

# Kinetic simulations by the DSBGK method with noise reduction

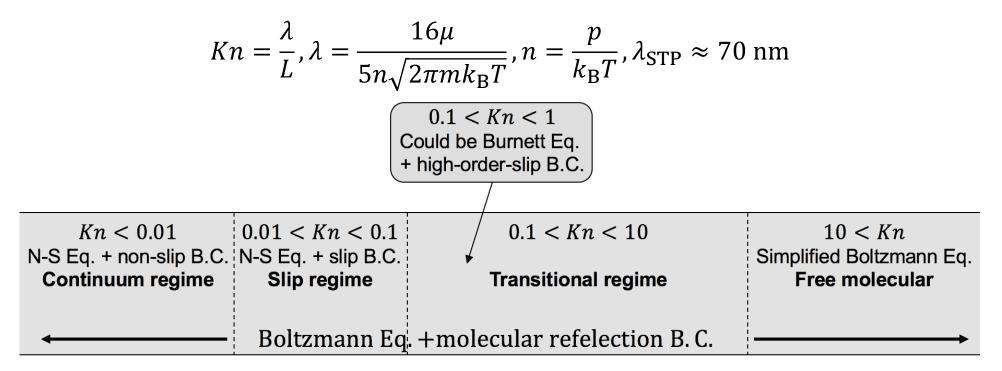
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## **Contents:**

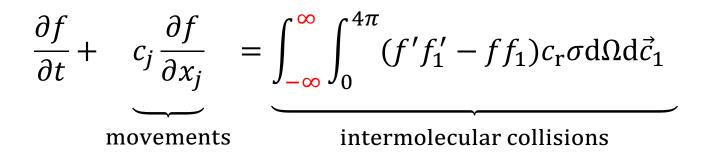
- Kinetic theory
- Different simulation methods
- DSBGK and its noise reduction technique
- Validation by and comparison with other methods
- Evolution demonstration of some transient solutions

## **Regime division of gas flows**



- MEMS/NEMS (e.g., microchip)
- Vacuum system (e.g., lithography, film coating)
- Knudsen pump (without rotor)
- Shale gas ( $D_{\text{pore}} \approx 10 \text{ nm}$ )
- Gas diffusion through membranes, nanotubes
- Mars exploration (ppprox 610 Pa, about  $p_{\mathrm{Earth}}$  at Hpprox 35 km )
- Spacecraft ( $\lambda \approx 1 \text{ cm}$  at H = 85 km, space station at  $H \approx 400 \text{ km}$ )
- Supersonic flow (shockwave thickness  $\approx$  200 nm)

#### **Boltzmann equation and Bhatnagar-Gross-Krook model**



 $f(t, \vec{x}, \vec{c}) d\vec{x} d\vec{c}$  is the molecular number inside element  $d\vec{x} d\vec{c}$ 

Hard sphere molecular model: 
$$\sigma = \frac{D^2(\mu)}{4}$$
  
 $n(t, \vec{x}) = \int_{-\infty}^{\infty} f d\vec{c}$ , and  $p = nk_{\rm B}T$   
 $\vec{u}(t, \vec{x}) = \frac{1}{n} \int_{-\infty}^{\infty} \vec{c} f d\vec{c}$   
 $T(t, \vec{x}) = \frac{2}{3k_{\rm B}n} \int_{-\infty}^{\infty} \frac{m(\vec{c} - \vec{u})^2}{2} f d\vec{c}$ 

BGK collision model: RHS =  $v(f^{eq} - f)$  with  $v = \frac{nk_BT}{\mu}$  or  $v = \frac{5k_B}{2m}\frac{nk_BT}{\eta}$ , or  $v(\vec{c})$ 

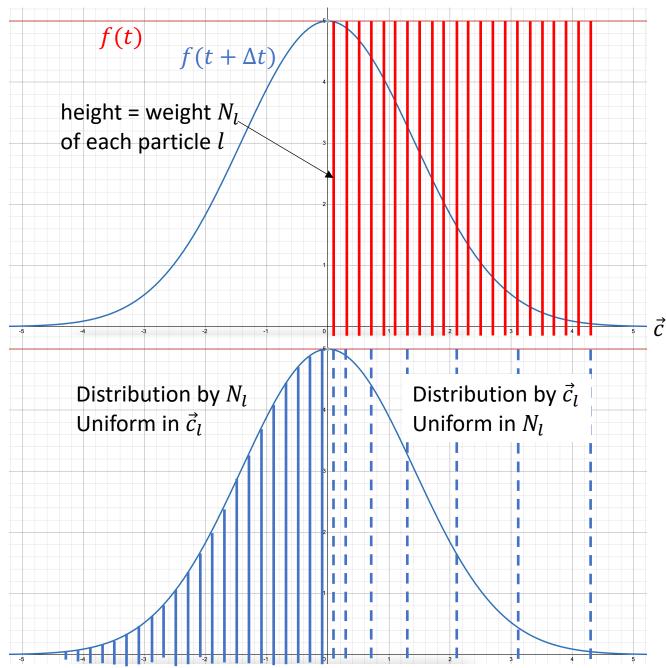
#### Simulation methods from different communities

