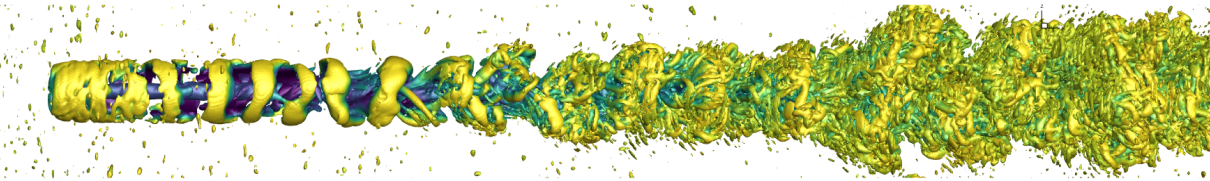


Controlling inflow boundary conditions in a lattice Boltzmann LES model for urban flows



Geir Evensen



Evaluate LBM as a scalable LES framework for real-world flows.

- Equinor project: Assess LBM for wind turbine wake simulations.
- UrbanAIR EU project: Explore data assimilation for LES.
- Open-source LBM implementation: <https://www.github.com/geirev/LBM>.

A gas with constant temperature and no external forcing equilibrates to the

Maxwell-Boltzmann equilibrium distribution

$$f(\mathbf{w}) \rightarrow f^{\text{eq}}(\mathbf{w}) = n \left(\frac{m}{2\pi k T} \right)^{3/2} \exp \left(-\frac{m (\mathbf{w} - \mathbf{v})^2}{2 k T} \right) \quad (1)$$

- Here $f(\mathbf{w})$ is the probability density of the particle velocities.
- Macroscopic variables are moments of $f(\mathbf{w})$.
- In the absence of forcing, the solution relaxes toward this state.
- LBM evolves discrete distributions that are close to this equilibrium.
- Average particle velocity in air is 470 m/s.

The probability density $f(\mathbf{w})$ evolves according to

Boltzmann equation (BGK approximation)

$$\frac{\partial f}{\partial t} + \mathbf{w} \cdot \frac{\partial f}{\partial \mathbf{r}} + \frac{\mathbf{X}}{m} \cdot \frac{\partial f}{\partial \mathbf{w}} = \tau^{-1}(f^{\text{eq}} - f) \quad (2)$$

- Collision term models relaxation toward equilibrium.
- Valid in the hydrodynamic limit (mean free path $\ll$ system size).
- Mean free path in air: $6.8 \cdot 10^{-8}$ m.
- Moments of this equation recover the Navier–Stokes equations.
- LBM solves a discrete, low-Mach approximation of this equation.

3'rd Order Hermite expansion



Represent $f(\mathbf{r}, \mathbf{w}, t)$ using a 3rd-order Hermite expansion in velocity

$$f(\mathbf{w}, \mathbf{r}, t) = e^{-\mathbf{w}^T \mathbf{w} / 2} \left[a^{(0)}(\mathbf{r}, t) + a_i^{(1)}(\mathbf{r}, t) H_i^{(1)}(\mathbf{w}) + \frac{1}{2} a_{ij}^{(2)}(\mathbf{r}, t) H_{ij}^{(2)}(\mathbf{w}) + \frac{1}{6} a_{ijk}^{(3)}(\mathbf{r}, t) H_{ijk}^{(3)}(\mathbf{w}) \right] \quad (3)$$

$$= \omega(\mathbf{w}) p(\mathbf{w}, \mathbf{r}, t). \quad (4)$$

- Second-order truncation captures the macroscopic NS variables and stress tensor.
- **Third-order term** is needed for stability in high-Reynolds-number flows.

Macroscopic variables by Gauss–Hermite quadrature NORCE

Velocity moments become discrete sums

$$\int \phi(\mathbf{w}) f(\mathbf{r}, \mathbf{w}, t) d\mathbf{w} \approx \int \phi(\mathbf{w}) \omega(\mathbf{w}) p(\mathbf{r}, \mathbf{w}, t) d\mathbf{w} \quad (5)$$

$$= \sum_{l=1}^Q w_l \phi(\mathbf{c}_l) p(\mathbf{r}, \mathbf{c}_l, t) = \sum_{l=1}^Q \phi(\mathbf{c}_l) f_l(\mathbf{r}, t) \quad (6)$$

