

Transported Probability Density Function (PDF) Methods for Multiscale and Uncertainty Problems - Part III

A Solution Algorithm for the Fluid Dynamic Equations Based on a Stochastic Model for Molecular Motion (cont.)

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Fokker-Planck Solution Algorithm

$$\frac{\partial \mathcal{F}}{\partial t} + V_i \frac{\partial \mathcal{F}}{\partial x_i} + \frac{\partial F_i \mathcal{F}}{\partial V_i} = S^{(\text{FP})}(\mathcal{F}) = \frac{\partial}{\partial V_i} \left(\frac{1}{\tau_{\text{FP}}} (V_i - U_i) \mathcal{F} \right) + \frac{\partial^2}{\partial V_k \partial V_k} \left(\frac{2e_s}{3\tau_{\text{FP}}} \mathcal{F} \right)$$

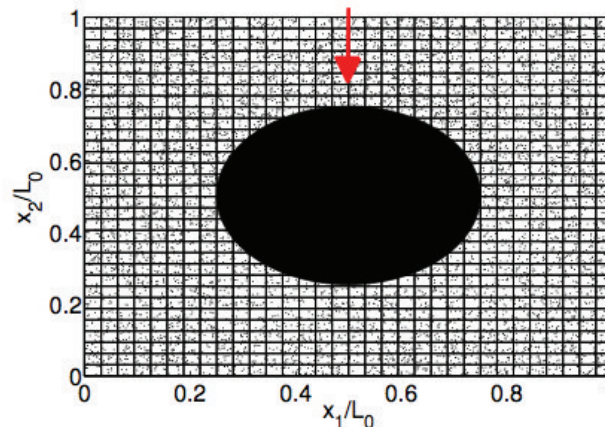
solved through stochastic motion of notional particles

$$\begin{aligned} \frac{dX_i}{dt} &= M_i \quad \text{with} \\ \frac{dM_i}{dt} &= -\frac{1}{\tau} (M_i - U_i) + \left(\frac{4e_s}{3\tau} \right)^{1/2} \frac{dW_i(t)}{dt} + F_i \end{aligned}$$

Fokker-Planck Solution Algorithm

n_t time steps are performed

- (1) U and e_s at time t are estimated at each grid node and interpolated to the particle positions,
- (2) the time step size Δt is determined,
- (3) a first half-step is performed to estimate the particle mid-points,
- (4) mid-point boundary conditions are applied,
- (5) U and e_s at time $t + \Delta t/2$ are interpolated from the grid nodes to the particle mid-point positions,
- (6) the new particle velocities and positions are computed, and
- (7) the boundary conditions are enforced.



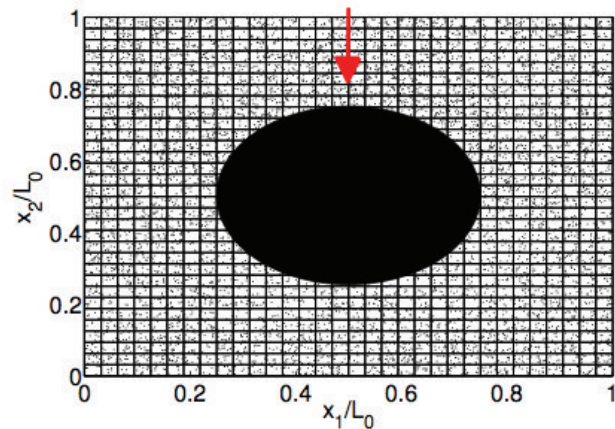
in statistical steady state U and e_s do not depend on the time.

Estimation of Statistical Moments

$$\mathcal{U}^J(t) = \sum_{j=1}^{N_p} \left\{ \hat{g}^J(\mathbf{X}^j(t)) \mathbf{M}^j(t) \right\},$$

$$\mathcal{E}^J(t) = \sum_{j=1}^{N_p} \left\{ \hat{g}^J(\mathbf{X}^j(t)) \mathbf{M}^j(t) \cdot \mathbf{M}^j(t) \right\}$$

$$\mathcal{W}^J(t) = \sum_{j=1}^{N_p} \left\{ \hat{g}^J(\mathbf{X}^j(t)) \right\},$$



$$\mathbf{U}(\mathbf{x}^J, t) = \frac{\mathcal{U}^J(t)}{\mathcal{W}^J(t)}$$

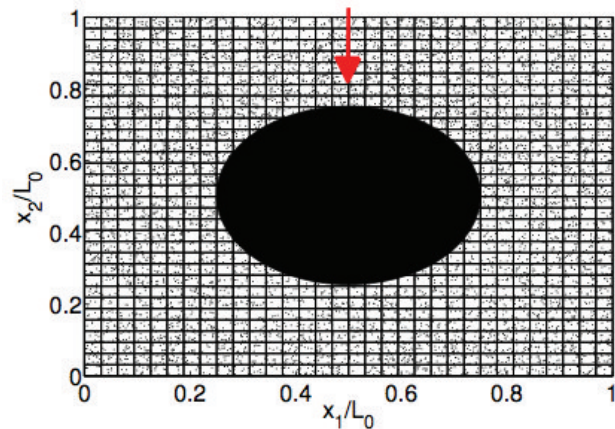
$$\mathbf{e}_s(\mathbf{x}^J, t) = \frac{1}{2} \left(\frac{\mathcal{E}^J(t)}{\mathcal{W}^J(t)} - \mathbf{U}^J(t) \cdot \mathbf{U}^J(t) \right)$$