Direct Simulation Monte Carlo Method (DSMC): Lagrangian particles (at 2 <sup>nd</sup> RGD, 1960)	Discrete Velocity Method (DVM): Eulerian grids (1964)	Lattice Boltzmann Method (LBM): Eulerian grids (1992~1997)	Direct Simulation BGK Method (DSBGK): Lagrangian particles (at 2 <sup>nd</sup> FCMC, 2009)	
<ul> <li>Converge to Boltzmann equation</li> <li>Standard at arbitrary <i>Kn</i></li> <li>Costly at low speed due to noise</li> </ul>	<ul> <li>Solve kinetic equations</li> <li>Accurate by FINE grids</li> <li>Efficient in 2D</li> <li>Costly in 3D</li> </ul>	<ul> <li>Solve BGK-type equation</li> <li>Use VERY coarse velocity grids</li> <li>Converge to the N-S equation</li> <li>Valid at low <i>Kn</i></li> <li>Efficient</li> </ul>	<ul> <li>Converge to the BGK equation</li> <li>Accurate at low speed (<i>Ma</i>&lt;1)</li> <li>Significant noise reduction</li> <li>Efficient particularly in large-scale 3D problems at low speed</li> </ul>	
• PIC-DSMC with the cell mean value method to deposit from particle to grid		<ul> <li>Easy extension to two-phase flows, Darcy flows, boiling, and MHD, etc</li> </ul>		
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[1] J Li, Multiscale and Multiphysics Flow Simulations of Using the Boltzmann Equation, Springer, 2020.

# DSBGK and its noise reduction technique

#### **Extrapolation of the acceptance-rejection scheme**

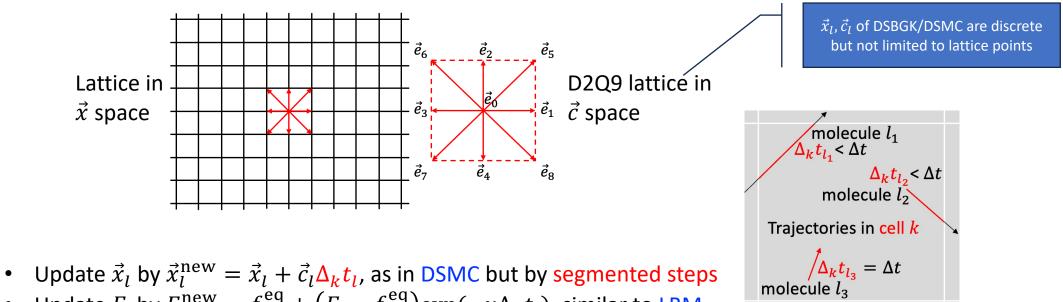


#### Model the change of f in the $\vec{c}$ space:

- Acceptance-rejection scheme: solidred → dashed-blue
- Extrapolation of the scheme: solid-red → solid-blue
- Generally valid for changing a representative sample  $(\vec{x}_l, \vec{c}_l, N_l)$  of arbitrary f(t) to  $(\vec{x}_l, \vec{c}_l, N_l \frac{f(t+\Delta t)}{f(t)})$  for  $f(t + \Delta t)$
- Statistically valid if  $(\vec{x}_l, \vec{c}_l, N_l)$  has noise but its ensemble-average is representative of f(t)

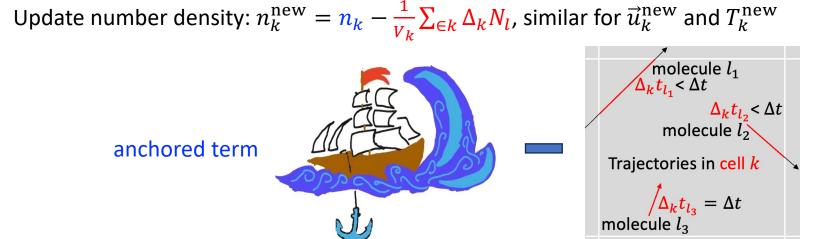
### **DSBGK with a duality of DSMC-LBM**

