

Implicit kinetic schemes for fluid models

J. Badwaik, David Coulette, E. Franck, P. Helluy, C. Hillairet,
H. Oberlin, M. Mehrenberger, L. Mendoza L. Navoret

IRMA Strasbourg & Inria TONUS, France

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Outline

- 1 Context and overall approach
 - Kinetic schemes
 - Lattice Boltzmann schemes
- 2 A tour in DG Lattice Boltzmann schemes family
 - A toy model
 - MHD : Dellar' approach
 - General framework
- 3 Implementation of an implicit DG-LBM solver
 - High order implicit DG scheme
 - Getting efficiency
- 4 Conclusion and prospects

Section 1

Context and overall approach

Target problem

Typical problem

Unknowns $w_k(t, \mathbf{x})$, $k = 1, N$ macroscopic fields.
System of (mostly) hyperbolic conservation laws

$$\partial_t w_k + \nabla \cdot \Phi_k(w) = \mathcal{D}_k(w) + \mathcal{S}_k$$

Φ_k : fluxes (nonlinear) - \mathcal{D}_k : parabolic terms (diffusion).
Systems of interest : Euler/Navier Stokes - MHD

Numerical challenges

- explicit schemes
 - CFL conditions :time scale constrained by space grid
 - forces to resolve possibly unwanted fast times scales
- implicit schemes
 - large nonlinear system
 - costly matrix assembly/storage/inversion.

Kinetic schemes

Distribution function $f(t, \mathbf{x}, \mathbf{v})$.

Boltzmann-BGK equation

$$\partial_t f + \nabla \cdot (f \mathbf{v}) = \frac{1}{\tau} (F^{eq}(Mf), \mathbf{v}) - f$$

with

- macroscopic data $m(t, \mathbf{x}) = Mf = \int K(\mathbf{v}) f(t, \mathbf{x}, \mathbf{v}) d^3 v$ obtained by **linear** map from f
- collision vector $K(\mathbf{v}) \in \mathbb{R}^N$.
- $F^{eq}(m)$ equilibrium state
 - $\int K(\mathbf{v}) F^{eq}(m, \mathbf{v}) d^3 v = m, \forall m$ (macro conservation)
 - $\int s(F^{eq}(m)) d^3 v = \max_{Mf=m} \int s(f) d^3 v$, s entropy

Kinetic schemes

In the limit of short relaxation times $\tau \rightarrow 0$

$$\partial_t m + \nabla \cdot \Phi(m) = 0, \quad \text{with } \Phi(m) = \int \mathbf{v} F^{eq}(m, \mathbf{v}) d^3 \mathbf{v}$$

Basic idea

Solving the **split** transport/relaxation kinetic system for small τ provides a natural scheme to approximate the relaxed system.

Interesting Features

- transport stage (T) is **linear**

$$\partial_t f + \nabla \cdot (f \mathbf{v}) = 0$$

- nonlinearities in the relaxation stage (R) are **local** :

$$\partial_t f = (1/\tau)(F^{eq}(Mf) - f)$$

- finite Δt (splitting) or/and τ generate additional diffusive terms.

Lattice Boltzmann schemes

A particular discretization of $f(t, \mathbf{x}, \mathbf{v})$

- v : small set of discrete velocities $\mathbf{v}_i, i = 1, q$
- Finite set of Boltzmann-BGK equations coupled only through relaxation

$$\partial_t f_i + \nabla \cdot \mathbf{v}_i f_i = \frac{1}{\tau} (f_i^{eq} - f_i), \forall i$$

- x : structured cartesian mesh \mathbf{x}_k generated by the velocity set for a given time scale Δt

$$\forall (k, k'), \exists (i, j) \in [1, q] \times \mathbb{Z}, \mathbf{x}_k - \mathbf{x}_{k'} = j \Delta t \mathbf{v}_i$$

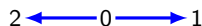
- splitting scheme

- 1 exact transport : $f_i^*(\mathbf{x}) = f_i(t^n, \mathbf{x} - \Delta t \mathbf{v}_i)$
- 2 local relaxation : $f_i(t^n + \Delta t) = (1 - s) f_i^* + s f_i^{eq}(m(f^*))$
relaxation parameter $s = \frac{2\Delta t}{2\tau + \Delta t}$ (Crank-Nicolson)

