

Consistent Lattice Boltzmann schemes for gas mixture modeling in the low Mach number limit

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Outline of this talk

- 1 Lattice Boltzmann Method
 - Mesoscopic Numerical Methods for ICNS
 - Microscopic Velocity Discretization
 - Space and Time Discretization
- 2 Homogeneous mixture flow modeling
 - Mixture characterization
 - Macroscopic modeling
 - Kinetic modeling
- 3 LBM scheme for mixture modeling
 - Simplified AAP model
 - Design of the discrete local equilibrium
 - Numerical results

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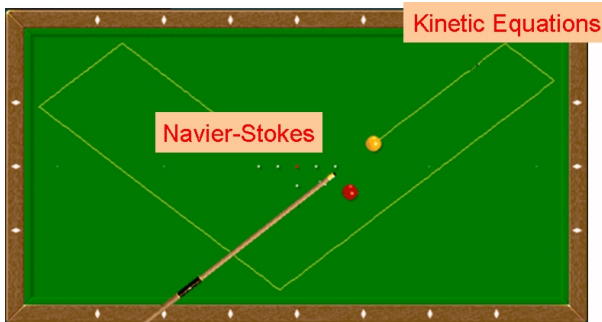
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Mesososcopic Numerical Methods for ICNS

Most of LBM models points to **kinetic equations** in order to solve fluidynamic equations in continuous regime (**Incompressible Navier-Stokes – ICNS**) ? **Does it worth the effort ?**



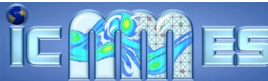
Main Categories

- Mesoscopic methods may be distinguished in two main categories:
 - (**Primitive**) Numerical methods using expressions for the numerical fluxes, derived by **simplified solutions of kinetic equations** (equilibrium and/or small-deviation solutions), for example GKS → they are not truly kinetic schemes, because the kinetic expressions are used for **physically-based macroscopic averaging**
 - (**Kinetic**) Numerical methods formulated directly in terms of **kinetic variables**, for example LBM → they are truly kinetic schemes, if and only if the adopted discretization allows to catch the kinetic phenomena, otherwise their kinetic content is questionable

Lattice Boltzmann Method (LBM) in a Nutshell

- number of papers on International Journals: **2,000 in the period 1988-2007** (for comparison, 28,000 papers on "Energy Saving"...)
- number of books: **14 in the period 2000-2007**
- main international conferences:
 - International Conference on Mesoscopic Methods in Engineering and Science, **ICMMES**
 - Discrete Simulation of Fluid Dynamics in Complex Systems, **DSFD**
- commercial codes: **PowerFLOW (EXA, spin-off MIT)**
- patents: mainly in bio-fluidics for medical applications

Conference: July 16 - 20, 2007
Location: Munich, Germany



Simplified BGK Model Equation

In the incompressible continuum limit, the **Mach number** as well as the **Knudsen number** is vanishingly small and the deviations of temperature and density are vanishingly small. Then, we can employ the simplified BGK equation, i.e.

- the **isothermal equilibrium distribution** $f_e^* \rightarrow f_e$, namely

$$f_e = \frac{\rho}{2\pi/3} \exp \left[-\frac{3(\xi_i - u_i)^2}{2} \right], \quad (1)$$

where $m = 1$ (since it is a constant), $d = 2$ (for two dimensional case) and $e = 1/3$ are assumed;

- and the **collision frequency independent** of the local state, namely $\lambda(\rho) \rightarrow \lambda$.

Gaussian Quadratures

We need to accurately compute integrals such as $\ll \phi(\boldsymbol{\xi}) f_e \gg$:
in particular, for the hydrodynamic conserved moments
 $\phi = 1, \xi_i$. The previous integrals can be expressed as

$$\ll \phi(\boldsymbol{\xi}) f_e \gg = \frac{\rho}{\pi} \int \phi(\boldsymbol{\xi}) \exp [-(\boldsymbol{\xi}^* - \mathbf{u})^2] d\xi_x^* d\xi_y^*, \quad (2)$$

where $d\xi_i^* = d\xi_i \sqrt{3/2}$. Let us apply a **Gaussian quadrature** in order to numerically solve the previous integrals by means of N discrete point along each direction

$$\ll \phi(\boldsymbol{\xi}) f_e \gg = \sum_{i=1}^N \sum_{j=1}^N (\zeta_i \zeta_j \phi f_e) (\xi_x^* = \xi_i^*, \xi_y^* = \xi_j^*) + E_N, \quad (3)$$

where ζ_i and ζ_j are proper **weighting functions**.