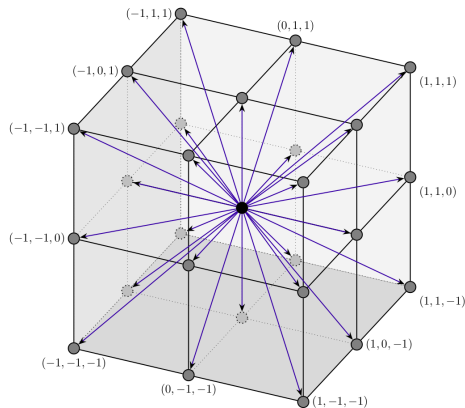
when defining discrete populations

$$f_l(\mathbf{r}, t) = w_l p(\mathbf{r}, \mathbf{c}_l, t) = w_l \frac{f(\mathbf{r}, \mathbf{c}_l, t)}{\omega(\mathbf{c}_l)} \quad (7)$$

- Velocity moments (up to quadrature order) are evaluated exactly using discrete velocities $\mathbf{c}_l$.

LBM discretization in 3D (D3Q27)

- Discrete velocities c_l arise from Gauss–Hermite quadrature.
- At each grid point (i, j, k) , we store populations $f_l(i, j, k)$ for $l = 1, \dots, 27$.
- f_l represents the population moving with velocity c_l .
- Choice of lattice units: $\Delta x = \Delta t = 1$
 - $c_l \in \{-1, 0, 1\}$
 - Exact streaming to neighboring nodes in one timestep
- Advection is exact in lattice units!
- D3Q27 better than D3Q19 for high-Re flows



Density

$$\rho(\mathbf{r}, t) = \sum_l f_l(\mathbf{r}, t) \quad (8)$$

Fluid velocity

$$\mathbf{v}(\mathbf{r}, t) = \frac{1}{\rho} \sum_l \mathbf{c}_l f_l(\mathbf{r}, t) \quad (9)$$

Isothermal equation of state

$$p(\mathbf{r}, t) = c_s^2 \rho(\mathbf{r}, t), \quad c_s^2 = \frac{1}{3} \quad (10)$$

Decompose the distribution:

$$f_l = f_l^{\text{eq}} + f_l^{\text{neq}}. \quad (11)$$

- Collision (BGK relaxation)

$$f_l^* = f_l - \frac{1}{\tau}(f_l - f_l^{\text{eq}}) + F_l \quad (12)$$

$$= f_l^{\text{eq}} + \left(1 - \frac{1}{\tau}\right) f_l^{\text{neq}} + F_l. \quad (13)$$

- Regularization

$$f_l^{\text{neq}} \longrightarrow \mathcal{R}(f_l^{\text{neq}}) \quad (14)$$

removes high-order (non-hydrodynamic) components that can lead to instabilities

- Exact advection of populations

$$f_l(t + \Delta t, \mathbf{x}) = f_l^*(t, \mathbf{x} - \mathbf{c}_l \Delta t) \quad (15)$$

- Due to $\Delta x = \Delta t = 1$:
 - particles move exactly to neighboring nodes
 - no interpolation or numerical diffusion

- Numerical method
 - D3Q27 velocity discretization
 - 3rd-order regularized BGK collision operator ([Jacob et al., 2018](#))
 - Forcing scheme ([Kupershtokh et al., 2009](#))
- Turbulence modeling
 - Vreman's Smagorinsky-type SGS model ([Vreman, 2004](#))
- Boundary conditions
 - Closed, periodic, and inflow–outflow boundaries
 - Bounce-back scheme for solid objects
 - Turbulent inflow via smooth pseudo-random perturbations
- Wind-turbine forcing
 - Uses actuator-line method (ALM) by [Sørensen and Shen \(2002\)](#).
- Implementation
 - NVIDIA CUDA Fortran with GPU + MPI parallelization

