-1}(X_1 - B\tilde{F}^{-1} X_0).$$

Using this preconditioner the system is solved using GMRES.

Navier–Stokes Solver Verification

We verify the solver using the following forced solution

$$\mathbf{u}(x, y, t) = \begin{pmatrix} t \sin(\pi x) \sin(\pi(y + .5)) \\ t \cos(\pi x) \cos(\pi(y + .5)) \end{pmatrix}, \quad p(x, y, t) = \sin(2\pi(x - y) + t),$$

on $\Omega = [0, 1]^2$. The model parameters are $\mu = 1$, $\lambda = .05$, $\epsilon = .2$. The system was solved until $t = 2$ with 1000 time steps. The system was solved on a uniformly refined grid with levels 3, 4, 5.

n cells		L^2 -error
3	64	0.642479
4	256	0.642518
5	1024	0.64252

Deliverables

- Cahn–Hilliard solver and unit tests
- Navier–Stokes solver and unit test
- Summary of the capabilities of the code will be published on the github repo
- List of software requirements
- Instructions on how to run each of the codes and how to alter initial conditions and forcing functions

Figure References

- Slide 1: https://youtu.be/wHZDgSFzQ_s?t=12
- Slide 2:
 - https://www.researchgate.net/profile/Vikram_Raghavan2/post/What_is_the_effect_of_magnetic_field_on_alignment_of_ferro_fluid_droplet/attachment/59d622166cda7b8083a1b9a2/AS%3A273810673078272%401442292959057/download/Effect+of+Magnetic+field.jpg
 - https://ksr-ugc.imgix.net/assets/003/310/641/f0ef73d1fd99f6aa5d96872168478df4_original.png?v=1424378871&w=680&fit=max&auto=format&lossless=true&s=c183d857603c12de82a71f3139283d9e
 - https://opentextbc.ca/chemistry/wp-content/uploads/sites/150/2016/05/CNX_Che_11_05_Colloid.jpg
- Slide 3: [2]
- Slide 12: [3]
- Slide 13: [3]
- Slide 18: [3]

References I



H. ELMAN, V. HOWLE, J. SHADID, R. SHUTTLEWORTH, AND R. TUMINARO, *Block preconditioners based on approximate commutators*, SIAM Journal on Scientific Computing, 27 (2006), pp. 1651–1668.



C. JANKO, S. DÜRR, L. E. MUNOZ, S. LYER, R. CHAURIO, R. TIETZE, S. V. LÖHNEYSSEN, C. SCHORN, M. HERRMANN, AND C. ALEXIOU, *Magnetic drug targeting reduces the chemotherapeutic burden on circulating leukocytes*, International Journal of Molecular Sciences, 14 (2013), pp. 7341–7355.



R. H. NOCHETTO, A. J. SALDAGO, AND I. TOMAS, *A diffuse interface model for two-phase ferrofluid flows*, Computer Methods in Applied Mechanics and Engineering, 309 (2016), pp. 497–531.