Estimation of Statistical Moments

$$\mathcal{U}^J(t) = \mu \mathcal{U}^J(t - \Delta t) + (1 - \mu) \sum_{j=1}^{N_p} \left\{ \hat{g}^J(\mathbf{X}^j(t)) \mathbf{M}^j(t) \right\},$$

$$\mathcal{E}^J(t) = \mu \mathcal{E}^J(t - \Delta t) + (1 - \mu) \sum_{j=1}^{N_p} \left\{ \hat{g}^J(\mathbf{X}^j(t)) \mathbf{M}^j(t) \cdot \mathbf{M}^j(t) \right\}$$

$$\mathcal{W}^J(t) = \mu \mathcal{W}^J(t - \Delta t) + (1 - \mu) \sum_{j=1}^{N_p} \left\{ \hat{g}^J(\mathbf{X}^j(t)) \right\},$$



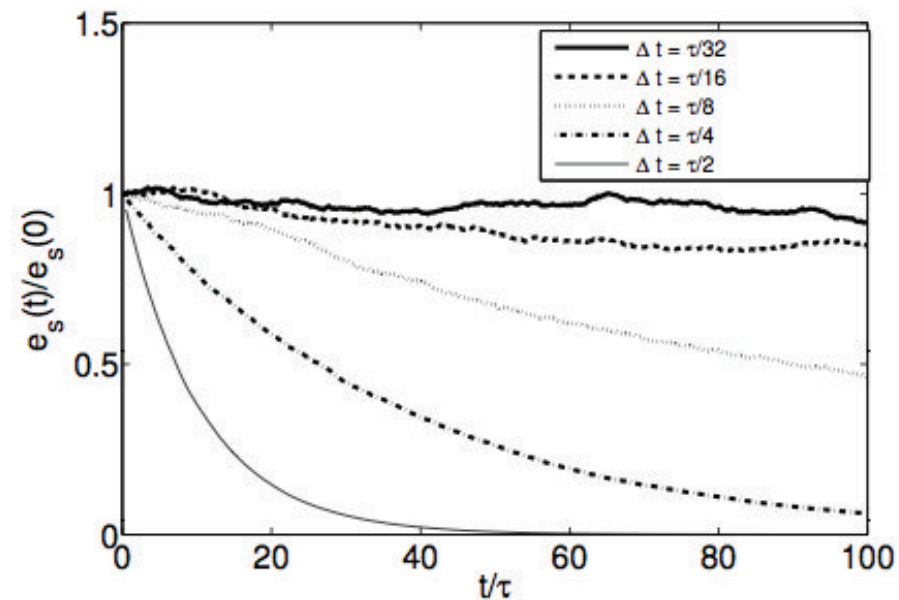
$$\mathbf{U}(\mathbf{x}^J, t) = \frac{\mathcal{U}^J(t)}{\mathcal{W}^J(t)}$$

$$\mathbf{e}_s(\mathbf{x}^J, t) = \frac{1}{2} \left(\frac{\mathcal{E}^J(t)}{\mathcal{W}^J(t)} - \mathbf{U}^J(t) \cdot \mathbf{U}^J(t) \right)$$

Particle Evolution

$$\frac{dX_i}{dt} = M_i \quad \text{with}$$

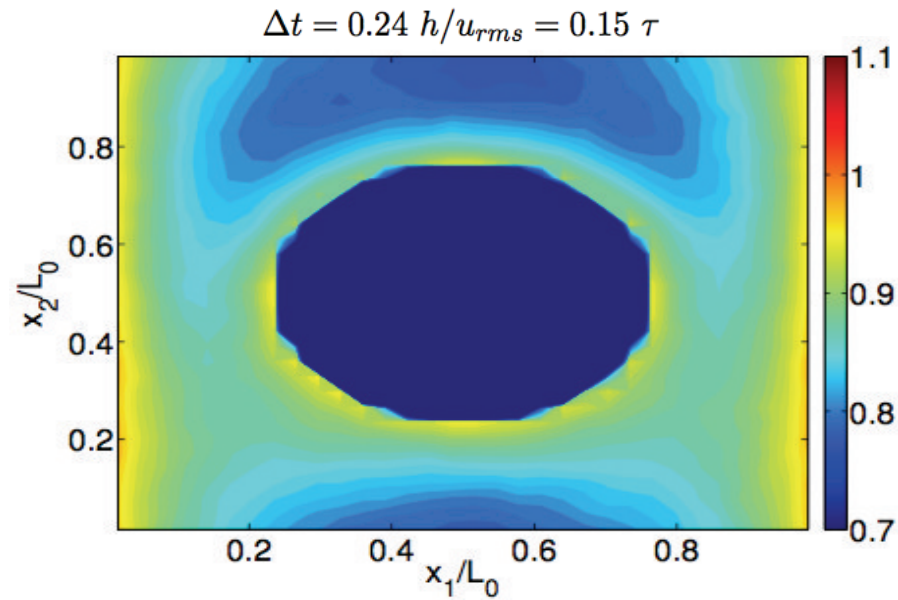
$$\frac{dM_i}{dt} = -\frac{1}{\tau} (M_i - U_i) + \left(\frac{4e_s}{3\tau}\right)^{1/2} \frac{dW_i(t)}{dt} + F_i$$