- Start with  $(\vec{x}_l, \vec{c}_l, N_l \equiv N_0)$  and  $F_l = f(\vec{c}_l) = f^{eq}(\vec{c}_l)$ , and  $n_k = n_0$ ,  $\vec{u}_k = \vec{u}_0$ ,  $T_k = T_0$  as in DSMC
- Adopt idea from lattice-based LBM:  $f_{\alpha}(\vec{x} + \vec{e}_{\alpha}\Delta t, t + \Delta t) = f_{\alpha}(\vec{x}, t) + [f_{\alpha}^{eq}(\vec{x}, t) f_{\alpha}(\vec{x}, t)]/\tau$



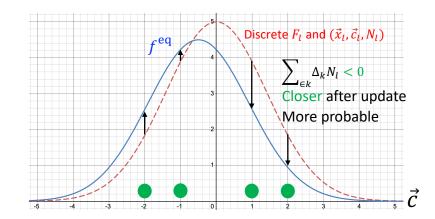
- Update  $F_l$  by  $F_l^{\text{new}} = f_k^{\text{eq}} + (F_l f_k^{\text{eq}}) \exp(-v\Delta_k t_l)$ , similar to LBM
- Update  $N_l$  by  $N_l^{\text{new}} = N_l F_l^{\text{new}} / F_l$ , according to extrapolation of the acceptance-rejection scheme
- No change of  $\vec{c}_l$  because the evolution of f due to collisions has been reflected by changing  $N_l$  (extrapolation idea)
- Regulate  $n_k$ ,  $\vec{u}_k$ ,  $T_k$  of  $f_k^{eq}$  using  $\sum_{\in k} \Delta_k N_l$ ,  $\sum_{\in k} (\Delta_k N_l m_l \vec{c}_l)$ ,  $\sum_{\in k} (\Delta_k N_l m_l \vec{c}_l^2/2)$  with  $\Delta_k N_l = N_l^{new} N_l$ , for conservations
- Update  $\vec{c}_l$ , as in DSMC, only after reflecting at walls and then  $F_l^{\text{new}} = f_{\text{wall}}(\vec{c}_l^{\text{new}})$ , but keep  $N_l$  for conservation
- Delete/generate outgoing/incoming particles at open boundaries, as in DSMC, additionally assign  $F_l = f_{BC}^{eq}(\vec{c}_l), N_l \equiv N_0$

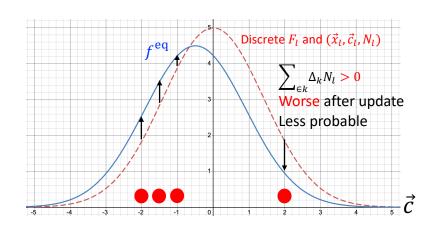
#### Noise reduction technique



increment due to physics (left-bottom) and noise (right-bottom)

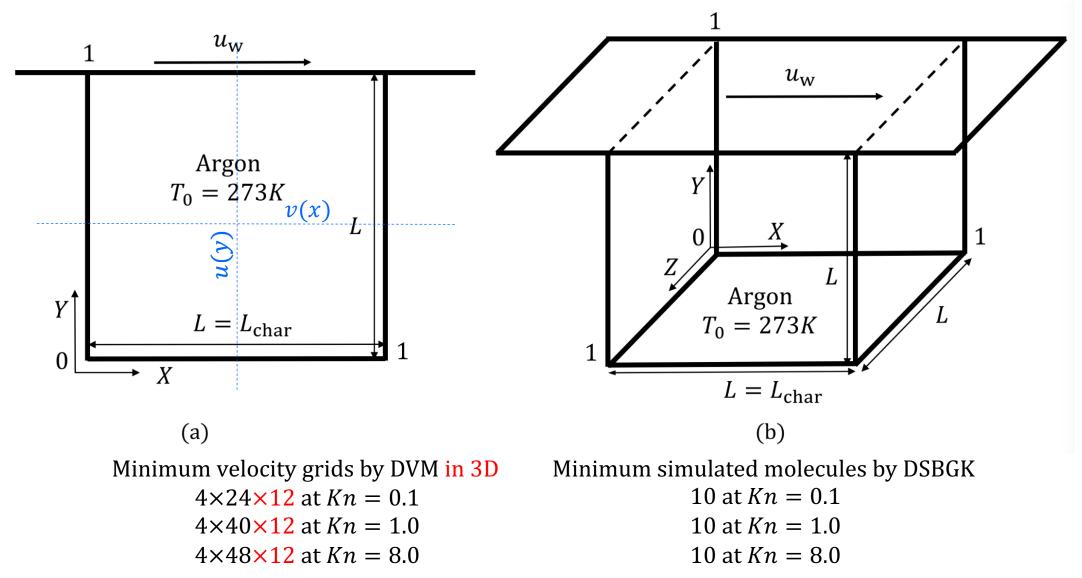
- Special case with  $(\vec{x}_l, \vec{c}_l, N_l)$  and  $F_l$  of  $f(\forall t, \forall \vec{x}) \equiv f^{eq}, \rightarrow \Delta_k N_l \equiv 0$  and  $n_k \equiv n_0$  as expected,  $\rightarrow$  zero noise
- General case with  $(\vec{x}_l, \vec{c}_l, N_l)$  and  $F_l$  of  $f \neq f^{eq}$ 
  - Avoid noise from number randomness (e.g., 4 < nV = 10 moving symmetrically, bottom-left figure)
  - Bring noise by asymmetric/random distribution of moving particles, bottom-right figure
  - In general, have noise reduction due to  $\sum_{\in k} \Delta_k N_l \ll \sum_{\in k} N_l$ , as  $\Delta_k N_l \propto (f f^{eq}), \propto \Delta t, \propto \Delta x$ , and  $\sum_{\in k} \Delta \pm dx$