(1) Three-point Gauss-Hermite Formula

Let us rewrite the previous expression as

$$\ll \phi(\boldsymbol{\xi}) f_e \gg = \frac{\rho}{\pi} \int \psi(\boldsymbol{\xi}^*) \exp [-(\boldsymbol{\xi}^*)^2] d\xi_x^* d\xi_y^*, \quad (4)$$

where $\psi(\boldsymbol{\xi}^*) = \phi(\boldsymbol{\xi}) \exp [-\mathbf{u} \cdot (\mathbf{u} + 2\boldsymbol{\xi}^*)]$. Because of the weighting factor, among all the Gaussian quadratures, it is convenient to adopt the **Gauss-Hermite formula**. For example, in the case of the **three-point formula**, the three abscissas and the corresponding weighting functions of the quadrature are

$$\xi_1^* = -\sqrt{3/2} \ , \ \xi_2^* = 0 \ , \ \xi_3^* = +\sqrt{3/2} \ , \quad (5)$$

and $\zeta_i = w_i \exp [(\xi_i^*)^2]$ respectively, where

$$w_1 = \sqrt{\pi} / 6 \ , \ w_2 = 2 \sqrt{\pi} / 3 \ , \ w_3 = \sqrt{\pi} / 6 \ . \quad (6)$$

Obviously three points are **very few** and **large** E_N is expected.

(2) Incompressible Limit

The low Mach number limit $|\mathbf{u}| \ll |\boldsymbol{\xi}|$ allows one to **expand** $\psi = \psi_0 + O(\mathbf{u}^3)$, where

$$\psi_0(\boldsymbol{\xi}^*) = \phi(\boldsymbol{\xi}) [1 - \mathbf{u} \cdot (\mathbf{u} + 2\boldsymbol{\xi}^*) + 2(\mathbf{u} \cdot \boldsymbol{\xi}^*)]. \quad (7)$$

In the following, the **truncated expansion** ψ_0 will be used instead of ψ . In case of the **hydrodynamic moments** $\phi = 1, \xi_i$, the quadrature formula yields

$$\ll f_e \gg = \rho \sum_{i=1}^3 \sum_{j=1}^3 W_{ij} \psi_0(\xi_x^* = \xi_i^*, \xi_y^* = \xi_j^*) = \rho, \quad (8)$$

$$\ll \boldsymbol{\xi} f_e \gg = \rho \sum_{i=1}^3 \sum_{j=1}^3 W_{ij} (\boldsymbol{\xi} \psi_0)(\xi_x^* = \xi_i^*, \xi_y^* = \xi_j^*) = \rho \mathbf{u}. \quad (9)$$

where $W_{ij} = w_i w_j / \pi$. In this case, **the error E_N is zero !!**

D2Q9 Lattice Definition

- The previous **unexpected result** is valid for all ϕ **up to second order** with regards to the particle velocities: higher order moments show the limits of the poor quadrature formula in terms of its **symmetry properties** and the numerical calculation is not exact any more
- Let us rearrange the velocities of the quadrature formula $\{(\xi_x^* = \xi_i^*, \xi_y^* = \xi_j^*) | i, j = 1, 2, 3\}$ in a finite set of $Q = 9$ particle velocities, called **D2Q9 lattice**, i.e. equivalently $\{\xi_q | 0 \leq q \leq (Q - 1)\}$, and let us collect **the velocity components** in a second order **tensor V** , i.e.
 $V = [\xi_0, \xi_1, \dots, \xi_{(Q-1)}]^T$. The result is **very simple**, namely

$$V^T = \begin{bmatrix} 0 & 1 & 0 & -1 & 0 & 1 & -1 & -1 & 1 \\ 0 & 0 & 1 & 0 & -1 & 1 & 1 & -1 & -1 \end{bmatrix}. \quad (10)$$

(3) Local Equilibrium Definition

- Let us introduce a **new definition of local equilibrium** for the considered lattice as a vector of polynomials $\mathbf{f}_e \in \mathbb{R}^9$, defined in such a way that the generic component q is

$$(\mathbf{f}_e)_q = \rho W_q \psi_0(\xi_x = V_{q1}, \xi_y = V_{q2}). \quad (11)$$

