Application of Lattice Boltzmann Method in Fluid Flow and Heat Transfer

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1. Introduction

1.1 Computational Fluid Dynamics (CFD) methods

In fluid dynamics, there are three levels to describe the motion of a fluid: microscopic level, mesoscopic level and macroscopic level [1]. On different levels, there exist the corresponding models to represent the fluid flow. No matter what kind of model used, eventually each of these models should satisfy the general conservation laws, i.e., conservation of mass, conservation of momentum and conservation of energy in the macroscopic world, in which the corresponding macroscopic variables (velocity, pressure and temperature) could be measured by using various kinds of sensor.

Fig. 1. Various approaches to describe fluid flow at different level

As shown in Figure 1 [2], the molecular dynamics (MD) method could be adopted to describe the fluid flow at the microscopic level. At this level, all the fluids are treated as the cluster of simple particles, such as molecule, atom and so on. All these particles are assumed to comply with the classical Newton's law of motion. Therefore, integrating the Newton's equations of motion for a set of molecules on the basis of an intermolecular potential, all the basic information (e.g., microscopic velocities) for each particle at the microscopic level could be obtained; After that if the coarse-grained procedure is introduced, all the macroscopic variables (e.g., macroscopic velocities, pressure and etc.) could be evaluated. For other models (such as dissipative particle dynamics and direct simulation Monte Carlo) at this level, they are all off-lattice pseudo-particle methods [3-8] in conjunction with Newtonian dynamics at microscopic level. In a word, at microscopic level, all the fluids are regarded as cluster of simple particles and this assumption is consistent with the reality.

At the second higher level (i.e., mesoscopic level), lattice gas and lattice Boltzmann methods treat flows in terms of coarse-grained fictitious particles which reside on a mesh and conduct translation as well as collision steps entailing overall fluid-like behavior. Therefore, it is safe to say that the fluid is still treated as a bunch of simple particles at mesoscopic level; and there is not any more particle motion equations, such as Newton's law of motion, involved except the special treatment has to be adopted for the collision procedure. At the meantime, the coarse-grained or statistical averaging process is introduced, and then the connections between microscopic and macroscopic variables (e.g., velocity, pressure and temperature) are developed.

The highest level as shown in Figure 1 is the macroscopic level. At this level, the fluid is treated as continuous medium, which means there is no space between the fluid particles and the entire domain are fully filled. Based on the continuous assumption and Newton's law of motion, the Navier-Stokes equations, which are the second order nonlinear partial differential equations (PDEs), are derived according to the general conservation laws (i.e., conservation of mass, momentum and energy).

Fig. 2. Top-down versus bottom-up model

According to the methods mentioned above, it is easy to say that all of these approaches, which are at different levels, belong to two different categories: bottom-up and top-down models [1], as shown in Figure 2. At macroscopic level, the Navier-Stokes equations associated with continuity equation are adopted to describe the fluid flow. For the specific flow problem, all the boundary conditions and the corresponding initial conditions if it

needs are usually given, after solving these nonlinear PDEs, all the macroscopic variables on the specific interesting regions are obtained. Since the general procedure to solve such complicated PDEs is to discretize these governing equations in the computational domain by different methods such as finite difference, finite element or spectral method, and then the corresponding algebraic equations are solved by iteration method for the macroscopic variables at each mesh point in the interested region. This approach, which is based on the macroscopic governing equations to obtain the specific discrete macroscopic variables, is called as top-down method. Although this top-down approach seems to be straightforward, it is not without any difficulties. In many books of the numerical solution of PDEs, the author put much emphasis on the truncation error which is due to the truncation of Taylor series when going from differential equations to finite difference equations. However, engineers are usually more concerned about whether certain quantities are conserved by the discretized form of the equations. The latter property is most important for integrations over long time scales in a closed domain like in the simulation of the world oceans. A small leakage would transform the ocean into an empty basin after a long time integration. In addition, numerical instability to solve the discretized algebraic equations is another problem of this type of numerical methods.

As compared to the top-down approach, the bottom-up method that includes all those models mentioned above (i.e., Molecular Dynamics (MD) method, Lattice Gas Cellular Automata (LGCA) and Lattice Boltzmann Equation (LBE) methods), are based on the fact that all the fluids are not perfectly continuous medium, but consist of simple particles, such as molecular, atom and so on. In MD approach, one tries to simulate macroscopic behavior of real fluids by setting up a model which describes the microscopic interactions for each particle as well as possible by using classical Newton's law of motion. This usually requires large amount of particles to describe a macroscopic behavior of the fluid flow. Unfortunately, the complexity of the interactions in MD restricts the number of particles and the time of integration that could be consumed in a simulation. As compared to MD, LGCA and LBE models are quite different variants of the bottom-up approach where the starting point is a discrete microscopic model which was developed by conserving the desired quantities (conservation of mass and momentum). These models are unconditionally stable (for LGCA) or exhibit good stability properties (for LBE). However, the derivation of the corresponding macroscopic equations (Navier-Stokes equation) requires lengthy calculations by multi-scale analysis. A major problem with the bottom-up approach is to detect and avoid spurious invariant which is also a problem for the models derived by the top-down approach.

2. Lattice Boltzmann Equation (LBE) model

2.1 Background of LBE model

The LBE model historically is originated from a Boolean fluid model known as the Lattice Gas Cellular Automata (LGCA), which simulates the motion of fluids by particles moving and colliding on a regular lattice. The averaged fluid variables, such as mass density and velocity, are shown to satisfy the Navier-Stokes equations. It has been proved that the LBE could be derived from the continuous Boltzmann equation directly as a discrete equation, in which velocity space is discretized with the minimum set of values. The latter derivation implies a compatibility of the LBE with the Boltzmann equation. In other words, the simulation of fluid flows by using the LBE method is based on kinetic equations and statistical physics, unlike those of conventional methods which are based on continuum mechanics.

The theoretical premises of the LBE method are as follows: (1) hydrodynamics are insensitive to the details of microscopic physics, and (2) hydrodynamics could be preserved as long as the conservation laws and associated symmetries of lattice are respected locally at the microscopic or mesoscopic level. Therefore, the computational advantage of the LBE method is attained by drastically reducing the particle velocity space to only a few discrete directions without seriously degrading hydrodynamics. At the same time, the interactions between particles are restricted locally (which means the LBE model has authentically parallelism characteristic). This is possible because the LBE method rigorously preserves the hydrodynamic moments of the distribution function, such as mass density, momentum fluxes, and the necessary lattice symmetries [9-11].

The LBE method eliminates the time consuming statistical averaging step in the original LGCA due to its kinetic nature. Therefore, the LBE method is able to simulate the complicated fluid flows such as multiphase flows, chemically reacting flows, visco-elastic non-Newtonian flows. In addition, simplified collision models are developed and used to replace the collision operator derived from the LGCA to improve both the computational efficiency and accuracy. It is worth noting that the simple collision model of Bhatnagar., Gross., and Krook (BGK) was applied to the lattice Boltzmann equation and yielded the so-called lattice BGK model. Since the extra flexibility in this approach is to allow the removal of the artifacts of the LGCA (i.e., the lack of Galilean invariance and the dependence between velocity and pressure in LGCA), eventually this method was numerically found to be at least as stable, accurate, and computationally efficient as traditional CFD methods for simulation of simple single-phase incompressible fluid flows. More importantly, the microscopic physics of the fluid particles could be incorporated as easily as in other particle collision methods since fluid motion is simulated at the level of distribution functions. Due to interparticle interactions such as capillary phenomena, multiphase flows and nonlinear diffusion, many complex fluid phenomena could be simulated naturally by LBE model.

In conclusion, from a computational point of view, the notable advantages of LBE method are parallelism of algorithm (local collision), simplicity of programming (only collision and streaming steps), ease of incorporating microscopic interactions and the simplicity of modeling complex geometry flow problems.

2.2 From Boltzmann equation to LBE model

The Boltzmann equation, devised by Ludwig Boltzmann, describes the statistical distribution of particles in a fluid. It is one of the most important equations for non-equilibrium statistical mechanics that deals with systems far from thermodynamic equilibrium. For example, if a temperature gradient or electric field is applied on a system, the Boltzmann equation is used to study how a fluid transports physical quantities such as heat or charge, and to derive the transport properties such as thermal conductivity and electrical conductivity.

It is well known that the Boltzmann equation is an integro-differential equation for the single particle distribution function $f(\mathbf{x}, \mathbf{v}, t)$:

$$\partial_t f + \mathbf{v} \cdot \partial_{\mathbf{x}} f + \frac{\mathbf{K}}{m} \cdot \partial_{\mathbf{v}} f = Q(f, f) \quad (1)$$

where, $\mathbf{v}$ is the particle velocity, $\mathbf{K}$ is the external body force, $f(\mathbf{x}, \mathbf{v}, t) d^3 \mathbf{x} d^3 \mathbf{v}$ is the probability to find a particle in the volume $d^3 \mathbf{x}$ around $\mathbf{x}$ and with velocity between $\mathbf{v}$ and

$\mathbf{v} + d\mathbf{v}$, and the collision integral with $\sigma(\Omega)$ the differential collision cross section for the two-particle collision which transforms the velocities from $\{\mathbf{v}, \mathbf{v}_1\}$ (incoming) into $\{\mathbf{v}', \mathbf{v}_1'\}$ (outgoing) is as follows.

$$Q(f, f) = \int d^3\mathbf{v}_1 \int d\Omega \sigma(\Omega) |\mathbf{v} - \mathbf{v}_1| \left[f(\mathbf{v}') f(\mathbf{v}_1') - f(\mathbf{v}) f(\mathbf{v}_1) \right] \quad (2)$$

In order to solve Eq.1, one can start with arbitrary velocity distribution and make this distribution evolve according to the Boltzmann equation; eventually it will relax to equilibrium Maxwell distribution or Maxwell-Boltzmann distribution

$$f^{(M)} = f(\mathbf{x}, \mathbf{v}) = \rho_0 \left(\frac{m}{2\pi k_B T} \right)^{D/2} \exp \left[-\frac{m}{2k_B T} (\mathbf{v} - \mathbf{u})^2 \right] \quad (3)$$

where, ρ_0 is the particle density, m particle mass, D dimension (for example, $D=2$ for 2D case and $D=3$ for 3D case), $\mathbf{v}$ microscopic particle speed, $\mathbf{u}$ macroscopic fluid velocity, k_B Boltzmann constant, T temperature. This equilibrium distribution is guaranteed by the famous Boltzmann H-theorem.

