

A Multi-Relaxation-Time Lattice Boltzmann Method on Non-Uniform Grids for Large Eddy Simulation of Rayleigh-Bénard Convection Using Two Sub-Grid Scale Models

A. R. Rahmati^{1†}, M. Ashrafizaadeh² and E. Shirani³

¹*Department of Mechanical Engineering, Faculty of Engineering, University of Kashan, Kashan, Iran*

²*Department of Mechanical Engineering, Isfahan University of Technology, Isfahan, Iran*

³*Foolad Institute of Technology, Fooladshahr, Isfahan, 8491663763, Iran*

†*Corresponding Author Email: ar_rahmati@kashanu.ac.ir*

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ABSTRACT

In the present work, for the first time, the application of a Multi-Relaxation-Time Lattice Boltzmann (MRT-LB) model for large-eddy simulation (LES) of turbulent thermally driven flows on non-uniform grids is considered. A Taylor series expansion and Least square based Lattice Boltzmann method (TLLBM) has been implemented in order to use a non-uniform mesh. It permits to reduce the required mesh size and consequently the computational cost to simulate the turbulent buoyant flow fields. The implementation is discussed in the context of a MRT-LB model in conjunction with both Smagorinsky and mixed scale viscosity sub-grid models. At first, to validate the code, a multi-relaxation-time lattice Boltzmann method on non-uniform grid is utilized to simulate a lid-driven cavity flow. Then large eddy simulation of this model is applied to simulate a turbulent Rayleigh-Bénard convection at different Rayleigh numbers in ranging 10^4 to 10^{15} for Prandtl number of 0.71. The simulation results show that lattice Boltzmann method is capable to simulate turbulent convection flow problems at high Rayleigh numbers.

Keywords: Lattice Boltzmann method, MRT-LBM, LES, TLLBM.

NOMENCLATURE

b	total number of lattice streaming vectors	\mathbb{V}	discrete velocity space
c	magnitude of lattice streaming vectors	\mathbf{u}	flow velocity vector
c_s	speed of sound	\hat{S}	relaxation matrix
C_S	Smagorinsky constant	\bar{S}	filtered mean momentum flux
D	dimension of system	S_{ij}	strain rate tensor
\mathbf{e}_α	lattice streaming vector in α direction	t	time
e	kinetic energy	T	temperature
f_α	density distribution function	x, y	Cartesian coordinate system
f_α^{eq}	equilibrium density distribution function	<i>Greek symbols</i>	
g	acceleration due to gravity	β	thermal expansion coefficient
L_y	height or vertical length	δr	lattice spacing
L_x	horizontal length	δt	time step
\mathbb{M}	moment space	ε	square of kinetic energy
M	transformation matrix	κ	fluid thermal diffusivity

m	moment	κ_t	turbulent heat diffusivity
p	pressure	ρ	local density
P_{ij}	second-order moments of the distribution function	τ_0	molecular relaxation time
Pr	v/α , Prandtl number	τ_t	the relaxation time corresponding to the turbulence or eddy viscosity ν_t
\mathbf{q}	heat flux vector	τ_{total}	effective relaxation time
Q_{ij}	momentum fluxes	ν_0	molecular kinematic viscosity
\bar{Q}	the characteristic filtered rate of strain	ν_t	turbulence or eddy viscosity
\mathbf{r}	position vector	ν_{total}	effective kinematic viscosity
Ra	$\frac{g\beta\Delta TL_y^3}{\nu\alpha}$, Rayleigh number	<i>Subscripts and superscripts</i>	
U	top lid velocity	eq	equilibrium state
u, v	x, and y components of the flow velocity vector	α	lattice streaming vector direction

1. INTRODUCTION

Thermally driven convection flow in enclosures which are heated from below (Rayleigh–Bénard convection) is an important problem that has been investigated for several decades (Siggia, 1994; Kadanoff, 2001; Kerr, 1996; Sreenivasan and Donnelly, 2001). The kinetic-based lattice Boltzmann method is a powerful tool for simulating fluid flows and modeling the physics in fluids (Chen and Doolen, 1998a; Rahmati and Ashrafizaadeh, 2010). However, the application of LB model to thermal problems has not attained great success for the thermal models due to the severe numerical instability caused by breaking the isothermal condition (Lallemand and Luo, 2003). The existing approaches for creating thermal LB models can be categorized into three categories, i.e., the multi-speed lattice Boltzmann scheme (Teixeira *et al.*, 2000), the double-density-distribution-function lattice Boltzmann (DDDFLB) approach (Chen and Doolen, 1998b), and the hybrid thermal lattice Boltzmann (HTLB) technique (Lallemand and Luo, 2003). The multi-speed scheme is a straightforward extension of the isothermal LB models, in which only the density distribution function is used; the DDDFLB approach uses two different density distribution functions, one for the velocity field and the other for the internal energy field; the HTLB technique is similar to the DDDFLB approach except that the internal energy equation is solved by finite-difference methods, rather than by solving the lattice Boltzmann method.

As mentioned above, the main difficulty of the thermal lattice Boltzmann models is their numerical instability. In the present work, for the first time, the important issues on the stability of lattice Boltzmann models have been studied and a stable combination model has been proposed and applied to simulate turbulent convective flows (Rahmati and Ashrafizaadeh, 2010). Using various thermal LB models, different numerical simulations have been performed to study 2D Rayleigh–Bénard convection

(Shan, 1997; Nwachok *et al.*, 2010; Kao and Yang, 2007). Although the results provided by these studies for low Rayleigh number are in good agreement with the CFD existing data, however, investigation about this flow at high Rayleigh numbers was not conducted. The present study employs, for the first time, HTLB model, which is based on the multi-relaxation-time lattice Boltzmann method proposed by Du *et al.* (2006), with the Boussinesq approximation to simulate turbulent Rayleigh–Bénard convection on a non-uniform mesh using Taylor series expansion–and least square–based Lattice Boltzmann method (Niu *et al.*, 2003).

2. GOVERNING EQUATION

2.1 Multi-Relaxation-Time Lattice Boltzmann Method for Large Eddy simulation

The lattice Boltzmann method utilizes a particle distribution function to describe the collective behavior of fluid molecules. In LB methods, particles are assumed to move synchronously along the bonds of a regular lattice, and satisfy the discrete form of the lattice Boltzmann equation. Perhaps, the simplest and consequently the most popular form of the lattice Boltzmann method is the Single-Relaxation-Time (SRT) lattice Boltzmann method (Chen and Doolen, 1998a). However, this simplicity comes at the expense of some deficiencies (e.g. numerical instability and inaccuracy in implementing boundary conditions). To overcome some of the SRT-LBM deficiencies, the multi-relaxation-time lattice Boltzmann method (Rahmati and Ashrafizaadeh, 2010) has been developed.

The multi-relaxation-time lattice Boltzmann equation can be written as (Chen and Doolen, 1998a; Du *et al.*, 2006):

$$\left| f(\mathbf{r} + \mathbf{e}_\alpha \delta t, t_n + \delta t) \right| - \left| f(\mathbf{r}, t_n) \right| = -M^{-1} S \left[\left| m(\mathbf{r}, t_n) \right| - \left| m^{(eq)}(\mathbf{r}, t_n) \right| \right], \quad (1)$$

where elements of the diagonal matrix \hat{S} are relaxation rates $\{s_\alpha \mid \alpha=1,2,\dots,(b+1)\}$, $|m^{(eq)}\rangle$ is the equilibrium-moment vector and its components are the equilibria of the moments, $|f(\mathbf{r}, t_n)\rangle$ is the distribution function vector and M is transformation matrix.