- Moreover let us introduce the following **discrete operator** $\langle \cdot, \cdot \rangle$, which involves a sum on the lattice discrete velocities, namely

$$\langle A_{i j \dots q}, B_{m n \dots q} \rangle = \sum_{q=0}^Q A_{i j \dots q} B_{m n \dots q} = C_{i j m n \dots}$$

- Hence the previous results can be expressed as

$$\langle\langle \mathbf{f}_e \rangle\rangle = \langle 1, \mathbf{f}_e \rangle = \rho \text{ and } \langle\langle \boldsymbol{\xi} \mathbf{f}_e \rangle\rangle = \langle \mathbf{V}, \mathbf{f}_e \rangle = \rho \mathbf{u}$$

(4) Method of Characteristics (MOC)

- In the discretization of the lattice BGK equation, let us apply the **method of characteristics (MOC)** → Let us consider the streamlines defined by the condition $V_{qi} = d\hat{x}_i/d\hat{t}$, i.e. $\hat{x}_i^*(\hat{t}) = V_{qi}(\hat{t} - \hat{t}_0) + \hat{x}_{i0}^*$, where \hat{x}_{i0}^* is a proper constant
- Along these streamlines, the following notation holds

$$\frac{\partial f_q}{\partial \hat{t}} + V_{q1} \frac{\partial f_q}{\partial \hat{x}_1} + V_{q2} \frac{\partial f_q}{\partial \hat{x}_2} = \sum_{\alpha \in A} \frac{\partial f_q}{\partial \hat{\alpha}} \frac{d\hat{\alpha}}{d\hat{t}} = \frac{Df_q}{D\hat{t}} = \lambda(f_{eq} - f_q), \quad (12)$$

where $A = \{\hat{t}, \hat{x}_1, \hat{x}_2\}$.

- The theory of characteristics for this case is **extremely simplified**, because V is made of constants, and in particular $|V_{qi}| = 0, 1$ for $\forall q, i$

Forward Euler Time Integration Formula

- Let us introduce an **homogeneous space discretization** with $\delta\hat{x}$ spacing and let \hat{x}_{i0}^* be a discretization grid node
- If the time discretization step is assumed $\delta\hat{t} = \delta\hat{x}$, then moving along the previously defined characteristic yields $\hat{x}_i^*(n) = n h V_{qi} + \hat{x}_{i0}^*$ and, taking into account that $|V_{qi}| = 0, 1$ for $\forall q, i$, then $\hat{x}_i^*(n)$ is **again a discretization grid node** at any discrete time
- During an elementary time step, the particles **jump to the neighboring nodes** according to their discrete velocity
- Applying the forward Euler time integration formula to the discrete BGK equation and taking $\delta\hat{t} = 1$ yields the **simplest LBM scheme**

$$f_q(\hat{t} + 1, \hat{x}_i) = f_q(\hat{t}, \hat{x}_i - V_{qi}) + \lambda (f_{eq} - f_q)(\hat{t}, \hat{x}_i - V_{qi}). \quad (13)$$

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Concentration measures

- The **mass** concentration is defined as

$$x_{\sigma} = \rho_{\sigma} / \rho, \quad (14)$$

where ρ_{σ} is the single species density, while $\rho = \sum_{\varsigma} \rho_{\varsigma}$ is the total mixture density.

- The molar density as

$$n_{\sigma} = \rho_{\sigma} / m_{\sigma}, \quad (15)$$

where m_{σ} is the molecular weight, i.e. the weight of one mole of molecules.

- Consequently the **molar** concentration as

$$y_{\sigma} = n_{\sigma} / n, \quad (16)$$

where $n = \sum_{\varsigma} n_{\varsigma}$ is the total mixture *molar* density.

Mixture velocities

- The **mass**-averaged mixture velocity is defined as

$$\mathbf{u} = \sum_{\varsigma} x_{\varsigma} \mathbf{u}_{\varsigma}, \quad (17)$$

where \mathbf{u}_{ς} is the single species velocity. Since the *mass* concentrations were used, the previous quantity is also called **barycentric (mixture) velocity**.

- Similarly, by means of the molar concentrations, it is possible to define a **mole**-averaged mixture velocity, namely

$$\mathbf{v} = \sum_{\varsigma} y_{\varsigma} \mathbf{u}_{\varsigma}. \quad (18)$$

Since the *molar* concentrations were used, the previous quantity is also called **molar (mixture) velocity**.

