

Moment-of-Fluid Analytic Reconstruction on 2D Cartesian grids

Antoine LEMOINE¹ Stéphane GLOCKNER² Jérôme BREIL³

¹Univ. Bordeaux, I2M, UMR 5295, F-33400 Talence, France.

²Bordeaux INP, I2M, UMR 5295, F-33400 Talence, France.

³CEA CESTA, 15 Avenue des Sablières, CS 60001, 33116 Le Barp Cedex, France.

June 6th, 2016

Introduction: Notus



Overview:

- Open-source CFD code <http://notus-cfd.org>
- Dedicated to modeling and simulation of incompressible fluid flows
- Designed to run on massively parallel supercomputers
- Finite volumes, 2D & 3D Cartesian staggered grid

Multiphysic applications:

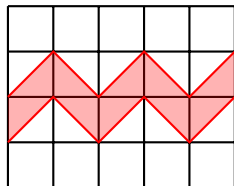
- Incompressible Navier-Stokes equations
- Multiphase flows → breaking waves
- Energy equation → energy storage, phase change
- Fluid – structure interactions (elasticity)

Many applications require interface **representation** and **reconstruction**

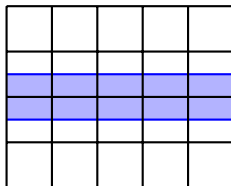
Introduction: objective

Objective: sharp interface reconstruction \rightarrow Piecewise Linear Construction (PLIC)

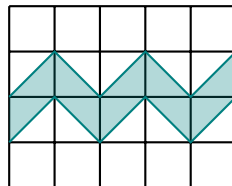
VOF-PLIC vs MOF:



Original interface



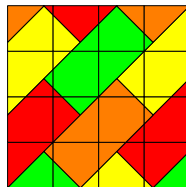
VOF-PLIC reconstruction



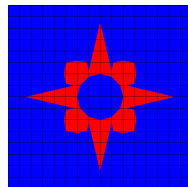
MOF reconstruction

Moment-of-Fluid (MOF) :

- Uses volume fraction + **centroids**
- Stencil reduced to only 1 cell
- 2nd order of convergence
- Multimaterial reconstruction



*Dyadechko & Shashkov
(2006)*



Example: compass

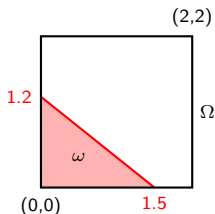
Moment-of-fluid: Moment?

0th order momentum (volume)

$$M_0(\omega) = \int_{\omega} d\mathbf{x} = |\omega|$$

Volume fraction (relative to a cell Ω)

$$\mu(\omega) = \frac{M_0(\omega)}{M_0(\Omega)}$$



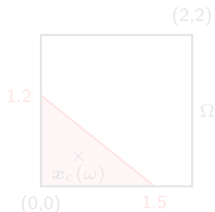
- $M_0(\Omega) = 4$
- $M_0(\omega) = 0.9$
- $\mu(\omega) = 0.225$

1st order momentum

$$M_1(\omega) = \int_{\omega} \mathbf{x} d\mathbf{x}$$

Centroid

$$\mathbf{x}_c(\omega) = \frac{M_1(\omega)}{M_0(\omega)}$$



- $M_0(\omega) = 0.9$
- $M_1(\omega) = (0.45, 0.36)$
- $\mathbf{x}_c(\omega) = (0.5, 0.4)$

MOF uses volume fraction and centroids to reconstruct interfaces

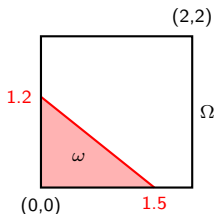
Moment-of-fluid: Moment?