For D2Q9 model (Fig. 1), the lattice velocities are defined as:

$$\mathbf{e}_\alpha = \begin{cases} 0, & \alpha=0, \\ \left(\cos[(\alpha-1)\frac{\pi}{2}], \sin[(\alpha-1)\frac{\pi}{2}] \right) c, & \alpha=1,2,3,4, \\ \left(\cos[(\alpha-\frac{9}{2})\frac{\pi}{2}], \sin[(\alpha-\frac{9}{2})\frac{\pi}{2}] \right) c, & \alpha=5,6,7,8. \end{cases} \quad (2)$$

where c is the particle velocity and is given by $\sqrt{3RT_m}$ where R is the gas constant and T_m is average value of temperature.

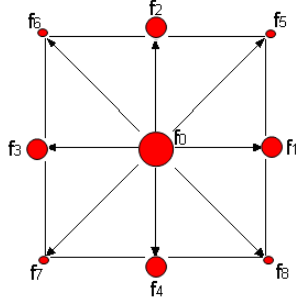


Fig. 1. Nine- particle velocity \mathbf{e}_α in the 2D square lattice

The transformation matrix M for D2Q9 model is given by:

$$M = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ -4 & -1 & -1 & -1 & -1 & 2 & 2 & 2 & 2 \\ 4 & -2 & -2 & -2 & -2 & 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & -1 & 0 & 1 & -1 & -1 & 1 \\ 0 & -2 & 0 & 2 & 0 & 1 & -1 & -1 & 1 \\ 0 & 0 & 1 & 0 & -1 & 1 & 1 & -1 & -1 \\ 0 & 0 & -2 & 0 & 2 & 1 & 1 & -1 & -1 \\ 0 & 1 & -1 & 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & -1 & 1 & -1 \end{pmatrix}. \quad (3)$$

The relaxation matrix \hat{S} is diagonal in the moment space \mathcal{M} :

$$\hat{S} \equiv \text{diag}(1 \quad s_1 \quad s_2 \quad 1 \quad s_4 \quad 1 \quad s_6 \quad s_7 \quad s_8). \quad (4)$$

The nine moments for D2Q9 model are:

$$|m\rangle = (\rho_0 \quad e \quad \varepsilon \quad u_x \quad q_x \quad u_y \quad q_y \quad p_{xx} \quad p_{xy})^T, \quad (5)$$

where ρ_0 the mass density, e the energy, ε square of the energy, $\mathbf{u} = (u_x \quad u_y)$ the momentum density, and $\mathbf{q} = (q_x \quad q_y)$ is the heat flux. Also, p_{xx} , and p_{xy} are the stresses.

The equilibria of the moments, $m^{(eq)}$, are the functions of the invariant moments, which are the mass density ρ_0 and the velocity vector \mathbf{u} for athermal fluids. However, in the current method the pressure p has been used instead of the mass density ρ_0 , i.e., $m^{(eq)}(\mathbf{r}, t_n) = m^{(eq)}(p(\mathbf{r}, t_n), u(\mathbf{r}, t_n))$. For the D2Q9 model, the equilibria for the non-invariant moments are given by:

$$\begin{aligned} e^{(eq)} &= 6p + 3(u_x u_x + u_y u_y), \\ \varepsilon^{(eq)} &= -9p - 3(u_x u_x + u_y u_y), \\ q_x^{(eq)} &= -u_x, \\ q_y^{(eq)} &= -u_y, \\ p_{xx}^{(eq)} &= \frac{1}{3}(u_x^2 - u_y^2), \\ p_{xy}^{(eq)} &= \frac{1}{3}u_x u_y, \end{aligned} \quad (6)$$

The kinematic viscosity ν of the model is given by:

$$\nu = \frac{1}{3} \left(\frac{1}{s_8} - \frac{1}{2} \right) = \frac{1}{3} \left(\frac{1}{s_7} - \frac{1}{2} \right), \quad (7)$$

2.2 Multi-Relaxation-Time Lattice Boltzmann Method for Large Eddy simulation

Large eddy simulation of multi-relaxation-time lattice Boltzmann method results in the filtered form of the MRT-LB method (Krafczyk *et al.*, 2003; Van Treeck *et al.*, 2006) and it is given below:

$$\begin{aligned} &|\bar{f}(\mathbf{r} + \mathbf{e}_\alpha \delta t, t_n + \delta t)\rangle - |\bar{f}(\mathbf{r}, t_n)\rangle = \\ &-M^{-1} S_{total} \left[|\bar{m}(\mathbf{r}, t_n)\rangle - |\bar{m}^{(eq)}(\mathbf{r}, t_n)\rangle \right], \end{aligned} \quad (8)$$

where $|\bar{f}(\mathbf{r}, t_n)\rangle$, $|\bar{m}(\mathbf{r}, t_n)\rangle$, $|\bar{m}^{(eq)}(\mathbf{r}, t_n)\rangle$, and

S_{total} represent the distribution function vector, the moment vector, the equilibrium moment vector of the resolved scales and effective relaxation time matrix, respectively.

We note that the second-order moments of the distribution function,

$$P_{ij} = \sum_{\alpha} e_{\alpha i} e_{\alpha j} f_{\alpha} = p \delta_{ij} + u_i u_j - \frac{1}{s_{xx}} 2c_s^2 S_{ij}, \quad (9)$$

where $e_{\alpha i}$ denotes the i -th Cartesian component of a discrete velocity \mathbf{e}_{α} , are in fact related to the second-order moments p_{xx} and p_{xy} . In above formula, s_{xx} is the relaxation rate for these second-order moments and c_s is the sound speed. (In the setting of SRT-LB model, $s_{xx} = 1/\tau$) Therefore,

$$S_{ij} = \frac{s_{xx}}{2c_s^2} [p \delta_{ij} + u_i u_j - P_{ij}] = -\frac{s_{xx}}{2c_s^2} Q_{ij}. \quad (10)$$

The second-order monomials $\{e_{\alpha i} e_{\alpha j} | i, j \in \{x, y\}\}$ can be projected to the orthogonal basis vectors $\{|\phi_{\beta}\rangle | \beta = 0, 1, \dots, N\}$, that is, the dual eigenvectors of M :

$$\begin{aligned} e_{\alpha x}^2 &= \frac{1}{6} [|\phi_1\rangle_{\alpha} + 4|\phi_0\rangle_{\alpha} + 3|\phi_7\rangle_{\alpha}], \\ e_{\alpha y}^2 &= e_{\alpha x}^2 - |\phi_7\rangle_{\alpha}, \\ e_{\alpha x} e_{\alpha y} &= |\phi_8\rangle_{\alpha}. \end{aligned} \quad (11)$$

Thus in this case the components of tensor Q can be explicitly given in terms of the moments:

$$\begin{aligned} Q_{xx} &= P_{xx} - (p + u_x u_x), \\ Q_{yy} &= P_{yy} - (p + u_y u_y), \\ Q_{xy} &= P_{xy} - u_x u_y, \\ Q_{xy} &= Q_{xy}, \\ P_{xx} &= \frac{1}{6} [e + 4p + 3p_{xx}], \\ P_{yy} &= -p_{xx} + P_{xx}, \\ P_{xy} &= p_{xy}. \end{aligned} \quad (12)$$