$$M_i^{n+1} - M_i^n = \left[-\frac{\Delta t}{\tau} M_i^n + \left(\frac{4e_s}{3\tau}\right)^{1/2} \xi_i \right] \left[1 - \frac{\Delta t}{2\tau} \right] \quad \text{and}$$

$$\Delta X_i^{n+1} = \frac{\Delta t}{2} (M_i^n + M_i^{n+1})$$

Particle Evolution



$$M_i^{n+1} - M_i^n = \left[-\frac{\Delta t}{\tau} M_i^n + \left(\frac{4e_s}{3\tau} \Delta t \right)^{1/2} \xi_i \right] \left[1 - \frac{\Delta t}{2\tau} \right] \quad \text{and}$$
$$\Delta X_i^{n+1} = \frac{\Delta t}{2} (M_i^n + M_i^{n+1})$$

Derivation of New Particle Evolution Scheme

(up to 2nd moments)

statistically exact for constant U and e_s for any time step Δt :

without loss of generality, $U_i = F_i = 0$ is assumed

solution of $\frac{dX_i}{dt} = M_i$ and

$$\frac{dM_i}{dt} = -\frac{1}{\tau}M_i + \left(\frac{4e_s}{3\tau}\right)^{1/2} \frac{dW_i(t)}{dt} \quad \text{is considered}$$

$$M_i^{n+1} = M_i^n e^{-\Delta t/\tau} + \lim_{N \rightarrow \infty} \sum_{k=1}^N \xi_{k,i} \left(\frac{4e_s}{3\tau} \frac{\Delta t}{N}\right)^{1/2} e^{-k\Delta t/(N\tau)}$$

$\xi_{k,i}$ are independent, normal distributed random variables

leads to the conditional expectation $\langle M_i^{n+1} M_j^{n+1} | \mathbf{M}^n \rangle = M_i^n M_j^n e^{-2\Delta t/\tau} + \delta_{ij} \lim_{N \rightarrow \infty} \sum_{k=1}^N \frac{4e_s}{3\tau} \frac{\Delta t}{N} e^{-2k\Delta t/(N\tau)}$

cross products disappear, since $\langle \xi_{k,i} \xi_{h,j} \rangle = \delta_{kh} \delta_{ij}$ and $\langle \xi_{k,i} \rangle = 0$

Derivation of New Particle Evolution Scheme

(up to 2nd moments)

statistically exact for constant U and e_s for any time step Δt :

without loss of generality, $U_i = F_i = 0$ is assumed

solution of $\frac{dX_i}{dt} = M_i$ and

$$\frac{dM_i}{dt} = -\frac{1}{\tau}M_i + \left(\frac{4e_s}{3\tau}\right)^{1/2} \frac{dW_i(t)}{dt} \quad \text{is considered}$$

$$M_i^{n+1} = M_i^n e^{-\Delta t/\tau} + \lim_{N \rightarrow \infty} \sum_{k=1}^N \xi_{k,i} \left(\frac{4e_s}{3\tau} \frac{\Delta t}{N}\right)^{1/2} e^{-k\Delta t/(N\tau)}$$

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$$\begin{aligned} \langle M_i^{n+1} M_j^{n+1} | M^n \rangle &= M_i^n M_j^n e^{-2\Delta t/\tau} + \delta_{ij} \frac{4e_s}{3\tau} \int_0^{\Delta t} e^{-2t/\tau} dt \\ &= M_i^n M_j^n e^{-2\Delta t/\tau} + \delta_{ij} \frac{2e_s}{3} (1 - e^{-2\Delta t/\tau}) \end{aligned}$$

therefore
$$M_i^{n+1} = M_i^n e^{-\Delta t/\tau} + \left(\frac{2e_s}{3} (1 - e^{-2\Delta t/\tau}) \right)^{1/2} \xi_{M,i}$$

Derivation of New Particle Evolution Scheme

$$M_i^{n+1} = M_i^n e^{-\Delta t/\tau} + \left(\frac{2e_s}{3} (1 - e^{-2\Delta t/\tau}) \right)^{1/2} \xi_{M,i}$$

preserves the internal energy $e_s = \frac{\langle M_i^n M_i^n \rangle}{2}$ independent of the time step size Δt