One of the major problems of dealing with Boltzmann equation is the complicated nature of the collision integral. It is therefore not surprising that alternative, simpler expressions have been developed. The idea behind this replacement is that the large amount of detail of two-body interactions is not likely to significantly influence the values of many experimentally measured quantities. However, the simpler operator $\mathcal{J}(f)$ which replaces the collision operator $Q(f, f)$ should satisfy two constraints [1]:

1. $\mathcal{J}(f)$ conserves the five collision invariants ψ_k of $Q(f, f)$, that is

$$\int \psi_k \mathcal{J}(f) d^3\mathbf{x} d^3\mathbf{v} = 0 \quad (k=0, 1, 2, 3, 4) \quad (4)$$

where, $\psi_0 = 1$ for conservation of mass, $\psi_{1,2,3} = v$ for conservation of momentum and $\psi_4 = v^2$ for conservation of energy.

2. The collision term has the capability to lead the velocity distribution toward a Maxwellian distribution (H-theorem)

Both constraints are fulfilled by the most widely known model usually called the BGK approximation. The simplest way to take the second constraint into account is to imagine that each collision changes the distribution function $f(\mathbf{x}, \mathbf{v}, t)$ by an amount proportional to the departure of $f(\mathbf{x}, \mathbf{v}, t)$ from a Maxwellian $f^M(\mathbf{x}, \mathbf{v})$:

$$\mathcal{J}(f) = \omega \left[f^M(\mathbf{x}, \mathbf{v}) - f(\mathbf{x}, \mathbf{v}, t) \right] = -\frac{1}{\tau} \left(f(\mathbf{x}, \mathbf{v}, t) - f^M(\mathbf{x}, \mathbf{v}) \right) \quad (5)$$

The coefficient ω is called the collision frequency and τ is called the collision time or relaxation time. From the first constraint it follows that

$$\int \psi_k \mathcal{J}(f) d^3\mathbf{x} d^3\mathbf{v} = \omega \left(\int \psi_k f^M(\mathbf{x}, \mathbf{v}) d^3\mathbf{x} d^3\mathbf{v} - \int \psi_k f(\mathbf{x}, \mathbf{v}, t) d^3\mathbf{x} d^3\mathbf{v} \right) = 0 \quad (6)$$

At any space point and time instant, the Maxwellian $f^M(\mathbf{x}, \mathbf{v})$ must have exactly the same density, velocity and temperature of the fluid as given by the distribution $f(\mathbf{x}, \mathbf{v}, t)$. Since

these values will, in general, vary with space and time, $f^M(\mathbf{x}, \mathbf{v})$ is called the local Maxwellian distribution function.

Therefore, if the external body force is neglected and the BGK model is used to replace the collision integral, the following Boltzmann equation with BGK approximation can be obtained:

$$\frac{\partial f}{\partial t} + \mathbf{v} \cdot \nabla f = -\frac{1}{\tau} (f - f^M)$$

or

$$\frac{\partial f}{\partial t} + \mathbf{v} \cdot \nabla f = -\frac{1}{\tau} (f - f^{(eq)}) \quad (7)$$

where, $f^{(eq)}$ is the local equilibrium distribution or local Maxwellian distribution.

Based on the above equation, it could be seen that the mass density distribution function, $f(\mathbf{x}, \mathbf{v}, t)$, depends on the space, velocity and time. At the same time, the $\mathbf{v}$ -space could be discretized by introducing a finite set of velocities, $\mathbf{v}_i$, and associated distribution functions, $f_i(\mathbf{x}, t)$, which are governed by the discrete Boltzmann equation:

$$\frac{\partial f_i}{\partial t} + \mathbf{v}_i \cdot \nabla f_i = -\frac{1}{\tau} (f_i - f_i^{(eq)}) \quad (8)$$

Introducing the reference parameters such as characteristic length scale, L , the reference speed, U , the reference density, n_r , and the time interval between particle collisions, t_c , the following discrete dimensionless Boltzmann equation could be obtained:

$$\frac{\partial F_i}{\partial \hat{t}} + \mathbf{c}_i \cdot \hat{\nabla} F_i = -\frac{1}{\hat{\tau} \cdot \varepsilon} (F_i - F_i^{(eq)}) \quad (9)$$

where, $\mathbf{c}_i = \frac{\mathbf{v}_i}{U}$, $\hat{\nabla} = L \nabla$, $\hat{t} = \frac{t \cdot U}{L}$, $\hat{\tau} = \frac{\tau}{t_c}$, $F_i = \frac{f_i}{n_r}$ and $\varepsilon = t_c \frac{U}{L}$.

The parameter ε could be interpreted as either the ratio of collision time to flow time or as the ratio of mean free path to the characteristic length (i.e., Knudsen number).

A discretization of the above equation is given by:

$$\begin{aligned} & \frac{F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta \hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t})}{\Delta \hat{t}} \\ & + c_{ix} \frac{F_i(\hat{x} + \Delta \hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta \hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta \hat{t})}{\Delta \hat{x}} \\ & + c_{iy} \frac{F_i(\hat{x}, \hat{y} + \Delta \hat{y}, \hat{z}, \hat{t} + \Delta \hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta \hat{t})}{\Delta \hat{y}} \\ & + c_{iz} \frac{F_i(\hat{x}, \hat{y}, \hat{z} + \Delta \hat{z}, \hat{t} + \Delta \hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta \hat{t})}{\Delta \hat{z}} = -\frac{1}{\hat{\tau} \cdot \varepsilon} (F_i - F_i^{(eq)}) \end{aligned} \quad (10)$$

where, $\Delta \hat{t} = \frac{\Delta t \cdot U}{L}$. And then, Lagrangian behavior could be obtained by the selection of the

corresponding lattice spacing ($\Delta\hat{x}$, $\Delta\hat{y}$ and $\Delta\hat{z}$) divided by the time step to equal the lattice velocity ($c_{ix} = \frac{\Delta\hat{x}}{\Delta\hat{t}}$, $c_{iy} = \frac{\Delta\hat{y}}{\Delta\hat{t}}$ and $c_{iz} = \frac{\Delta\hat{z}}{\Delta\hat{t}}$ or $c_i = \frac{\Delta\hat{x}}{\Delta\hat{t}}$).

Therefore,

$$\begin{aligned} & \frac{F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta\hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t})}{\Delta\hat{t}} \\ & + \frac{F_i(\hat{x} + c_{ix} \cdot \Delta\hat{t}, \hat{y}, \hat{z}, \hat{t} + \Delta\hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta\hat{t})}{\Delta\hat{t}} \\ & + \frac{F_i(\hat{x}, \hat{y} + c_{iy} \cdot \Delta\hat{t}, \hat{z}, \hat{t} + \Delta\hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta\hat{t})}{\Delta\hat{t}} \\ & + \frac{F_i(\hat{x}, \hat{y}, \hat{z} + c_{iz} \cdot \Delta\hat{t}, \hat{t} + \Delta\hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta\hat{t})}{\Delta\hat{t}} = -\frac{1}{\hat{\tau} \cdot \varepsilon} (F_i - F_i^{(eq)}) \end{aligned} \quad (11)$$

Since,

$$\begin{aligned} & \frac{F_i(\hat{x} + c_{ix} \cdot \Delta\hat{t}, \hat{y}, \hat{z}, \hat{t} + \Delta\hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta\hat{t})}{\Delta\hat{t}} \\ & + \frac{F_i(\hat{x}, \hat{y} + c_{iy} \cdot \Delta\hat{t}, \hat{z}, \hat{t} + \Delta\hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta\hat{t})}{\Delta\hat{t}} \\ & + \frac{F_i(\hat{x}, \hat{y}, \hat{z} + c_{iz} \cdot \Delta\hat{t}, \hat{t} + \Delta\hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta\hat{t})}{\Delta\hat{t}} \\ & = \frac{F_i(\hat{x} + c_{ix} \cdot \Delta\hat{t}, \hat{y} + c_{iy} \cdot \Delta\hat{t}, \hat{z} + c_{iz} \cdot \Delta\hat{t}, \hat{t} + \Delta\hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta\hat{t})}{\Delta\hat{t}} \end{aligned} \quad (12)$$

Then,

$$\begin{aligned} & \frac{F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta\hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t})}{\Delta\hat{t}} \\ & + \frac{F_i(\hat{x} + c_{ix} \cdot \Delta\hat{t}, \hat{y} + c_{iy} \cdot \Delta\hat{t}, \hat{z} + c_{iz} \cdot \Delta\hat{t}, \hat{t} + \Delta\hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t} + \Delta\hat{t})}{\Delta\hat{t}} \\ & = \frac{F_i(\hat{x} + c_{ix} \cdot \Delta\hat{t}, \hat{y} + c_{iy} \cdot \Delta\hat{t}, \hat{z} + c_{iz} \cdot \Delta\hat{t}, \hat{t} + \Delta\hat{t}) - F_i(\hat{x}, \hat{y}, \hat{z}, \hat{t})}{\Delta\hat{t}} \\ & = \frac{F_i(\hat{x} + c_i \cdot \Delta\hat{t}, \hat{t} + \Delta\hat{t}) - F_i(\hat{x}, \hat{t})}{\Delta\hat{t}} = -\frac{1}{\hat{\tau} \cdot \varepsilon} (F_i - F_i^{(eq)}) \end{aligned} \quad (13)$$

Therefore, the two terms on the left hand side could be canceled out and thereby the method becomes explicit. Choosing $\Delta t = t_c$, multiplying the above equation by $\Delta\hat{t}$ and dropping all carets, one obtains the dimensionless lattice Boltzmann BGK equation:

$$F_i(\mathbf{x} + \mathbf{c}_i \cdot \Delta t, t + \Delta t) - F_i(\mathbf{x}, t) = -\frac{1}{\tau} (F_i - F_i^{(eq)}) \quad (14)$$

As we can see, this equation has a particularly simple physical interpretation in which the collision term is evaluated locally and there is only one streaming step or 'shift' operation per lattice velocity. This stream-and-collide particle interpretation is a result of the fully Lagrangian character of the equation for which the lattice spacing is the distance traveled by the particle during a time step. Higher order discretization of the discrete Boltzmann equation typically requires several 'shift' operations to evaluate each derivative and a particle interpretation is less obvious than the above one.

