



# Lattice Boltzmann Method

Presented by:  
Keith Doyle  
Liwei Li  
Chaoyi Wang  
Cooper Elsworth



# Introduction & History

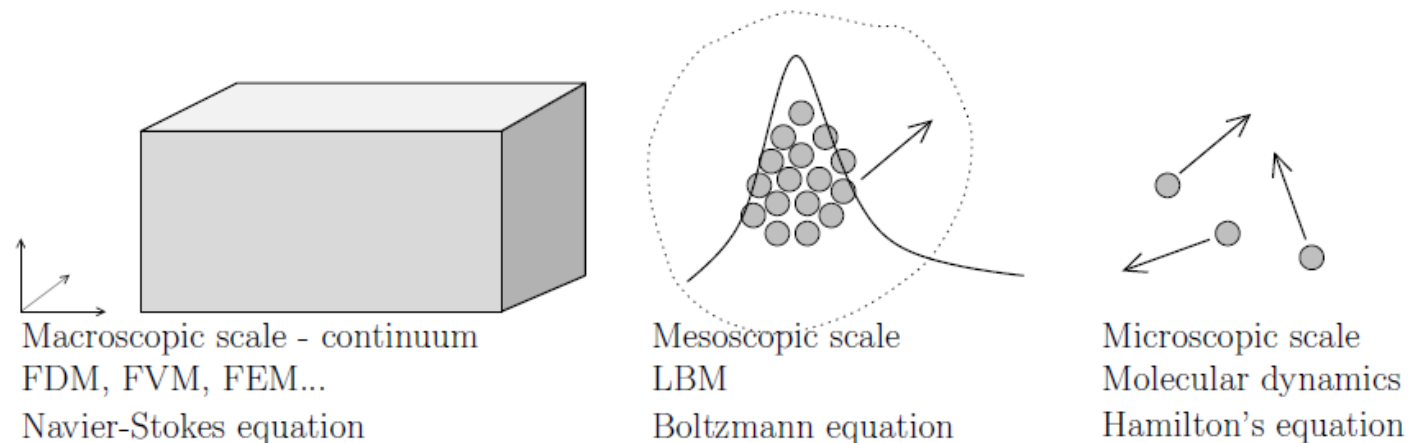
Let's get to know LBM

# Introduction

- Two extreme scales for modeling fluid flow (Mele 2013)
  - Macro-Scale
    - Uses PDE Equations such as Navier Stokes equation
    - Normally solved numerically using FDM, FEM, or FVM
  - Micro-Scale
    - Models individual molecules
    - Behavior governed by Hamilton's equation
    - There are too many molecules to practically model virtually anything useful

# Introduction

- LBM splits the gap between these two scales
  - Considers a collection of molecules as a unit
  - Able to accurately model macro-scale behavior by considering average behavior of these collections of molecules
  - Behavior governed by Boltzmann equation

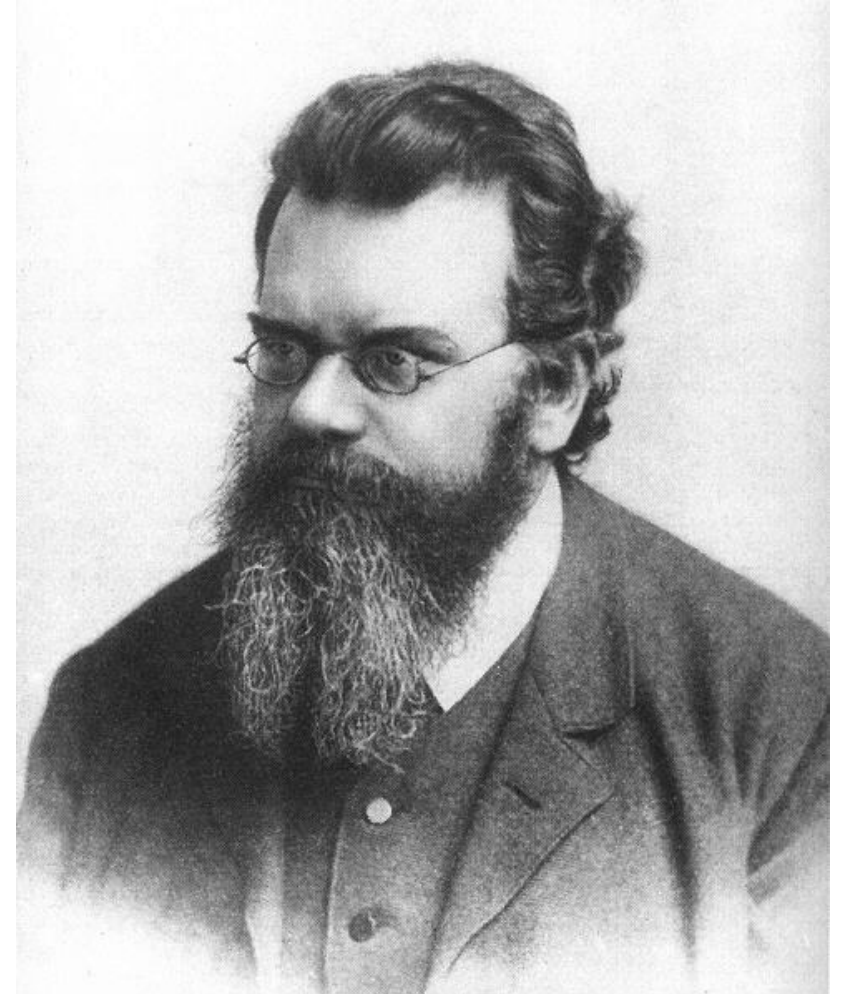


# Introduction

- Advantages
  - LBM solved locally so it is easy to break the problem into calculations that can be done in parallel by multiple computer processors (Mele 2013).
  - Meshing is quasi-instantaneous and computationally simple
- Disadvantages
  - Difficult to simulate scenarios with a high Mach numbers
  - Thermo-hydrodynamic scheme is absent

# Historical Perspective

- Boltzmann Equation (1800's)
  - Developed by Ludwig Boltzmann
  - Describes the dynamics of an ideal gas
  - The Lattice Boltzmann Equation, which governs behavior in the LBM, is a discretized form of the Boltzmann Equation



# Historical Perspective

- Lattice Gas Automata
  - Precursor to LBM
  - Developed by Hardy, Pomeau, and de Pazzis in the 1970's
  - Initially was widely praised as a revolutionary technique.
  - Featured on front page of Washington Post on November 19, 1985
  - Problems with LGA led to the need for the development of LBM

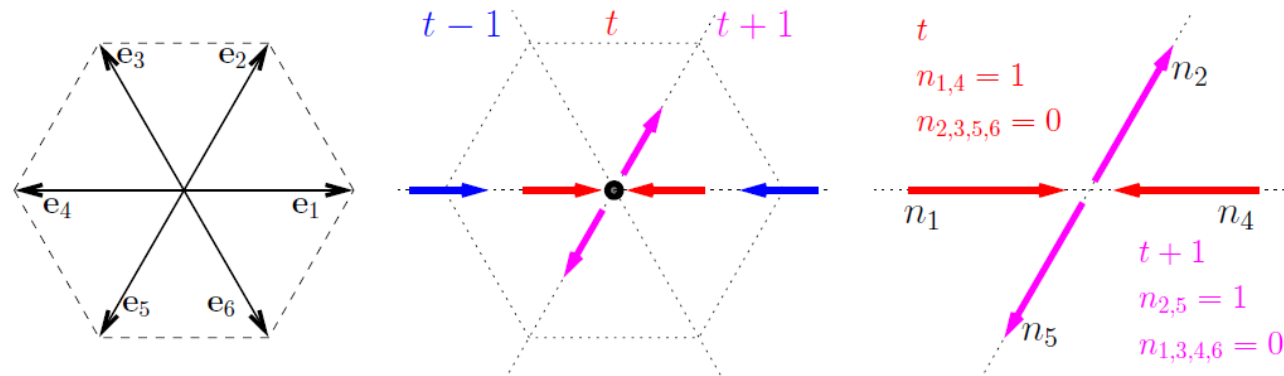


Image : Mele, 2013

# Historical Perspective

- Lattice Gas Automata
  - Disadvantages
    - Statistical noise
    - Needs to simulate a large number of particles in order to reach an acceptable solution
    - Computationally inefficient due to its discrete state calculations

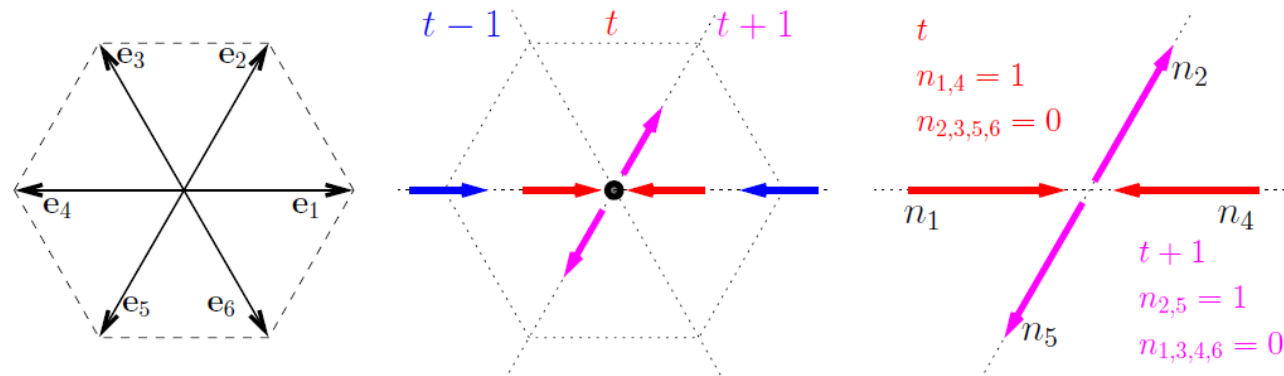


Image : Mele, 2013



# Historical Perspective

- Lattice Boltzmann Method
  - Developed incrementally in the 1980's
  - Overcomes statistical noise associated with LGA by replacing boolean particle occupation variables with single particle distribution functions
  - Distribution functions are an averaged quantity, so there is no need to average the state of a large quantity of cells to define macroscopic behavior

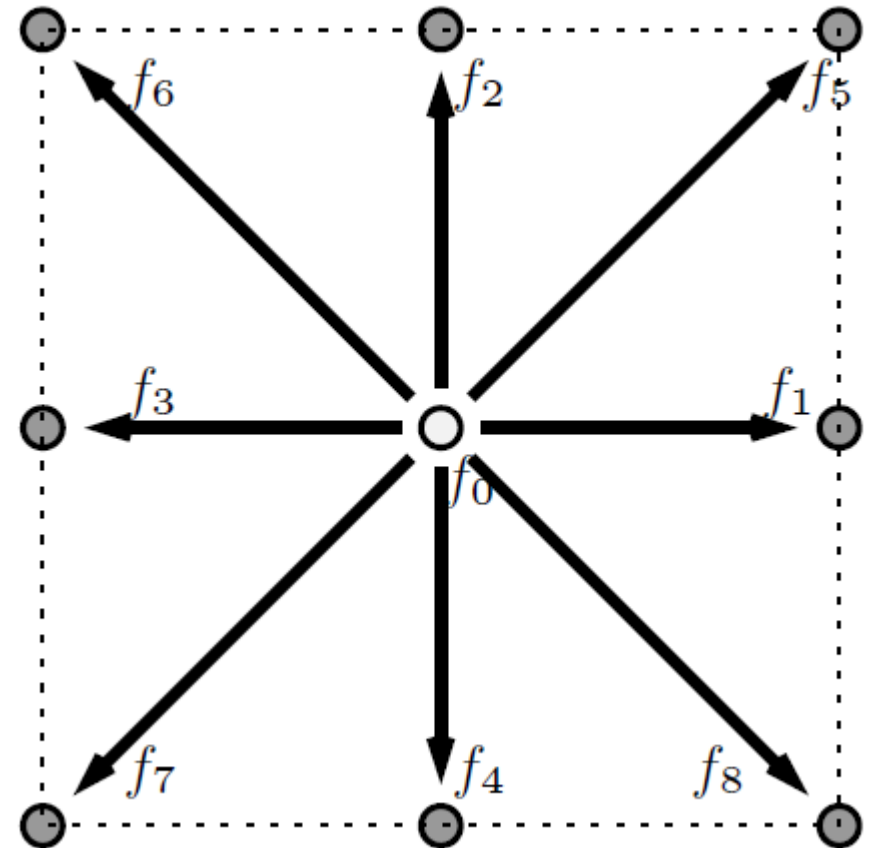


Image : Mele, 2013



# General Principle

Essence of Lattice Boltzmann Method

# General Principles

- LGA (Lattice Gas Automata)
- Lattice Boltzmann Equation
- Collision and Streaming Stages
- LBM vs. CFD (Computational Fluid Dynamics)
- Validation of LBM

# Lattice Gas Automata (LGA)

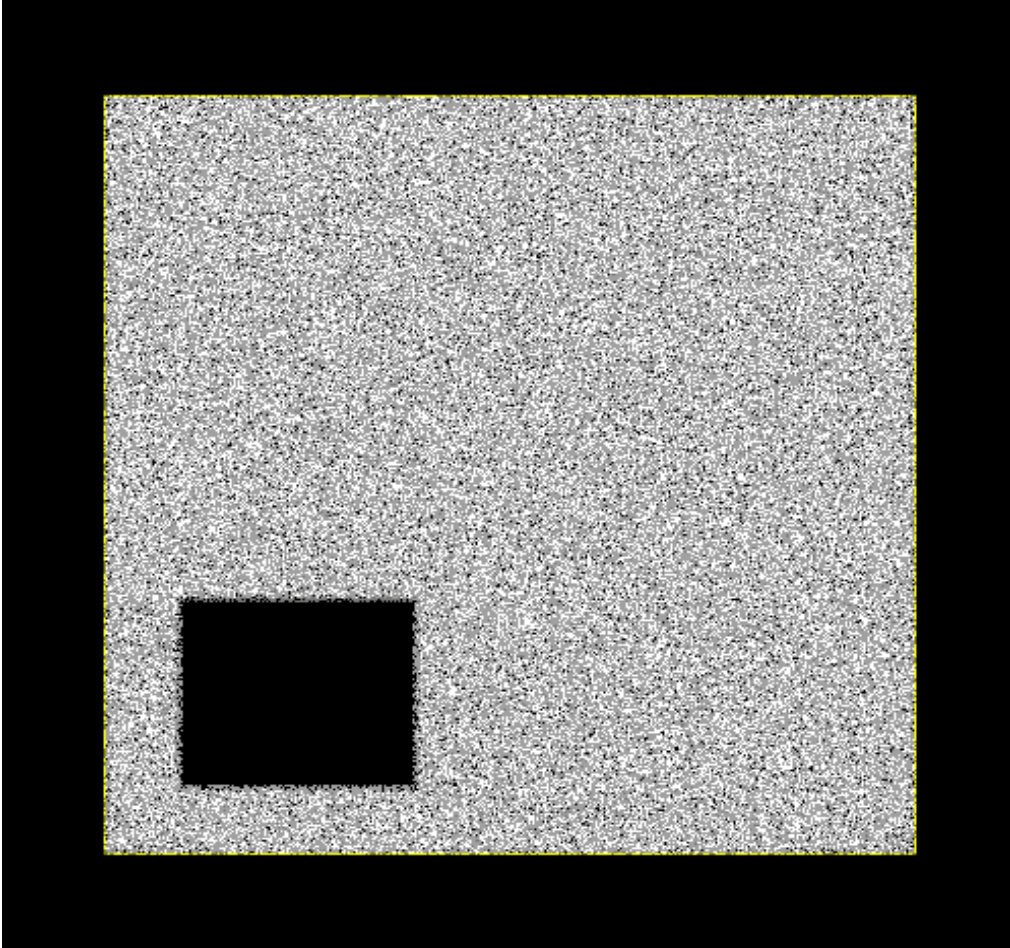
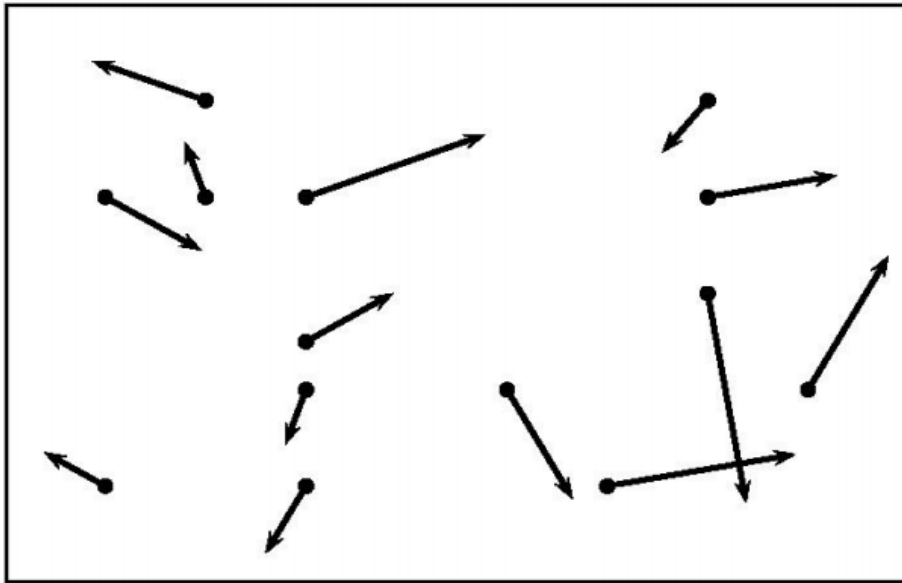


Figure from wiki

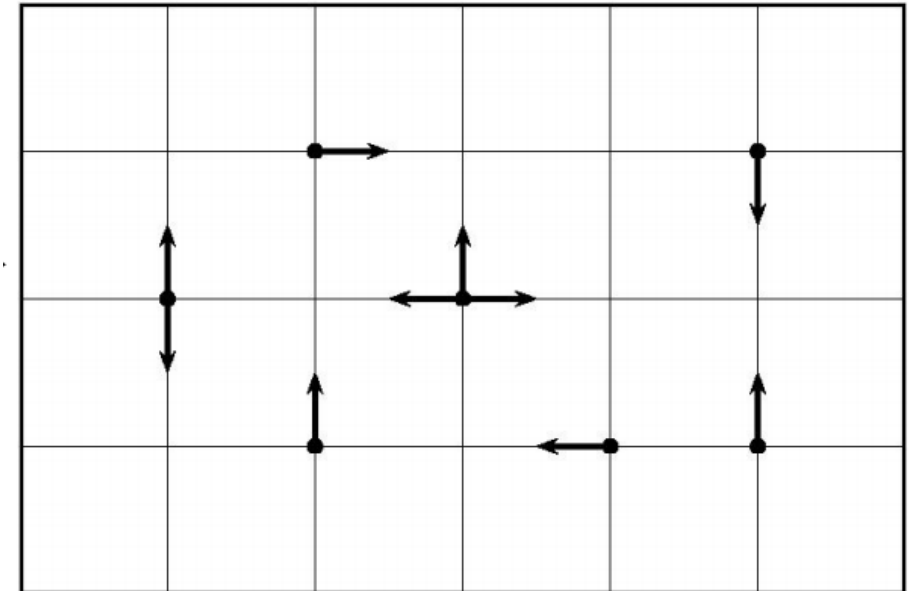
- Originated from early 1990's.
- Lattice Automaton used to simulate fluid flows
- Comprises of a lattice with different states on sites.
- Lattice Gas: states are represented by particles with certain velocities.
- State at each site is purely boolean: there either *is* or *is not* a particle travel in each direction.
- Evolution is done by two steps in each time step: streaming and collision
- Precursor to Lattice Boltzmann Method

# Microscopic Dynamics

Microscopic particles inside fluids



Fictitious particles moving along lattice links

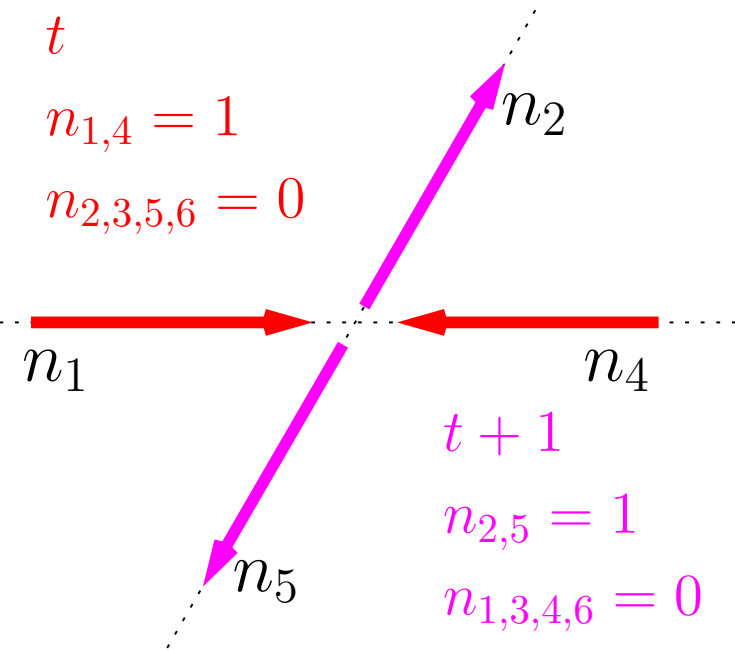
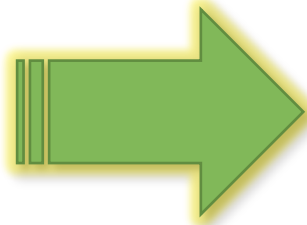
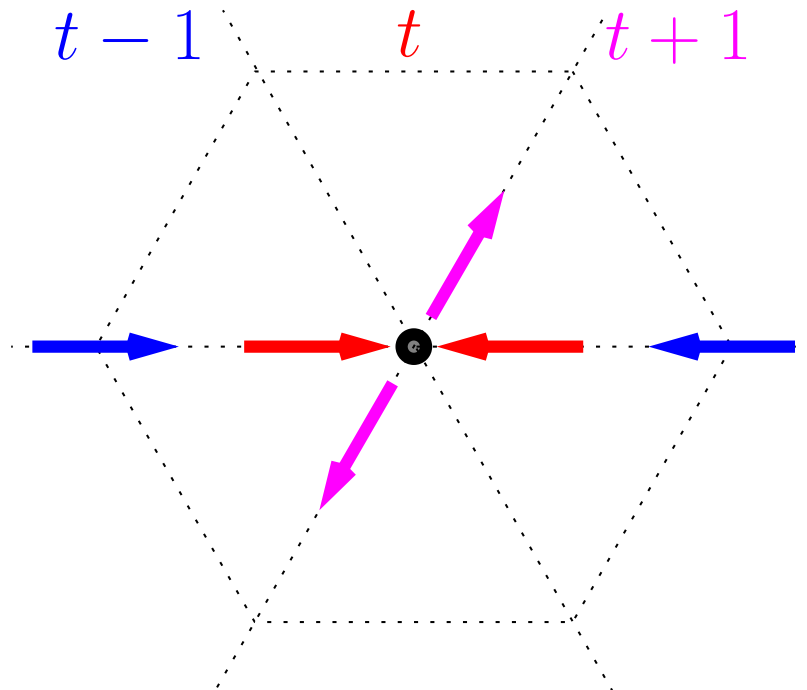


# Lattice Gas Automata (LGA)

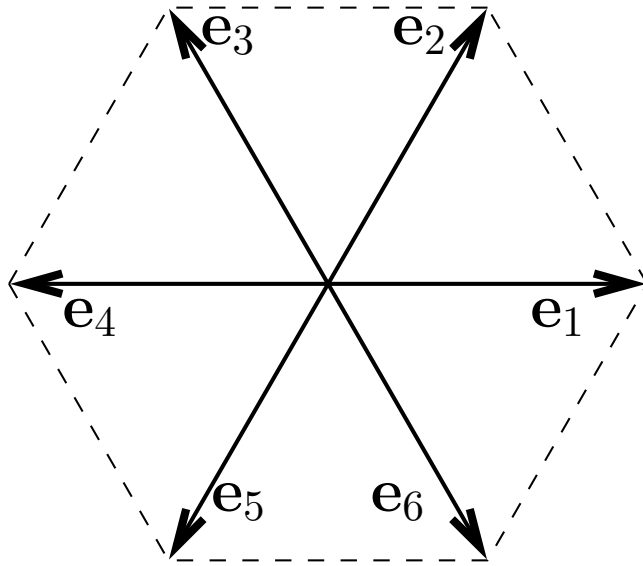
Evolution Equation

$$n_i(x + e_i \delta t, t + \delta t) = n_i(x, t) + \Omega_i(n(x, t))$$

Collision term  
-1,0,1



# Lattice Gas Automata (LGA)



FHP model  
First introduced in 1986 by  
Frisch, Hasslacher and Pomeau

- Consider a set of boolean variables:

$$n_i(x, t), \quad i = 0, 1, \dots, M$$

- Particle presentation

$$n_i(x, t) = 0 \quad \text{No particles at site } x \text{ and time } t$$

$$n_i(x, t) = 1 \quad \text{A particle is present at site } x \text{ and time } t$$

- Collision rules

$$\Omega_i(n(x, t)) = -1, 0, 1$$

- Mass conservation
- Momentum conservation

# Lattice Boltzmann Equation

**LGA**  $n_i(x + e_i \delta t, t + \delta t) = n_i(x, t) + \Omega_i(n(x, t))$

**Probability distribution function**

$$f_i = f_i(x, \xi, t)$$

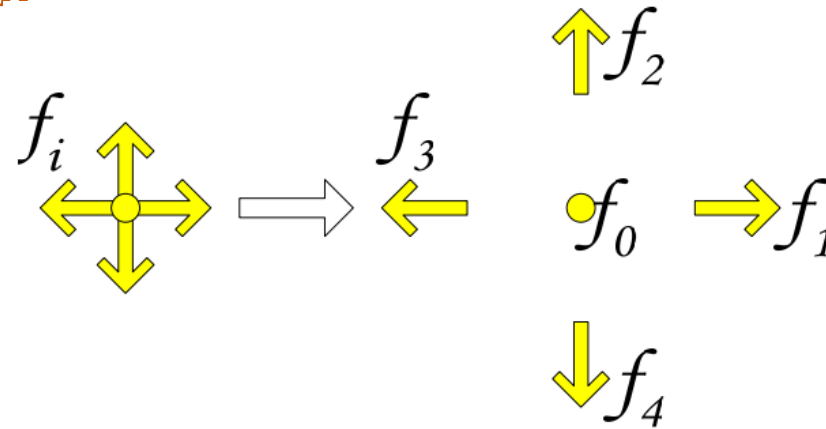
**LBM**  $f_i(x + e_i \delta t, t + \delta t) = f_i(x, t) + \Omega_i(f(x, t))$



