

closures-2d

Tim Shaffer¹

August 5, 2015

¹Mentor: C. Kristopher Garrett

We are releasing software that tests angular approximations in kinetic transport simulations

- Based on code used for a previous publication (citation)
- Open source so that others can use/collaborate
- Makes it easy to implement new features
- Implements S_N , P_N , FP_N , M_N , PP_N , D_N (experimental)

[pictures]

Improvements

Algorithmic changes:

- Added experimental Lebedev quadrature
- Removed γ factor for ansatz correction in momopt
- Changed momopt's regularization in case of bad condition number

Software-related improvements:

- Cross-platform build system
- Automated testing
- Better MPI communication
- Improved documentation
- Improved interface
- Bugfixes
- Profiling and optimization

We are releasing this code as open source software.

Now uses SCons, a Python-based build system, rather than Makefiles

- Cross platform
- Sets up library search paths
- Intelligent compilation

Depends on:

- | | |
|----------|-----------------------|
| • GSL | • OpenMP (optional) |
| • BLAS | • Open MPI (optional) |
| • LAPACK | • PAPI (optional) |

Testing

We implemented several automated tests for the software. The testing code is written in Python and integrated with the build system.

The testing code carries out

- Regression tests – compare output to reference data included with the software
- Mass Conservation tests – check that total density remains the same
- Convergence tests – measure the effect of decreasing the cell size on the precision of the output

convergence tests:

- moment(sspline)

-- dx / 16	35.922s	
-- dx / 8	4.580s	
-- dx / 4	0.603s	2.340759
-- dx / 2	0.092s	2.082828
-- dx / 1	0.022s	1.954163

Added timing and profiling-related measurement capabilities

Tested additional optimizations

- Spatial blocking (reduce scheduling overhead, cache misses)
- CPU pinning (ccNUMA)
- Memory alignment

The Importance of Caching

Blocking

What Are Kinetic Equations?

Macroscopic

- $\rho(\mathbf{x}, t)$ – Density
- $\mathbf{u}(\mathbf{x}, t)$ – Velocity
- $E(\mathbf{x}, t)$ – Kinetic Energy

Discretize \mathbf{x} , t into 100 values:
4GB memory requirement

Mesoscopic

- $f(\mathbf{x}, \mathbf{v}, t)$ – Density with respect to space *and* velocity

Discretize \mathbf{x} , \mathbf{v} , t into 100 pieces:
800TB memory requirement

Macroscopic can be derived from mesoscopic

- $\rho(\mathbf{x}, t) = \int_{\mathbb{R}^3} f \, d\mathbf{v}$
- $\mathbf{u}(\mathbf{x}, t) = \frac{1}{\rho} \int_{\mathbb{R}^3} \mathbf{v} f \, d\mathbf{v}$
- $E(\mathbf{x}, t) = \frac{1}{2} \int_{\mathbb{R}^3} \|\mathbf{v} - \mathbf{u}\|^2 \, d\mathbf{v}$

What Are Kinetic Equations?

First used for rarefied gas dynamics (e.g. high altitude gases where collisions do not dominate the physics)

$$\partial_t f + \mathbf{v} \cdot \nabla_{\mathbf{x}} = C(f)$$

where $\int C(f) d\mathbf{v} = 0$.

Integrate against \mathbf{v} to get First Euler/Navier-Stokes equation

$$\partial_t \rho + \nabla_{\mathbf{x}} \cdot (\rho \mathbf{u}) = 0$$

Other areas:

- Radiation transport
- Plasma simulations

Unit Speed, Isotropic Scattering

$$\partial_t f + \Omega \cdot \nabla_{\mathbf{x}} = \frac{\sigma}{4\pi} \langle f \rangle - \sigma f$$

where $x \in \mathbb{R}^3$, $\Omega \in S^2$, σ is the scattering cross section, and $\langle \cdot \rangle = \int_{S^2} \cdot d\Omega$.

kinetic

- Implements S_N
- Easy to compute
- Permits negative densities
- Suffers from ray effects at low order

[picture]

moment

- Implements P_N
- Somewhat easy to compute
- Permits negative densities
- Suffers from oscillatory artifacts
- Filters can improve performance

[picture]

momopt

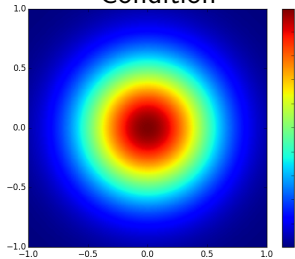
- Implements M_N and PP_N
- Difficult to compute
- Ensures positivity
- Requires additional optimization procedure

[picture]

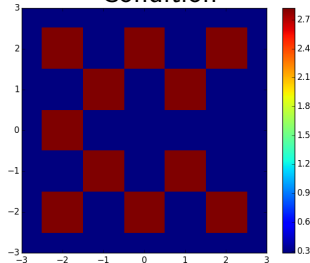
[picture]

Initial Conditions

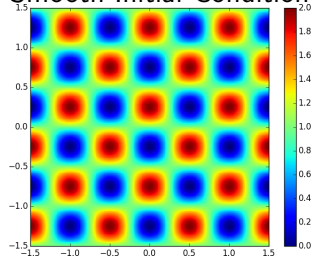
Gaussian Initial Condition



σ_T for Lattice Initial Condition



Smooth Initial Condition



Time Stepping

CFL Condition

A small enough time step is necessary for convergence. In this case,

$$\frac{\Delta t}{\Delta x} + \frac{\Delta t}{\Delta y} \leq 1$$

Heun's Method

To compute a first estimate, carry out two Euler steps. Now use the average of the initial state and the estimate.

- Explicit, two-stage Runge-Kutta method
- Second order accurate
- Strong stability preserving

Currently uses Chebyshev-Legendre quadrature.

- Constructed from n Gauss-Legendre rule
- Optimized for symmetry
- Exactly integrates to order $2n - 1$ moments

Added (experimental) Lebedev quadrature

- Constructed based on octahedral symmetry group
- Optimal with respect to number of points
- Structure does not lend itself to symmetry optimizations
- Negative weights

Optimization Problem

`momopt` uses nonlinear spectral methods, so updates entail solving an optimization problem for a given \mathbf{u} and moments \mathbf{m} .

$$\min_{\alpha} \left\langle \exp(\alpha^T \mathbf{m}) \right\rangle - \alpha^T \mathbf{u}$$

To use a Newton solver, we need

- An objective function $F(\alpha) = \left\langle \exp(\alpha^T \mathbf{m}) \right\rangle - \alpha^T \mathbf{u}$
- Gradient $g(\alpha)$ of $F(\alpha)$
- Hessian $H(\alpha)$ of $F(\alpha)$

Now the estimated $\alpha' = \alpha + t\mathbf{d}$ where $\mathbf{d} = -H(\alpha)^{-1}g(\alpha)$.

momopt computes ansatz for the optimization, yielding an *approximate* solution $\mathbf{u} \approx \langle \exp(\alpha^\top) \rangle$.

Needed to estimate error and try to correct via γ ratio.

Required fine mesh and low tolerance on optimization to prevent non-realizability.

Instead, compute $\hat{\mathbf{u}}$ such that $\hat{\mathbf{u}} = \langle \exp(\alpha^\top) \rangle$.

Fixed regularization

In case of bad condition number (i.e. the system is too sensitive to errors) fall back to isotropic α to allow the algorithm to complete successfully

The source code is available on Github

<https://github.com/ckrisgarrett/closures-2d>

cite that presentation, linesource paper