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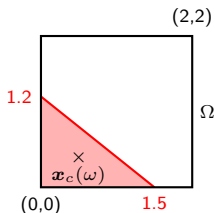
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- $\mathbf{x}_c(\omega) = (0.5, 0.4)$

MOF uses volume fraction **and** centroids to reconstruct interfaces

Moment-of-fluid: formulation

Two unknowns to reconstruct the interface: **angle** and **distance**

Available information:

- **Volume fraction** of any portions of fluid μ in each cells
- **Centroid** of any portions of fluid x_c in each cells

Reconstruction method:

- VOF-PLIC: $|\omega^\ell| = |\omega^*|$ for each cell
→ *under-determined* problem!
- MOF: $|\omega^\ell| = |\omega^*|$ and $x_c(\omega^\ell) = x_c(\omega^*)$ for each cell
→ *over-determined* problem!

Minimization problem:

- Find $\omega^\ell = \operatorname{argmin}_{\omega^\ell} |x_c(\omega^\ell) - x_c(\omega^*)|^2$ → Minimize centroid distance
- Under constraint $|\omega^\ell| = |\omega^*|$ → Preserve volume

Reference: Dyadechko, V., Shashkov, M. (2007)

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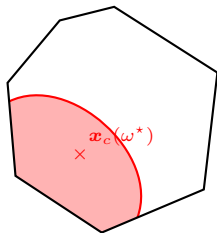
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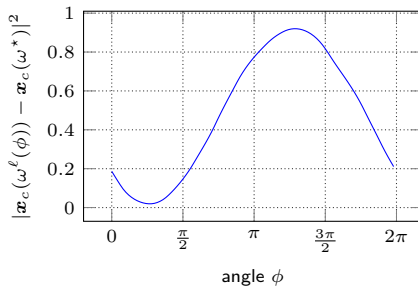
Minimization: example

Example parameters:

- $\mu(\omega^*) = 0.3$
- $\mathbf{x}_c(\omega^*) = (-0.3, -0.3)$



Objective function: \triangle possible local minima



Solution: $\phi \approx 0.841$

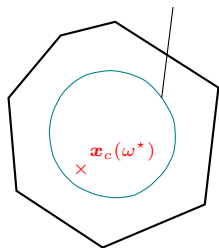
Portion of fluid with curved interface and its centroid $\mathbf{x}_c(\omega^*)$

Minimization: example

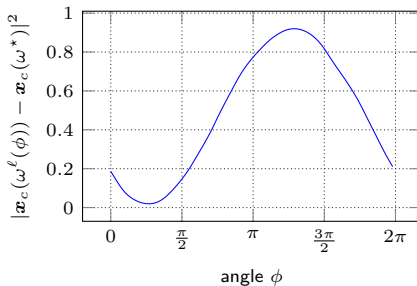
Example parameters:

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locus of the centroids for $\mu = 0.3$



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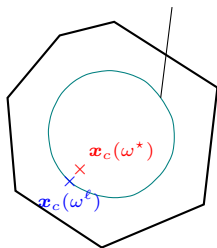
Locus of all possible centroids for piecewise linear reconstructions for $\mu = 0.3$

Minimization: example

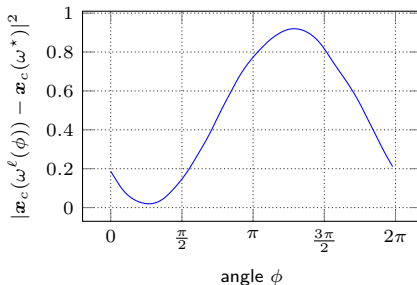
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Objective function: \triangle possible local minima



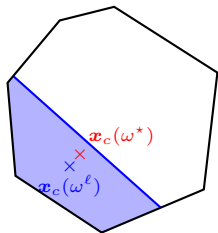
Solution: $\phi \approx 0.841$

Find the closest point $\mathbf{x}_c(\omega^\ell)$ on the curve to the reference centroid \rightarrow **minimization**
 \rightarrow The minimization algorithm stops when the **error on the angle** is small enough

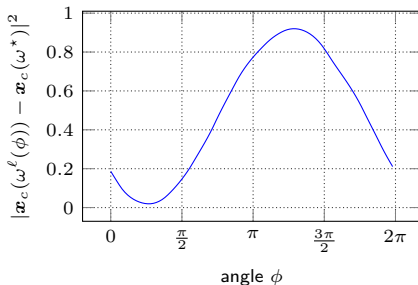
Minimization: example

Example parameters:

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Objective function: \triangle possible local minima



Solution: $\phi \approx 0.841$

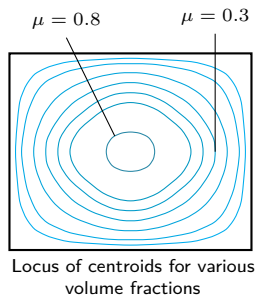
Compute the position of the interface with a **flood algorithm** (I used Breil *et al.* (2011))

Remark

One interface reconstruction per minimization iteration: highly time consuming

Analytic reconstruction: motivations & proposal

Objective: optimize the reconstruction on Cartesian grids

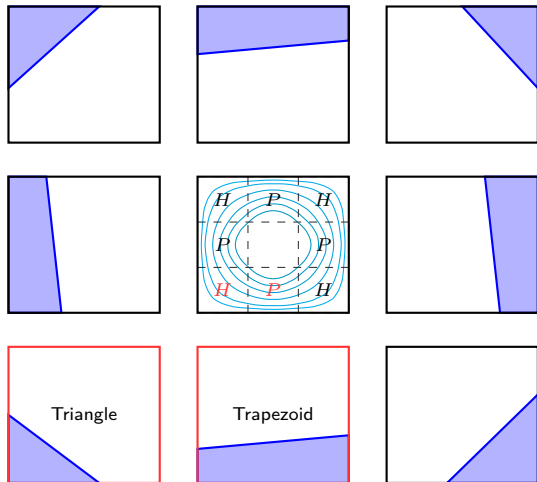


Idea:

- On Cartesian grids, cells are **rectangles**
- Use **symmetry** to reduce the number of configurations
- Possible **parametrization**
- **Analytic solution** to minimization problem

Bonus:

- Upgrade a VOF-PLIC algorithm
- Easier to implement than original method
- Faster for equivalent result
- Compatible with any rectangular meshes (e.g. AMR)

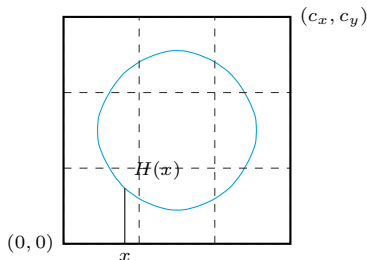
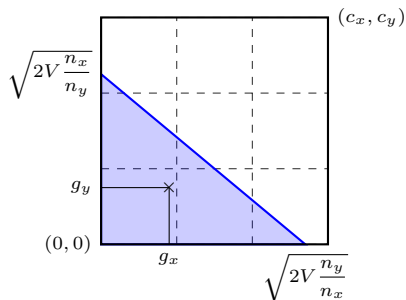
Analytic reconstruction: possible configurations ($\mu \leq 0.5$)

- Consider $\mu \leq 0.5$
- 8 configurations
- Reduced to 2 by symmetry
- Triangle \rightarrow **Hyperbola** H
- Trapezoid \rightarrow **Parabola** P

Find a **parametrization** of the parabola and the hyperbola

Analytic solution: parametrization of the hyperbola

Corner configuration \rightarrow shape: triangle \rightarrow centroid locus: hyperbola



Parameters

- Normal $\mathbf{n} = (n_x, n_y)$
- Volume V

Parametrization

$$\begin{cases} g_x = \frac{1}{3} \sqrt{2V \frac{n_y}{n_x}} \\ g_y = \frac{1}{3} \sqrt{2V \frac{n_x}{n_y}} \end{cases} \Rightarrow g_y = \frac{9V}{2g_x}$$

Analytic solution: closest point to the hyperbola

Problem

- Let $\mathbf{p} = (p_x, p_y)$ any point of \mathbb{R}^2 (e.g. the reference centroid)
- Find the closest point of \mathbf{p} to the hyperbola H
- For all $x \in \left[\frac{2V}{3c_x}, \frac{c_x}{3} \right]$ $H(x) = \frac{9V}{2x}$