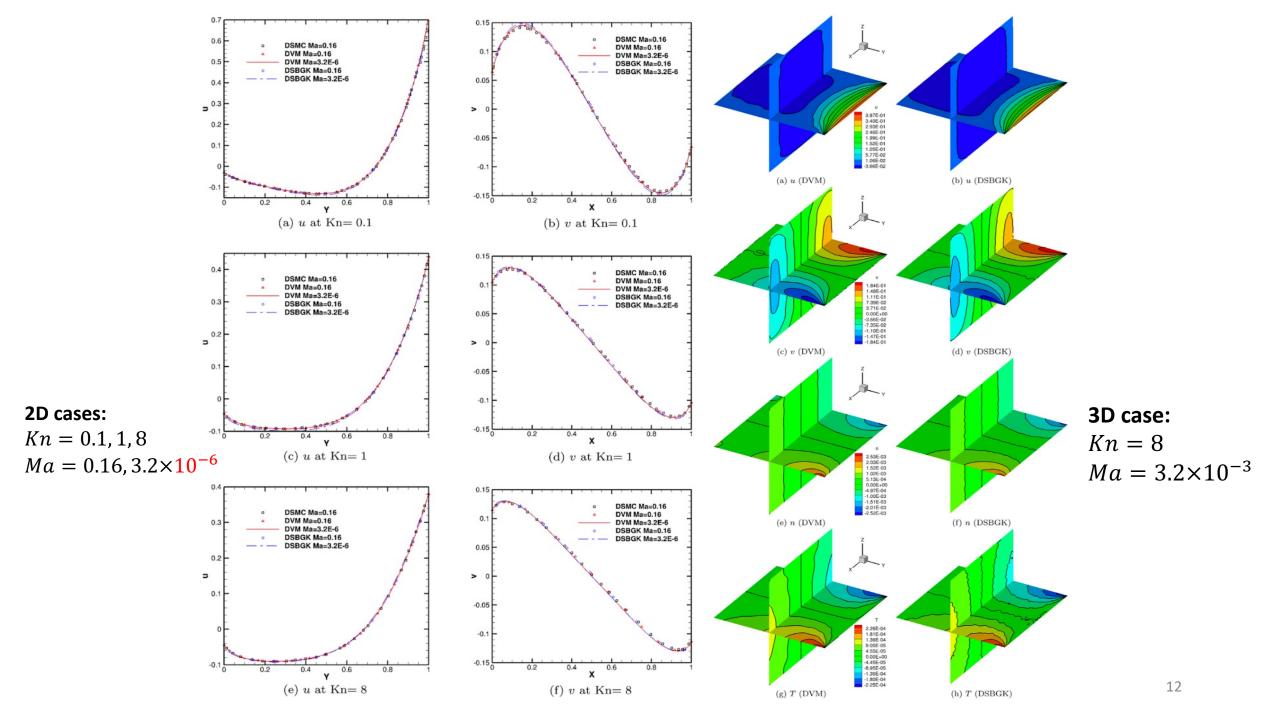


# Validations, comparisons and demonstrations

#### Lid-driven cavity flow



[1] MT Ho, J Li, L Wu, JM Reese, YH Zhang, *Computers & Fluids*, **181**: 143-159, 2019.