The effect of the unresolved scales is modeled through an effective relaxation time scale τ_t . Thus in Eq. (1) the total LES effective relaxation time should be $\tau_{total} = \tau_t + \tau_0$ where τ_0 and τ_t are the relaxation times corresponding to the molecular viscosity ν_0 and the turbulence or eddy viscosity ν_t , respectively. Accordingly ν_{total} is given by (Krafczyk *et al.*, 2003; Van Treeck *et al.*, 2006):

$$\begin{aligned} \nu_{total} &= \nu_0 + \nu_t = \\ \frac{1}{3} \left(\tau_{total} - \frac{1}{2} \right) c^2 \delta t &= \frac{1}{3} \left(\tau_0 + \tau_t - \frac{1}{2} \right) c^2 \delta t, \\ \nu_t &= \frac{1}{3} \tau_t c^2 \delta t, \end{aligned} \quad (13)$$

where ν_t depends on the sub-grid model used in the simulation. We use the Smagorinsky model and the mixed scale viscosity model for sub-grid closure.

In the Smagorinsky model, the eddy viscosity ν_t is calculated from the filtered strain rate tensor, and a filter length scale, $\bar{\Delta}$, as:

$$\begin{aligned} \nu_t &= (C_s \bar{\Delta})^2 |\bar{S}|, \\ |\bar{S}| &= \frac{|\bar{Q}|}{2\rho_0 c_s^2 \tau_{total}}, \quad |\bar{Q}| = \sqrt{2\bar{Q}_{xy} \bar{Q}_{xy}} \end{aligned} \quad (14)$$

where \bar{S} and \bar{Q} are the characteristic filtered rate of strain and the filtered mean momentum flux, respectively, and C_s is the Smagorinsky constant.

The mixed scale viscosity model stems from a Smagorinsky model in which the local adaptation is achieved by taking into account the kinetic energy at the cut-off, q_c , explicitly. The sub-grid viscosity is then measured as (Sergent *et al.*, 2006):

$$\nu_t = 0.07 |\bar{S}|^2 (q_c)^{\frac{1}{4}} \bar{\Delta}^{\frac{3}{2}} \quad (15)$$

where q_c stands for the kinetic energy at the cut-off,

$$q_c = \frac{1}{2} \bar{u}'_i \bar{u}'_i. \quad \text{Following Bardina's similarity}$$

hypothesis, the velocity field at the cut-off, \bar{u}'_i , can be estimated by filtering the resolved velocity field with a test filter coarser than the implicit one, $\tilde{\Delta} > \bar{\Delta}$, and so $\bar{u}' = \bar{u} - \tilde{u}$.

The subgrid-scale dependency of this model ensures that it will adapt to the local state of the flow, and vanish in fully resolved regions of the flow and near the walls.

2.3 Accomplishment on Non-Uniform Grid

In standard lattice Boltzmann method, the grid is described as a regular lattice with identical spaces. But for high Rayleigh number flows, the thermal boundary layer is very thin. Hence it needs sizeable number of nodes which waste computational time and memory size. The Taylor series expansion and least square based lattice Boltzmann method (Niu *et al.*, 2003) is based on the reality that density distribution function is a continuous function in the physical space and can be well defined in any mesh system. It is obtained from the standard LBM by using Taylor series expansion and optimized by the least squares technique. This method updates the density distribution functions at mesh points by an algebraic formulation and the pertinent coefficients are pre-calculated from the coordinates of

mesh points. TLLB method allows mesh refinement near the walls.

Let us to consider a particle is initially at the grid point (x, y, t) . Along the i -direction, this particle will propagate to the position $(x + e_{\alpha x} \delta t, y + e_{\alpha y} \delta t, t + \delta t)$. For a uniform lattice, $\delta x = e_{\alpha x} \delta t$, $\delta y = e_{\alpha y} \delta t$. So, $(x + e_{\alpha x} \delta t, y + e_{\alpha y} \delta t)$ is on the grid point. In other words Eq. (1) can be used to update the density distribution functions exactly at the grid points. However, for a non-uniform grid, $(x + e_{\alpha x} \delta t, y + e_{\alpha y} \delta t)$ is usually not at the grid point $(x + \delta x, y + \delta y)$. In the numerical simulation, only the density distribution function at the mesh points for all the time levels are needed, so that the macroscopic properties such as the density, flow velocity and temperature can be evaluated at every mesh point. To get the density distribution function at the grid point $(x + \delta x, y + \delta y)$ and the time level $t + \delta t$, the Taylor series expansion in the spatial direction is applied.

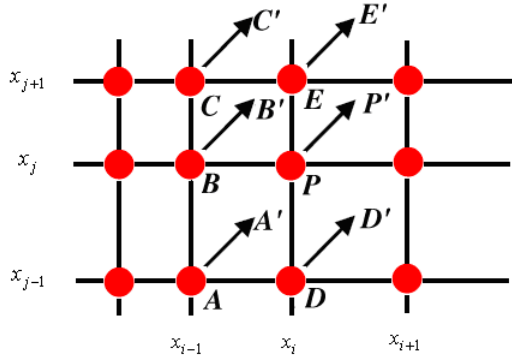


Fig. 2. Configuration of particle movement along the α -direction.

As shown in Fig. 2, for simplicity, the point A represents the grid point (x_A, y_A, t) , point A' represents the position $(x_A + e_{\alpha x} \delta t, y_A + e_{\alpha y} \delta t, t + \delta t)$, and point P represents the position $(x_P, y_P, t + \delta t)$ with $x_P = x_A + \delta x$, $y_P = y_A + \delta y$. So Eq. (1) gives

$$f_\alpha(A', t + \delta t) = f_\alpha(A, t) + \frac{f_\alpha^{eq}(A, t) - f_\alpha(A, t)}{\tau}. \quad (16)$$

For the general case, A' may not coincide with the mesh point P. We truncate the Taylor series expansion to the second order derivative terms. So $f_\alpha(A', t + \delta t)$ can be approximated by the corresponding function and its derivatives at the mesh point P as

$$\begin{aligned} f_\alpha(A', t + \delta t) &= f_\alpha(P, t + \delta t) + \\ &\Delta x_A \frac{\partial f_\alpha(A, t + \delta t)}{\partial x} + \Delta y_A \frac{\partial f_\alpha(A, t + \delta t)}{\partial y} + \\ &\frac{1}{2} (\Delta x_A)^2 \frac{\partial^2 f_\alpha(A, t + \delta t)}{\partial x^2} + \frac{1}{2} (\Delta y_A)^2 \frac{\partial^2 f_\alpha(A, t + \delta t)}{\partial y^2} + \\ &(\Delta x_A \Delta y_A) \frac{\partial^2 f_\alpha(A, t + \delta t)}{\partial x \partial y} + O[(\Delta x_A)^3, (\Delta y_A)^3] \end{aligned} \quad (17)$$

where $\Delta x_A = x_A + e_{\alpha x} \delta t - x_P$, $\Delta y_A = y_A + e_{\alpha y} \delta t - y_P$. For the two-dimensional case, this expansion involves six unknowns, that is, one density distribution function at the time level $t + \delta t$, two first-order derivatives, and three second-order derivatives. To solve for these unknowns, six equations are needed to close the system. This can be done by applying the second-order Taylor series expansion at six points: P, A, B, C, D, E. The following equation system can be obtained:

$$f_k' = \{s_k\}^T \{W\} = \sum_{i=1}^6 s_{ki} W_i, \quad k = P, A, B, C, D, E, \quad (18)$$

where

$$\begin{aligned} f_k' &= \\ f_\alpha(x, y, t) &+ \frac{f_\alpha^{eq}(x, y, t) - f_\alpha(x, y, t)}{\tau}, \\ \{s_k\}^T &= \\ \{1, \Delta x_k, \Delta y_k, (\Delta x_k)^2 / 2, (\Delta y_k)^2 / 2, \Delta x_k \Delta y_k\}, \\ \{W\} &= \\ \{f_\alpha, \partial f_\alpha / \partial x, \partial f_\alpha / \partial y, \partial^2 f_\alpha / \partial x^2, \partial^2 f_\alpha / \partial y^2, \partial^2 f_\alpha / \partial x \partial y\}^T. \end{aligned} \quad (19)$$