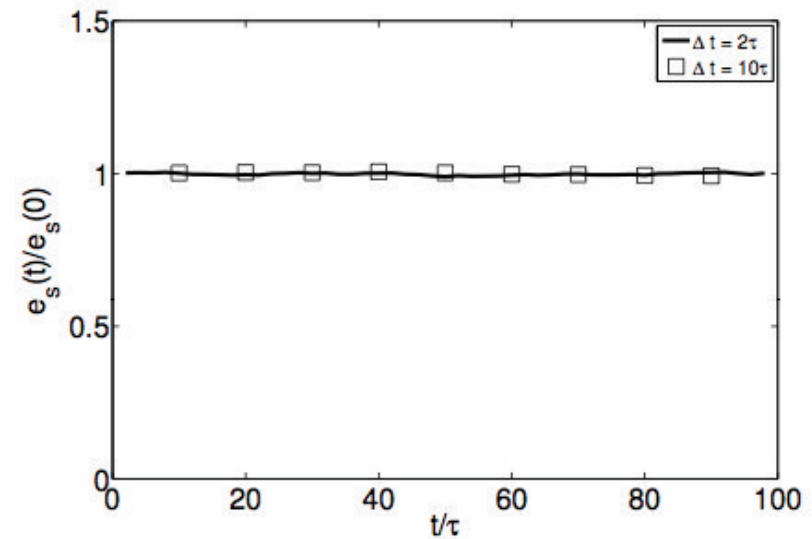
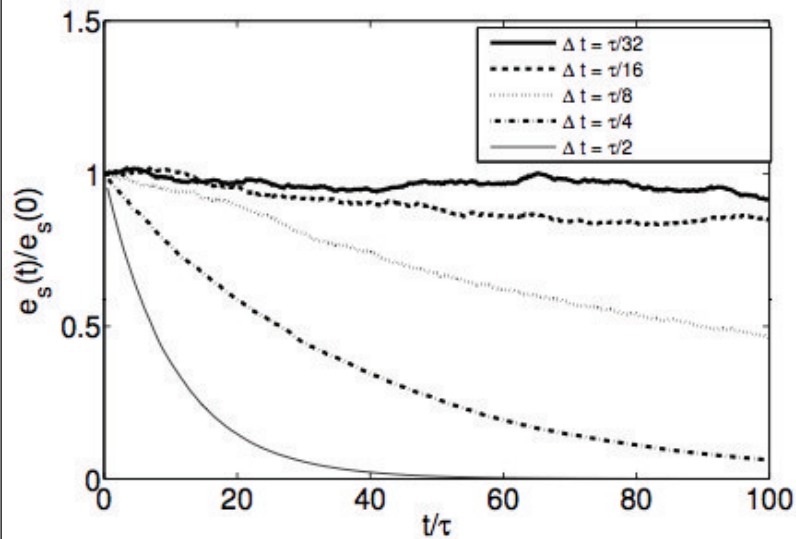
$$\begin{aligned} \langle M_i^{n+1} M_i^{n+1} \rangle &= \langle M_i^n M_i^n \rangle e^{-2\Delta t/\tau} + 2e_s (1 - e^{-2\Delta t/\tau}) \\ &= (\langle M_i^n M_i^n \rangle - 2e_s) e^{-2\Delta t/\tau} + 2e_s, \end{aligned}$$

analytically correct correlation coefficient

$$\frac{\langle M_i^{n+1} M_i^n \rangle}{\langle M_i^n M_i^n \rangle} = e^{-\Delta t/\tau}$$

Validation of the Exact Scheme

internal energy:



Derivation of New Particle Evolution Scheme

$$\frac{dX_i}{dt} = M_i \quad \text{and}$$
$$\frac{dM_i}{dt} = -\frac{1}{\tau} M_i + \left(\frac{4e_s}{3\tau}\right)^{1/2} \frac{dW_i(t)}{dt}$$

exact particle position scheme

$$\Delta X_i^{n+1} = X_i^{n+1} - X_i^n = M_i^n \tau (1 - e^{-\Delta t/\tau})$$
$$+ \lim_{N \rightarrow \infty} \sum_{k=1}^N \xi_{k,i} \left(\frac{4e_s}{3\tau} \frac{\Delta t}{N}\right)^{1/2} \tau (1 - e^{-k\Delta t/(N\tau)})$$

$$\langle \Delta X_i^{n+1} \Delta X_j^{n+1} | \mathbf{M}^n \rangle = M_i^n M_j^n \tau^2 (1 - e^{-\Delta t/\tau})^2$$
$$+ \delta_{ij} \lim_{N \rightarrow \infty} \sum_{k=1}^N \frac{4e_s}{3\tau} \frac{\Delta t}{N} \tau^2 (1 - e^{-k\Delta t/(N\tau)})^2$$

Derivation of New Particle Evolution Scheme

$$\begin{aligned} \langle \Delta X_i^{n+1} \Delta X_j^{n+1} | \mathbf{M}^n \rangle &= M_i^n M_j^n \tau^2 (1 - e^{-\Delta t/\tau})^2 \\ &+ \delta_{ij} \lim_{N \rightarrow \infty} \sum_{k=1}^N \frac{4e_s}{3\tau} \frac{\Delta t}{N} \tau^2 (1 - e^{-k\Delta t/(N\tau)})^2 \end{aligned}$$

which can be written as

$$\begin{aligned} \langle \Delta X_i^{n+1} \Delta X_j^{n+1} | \mathbf{M}^n \rangle &= M_i^n M_j^n \tau^2 (1 - e^{-\Delta t/\tau})^2 \\ &+ \delta_{ij} \frac{4e_s}{3} \tau \int_0^{\Delta t} (1 - 2e^{-t/\tau} + e^{-2t/\tau}) dt \\ &= M_i^n M_j^n \tau^2 (1 - e^{-\Delta t/\tau})^2 \\ &+ \delta_{ij} \frac{2e_s \tau^2}{3} \left(\frac{2\Delta t}{\tau} - (1 - e^{-\Delta t/\tau}) (3 - e^{-\Delta t/\tau}) \right) \end{aligned}$$

$$\Delta X_i^{n+1} = M_i^n \tau (1 - e^{-\Delta t/\tau})$$

$$+ \left(\frac{2e_s \tau^2}{3} \left(\frac{2\Delta t}{\tau} - (1 - e^{-\Delta t/\tau}) (3 - e^{-\Delta t/\tau}) \right) \right)^{1/2} \xi_{X,i}$$