# Macroscopic Properties

## Probability distribution function

$$f_i = f_i(x, \xi, t)$$



Flow properties easily computed from particle distribution values per time step

$$\rho = \sum_i f_i \quad u = \frac{\sum_i f_i e_i}{\rho} \quad v = \frac{2\tau - 1}{6}$$

# Streaming on Lattice

## A D2Q9 Lattice Model

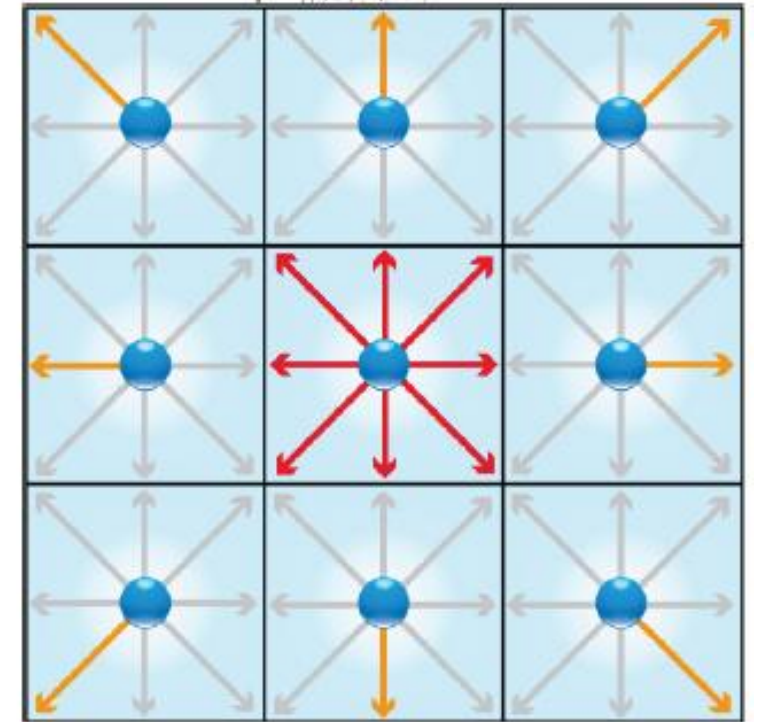
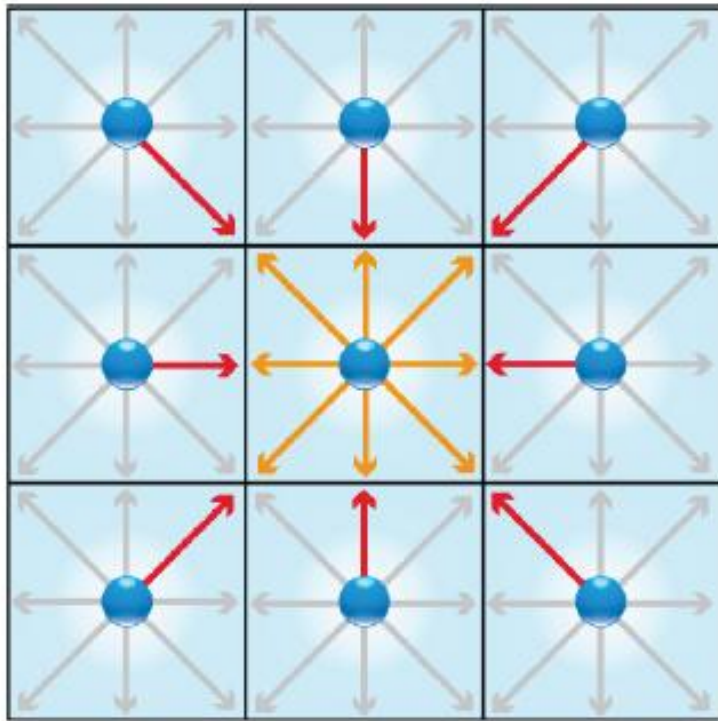


Image from Indo-German winter academy 2011



# Collision Computation

Bhatnagar-Gross-Krook (BGK) collision operator for equilibrium

$$\Omega_i = -\tau^{-1} \left( f_i(x, t) - f_i^{EQ}(\rho, u) \right)$$

where,  $\tau$  is the relaxation time

$$f_i^{EQ}(\rho, u) = \rho \left( A + B(e_i \cdot u) + Cu^2 + D(e_i \cdot u)^2 \right)$$

where, - A, B, C, D are constants defined by lattice geometry

# Boundary Handling

## Microscopic Numerical Fluid Solver

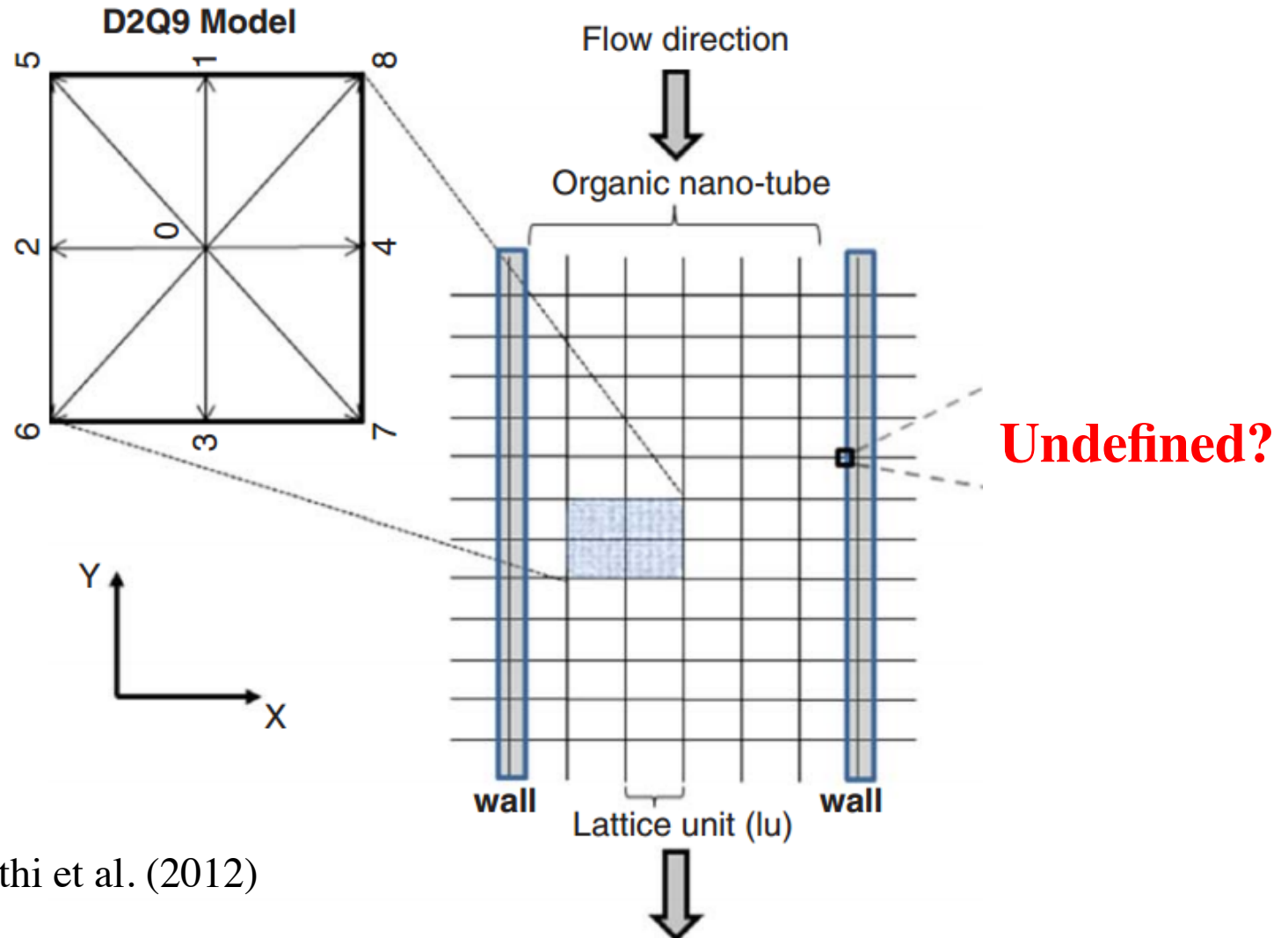
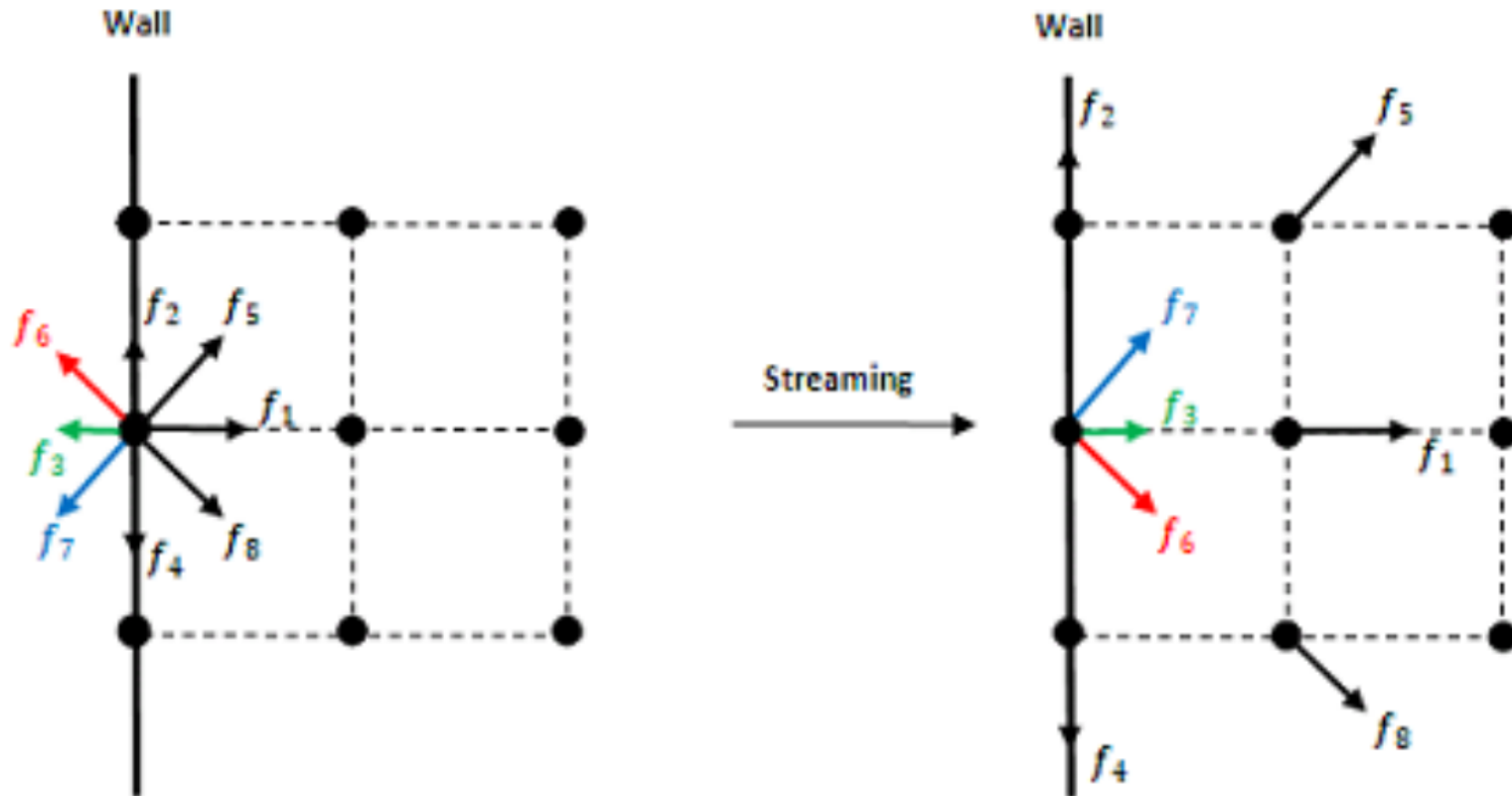


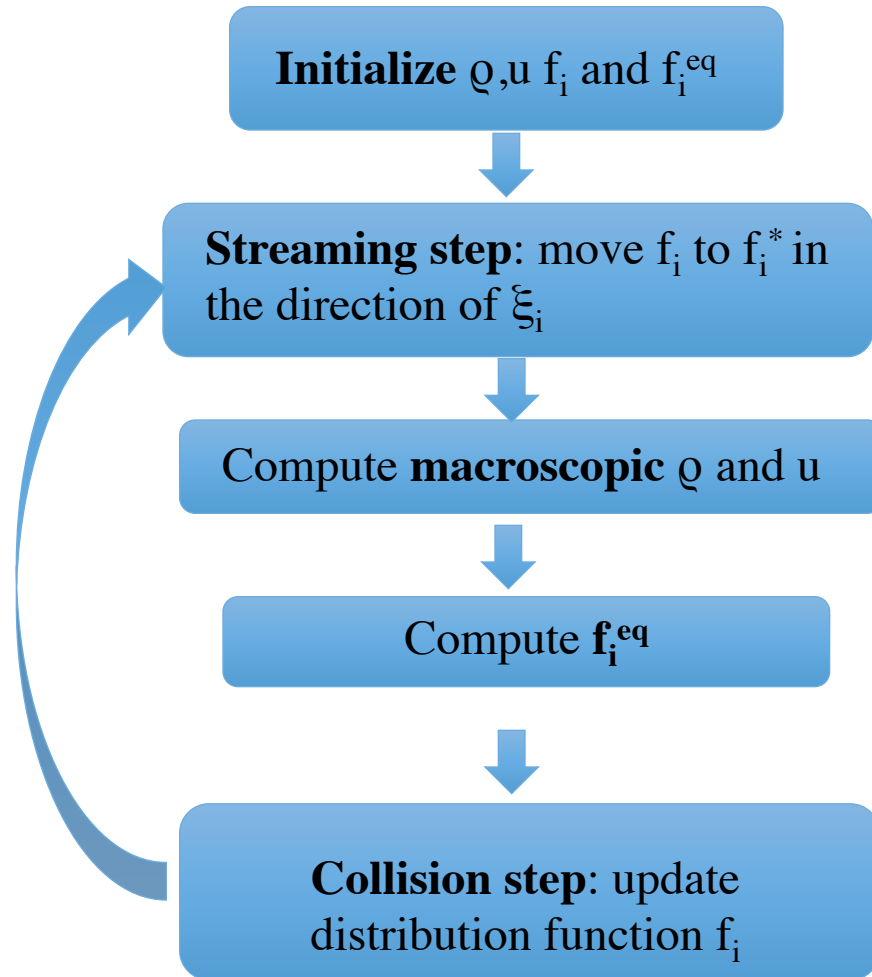
Figure from Fathi et al. (2012)

# Bounce-Back Method

$$f_{-i}(x, t + 1) = f_i(x, t)$$



# Algorithm of LBM



$$\rho = \sum_{\alpha} f_{\alpha}$$

$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) = f(x, \xi, t)$$

$$\rho = \sum_{\alpha} f_{\alpha}$$

$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f^{EQ} = \rho w_{\alpha} \left[ 1 + \frac{\xi \cdot u}{RT} - \frac{u \cdot u}{2RT} + \frac{(\xi \cdot u)^2}{2(RT)^2} \right]$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) - f(x, \xi, t) = -\frac{1}{\tau} [f(x, \xi, t) - f^{EQ}(x, \xi, t)]$$

# LBM vs. CFD

## Conventional CFD Method

---

### Construction of fluid equations

*Navier-Stokes equation*

2<sup>nd</sup>-order PDE, nonlinear convective term

### Discrete approximation of PDE

Finite difference, finite element, etc

### Numerical integration

Solve the equations on a given mesh and apply PDE boundary conditions

## Lattice Boltzmann Method

---

### Discrete formulation of kinetic theory

*Lattice Boltzmann equation*

1st-order PDE, simple advection

### No further approximation

The equations are already in discrete form

### Numerical integration

Solve on lattices and apply kinetic based BC

### Simple conversion to fluid variables

These are theoretically shown to obey the required fluid equations

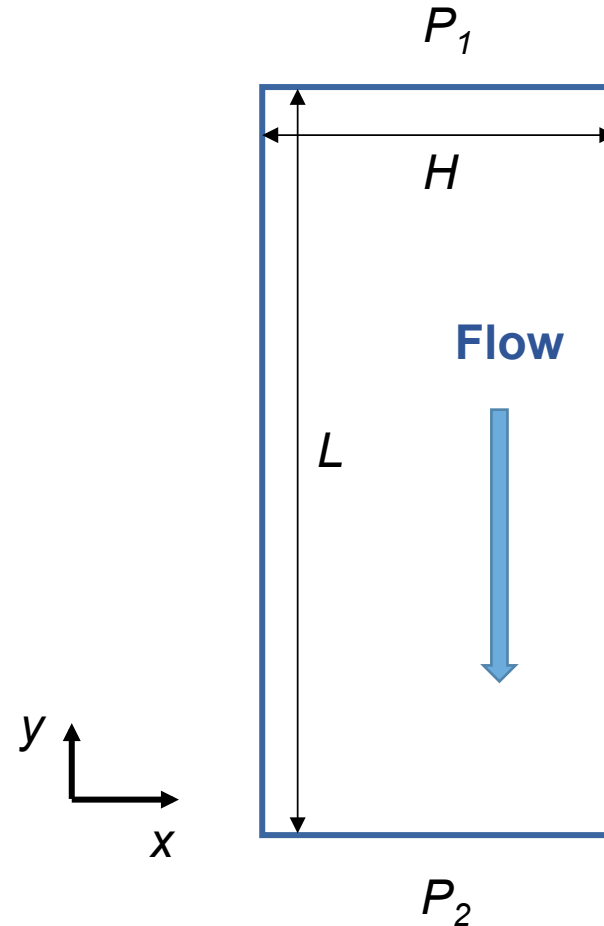
# Steady Poiseuille Flow

Navier-Stokes equation for incompressible flow

$$\mu \frac{\partial^2 u}{\partial x^2} = \frac{\partial p}{\partial y}$$

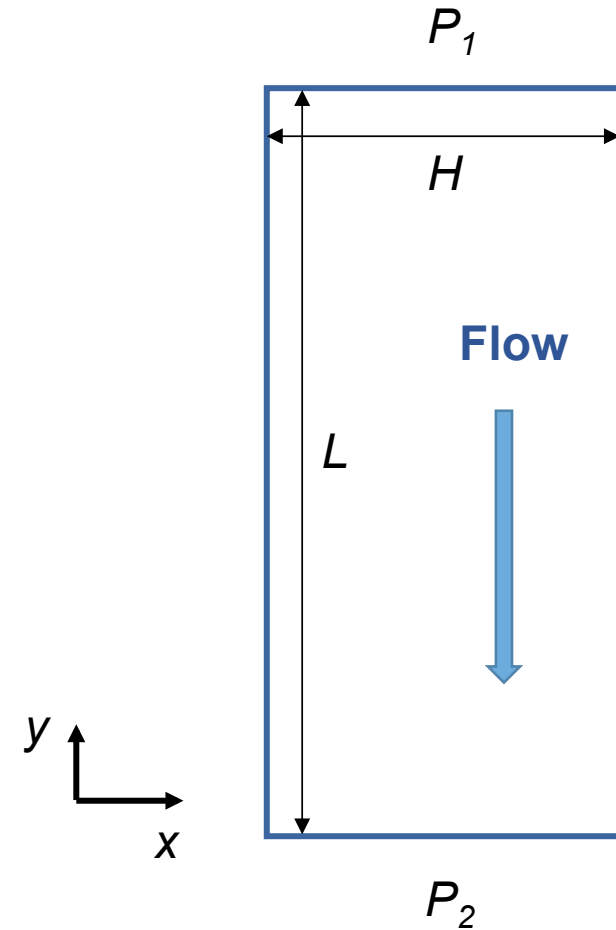
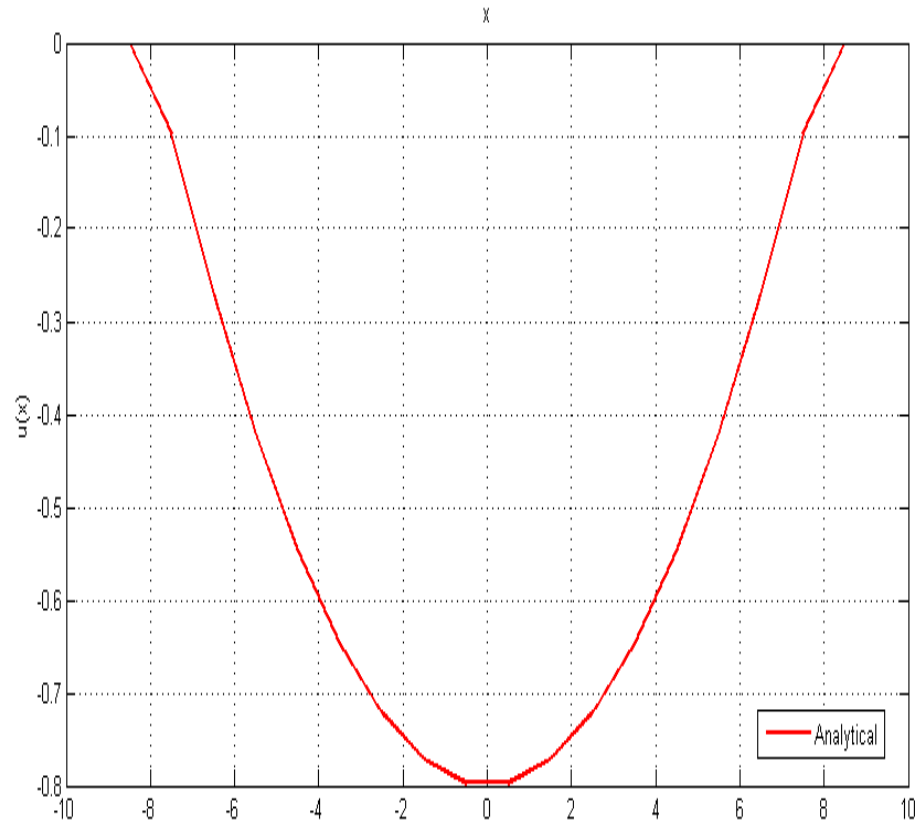
Available exact analytical solution

$$u(x) = \frac{\Delta P}{2\mu L} x(x - H)$$

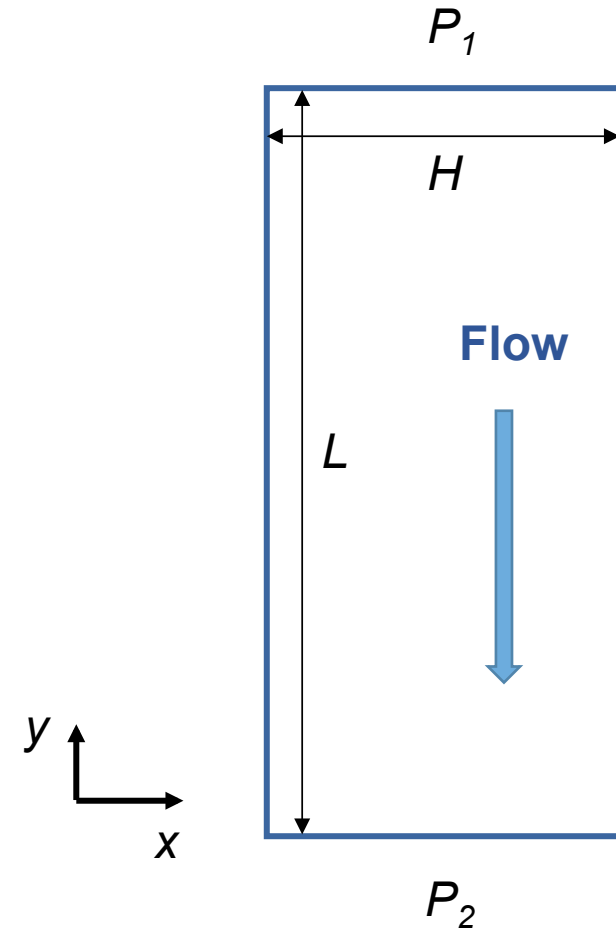
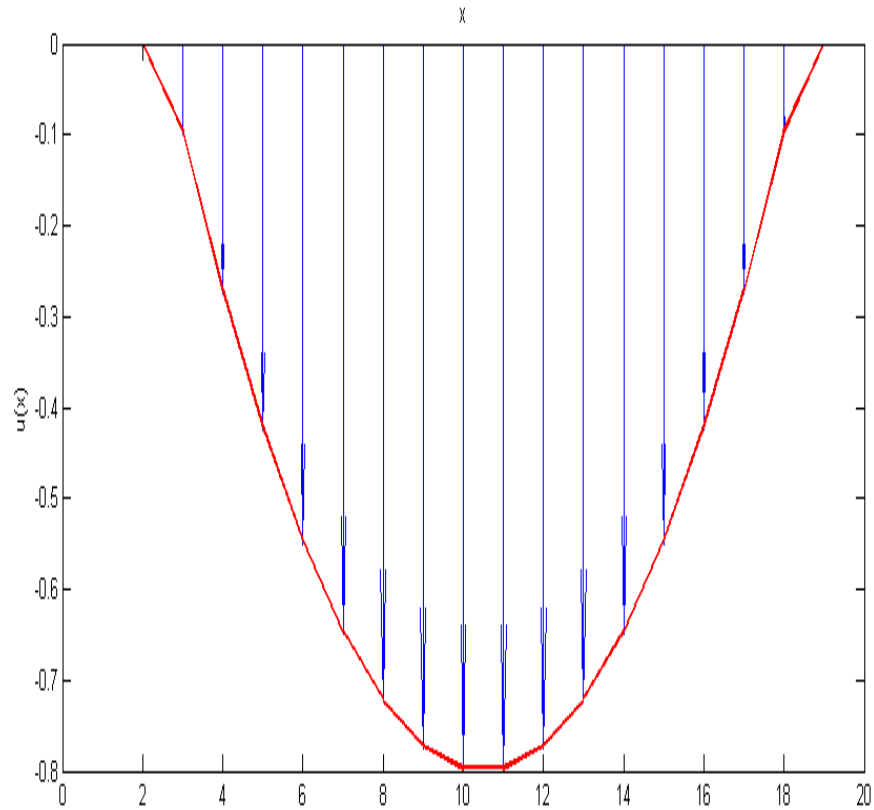