When a specific lattice scheme is introduced, the corresponding equilibrium distribution function and the related coefficients could be derived. In what follows, a D2Q9 lattice will be adopted as an example to show this derivation procedure.

Fig. 3. D2Q9 lattice scheme

As shown in Figure 3, the corresponding lattice velocities for D2Q9 lattice are as follows:

$$\begin{aligned} c_0 &= (0,0) \\ c_{1,3,2,4} &= (\pm c, 0), (0, \pm c) \\ c_{5,6,7,8} &= (\pm c, \pm c) \end{aligned} \quad (15)$$

The mass density, ρ , and the momentum density, $\mathbf{j}$, are defined by the sums over the distribution function $F_i(\mathbf{x}, t)$

$$\rho(\mathbf{x}, t) = \sum_i F_i(\mathbf{x}, t) \quad (16)$$

$$\mathbf{j}(\mathbf{x}, t) = \rho(\mathbf{x}, t) \cdot \mathbf{u}(\mathbf{x}, t) = \sum_i \mathbf{c}_i \cdot F_i(\mathbf{x}, t) \quad (17)$$

For vanishing velocities, a global equilibrium distribution W_i ("fluid at rest") is defined. In the vicinity (small Mach numbers) of this equilibrium, distribution functions could be written as sums of the W_i and small perturbations $f_i(\mathbf{x}, t)$

$$F_i(\mathbf{x}, t) = W_i + f_i(\mathbf{x}, t) \quad (18)$$

With $|f_i(\mathbf{x}, t)| \ll W_i$.

The W_i should be positive to guarantee the positive mass density, as shown in Eq.16. They are chosen of Maxwell type in the following sense. The lattice velocity moments up to fourth order over the W_i shall be identical to the respective velocity moments over the Maxwell distribution as shown in Eq.3.

Thus, the odd moments vanish:

$$\sum_{i=0}^8 W_i c_{i\alpha} = 0 \quad (19)$$

$$\sum_{i=0}^8 W_i c_{i\alpha} c_{i\beta} c_{i\gamma} = 0 \quad (20)$$

And the even moments read:

$$\sum_{i=0}^8 W_i = \int d\mathbf{v} \cdot f^{(M)}(\mathbf{x}, \mathbf{v}) = \rho \quad (21)$$

$$\sum_{i=0}^8 W_i c_{i\alpha} c_{i\beta} = \int d\mathbf{v} \cdot f^{(M)}(\mathbf{x}, \mathbf{v}) \cdot \mathbf{v}_\alpha \cdot \mathbf{v}_\beta = \rho \frac{k_B T}{m} \delta_{\alpha\beta} \quad (22)$$

$$\sum_{i=0}^8 W_i c_{i\alpha} c_{i\beta} c_{i\gamma} c_{i\delta} = \int d\mathbf{v} \cdot f^{(M)}(\mathbf{x}, \mathbf{v}) \mathbf{v}_\alpha \mathbf{v}_\beta \mathbf{v}_\gamma \mathbf{v}_\delta = \rho \left(\frac{k_B T}{m} \right)^2 (\delta_{\alpha\beta} \delta_{\gamma\delta} + \delta_{\alpha\gamma} \delta_{\beta\delta} + \delta_{\alpha\delta} \delta_{\beta\gamma}) \quad (23)$$

Note that the last constraint (i.e., Eq.23) is more rigorous than the requirement of pure lattice isotropy [1].

Nonnegative solutions of the above three equations, Eq. 21-23, for W_i could be obtained whenever the number of lattice velocities c_i is large enough [1]. For the D2Q9 lattice, after going through the calculations, one obtains

$$\frac{W_i}{\rho} = \frac{4}{9} \quad i = 0 \quad (24)$$

$$\frac{W_i}{\rho} = \frac{1}{9} \quad i = 1, 2, 3, 4 \quad (25)$$

$$\frac{W_i}{\rho} = \frac{1}{36} \quad i = 5, 6, 7, 8 \quad (26)$$

$$\frac{k_B T}{m} = \frac{c^2}{3} \quad (27)$$

The evolution of the LBE model consists of the recurring alternation between translation to the local equilibrium and propagation of the distributions to neighboring sites according to the lattice velocities. From Eq.12 the dimensionless BGK kinetic equation reads

$$F_i(\mathbf{x} + \mathbf{c}_i \cdot \Delta t, t + \Delta t) = (1 - \omega) F_i(\mathbf{x}, t) + \omega \cdot F_i^{(eq)}(\mathbf{x}, t) \quad (28)$$

Where, $\omega = \frac{1}{\tau}$ is the dimensionless collision frequency. The local equilibrium distributions $F_i^{(eq)}$ depend only on the local values of mass and momentum density

$$F_i^{(eq)}(\mathbf{x}, t) = F_i^{(eq)}(\rho(\mathbf{x}, t), \mathbf{j}(\mathbf{x}, t)) \quad (29)$$

They could be derived by applying the maximum entropy principle [1] under the constraints of mass and momentum conservations. Up to second order in momentum density, one obtains

$$F_i^{(eq)}(\rho, \mathbf{j}) = \frac{W_i}{\rho} \left\{ \rho + \frac{m}{k_B T} c_i \cdot \mathbf{j} + \frac{m}{2\rho \cdot k_B T} \left[\frac{m}{k_B T} (c_i \cdot \mathbf{j})^2 - \mathbf{j}^2 \right] \right\} \quad (30)$$

or more explicitly

$$F_i = \frac{4}{9} \rho \cdot \left[1 - \frac{3}{2} \cdot \frac{\mathbf{u}^2}{c^2} \right] \quad i = 0 \quad (31)$$

$$F_i = \frac{1}{9} \rho \cdot \left[1 + 3 \frac{c_i \cdot \mathbf{u}}{c^2} + \frac{9}{2} \frac{(c_i \cdot \mathbf{u})^2}{c^4} - \frac{3}{2} \cdot \frac{\mathbf{u}^2}{c^2} \right] \quad i = 1, 2, 3, 4 \quad (32)$$

$$F_i = \frac{1}{36} \rho \cdot \left[1 + 3 \frac{c_i \cdot \mathbf{u}}{c^2} + \frac{9}{2} \frac{(c_i \cdot \mathbf{u})^2}{c^4} - \frac{3}{2} \cdot \frac{\mathbf{u}^2}{c^2} \right] \quad i = 5, 6, 7, 8 \quad (33)$$

Based on the equilibrium distributions and lattice velocities, the lengthy and complicated multi-scale technique (Chapman-Enskog expansion [1]) yields the Navier-Stokes equation with pressure $p = \rho \cdot k_B T / m$, and the corresponding kinetic shear viscosity is

$$\nu = \frac{k_B T}{m} \left(\frac{1}{\omega} - \frac{1}{2} \right) \Delta t = \frac{c^2}{3} \left(\frac{1}{\omega} - \frac{1}{2} \right) \Delta t = \frac{2 - \omega}{6\omega} c^2 \Delta t = \frac{c^2}{3} \left(\tau - \frac{1}{2} \right) \Delta t \quad (34)$$

The above given presentation of the LBE model contains all information necessary to set up the computer code. The algorithm proceeds as follows:

1. For given initial values of mass $\rho(\mathbf{x}, t)$ and momentum density $\mathbf{j}(\mathbf{x}, t)$, evaluate the equilibrium distributions $F_i^{(eq)}(\rho(\mathbf{x}, t), \mathbf{j}(\mathbf{x}, t))$ from Eq.30 or Eq.31-33, and set $F_i = F_i^{(eq)}$.
2. Apply the kinetic equation (Eq.28), i.e., add the the (non-equilibrium) distribution function $F_i(\mathbf{x}, t)$ and the equilibrium distribution function $F_i^{(eq)}(\mathbf{x}, t)$ with the appropriate weights $(1 - \omega)$ and ω , and then propagate it to the next neighbor (except for the distribution of 'rest particles' with $c_0 = 0$).
3. Evaluate the new values of $\rho(\mathbf{x}, t)$ and $\mathbf{j}(\mathbf{x}, t)$ from the propagated distributions according to the corresponding definitions, Eq.16 and Eq.17.
4. Check the convergence criterion and calculate the new equilibrium distributions with the latest values of $\rho(\mathbf{x}, t)$ and $\mathbf{j}(\mathbf{x}, t)$, and then go back to the second step of the algorithm until the convergence criterion is satisfied.

As shown in BGK-LBE model, the collisions are not explicitly defined like that in LGCA, but are accomplished by the transition to local equilibrium (the term $\omega \cdot (\mathbf{F}_i^{(eq)}(\mathbf{x}, t))$ in the kinetic equation).

2.3 Incompressible D3Q27 LBE model

For the 3D fluid flow problem, the lattice D3Q15, D3Q19 and D3Q27 are commonly adopted for the different situations. However, since the lattice D3Q27 has more isotropy than other 3D lattice schemes, it is more stable and more accurate than other 3D schemes for the same fluid flow problem. Therefore, in what follows, the D3Q27 lattice model will be used as an example to show how to develop the incompressible LBE model.