Diffusion fluxes

- It is possible to define a specific **mass** diffusion flux for each species σ as

$$\mathbf{j}_\sigma = \rho_\sigma \mathbf{w}_\sigma, \quad (19)$$

where $\mathbf{w}_\sigma = \mathbf{u}_\sigma - \mathbf{u}$ is the *mass* diffusion velocity and clearly $\sum_\varsigma \mathbf{j}_\varsigma = 0$.

- Similarly, it is possible to define a specific **molar** diffusion flux for each species σ as

$$\mathbf{k}_\sigma = n_\sigma \mathbf{z}_\sigma, \quad (20)$$

where $\mathbf{z}_\sigma = \mathbf{u}_\sigma - \mathbf{v}$ is the *molar* diffusion velocity and clearly $\sum_\varsigma \mathbf{k}_\varsigma = 0$.

Maxwell-Stefan model

- In case of more than two species, the diffusion fluxes can be described macroscopically by the Maxwell-Stefan model, namely

$$\nabla y_\sigma = \sum_{\varsigma} B_{\sigma\varsigma} y_\sigma y_\varsigma (\mathbf{u}_\varsigma - \mathbf{u}_\sigma) = \frac{1}{n} \sum_{\varsigma} B_{\sigma\varsigma} (y_\sigma \mathbf{k}_\varsigma - y_\varsigma \mathbf{k}_\sigma), \quad (21)$$

where $B_{\sigma\varsigma} = B(m_\sigma, m_\varsigma)$ is the **binary Maxwell-Stefan diffusion resistance** coefficient. An important comment is that the previous parameter only depends (according to the results of the kinetic theory) on the **molecular weights** of considered species and on the **total pressure** and **(total) temperature** (thermodynamic variables identifying the mixture equilibrium state).

Full Boltzmann equations

- The simultaneous Boltzmann equations for a mixture without external force can be written as:

$$\partial_t f_\sigma + \xi \cdot \hat{\nabla} f_\sigma = Q_\sigma, \quad (22)$$

where $Q_\sigma = \sum_\varsigma Q_{\sigma\varsigma}$ and $Q_{\sigma\varsigma} = Q_{\varsigma\sigma}$, $\varsigma \neq \sigma$, is the **cross collision term** for two different species σ and ς . Obviously, for an N -component system, there will be N such equations. In general, the collision term is

$$Q_{\sigma\varsigma} = \int d\xi_\varsigma d\Theta d\varepsilon B(\Theta, \|\xi_{\sigma\varsigma}\|) [\textcolor{red}{f'_\sigma f'_\varsigma} - \textcolor{blue}{f_\sigma f_\varsigma}], \quad (23)$$

where f'_σ (f'_ς) and f_σ (f_ς) denote the **post-collision** and **pre-collision** state of the particle of species σ (ς), respectively, $\xi_{\sigma\varsigma} = \xi - \xi_\varsigma$.

Momentum transfer among the species

- Clearly the momentum of the single species is **not conserved**, because the species are interacting each other by transferring momentum, in such a way that the total mixture momentum is **conserved**.
- Hence it is worth the effort to compute the following integral, which describes the **momentum transfer** prescribed by full Boltzmann equations, namely

$$\int \xi Q_{\sigma} d\xi = p \sum_{\varsigma} B_{\sigma\varsigma} y_{\sigma} y_{\varsigma} (\mathbf{u}_{\varsigma} - \mathbf{u}_{\sigma}), \quad (24)$$

where now the Maxwell-Stefan diffusion resistance coefficient $B_{\sigma\varsigma}$ can be interpreted as macroscopic consequence of the **interaction potential** between species σ and ς .