CLBM

NS

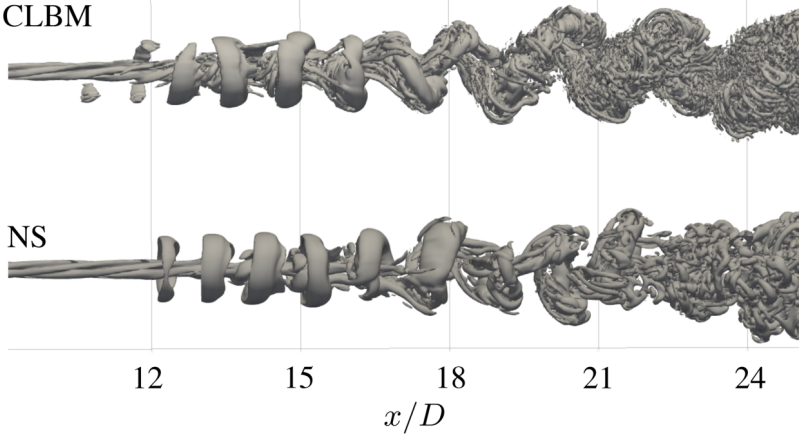
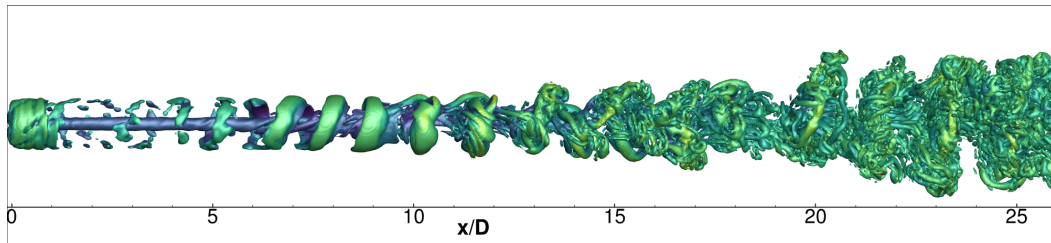


Figure 11. Rendering of the instantaneous contours of the Q criterion ($Q = 0.0005$) in the far wake with the CLBM and NS with $\Delta x = D/32$.

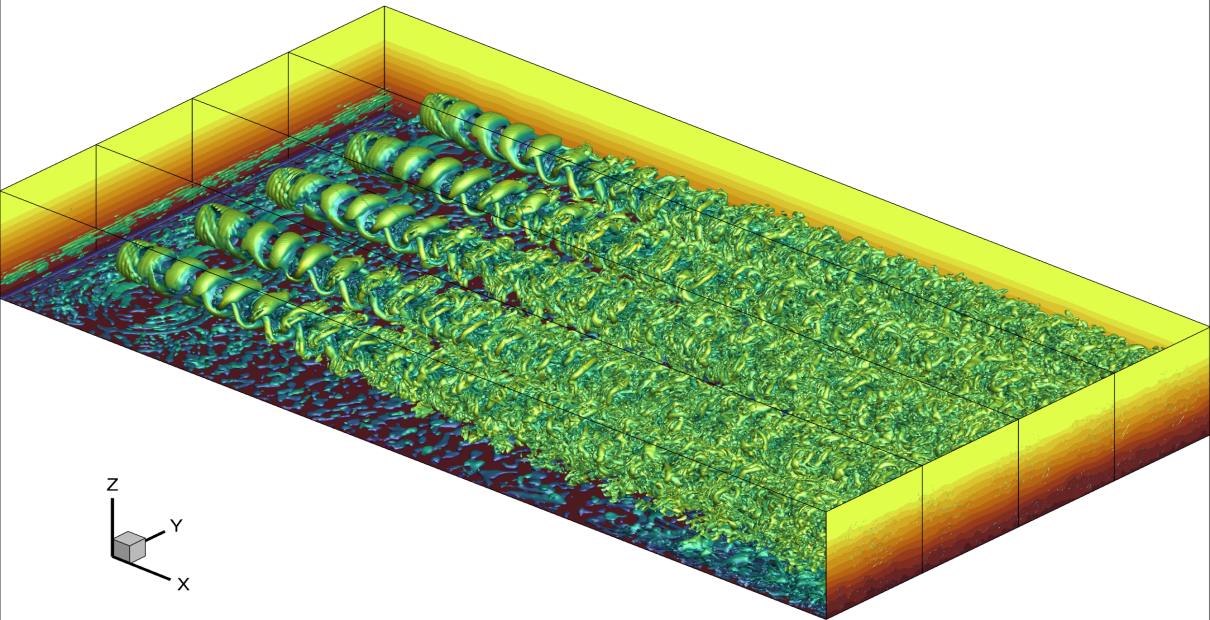
Q-criterion plot

- Rich small-scale structures in the wake
- Low numerical dissipation preserves vortical structures
- Exact streaming avoids artificial diffusion
- D3Q27 + regularization improves isotropy and stability