Fig. 4. D3Q27 lattice model

For the D3Q27 LBE model as shown in Figure 4, the corresponding discretized Boltzmann equation and equilibrium distribution functions are as follows:

$$f_{\alpha}(\mathbf{x} + \mathbf{e}_{\alpha} \cdot \Delta t, t + \Delta t) - f_{\alpha}(\mathbf{x}, t) = -\frac{1}{\tau} [f_{\alpha}(\mathbf{x}, t) - f_{\alpha}^{(eq)}(\mathbf{x}, t)] \quad (35)$$

$$f_{\alpha}^{(eq)} = \omega_{\alpha} \rho \left[1 + \frac{3}{c^2} \mathbf{e}_{\alpha} \cdot \mathbf{u} + \frac{9}{2c^4} (\mathbf{e}_{\alpha} \cdot \mathbf{u})^2 - \frac{3}{2c^2} (\mathbf{u} \cdot \mathbf{u}) \right] \quad (36)$$

Where, $\mathbf{e}_{\alpha}$ is the discrete velocity, $c = \Delta x / \Delta t$ the lattice speed, $\mathbf{u}$ the macroscopic velocity, ω_{α} the weighting factors, and Δx and Δt are the lattice constant and time step, respectively. The lattice velocities and corresponding weighting factors for equilibrium distribution functions are as follows:

$$\mathbf{e}_{\alpha} = \begin{cases} (0, 0, 0), & \alpha = 0 \\ (\pm 1, 0, 0)c, (0, \pm 1, 0)c, (0, 0, \pm 1)c, & \alpha = 1, 2, \dots, 6 \\ (\pm 1, \pm 1, 0)c, (0, \pm 1, \pm 1)c, (\pm 1, 0, \pm 1)c, & \alpha = 7, 8, \dots, 18 \\ (\pm 1, \pm 1, \pm 1)c & \alpha = 19, 20, \dots, 26 \end{cases} \quad (37)$$

$$\omega_\alpha = \begin{cases} 8/27, & \alpha = 0 \\ 2/27, & \alpha = 1, 2, \dots, 6 \\ 1/54, & \alpha = 7, 8, \dots, 18 \\ 1/16, & \alpha = 19, 20, \dots, 26 \end{cases} \quad (38)$$

Using the Chapman-Enskog expansion, the viscosity in the corresponding Navier-Stokes equation could be derived based on the Eq.35-38.

$$\nu = \frac{c^2}{3} \left(\tau - \frac{1}{2} \right) \Delta t \quad (39)$$

This choice for the viscosity makes the BGK-LBE scheme formally a second order method for solving compressible fluid flow and the fluid flow problem could be solved by the following two steps:

Collision step:

$$\bar{f}_\alpha(\mathbf{x}, t) = f_\alpha(\mathbf{x}, t) - \frac{1}{\tau} [f_\alpha(\mathbf{x}, t) - f_\alpha^{(eq)}(\mathbf{x}, t)] \quad (40)$$

Streaming step:

$$f_\alpha(\mathbf{x} + \mathbf{e}_\alpha \cdot \Delta t, t + \Delta t) = \bar{f}_\alpha(\mathbf{x}, t) \quad (41)$$

Where, $\bar{f}_\alpha$ denotes the post-collision state of the distribution function. Note that the collision step is completely local, and the streaming step is uniform and requires little computational effort.

It is worth mentioning that the above equations are based on the compressible fluid flow assumptions, i.e., the density is one of the independent variables. Since the incompressible fluid assumption is widely used in the real industrial applications and the corresponding error is small enough and acceptable, the incompressible LBE model is necessary to derive for the industrial applications. Because the pressure could be treated as an independent variable instead of density, the incompressible D3Q27 LBE model [12] could be obtained through the following procedures.

It is well understood that in an incompressible fluid the density is approximately a constant, ρ_0 , and the density fluctuation, $\delta\rho$, should be of the order $O(M^2)$ in the limit of Mach number $M \rightarrow 0$ [13]. If we explicitly substitute $\rho = \rho_0 + \delta\rho$ into the equilibrium distribution function, $f_\alpha^{(eq)}$, and neglect the terms proportional to $\delta\rho(u/c)$, and $\delta\rho(u/c)^2$, which are of the order $O(M^3)$ or higher, then the equilibrium density distribution function becomes:

$$f_\alpha^{(eq)} = \omega_\alpha \left\{ \rho + \rho_0 \left[\frac{3}{c^2} \mathbf{e}_\alpha \cdot \mathbf{u} + \frac{9}{2c^4} (\mathbf{e}_\alpha \cdot \mathbf{u})^2 - \frac{3}{2c^2} (\mathbf{u} \cdot \mathbf{u}) \right] \right\} \quad (42)$$

If a local pressure distribution function, $p_\alpha \equiv c_s^2 f_\alpha$, is introduced into the above equilibrium equation, the pressure representative equilibrium equation for an incompressible fluid could be evaluated as follows:

$$p_{\alpha}^{(eq)} = c_s^2 f_{\alpha}^{(eq)} = \omega_{\alpha} \left\{ p + p_0 \left[3 \frac{(\mathbf{e}_{\alpha} \cdot \mathbf{u})}{c^2} + \frac{9}{2} \frac{(\mathbf{e}_{\alpha} \cdot \mathbf{u})^2}{c^4} - \frac{3}{2} \frac{\mathbf{u}^2}{c^2} \right] \right\} \quad (43)$$

Where, c_s is the speed sound, and $c_s = \frac{c}{\sqrt{3}}$ for D3Q27 model, p_0 is the average pressure and the equation of state for this incompressible fluid is $p = c_s^2 \rho$. Accordingly, the evolution equation of the LBE system for the incompressible fluid becomes,

$$p_{\alpha}(\mathbf{x} + \mathbf{e}_{\alpha} \cdot \Delta t, t + \Delta t) - p_{\alpha}(\mathbf{x}, t) = -\frac{1}{\tau} [p_{\alpha}(\mathbf{x}, t) - p_{\alpha}^{(eq)}(\mathbf{x}, t)] \quad (44)$$

The macroscopic parameters of pressure, p , and velocity, $\mathbf{u}$, are given by

$$p = \sum_{\alpha} p_{\alpha} \quad (45)$$

$$p_0 \mathbf{u} = \sum_{\alpha} \mathbf{e}_{\alpha} p_{\alpha} \quad (46)$$

Since the equilibrium equation and evolution equation (i.e., Eq.43 and Eq.44) for incompressible fluid have been obtained, the pressure representative incompressible LBE system could be solved by following the same procedure (i.e., collision and streaming steps as mentioned before) as the density representative compressible LBM in Eq.40 and Eq.41.

3. Thermal Lattice Boltzmann model

In the past 15 years, there has been rapid progress in developing the method of the lattice Boltzmann equation (LBE) for solving a variety of fluid flow problems. However, the effort for construction of stable thermal lattice Boltzmann equation (TLBE) models in order to simulate heat transfer has been initiated more recently. McNamera and Alder first succeeded in simulating heat transfer phenomena by adopting multispeed thermal fluid lattice Boltzmann models [14]. In general, the thermal lattice Boltzmann models could be classified into two categories: the multispeed or double-population model [15] and the passive-scalar approach [16].

3.1 Multispeed approach for thermal lattice Boltzmann model

The multispeed approach or internal energy density distribution function (IEDDF) model is also called double-population method. In this approach, an independent internal energy density distribution function was introduced to obtain the temperature field. Therefore, two independent density distribution functions, i.e., mass density distribution function and internal energy density distribution function, are adopted to describe the fluid flow and heat transfer, respectively. As compared to the passive-scalar method, this model can exactly recover the energy conservation equation for compressible fluid flow at the macroscopic level, i.e., the viscous dissipation and compression work done by the pressure could be taken into account in this method.

Since the different independent density distribution functions are adopted to describe the fluid flow and heat transfer, these two density distribution functions may not share the same lattice model (or lattice velocity) and corresponding equilibrium distribution

functions. For example, in the 3D case the fluid flow might use D3Q27 lattice model, however, the internal energy could use D3Q19 or higher lattice model and associated equilibrium distribution function.

For the incompressible thermal problem, He *et al.* [17] proposed two distribution functions: mass density distribution function and internal energy density distribution function. The governing equations for these two functions are:

$$f_\alpha(\mathbf{x} + \mathbf{e}_\alpha \Delta t, t + \Delta t) - f_\alpha(\mathbf{x}, t) = -\frac{1}{\tau_v} [f_\alpha(\mathbf{x}, t) - f_\alpha^{(eq)}(\mathbf{x}, t)] + \delta \mathbf{F} \quad (47)$$

$$g_\alpha(\mathbf{x} + \mathbf{e}_\alpha \Delta t, t + \Delta t) - g_\alpha(\mathbf{x}, t) = -\frac{1}{\tau_T} [g_\alpha(\mathbf{x}, t) - g_\alpha^{(eq)}(\mathbf{x}, t)] \quad (48)$$

Where, $\mathbf{F}$ is an external body force term, τ_v and τ_T are the fluid flow and heat transfer relaxation times, respectively.

Fig. 5. D3Q19 Lattice Model

As shown in Figure 5, the D3Q19 lattice model is presented. The corresponding lattice velocities [18] of D3Q19 lattice model are defined as:

$$\mathbf{e}_\alpha = \begin{cases} (0,0,0), & \alpha = 0 \\ (\pm 1, 0, 0)c, (0, \pm 1, 0)c, (0, 0, \pm 1)c, & \alpha = 1, 2, \dots, 6 \\ (\pm 1, \pm 1, 0)c, (0, \pm 1, \pm 1)c, (\pm 1, 0, \pm 1)c, & \alpha = 7, 8, \dots, 18 \end{cases} \quad (49)$$

And the associated mass density equilibrium distribution functions are given as:

$$f_\alpha^{(eq)} = \omega_\alpha \rho \left[1 + \frac{3}{c^2} \mathbf{e}_\alpha \cdot \mathbf{u} + \frac{9}{2c^4} (\mathbf{e}_\alpha \cdot \mathbf{u})^2 - \frac{3}{2c^2} (\mathbf{u} \cdot \mathbf{u}) \right] \quad (50)$$

The weighting factors for each lattice direction are:

$$\omega_\alpha = \begin{cases} 1/3, & \alpha = 0 \\ 1/18, & \alpha = 1, 2, \dots, 6, \\ 1/36, & \alpha = 7, 8, \dots, 18 \end{cases} \quad (51)$$

In this model, the fluid pressure is determined by the equation of state $p = c_s^2 \rho$ and the sound speed is $c_s^2 = \frac{c^2}{3}$, the viscosity in this model is calculated from the equation $\nu = \frac{c^2}{3} \left(\tau_v - \frac{1}{2} \right) \Delta t$, in which τ_v is the fluid flow relaxation time. The corresponding mass density and macroscopic velocities could be evaluated by the following equations:

$$\rho = \sum_\alpha f_\alpha \quad (52)$$

$$\rho \cdot \mathbf{u} = \sum_\alpha f_\alpha \mathbf{e}_\alpha \quad (53)$$

Usually, the higher order quadrature for velocity is required for the thermal LBE model as compared to the fluid flow LBE model. Following the similar procedure of fluid flow LBE model, the LBE thermal models could be derived by properly discretizing the continuous evolution equation (Boltzmann equation) for the internal energy density distribution in temporal, spatial and velocity spaces. Therefore, the continuous equilibrium distribution function for the internal energy density distribution function can be represented by [19]

$$\begin{aligned} g^{(eq)} = & \frac{\rho \varepsilon}{(2\pi RT)^{D/2}} \exp\left(-\frac{\mathbf{e}^2}{2RT}\right) \left[\frac{\mathbf{e}^2}{DRT} + \left(\frac{\mathbf{e}^2}{DRT} - \frac{2}{D}\right) \frac{(\mathbf{e} \cdot \mathbf{u})}{RT} + \frac{(\mathbf{e} \cdot \mathbf{u})^2}{2(RT)^2} - \frac{\mathbf{u}^2}{2RT} \right] \\ & + \frac{\rho \varepsilon}{(2\pi RT)^{D/2}} \exp\left(-\frac{\mathbf{e}^2}{2RT}\right) \left[\left(\frac{\mathbf{e}^2}{DRT} - \frac{D+4}{D}\right) \frac{(\mathbf{e} \cdot \mathbf{u})^2}{2(RT)^2} - \left(\frac{\mathbf{e}^2}{DRT} - \frac{D+2}{D}\right) \frac{\mathbf{u}^2}{2RT} \right] \end{aligned} \quad (54)$$