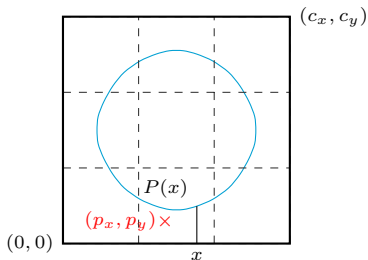
Solution

- The closest point of \mathbf{p} to the hyperbola is its **orthogonal projection**
- Tangent to the curve for the coordinate $g_x: (1, H'(g_x))$
- Orthogonal projection: $(g_x - p_x, H(g_x) - p_y) \cdot (1, H'(g_x)) = 0$

The x coordinate of $\mathbf{x}_c(\omega^\ell) = (x, H(x))$ is one of the solution of a **quartic** equation:

$$x^4 - p_x x^3 + \frac{2}{9} V p_y x - \left(\frac{2V}{9} \right)^2 = 0$$

Analytic solution: parabola



For all $x \in \left[\frac{c_x}{3}, \frac{2c_x}{3} \right]$

$$P(x) = \frac{V}{2c_x} + \frac{6V}{c_x} \left(\frac{1}{2} - \frac{x}{c_x} \right)^2$$

Let $\mathbf{p} = (p_x, p_y)$ any point of \mathbb{R}^2

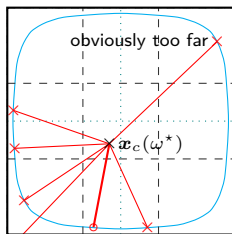
The closest point of \mathbf{p} to the parabola P is its orthogonal projection

The x coordinate of $\mathbf{x}_c(\omega^\ell) = (x, P(x))$ is one of the solution of a **cubic** equation:

$$x - p_x - \frac{12V}{c_x^2} \left(\frac{1}{2} - \frac{x}{c_x} \right) \left(\frac{V}{2c_x} - p_y \right) - \frac{72V^2}{c_x^3} \left(\frac{1}{2} - \frac{x}{c_x} \right)^3 = 0$$

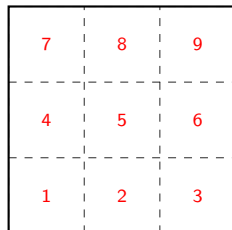
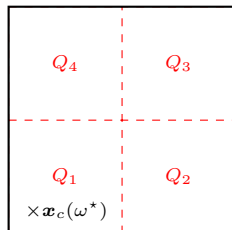
Analytic reconstruction: algorithm

Cubic and quartic equations \Rightarrow possible analytic solution



- Multiple solutions inside one configuration
 \rightarrow Maybe outside their definition domain
- Solutions found in many configurations
- Limit the search of solution in one quadrant

Analytic solution: algorithm



- ① If $\mu > 0.5$ solve the dual problem
- ② Locate the quadrant where $\mathbf{x}_c(\omega^*)$ is
 - $\mathbf{x}_c(\omega^*) \in Q_1$ try $\{1, 2, 4\}$
 - $\mathbf{x}_c(\omega^*) \in Q_2$ try $\{2, 3, 6\}$
 - $\mathbf{x}_c(\omega^*) \in Q_3$ try $\{6, 8, 9\}$
 - $\mathbf{x}_c(\omega^*) \in Q_4$ try $\{4, 7, 8\}$
- ③ Solve 2 cubic and 1 quartic
 - Strobach, Fast quartic solver (2010)
 - Strobach, Solving cubics by polynomial fitting (2011)
- ④ Eliminate wrong solutions
- ⑤ Find the closest solution
- ⑥ Compute n and d from the solution

Results

About 30% to 300% faster than minimization
 2nd order verified in time and space

Minimization vs Analytic: static reconstruction

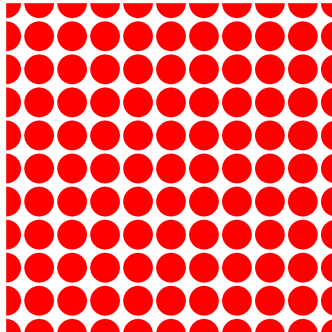
Static reconstruction (2 materials)