# Analytical Fluid Parabolic Velocity Profile



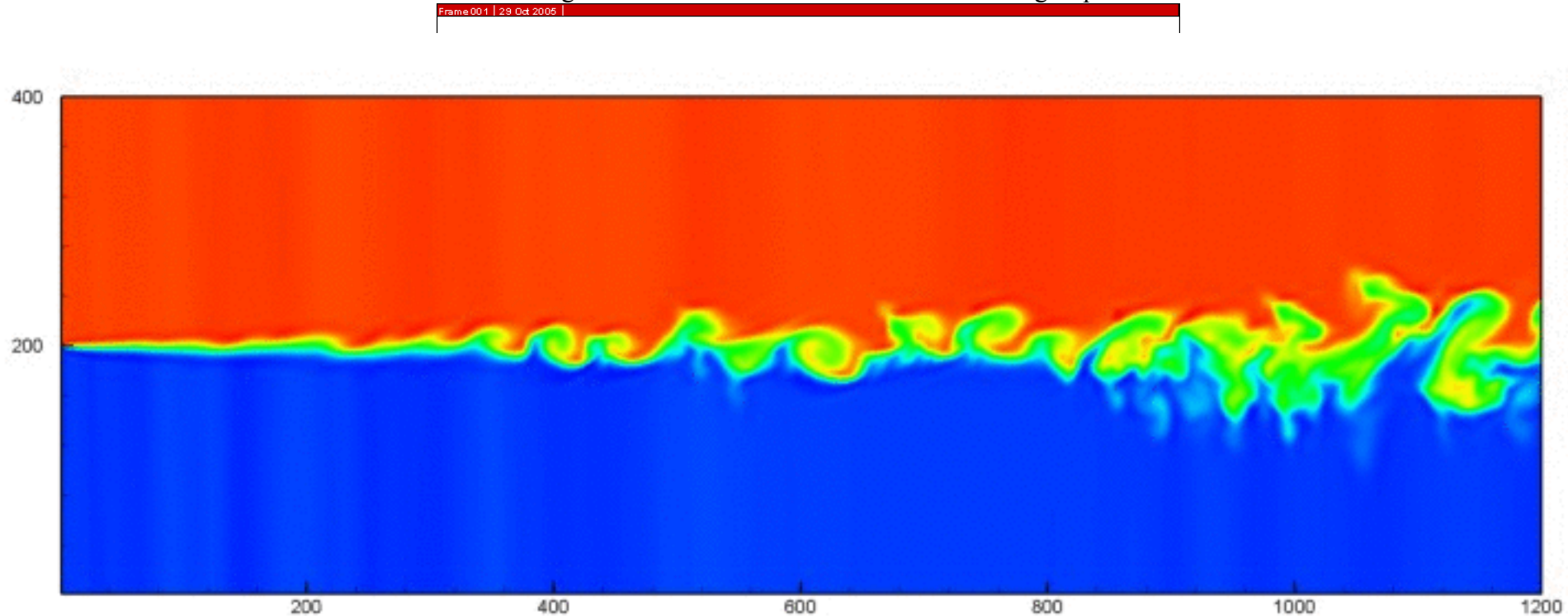
# Validation of the LBM



# Lattice Boltzmann Equation

- We can solve various Fluid Dynamics problems with LBM

Image from nus.edu & combustion fundamental group



Simulation of turbulent mixing in a binary mixture  
Fish motion simulation by LBM



# Governing Equation of LBM

Manipulation & Interpretation

# Lattice Boltzmann Equation

We start from general Boltzmann Equation

$$\frac{\partial f}{\partial t} + \frac{p}{m} \cdot \nabla f + F \cdot \frac{\partial f}{\partial p} = \left( \frac{\partial f}{\partial t} \right)_{coll} \quad \text{--- (1)}$$

In which:

- $f$  is a particle distribution function
- $F$  is external force field acting on the particle
- $m$  is particle mass
- $p$  is particle momentum
- $t$  is time

# Lattice Boltzmann Equation

To derive LB equation, assume zero force field  
Also note that momentum over mass is particle velocity

Together with (1) yields:

$$\frac{\partial f}{\partial t} + \vec{\xi} \cdot \nabla f = \left( \frac{\partial f}{\partial t} \right)_{coll} \quad \text{--- (2)}$$

In which:

- $\xi$  is microscopic velocity  $\frac{p}{m} = \xi$

$$\left( \frac{\partial f}{\partial t} \right)_{coll} = ?$$

Collision term is usually complex and hard to interpret

Collision term is usually approximated using Bhatnagar-Gross-Krook (BGK) collision operator

$$\Omega_i = -\tau^{-1} (n_i - n_i^{EQ})$$

In which:

- $\Omega$  is the collision term
- $\tau$  or  $\lambda$  is a relaxation time representing the amount of time it consumed to return to equilibrium state.
- $n$  or  $f$  is the particle distribution function
- $n^{EQ}$  or  $g$  is the distribution function in equilibrium state.

$$\left( \frac{\partial f}{\partial t} \right)_{coll} = -\frac{1}{\lambda} (f - g)$$

# Lattice Boltzmann Equation

Assemble BGK collision term with LHS yields the general Lattice Boltzmann Equation:

$$\frac{\partial f}{\partial t} + \xi \cdot \nabla f = -\frac{1}{\lambda}(f - g)$$

in which:

- $f$  is the single particle distribution function.
- $\xi$  is the microscopic velocity vector
- $\lambda$  is the relaxation time due to collision
- $g$  is the Boltzmann-Maxwellian distribution function.

$$g \equiv \frac{\rho}{(2\pi RT)^{D/2}} \exp\left(-\frac{(\xi - u)^2}{2RT}\right)$$

in which:

- $D$  is the dimension of space
- $R$  is the ideal gas constant
- $\rho$ ,  $T$  and  $u$  are the macroscopic density of mass, temperature and velocity respectively. They are moments of distribution function  $f$ .

# Lattice Boltzmann Equation

Compute macroscopic quantities (moments of distribution function  $f$ )

$$\rho = \int f d\xi = \int g d\xi$$

$$\rho u = \int \xi f d\xi = \int \xi g d\xi$$

$$\rho \varepsilon = \frac{1}{2} \int (\xi - u)^2 f d\xi = \frac{1}{2} \int (\xi - u)^2 g d\xi$$

Macroscopic quantities can be represented by integrating the distribution function in proper order

**That's the beauty of LBM**



# Discretized LB Equation

Chapman-Enskog assumption

$$\int h(\xi) f(x, \xi, t) d\xi = \int h(\xi) g(x, \xi, t) d\xi$$

$$h(\xi) = A + B \cdot \xi + C \xi \cdot \xi$$

in which:

- A and C are arbitrary constants, B is an arbitrary constant vector

By writing LB equation in an ODE form and implementing Chapman-Enskog assumption

**We can discretize LB equation in time**

$$\frac{df}{dt} + \frac{1}{\lambda} f = \frac{1}{\lambda} g$$

$$\frac{d}{dt} \equiv \frac{\partial}{\partial t} + \xi \cdot \nabla$$

The Equation can be formally integrated over time step  $\delta_t$

# Discretized LB Equation

$$f(\mathbf{x} + \boldsymbol{\xi}\delta_t, \boldsymbol{\xi}, t + \delta_t) = \frac{1}{\lambda} e^{-\delta_t/\lambda} \int_0^{\delta_t} e^{t'/\lambda} g(\mathbf{x} + \boldsymbol{\xi}t', \boldsymbol{\xi}, t + t') dt' + e^{-\delta_t/\lambda} f(\mathbf{x}, \boldsymbol{\xi}, t).$$

Assuming that  $\delta_t$  is small enough and  $g$  is smooth enough locally, the following approximation can be made:

$$g(\mathbf{x} + \boldsymbol{\xi}t', \boldsymbol{\xi}, t + t') = \left(1 - \frac{t'}{\delta_t}\right) g(\mathbf{x}, \boldsymbol{\xi}, t) + \frac{t'}{\delta_t} g(\mathbf{x} + \boldsymbol{\xi}\delta_t, \boldsymbol{\xi}, t + \delta_t) + O(\delta_t^2), \quad 0 \leq t' \leq \delta_t.$$

The leading terms neglected in the above approximation are of the order of  $O(\delta_t^2)$ . With this approximation, Eq. (8) becomes

$$\begin{aligned} f(\mathbf{x} + \boldsymbol{\xi}\delta_t, \boldsymbol{\xi}, t + \delta_t) - f(\mathbf{x}, \boldsymbol{\xi}, t) &= (e^{-\delta_t/\lambda} - 1)[f(\mathbf{x}, \boldsymbol{\xi}, t) - g(\mathbf{x}, \boldsymbol{\xi}, t)] \\ &+ \left(1 + \frac{\lambda}{\delta_t} (e^{-\delta_t/\lambda} - 1)\right) \\ &\times [g(\mathbf{x} + \boldsymbol{\xi}\delta_t, \boldsymbol{\xi}, t + \delta_t) - g(\mathbf{x}, \boldsymbol{\xi}, t)]. \end{aligned}$$

If we expand  $e^{-\delta_t/\lambda}$  in its Taylor expansion and, further, neglect the terms of order  $O(\delta_t^2)$  or smaller on the right-hand side of Eq. (10), then Eq. (10) becomes

$$f(\mathbf{x} + \boldsymbol{\xi}\delta_t, \boldsymbol{\xi}, t + \delta_t) - f(\mathbf{x}, \boldsymbol{\xi}, t) = -\frac{1}{\tau} [f(\mathbf{x}, \boldsymbol{\xi}, t) - g(\mathbf{x}, \boldsymbol{\xi}, t)],$$

# Discretized LB Equation

$$f(x + \xi \delta_t, \xi, t + \delta_t) - f(x, \xi, t) = \delta_t [f(x, \xi, t) - g(x, \xi, t)]$$

Above is the evolution of the distribution function over time

$$\rho = \int f d\xi = \int g d\xi$$

$$\rho u = \int \xi f d\xi = \int \xi g d\xi$$

$$\rho \varepsilon = \frac{1}{2} \int (\xi - u)^2 f d\xi = \frac{1}{2} \int (\xi - u)^2 g d\xi$$

with discrete

In which  $\tau$  is the dimensionless relaxation time  
 $\delta_t$

Recall: we can calculate macroscopic quantities by integrating in momentum space.

**The integration can be approximated by quadrature up to a certain degree of accuracy.**

# Discretized LB Equation

The approximating quadrature takes the form:

$$\int \Psi(\xi) g(x, \xi, t) d\xi = \sum_{\alpha} W_{\alpha} \Psi(\xi_{\alpha}) g(x, \xi_{\alpha}, t)$$

Where  $\Psi(\xi)$  is a polynomial of  $\xi$ ,  $W_{\alpha}$  is the weight coefficient of the quadrature, and  $\xi_{\alpha}$  is the discrete velocity set. Accordingly, the hydrodynamic moments can be computed by:

$$\rho = \sum_{\alpha} f_{\alpha} = \sum_{\alpha} g_{\alpha}$$

$$\rho u = \sum_{\alpha} \xi_{\alpha} f_{\alpha} = \sum_{\alpha} \xi_{\alpha} g_{\alpha}$$

$$\rho \varepsilon = \frac{1}{2} \sum_{\alpha} (\xi_{\alpha} - u)^2 f_{\alpha} = \frac{1}{2} \sum_{\alpha} (\xi_{\alpha} - u)^2 g_{\alpha}$$

Where:

$$f_{\alpha} \equiv f_{\alpha}(x, t) \equiv W_{\alpha} f(x, \xi_{\alpha}, t)$$

$$g_{\alpha} \equiv g_{\alpha}(x, t) \equiv W_{\alpha} g(x, \xi_{\alpha}, t)$$

**Question becomes finding:**

- 1. A approximation of distribution function  $f$**
- 2. Weight coefficients**

# Approximation of Distribution Function

Recall Boltzmann-Maxwellian distribution function:  $f \equiv \frac{\rho}{(2\pi RT)^{D/2}} \exp\left(-\frac{(\xi - u)^2}{2RT}\right)$

Assume D=2, which means a 2-D case

$$\begin{aligned} f &= \frac{\rho}{(2\pi RT)^{D/2}} \exp\left(-\frac{(\xi - u)^2}{2RT}\right) \\ &= \frac{\rho}{(2\pi RT)} \exp\left(-\frac{\xi \cdot \xi}{2RT}\right) \exp\left(\frac{2\xi \cdot u - u \cdot u}{2RT}\right) \\ &\approx \frac{\rho}{(2\pi RT)} \exp\left(-\frac{\xi \cdot \xi}{2RT}\right) \left[1 + \frac{\xi \cdot u}{RT} - \frac{u \cdot u}{2RT} + \frac{(\xi \cdot u)^2}{2(RT)^2}\right] \\ &= \rho w_\alpha \left[1 + \frac{\xi \cdot u}{RT} - \frac{u \cdot u}{2RT} + \frac{(\xi \cdot u)^2}{2(RT)^2}\right] \end{aligned}$$



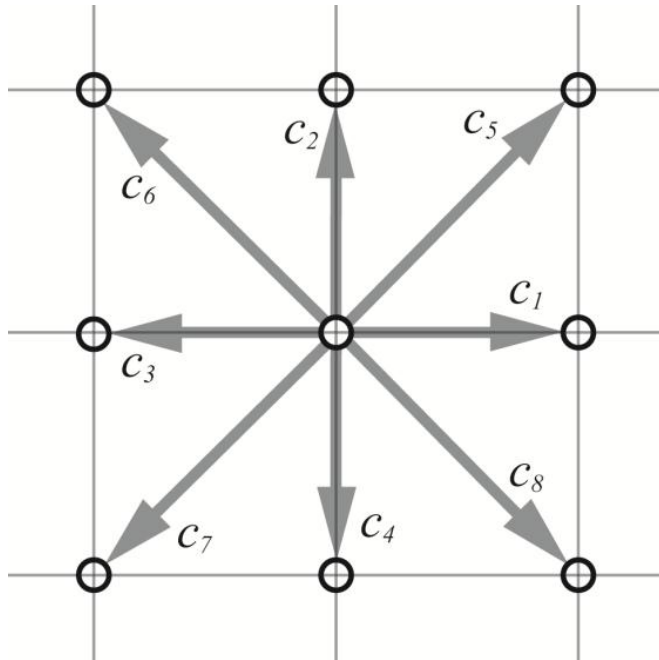
$$\begin{aligned} f_\alpha &= \rho w_\alpha \left[1 + \frac{\xi \cdot u}{RT} - \frac{u \cdot u}{2RT} + \frac{(\xi \cdot u)^2}{2(RT)^2}\right] \\ f_\alpha^{EQ} &= \rho w_\alpha \left[1 + \frac{\xi \cdot u'}{RT} - \frac{u' \cdot u'}{2RT} + \frac{(\xi \cdot u')^2}{2(RT)^2}\right] \end{aligned}$$

$$RT = c_s^2 = c^2/3$$

$c_s$  is the sound speed of the system

# Weighting Coefficients

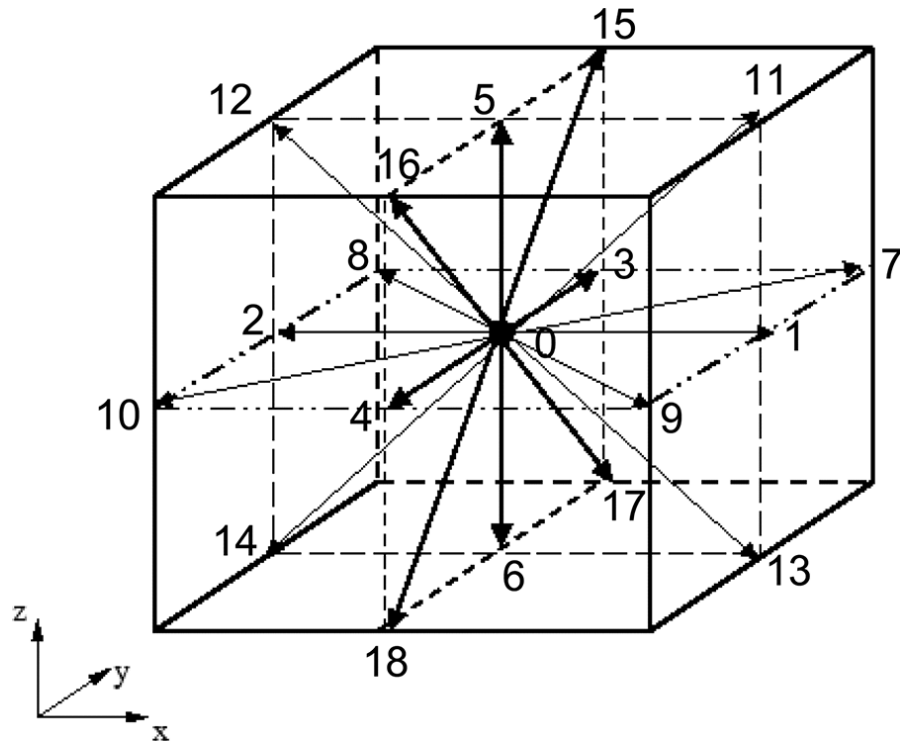
Weight  $w_\alpha$  depends on Lattice arrangements



D2Q9 model  
Image from nus.edu

$$w_\alpha = \begin{cases} \frac{4}{9}, & \alpha = 9 \\ \frac{1}{9}, & \alpha = 1, 2, 3, 4 \\ \frac{1}{36}, & \alpha = 5, 6, 7, 8 \end{cases}$$

# Lattice Boltzmann Method



$$w_{\alpha} = \begin{cases} \frac{2}{36} & \alpha = 1 \dots 6 \\ \frac{1}{36} & \alpha = 7 \dots 18 \\ \frac{12}{36} & \alpha = 19 \end{cases}$$

D3Q19 Lattice model  
Image from ASME Digital Collection



# Summary

$$f = \rho w_\alpha \left[ 1 + \frac{\xi \cdot u}{RT} - \frac{u \cdot u}{2RT} + \frac{(\xi \cdot u)^2}{2(RT)^2} \right]$$

$$f^{EQ} = \rho w_\alpha \left[ 1 + \frac{\xi \cdot u}{RT} - \frac{u \cdot u}{2RT} + \frac{(\xi \cdot u)^2}{2(RT)^2} \right]$$

$$w_\alpha = \begin{cases} \frac{4}{9}, & \alpha = 0 \\ \frac{1}{9}, & \alpha = 1, 2, 3, 4 \\ \frac{1}{36}, & \alpha = 5, 6, 7, 8 \end{cases}$$

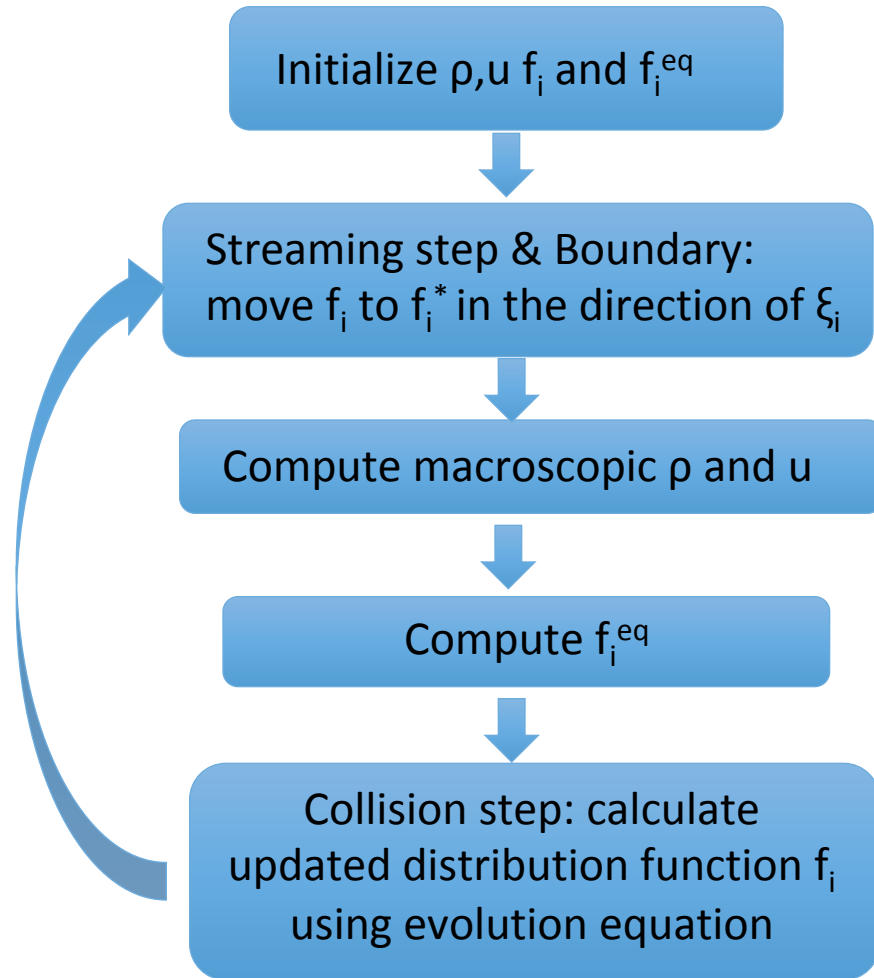
$$\rho = \sum_\alpha f_\alpha$$

$$u = \frac{1}{\rho} \sum_\alpha \xi_\alpha f_\alpha$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) - f(x, \xi, t) = -\frac{1}{\tau} [f(x, \xi, t) - f^{EQ}(x, \xi, t)]$$



# Algorithm of LBM



$$\rho = \sum_{\alpha} f_{\alpha}$$

$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) = f(x, \xi, t)$$

$$\rho = \sum_{\alpha} f_{\alpha}$$

$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f^{EQ} = \rho w_{\alpha} \left[ 1 + \frac{\xi \cdot u}{RT} - \frac{u \cdot u}{2RT} + \frac{(\xi \cdot u)^2}{2(RT)^2} \right]$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) - f(x, \xi, t) = -\frac{1}{\tau} [f(x, \xi, t) - f^{EQ}(x, \xi, t)]$$



# Calculation Example

Steady Channel Flow

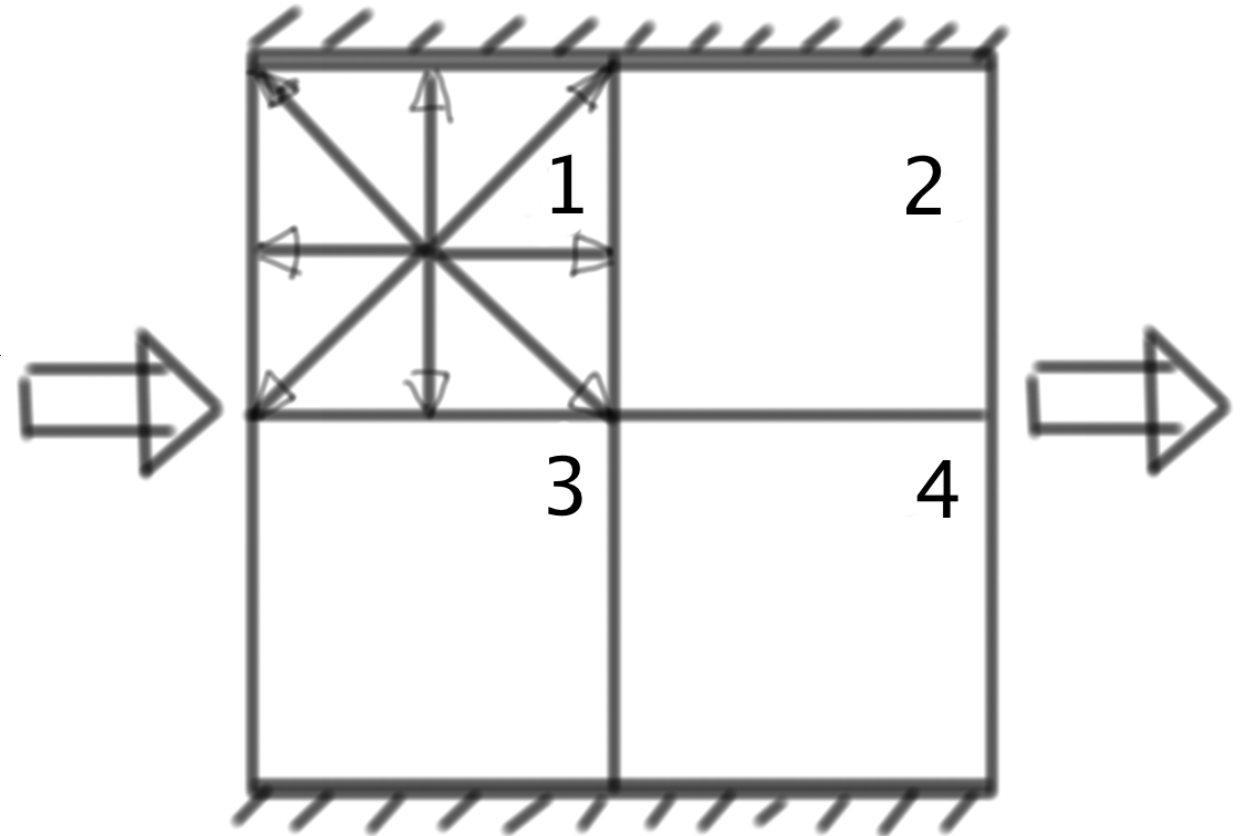
# Problem Description

- D2Q9 model
- 2 by 2 system, 4 lattices
- Channel flow from left to right
- Boundary condition--bounce back
- Initial parameter