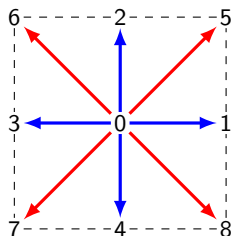
- $\tau \ll \Delta t$ (over-relaxation) fast oscillations around equilibrium manifold

Standard Lattice Boltzmann models

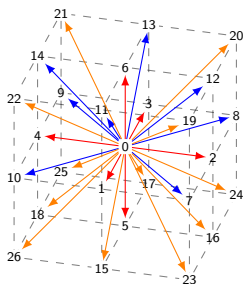
Notation : DdQq with $d = 1, 2, 3$ space dimension q number of velocities.



D1Q3



D2Q9



D3Q27

- K built from low-order polynomials : macro quantities are moments.
- splitting error generates diffusive terms : can mimic physical diffusion
- Can be applied to any hyperbolic system of conservation laws (fluid mechanics [CD98], Maxwell [Gra14], MHD [Del02], etc.)
- transport is easy but $\Delta t/\Delta x$ linked : integer CFL-like condition

Section 2

A tour in DG Lattice Boltzmann schemes family

Simple example D1Q3 (1)

- $v_i = i\lambda, i \in \{-1, 0, 1\}$.
- $K(v) = (1, v)^T \rightarrow w = (\rho, q)$. Density ρ , momentum $q = \rho u$.
- limit system is the Euler isothermal equation

$$\begin{cases} \sum_i [\partial_t f_i + \partial_x (v_i f_i) - \tau^{-1} (f_i^{eq} - f_i)] = 0 \\ \sum_i [\partial_t (f_i v_i) + \partial_x (v_i^2 f_i) - \tau^{-1} (v_i f_i^{eq} - v_i f_i)] = 0 \end{cases} \rightarrow \begin{cases} \partial_t \rho + \partial_x q = 0 \\ \partial_t q + \partial_x (q^2 / \rho + c^2 \rho) = 0 \end{cases}$$

- f^{eq} is an equilibrium (Maxwellian) state if:

$$\rho = \sum_i f_i^{eq} \quad q = \sum_i f_i^{eq} v_i \quad \frac{q^2}{\rho} + \rho c^2 = \sum_i f_i^{eq} v_i^2$$

- solving the linear system for f^{eq} we obtain

$$f_0^{eq} = \rho(\lambda^2 - u^2 - c^2)/\lambda^2 \quad f_{\pm 1}^{eq} = \frac{\rho}{2} (\pm \lambda u + u^2 + c^2) / \lambda^2$$

Simple example D1Q3 (2)

Let's consider the extended moment set (ρ, q, z) , with $z = \sum_i f_i v_i^2$.
In moment space, the D1Q3 model reads

$$\begin{aligned}\partial_t \rho + \partial_x q &= 0, \\ \partial_t q + \partial_x z &= 0, \\ \partial_t z + \lambda^2 \partial_x q &= \tau^{-1}(z^{eq}(\rho, q) - z) = \tau^{-1}(q^2/\rho + c^2 \rho - z).\end{aligned}$$

Chapman-Enskog method shows that for small τ we have

$$\begin{aligned}\partial_t \rho + \partial_x q &= 0, \\ \partial_t q + \partial_x \left(\frac{q^2}{\rho} + c^2 \rho \right) &= \tau \partial_x \left(\underbrace{(\lambda^2 - c^2 - 3u^2)}_{\text{sign!}} \partial_x q + 2u(u^2 - c^2) \partial_x \rho \right).\end{aligned}$$

The viscosity terms are not entropy dissipative \rightarrow small Mach flows.
Different from the Jin-Xin relaxation [JX95, Dub13].