#### Mass flow rate through short channels

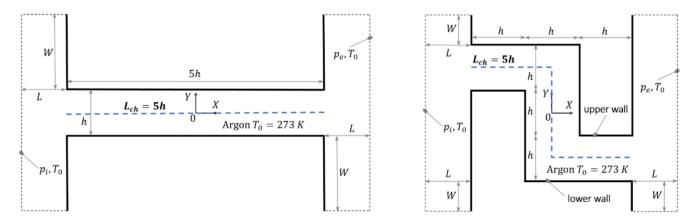
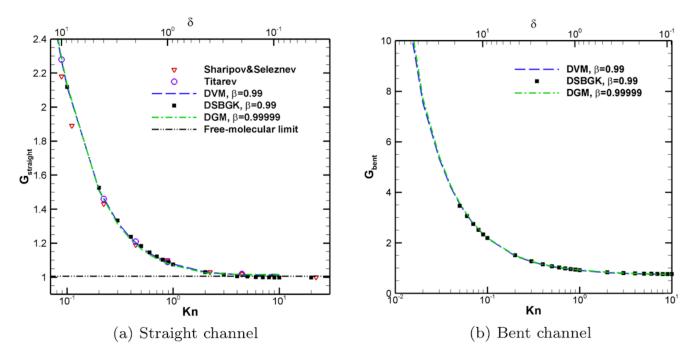
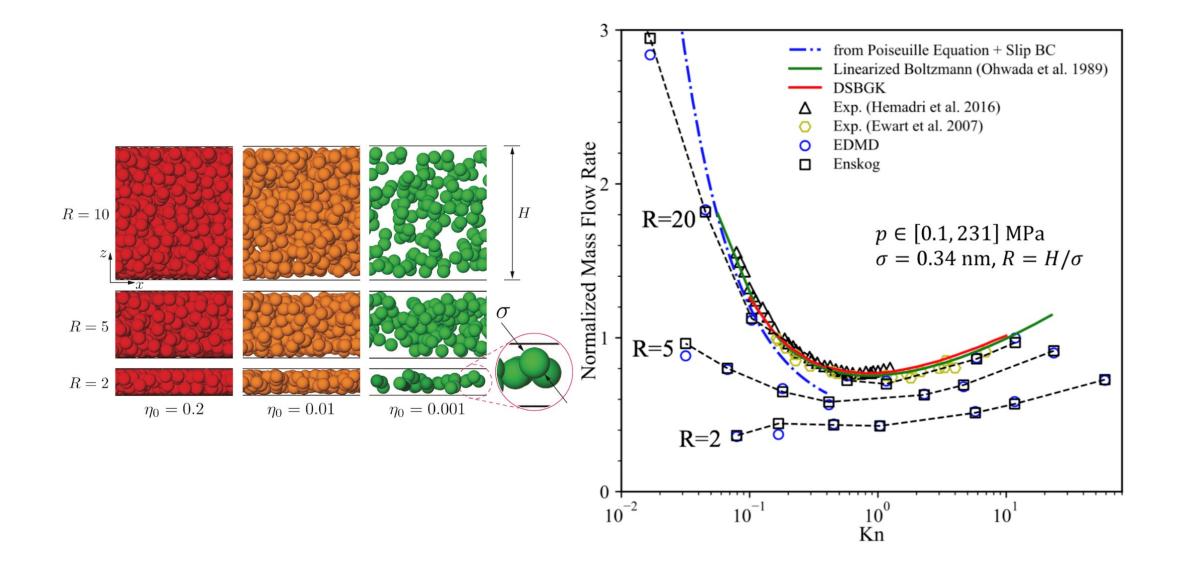


Fig. 1. Pressure-driven gas flows from the left-hand reservoir to the right-hand reservoir through the straight channel (left) and the bent channel (right) of the same height h and axis length  $L_{ch} = 5h$ . Note that only small parts of the reservoirs are illustrated here [12].