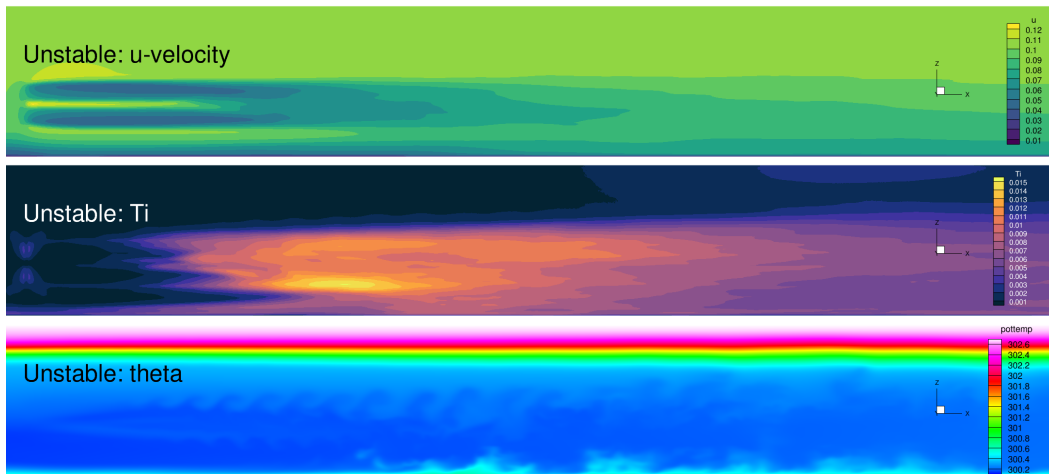


von Kármán vortex street





Unstable atmospheric boundary layer



UrbanAIR: Motivation

- High fidelity city-LES (4 m res)
- Model wind and temperature
- Uncertain inflow conditions from meteo (1 km res)
- **Improve inflow conditions by data assimilation**
- Observations are wind measurements in the city



Flow through a city



Flow through a city



Time varying inflow



- Estimate **uncertain inflow wind speed and direction**.
- Interior observations collected at **four locations within the city**.
- Measurements available every **60 s** with **2% relative observation error**.
- Recursive data assimilation with **5-, 10-, or 20-minute assimilation windows**.
- **4-step ESMDA** for updating the inflow forcing.
- Prior inflow from **AR(2) simulations continued from the previous window**.

Generate a **continuous and smooth prior ensemble** for the next DA window.

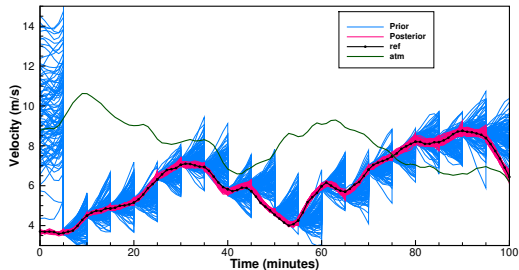
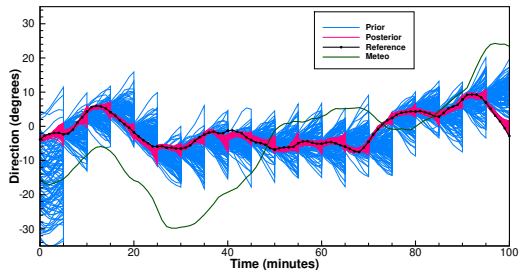
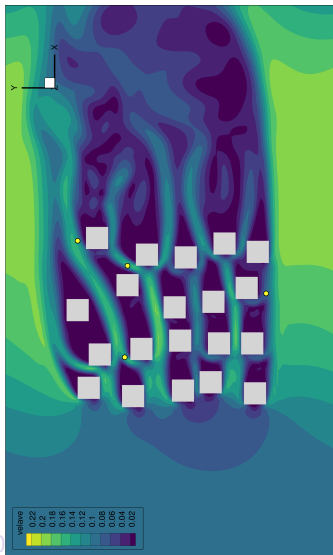
The prior ensemble for the next assimilation window is

$$X_j(t_k) = (1 - \alpha_k) \text{meteo}(t_k) + \alpha_k \bar{X} + \sigma z_j(t_k), \quad (16)$$