Derivation of New Particle Evolution Scheme

$$M_i^{n+1} = M_i^n e^{-\Delta t/\tau} + \lim_{N \rightarrow \infty} \sum_{k=1}^N \xi_{k,i} \left(\frac{4e_s}{3\tau} \frac{\Delta t}{N} \right)^{1/2} e^{-k\Delta t/(N\tau)}$$

$$\begin{aligned} \Delta X_i^{n+1} &= X_i^{n+1} - X_i^n = M_i^n \tau (1 - e^{-\Delta t/\tau}) \\ &\quad + \lim_{N \rightarrow \infty} \sum_{k=1}^N \xi_{k,i} \left(\frac{4e_s}{3\tau} \frac{\Delta t}{N} \right)^{1/2} \tau (1 - e^{-k\Delta t/(N\tau)}) \end{aligned}$$

$$\begin{aligned} \left\langle M_i^{n+1} \Delta X_j^{n+1} \middle| \mathbf{M}^n \right\rangle &= M_i^n M_j^n \tau (e^{-\Delta t/\tau} - e^{-2\Delta t/\tau}) \\ &\quad + \delta_{ij} \lim_{N \rightarrow \infty} \sum_{k=1}^N \frac{4e_s}{3\tau} \frac{\Delta t}{N} \tau (e^{-k\Delta t/(N\tau)} - e^{-2k\Delta t/(N\tau)}) \\ &= M_i^n M_j^n \tau (e^{-\Delta t/\tau} - e^{-2\Delta t/\tau}) \\ &\quad + \delta_{ij} \frac{4e_s}{3} \int_0^{\Delta t} (e^{-t/\tau} - e^{-2t/\tau}) dt \\ &= M_i^n M_j^n \tau (e^{-\Delta t/\tau} - e^{-2\Delta t/\tau}) \\ &\quad + \delta_{ij} \underbrace{\frac{2e_s \tau}{3} (1 - e^{-\Delta t/\tau})^2}_C. \end{aligned}$$

Derivation of New Particle Evolution Scheme

$$M_i^{n+1} = M_i^n e^{-\Delta t/\tau} + \overbrace{\left(\frac{2e_s}{3} (1 - e^{-2\Delta t/\tau}) \right)^{1/2}}^{S_{M,i}} \xi_{M,i}$$

$$\Delta X_i^{n+1} = M_i^n \tau (1 - e^{-\Delta t/\tau})$$

$$+ \overbrace{\left(\frac{2e_s \tau^2}{3} \left(\frac{2\Delta t}{\tau} - (1 - e^{-\Delta t/\tau}) (3 - e^{-\Delta t/\tau}) \right) \right)^{1/2}}^{S_{X,i}} \xi_{X,i}$$

$$S'_{M,i} = \left(\frac{C^2}{B} \right)^{1/2} \xi_{1,i} + \left(A - \frac{C^2}{B} \right)^{1/2} \xi_{2,i} \quad \text{and}$$