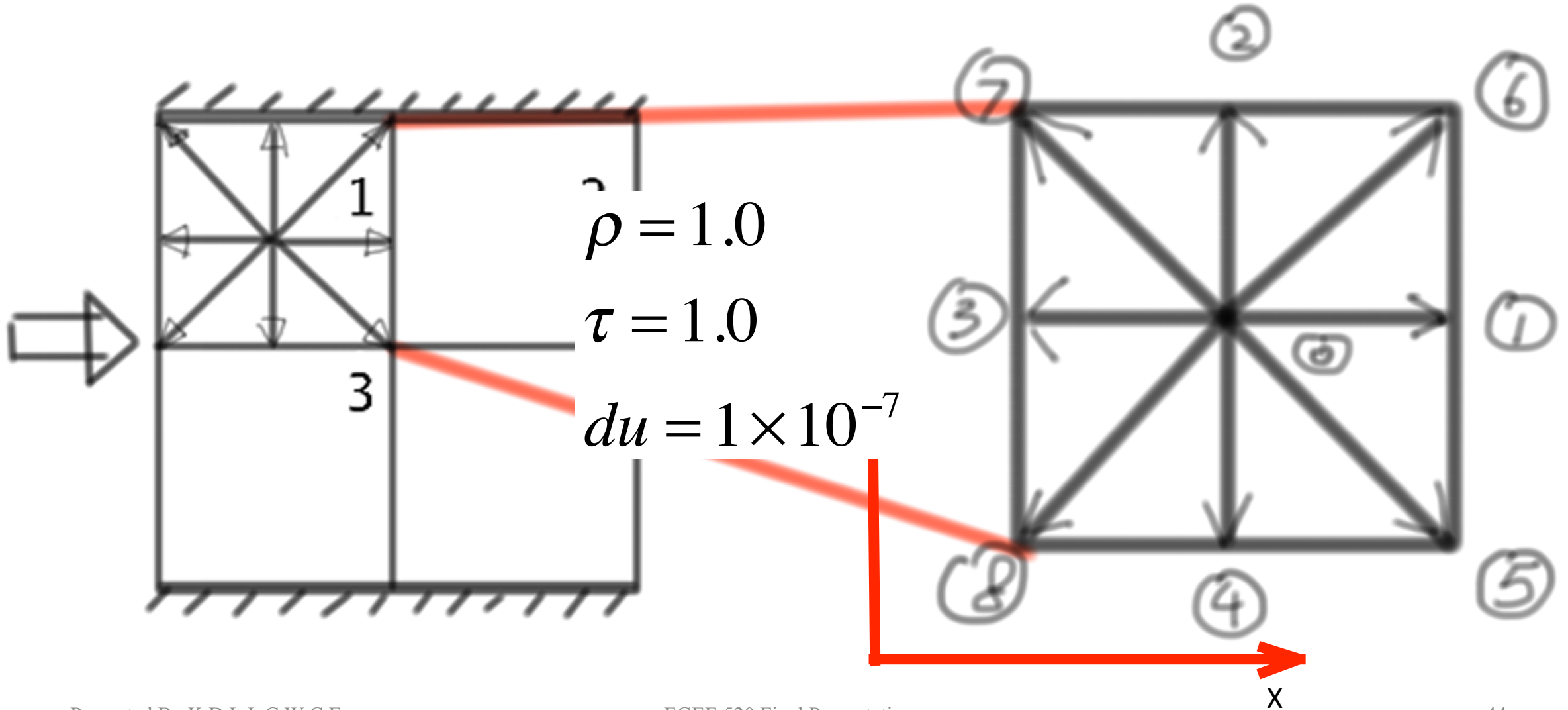
$$\rho = 1.0$$

$$\tau = 1.0$$

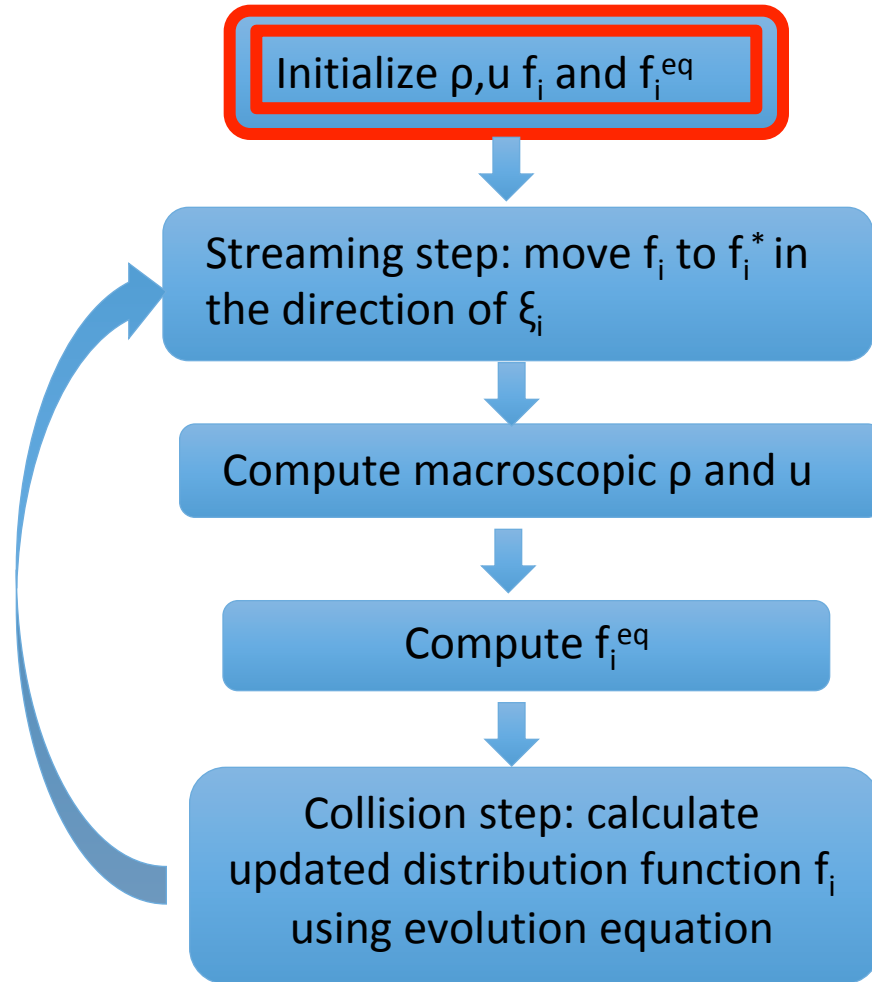
$$du = 1 \times 10^{-7}$$



# Hand Calculation Example



# Algorithm of LBM



$$\rho = \sum_{\alpha} f_{\alpha}$$

$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) = f(x, \xi, t)$$

$$\rho = \sum_{\alpha} f_{\alpha}$$

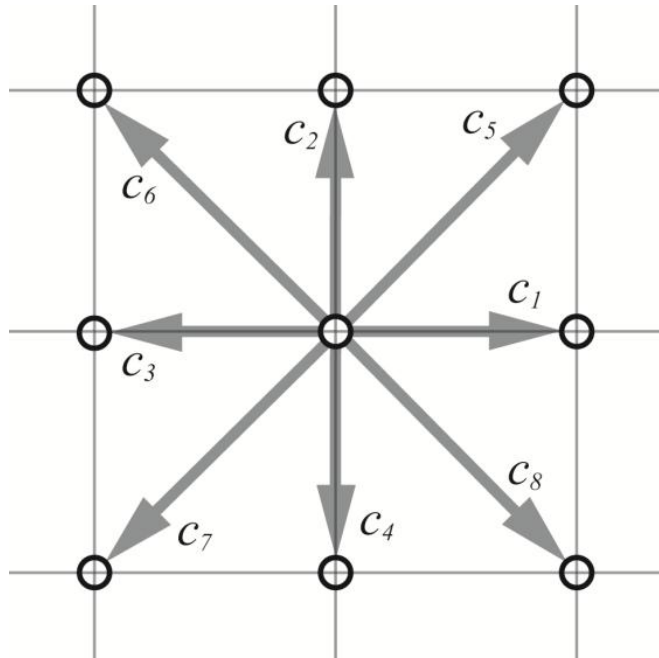
$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f^{EQ} = \rho w_{\alpha} \left[ 1 + \frac{\xi \cdot u}{RT} - \frac{u \cdot u}{2RT} + \frac{(\xi \cdot u)^2}{2(RT)^2} \right]$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) - f(x, \xi, t) = -\frac{1}{\tau} [f(x, \xi, t) - f^{EQ}(x, \xi, t)]$$

# Hand Calculation Example

For each lattice in D2Q9 model, we present velocity by combination of 9 matrices, each matrix contains distribution function  $f_\alpha$ ,  $\alpha=1,2,\dots,9$

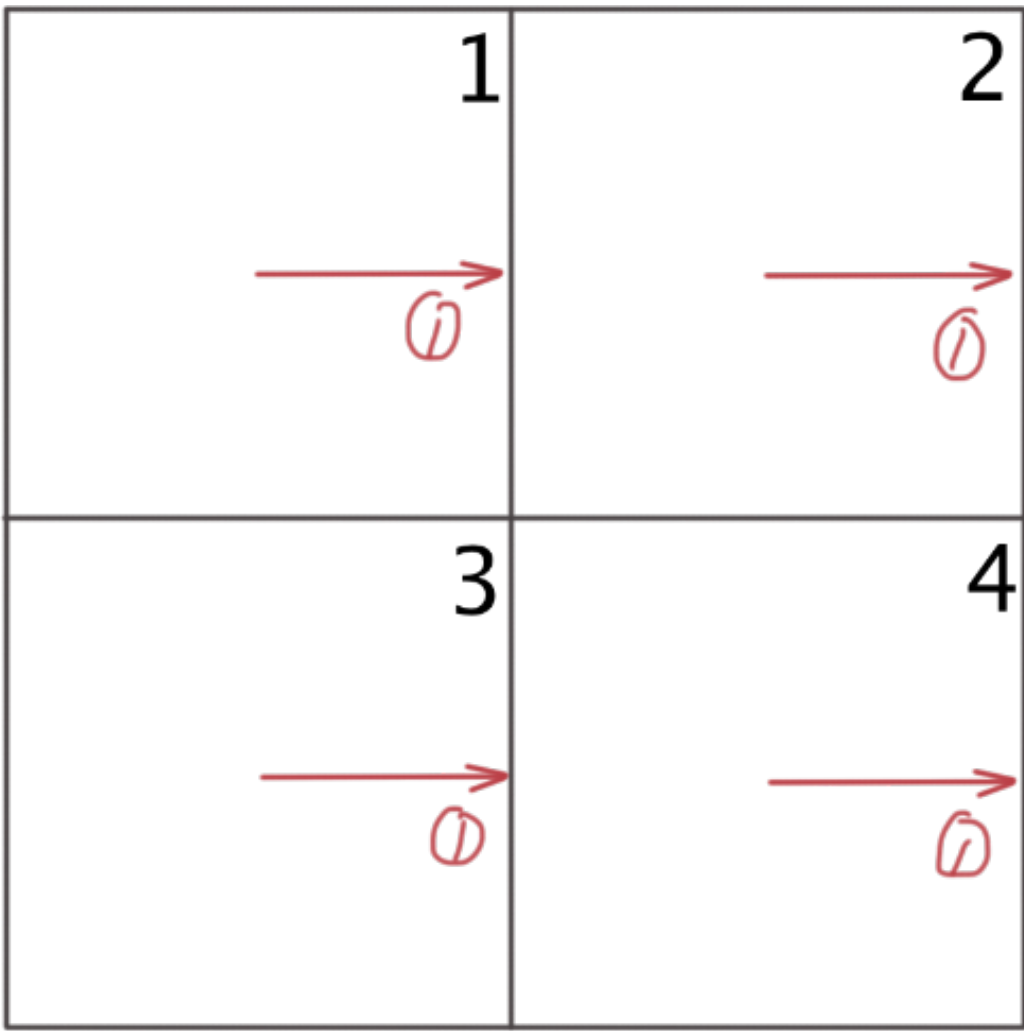


Assume a initial state:

$$f_1 = f_2 = \dots = f_9$$

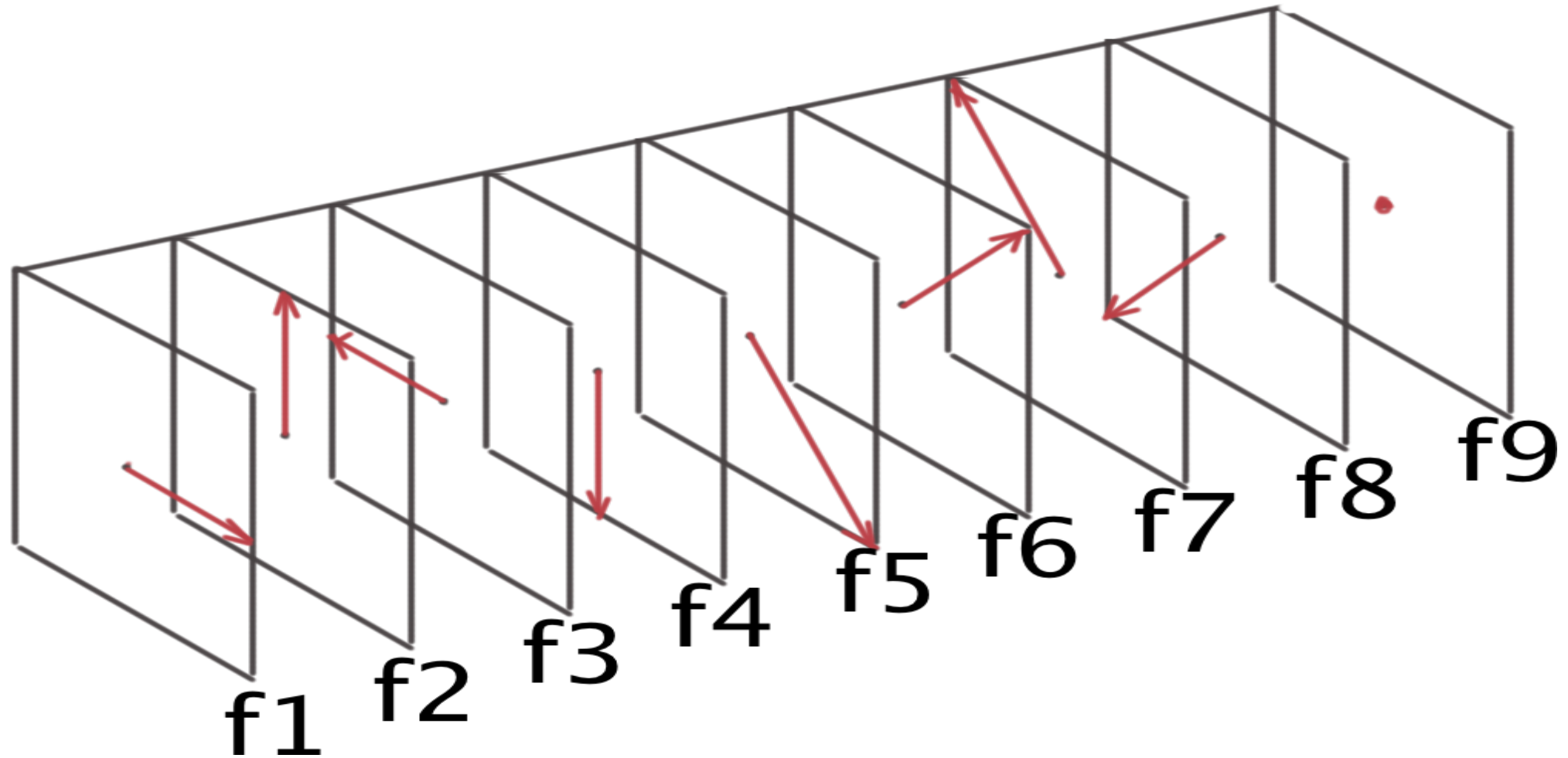
$$f_\alpha = \rho / 9, \alpha = 1, 2, 3, \dots, 9$$

# Hand Calculation Example



$\Rightarrow f_1 = \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix}$

# Hand Calculation Example





# Hand Calculation Example

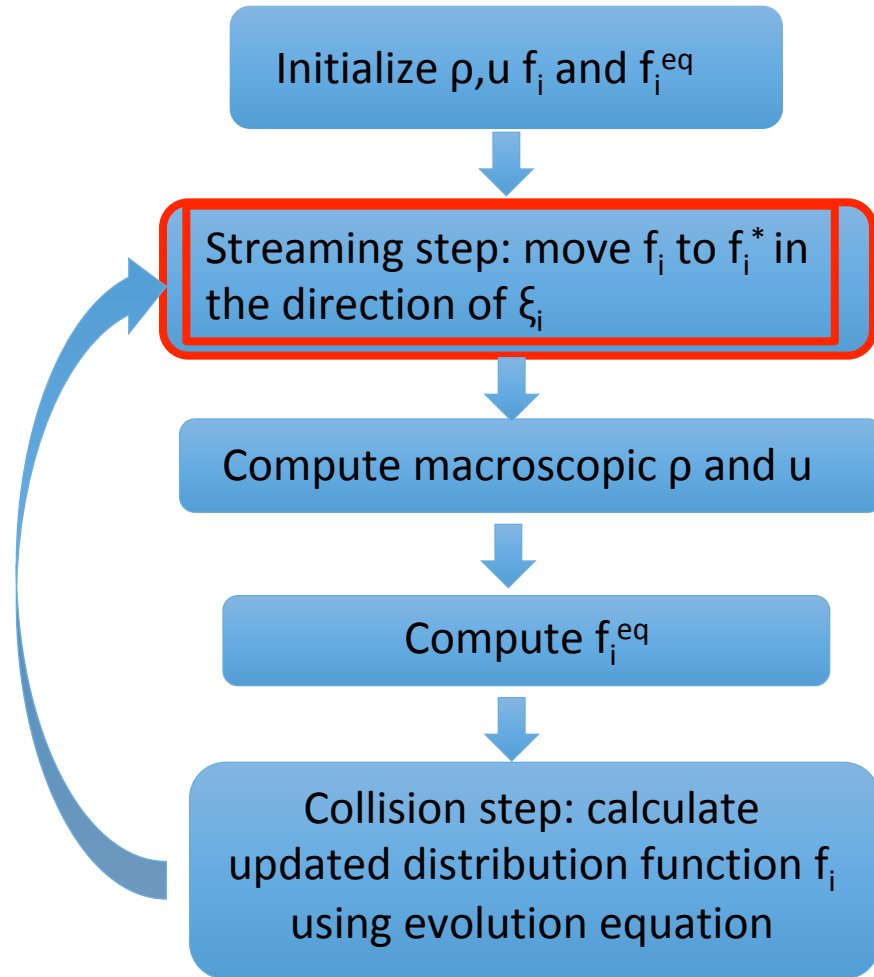
## Distribution Function

$$(f1) = \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \quad (f2) = \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \quad (f3) = \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix}$$

$$(f4) = \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \quad (f5) = \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \quad (f6) = \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix}$$

$$(f7) = \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \quad (f8) = \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \quad (f9) = \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix}$$

# Algorithm of LBM



$$\rho = \sum_{\alpha} f_{\alpha}$$

$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) = f(x, \xi, t)$$

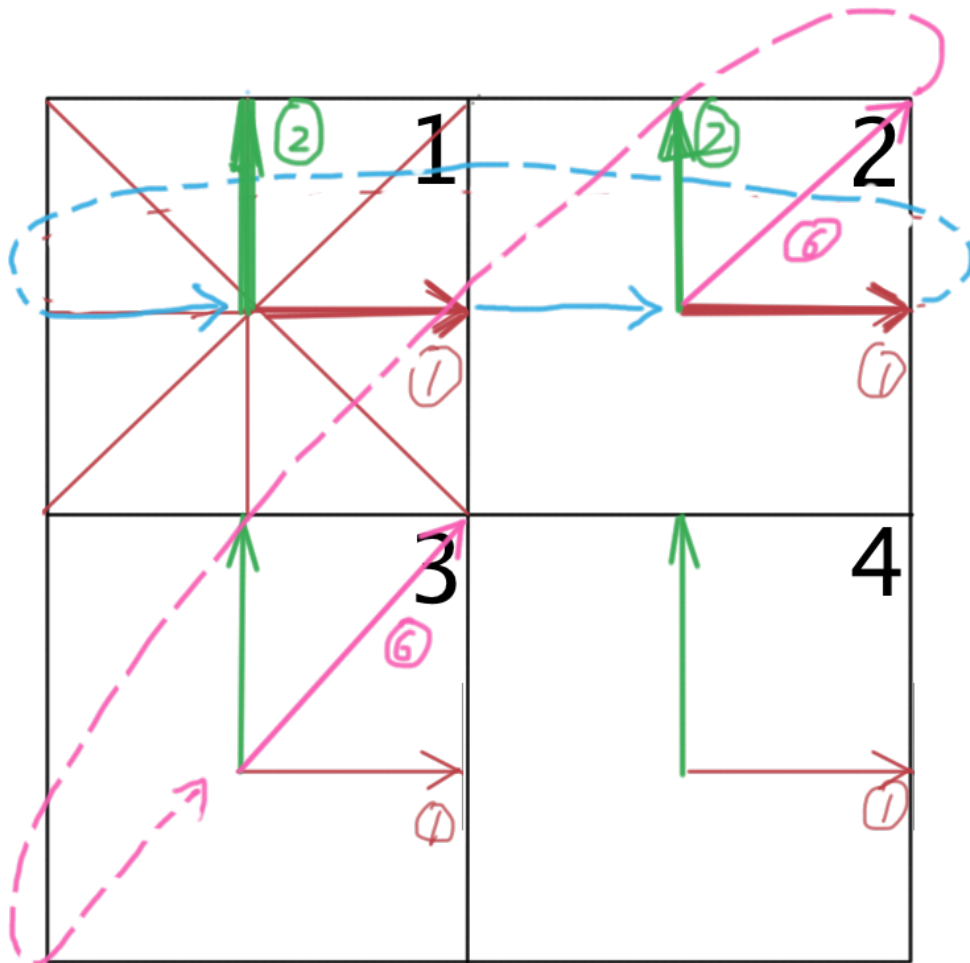
$$\rho = \sum_{\alpha} f_{\alpha}$$

$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

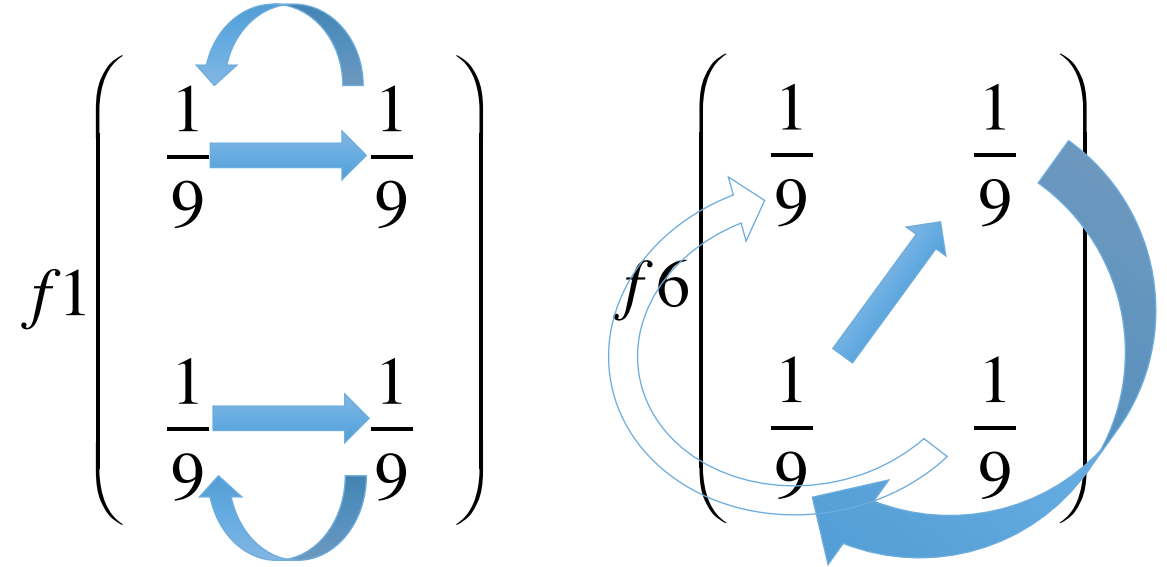
$$f^{EQ} = \rho w_{\alpha} \left[ 1 + \frac{\xi \cdot u}{RT} - \frac{u \cdot u}{2RT} + \frac{(\xi \cdot u)^2}{2(RT)^2} \right]$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) - f(x, \xi, t) = -\frac{1}{\tau} [f(x, \xi, t) - f^{EQ}(x, \xi, t)]$$

# Hand Calculation Example



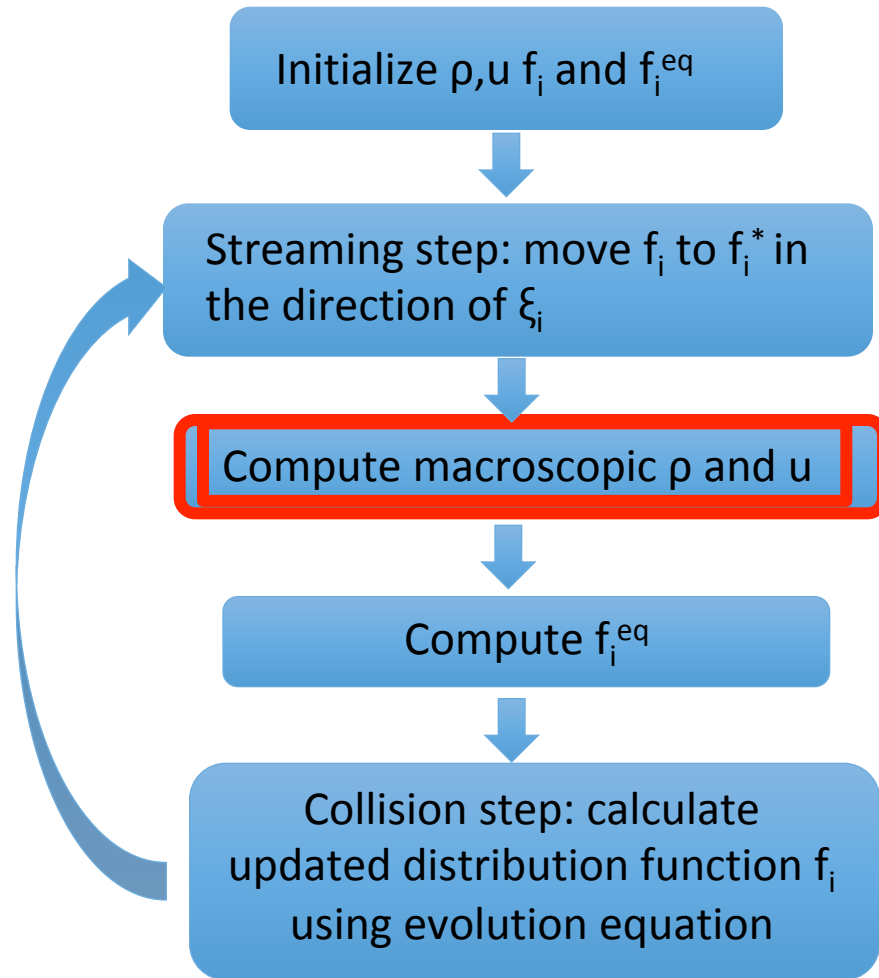
Streaming



# Hand Calculation Example

$$\begin{array}{l}
 (f1 \rightarrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} (f2 \uparrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} (f3 \leftarrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \\
 (f4 \downarrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} (f5 \searrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} (f6 \nearrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \\
 (f7 \swarrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} (f8 \searrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} (f9 \circ) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix}
 \end{array}$$