The zeroth- through second-order [17] moment of the second term in the above equation vanishes. Consequently, this term can be eliminated without affecting the recovery of the macroscopic energy equation from the energy evolution equation. The zeroth- through second-order moment of remaining part of the energy equilibrium distribution involves only zeroth- through fifth-order moment of $\int \xi^m \exp(-\xi^2) d\xi = \sum w_\alpha \xi_\alpha^m$. Therefore, the third-order Gauss-Hermite quadrature is still valid. Thus, the same lattice models for the internal energy density distribution function as those used for the mass density distribution function can be adopted (e.g., D3Q19 lattice model). At the same time, it can be seen from the thermal equilibrium distribution function that after omitting the second term, the equilibrium internal energy density distribution function has the similar form as the equilibrium mass density distribution function. Following the same derivation procedure, the equilibrium internal energy density distribution functions for D3Q19 can be obtained.

$$g_0^{(eq)} = -\frac{\rho \varepsilon}{2} \frac{\mathbf{u}^2}{c^2} \quad (55)$$

$$g_{1-6}^{(eq)} = \frac{\rho \varepsilon}{18} \left[1 + \frac{\mathbf{e}_\alpha \cdot \mathbf{u}}{c^2} + \frac{9}{2} \frac{(\mathbf{e}_\alpha \cdot \mathbf{u}^2)}{c^4} - \frac{3}{2} \frac{\mathbf{u}^2}{c^2} \right] \quad (56)$$

$$g_{7-18}^{(eq)} = \frac{\rho \varepsilon}{36} \left[2 + 4 \frac{\mathbf{e}_\alpha \cdot \mathbf{u}}{c^2} + \frac{9}{2} \frac{(\mathbf{e}_\alpha \cdot \mathbf{u}^2)}{c^4} - \frac{3}{2} \frac{\mathbf{u}^2}{c^2} \right] \quad (57)$$

The internal energy is related to the temperature by $\varepsilon = 3RT/2$, where R is the gas constant. Then the macroscopic density, velocity and temperature could be evaluated by

$$\rho = \sum_{\alpha} f_{\alpha} \quad (58)$$

$$\rho \cdot \mathbf{u} = \sum_{\alpha} \mathbf{e}_{\alpha} \cdot f_{\alpha} \quad (59)$$

$$\rho(3RT/2) = \sum_{\alpha} g_{\alpha} \quad (60)$$

The corresponding thermal diffusivity α for D3Q19 model is determined by

$$\alpha = \frac{5}{9} \left(\tau_T - \frac{1}{2} \right) \Delta t \quad (61)$$

3.2 Passive-scalar approach for thermal lattice Boltzmann model

The passive-scalar approach [16] utilizes the fact that the macroscopic temperature satisfies the same evolution equation as a passive scalar if the viscous dissipation and compression work done by the pressure are negligible. Therefore, in the passive-scalar thermal LBE model, the temperature is simulated by using a separate distribution function which is independent of the mass density distribution, however, this independent distribution function shares the same lattice model and equilibrium distribution functions with the mass density distribution function, this is different than the double-population method or multispeed LBE thermal model.

The main advantage of this method is the enhancement of numerical stability as compared to the previous multispeed thermal lattice Boltzmann model. It has been shown that this passive-scalar method has the same stability with the fluid flow LBE model. However, since the temperature distribution function shares the same lattice model and equilibrium distribution functions with fluid flow LBE model, the viscous dissipation and compression work done by the pressure can not be taken into account in this model.

The passive-scalar thermal LBE model is originated from the multiple component LBE model with interparticle interaction [20], which was developed for the simulation of multiphase flow and phase transitions. In the multiple components LBE model, the components can be miscible or partially immiscible depending on the strength of the interaction. When the interaction is weak or in a single phase region of a multiphase system, this model can be used to simulate diffusion due to various driving mechanisms [21]. In this passive-scalar approach, the distribution function of each component evolves according to the discretized Lattice Boltzmann equation. The same form of the equilibrium distribution

function is used for all the components except that density and velocity are calculated separately for each component. In the absence of any interaction and external forces, the distribution functions of all the components were assumed to have a common velocity $\mathbf{u}'$, and the conservation of the total momentum at each collision requires that

$$\mathbf{u}' = \sum_{\sigma=1}^S \frac{m_{\sigma} n_{\sigma} \mathbf{u}_{\sigma}}{\tau_{\sigma}} \bigg/ \sum_{\sigma=1}^S \frac{m_{\sigma} n_{\sigma}}{\tau_{\sigma}} \quad (62)$$

Where S is the number of components in the system, m_{σ} , τ_{σ} and $n_{\sigma} = \sum_{\alpha} n_{\alpha}^{\sigma}$ are the molecular mass, the relaxation time, the number density of the component σ , respectively. And $m_{\sigma} n_{\sigma} \mathbf{u}_{\sigma} = m_{\sigma} \sum_{\alpha} n_{\alpha}^{\sigma} \mathbf{e}_{\alpha}$ is the momentum of component σ calculated from its distribution function n_{α}^{σ} . When the force term, $\mathbf{F}_{\sigma}$, applied to component σ , the momentum has to be correspondingly increased, and this was done by replacing velocity in the equilibrium distribution functions with the summation of old velocity and related velocity increment. The force $\mathbf{F}_{\sigma}$ in general includes both interparticle forces and external forces.

In the most general multiple components LBE model with interparticle interaction and external forces, there are three types of diffusions due to different driving mechanisms [21]. They are ordinary diffusion, pressure diffusion and forced diffusion. With the equilibrium distribution functions, the pressure diffusion does not appear. If a common acceleration is applied to all the components, namely $\mathbf{F}_{\sigma} = \rho_{\sigma} \mathbf{g}$, forced diffusion is also absent. The only type of diffusion left is the ordinary diffusion due to concentration gradients which obeys Fick's Law. In addition, a component (e.g., component S) can be made to behave as a passive scalar by setting its molecular mass to zero together with its interaction with all the other components. Therefore, this component will not contribute to the total momentum of the mixture. It is simply advected "passively" and diffuses into the main flow, having no effect on the flow. Similar to the kinetic viscosity in fluid flow LBE model, the corresponding thermal diffusivity in the passive scalar model is defined as:

$$\alpha = \frac{c^2}{3} \left(\tau_T - \frac{1}{2} \right) \Delta t \quad (63)$$

Where, τ_T is the corresponding heat transfer relaxation time. Since the viscosity in the LBE model is defined as

$$\nu = \frac{c^2}{3} \left(\tau_v - \frac{1}{2} \right) \Delta t \quad (64)$$

Therefore, the corresponding Prandtl number of the fluid is

$$\text{Pr} = \frac{\nu}{\alpha} = \frac{2\tau_v - 1}{2\tau_T - 1} \quad (65)$$

In a word, to compare these two thermal LBE models, it is easy to make the conclusion that the passive scalar method is much easier and better than multispeed approach if the viscous dissipation and compression work done by the pressure can be negligible.

4. Boundary treatments for LBE model

In the previous sections, the fluid flow lattice Boltzmann equation model, the thermal lattice Boltzmann model and corresponding algorithm to solve for the macroscopic variables (i.e., velocity, pressure and temperature) are demonstrated. In order to properly proceed the algorithm and eventually get the meaningful results, special attentions must be paid at the boundary treatments, which is as important as it usually does for solving the Navier-Stokes equations by finite difference, finite elements or any other top-down models.

For the top-down model, since it starts with macroscopic variables, such as velocity, temperature, pressure, and ends up with the same macroscopic variable distributions. It is straightforward to specify the macroscopic variables on the boundary to the corresponding algebraic equations. However, for the bottom-up models (such as LBE model, MD model), all the calculations are proceeded at the microscopic or mesoscopic level and we do not have the corresponding microscopic or mesoscopic information to specify, but only the macroscopic boundary information in the computational domain. Therefore, how to develop the connections between the microscopic or mesoscopic and macroscopic levels on the specific physical boundaries is one of the necessary steps before these bottom-up methods can be used to do some real case simulations.

Specifically, for the LBE model, since the calculation process is very simple, i.e., streaming and collision, the boundary treatment schemes focus on how to find out the directional density distribution functions on the boundary nodes pointing from outside of domain (or solid) to inside of computational domain (or fluid). However, the difficulties arise from that there is no essential information provided by streaming process at the previous time level on the boundary nodes, and how to transfer the macroscopic information (e.g., velocity and pressure) to the mesoscopic information, i.e., density distributions functions for the boundary nodes.

4.1 Periodic boundary scheme

The simplest boundary condition is “periodic” in that the system becomes closed by the edges being treated as if they are attached to opposite sides, as shown in Fig. 6. The highlight dashed edges at circumference of cylinder are attached seamlessly.

Fig. 6. Periodic boundary illustration

In the literatures, most early papers used these periodic conditions at inlet/outlet boundary along with bounce back boundaries at wall. In simulating flow in a duct for example, bounce back boundaries would be applied at the duct walls and periodic boundaries would be applied to the “open” ends of the inlet and outlet cross sections.

Fully periodic boundaries are also useful in some cases (for example, simulation of an infinite domain of multiphase fluids). In this case, the computational domain topology is like a torus, i.e., there are no specific edges or all the edges are connected together in the specific direction.

The implementation of periodic boundary treatment is very easy. On the streaming step, once the computational boundary nodes are detected, the boundary neighboring nodes, which lie outside of computational domain, will be pointed to the appropriate nodes on the opposite boundary in the same orientation.

4.2 Bounce back boundary scheme

As shown in Figure 7, the so-called bounce back scheme is when a particle reaches a wall node, the particle will scatter back to the fluid nodes along its incoming direction. The collision process does not occur at the boundary, but only at the internal fluid region.