Our aim is to find its first element $W_1 = f_\alpha(P, t + \delta t)$. Eq. (18) can be put into the following matrix form

$$[S] \{W\} = \{f'\}. \quad (20)$$

In practical applications, it was found that the matrix [S] might be singular or ill-conditioned. To overcome this difficulty and make the method be more general, the least squares approach was introduced to optimize the approximation by Eq. (20). As a result, the equation system for {W} becomes

$$\{W\} = ([S]^T [S])^{-1} [S]^T \{f'\} = [A] \{f'\}. \quad (21)$$

From Eq. (21), we can have

$$\begin{aligned} f_\alpha(x_0, y_0, t + \delta t) &= W_1 = \sum_{k=1}^{M+1} a_{1,k} f_{k-1}', \\ |f(\mathbf{r}_p, t + \delta t)| &= \sum_{k=1}^{M+1} a_{1,k} |f_{k-1}'|, \\ |f_k'| &= |f(\mathbf{r}_k, t_n) - M^{-1} S [m(\mathbf{r}_k, t_n)] - m^{eq}(\mathbf{r}_k, t_n)|. \end{aligned} \quad (22)$$

where $a_{1,k}$'s are the elements of the first row of the matrix A, which are defined by the coordinates of mesh points, the particle velocity and time step. They are not varied throughout the calculation. We can calculate once and store in advance, so little computational effort is introduced as compared with the standard LB method. On the other hand, Eq. (22) is nothing to do with the mesh structure. It needs only the information of coordinates of the mesh points. Thus, we can say that Eq. (22) can be consistently used for any kind of mesh structure.

2.4 HTLB Scheme with a Sub-grid Model

In the HTLB method (Van Treec *et al.*, 2006), the flow field is solved by the isothermal lattice Boltzmann equation while the advection-diffusion equation for temperature is solved separately by a finite difference scheme. The equation is used for solving the flow field is given by Eq. (8). The equation used for solving the temperature field is given by:

$$\frac{\partial \bar{T}}{\partial t} + \bar{u} \frac{\partial \bar{T}}{\partial x} + \bar{v} \frac{\partial \bar{T}}{\partial y} = (\kappa + \kappa_t) \left(\frac{\partial^2 \bar{T}}{\partial x^2} + \frac{\partial^2 \bar{T}}{\partial y^2} \right), \quad (23)$$

where κ is the fluid thermal diffusivity and κ_t is turbulent heat diffusivity. The energy equation is solved by a finite difference discretization on the non-uniform mesh. Both velocity components, \bar{u} and \bar{v} are obtained from Eq. (8). The relation between turbulent viscosity ν_t and the turbulent heat diffusivity κ_t is given by the turbulent Prandtl number, i.e. $Pr_t = \nu_t / \kappa_t$ which is assumed to be constant and equal to its theoretical value, i.e. 0.87 (Van Treec *et al.*, 2006).

3. RESULTS AND DISCUSSION

In the present study, an incompressible LBM solver is developed. At first a cavity driven flow is simulated to validate the code for simulation of isothermal incompressible flows on non-uniform grid. Then a Rayleigh-Bénard convective flow has been simulated at different Rayleigh numbers with Prandtl number of 0.71.

3.1 Lid-driven cavity flow

Using the presented 2D TLLBM, numerical simulation of lid-driven flows in a cavity are considered for $Re=1000$, where $Re = U L_y / \nu$ is the Reynolds number,

based on the lid velocity and the height of the cavity. The configuration of the problem is shown in Fig. 3. This problem has been widely used for validation and comparison purposes. Non-uniform mesh of 101×101 for $Re=1000$ is used. A typical non-uniform mesh in a square cavity for a 101×101 grid with stretch ratio of 2.54, which is defined as the ratio of the maximum mesh spacing over the minimum mesh spacing, is shown in Fig. 4. The employment of non-uniform mesh

is popular, particularly for simulation of the flow field at higher Reynolds numbers. This is because the boundary layer near the solid boundaries is thin. Hence, the mesh spacing near the wall should be very small to capture the thin boundary layer. The mesh spacing apart from the solid wall can be use relatively large. In this method, we can capture the thin boundary layer, and in the meantime, we can save the computational effort. Furthermore, the fluctuations which are created in the pressure domain at high Reynolds numbers can be reduced due to a decrease in the time step used for the simulation.

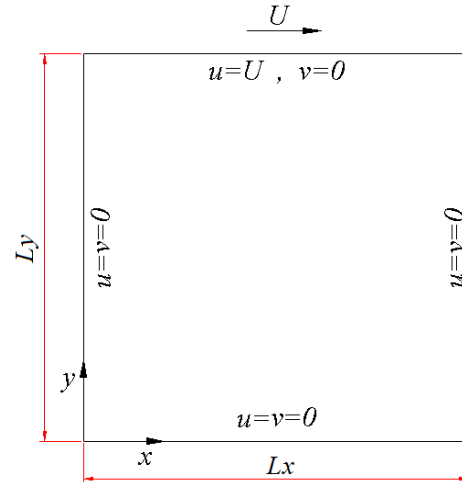


Fig. 3. 2D cavity flow.

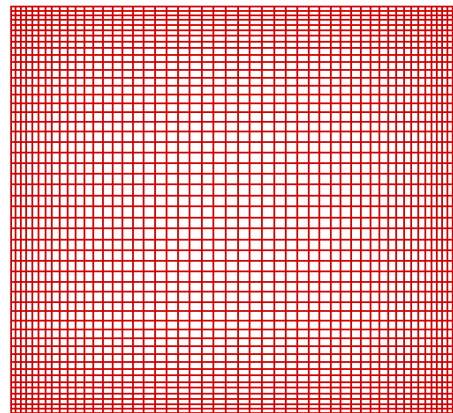


Fig. 4. The solution grid (a 101×101 non-uniform grid with stretch ratio of 2.54)

Initially, a constant pressure, $p = 1/3$, is prescribed in the whole field, and the velocities in the interior of the cavity are set to zero. On the top, the x component of the velocity is U , which is set to 0.19, and the y component of the velocity is zero. At the end of each time step, the density distribution function f_α on the moving wall is obtained by non-equilibrium boundary condition. The node wall bounce back boundary conditions are used on the other three stationary walls.