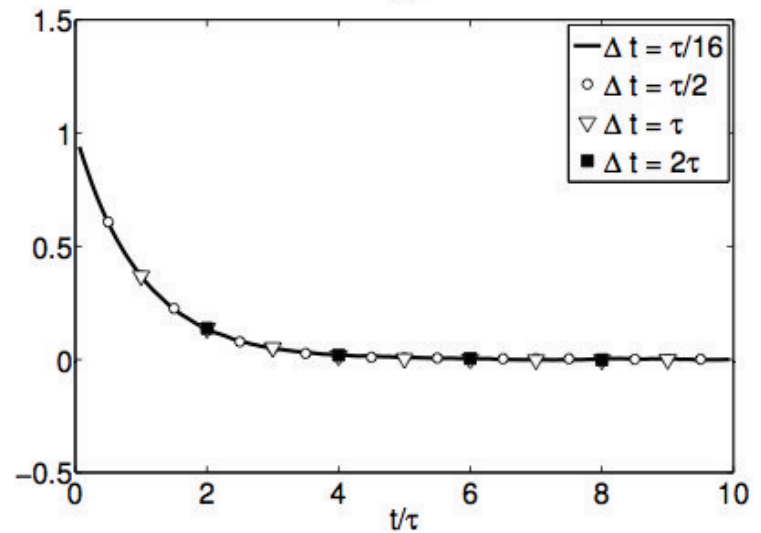
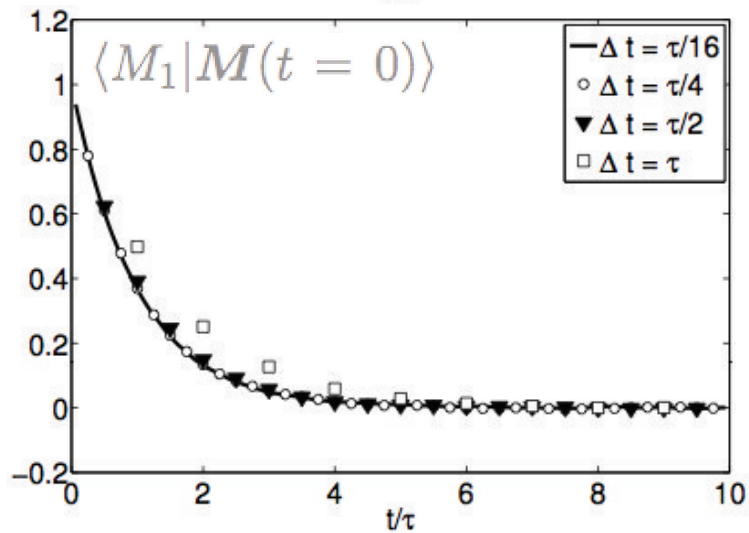
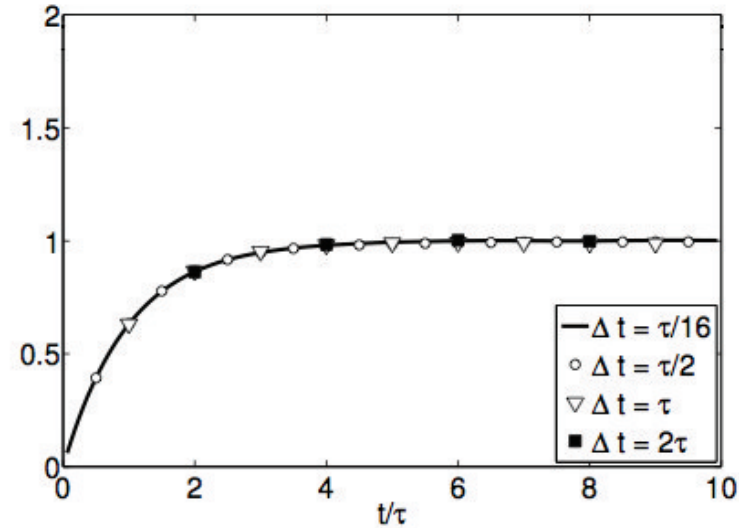
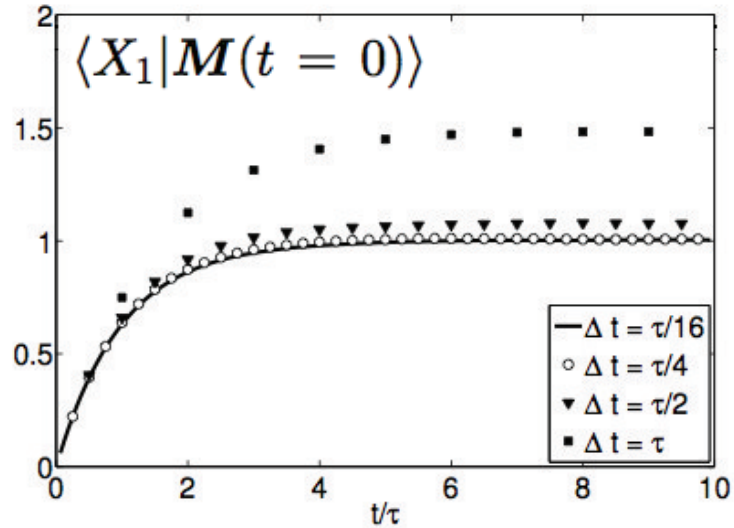
$$S'_{X,i} = (B)^{1/2} \xi_{1,i},$$

$$\langle S'_{M,i} S'_{M,j} \rangle = \langle S_{M,i} S_{M,j} \rangle = \delta_{ij} A \quad \text{and}$$