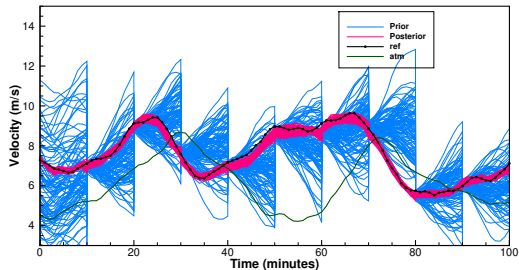
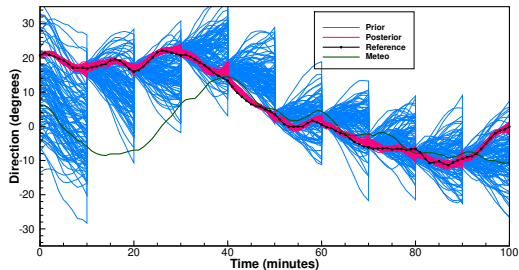
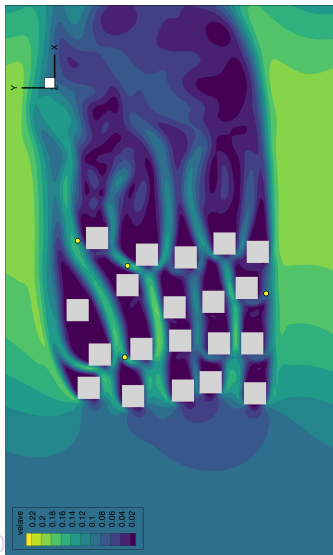
with a relaxation towards the atmospheric prediction determined by

$$\alpha_k = \exp \left[- \left(\frac{k \Delta t}{l_{\text{corr}}} \right)^2 \right]. \quad (17)$$

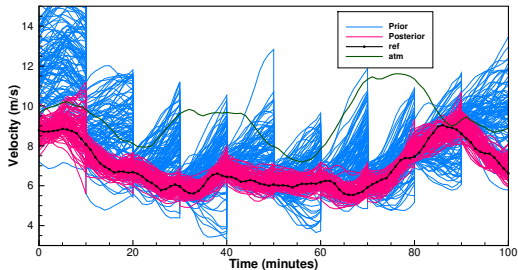
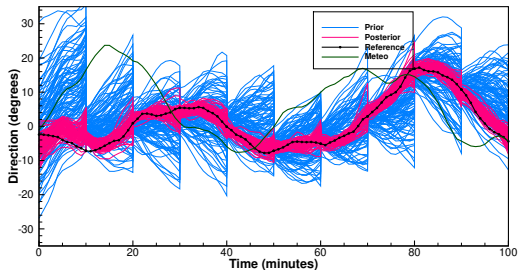
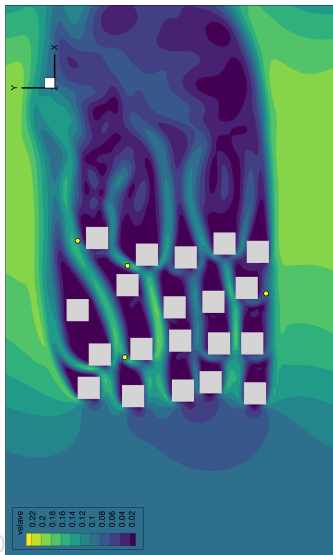
Estimating inflow conditions, ESMDA-4



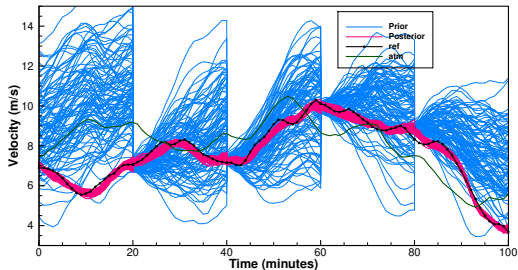
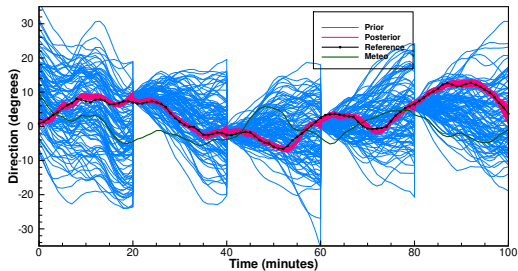
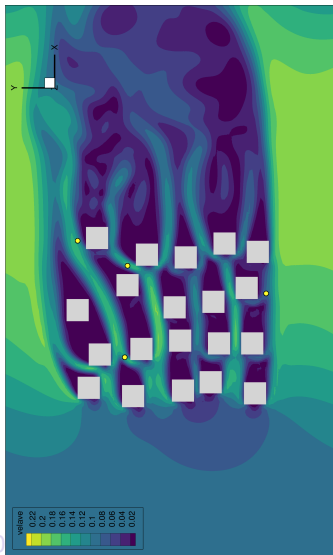
Estimating inflow conditions, ESMDA-4