Dellar's approach : D2Q9 + 2 x D2Q5 for 2D MHD

Bibliography [Del02]

$$\text{Basic resistive MHD} \left\{ \begin{array}{l} \partial_t \rho + \nabla \cdot (\rho \mathbf{u}) = 0 \\ \partial_t (\rho \mathbf{u}) + \nabla \cdot [(\rho + B^2/2)\mathbb{I} + \rho \mathbf{u} \mathbf{u} - \mathbf{B} \mathbf{B}] = \nabla \cdot (\mu \mathcal{S}) \\ \partial_t \mathbf{B} + \nabla \times [\eta \nabla \times \mathbf{B} - \mathbf{u} \times \mathbf{B}] = 0 \end{array} \right.$$

LBM modelisation

- fluid part : standard Euler/Navier Stokes ; Lorentz force \rightarrow included in the equilibrium flux
- induction equation : \mathbf{B} cannot be directly cast as a first order moment (due to antisymmetry).
- Dellar's approach
 - associate each component of \mathbf{B} to a separate $DsQq$ model
 - $\alpha = 1, d : B^\alpha = \sum_i g_i^\alpha$.
 - coupling is done through the equilibrium.
 - extension to more complex MHD models

A (too) general approach : $\Pi_k(DdQq_k)$ schemes

Generic Multi-LBM Scheme building

For each $k = 1, N$ consider a $DsQq_k$ model given by

- q_k velocities $v_{k,i}$
- a $q \times q$ invertible square matrix P_k mapping micro/macro variables $m = Pf$
- a subset of $1 \leq c_k \leq q_k - 1$ conserved variables.
- q equilibrium functions f_i^{eq} (or equivalently $m_i^{eq} = [Pf_{eq}]_i$)

In moment space we have

$$\partial_t m_k + \nabla \cdot \left[P_k \text{diag}[v_k] P_k^{-1} m_k \right] = \frac{1}{\epsilon} \sum_{j=1}^N P_k \Omega_{kj} P_j^{-1} (m_j^{eq} - m_j)$$

With Ω the linear relaxation matrix made of $q_k \times q_j$ blocks. Ω may depend on conserved variables.

- limit system yields $C = \sum_k c_k$ equations on the conserved variables.
- diffusive terms are shaped by equilibrium and the structure of Ω .

$(DdQ(d+1))^n$ schemes

Bibliography [Gra14]

Target problem

$$\partial_t w_k + \nabla \cdot \Phi_k(w) = \mathcal{D}_k(w), k = 1, n$$

For each k one conserved quantity and d flux components : $d+1$ scalar fields

$(DdQ(d+1))^n$ scheme building

For each $k = 1, n$ we consider the "same" $DdQ(d+1)$ model given by

- $d+1$ velocities v_i forming a simplex, $\sum_i v_i = 0$
- 1 conserved quantity $m_{k,0} = \sum_i f_{k,i}$. ($w = [m_{1,0}, m_{2,0}, \dots, m_{n,0}]$)
- d non-conserved quantities are the $m_{k,\alpha} = \sum_i f_{k,i} v_i^\alpha, \alpha = 1, d$
- $q = d+1$ equilibrium functions f_i^{eq} obtained by solving $w_k = \sum_i f_{i,k}^{eq}(w)$ and $F_k^\alpha(w) = \sum_i f_{i,k}^{eq}(w) v_i^\alpha$.
- dissipative provided $|v_i| > \sup\{|\lambda|, \lambda \in \text{spec}(d_w \Phi)\}$

Section 3

Implementation of an implicit DG-LBM solver

Back to our problem

Guiding principle

- Replace strongly coupled non-linear hyperbolic system by a (larger) set of more loosely coupled ones.
- in essence : split spatial coupling / inter-variable nonlinear coupling.
- compensate for larger problem size by efficient parallelization.

Requisites

- unstructured meshes to handle complex geometry
- no CFL
- high order in space and time.

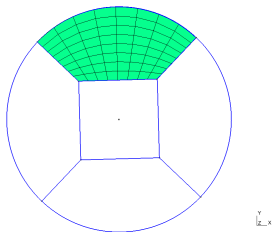
Project guidelines

- unstructured mesh : DG space discretization (h,p refinement, locality)
- no CFL? → implicit schemes
- control diffusive terms : high space and time order + relaxation tweaking.