# Algorithm of LBM



$$\rho = \sum_{\alpha} f_{\alpha}$$

$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) = f(x, \xi, t)$$

$$\rho = \sum_{\alpha} f_{\alpha}$$

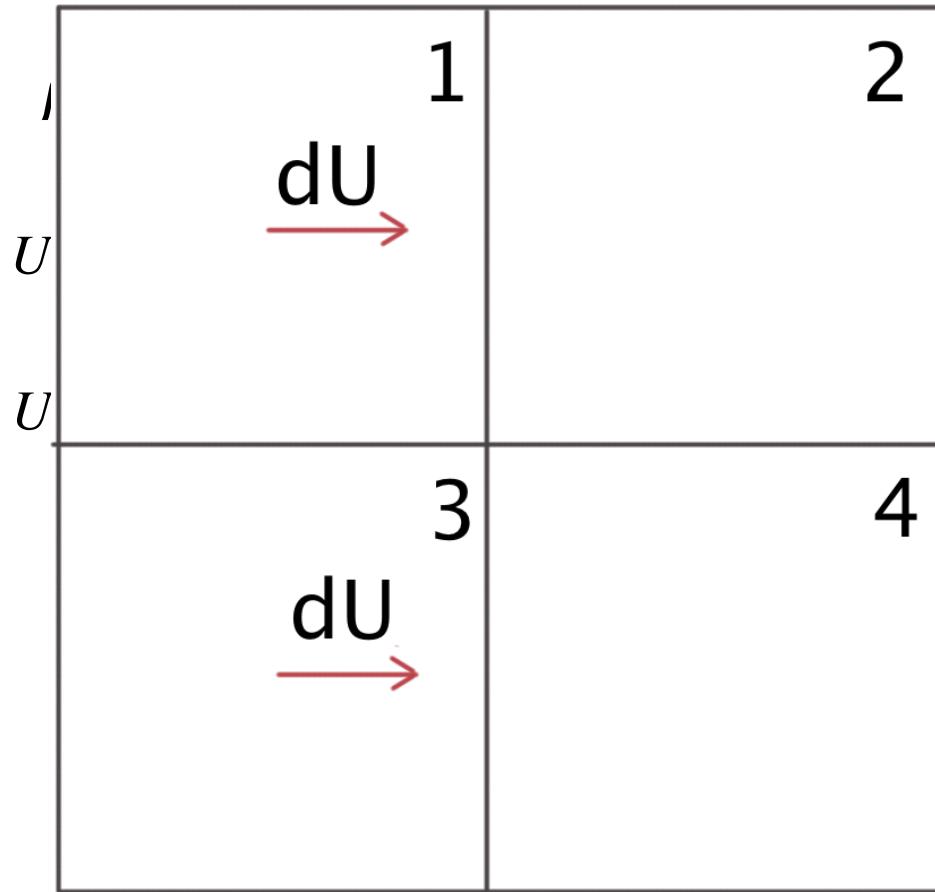
$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f^{EQ} = \rho w_{\alpha} \left[ 1 + \frac{\xi \cdot u}{RT} - \frac{u \cdot u}{2RT} + \frac{(\xi \cdot u)^2}{2(RT)^2} \right]$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) - f(x, \xi, t) = -\frac{1}{\tau} [f(x, \xi, t) - f^{EQ}(x, \xi, t)]$$

# Hand Calculation Example

Calculate Macroscopic Quantities



$$\rho = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \quad \begin{matrix} U_x(1) + du \\ U_x(3) + du \end{matrix}$$

$$U_x = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \quad U_y = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$U \nearrow = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \quad U \swarrow = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$U \nwarrow = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \quad U \searrow = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

# Hand Calculation Example

Calculate Macroscopic Velocities

$$\rho_\alpha = \sum_\alpha f_\alpha = \frac{1}{9} \times 9 = 1$$

$$U_x = \frac{1}{\rho} ((f_1 + f_5 + f_6) - (f_3 + f_7 + f_8))$$

$$U_y = \frac{1}{\rho} ((f_6 + f_2 + f_7) - (f_5 + f_4 + f_8))$$

$$U = U_x^2 + U_y^2$$

$$U \nearrow = U_x + U_y$$

$$U \searrow = U_x - U_y$$

$$U \nwarrow = -U \swarrow$$

$$U \swarrow = -U \nwarrow$$

$$\rho = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \quad U_x(1) + du$$

$$U_x = \begin{pmatrix} 1e-7 & 0 \\ 1e-7 & 0 \end{pmatrix} \quad U_x(3) + du$$

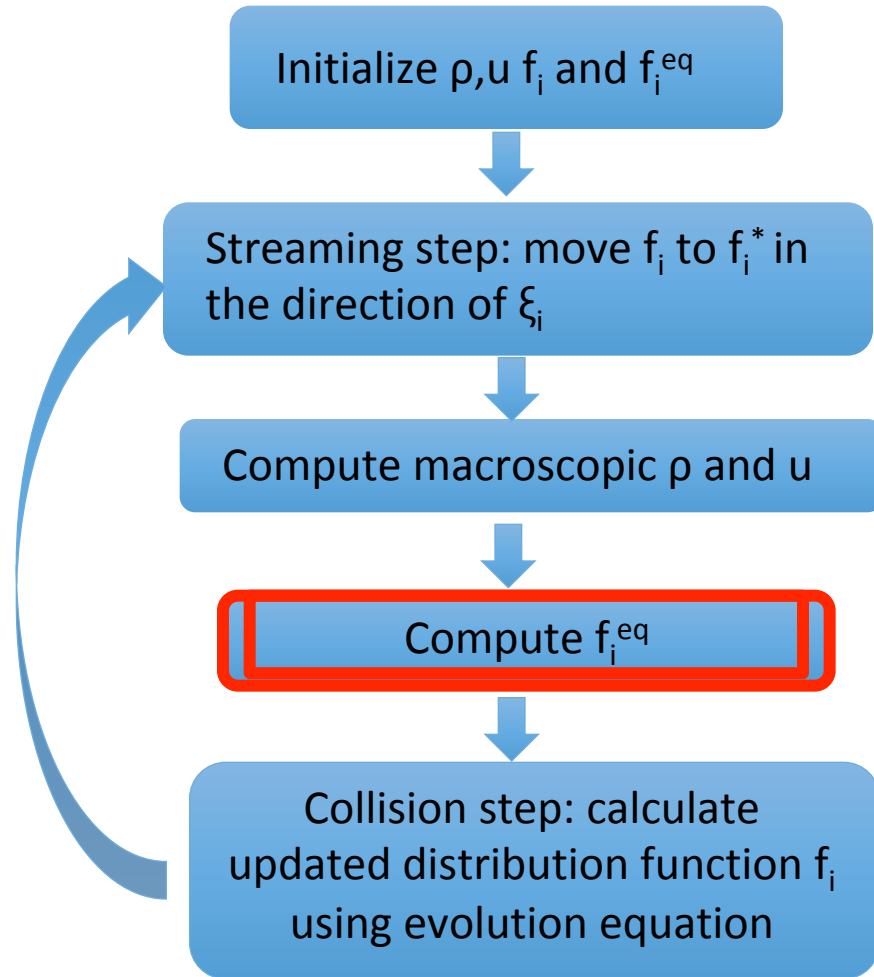
$$U_y = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$U = \begin{pmatrix} 1e-7 & 0 \\ 1e-7 & 0 \end{pmatrix}$$

$$U \nearrow = \begin{pmatrix} 1e-7 & 0 \\ 1e-7 & 0 \end{pmatrix} \quad U \swarrow = \begin{pmatrix} -1e-7 & 0 \\ -1e-7 & 0 \end{pmatrix}$$

$$U \nwarrow = \begin{pmatrix} -1e-7 & 0 \\ -1e-7 & 0 \end{pmatrix} \quad U \searrow = \begin{pmatrix} 1e-7 & 0 \\ 1e-7 & 0 \end{pmatrix}$$

# Algorithm of LBM



$$\rho = \sum_{\alpha} f_{\alpha}$$

$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) = f(x, \xi, t)$$

$$\rho = \sum_{\alpha} f_{\alpha}$$

$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f^{EQ} = \rho w_{\alpha} \left[ 1 + \frac{\xi \cdot u}{RT} - \frac{u \cdot u}{2RT} + \frac{(\xi \cdot u)^2}{2(RT)^2} \right]$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) - f(x, \xi, t) = -\frac{1}{\tau} [f(x, \xi, t) - f^{EQ}(x, \xi, t)]$$



# Hand Calculation Example

Calculate Equilibrium State Distribution Function

$$f_{\alpha}^{EQ} = \rho w_{\alpha} \left[ 1 + \frac{\xi \cdot u'}{RT} - \frac{u' \cdot u'}{2RT} + \frac{(\xi \cdot u')^2}{2(RT)^2} \right]$$

$$\xi \cdot u' = U_x; U_y; U \nearrow; U \swarrow; U \searrow; U \nwarrow.$$

$$u' \cdot u' = U^2$$

$$RT = c_s^2 = c^2/3$$

$$c = 1.0$$

$$f_1^{EQ} = \rho w_1 \left[ 1 + \frac{U_x}{RT} - \frac{U^2}{2RT} + \frac{(U_x)^2}{2(RT)^2} \right] \quad f_5^{EQ} = \rho w_5 \left[ 1 + \frac{U \nearrow}{RT} - \frac{U^2}{2RT} + \frac{(U \nearrow)^2}{2(RT)^2} \right]$$

$$f_2^{EQ} = \rho w_2 \left[ 1 + \frac{U_y}{RT} - \frac{U^2}{2RT} + \frac{(U_y)^2}{2(RT)^2} \right] \quad f_6^{EQ} = \rho w_6 \left[ 1 + \frac{U \nwarrow}{RT} - \frac{U^2}{2RT} + \frac{(U \nwarrow)^2}{2(RT)^2} \right]$$

$$f_3^{EQ} = \rho w_3 \left[ 1 + \frac{-U_x}{RT} - \frac{U^2}{2RT} + \frac{(-U_x)^2}{2(RT)^2} \right] \quad f_7^{EQ} = \rho w_7 \left[ 1 + \frac{U \swarrow}{RT} - \frac{U^2}{2RT} + \frac{(U \swarrow)^2}{2(RT)^2} \right]$$

$$f_4^{EQ} = \rho w_4 \left[ 1 + \frac{-U_y}{RT} - \frac{U^2}{2RT} + \frac{(-U_y)^2}{2(RT)^2} \right] \quad f_8^{EQ} = \rho w_8 \left[ 1 + \frac{U \searrow}{RT} - \frac{U^2}{2RT} + \frac{(U \searrow)^2}{2(RT)^2} \right]$$

$$f_9^{EQ} = \rho w_9 \left[ 1 + \frac{-U_{\circ}}{RT} - \frac{U^2}{2RT} + \frac{(-U_{\circ})^2}{2(RT)^2} \right]$$

# Hand Calculation Example

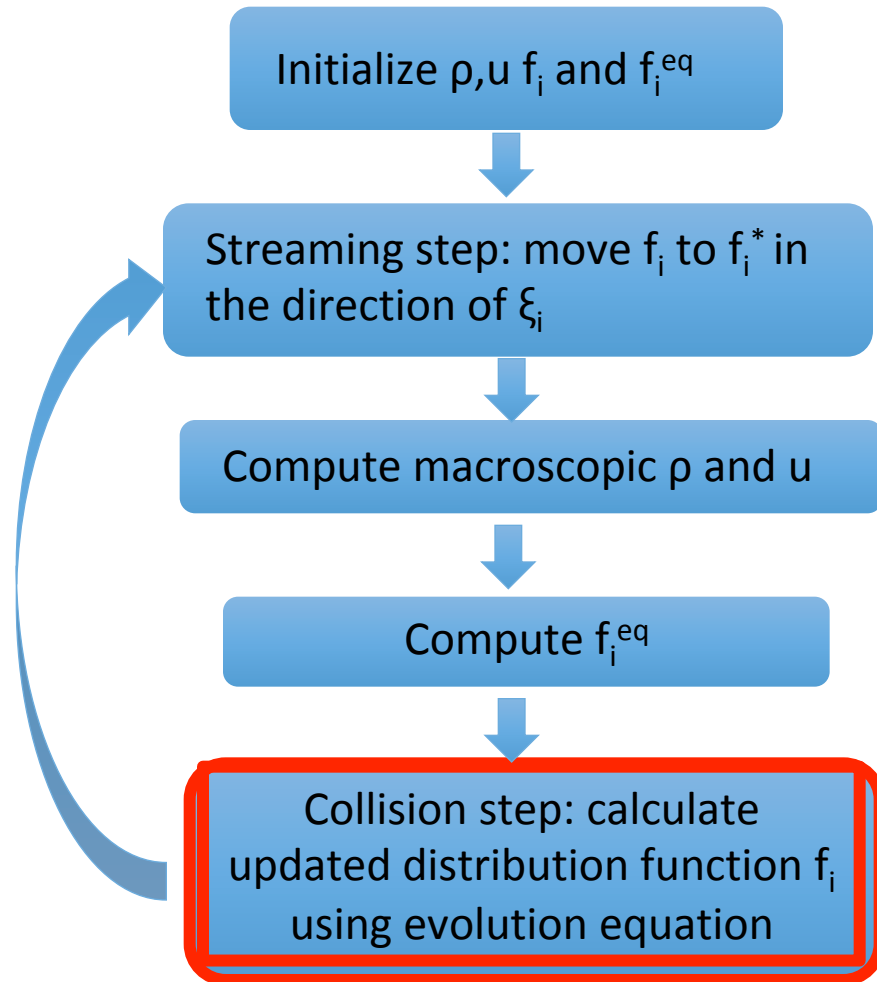
## Calculated Equilibrium State Distribution Function

$$(f1^{EQ} \rightarrow) = \begin{pmatrix} 0.01234568 & 0.012345679 \\ 0.01234568 & 0.012345679 \end{pmatrix} (f2^{EQ} \uparrow) = \begin{pmatrix} 0.012345679 & 0.012345679 \\ 0.012345679 & 0.012345679 \end{pmatrix} (f3^{EQ} \leftarrow) = \begin{pmatrix} 0.01234567 & 0.012345679 \\ 0.01234567 & 0.012345679 \end{pmatrix}$$

$$(f4^{EQ} \downarrow) = \begin{pmatrix} 0.012345679 & 0.012345679 \\ 0.012345679 & 0.012345679 \end{pmatrix} (f5^{EQ} \searrow) = \begin{pmatrix} 0.003086 & 0.003086 \\ 0.003086 & 0.003086 \end{pmatrix} (f6^{EQ} \nearrow) = \begin{pmatrix} 0.003086 & 0.003086 \\ 0.003086 & 0.003086 \end{pmatrix}$$

$$(f7^{EQ} \nwarrow) = \begin{pmatrix} 0.003086 & 0.003086 \\ 0.003086 & 0.003086 \end{pmatrix} (f8^{EQ} \swarrow) = \begin{pmatrix} 0.003086 & 0.003086 \\ 0.003086 & 0.003086 \end{pmatrix} (f9^{EQ} \circ) = \begin{pmatrix} 0.0493827 & 0.0493827 \\ 0.0493827 & 0.0493827 \end{pmatrix}$$

# Algorithm of LBM



$$\rho = \sum_{\alpha} f_{\alpha}$$

$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) = f(x, \xi, t)$$

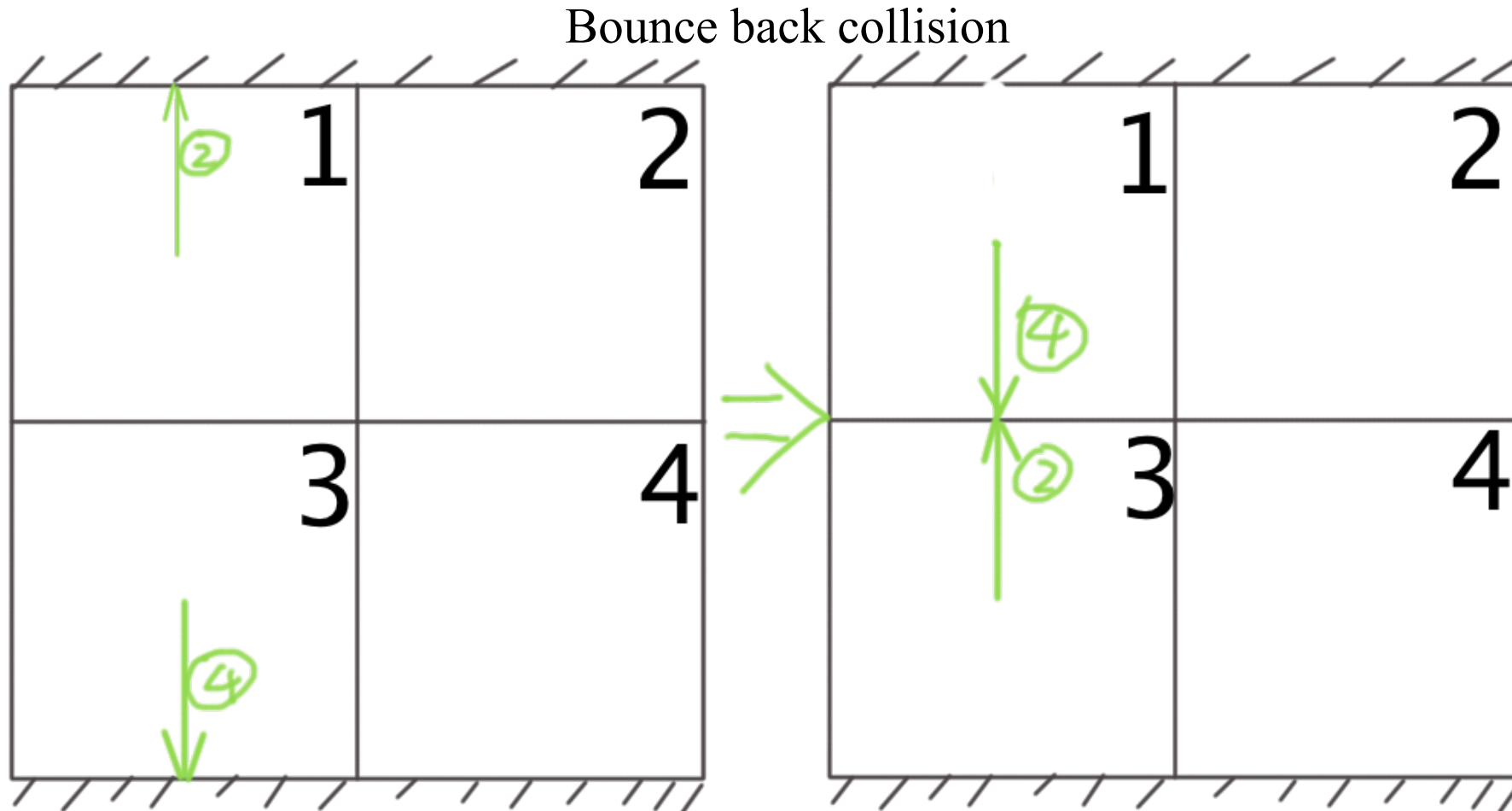
$$\rho = \sum_{\alpha} f_{\alpha}$$

$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f^{EQ} = \rho w_{\alpha} \left[ 1 + \frac{\xi \cdot u}{RT} - \frac{u \cdot u}{2RT} + \frac{(\xi \cdot u)^2}{2(RT)^2} \right]$$

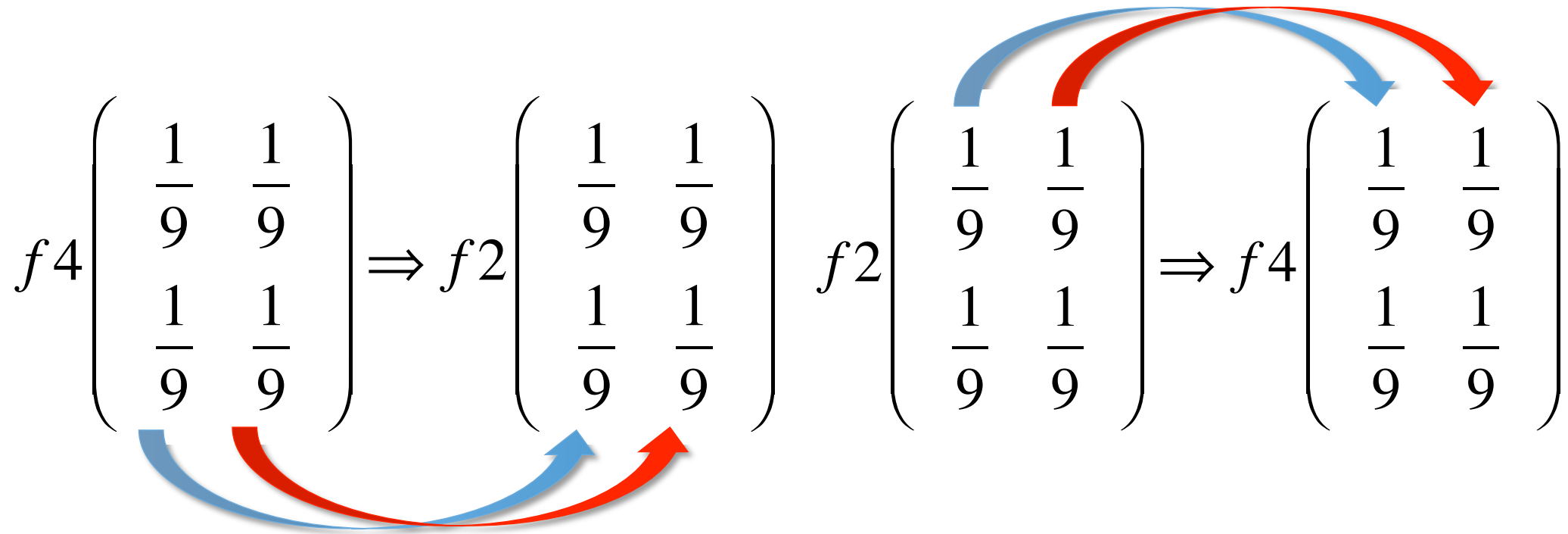
$$f(x + \xi \delta_t, \xi, t + \delta_t) - f(x, \xi, t) = -\frac{1}{\tau} [f(x, \xi, t) - f^{EQ}(x, \xi, t)]$$

# Hand Calculation Example

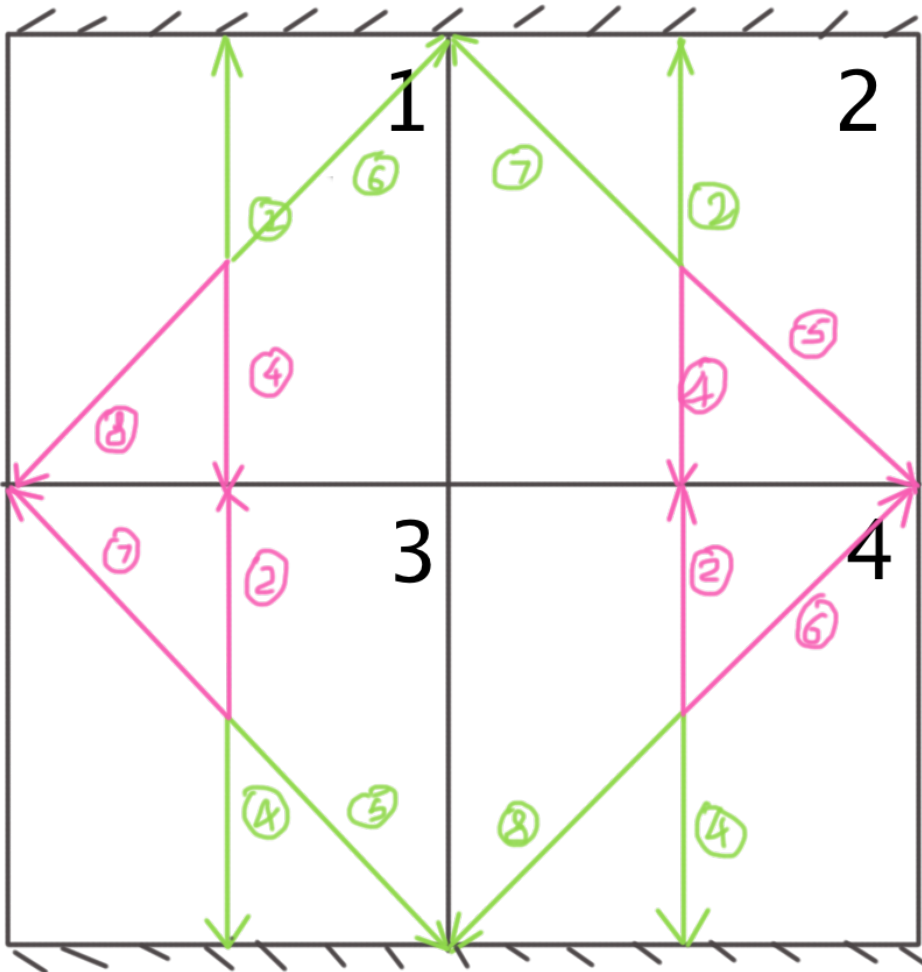


# Hand Calculation Example

Bounce back collision



# Hand Calculation Example



$$\begin{aligned}
 (f2 \uparrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} &\Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} & (f4 \downarrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} &\Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \\
 (f5 \searrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} &\Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} & (f6 \nearrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} &\Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} \\
 (f7 \swarrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} &\Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} & (f8 \searrow) \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix} &\Rightarrow \begin{pmatrix} \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} \end{pmatrix}
 \end{aligned}$$

# Hand Calculation Example

$$f(x + \xi\delta_t, \xi, t + \delta_t) - f(x, \xi, t) = -\frac{1}{\tau} [f(x, \xi, t) - f^{EQ}(x, \xi, t)]$$



$$\because \tau = 1.0$$

$$f(t + \delta t) - f = -\frac{1}{\tau} [f - f^{EQ}]$$

$$\Rightarrow f(t + \delta t) = f - [f - f^{EQ}]$$

$$\Rightarrow f(t + \delta t) = f^{EQ}$$

$T=1.0$  indicates that distribution function goes to equilibrium within the current time step