Estimating inflow conditions, ES



Estimating inflow conditions, ESMDA-4



- **LBM is a competitive framework for high-Reynolds-number LES:**
 - D3Q27 lattice and 3rd-order Hermite expansion improve stability for turbulent flows.
 - Regularization removes non-hydrodynamic higher-order components.
 - Stability with BGK simplicity (Jacob et al., 2018).
 - Explicit advection and local operations are highly GPU-friendly.
- **Open-source LES framework:**
 - Model buoyancy forcing by advecting potential temperature.
 - Vreman subgrid-scale model (Vreman, 2004).
 - Efficient GPU implementation (CUDA Fortran + MPI).
 - Validated for wind-turbine wake simulations.
 - Open-source code: <https://www.github.com/geirev/LBM>.
- **Practical framework for continuous updating of urban flows:**
 - We can continuously update uncertain inflow conditions using internal wind observations.
 - Next step: realistic urban-scale applications.

LATTICE BOLTZMANN METHOD: DERIVATION & PROPERTIES

From the Boltzmann equation to a discrete, efficient and robust fluid solver



1. ORIGIN: FROM STATISTICAL MECHANICS TO BOLTZMANN

Statistical mechanics

A dilute gas of many particles is described by the s-particle distribution function $F_s(\mathbf{r}_1, \mathbf{v}_1, \dots, \mathbf{r}_s, \mathbf{v}_s, t)$.

Liouville's theorem

Conservation of probability in 6Ns-dimensional phase space:

$$\frac{dF_s}{dt} = 0$$

BBGKY hierarchy

Leads to an infinite hierarchy of equations for reduced distributions $f^{(n)}$.

Boltzmann equation (assumptions)

Molecular chaos (Stosszahlansatz) and binary, short-range collisions $\Rightarrow$ closed equation for the one-particle distribution $f(\mathbf{r}, \mathbf{v}, t)$:

$$\frac{\partial f}{\partial t} + \mathbf{v} \cdot \nabla_{\mathbf{r}} f + \mathbf{F} \cdot \nabla_{\mathbf{v}} f = \Omega(f)$$

f : particle distribution function
 $\mathbf{r}$: position, $\mathbf{v}$: velocity, t : time
 $\mathbf{F}$: external force (e.g., gravity)

2. HERMITE EXPANSION

Factorize the distribution

$$f(\mathbf{r}, \mathbf{v}, t) = \omega(\mathbf{v}) p(\mathbf{r}, \mathbf{v}, t)$$

$$\omega(\mathbf{v}) = \exp\left(-\frac{\mathbf{v}^T \mathbf{v}}{2c_s^2}\right)$$

$\omega(\mathbf{v})$: Gaussian weight
(c_s : lattice sound speed)

$p(\mathbf{r}, \mathbf{v}, t)$: polynomial in $\mathbf{v}$ (Hermite expansion)

$$p(\mathbf{r}, \mathbf{v}, t) = \sum_{n=0}^N \frac{1}{n!} a^{(n)}(\mathbf{r}, t) : \mathbf{H}^{(n)}(\mathbf{v})$$

$\mathbf{H}^{(n)}$: n-th order Hermite tensors

Truncate at order N
(e.g., $N = 2$ for NS, $N = 3$ for improved stability)

3. GAUSS-HERMITE QUADRATURE

Discretize velocity space

$$\int_{RD} \omega(\mathbf{v}) q(\mathbf{v}) d\mathbf{v} \approx \sum_{l=1}^Q w_l q(\mathbf{c}_l)$$

($\mathbf{c}_l, w_l$): abscissas (discrete velocities) and weights of Gauss-Hermite quadrature

Exact for all polynomials $q(\mathbf{v})$ up to degree $2M - 1$
(M : order of the quadrature)