All results which are putted on show in the current work are steady state solutions. In order to reach the steady state, a number of iterations are carried out. The criterion to steady sate is

$$\sqrt{\frac{1}{N} \sum_{i=1}^N (u^{n+1} - u^n)^2 + (v^{n+1} - v^n)^2} \leq 10^{-10}, \quad (24)$$

where N is the total number of nodes in the solution domain; (u^n, u^{n+1}) and (v^n, v^{n+1}) are x- and y-component of the velocity for the old and new time levels, respectively.

Figures 5 and 6 present the U and V velocities along their respective centerline of the cavity flow at $Re = 1000$, respectively. As shown in those figures, the agreement between the Ghia's results (Ghia *et al.*, 1982) and the present numerical results is very good.

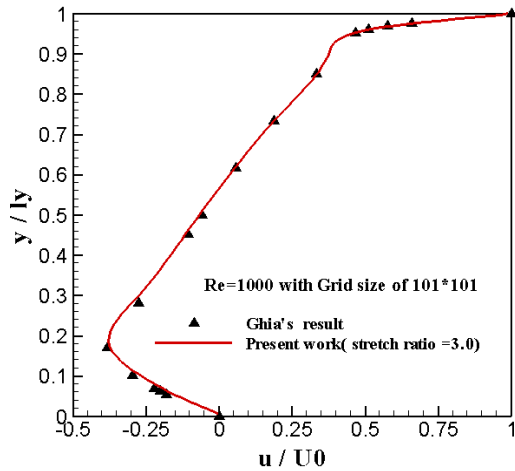


Fig. 5. The x-component velocity along vertical centerline cavity driven flow for $Re=1000$.

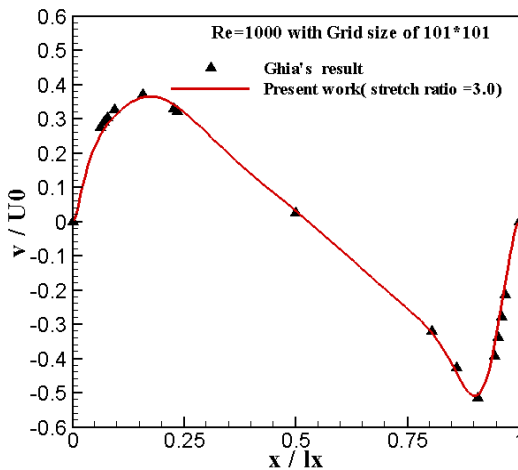


Fig. 6. The y-component velocity along horizontal centerline cavity driven flow for $Re=1000$.

3.2. Rayleigh-Bénard convection flow

3.2.1. Description of the problem

Rayleigh-Bénard convection occurs in a wide variety of engineering applications. The flows are driven by the buoyancy effect due to the presence of gravity and density variations from low layer to another. When difference of temperature is lower, the flow is stationary. As difference of temperature is increased, the flow motion is regular and steady. As Ra is increased beyond the second critical Rayleigh number, fluctuation and instability of flows can be observed. In this natural convection problem, the initially static flow is heated from the bottom boundary, and a lower temperature is maintained at the upper wall. Additionally, a vertical gravitational force is applied to the y-direction of the computational domain. As the temperature difference between upper and bottom boundaries is increased, the stationary conduction state becomes unstable by any small disturbance.

According to linear stability theory, the critical wave number for Rayleigh-Bénard convection is $a_c = 3.117$. So that, the convective cells would be developed readily with an aspect ratio of $2\pi/a_c = 2.016$. An appropriate value of aspect ratio, $AR \equiv L_x / L_y = 2$, for simulations can be determined.

The temperature difference between the hot and cold walls introduces a temperature gradient in a fluid, and the consequent density difference induces a fluid motion, that is, convection. In the simulation, the Boussinesq approximation is applied to the buoyancy force term.

$$\rho \mathbf{G} = \rho \beta g T - T_m \mathbf{j}, \quad (25)$$

where β is the thermal expansion coefficient, g is the acceleration due to gravity, T_m is the average temperature and \mathbf{j} is the vertical direction opposite to that of gravity.

The dynamical similarity depends on two dimensionless parameters: the Prandtl number Pr and the Rayleigh number Ra defined as:

$$Pr = \frac{\nu}{\alpha}, Ra = \frac{g \beta \Delta T L_y^3}{\nu \alpha}, \quad (26)$$

where L_y is the channel height, and ΔT is the applied temperature difference between the bottom and top walls.

3.2.2. Numerical Results

3.2.2.1. Laminar flow simulation

Using the HT-LB model, we simulate Rayleigh-Bénard convection between two horizontal stationary walls for $Ra=10^4$ with a $Pr = 0.71$ on different size of lattice (Fig. 7). To capture the thin boundary layer, a non-uniform mesh distribution is used (Fig. 8). Non-slip boundary

conditions are implemented at the bottom and top boundaries by reversing the flow velocities in the ghost cell next to the simulation domain. The temperature along two opposite horizontal walls is maintained at constant temperatures 0 and 1, respectively. Periodic boundary conditions are assigned for both the flow and the temperature fields along the side boundaries of the channel, see Fig. 7. A small perturbation in the form of a cosine wave with amplitude of 1×10^{-3} is used to the pressure population. A linear distribution applied to the temperature field. Hence, at initial condition temperature is given by:

$$T = T_{cold} - \frac{y - l_y}{l_y} (T_{hot} - T_{cold}). \quad (27)$$

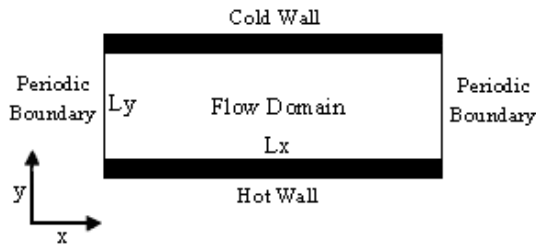


Fig. 7. Schematic of 2D enclosure Rayleigh- Bénard convection cell and boundary condition.

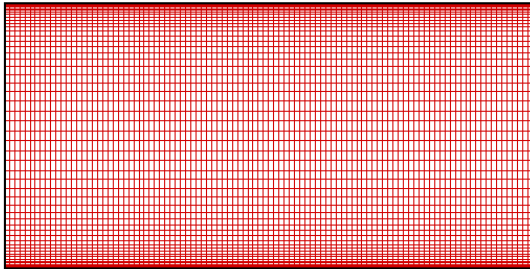


Fig. 8. The solution grid (a 101×51 non-uniform grid with stretch ratio of 3.15).