Mesh cells number: 2048^2

The minimization error is relative to the **angle**

Time ratio minimization / analytic:

	Min. 10^{-15}	Min. 10^{-11}	Min. 10^{-6}
Ana.	2.59	2.10	1.44



44 % to 159 % faster than minimization for static reconstruction

Minimization vs Analytic: dynamic reconstruction

Advection of 5 materials in a sheared flow

For mixed cells:

- First material uses **analytic reconstruction**
- Use **minimization** for remaining materials

Mesh: 128^2

Time ratio minimization / analytic :

	Min. 10^{-15}	Min. 10^{-11}	Min. 10^{-6}
Ana. & Min. 10^{-15}	2.27	1.84	1.33
Ana. & Min. 10^{-11}	2.43	1.96	1.42
Ana. & Min. 10^{-6}	2.67	2.16	1.56

Time ratio increases when mesh size increases

Conclusion & perspectives

Conclusion

- Faster than minimization
- Can be applied to any rectangular meshes (not only Cartesian grids)

Perspectives

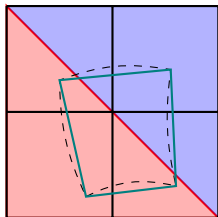
- Extension to 3D
 - No analytic solution (yet), but a parametrization
- Coupling with other methods: level-set for surface tensions computation

Thank you

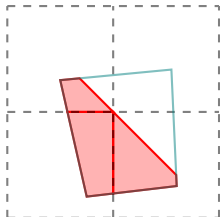
Appendix

Backward advection: Lagrangian remap

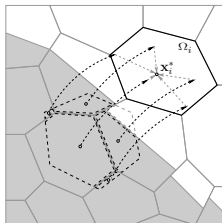
Backward advection: Compute the pre-image Ω^{n-1} of Ω^n with a Runge-Kutta 2 method.



Backward advection



Polygonal intersection
red fluid



Source: Dyadechko, V., Shashkov, M. (2007)

We can show that the centroids almost follow an advection equation:

$$\frac{d}{dt} \mathbf{x}_c(\omega) = \mathbf{v}(\mathbf{x}_c(\omega)) + \mathcal{O}(h^2)$$

→ Forward advection of the centroids (RK2)

Remark

Requires a polygon/polygon intersection algorithm

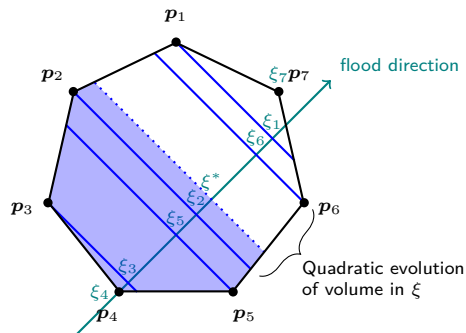
Fluid domain $\omega(t)$. Eulerian velocity $\mathbf{u}(\mathbf{x}, t)$. $\operatorname{div} \mathbf{u} = 0$.

$$\begin{aligned}
 \frac{d}{dt} \int_{\omega(t)} \mathbf{x} d\mathbf{x} &= \int_{\omega(t)} \left(\frac{\partial}{\partial t} \operatorname{Id}(\mathbf{x}) + \mathbf{u}(\mathbf{x}, t) \cdot \nabla \operatorname{Id}(\mathbf{x}) + \operatorname{Id}(\mathbf{x}) \operatorname{div} \mathbf{u}(\mathbf{x}, t) \right) d\mathbf{x} \\
 &= \int_{\omega(t)} \mathbf{u}(\mathbf{x}, t) d\mathbf{x} \\
 &= \int_{\omega(t)} \left(\mathbf{u}(\mathbf{x}_c, t) + [\nabla \mathbf{u}(\mathbf{x}_c, t)] (\mathbf{x} - \mathbf{x}_c) + \mathcal{O}(|\mathbf{x} - \mathbf{x}_c|^2) \right) d\mathbf{x} \\
 &= |\omega(t)| \mathbf{u}(\mathbf{x}_c, t) + \nabla \mathbf{u}(\mathbf{x}_c, t) \underbrace{\int_{\omega(t)} (\mathbf{x} - \mathbf{x}_c) d\mathbf{x}}_{=0} + \mathcal{O}(h^2)
 \end{aligned}$$