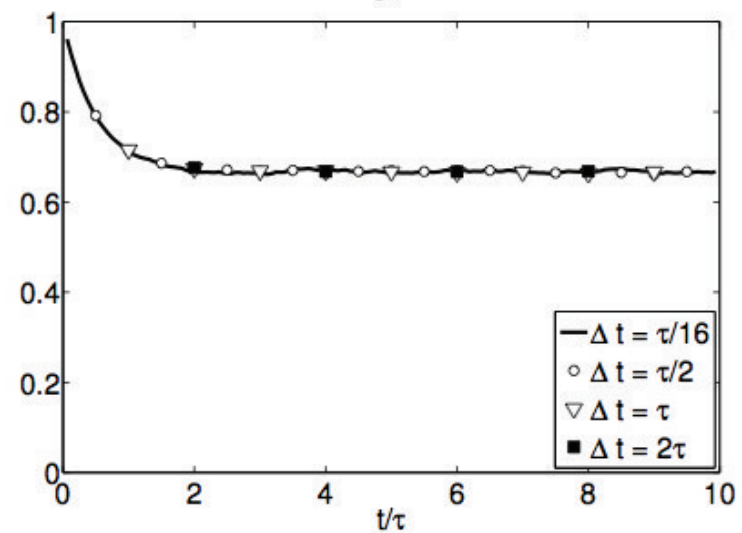
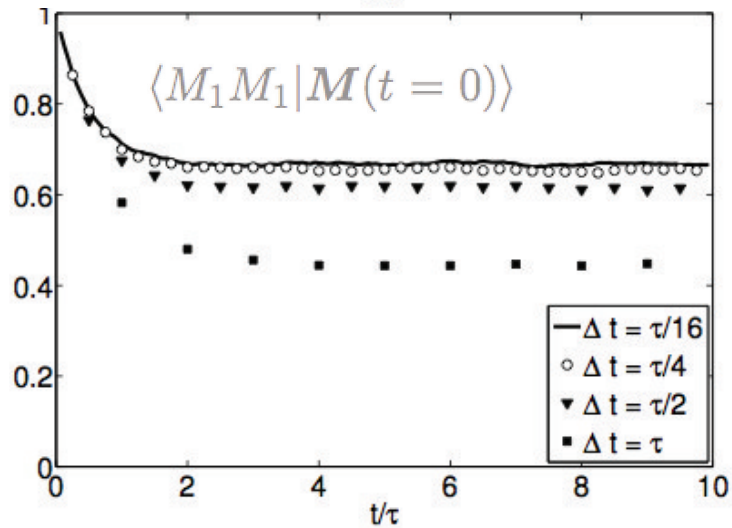
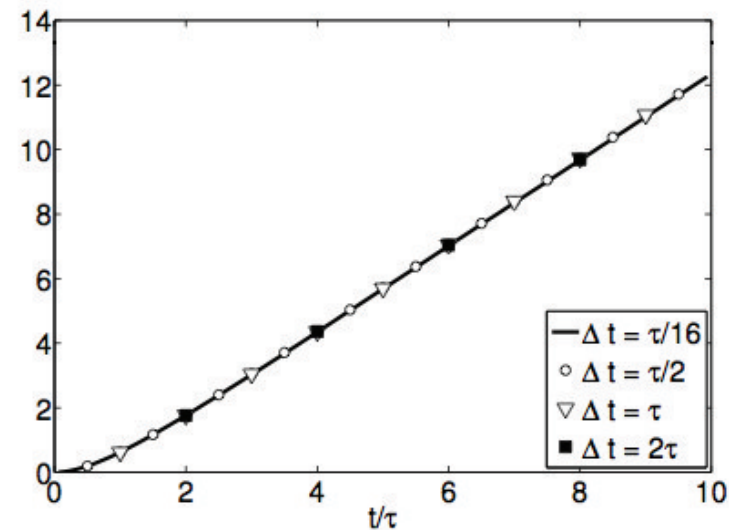
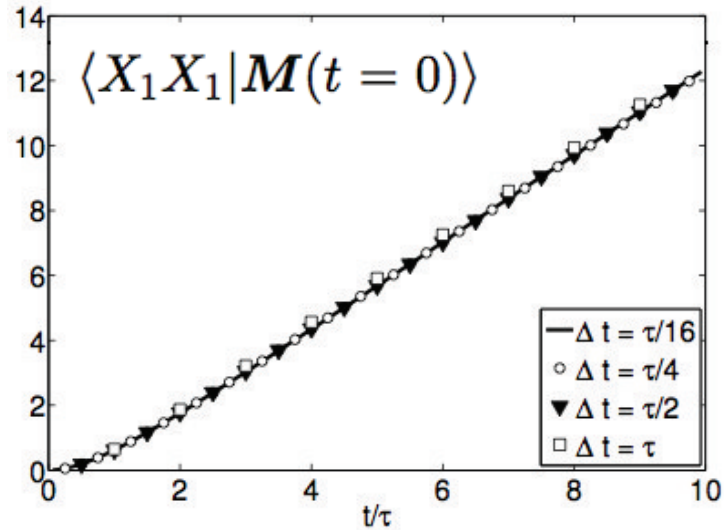
$$\langle S'_{X,i} S'_{X,j} \rangle = \langle S_{X,i} S_{X,j} \rangle = \delta_{ij} B,$$

$$\langle S'_{M,i} S'_{X,j} \rangle = C$$

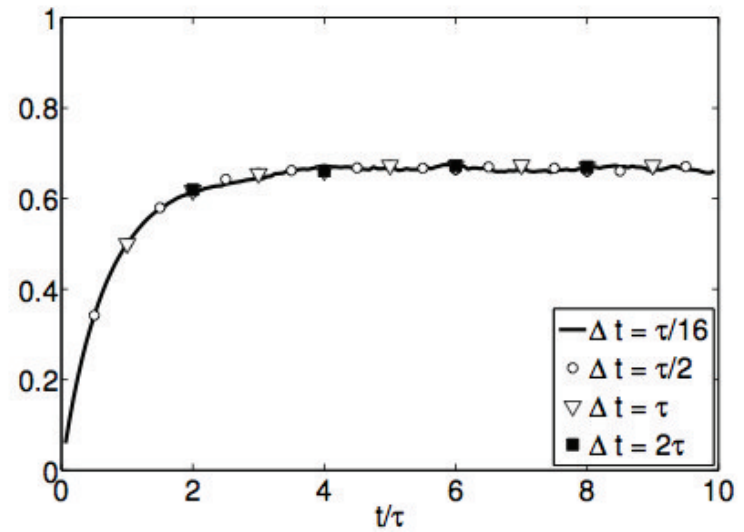
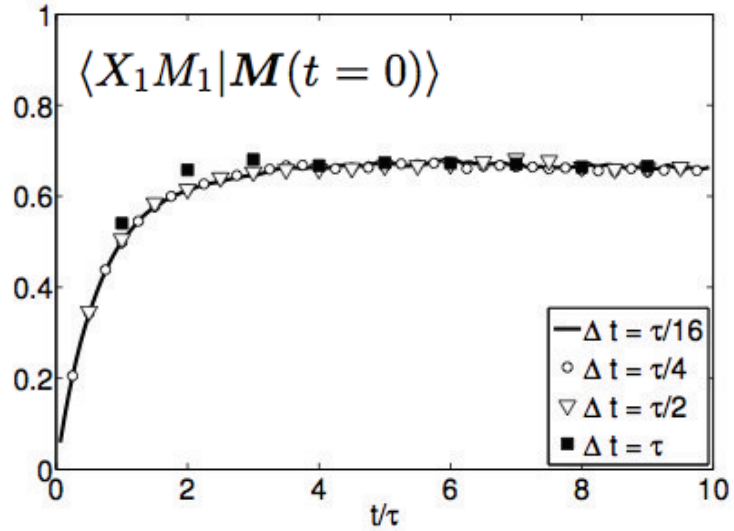
Validation of the Exact Scheme