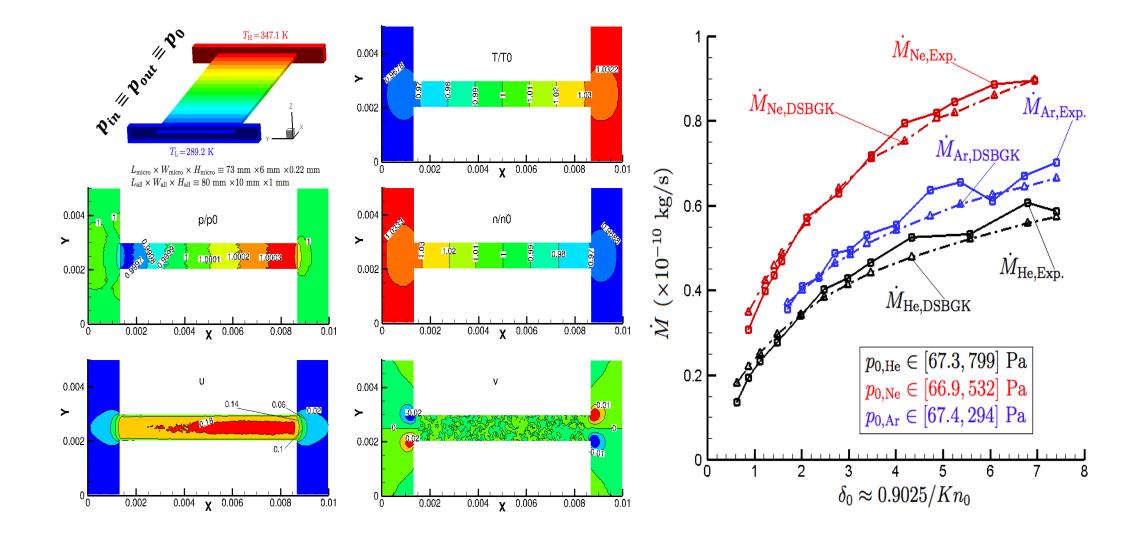


[1] MT Ho, J Li, W Su, L Wu, MK Borg, ZH Li, YH Zhang, *Journal of Fluid Mechanics*, **901**: A26, 2020.

#### Mass flow rate through nano-channel with confinement

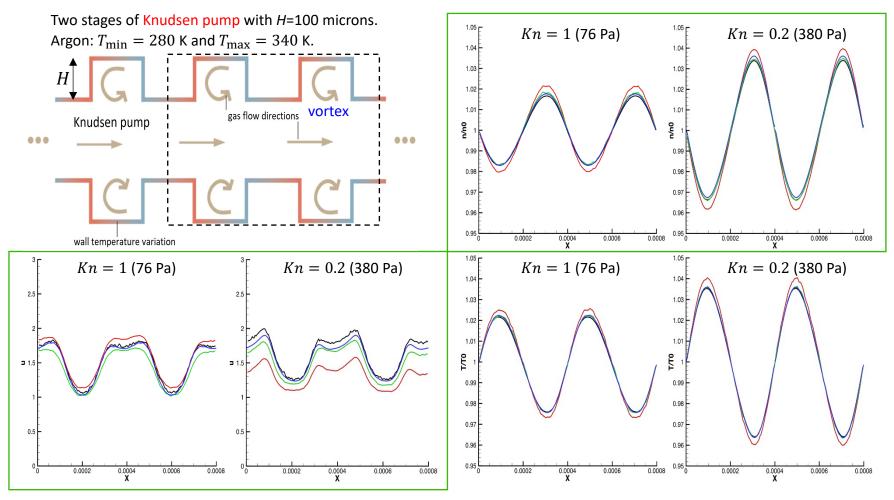


#### **Thermal transpiration flow**



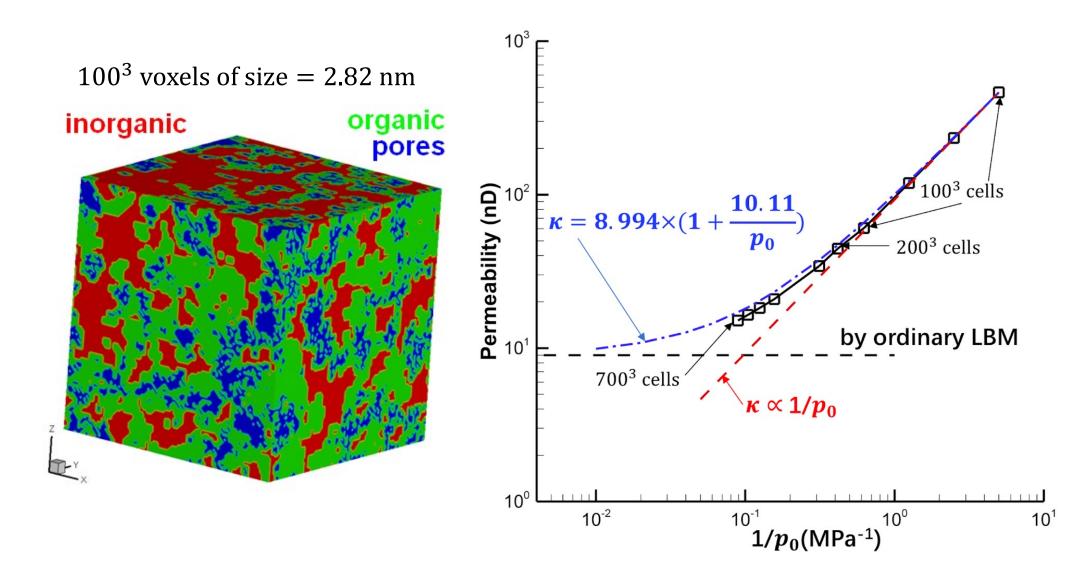
[1] H Yamaguchi, P Perrier, MT Ho, JG Meolans, T Niimi, I Graur, *Journal of Fluid Mechanics*, **795**: 690-707, 2016.
[2] J Li, CP Cai and ZH Li, *International Communications in Heat and Mass Transfer*, **119**: 104924, 2020.