DG - Implicit upwind transport scheme 1

We consider a coarse mesh made of hexahedral curved macrocells

- Each macrocell is itself split into smaller subcells of size h .
- In each subcell L we consider polynomial basis functions ψ_k^L of degree p .
- Expansion on the polynomial basis: discontinuous approximation of f .



$$f(x, v, p\Delta t) \simeq f_L^p(x, v) = \sum_k f_{L,k}^p(v) \psi_k^L(x), \quad x \in L.$$

DG - Implicit upwind transport scheme 2

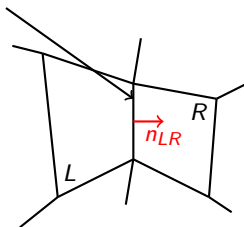
Implicit DG approximation scheme

$\forall L, \forall k$

$$\int_L \frac{f_L^p - f_L^{p-1}}{\Delta t} \psi_k^L - \int_L v \cdot \nabla \psi_k^L f_L^p + \int_{\partial L} (v \cdot n^+ f_L^p + v \cdot n^- f_R^p) \psi_k^L = 0.$$

- time step index: p
- R denotes the neighbor cells along ∂L .
- $v \cdot n^+ = \max(v \cdot n, 0)$,
 $v \cdot n^- = \min(v \cdot n, 0)$.
- n_{LR} is the unit normal vector on ∂L oriented from L to R .

$\partial L \cap \partial R$



Features

- implicit scheme, unconditionally stable, (h, p) refinement
- requires a priori the resolution of a large linear system for each v .

Getting high order in time : symmetric splitting

Bibliography: [MQ02]

Example

- D2Q9 model ,Euler stationary state in a constant gravity field $\mathbf{g} = g\mathbf{e}_y$.
- Analytical solution $\rho = \rho_0 e^{-gy/T}$

Splitting schemes made of **symmetric** building blocks

All steps implemented as θ weighted schemes.

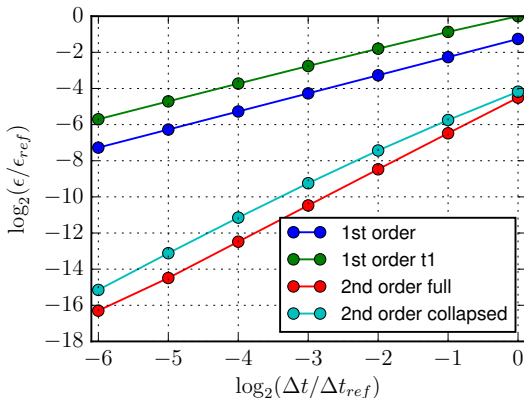
$\theta = 0$ explicit , $\theta = 1$ implicit ; $\theta = 1/2$ (Crank-Nicolson) \rightarrow symmetric

- Transport (T)
- Macroscopic source (S) (gravity)
- BGK Relaxation (R).

1 st order	$T(\Delta t)$	$S(\Delta t)$	$R(\Delta t)$		
2 nd order	$T(\Delta t/2)$	$R(\Delta t/2)$	$S(\Delta t)$	$R(\Delta t/2)$	$T(\Delta t/2)$
2 nd order collapsed		$R(\Delta t/2)$	$S(\Delta t)$	$R(\Delta t/2)$	$T(\Delta t)$

Euler gravity stationary

Δt Convergence of $L2$ error on macroscopic data wrt analytical solution.

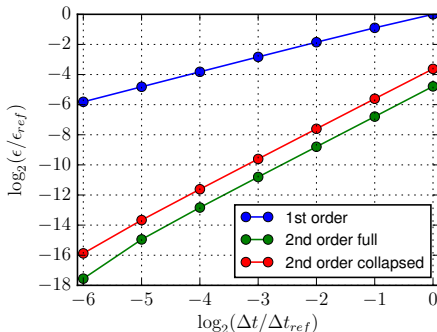


First order t1 : first order splitting with fully implicit $\theta = 1$ blocks.

First order : first order splitting with $\theta = 1/2$ second order symmetric blocks.