# Hand Calculation Example

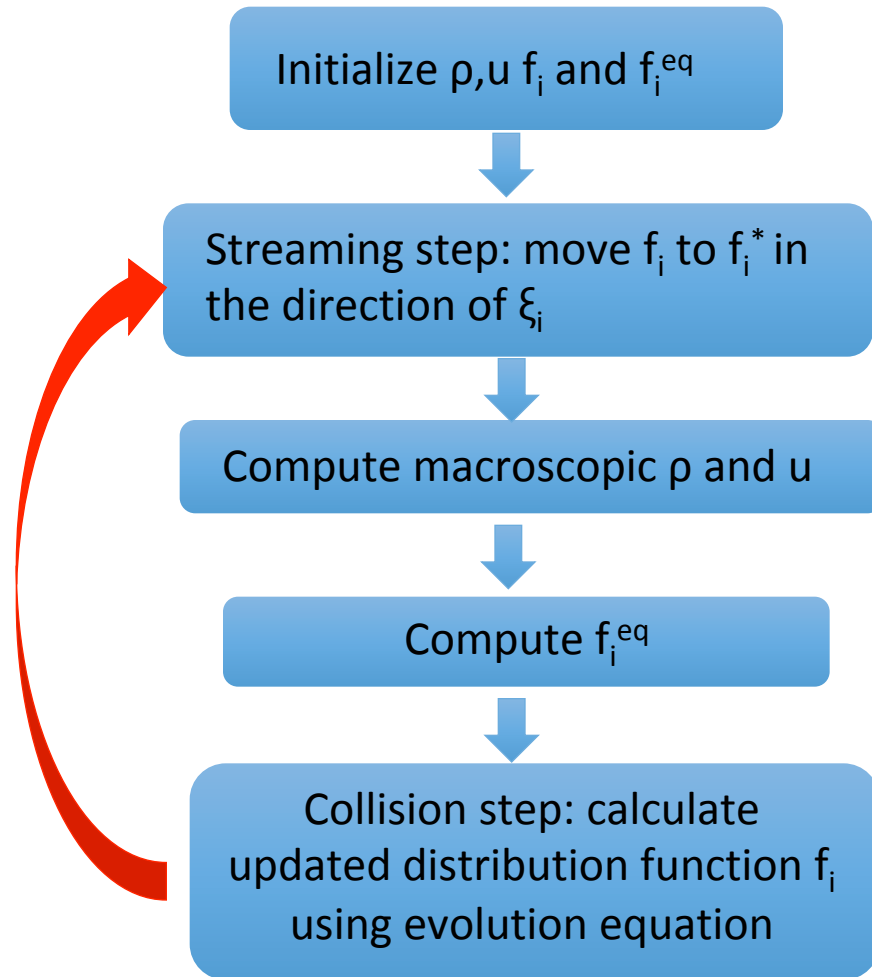
$$(f1^{EQ} \rightarrow)' = \begin{pmatrix} 0.01234568 & 0.012345679 \\ 0.01234568 & 0.012345679 \end{pmatrix} (f2^{EQ} \uparrow)' = \begin{pmatrix} 0.012345679 & 0.012345679 \\ 0.012345679 & 0.012345679 \end{pmatrix} (f3^{EQ} \leftarrow)' = \begin{pmatrix} 0.01234567 & 0.012345679 \\ 0.01234567 & 0.012345679 \end{pmatrix}$$

$$(f4^{EQ} \downarrow)' = \begin{pmatrix} 0.012345679 & 0.012345679 \\ 0.012345679 & 0.012345679 \end{pmatrix} (f5^{EQ} \searrow)' = \begin{pmatrix} 0.003086 & 0.003086 \\ 0.003086 & 0.003086 \end{pmatrix} (f6^{EQ} \nearrow)' = \begin{pmatrix} 0.003086 & 0.003086 \\ 0.003086 & 0.003086 \end{pmatrix}$$

$$(f7^{EQ} \swarrow)' = \begin{pmatrix} 0.003086 & 0.003086 \\ 0.003086 & 0.003086 \end{pmatrix} (f8^{EQ} \searrow)' = \begin{pmatrix} 0.003086 & 0.003086 \\ 0.003086 & 0.003086 \end{pmatrix} (f9^{EQ} \circ)' = \begin{pmatrix} 0.0493827 & 0.0493827 \\ 0.0493827 & 0.0493827 \end{pmatrix}$$



# Algorithm of LBM



$$\rho = \sum_{\alpha} f_{\alpha}$$

$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f(x + \xi \delta_t, \xi, t + \delta_t) = f(x, \xi, t)$$

$$\rho = \sum_{\alpha} f_{\alpha}$$

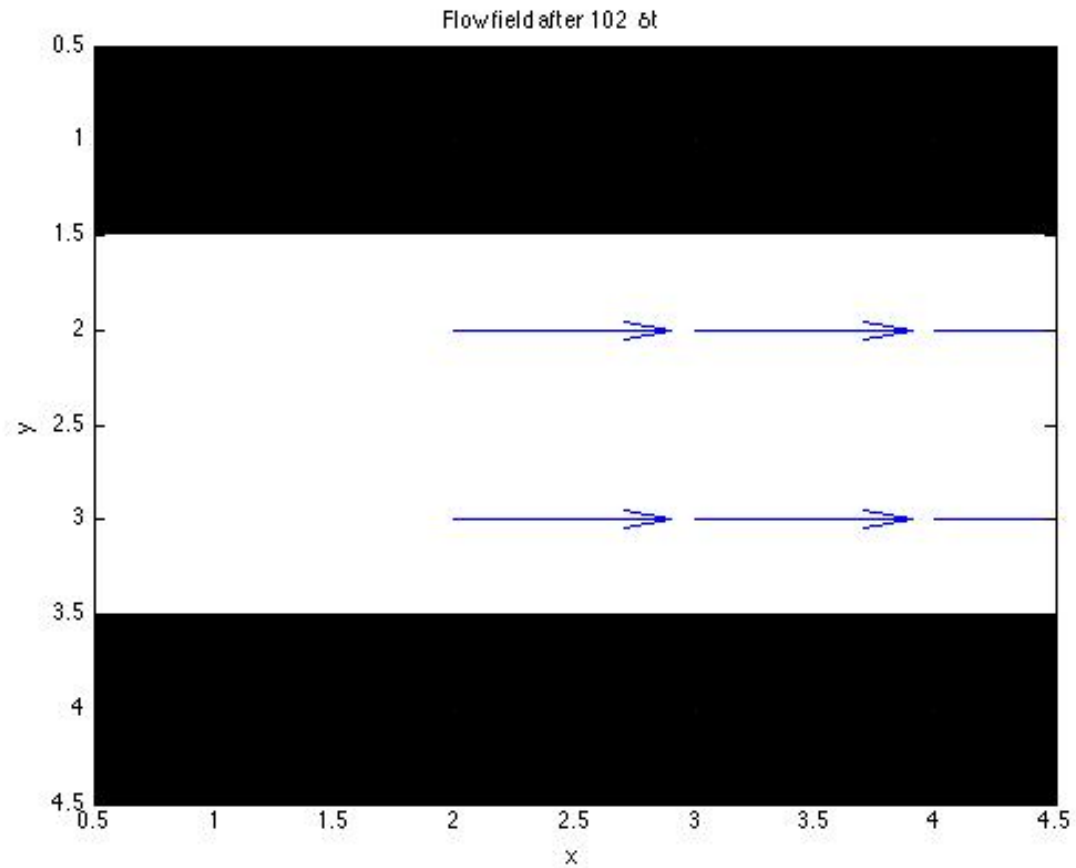
$$u = \frac{1}{\rho} \sum_{\alpha} \xi_{\alpha} f_{\alpha}$$

$$f^{EQ} = \rho w_{\alpha} \left[ 1 + \frac{\xi \cdot u}{RT} - \frac{u \cdot u}{2RT} + \frac{(\xi \cdot u)^2}{2(RT)^2} \right]$$

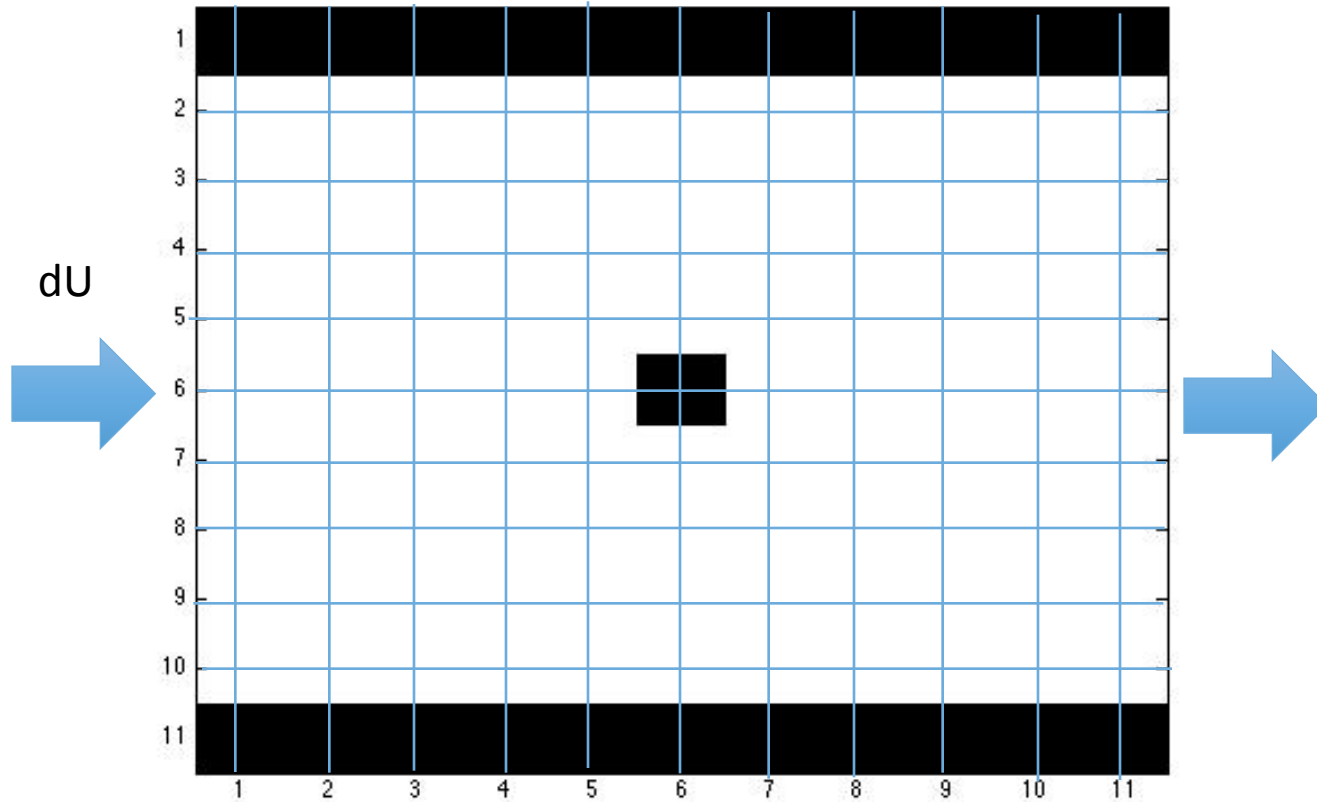
$$f(x + \xi \delta_t, \xi, t + \delta_t) - f(x, \xi, t) = -\frac{1}{\tau} [f(x, \xi, t) - f^{EQ}(x, \xi, t)]$$

# Hand Calculation Example

## Calculation results and visualization



# Problem Description



Steady Fluid Flow through a channel with a block in the middle

$$\tau = 1.0$$

$$\rho = 1.0$$

$$RT = 1/3$$

D2Q9 MODEL

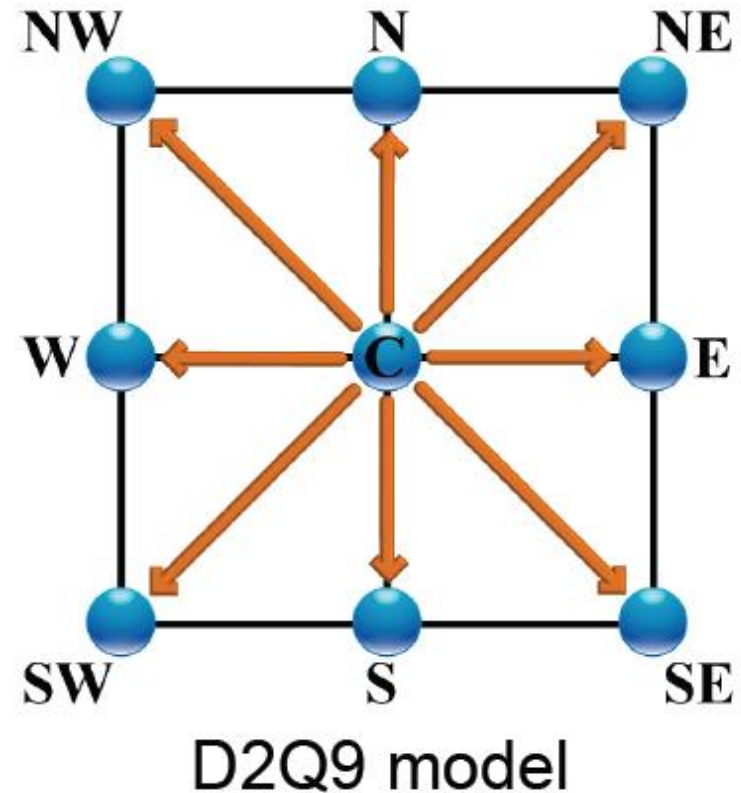
11×11 mesh

100 active lattice

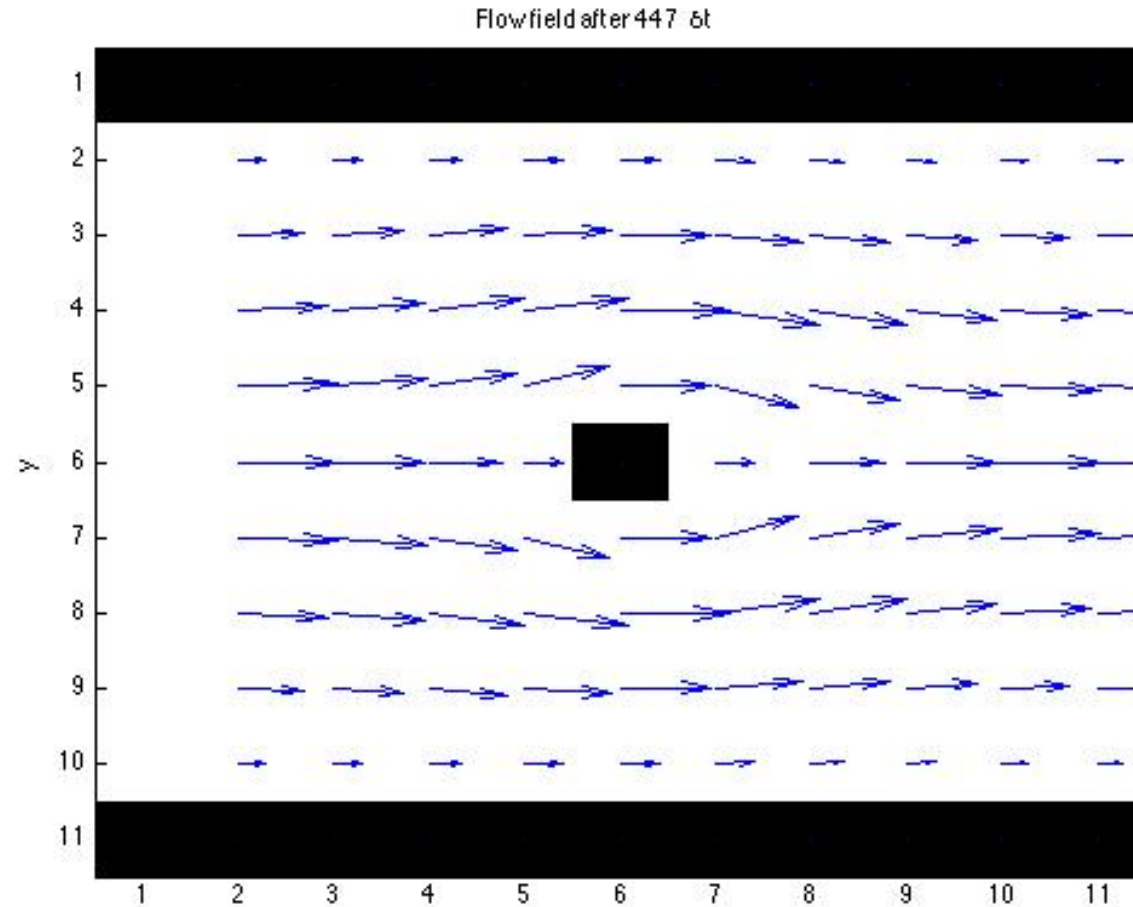
# MATLAB based calculation

Why?

- Every lattice contains a 9-dimensional matrix
- Lengthy calculation hard to present by hand
- In MATLAB multi-dimensional matrix can be easily presented



# MATLAB based calculation





# Numerical Example

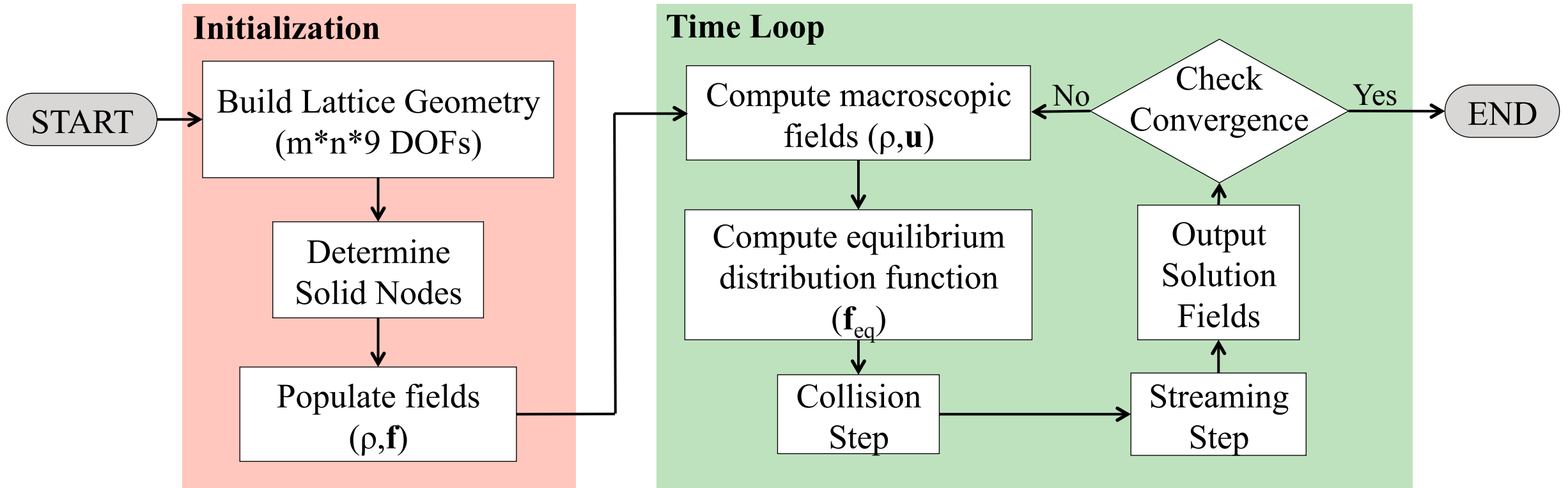
Implementation and Results

# Motivation

- Porous Media Flow
  - Discrete Simulation (NS) vs. Averaged Flow (Darcy)
  - Couple with transport/heat transfer
- Lattice-Boltzmann Methods
  - Incompressible Navier-Stokes
    - Water/Oil
  - Complex/Stochastic Geometries
    - Simple Meshing
  - Scalable
    - Large simulations