#### Thermal transpiration flow (cont'd)

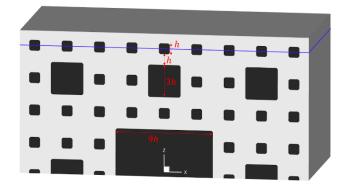


Distributions of  $n/n_0$ ,  $T/T_0$  and u on the centerline along the x axis obtained by the **DSMC** simulation (black), the **GKUA** simulation based on the Shakhov equation (blue), and the **DSBGK** simulation of using  $\tau = \tau(\mu)$  (red) and  $\tau = \tau(\eta)$  (green), respectively.

#### Shale gas permeability of a real digital rock

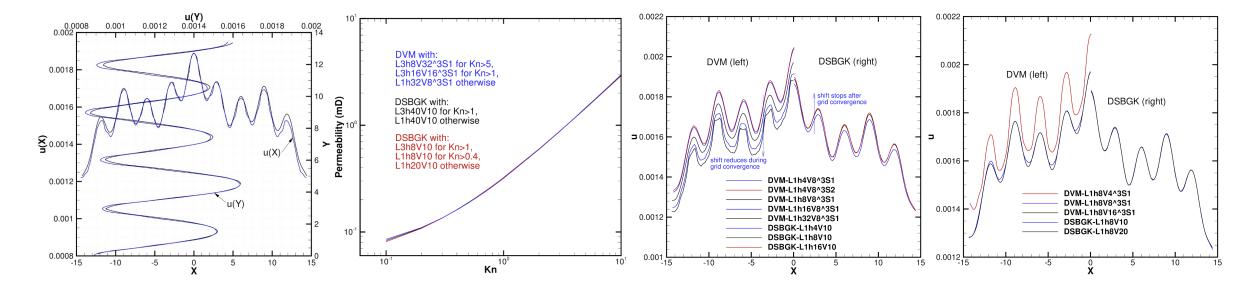


#### Performance comparison between DSBGK and DVM



<b>DSBGK</b>	L0h4V10	L3h4V10	L1h4V10	<b>L1h8V10</b>	L1h8V20
G	0.2125	0.2099	0.2110	0.2126	0.2126
Memory (GB)	0.3060	0.3780	0.3310	2.624	4.570
CPU time (sec)	1729 × 1	2270 × 1	1916 × 1	17518 × 2	17333 × 4
<b>DVM</b>	L1h4V8 <sup>3</sup>	L1h8V8 <sup>3</sup>	$L1h8V16^{3}$	L1h16V8 <sup>3</sup>	<i>L</i> 1 <i>h</i> 32 <i>V</i> 8 <sup>3</sup>
G	0.2299	0.2204	0.2189	0.2134	0.2090
Memory (GB)	2.304	15.23	110.3	109.6	≈ 7 × 109.6
CPU time (sec)	60.09 × 8	488.8 × 8	4759 × 8	3410 × 8	4770 × 64

#### *Kn*=1 by default



Minimum cost by **DVM** : 767 GB memory,  $2^3 \times 8$  CPU cores and  $4770 \times 64$  seconds of CPU time Minimum cost by **DSBGK**: 2.6 GB memory, 2 CPU cores and  $17518 \times 2$  seconds of CPU time

#### **Comparison with other kinetic methods**

#### DVM vs. other deterministic methods:

- DVM adopted in our performance comparison is cheaper than its variants for Kn > 0.01
- Only for continuum regime of Kn < 0.01, UGKS/DUGKS/GSIS can use coarser spatial grids, and GSIS can also use less iteration steps (see [1], [2], [3])

#### **DSBGK vs. other statistical particle methods:**

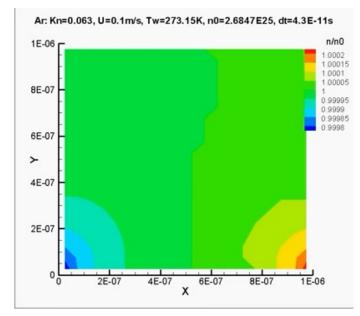
- DSBGK uses much less averaging samples than LVDSMC (see [4] and [5])
- DSBGK uses much, much less averaging samples than DSMC at low speeds (obviously)

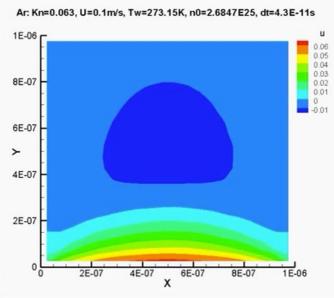
#### Fortran MPI solver based on DSBGK:

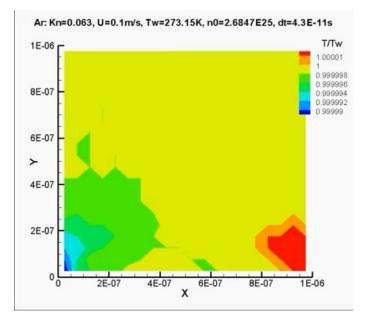
• More info @ https://sites.google.com/view/nanogassim/welcome

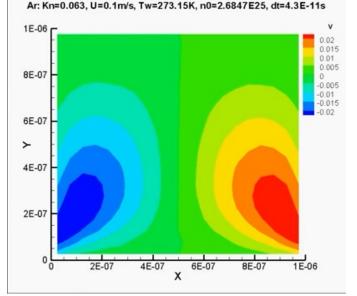
P Wang, MT Ho, L Wu, ZL Guo, YH Zhang, Computers & Fluids, 161: 33-46, 2018.
 Q Gu, MT Ho, YH Zhang, Computers & Fluids, 222: 104932, 2021.
 W Su, MT Ho, YH Zhang, L Wu, Computers & Fluids, 206: 104576, 2020.
 FD Bosco, YH Zhang, Transport in Porous Media, 138: 25-48, 2021.
 J Li, MT Ho, MK Borg, CP Cai, ZH Li, YH Zhang, Computers & Fluids, 226: 105017, 2021.

#### Lid-driven cavity flow (video online)

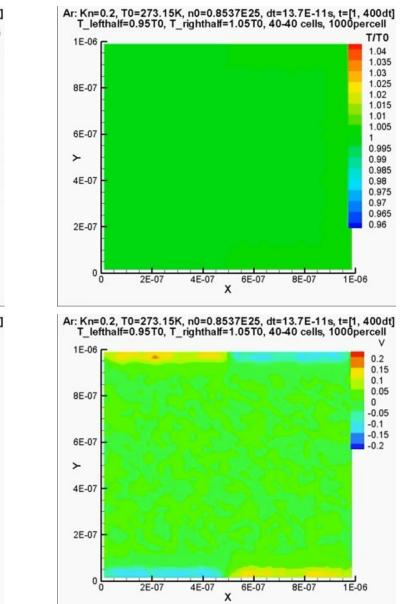


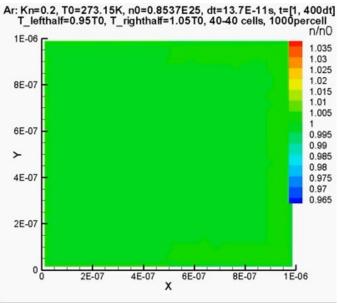


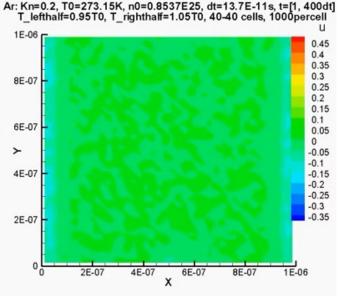




#### Thermal transpiration flow (video online)



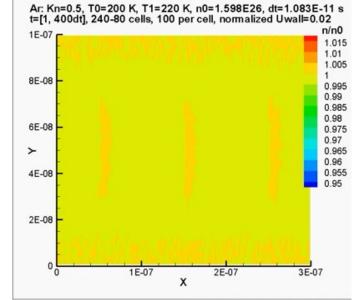


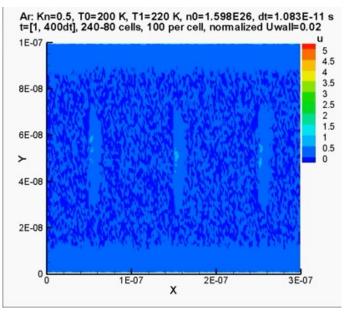


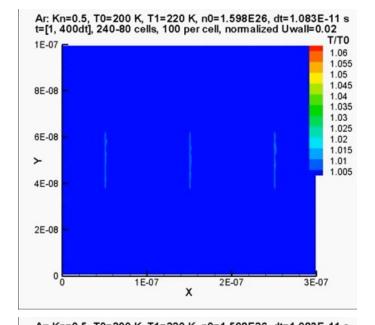
#### **Radiometer flow (video online)**

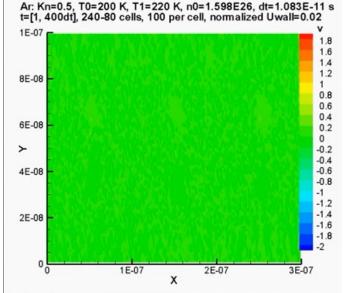


from Wikipedia









# Thank you