Fig. 7. Illustration of bounce back boundary treatment

In order to apply the bounce back boundary, the entire solids will be separated into two categories: boundary solids that lie at the solid-fluid interface and isolated solids that do not contact with the fluid. With this division, it is possible to eliminate unnecessary computations at inactive nodes, i.e., isolated nodes. For example, this treatment could be particularly important for the simulation of fluid flow in fractured media, where the fraction of the total domain occupied by the open space accessible to fluids might be very small.

Since bounce back boundary treatment is particularly simple, it has played a major role in making LBE model popular among modelers interested in simulating fluids in domains characterized by complex geometries such as those found in porous media. The beauty of this boundary scheme lies in that one simply needs to designate a particular node as a solid obstacle and no other special programming treatment is required. Therefore, it is trivial to incorporate images of porous media for example and immediately conduct the fluid flow simulation in them.

It has been found that the slip velocity between boundary surface and fluid is zero at the boundary as long as the momentum dissipated by boundaries is equal to the stress provided by the fluid, no matter what combination of distributions is chosen. Therefore, the bounce back boundary scheme would inevitably generate a nonzero slip velocity between the boundary surface and the fluid at the boundary because the collision step just happens only in interior fluid region. In addition, it has been already confirmed that the bounce back scheme is only of first order in numerical accuracy at the boundaries, which degrades the LBE, because the numerical accuracy of the LBE is of second order in the interior points. Even though there are different versions of bounce back boundary scheme (e.g., modified bounce back scheme, in which the collision process is introduced at the boundary nodes, and bounce back scheme with the wall located halfway [22] between a fluid node and a bounce back node) and the accuracy of these schemes was improved a lot as compared to the original bounce back boundary scheme, the no-slip velocity still exists except the case in which the relaxation time is 1.

4.3 Von Neumann (Flux) boundary

Von Neumann boundary conditions specify the flux at the boundaries. For example, in 2D case, a velocity vector consisting of x and y components is specified from which density/pressure is computed on the basis of conditions inside the computational domain. Only the macroscopic density/pressure and the unknown directional density distributions need to be computed. After the streaming step, there are three unknown directional densities at each lattice node pointing from the boundary into the computational domain, i.e., the highlighted density distributions f_4 , f_7 and f_8 as shown in Figure 8. These unknowns can be solved in a way that maintains the specified macroscopic velocities at their lattice nodes.

Fig. 8. Von Neumann Boundary for D2Q9 lattice

As shown in Figure 8, assuming the boundary conditions are vertical velocity $u_y = u_{y0}$ and horizontal velocity $u_x = 0$. After streaming at solid-fluid boundary, there are three unknown directional density distributions, i.e., f_4 , f_7 and f_8 . The other distributions are already known because they arrived from other nodes inside the computational domain. Four equations are needed to solve for ρ , f_4 , f_7 and f_8 .

The macroscopic density formula can be written as:

$$\rho = \sum_{\alpha} f_{\alpha} \quad (66)$$

By considering the individual f_s that can contribute to x and y velocities, the equation for macroscopic velocity is:

$$\mathbf{u} = \frac{1}{\rho} \sum_{\alpha} f_{\alpha} \mathbf{e}_{\alpha} \quad (67)$$

This gives two equations, one for each direction:

X-direction:

$$0 = f_1 - f_3 + f_5 - f_6 - f_7 + f_8 \quad (68)$$

and

Y-direction:

$$\rho \cdot u_{y0} = f_2 - f_4 + f_5 + f_6 - f_7 - f_8 \quad (69)$$

A fourth equation can be written by assuming that the bounce back condition holds in the direction normal to the boundary

$$f_2 - f_2^{(eq)} = f_4 - f_4^{(eq)} \quad (70)$$

as proposed by Zou and He (1997) [23].

Based on Eq.66, Eq.68, Eq.69 and Eq.70, the four unknowns (ρ , f_4 , f_7 and f_8) can be solved as follows:

$$\rho = \frac{f_0 + f_1 + f_3 + 2(f_2 + f_5 + f_6)}{1 + u_{y0}} \quad (71)$$

$$f_4 = f_2 - f_2^{(eq)} + f_4^{(eq)} = f_2 - \frac{2}{3} \rho \cdot u_{y0} \quad (72)$$

$$f_7 = f_5 + \frac{1}{2}(f_1 - f_3) - \frac{1}{6} \rho \cdot u_{y0} \quad (73)$$

$$f_8 = f_5 - \frac{1}{2}(f_1 - f_3) - \frac{1}{6} \rho \cdot u_{y0} \quad (74)$$

To briefly summarize the procedure, a velocity is specified at the boundary and solve for the macroscopic density and three unknown directional density distributions by Eq.66, Eq.68, Eq.69 and Eq.70. The equations come from the usual macroscopic variable formula and the assumption that bounce back is still satisfied in the direction normal to the boundary.

4.4 Dirichlet (pressure) boundary

Dirichlet boundary conditions specify the pressure/density at the boundaries. The solution for these boundaries is closely related to that discussed above for the velocity boundaries. A density ρ_0 is specified from which velocity is computed. Since the relationship between mass density and pressure can be corrected by equation of state, specifying density is equivalent to specifying pressure.

Assuming velocity tangent to the boundary is zero and solve for the component of velocity normal to the boundary. In 2D case, the velocity and a proper distribution function at the boundary nodes need to be determined (as shown in Figure 8). After the streaming step, there are still three unknown directional density distributions at each lattice node pointing from the boundary into the computational domain. These unknowns can be solved in a way that maintains the specified pressure/density ρ_0 at their lattice nodes.

Given the boundary condition $\rho = \rho_0$ and the known directional densities ($f_0, f_1, f_2, f_3, f_5, f_6$), four unknowns (u_y, f_4, f_7 and f_8) need to be solved. Similarly, this can be done by: Macroscopic density equation:

$$\rho_0 = \sum_{\alpha} f_{\alpha} \quad (75)$$

Macroscopic velocity:

$$0 = f_1 - f_3 + f_5 - f_6 - f_7 + f_8 \quad (76)$$

$$\rho_0 u_y = f_2 - f_4 + f_5 + f_6 - f_7 - f_8 \quad (77)$$

And the assumption that bounce back [23] still holds in the direction normal to the boundary:

$$f_2 - f_2^{(eq)} = f_4 - f_4^{(eq)} \quad (78)$$

Based on the above four equations (i.e., Eq.75-78) to solve the four unknowns (u_y, f_4, f_7 and f_8), it yields:

$$u_y = -1 + \frac{f_0 + f_1 + f_3 + 2(f_2 + f_5 + f_6)}{\rho_0} \quad (79)$$

$$f_4 = f_2 - f_2^{(eq)} + f_4^{(eq)} = f_2 - \frac{2}{3} \rho_0 u_y \quad (80)$$

$$f_7 = f_5 + \frac{1}{2}(f_1 - f_3) - \frac{1}{6} \rho_0 u_y \quad (81)$$

$$f_8 = f_5 - \frac{1}{2}(f_1 - f_3) - \frac{1}{6} \rho_0 u_y \quad (82)$$

It can be seen that the calculation procedure is similar to the Von Neumann boundary case.

4.5 Extrapolation boundary scheme

Since the lattice Boltzmann scheme can be regarded as a special finite difference of the kinetic equation, this scheme allows us to use the extrapolation scheme for the boundary [24], which is similar to the approach for boundary conditions used in traditional finite difference schemes. The idea of this procedure is as follows: for any given fluid flow, it is assumed that there is one additional layer of sites, beyond the boundary, inside the wall (such as layer F-G-H as shown in Figure 9).

quantities, such as density, momentum flux and the strain field, are simply weighted summations of the distribution functions over particle velocity directions, it can be concluded that the corresponding macroscopic quantities have the same numerical accuracy as the distribution functions. Since the collision process occur not only on the nodes in the internal fluid region, but also on the nodes at the boundary, and the macroscopic boundary conditions are directly applied to the corresponding equilibrium distribution function for the boundary nodes, the extrapolation boundary scheme is also a no-slip boundary scheme.

4.6 Thermal boundary treatment

Considering the thermal boundary conditions, the fluid flow boundary treatments (such as bounce back boundary, extrapolation boundary) can also be used by the temperature variable. However, it has been confirmed that [25-27], to date, the model of assuming a counter slip thermal energy density boundary is of the highest accurate boundary treatments because it can guarantee the fixed velocity and temperature or heat flux at the wall exactly.

For the isothermal wall boundary case, the temperature is fixed as T_0 at the solid surface pointing to the fluid region. After streaming, g_2 , g_5 and g_6 are unknowns, as shown in Figure 10.

Fig. 10. Counter slip thermal boundary for D2Q9 lattice

Assume these unknown temperature density distributions are equal to their temperature density equilibrium distributions with unknown temperature T^* . Adding these three unknown temperature density distributions together (e.g., for the passive scalar thermal LBE model), it yields:

$$g_2 + g_5 + g_6 = \frac{1}{6} T^* (1 + 3u_y + 3u_y^2) \quad (84)$$

where u_y is the velocity normal to the wall and the x velocity (u_x) is assumed to be zero. If the value of T^* is known, the unknown temperature density distributions (g_2 , g_5 and g_6) can be solved by using the corresponding temperature equilibrium distribution functions. Meanwhile, note that for the isothermal wall, it reads:

$$\sum_{\alpha} g_{\alpha} = T_0 \quad (85)$$

Substitute Eq.84 into Eq.85 and solve for T^* , it shows:

$$T^* = \frac{6}{1 + 3u_y + 3u_y^2} (T_0 - g_0 - g_1 - g_3 - g_4 - g_7 - g_8) \quad (86)$$

Eventually, the three unknown temperature density distributions (g_2 , g_5 and g_6) can be obtained by substituting T^* into the temperature density equilibrium distributions. Obviously, this method can be easily extended to the 3D case without any difficulties. For the heat flux boundary condition, a second order finite difference scheme is used to evaluate the temperature on the wall,

$$q = \left. \frac{\partial T}{\partial y} \right|_{y=0} = \frac{-3T_{i,1} + 4T_{i,2} - T_{i,3}}{2\Delta y} \quad (87)$$

After finding the wall temperature, the same procedure as described in the isothermal wall case is used to calculate the unknown temperature density distributions.

4.7 Body force treatment

Since there exist all kinds of body forces (e.g., gravity, centrifugal force) in the real case simulations, how to deal with the body force in the LBE model is another important issue needed to be done before this method can be widely used in the real applications.