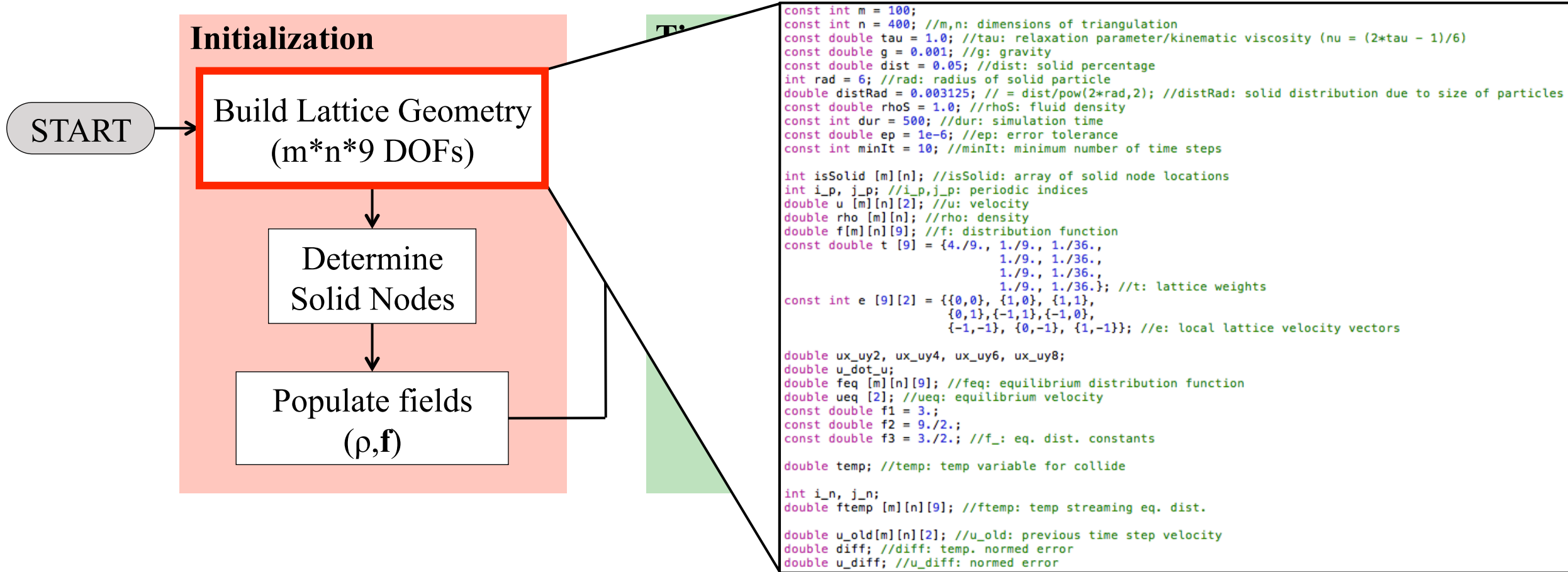


# Implementation



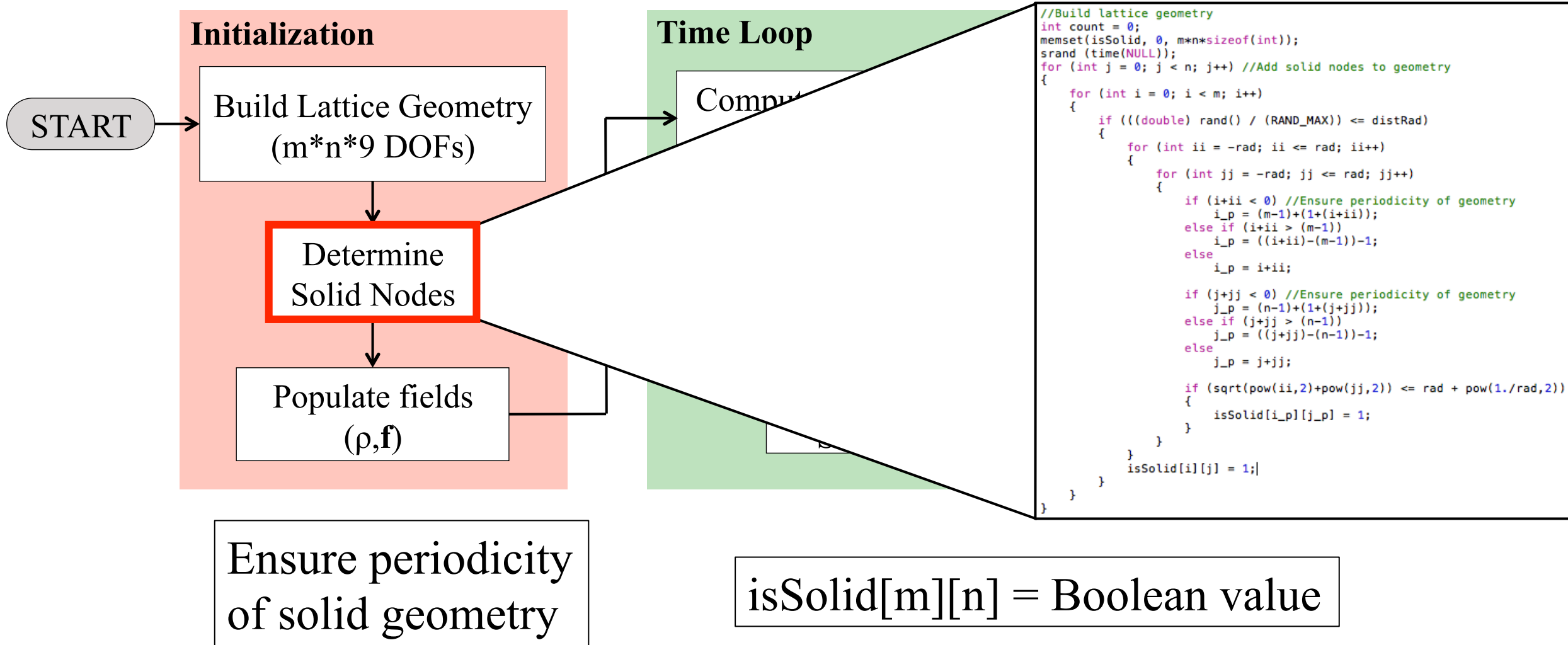


# Implementation

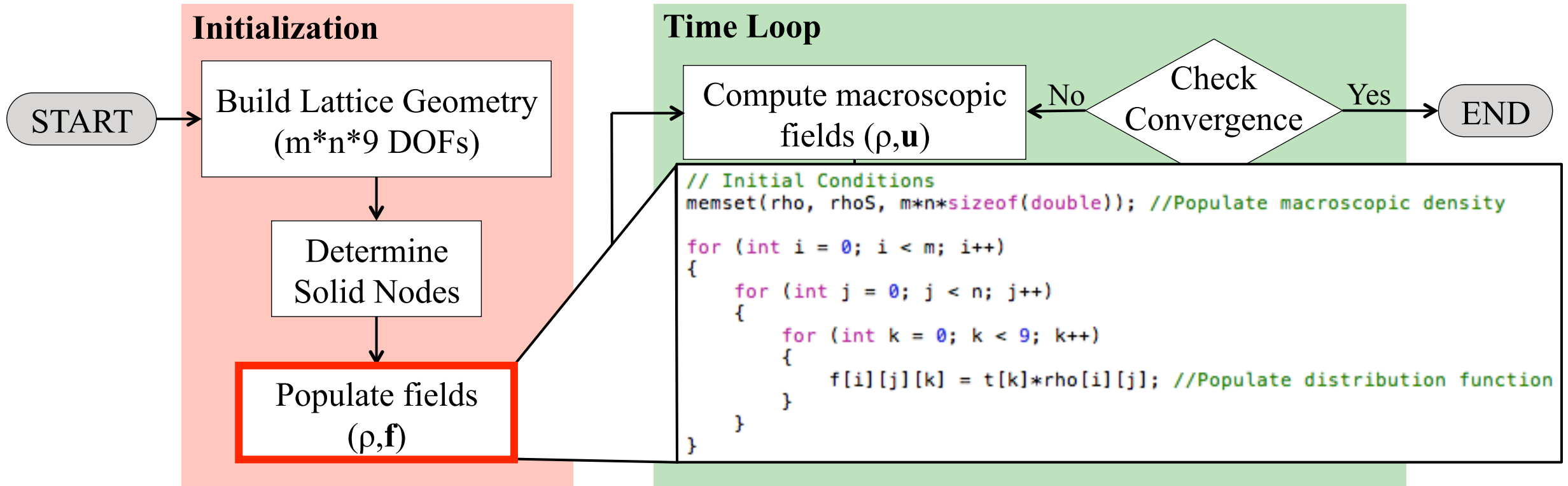


Initialize all fields and values for memory preallocation

# Implementation



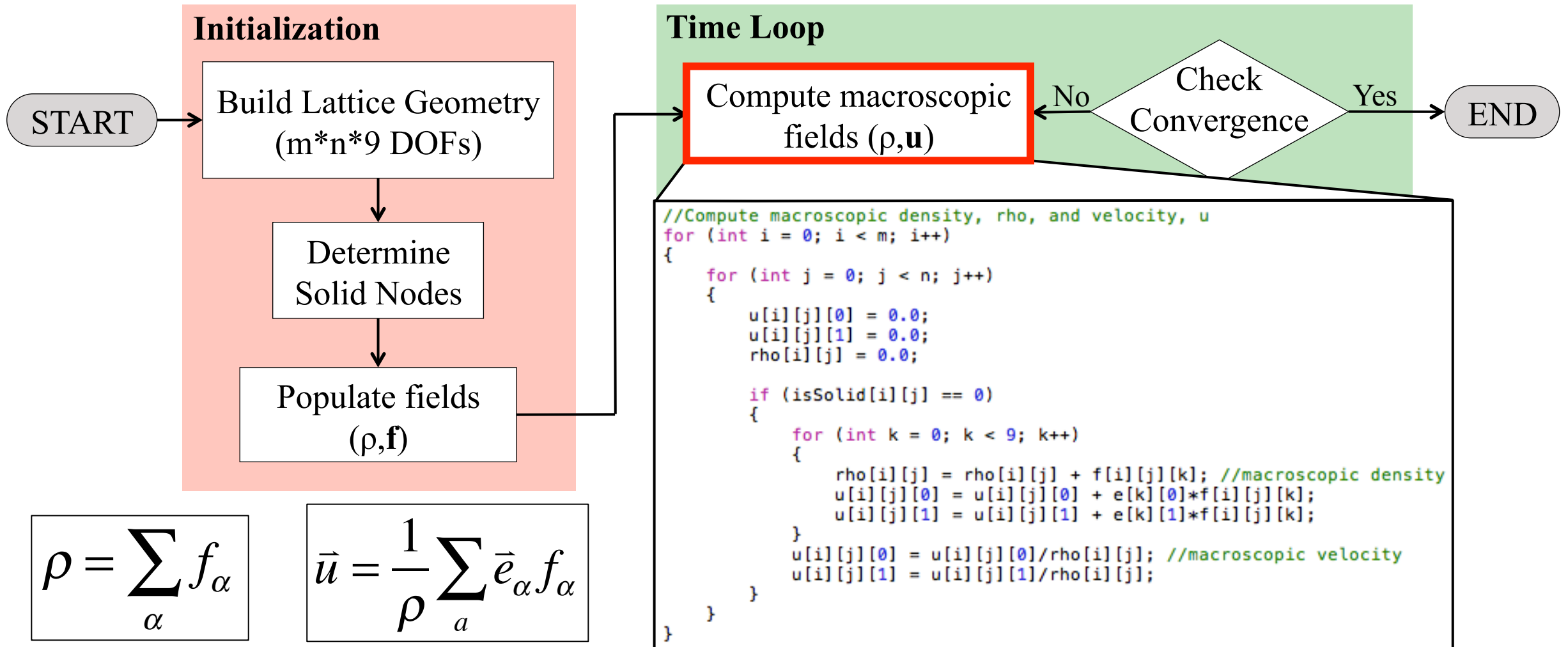
# Implementation



$\rho[m][n] = \text{constant and uniform density}$

$f[m][n][9] = \text{density} * \text{lattice weights}$

# Implementation



# Implementation

```

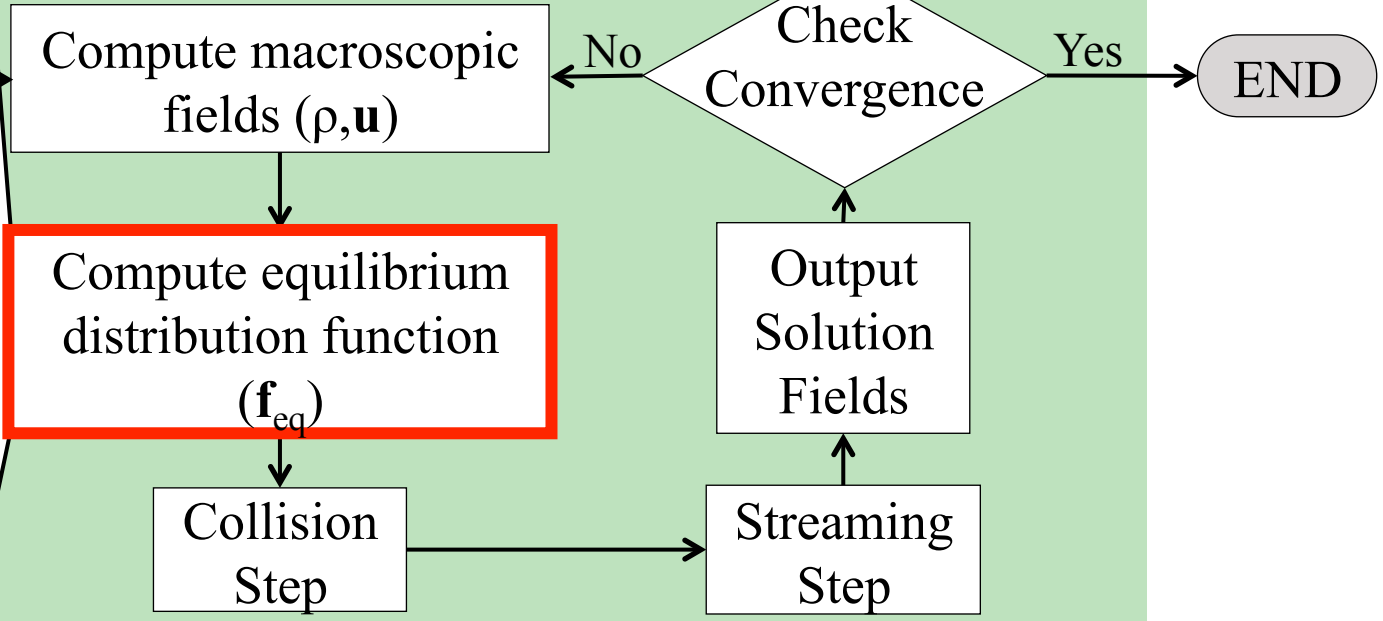
//Compute equilibrium distribution function
for (int i = 0; i < m; i++)
{
  for (int j = 0; j < n; j++)
  {
    ueq[0] = u[i][j][0];
    ueq[1] = u[i][j][1] - tau*g; //Add forcing term from gravity

    ux_uy2 = ueq[0] + ueq[1];
    ux_uy4 = -ueq[0] + ueq[1];
    ux_uy6 = -ueq[0] - ueq[1];
    ux_uy8 = ueq[0] - ueq[1];
    u_dot_u = pow(ueq[0],2) + pow(ueq[1],2);

    if (isSolid[i][j] == 0)
    {
      feq[i][j][0] = rho[i][j]*t[0]*(1 + 0 + 0 - f3*u_dot_u);
      feq[i][j][1] = rho[i][j]*t[1]*(1 + f1*ueq[0] + f2*pow(ueq[0],2) - f3*u_dot_u);
      feq[i][j][2] = rho[i][j]*t[2]*(1 + f1*ux_uy2 + f2*ux_uy2*ux_uy2 - f3*u_dot_u);
      feq[i][j][3] = rho[i][j]*t[3]*(1 + f1*ueq[1] + f2*pow(ueq[1],2) - f3*u_dot_u);
      feq[i][j][4] = rho[i][j]*t[4]*(1 + f1*ux_uy4 + f2*ux_uy4*ux_uy4 - f3*u_dot_u);
      feq[i][j][5] = rho[i][j]*t[5]*(1 - f1*ueq[0] + f2*pow(ueq[0],2) - f3*u_dot_u);
      feq[i][j][6] = rho[i][j]*t[6]*(1 + f1*ux_uy6 + f2*ux_uy6*ux_uy6 - f3*u_dot_u);
      feq[i][j][7] = rho[i][j]*t[7]*(1 - f1*ueq[1] + f2*pow(ueq[1],2) - f3*u_dot_u);
      feq[i][j][8] = rho[i][j]*t[8]*(1 + f1*ux_uy8 + f2*ux_uy8*ux_uy8 - f3*u_dot_u);
    }
  }
}
//Note here c = delta_x/delta_t = 1 and is not included
  
```

( $\rho, \mathbf{f}$ )

## Time Loop



Add gravity:  $\vec{u} = \vec{u} + \tau \vec{g}$

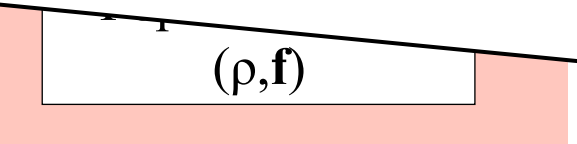
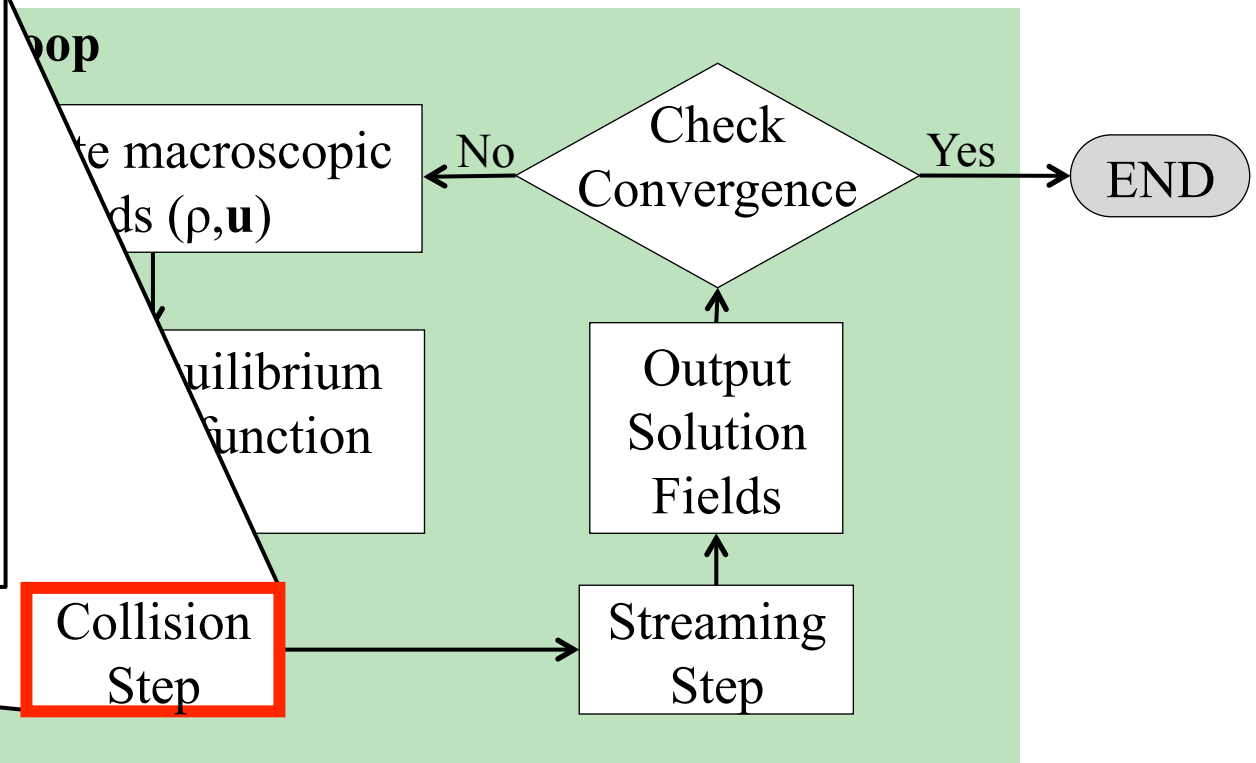
$$f_{eq} = w_{\alpha} \rho \left[ 1 + 3 \frac{\vec{e}_{\alpha} \cdot \vec{u}}{c^2} + \frac{9}{2} \frac{(\vec{e}_{\alpha} \cdot \vec{u})^2}{c^4} - \frac{3}{2} \frac{\vec{u}^2}{c^2} \right]$$

# Implementation

```

//Collision
for (int i = 0; i < m; i++)
{
  for (int j = 0; j < n; j++)
  {
    if (isSolid[i][j] == 1) //Bounceback from solid
    {
      temp = f[i][j][1]; f[i][j][1] = f[i][j][5]; f[i][j][5] = temp;
      temp = f[i][j][3]; f[i][j][3] = f[i][j][7]; f[i][j][7] = temp;
      temp = f[i][j][2]; f[i][j][2] = f[i][j][6]; f[i][j][6] = temp;
      temp = f[i][j][4]; f[i][j][4] = f[i][j][8]; f[i][j][8] = temp;
    }
    else //Collision between distributions
    {
      for (int ii = 0; ii < 9; ii++)
      {
        f[i][j][ii] = f[i][j][ii] - (1/tau)*(f[i][j][ii] - feq[i][j][ii]);
      }
    }
  }
}

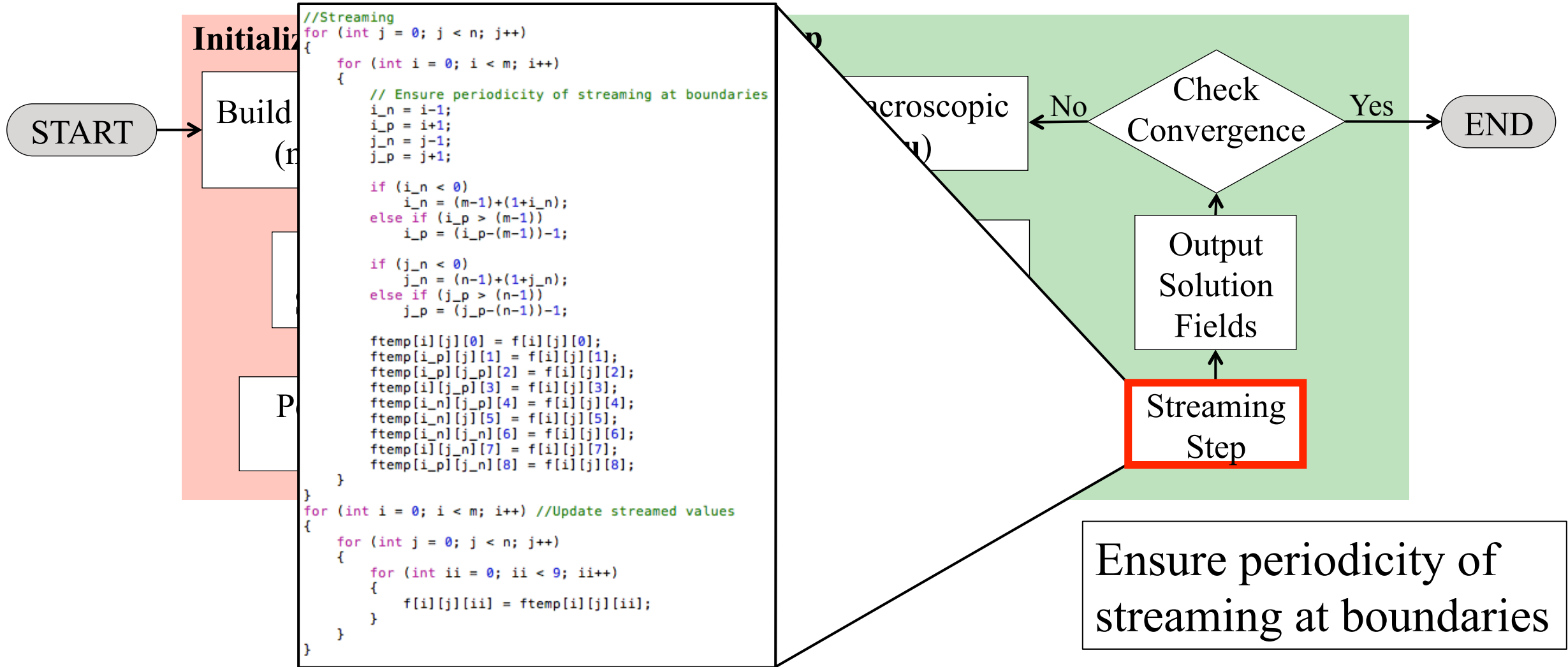
```