The Nusselt number can be calculated by the following equation:

$$Nu = 1 + \frac{\langle vT \rangle}{\alpha \Delta T / L_y}, \quad (28)$$

where L_y is the vertical height of the channel, ΔT is fixed and denotes the temperature difference between the upper and bottom boundaries, v is the flow velocity in the y -direction and $\langle \cdot \rangle$ denotes the average over the whole flow domain. Table 1 presents the calculated Nusselt numbers using HT-LB model at $Ra = 10^4$, as well as the values obtained by the empirical formula (Chen and Doolen, 1998b; Shan, 1997), $Nu = 1.56(Ra / Ra_c)^{0.296}$ where Ra_c is the critical

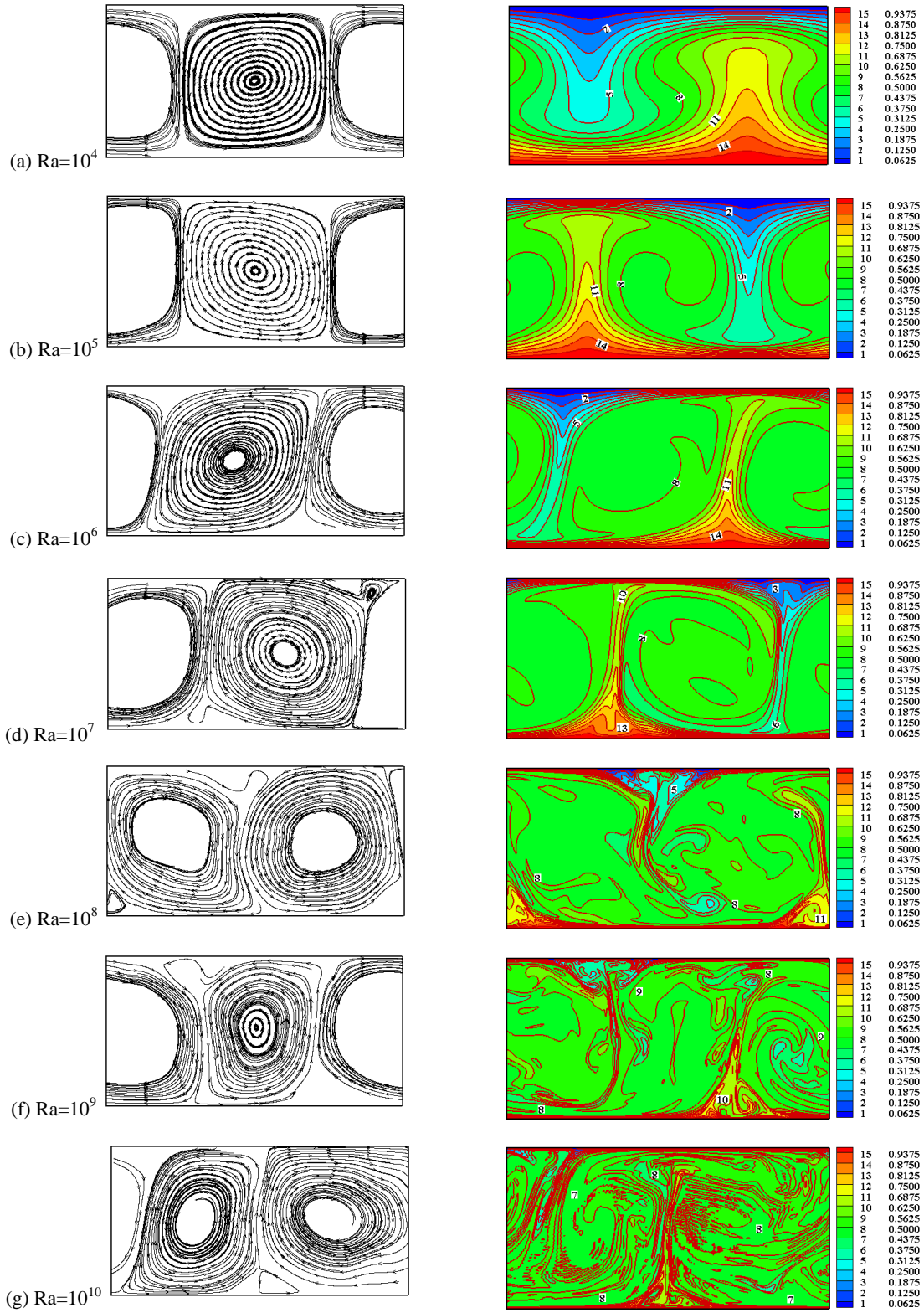
Rayleigh number and is equal to $Ra_c = 1707.76$. From the presented results in Table 1, it can be seen that as the size of the grid is increased from 35×17 to 141×71 , the discrepancy between the computed quantities is reduced from 14.0% to 0.65% for Nusselt number. This indicates that the employment of 101×51 grid can give accurate numerical results for $Ra = 10^4$.

Table 1 Grid dependence study for the Rayleigh-Bénard convection at $Ra = 10^4$.

Ra no.	Size of mesh	Present work	$Nu = 1.56(Ra / Ra_c)^{0.296}$
10^4	35×17	2.264	2.632
10^4	71×35	2.423	2.632
10^4	101×51	2.609	2.632
10^4	141×71	2.615	2.632

3.2.2.2. Turbulent flow simulation

In this study, the hybrid thermal lattice Boltzmann scheme is used along with the large-eddy approach to simulate turbulent convective flows. For the mass and momentum equations, a multi-relaxation-time LB technique is used while the heat equation is solved by a finite difference scheme. Furthermore, we applied the hybrid model by two subgrid scale models for both the fluid flow and the heat flux. Using HTLB-LES model, we simulate turbulent Rayleigh-Bénard convection between two horizontal stationary walls for $Ra = 10^4 - 10^{15}$ with $Pr = 0.71$ (Fig. 7). To capture the thin boundary layer, a non-uniform mesh distribution is used. As shown in Fig. 8, non-uniform meshes of 101×51 for $Ra = 10^4$, 201×101 for $Ra = 10^5$, 301×151 for $Ra = 10^6$, 361×181 for $Ra = 10^7$, and 401×201 for $Ra = 10^8 - 10^{15}$, respectively. The stretch ratio is 3.15 for all cases. No-slip boundary conditions are performed at the bottom and top walls by using the mid-wall bounce-back scheme in LB simulation. The temperatures along two opposite horizontal walls are maintained constant, 0 and 1. Periodic boundary conditions are assigned for both the flow and the temperature fields at the inlet and outlet of the channel, see Fig. 7, Figs. 9 and 10 show streamlines and temperature contours at different Ra numbers. As shown in Figs. 9 and 10, the hot fluid near the bottom wall moves upward and intensifies the temperature in the central part of the flow domain, while the cold fluid close to the top wall streams downward and reduces the temperature near the side boundaries. When the Rayleigh number increases, two tendencies were perceived for the temperature contour distributions: 1) an increase in the temperature gradients close to the top and bottom boundaries and 2) intensified mixing of the hot and cold fluids. Both inclinations enhance the heat transfer in the flow domain.