Thus

$$\frac{d}{dt} \mathbf{x}_c = \mathbf{u}(\mathbf{x}_c) + \mathcal{O}(h^2)$$

Flood algorithm on convex cells



Initial condition

- Flood direction \mathbf{n}
- Volume of fluid $V^* = |\Omega|/2$

Algorithm (find ξ^*)

- 1 Project vertices
- 2 Start from the first point (p_4)
- 3 Find the closest neighbor point
- 4 Generate the section
- 5 If $V_{\text{total}} > V^*$ exit
- 6 Repeat 3

Result

- ξ^* given by quadratic interpolation

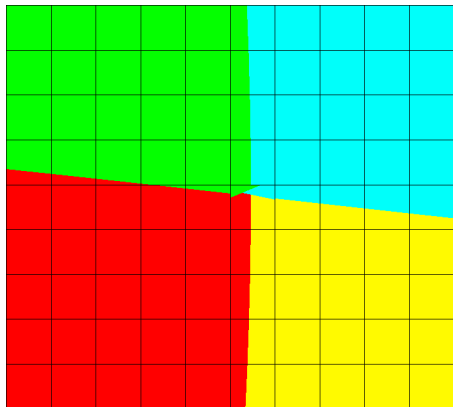
→ Convexity: no need to sort the ξ_n

$$\alpha = \frac{V - V_{\text{tot}}}{V_{\text{trapezoid}}} \quad \beta = \sqrt{\left(\frac{|\Gamma|}{|\Gamma_{\text{next}}| + |\Gamma|}\right)^2 + \alpha \frac{|\Gamma_{\text{next}}| - |\Gamma|}{|\Gamma_{\text{next}}| + |\Gamma|} + \frac{|\Gamma|}{|\Gamma_{\text{next}}| + |\Gamma|}}$$

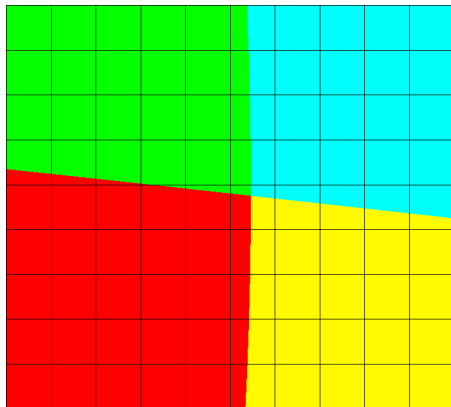
$$\xi^* = \xi + (\xi_{\text{next}} - \xi) \frac{\alpha}{\beta}$$

Reference: Breil, J., Geler, S., & Maire, P. H. (2011).

Multimaterial reconstruction: Remark on B-tree dissection

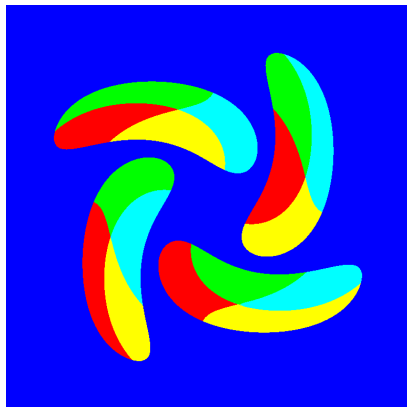
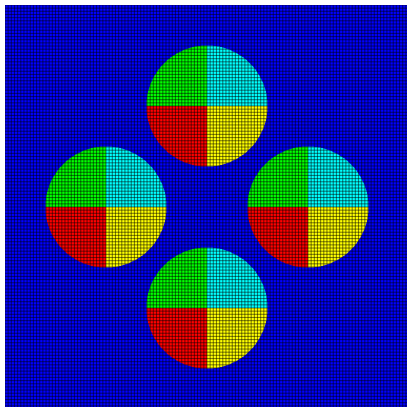