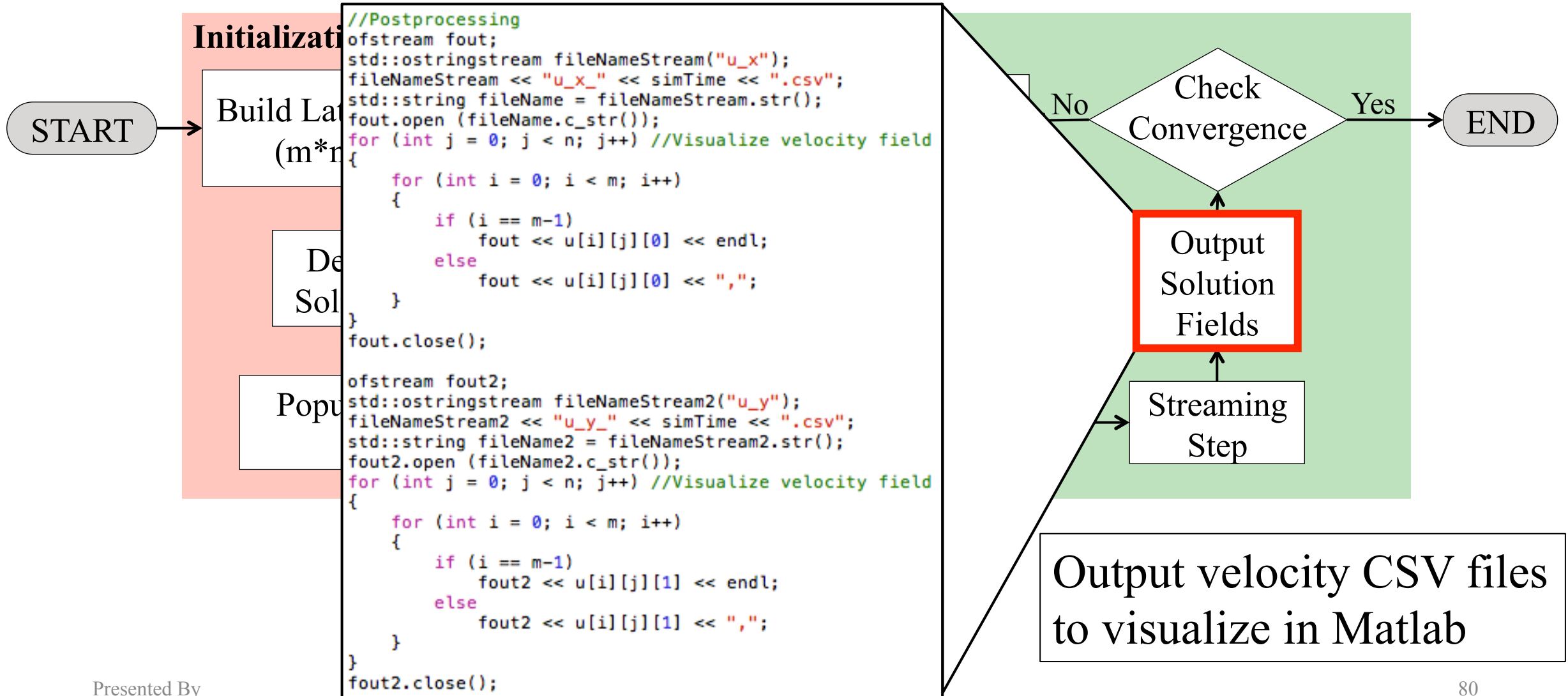
Bounceback condition on solid

$$f = f - \frac{1}{\tau} (f - f_{eq})$$

# Implementation

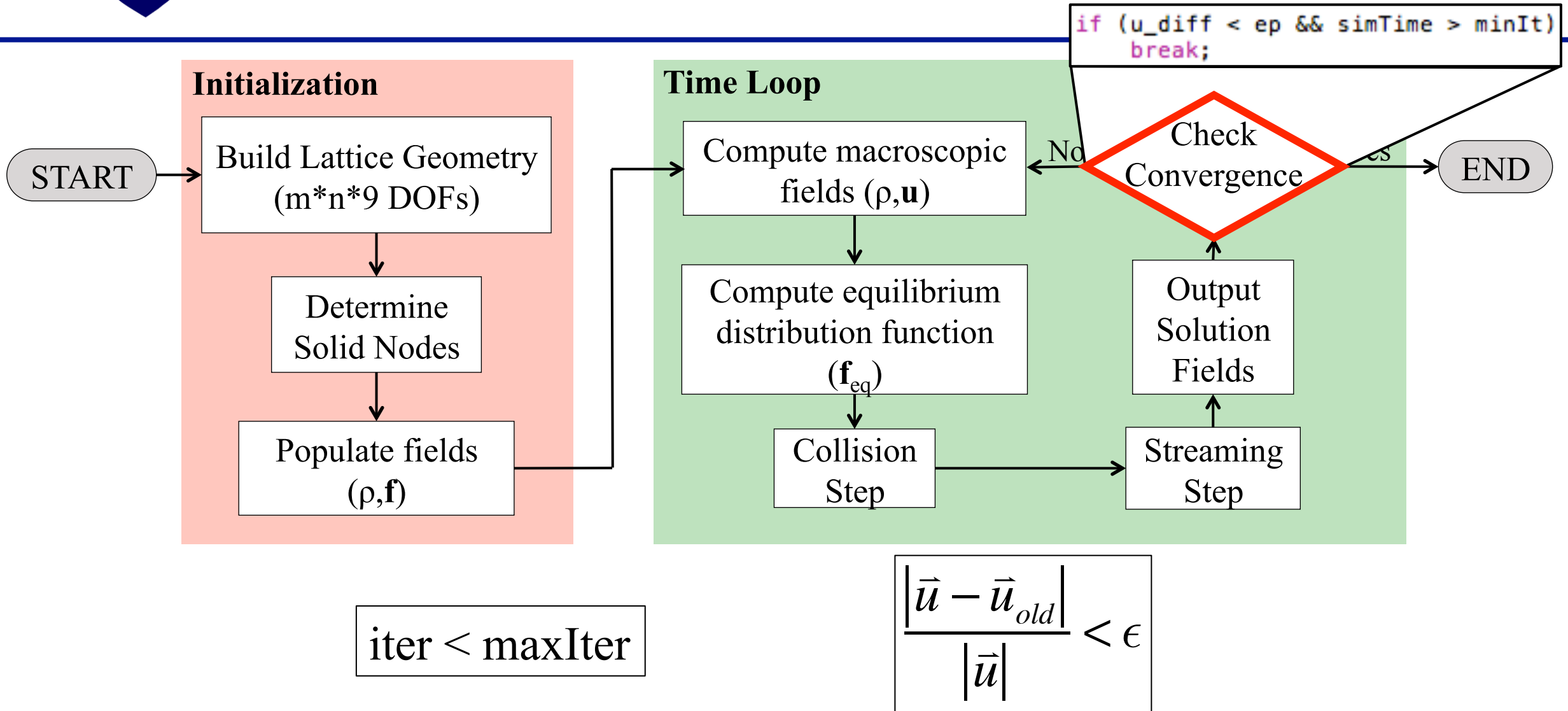


# Implementation

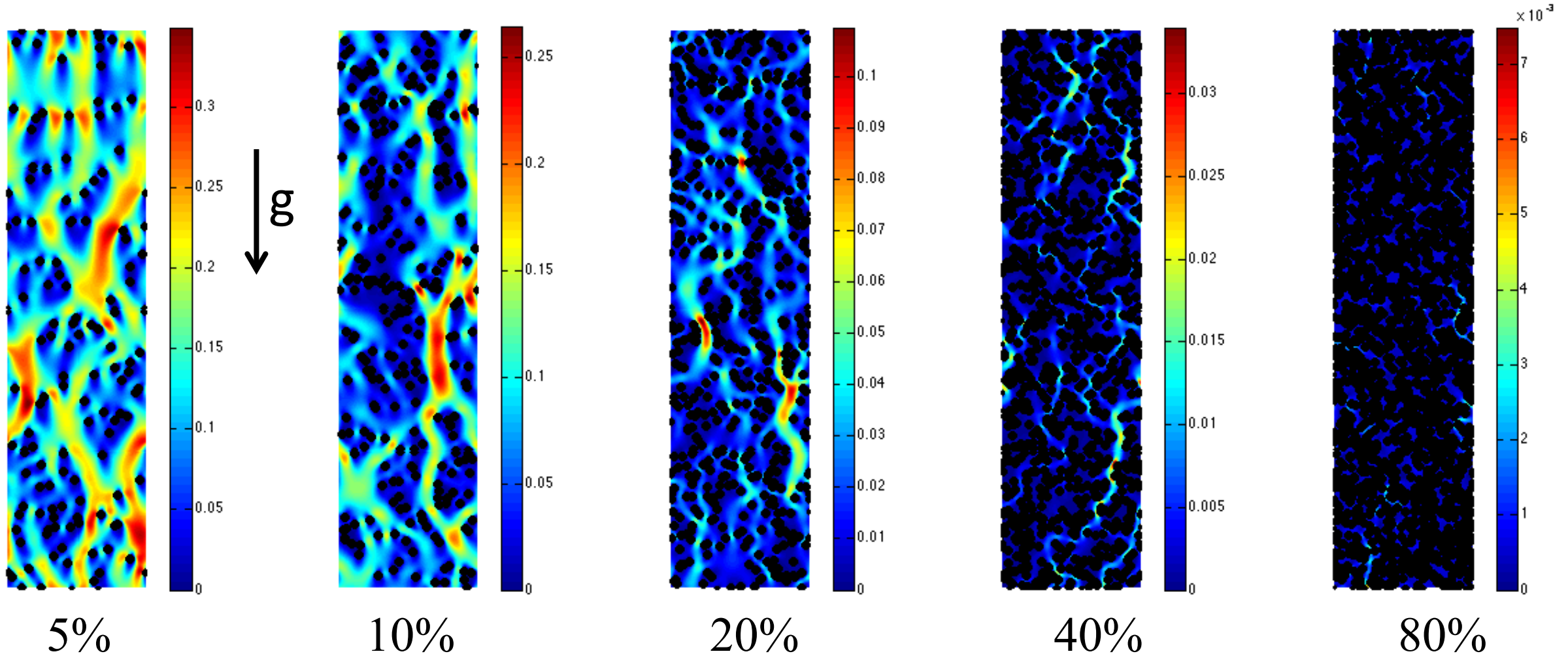




# Implementation



# Varying Grain Density ( $r = 2$ )



5%

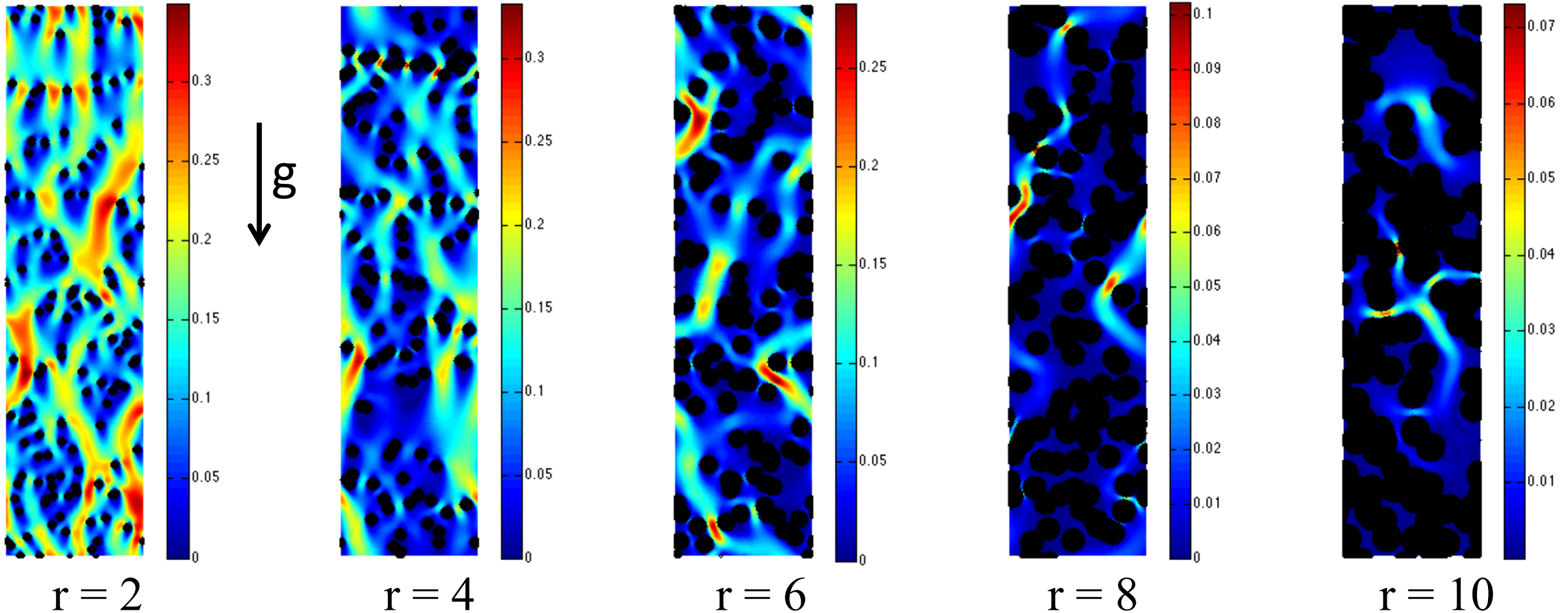
10%

20%

40%

80%

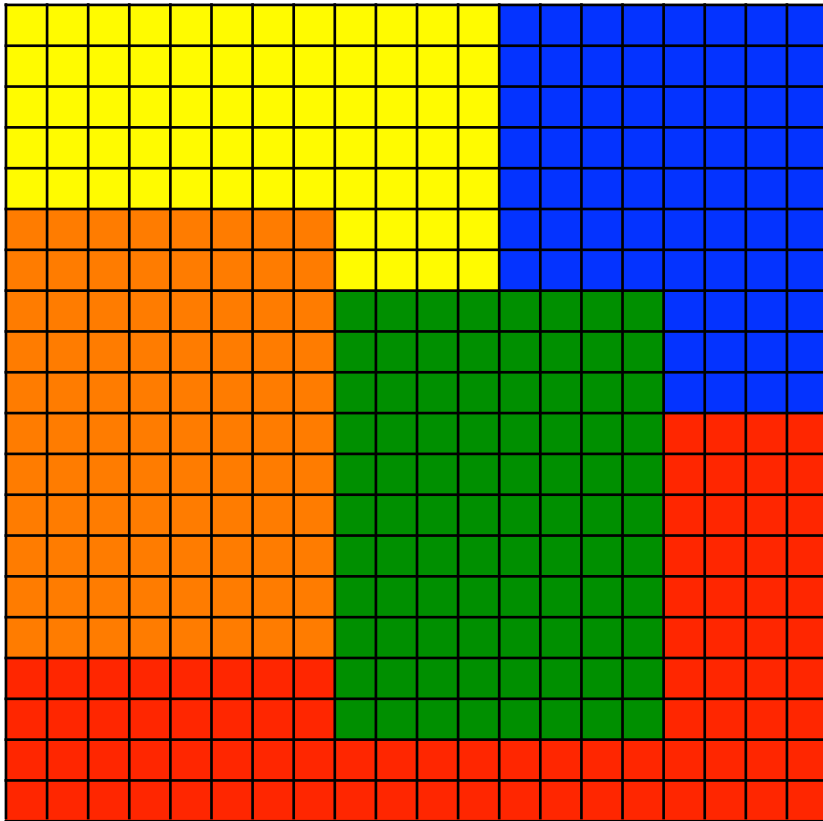
# Varying Grain Size (5%)



# Non-Uniform Porosity

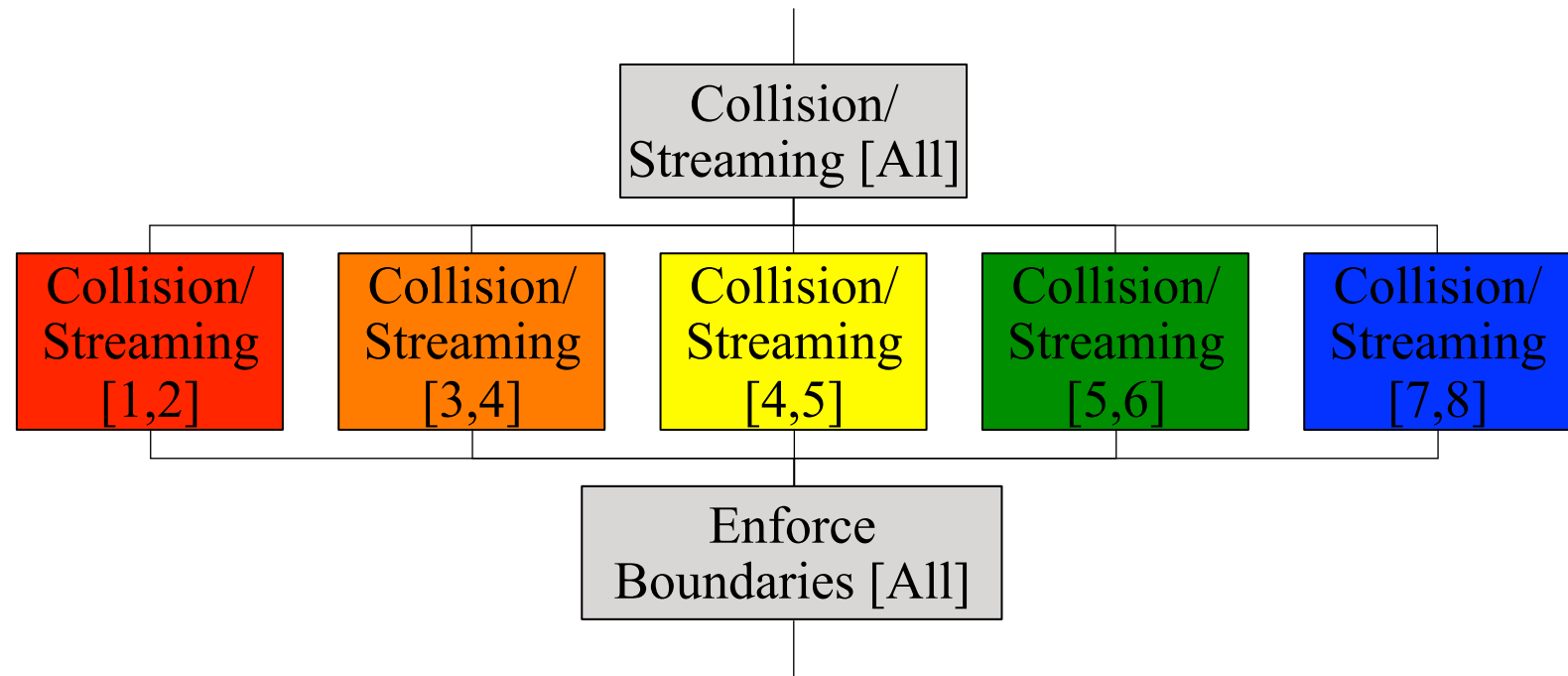


# Parallelism (Domain Decomposition)

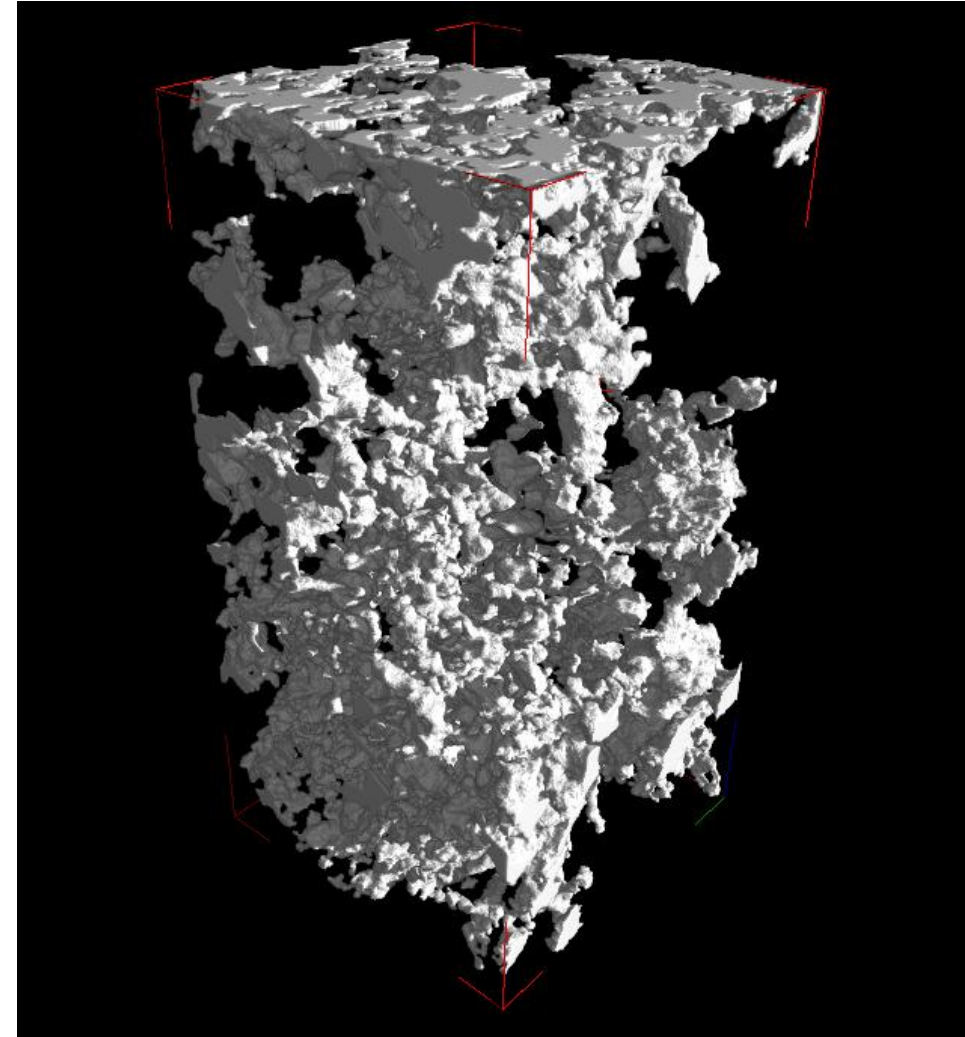
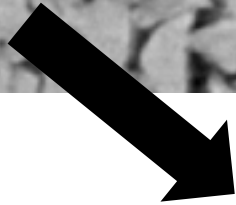
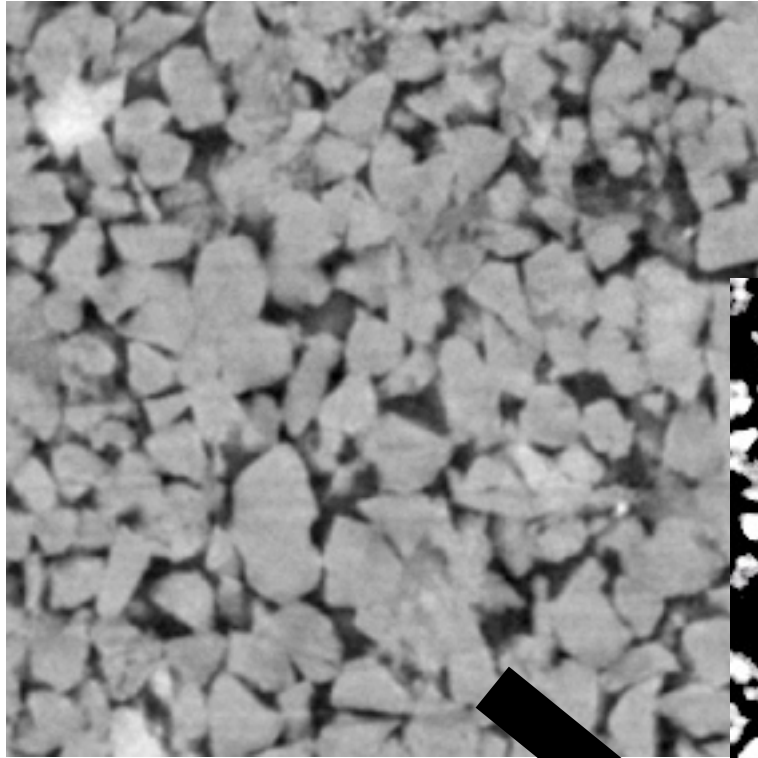


## Cluster

Proc 1	Proc 2	Proc 3	Proc 4	Proc 5	Proc 6	Proc 7	Proc 8
--------	--------	--------	--------	--------	--------	--------	--------



# Rock Sample Tomography





# Example Applications

Fancy stuff we can do with LBM

# Examples

- Air Conditioner
  - LBM simulates two conditions
    - 1. Fixed Fan Air Conditioner
    - 2. Sweeping Fan Air Conditioner
  - Parallelism
    - Extremely High Resolution Simulation Required
    - Parallel computations allow LBM to be extremely efficient with reasonable computer hardware
    - Allowed for the billion grid points required to accurately simulate this case

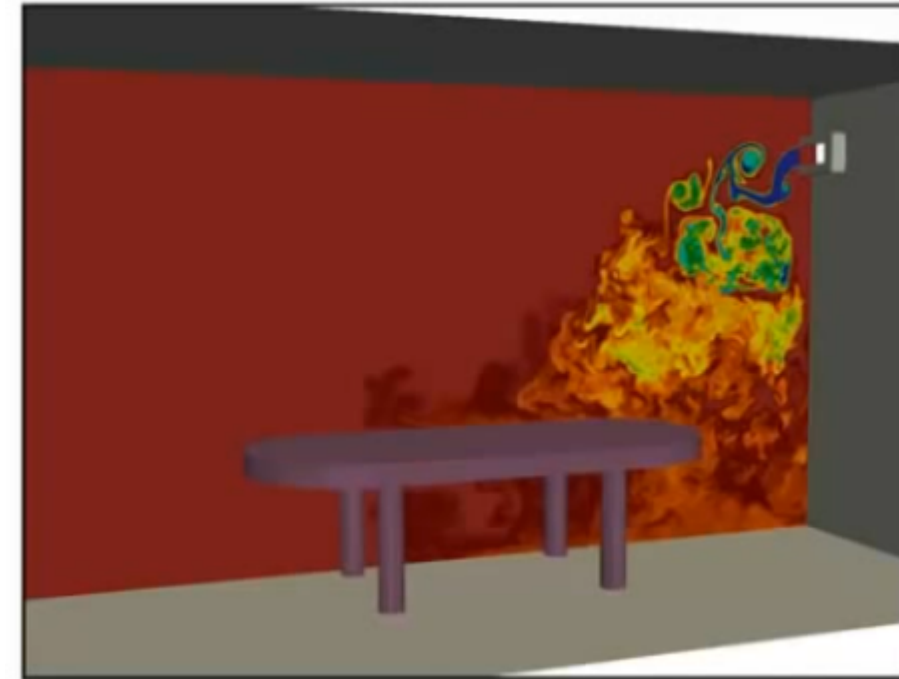


Image: <https://www.youtube.com/watch?v=I82uCa7SHSQ>

<http://youtu.be/I82uCa7SHSQ?t=21m20s>



# Examples

- Blood Clotting in a Human Artery
  - Arteries that have been affected by disease can be at high risk to rupture.
  - These ruptures can possibly be prevented by blood clotting in the vulnerable area
  - Need to simulate red blood cell changing from liquid to solid behavior and stick to artery wall.
    - LBM is effective at achieving this because of its hybrid particle/continuum nature



Image: <https://www.youtube.com/watch?v=l82uCa7SHSQ>

# Examples

- Blood Clotting in a Human Artery

- Parallelism

- Using LBM, it is very easy to send different calculations to different processors
    - Allows for high efficiency when using computers with a high number of processors

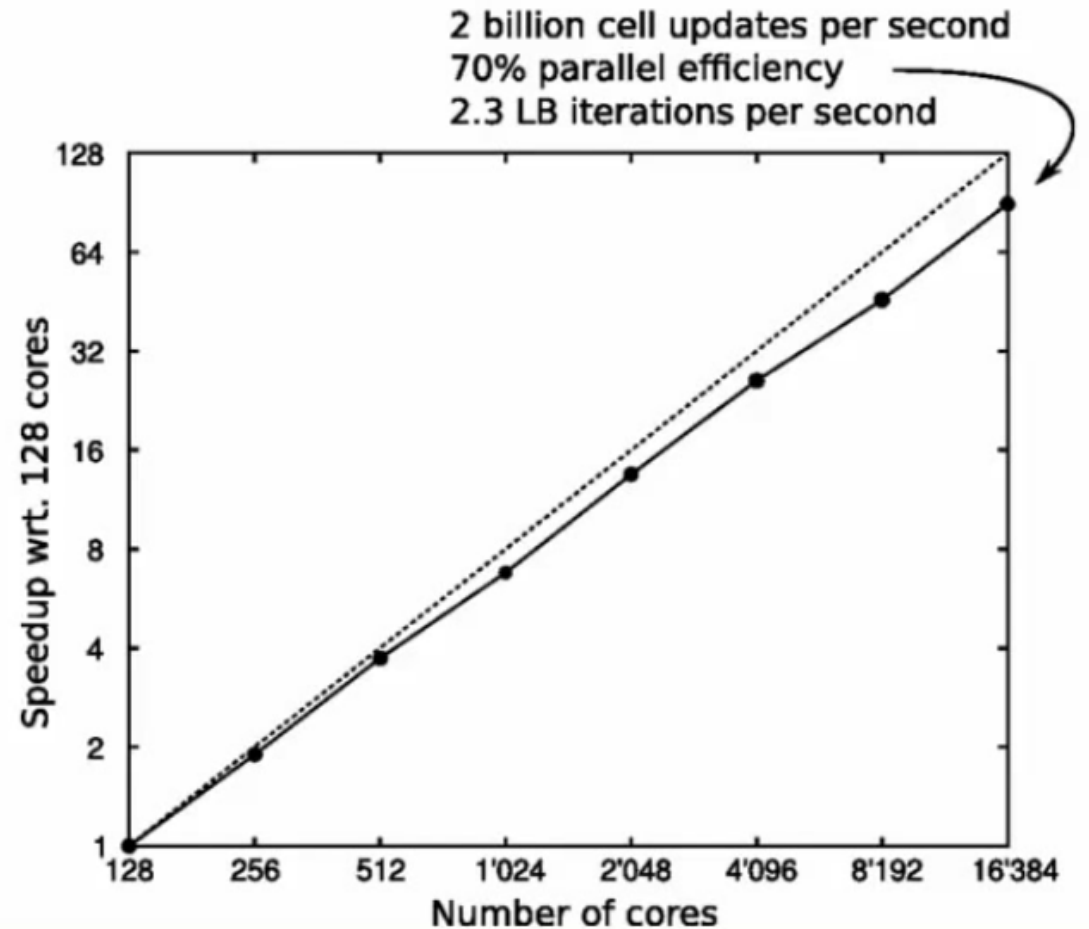


Image: <https://www.youtube.com/watch?v=I82uCa7SHSQ>

# Examples

- Turbulence Modeling
  - Models flow between two parallel plates
  - Large- eddy simulation approach
  - Replace Lattice Boltzmann equation with a filtered form
- Comparison to Spectral Method
  - Solution is virtually identical
  - LBM able to simulate with a 200x reduction in resolution
  - LBM much less computationally intensive in this case

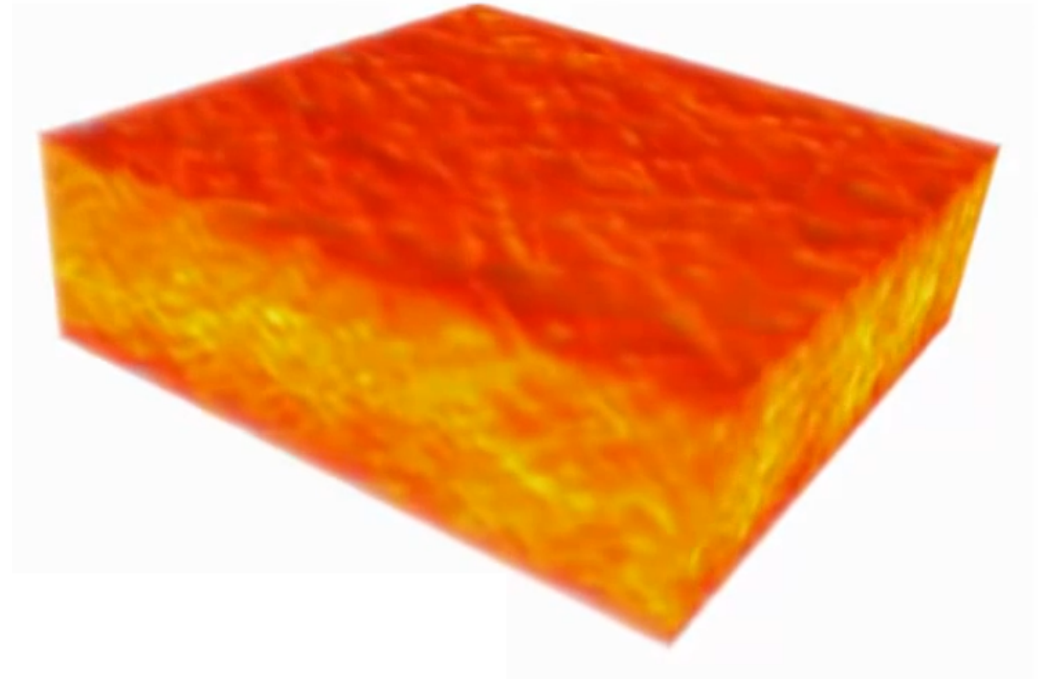


Image: <https://www.youtube.com/watch?v=l82uCa7SHSQ>

# References

- Latt, J. (2013). *Introduction to Lattice Boltzmann Method @ NasaGlenn 2013*
- [Video]. <https://www.youtube.com/watch?v=I82uCa7SHSQ>
- Mele, I. (2013). *Lattice Boltzmann Method*. Univerza v Ljubljani.
- Fathi, Ebrahim, and I. Yucel Akkutlu. "Lattice Boltzmann method for simulation of shale gas transport in kerogen." *SPE Journal* 18.01 (2012): 27-37.
- Engler, Simon T. "Benchmarking the 2D lattice Boltzmann BGK model." *Short communication. Amsterdam Center for Computational Science, Amsterdam, The Netherlands* (2003).
- Bao, Yuanxun Bill, and Justin Meskas. "Lattice Boltzmann Method for Fluid Simulations." (2011).
- Adhvaryu, Chinmay. "The Lattice Boltzmann Method For Computational Fluid Dynamics Applications." (2008).
- Xiaoyi He, Li-shi Luo. " Theory of the lattice Boltzmann method: From the Boltzmann equation to the lattice Boltzmann equation." (1997).



Thank You!

Questions?