4. DISCRETE POPULATIONS

Define f_l on a lattice

$$f_l(\mathbf{r}, t) = w_l p(\mathbf{r}, \mathbf{c}_l, t) = w_l \frac{f(\mathbf{r}, \mathbf{c}_l, t)}{\omega(\mathbf{c}_l)}$$

$\mathbf{c}_l$: discrete velocities (stencil)

w_l : quadrature weights

Q : number of discrete velocities

5. MOMENTS AS DISCRETE SUMS

Macroscopic fields

$$\int \phi(\mathbf{v}) f(\mathbf{r}, \mathbf{v}, t) d\mathbf{v} \approx \sum_{l=1}^Q \phi(\mathbf{c}_l) f_l(\mathbf{r}, t)$$

Examples:

- $\rho = \sum_l f_l$ (density)
- $\rho \mathbf{u} = \sum_l f_l \mathbf{c}_l$ (momentum)
- $\Pi = \sum_l f_l \mathbf{c}_l \mathbf{c}_l$ (stress tensor)
- (... higher moments similarly)

KEY PROPERTIES

RECOVERS NAVIER-STOKES



Second-order truncation is sufficient to recover the Navier-Stokes equations. Third-order improves stability at high Reynolds numbers.

MOMENT CONSERVATION



Mass and momentum are exactly conserved by the collision operator. Higher-order moments are accurate up to the quadrature order.

ISOTROPY



Discrete velocities and weights are chosen to preserve isotropy of tensors up to a given order, ensuring correct macroscopic behavior.

SIMPLE & LOCAL



Streaming and collision are local operators. No global solves. Ideal for parallel computing.

NATURAL BOUNDARIES



Bounce-back and other schemes are simple to implement and robust.

ALGORITHM



Two-step update:
1) Collision (relaxation)
2) Streaming (advection)

TYPICAL LBM MODEL (BGK) WITH REGULARIZATION

Collision (relaxation) with regularization:

$$f_l^*(\mathbf{r}, t) = f_l(\mathbf{r}, t) - \frac{1}{\tau} [f_l(\mathbf{r}, t) - f_l^{eq}(\mathbf{r}, t)]$$

Regularize the non-equilibrium part:

$$f_l^{nug} = f_l - f_l^{eq} \Rightarrow f_l^{nug, reg} = \mathcal{R}(f_l^{nug})$$

$\mathcal{R}(\cdot)$: projection onto the space spanned by Hermite polynomials up to order N ; removes components not represented by the truncated expansion.

Streaming (advection):

$$f_l(\mathbf{r} + \mathbf{c}_l \Delta t, t + \Delta t) = f_l^*(\mathbf{r}, t)$$

Macroscopic relations (from moments):

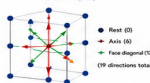
- $\rho = \sum_l f_l$
- $\rho \mathbf{u} = \sum_l f_l \mathbf{c}_l$
- $\Pi = \sum_l f_l \mathbf{c}_l \mathbf{c}_l$

(... higher moments similarly)

$$c_s^2 = \frac{1}{D} \quad (\text{lattice sound speed})$$
$$\nu = c_s^2 \left(\tau - \frac{1}{2} \right) \Delta t \quad (\text{kinematic viscosity})$$

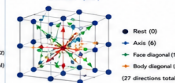
DISCRETE VELOCITY SETS (EXAMPLES)

D3Q19 (3D)



$Q = 19$ directions
2nd-order isotropy

D3Q27 (3D)



$Q = 27$ directions
3rd-order isotropy

KEY TAKEAWAYS

- ✓ LBM follows rigorously from the Boltzmann equation.
- ✓ Hermite expansion + Gauss-Hermite quadrature enable exact moments up to a chosen order.
- ✓ Second order $\Rightarrow$ Navier-Stokes level.
- ✓ Third order $\Rightarrow$ improved stability.
- ✓ Regularization of f_l^{nug} suppresses non-hydrodynamic modes and enhances stability.
- ✓ Simple, local, parallel, and naturally suited for complex geometries.



PHYSICALLY CONSISTENT



NUMERICALLY EFFICIENT



HIGHLY PARALLEL



ROBUST & VERSATILE

Thank you!

Asmuth, H., H. Olivares-Espinosa, and S. Ivanell. Actuator line simulations of wind turbine wakes using the lattice boltzmann method. *Wind Energy Science*, 5(2):623–645, 2020. doi:[10.5194/wes-5-623-2020](https://doi.org/10.5194/wes-5-623-2020).

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