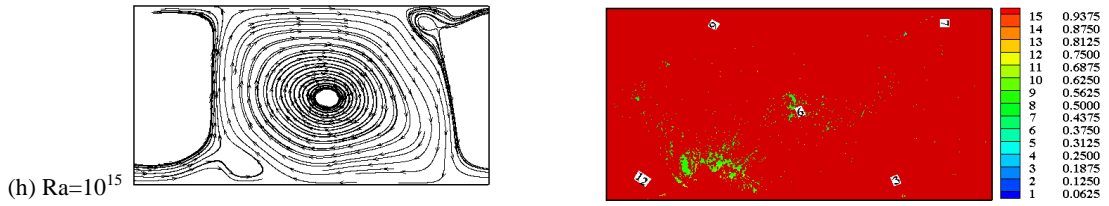
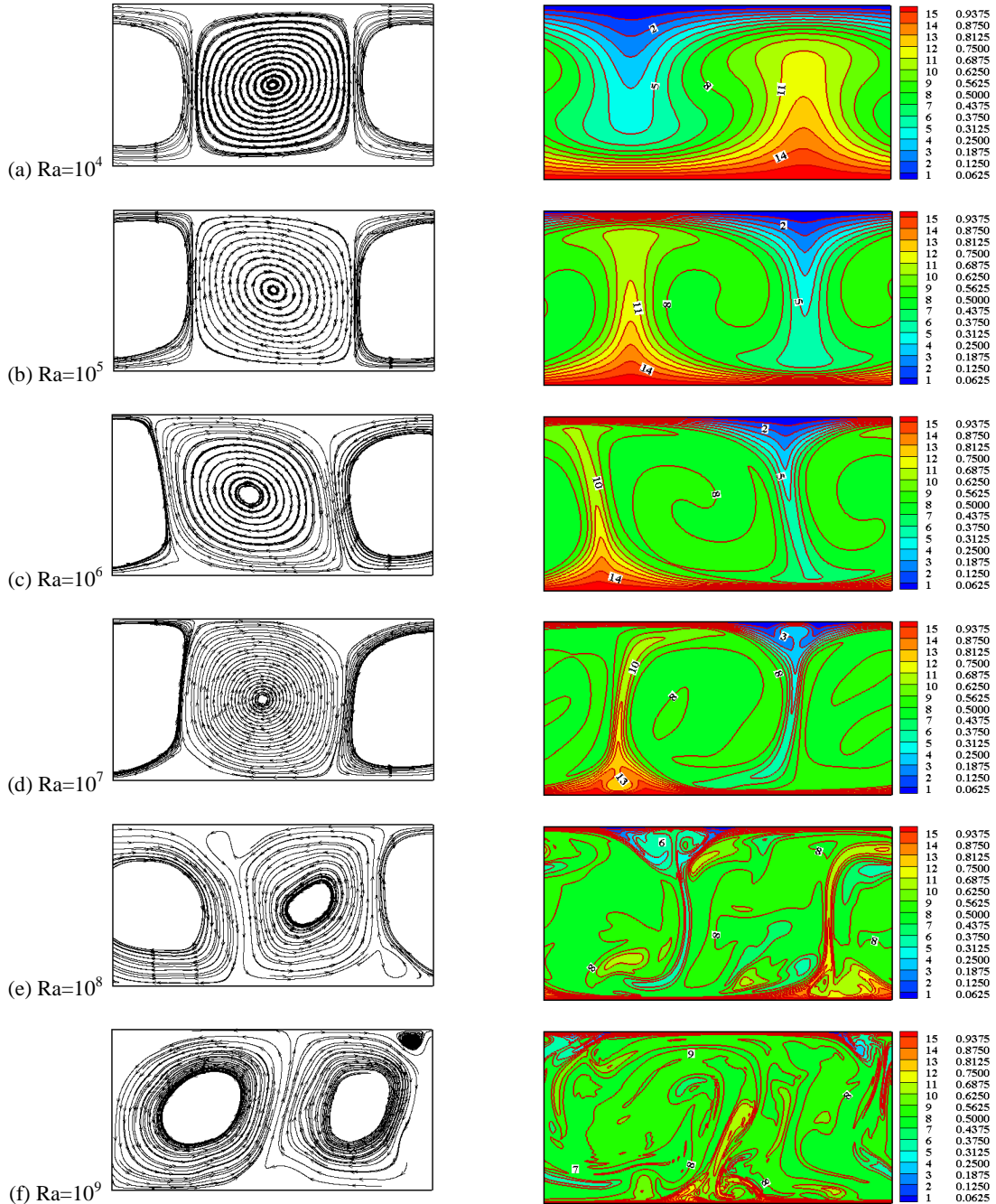


Fig. 9. Streamlines (left side) and temperature contours (right side) obtained by Smagorinsky model for Rayleigh-Bénard convection at the 8000000-th time step for different Rayleigh numbers.



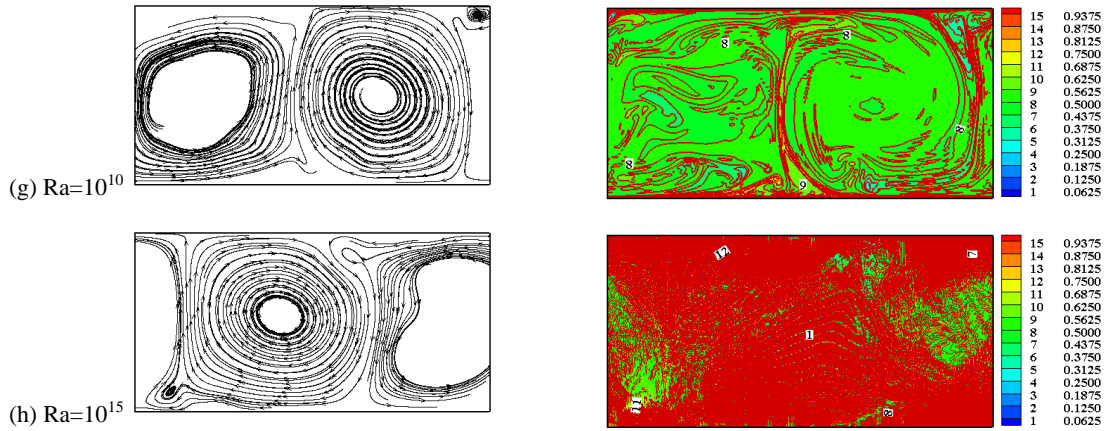


Fig. 10. Streamlines (left side) and temperature contours (right side) obtained by mixed scale model for Rayleigh-Bénard convection at the 8000000-th time step for different Rayleigh numbers.

Time evolution of the x and y components of the velocity vector, and the temperature obtained by both Smagorinsky model (left side) and mixed scale model (right side) at the mid of flow domain for $Ra=10^{10}$ is shown in Fig. 11. Furthermore, Fig. 12 presents power spectral density of the kinetic energy fluctuations at the mid of flow domain obtained by Smagorinsky model (left side) and mixed scale model (right side) for $Ra=10^{10}$. As shown in Fig. 12, slope of inertial subrange obtained by both Smagorinsky model and mixed scale model is in a better agreement with -3 power law compared to -5/3 power law (Frost and Moulden, 1977).

An important feature is the total heat transport that is described in terms of the Nusselt number $Nu = 1 + \langle \bar{v} \bar{T} \rangle L_y / \alpha \Delta T$. Here \bar{v} is the mean vertical velocity, and $\langle \rangle$ denotes the average over the whole domain. Table 2 presents the calculated Nusselt number at $Ra = 10^4 - 10^{10}$, as well as DNS results obtained by Kerr (1996). It is observed that our results are in good agreement with the empirical data for lower Rayleigh number (up to 10^4). As the Rayleigh number increases the computed Nusselt number is under predicted compared to the Nusselt number calculated by Kerr (1996). The under-prediction of the Nusselt number may be due to insufficient spatial resolution. Furthermore, it is observed that the results obtained using the mixed scale model, are in a better agreement with those reported in Ref. (Kerr, 1996), compared to the Smagorinsky model results.