Consider the Newton's second law of motion, we have

$$\mathbf{F} = m\mathbf{a} = m \frac{d\mathbf{u}}{dt} \quad (88)$$

Recognizing that the density is proportional to the mass, the lattice length is the unit length in the LBE model (which means the volume for each basic lattice is also unit) and that the relaxation time τ is the elementary time of collisions, then rearrange Eq.88, it yields:

$$\Delta \mathbf{u} = \frac{\tau \cdot \mathbf{F}}{\rho} \quad (89)$$

where $\Delta \mathbf{u}$ is a increment in velocity due to the external body force. Finally, it can be written down like:

$$\mathbf{u}^{(eq)} = \mathbf{u} + \Delta \mathbf{u} = \mathbf{u} + \frac{\tau \cdot \mathbf{F}}{\rho} \quad (90)$$

where $\mathbf{u}^{(eq)}$ is used to compute the equilibrium distribution functions, $f^{(eq)}$, and $\mathbf{u}$ comes from the summation over the local distribution functions, f_α , like Eq.17.

5. Application of Lattice Boltzmann model

In the previous sections, the LBE fluid flow, heat transfer models and the corresponding boundary treatments have been discussed in details. In this section, these models will be used to solve the fully developed fluid flow and heat transfer problem in a curved square duct to validate this new approach. The D3Q27 incompressible fluid flow and passive scalar

thermal LBE models are adopted to simulate the fully developed fluid flow and heat transfer in the curved square duct, and the flow driven force, i.e., pressure difference, and centrifugal force, will be taken into account in the models like body force acting on each lattice.

5.1 Fluid flow and heat transfer in curved square duct

The study of viscous flow in the curved ducts is of fundamental interest in fluid mechanics because of the numerous applications such as flows through turbo machinery blade passages, aircraft intakes, diffusers, heat exchangers etc. [28]. The major effect of curved duct on the fluid flow involves the strong secondary flow due to the longitudinal curvature in the geometry. The presence of longitudinal curvature generates the centrifugal force perpendicular to the main flow along the axis and produces the so-called secondary flow on the cross section of ducts. As a consequence of this centrifugal force, the axial velocity profile will be distorted (from the typical-parabolic velocity profile in straight ducts) with an outward shift of the peak axial velocity, and at the same time, the total flow rate will be reduced due to the decrease of average axial velocity.

Fig. 11. The geometry configuration of curved square duct

As shown in Figure 11, a curved square duct with side length L is placed on the horizontal plane and the radius of the curved duct is R , measured from the center of duct to the center of the curve. The fluid flows in the square duct about the center of curvature toward the inside of the plane of the paper.

The curvature ratio of this curved duct is defined as:

$$C = \frac{D_h}{R} \quad (91)$$

Where, D_h is the equivalent hydrodynamic diameter of square duct, which is defined as

$$D_h = \frac{4A}{P} = \frac{4L^2}{4L} = L \quad (92)$$

The non-dimensional characteristic Dean number is defined as the function of Reynolds number and curvature ratio.

$$D_n = R_e \cdot (C)^{0.5} \quad (93)$$

$$R_e = \frac{U \cdot D_h}{\nu} \quad (94)$$

where, R_e is the Reynolds number of the duct flow. A uniform grid is used in this article, and the corresponding relaxation factors for fluid flow and heat transfer are in the range of 0.6 ~1.0.

On each cross section of curved duct, a constant pressure gradient is introduced to drive fluid flow along the duct's axial direction. Due to the longitudinal curvature and axial velocity, the centrifugal force is involved and drives the fluid flow far away from the center of curved duct (i.e., point **O** as shown in Figure 11), and then the so-called secondary flow is formed on the cross section. In this article, both pressure gradient and centrifugal force are treated like body forces acting on each interior lattice with different directions on the cross section of duct (which is the projection of the computational domain), and the macroscopic momentum conservation method [23, 24] is used to apply these body forces to the LBE fluid flow and heat transfer models.

As shown in Figure 11, the fluid flows in the square duct pointing toward the inside of the plane of the paper. Since the fully developed fluid flow and heat transfer will be studied numerically in this section, the axial velocity profile and dimensionless temperature distribution on the cross section will not be changed along the main flow direction. Therefore, the original computation domain, which consists of the square cross section and curved duct, can be simplified to a domain including the square cross section and a few lattice lengths perpendicular to this section plane. As a result of this simplification, the basic LBM with uniform lattice can be used to solve this problem without any special treatment for the curved boundary. It is worth noting that although the original computation domain has been simplified, the fluid flow and heat transfer simulations are still in three dimensions, i.e., two dimensions on the cross section and one dimension in the axial flow direction.

For fluid flow in pipes, the Fanning friction factor is defined as [25]:

$$C_f = \frac{2 \cdot \tau_w}{\rho \cdot V_{avg}^2} \quad (95)$$

where τ_w is the average wall shear stress and is defined as:

$$\tau_w = \mu \cdot \frac{dU_z}{dn} \quad (96)$$

Plugging Eq.96 and Eq.94 into Eq.95 and rearranging, one obtains the following friction factor for the curved duct.

$$C_f \cdot R_e = \frac{2 \cdot D_h}{V_{avg}} \cdot \frac{dU_z}{dn} \quad (97)$$

As far as fluid flow in straight square duct is concerned, the analytical friction factor is available in literature and equals 14.25 [26].

For heat transfer in curved duct, the corresponding dimensionless temperature and Nusselt number are defined as follows:

$$T^* = \frac{T_w - T}{T_w - T_m} \quad (98)$$

$$N_u = \frac{\bar{h} \cdot D_h}{k} \quad (99)$$

where T_w and T_m are the wall temperature and mean temperature across the duct, receptively. $\bar{h}$ is the average heat transfer coefficient around the four wall sides and is defined as:

$$\bar{h} = \frac{-k \cdot \frac{\partial T}{\partial n}}{T_w - T_m} \quad (100)$$

Plugging Eq.98 and Eq.100 into Eq.99 and rearranging, one obtains:

$$N_u = D_h \cdot \frac{\partial T^*}{\partial n} \quad (101)$$

For the fluid flow and heat transfer in a straight square duct, the analytical Nusselt number with constant wall temperature is 2.98 [26].

In order to obtain more accurate results, the two-dimensional Simpson integration method is used to calculate the average axial velocity based on the uniform lattice on the cross section.

5.2 Boundary conditions

In the simplified computation domain, the boundary conditions have to be specified in two parts with three directions, i.e., axial flow direction and two directions on the cross section plane.

Axial flow direction: Since both the fluid flow and heat transfer are fully developed, the velocity and dimensionless temperature profiles will not change along the axial flow direction (which is perpendicular to the plane of the paper and toward inside). Therefore, the simple periodic boundary can be naturally applied for both fluid flow and heat transfer in this direction, without any special treatment for the curved boundary.

Cross section plane: On the duct cross section, four sides of wall are needed to specify the boundary conditions. In this mathematic model, the no-slip boundary and constant wall temperature conditions are applied on the four duct walls for fluid flow and heat transfer, respectively. The second-order extrapolation boundary treatment [27] and counter-slip thermal boundary treatment [28] are employed for the fluid flow and heat transfer to determine the unknown distributions (i.e., pressure and dimensionless temperature distributions) coming from outside the computation domain.

Once all these boundary conditions are established, the fully developed fluid flow and heat transfer in the straight square duct are simulated using the current LBM model by letting curvature ratio approach to zero (i.e., the radius of curved duct, R , is big enough, for example, 10^9), and the converging criterion in this test simulation is that the relative error of

total velocity (including three components) on each uniform lattice is less than 1.0×10^{-4} for every 400 consecutive time steps. After the converging criterion is reached, the simulation results show that both friction coefficient and Nusselt number are very in good agreement with analytical results (i.e., the relative errors versus analytical results are less than 0.1%). This validates that this LBM model is correct and the simplification of computation domain is feasible and applicable. In addition to this bench mark validation of the straight duct, a grid convergence test was implemented before any data were adopted. For a given fluid flow problem, different number of uniform lattices (i.e., $50 \times 50 \times 3$, $100 \times 100 \times 3$ and $160 \times 160 \times 3$, where three lattices are used for the periodic direction) for the same computation domain were used to compare the differences in terms of friction coefficient and flow pattern on the cross section. After a couple of comparisons, it was found that the uniform lattice $100 \times 100 \times 3$ has both good accuracy and less computing cost. Therefore, $100 \times 100 \times 3$ uniform lattice mesh is used in this article to simulate the fluid flow and heat transfer problems.

5.3 Simulation results and discussion

5.3.1 Fluid flow

As shown in Figure 12, the non-dimensional axial velocity distribution of cross section b-b is presented at different Dean number with a constant curvature ratio, $C = 0.05$. From this Figure, it is evident that, as the Dean number increases, the maximum axial velocity first shifts toward the outside of the duct from near the center position until it reaches a most outside point on the cross section; and then once the Dean number reaches a certain value (which is a function of the curvature ratio), the velocity profile on cross section b-b suddenly changes to a new pattern (as shown in Figure 12), and the location of maximum axial velocity is much closer to the center of duct than before. Moreover, as Dean number increases further, the maximum axial velocity moves toward the center of duct, which is the opposite moving direction compared to the case with smaller Dean number before the new velocity profile was observed.

Fig. 12. Velocity distributions along cross section b-b

With the same conditions, the dimensionless axial velocity profile along cross section a-a is shown in Figure 13. In this Figure, it is apparent that as the Dean number increases, the peak of velocity profile on cross section a-a changes from one point at the beginning to two symmetrical points, and eventually up to three points (one is at the center and the other two

are symmetrical about the center one). Meanwhile, the axial velocity on this cross section a-a is becoming more uniform as Dean number increases. The switchback movement of maximum axial velocity on cross section b-b in Figure 12 and the change of number of peak velocities on plane a-a in Figure 13 can be explained as follows. On the duct cross section, the centrifugal force (which is induced by axial velocity and duct curvature) drives the fluid flow from inner side wall to the outer side wall and this fluid flow causes a symmetrical fluid flow pattern on the cross section for the horizontal curved duct. Compared to the main axial fluid flow (which is perpendicular to cross section), the flow on the cross section is called secondary flow. When the Dean number is small, there is just one peak point on the axial velocity profile, and the secondary flow is one pair of weak symmetrical eddies due to the small centrifugal force. As Dean number keeps increasing, the axial maximum velocity moves toward the outside of wall and, at the same time, the two symmetrical eddies become stronger and stronger, eventually distorting the axial velocity from a single peak to two symmetrical peaks on cross section a-a. On the other hand, once Dean number exceeds a certain value, the secondary flow suddenly changes from one pair symmetrical eddies to two pairs of symmetrical eddies (having opposite rotating directions) due to the imbalance between centrifugal force and pressure gradient on the cross section. This is the so-called Dean instability [29, 30]. Therefore, the velocity distribution on section b-b suddenly changes and another peak velocity appears at cross section a-a, that is, there are now three peak velocities.