Euler Gaussian pulse

- D2Q9 model on a square : 8×8 elements, 10×10 subcells, 3^{rd} order
- Initial condition : narrow gaussian density bump
 $\rho = 1 + 0.1 \exp(-40 * (x^2 + y^2)).$
- Convergence evaluated from highly time-resolved solution.



The benefits of upwinding

upwind flux \rightarrow data dependencies follow the (constant) velocity

- transport operator can be cast into Block Triangular Form (BTF) by appropriate data renumbering.
- inversion : BTF + inversion of diagonal blocks.
- data blocks at the subcell scale : too small for efficient parallelism.

Coarse grain block structure at macrocell level

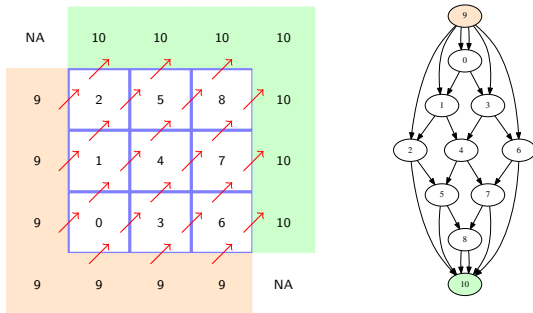
- L is *upwind* with respect to R if $\mathbf{v} \cdot \mathbf{n}_{LR} > 0$ on $\partial L \cap \partial R$.
- In a cell L , the solution depends only on the values of f in the upwind macrocells.



2	5	8
1	4	7
0	3	6

Dependency graph

For a given velocity v we can build a dependency graph. Vertices are associated to macrocells and edges to macrocells interfaces or boundaries. We consider two fictitious additional vertices: the “upwind” vertex and the “downwind” vertex.

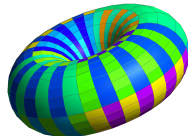


- The dependency graph yields a coarse block triangular ordering
- the local system in each macrocell is solved "on the fly" using the KLU library.
- no need to assemble, store, and factorize the global system !

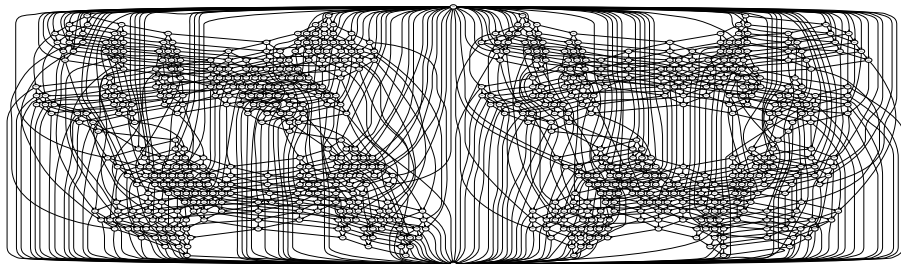
Transport solver parallelism

- ideal across velocities (uncoupled)
- across macrocells : can be high but load imbalance
- realistic mesh : complex to manage...

Toroidal mesh - 720 macrocells



Toroidal mesh - transport graph for $(1, 0, 0)$ velocity.



We need smart task scheduling

Kirsch : Task-Based parallel DG-LBM solver

Here comes StaPU

- StarPU is a task-based scheduling library developed at Inria Bordeaux [AAF⁺12]: <http://starpu.gforge.inria.fr>
- Task description : codelets, inputs (R), outputs (W or RW).
- The user submits tasks in a correct sequential order.
- StarPU schedules the tasks in parallel if possible.
- MPI extension easy : dispatch data and declare owner process : communications handled transparently.

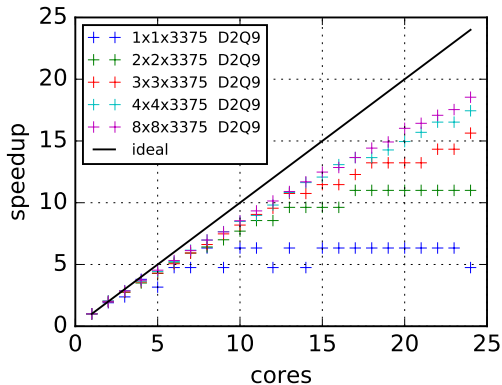
SCHNAPS + StaRPU + LBM = KIRSCH

- starting point SCHNAPS : general DG explicit solver.
- StarPU + Optimization for Kinetic LBM-like schemes
- KIRSCH : Kinetic Representation for SCHnaps