4. CONCLUSION

In this work, a new LBM solver is proposed to simulate turbulent convection flows problems at high Rayleigh

numbers. In order to illustrate the capability of the code to simulate isothermal flows, a two-dimensional cavity driven flow has first been simulated at Reynolds number of 1000. Then its capability to solve the heat transfer problem in the case of the Rayleigh-Bénard convective flow up to $Ra=10^{15}$ is shown. In the code, a MRT-LB model in conjunction with both Smagorinsky and mixed scale viscosity sub-grid models is used to simulate a turbulent Rayleigh-Bénard convection at different Rayleigh numbers. The combination of the MRT-LB method with the subgrid model allows us to simulate the turbulent convection flows up to a Rayleigh number of 10^{15} . In the proposed code, a space-filtered density distribution function is defined. In Smagorinsky subgrid model, the energy dissipation caused by the interaction between resolved and unresolved scales is included. In current method the local strain intensity stress is calculated locally at each time step using the non-equilibrium density distribution functions. In order to get proper near-wall behaviour or to simulate flows which are not fully turbulent, the subgrid model should be able to adjust itself to locally inhomogeneous flow. The mixed scale viscosity model ensures that it will adapt to the local state of the flow, and vanish in fully resolved regions of the flow and near the walls. Results show that the proposed method produce reasonably accurate results at low Rayleigh numbers and stable results at high Rayleigh numbers. Besides, it is observed that the results obtained using the mixed scale model, are in a better agreement with existing CFD data, compared to the Smagorinsky model results.

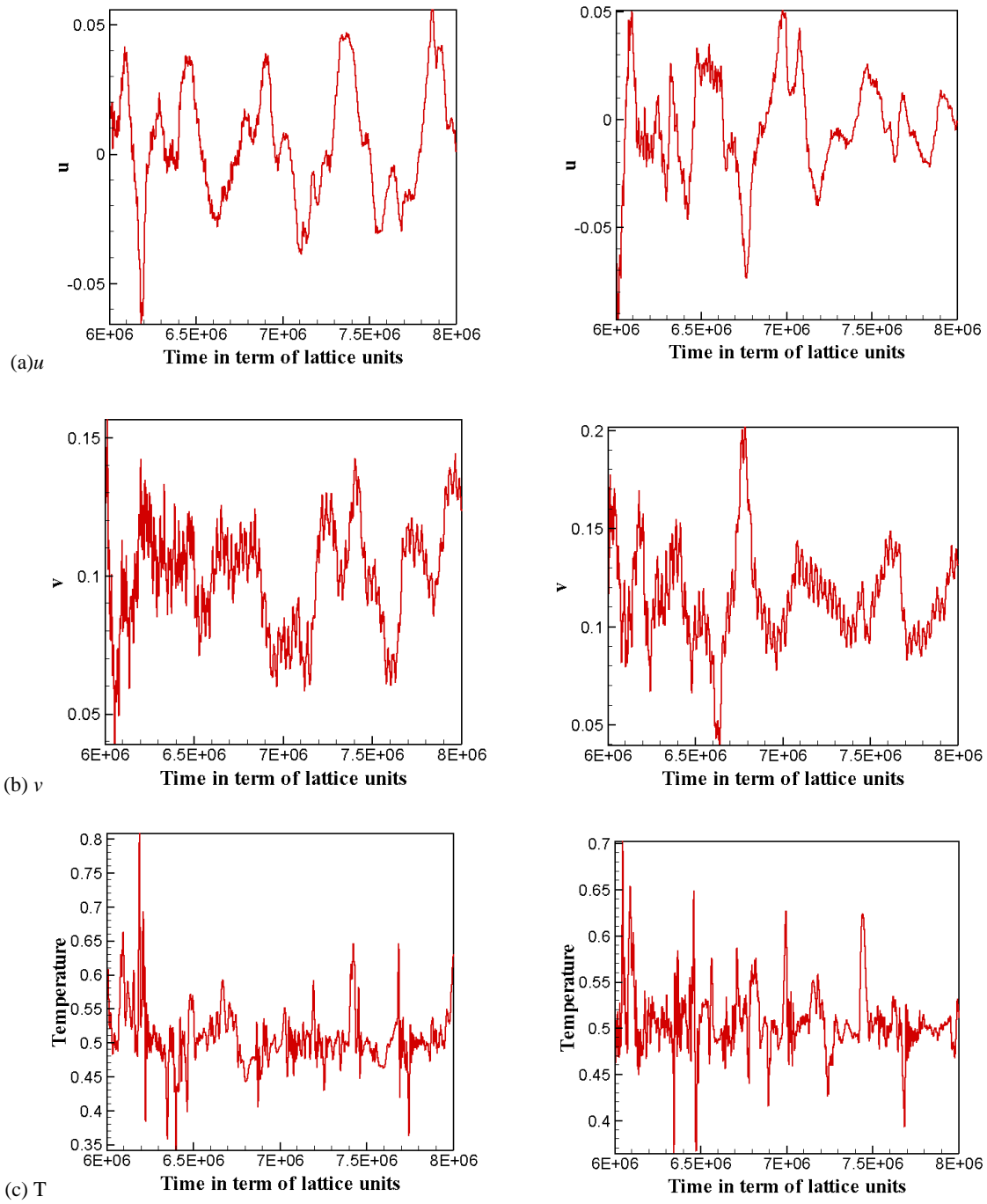


Fig. 11. Time evolution of the x and y components of the velocity vector, and the temperature obtained by both Smagorinsky model (left side) and mixed scale model (right side) at the mid of flow domain for $Ra=10^{10}$.

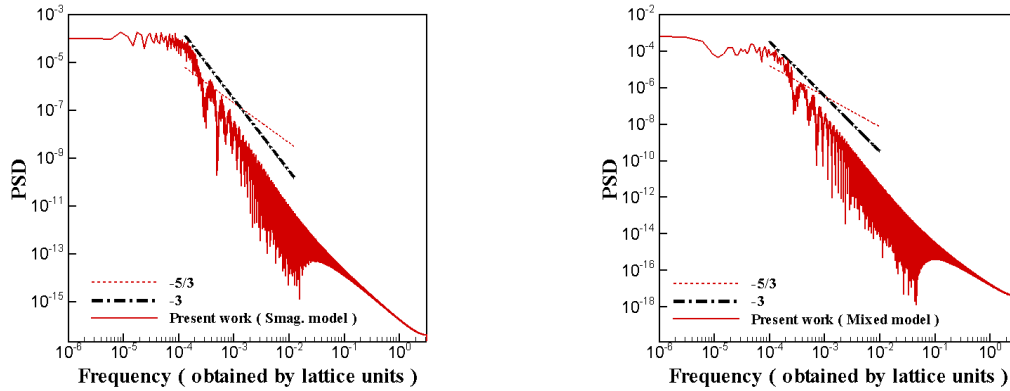


Fig. 12. Power spectral density of the kinetic energy fluctuations obtained by Smagorinsky model (left side) and mixed scale model (right side) at the mid of flow domain for $Ra=10^{10}$.

Table 2 Nusselt numbers computed for Rayleigh-Bénard convection for different Ra .

Ra no.	Size of mesh	Present work		Kerr's results
		Smagorinsky model	Mixed scale model	
10^4	101×51	2.6690	2.6687	2.363
10^5	201×101	4.971	4.968	4.462
10^6	301×151	8.169	8.3535	8.424
10^7	361×181	13.729	14.016	15.904
10^8	401×201	24.252	22.611	-
10^9	401×201	38.237	33.302	-
10^{10}	401×201	72.93	71.76	-

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