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Simplified AAP model

- Let us consider a **simplified** version of the AAP model [Andries, Aoki, and Perthame 2002], which is based on only one **global** (i.e., taking into account all the species ς) operator for each species σ , namely

$$\partial_t f_\sigma + \boldsymbol{\xi} \cdot \hat{\nabla} f_\sigma = \lambda_\sigma [f_{\sigma(*)} - f_\sigma], \quad (25)$$

where

$$f_{\sigma(*)} = \frac{\rho_\sigma}{(2\pi\varphi_\sigma/3)} \exp \left[-\frac{3 (\boldsymbol{\xi} - \mathbf{u}_\sigma^*)^2}{2 \varphi_\sigma} \right], \quad (26)$$

and

$$\mathbf{u}_\sigma^* = \mathbf{u}_\sigma + \sum_\varsigma \frac{m^2}{m_\sigma m_\varsigma} \frac{B_{\sigma\varsigma}}{B_{\sigma\sigma}} x_\varsigma (\mathbf{u}_\varsigma - \mathbf{u}_\sigma). \quad (27)$$

Properties of simplified AAP model

- The **target velocity** can be easily recasted as

$$\mathbf{u}_\sigma^* = \mathbf{u} + \sum_{\varsigma} \left(\frac{m^2}{m_\sigma m_\varsigma} \frac{B_{\sigma\varsigma}}{B_{\sigma\sigma}} - 1 \right) x_\varsigma (\mathbf{u}_\varsigma - \mathbf{u}_\sigma). \quad (28)$$

- If $m_\sigma = m$ for $\forall \sigma$, then (Property 1)

$$\mathbf{u}_\sigma^* = \mathbf{u} + \sum_{\varsigma} \left(\frac{m^2}{mm} \frac{B_{mm}}{B_{mm}} - 1 \right) x_\sigma x_\varsigma (\mathbf{u}_\varsigma - \mathbf{u}_\sigma) = \mathbf{u}. \quad (29)$$

- Clearly (Property 2)

$$\sum_{\sigma} x_\sigma \mathbf{u}_\sigma^* = \mathbf{u} + \sum_{\sigma} \sum_{\varsigma} \left(\frac{m^2}{m_\sigma m_\varsigma} \frac{B_{\sigma\varsigma}}{B_{\sigma\sigma}} - 1 \right) x_\sigma x_\varsigma (\mathbf{u}_\varsigma - \mathbf{u}_\sigma) = \mathbf{u}. \quad (30)$$

Asymptotic analysis of AAP model

- Let us consider a **regular (Hilbert) expansion** of the previous model. Collecting the leading terms of the momentum equation yields

$$\lambda_{\sigma} \rho_{\sigma}^{(0)} [\mathbf{u}_{\sigma}^{*(1)} - \mathbf{u}_{\sigma}^{(1)}] = \nabla p_{\sigma}^{(0)}, \quad (31)$$

where $p_{\sigma}^{(k)} = \varphi_{\sigma} \rho_{\sigma}^{(k)} / 3$.

- If λ_{σ} is selected as $\lambda_{\sigma} = p B_{\sigma\sigma} / \rho$, then the previous expression becomes

$$1/p^{(0)} \nabla p_{\sigma}^{(0)} = \sum_{\varsigma} B_{\sigma\varsigma} y_{\sigma} y_{\varsigma} [\mathbf{u}_{\varsigma}^{(1)} - \mathbf{u}_{\sigma}^{(1)}], \quad (32)$$

which clearly proves that the leading terms of the macroscopic equations recovered by means of the AAP model are **consistent with Maxwell-Stefan model**

Indifferentiability Principle

- If $m_\sigma = m$ for $\forall \sigma$, then $\mathbf{u}_\sigma^* = \mathbf{u}$ (Property 1) and, according to the selected tuning strategy, $\lambda_\sigma = \lambda = p B_{mm}/\rho$.
- Hence summing over all the species yields

$$\partial_t f + \boldsymbol{\xi} \cdot \nabla f = \lambda [f_{(m)} - f], \quad (33)$$

where $f = \sum_\sigma f_\sigma$ and $f_{(m)}$ is defined by

$$f_{(m)} = \frac{\rho}{(2\pi\varphi/3)} \exp \left[-\frac{3 (\boldsymbol{\xi} - \mathbf{u})^2}{2\varphi} \right]. \quad (34)$$

- This clearly proves that the AAP model satisfies the **Indifferentiability Principle**.

Continuous equilibrium moments

- Let us introduce the following function

$$f_e(\rho, \varphi, u_1, u_2) = \frac{\rho}{(2\pi\varphi/3)} \exp \left[-\frac{3 (\xi - \mathbf{u})^2}{2 \varphi} \right]. \quad (35)$$

- Let us define $\ll \cdot \gg = \int_{-\infty}^{+\infty} \cdot d\xi_1 d\xi_2$ and the **generic moment** $m_{pq} = \ll f_e \xi_1^p \xi_2^q \gg$.
- All the equilibrium moments appearing in the **Euler system of equations** are the following m_{00} , m_{10} , m_{01} , m_{20} , m_{02} , m_{11} . Unfortunately this set is made of 6 elements, but the dimension of the considered lattice (for symmetry reasons) is 9. Hence other 3 (=9-6) target equilibrium moments are missing. **Arbitrarily** they are selected as m_{21} , m_{12} and m_{22} .