Without B-tree dissection

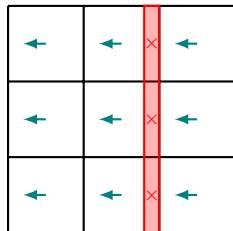


With B-tree dissection

Advection: 5 fluids on a sheared flow



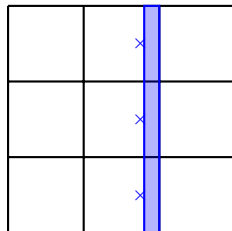
Limitations of MOF: Filaments



Initial configuration

0	0.2x	0
0	0.2x	0
0	0.2x	0

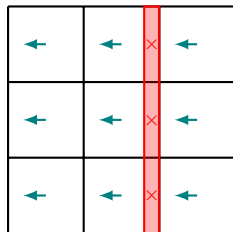
MOF representation after advection



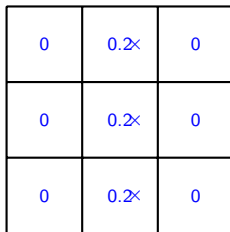
MOF reconstruction

- The filament does not move if the time step is too small!

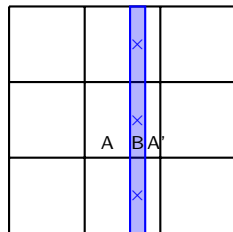
Limitations of MOF: Possible solution



Initial configuration



MOF representation after advection



MOF reconstruction

Virtual fluid A'

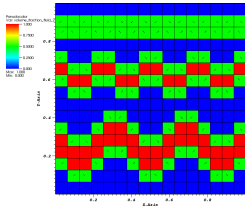
- Detection of filament at the advection step
- Introduction of a virtual fluid A'
- 3-fluid reconstruction

Reference: Jemison, M., Sussman, M., Shashkov, M. (2015)

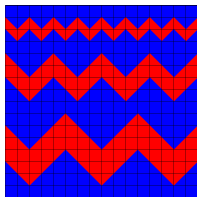
Examples of static reconstructions

Zigzags

Volume fraction

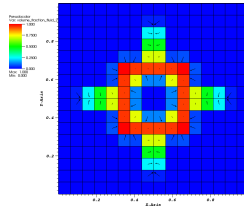


MoF reconstruction

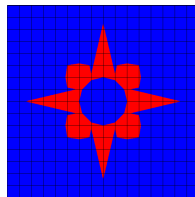


Compass

Volume fraction

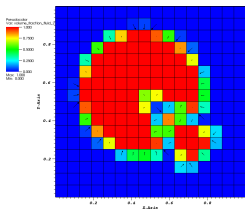


MoF reconstruction

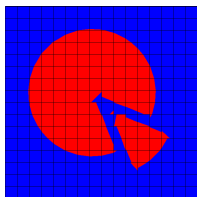


Pie

Volume fraction

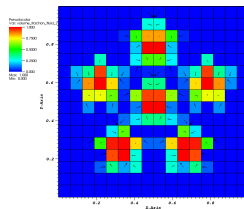


MoF reconstruction

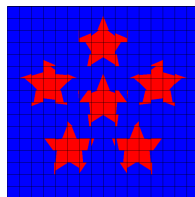


Stars

Volume fraction



MoF reconstruction

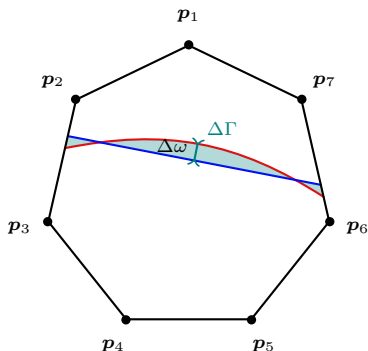


Volume fraction and centroids data from: Dyadechko, V., Shashkov, M. (2007)