D2Q9 multithread performance

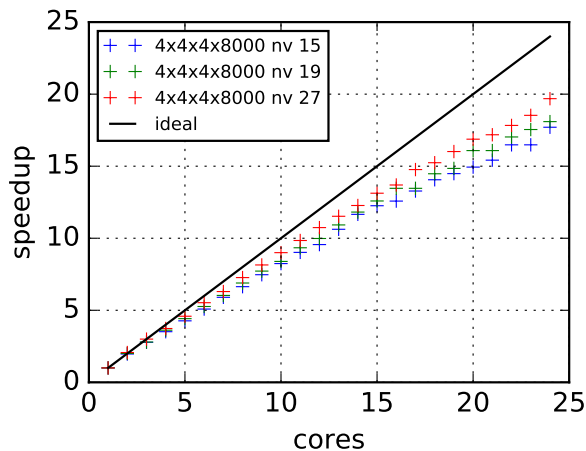
Full D2Q9 scheme on square grids. Constant dof number per macrocell. Number N of macrocells N from 1 to $64 = 8 \times 8$.

- for 1 macrocell : saturation at $n_{core} = n_v$. This is expected.
- efficiency grows with N due to topological parallelism.



D3Q* multithread performance

D3Q15, D3Q19, D3Q27 models on a cube with $4 \times 4 \times 4$ elements and 8000 dof per elements with eager scheduler.

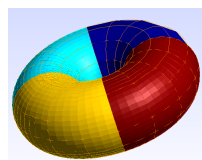
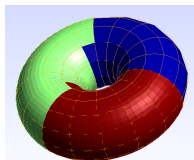
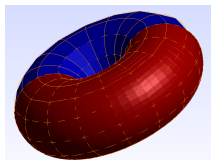
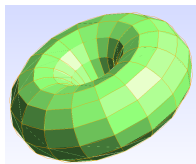


MPI Scaling : D3Q15 in a torus

Toroidal mesh : 720 macroelements \times 3335 dof
2064 interfaces - 192 boundary faces

Wall time in sec for 100 iterations.

Nthreads/Nmpi	1	2	3	4
14	6862	2772	1491	1014



Section 4

Conclusion and prospects

Conclusions and prospects

Current state

- DG-LBM parallel solver
- 2nd order in time
- validation tests on standard 2D LBM-BGK models (Fluids, 2D MHD)
- good MPI/Multithreaded scaling in both 2D and 3D.

Next steps

- $(DqQd + 1)^n$ approach is appealing : stable and generic.
- optimization : Transport Tasks Optimization / GPU codelets
- higher order in time (composition [CCDV09] , complex time steps) :mitigate diffusion
- validate 2D – 3D MHD models.
- benchmark wrt JOREK.

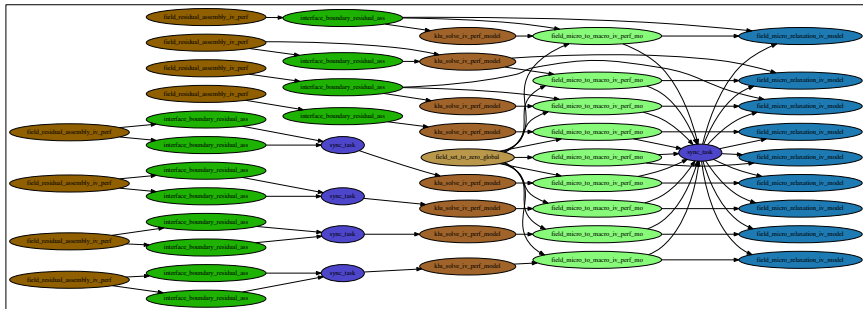
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What StarPU does for us

Task graph for 2DQ9 model

- a single 2D macrocell
- a single time step of the first order scheme (T + R)



Pretty simple...

What StarPU does for us

- 4 macrocells in a 2D square
- a single time step of the first order scheme (T + R)