Fig. 13. Velocity distributions along a-a cross section

In Figures 14 to 17 or 18 to 21, the detailed dimensionless axial velocity distributions are presented and the transition process from one velocity peak to two peaks and eventually to three peaks are all clearly shown (i.e., Figures 14- 16 or 18- 20 in 3-D view). Based on these axial velocity contours, it is obvious that the higher Dean number, the greater velocity gradient that will be observed around the duct walls, especially for the two vertical sides, and this is consistent with the previous results in Figure 13. Moreover, since the axial velocity with the same number of uniform contour lines is provided at different Dean numbers, the distribution of uniform contour lines shows how well the axial velocity distribute uniformly on the cross section. From Figures 15-17 or 19-21, one can conclude that the axial velocity trends to be more uniform on the cross section area as Dean number increases, which is also consistent with Figures 22 and 23.

Fig. 14. Velocity contour of cross section (Dn=50.0 C=0.05)

Fig. 15. Velocity contour of cross section (Dn=100.0 C=0.05)

Fig. 16. Velocity contour of cross section (Dn=150.0 C=0.05)

Fig. 17. Velocity contour of cross section (Dn=200.0 C=0.05)

Fig. 18. Axial velocity profile in 3D view (Dn=50.0 C=0.05)

Fig. 19. Axial velocity profile in 3D view (Dn=100.0 C=0.05)

Fig. 20. Axial velocity profile in 3D view
($Dn=150.0$ $C=0.05$)

Fig. 21. Axial velocity profile in 3D view
($Dn=200.0$ $C=0.05$)

Fig. 22. Velocity vector of cross section
($Dn=50.0$ $C=0.05$)

Fig. 23. Velocity vector of cross section
($Dn=100.0$ $C=0.05$)

Fig. 24. Velocity vector of cross section
($Dn=150.0$ $C=0.05$)

Fig. 25. Velocity vector of cross section
($Dn=200.0$ $C=0.05$)

In Figures 22-25, the detailed secondary flows on the cross section are presented at different Dean numbers. These figures also clearly show the flow transition from one pair of eddies (in Figures 22 and 23) to two pairs of symmetrical eddies (in Figures 24 and 25) when the Dean number increasing up to certain values. Compared to the single-pair symmetrical eddies, in the double-pair eddies case, the additional pair eddies rotate in the opposite direction on the main vortices (as shown in Figures 24 and 25). The imbalance between the pressure gradient and centrifugal force on the cross section causes the transition from a single-vortex-pair to the double-vortex-pair structure; therefore, the location of maximum axial velocity moves closer to the outer wall in symmetrical positions above and below the radial centerline in Figures 16 and 17 or 20 and 21.

Fig. 26. Pressure contour on cross section (Dn=50.0 C=0.05)

Fig. 27. Pressure contour on cross section (Dn=100.0 C=0.05)

Fig. 28. Pressure contour on cross section (Dn=150.0 C=0.05)

Fig. 29. Pressure contour on cross section (Dn=200.0 C=0.05)

In Figures 26-29 and Figures 30-33, the detailed pressure distributions are presented at different Dean numbers. It is apparent that the pressure contour patterns are considerably different for the double-pair eddies case (in Figures 28 and 29 or Figures 32 and 33) than the single-pair eddies' (in Figures 26 and 27 or 30 and 31). Furthermore, the pressure gradient on the cross section becomes more uniform as the Dean number continues increasing, as shown in Figures 26-28 or Figures 30-32. Since pressure is always a passive variable in fluid flow problems, the pressure profiles shown in Figures 26-29 or Figures 30-33 are a consequence of balance between centrifugal force and fluid viscous force.

Fig. 30. Pressure profile in 3D view
(Dn=50.0 C=0.05)

Fig. 31. Pressure profile in 3D view
(Dn=100.0 C=0.05)

Fig. 32. Pressure profile in 3D view
(Dn=150.0 C=0.05)

Fig. 33. Pressure profile in 3D view
(Dn=200.0 C=0.05)

In real engineering applications, one of the most important results of fully developed flow in curved square duct is estimation of the friction factor (or flow-rate reduction). The available well-known friction factor correlation in the literature [31] is:

$$f_c / f_s = 0.225 Dn^{0.39} \quad \text{for square duct with } 100 < Dn < 1500 \quad (102)$$

In Figure 34, the computed friction factor ratios for a straight square duct at different curvature ratio are presented. It is evident that the presented LBM simulation results are in good agreement with the experimental correlation. On the other hand, it is clearly shown in this Figure that the friction coefficient for each specific curvature ratio (from 0.05 to 1.00) changes suddenly when the Dean number is between 115 and 130. This is because when Dean number increases from 115 to 130, the flow pattern on cross section changes from one pair of eddies to two pairs of eddies. This Dean number which identifies the flow pattern is called critical Dean number. As a consequence of flow pattern change, the corresponding friction coefficient is increased.

Fig. 34. Friction coefficients vs Dean number at different curvature ratio

With regard to the transition from single-pair eddies to double-pair eddies, it was proposed in [32] that, for a curved square duct, the transition is a result of a complex structure of multiple, symmetric, and asymmetric solutions. The singular value of Dean number for the transition of the two- and four- vortex flow patterns is between 113 and 191, which is consistent with the current results obtained from the LBE model, i.e., the critical Dean number is between 115 and 130.

5.3.2 Heat transfer

In Figures 35-42 and Figures 43-50, the dimensionless temperature contours are presented at different Dean numbers and Prandtl numbers (two kinds of typical fluid, i.e., air and water, were chosen). From all these Figures, it is obvious that the Prandtl number plays a considerable role to the temperature field at a given Dean number and curvature ratio in a curved duct flow. For example, in Figures 35 and 36 or Figures 43 and 44, the fluid flow conditions are exactly same (i.e., same Dean number and curvature ratio) but the fluid physical properties (i.e., Prandtl number) are different. As a result, the dimensionless temperature contours are significantly different for these two cases with different Prandtl number. Similar results with different Dean numbers are presented in Figures 37-42 or Figures 45-50.

Fig. 35. Temperature contour of cross section (Dn=50.0 Pr=0.716 C=0.05)

Fig. 36. Temperature contour of cross section (Dn=50.0 Pr=6.587 C=0.05)

Fig. 37. Temperature contour of cross section ($Dn=100.0$ $Pr=0.716$ $C=0.05$)

Fig. 38. Temperature contour of cross section ($Dn=100.0$ $Pr=6.587$ $C=0.05$)

Fig. 39. Temperature contour of cross section ($Dn=150.0$ $Pr=0.716$ $C=0.05$)

Fig. 40. Temperature contour of cross section ($Dn=150.0$ $Pr=6.587$ $C=0.05$)

Fig. 41. Temperature contour of cross section ($Dn=200.0$ $Pr=0.716$ $C=0.05$)

Fig. 42. Temperature contour of cross section ($Dn=200.0$ $Pr=6.587$ $C=0.05$)

Fig. 43. Temperature profile in 3D view
($Dn=50.0$ $Pr=0.716$ $C=0.05$)

Fig. 44. Temperature profile in 3D view
($Dn=50.0$ $Pr=6.587$ $C=0.05$)

Fig. 45. Temperature profile in 3D view
($Dn=100.0$ $Pr=0.716$ $C=0.05$)

Fig. 46. Temperature profile in 3D view
($Dn=100.0$ $Pr=6.587$ $C=0.05$)

Fig. 47. Temperature profile in 3D view
($Dn=150.0$ $Pr=0.716$ $C=0.05$)

Fig. 48. Temperature profile in 3D view
($Dn=150.0$ $Pr=6.587$ $C=0.05$)

Fig. 49. Temperature profile in 3D view
($Dn=200.0$ $Pr=0.716$ $C=0.05$)

Fig. 50. Temperature profile in 3D view
($Dn=200.0$ $Pr=6.587$ $C=0.05$)

Figure 51 shows a comparison of Nusselt numbers between passive-scalar thermal LBE and Fluent software results with the same fluid flow and heat transfer parameters, including boundary conditions and geometry configurations. Before the thermal results by Fluent software were adopted in Figure 51, the grid convergence test was implemented for fluid flow and heat transfer. It is also apparent from this figure that the results of LBE are very consistent with those of Fluent software (i.e., the conventional CFD method).

Fig. 51. Nusselt number comparisons between LBM and Fluent results ($C=0.05$)

In this section, the fully developed fluid flow and heat transfer in a square duct with curvature ratio (0.05–1.0) and Dean number (0–200) have been thoroughly investigated using D3Q27 incompressible LBGK model and passive-scalar thermal model, respectively. Based on the simulation results, the following conclusions can be drawn:

1. The fluid flow simulation results of D3Q27 LBE model are in good agreement with experimental correlation at the given conditions;
2. The fully Dean stability range is obtained in this article and is consistent with results in the literature.
3. LBE is a new useful successive method to solve the hydrodynamic problems; compared to the conventional CFD approach, this method is very easy to understand and

implement. The unique parallelism characteristics can save a lot of calculation time if a special programming technique is applied.

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This book is planned to publish with an objective to provide a state-of-art reference book in the area of computational fluid dynamics for CFD engineers, scientists, applied physicists and post-graduate students. Also the aim of the book is the continuous and timely dissemination of new and innovative CFD research and developments. This reference book is a collection of 14 chapters characterized in 4 parts: modern principles of CFD, CFD in physics, industrial and in castle. This book provides a comprehensive overview of the computational experiment technology, numerical simulation of the hydrodynamics and heat transfer processes in a two dimensional gas, application of lattice Boltzmann method in heat transfer and fluid flow, etc. Several interesting applications area are also discusses in the book like underwater vehicle propeller, the flow behavior in gas-cooled nuclear reactors, simulation odour dispersion around windbreaks and so on.

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