Simplified continuous equilibrium moments

- Collecting the previous results yields

$$\begin{aligned} \bar{m}_c(\rho, \varphi, u_1, u_2) = & \rho [1, u_1, u_2, \\ & u_1^2 + \varphi/3, u_2^2 + \varphi/3, u_1 u_2, \\ & u_1 u_2^2 + u_1 \varphi/3, u_1^2 u_2 + u_2 \varphi/3, \\ & \varphi (u_1^2 u_2^2 + u_1^2 \varphi/3 + u_2^2 \varphi/3 + \varphi/9)]^T. \end{aligned}$$

- The previous analytical results involve **high order terms** (like $u_1 u_2^2$) which are not strictly required, in order to recover the macroscopic equations we are interested in.

$$\begin{aligned} m_c(\rho, \varphi, u_1, u_2) = & \rho [1, u_1, u_2, \\ & u_1^2 + \varphi/3, u_2^2 + \varphi/3, u_1 u_2, \\ & u_1/3, u_2/3, \\ & (u_1^2 + u_2^2)/3 + \varphi/9]^T \end{aligned}$$

(3) Design of discrete local equilibrium

- On the selected lattice, the discrete integrals $m_{\sigma(*)}$, corresponding to the previous continuous ones, can be computed by means of **simple linear combinations** of the discrete equilibrium distribution function $f_{\sigma(*)}$ (still unknown), namely $m_{\sigma(*)} = M f_{\sigma(*)}$ where M is a matrix defined as

$$M = [1, V_1, V_2, V_1^2, V_2^2, V_1 V_2, V_1 V_2^2, V_1^2 V_2, V_1^2 V_2^2]^T. \quad (36)$$

- We design the **discrete local equilibrium** such as $m_{\sigma(*)} = m_c(\rho_\sigma, \varphi_\sigma, u_{\sigma 1}^*, u_{\sigma 2}^*)$, or equivalently $f_{\sigma(*)} = M^{-1} m_c(\rho_\sigma, \varphi_\sigma, u_{\sigma 1}^*, u_{\sigma 2}^*)$. In particular the latter provides the operative formula for defining the local equilibrium and consequently the scheme.

Asymptotic analysis of LBM scheme

- Let us consider a **regular (Hilbert) expansion** of the previous model. Collecting the leading terms of the continuity equation yields

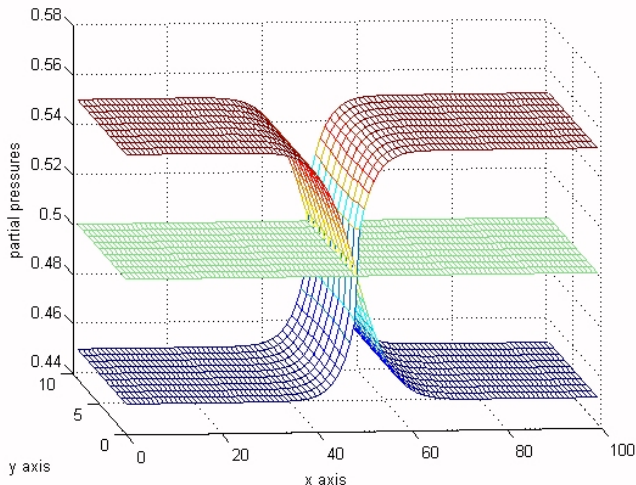
$$\partial_t \rho_\sigma^{(0)} + \nabla \cdot [\rho_\sigma^{(0)} \mathbf{u}_\sigma^{*(1)}] = \omega_\sigma \nabla^2 p_\sigma^{(0)}, \quad (37)$$