Dam break – Air / Water

Mesh 400×200 , domain dimensions $(0.993, 0.5)$

Numerical results: Error computation



Local errors

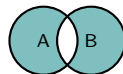
- Distance error

$$\Delta\Gamma = \max_{\mathbf{x}^* \in \Gamma^*} \min_{\mathbf{x} \in \Gamma^\ell} |\mathbf{x} - \mathbf{x}^*|$$

- Area of symmetric difference

$$\Delta\omega = |\omega^\ell \Delta\omega^*|$$

$$A\Delta B = (A \setminus B) \cup (B \setminus A)$$



Global error

- Average deviation (equivalent to $\Delta\Gamma$)

$$\Delta\Gamma_{avg} = \frac{1}{|\partial\omega^*|} \sum_{i=1}^N |\omega_i^\ell \Delta\omega_i^*|$$

Reference: Dyadechko, V., Shashkov, M. (2007)

Parameters

- Iterations: 1000
- Time step: 10^{-4} s
- Mesh: $N \times N$, $N \in \{16, 32, \dots, 4096\}$

Vector field

$$\mathbf{u}(x, y, t) = \begin{bmatrix} -2 \sin^2(\pi x) \sin(\pi y) \cos(\pi y) \\ 2 \sin^2(\pi y) \sin(\pi x) \cos(\pi x) \end{bmatrix} \cos\left(\pi \frac{t}{T}\right)$$

Spatial convergence

N	$\Delta\Gamma_{\text{avg}}$	order
512	$1.34 \cdot 10^{-6}$	2.03
1024	$3.09 \cdot 10^{-7}$	2.11
2048	$7.19 \cdot 10^{-8}$	2.10
4096	$1.60 \cdot 10^{-8}$	2.17

Numerical results: Sheared flow time convergence

Convergence with RK2

Time step	$\Delta\Gamma_{\text{avg}}$	order
$5 \cdot 10^{-4}$	$2.48 \cdot 10^{-7}$	—
$2.5 \cdot 10^{-4}$	$2.86 \cdot 10^{-7}$	-0.21
$1.25 \cdot 10^{-4}$	$3.11 \cdot 10^{-7}$	-0.12

→ Does not converge! (even with a thinner grid)

Convergence with Euler

Time step	$\Delta\Gamma_{\text{avg}}$	order
$5 \cdot 10^{-4}$	$2.93 \cdot 10^{-4}$	—
$2.5 \cdot 10^{-4}$	$1.46 \cdot 10^{-4}$	1.00
$1.25 \cdot 10^{-4}$	$7.32 \cdot 10^{-5}$	1.00

→ 1st order verified with Euler

Parameters

- Total time: 0.5 s
- Mesh: 1024×1024
- Time step: $\{5 \cdot 10^{-4}, \dots, 1.25 \cdot 10^{-4}\}$ s

→ Error RK2 < Error Euler

Conclusion

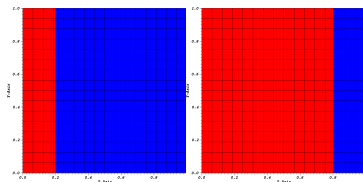
Spatial error dominates

→ Try a case without spatial error

Numerical results: Accelerated front flow time convergence

Convergence with RK2

Time step	$\Delta\Gamma_{\text{avg}}$	order
$1 \cdot 10^{-2}$	$4.93 \cdot 10^{-5}$	—
$5 \cdot 10^{-3}$	$1.23 \cdot 10^{-5}$	2.00
$2.5 \cdot 10^{-3}$	$3.08 \cdot 10^{-6}$	2.00
$1.25 \cdot 10^{-3}$	$7.71 \cdot 10^{-7}$	2.00
$6.25 \cdot 10^{-4}$	$1.93 \cdot 10^{-7}$	2.00



Vector field

$$u_x(x, y, t) = 0.3\pi \sin(\pi t)$$

Conclusion

2nd order with RK2