and applying the flux definition yields

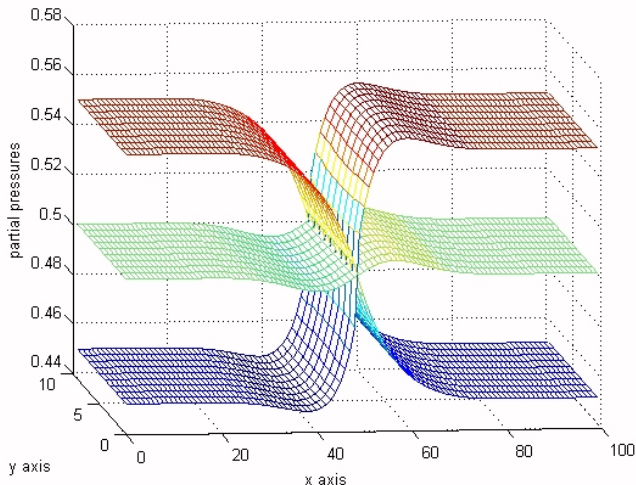
$$\partial_t \rho_\sigma^{(0)} + \nabla \cdot [\rho_\sigma^{(0)} \mathbf{u}_\sigma^{(1)}] = (\omega_\sigma - \tau_\sigma) \nabla^2 p_\sigma^{(0)} = -1/2 \nabla^2 p_\sigma^{(0)} \neq 0. \quad (38)$$

- The simple scheme **does not preserve the mass continuity for the single species**. Clearly this is due to the low accuracy of the forward Euler integration rule.
- The problem can be fixed by means of a **variable transformation** $f_\sigma \rightarrow \bar{f}_\sigma$ which is equivalent to apply the Crack-Nicholson integration rule.

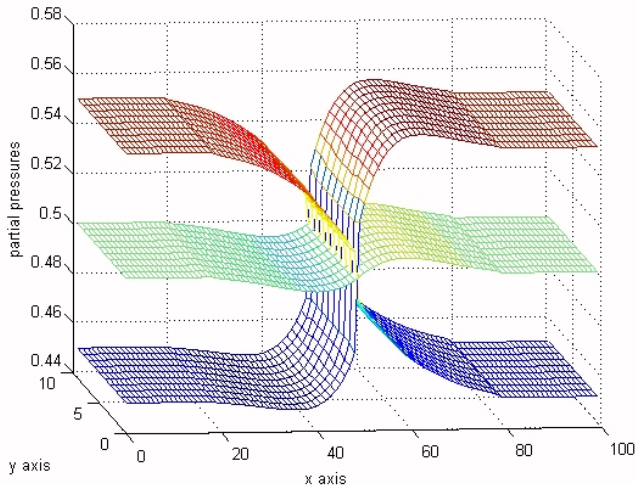
Passive scalar (no barycentric dynamics)



Fick model (with barycentric dynamics)



Maxwell-Stefan model (with barycentric dynamics)



Results due to Fick and Maxwell-Stefan are similar

- Actually it seems as there is **no much qualitative difference** between truly Fick model and Maxwell-Stefan model.
- Let us compare directly the two models, namely

$$(\nabla y_\sigma)_F = \frac{y_\sigma}{D_\sigma} (\mathbf{v} - \mathbf{u}_\sigma) = \sum_\varsigma \frac{y_\sigma y_\varsigma}{D_\sigma} (\mathbf{u}_\varsigma - \mathbf{u}_\sigma), \quad (39)$$

$$(\nabla y_\sigma)_{MS} = \sum_\varsigma B_{\sigma\varsigma} y_\sigma y_\varsigma (\mathbf{u}_\varsigma - \mathbf{u}_\sigma). \quad (40)$$

- The difference is not in the structure of the expressions, but only in the transport coefficients: D_σ depends only on species σ , while $B_{\sigma\varsigma}$ depends on the **interacting couple**.
- Passive scalar approach reduces a lot the potentiality of Fick model, by **simplifying the connection** among species.

Conclusions: why mesoscopic methods

- Mesoscopic numerical methods inherit the **(conceptual) simplicity** of kinetic formulation:
 - 1 they involve simple transport equations because the microscopic velocities are constrained on a **lattice**;
 - 2 in the pseudo-kinetic equations of these schemes, only **linear** differential operators appear;
 - 3 the non-linearities are **concentrated** in the definition of the local equilibrium.
- By improving the accuracy of the numerical discretization, it is possible to tune locally the scheme in order to realize **hybrid (kinetic –fluidynamic) solvers**.
- Numerical error preserves some flavors of the high-order kinetic dynamics → this makes **the error somehow more predictable** because it is physically based.

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