Validation of the Exact Scheme



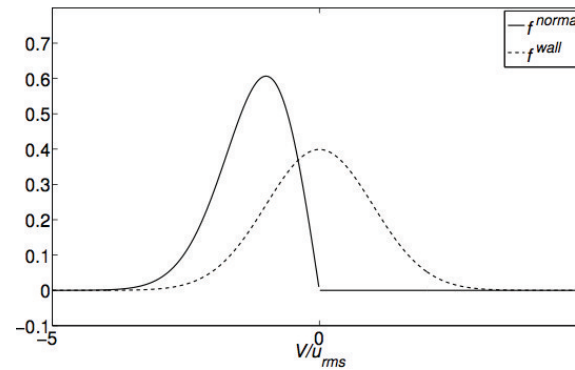
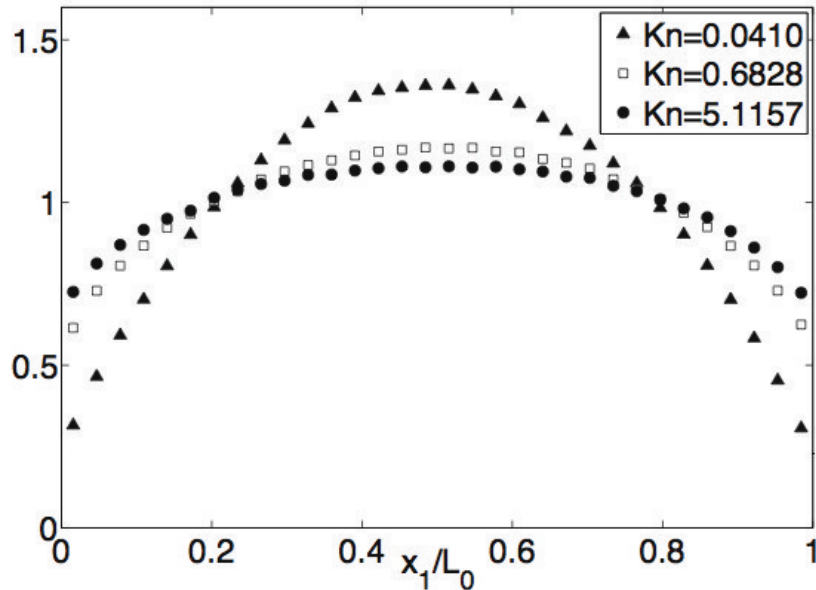
Validation of the Exact Scheme



Knudsen Paradox

Periodic-, open- and wall boundary conditions

isothermal wall is treated as an interface between computational and a virtual domain



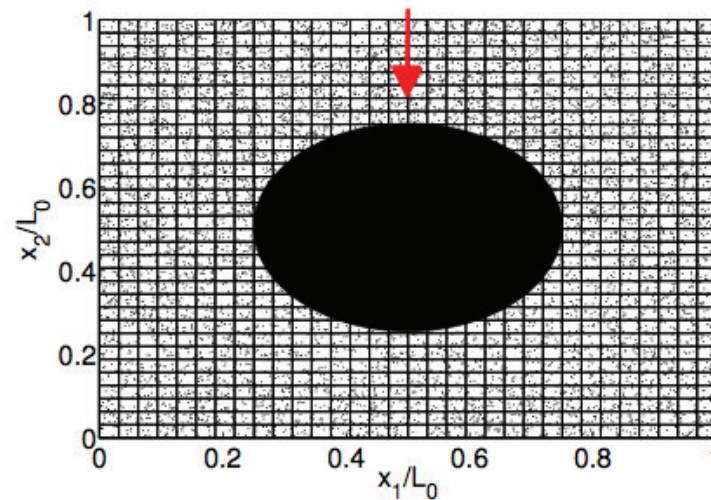
$$f^{wall} = \frac{1}{(2\pi kT^{wall}/m)^{1/2}} \exp\left(-\frac{V_i^2}{2kT^{wall}/m}\right)$$

$$f^{normal} = \frac{-H(-V_n)V_n}{Q} \frac{1}{(2\pi kT^{wall}/m)^{1/2}} \exp\left(-\frac{V_n^2}{2kT^{wall}/m}\right)$$

Flow Around a Cylinder

$$Kn = \lambda / (0.25L_0) = 2\sqrt{RT^{wall}}\tau / L_0 = 0.1$$

$$\hat{F} = FL_0 / (RT^{wall}) = 0.4$$

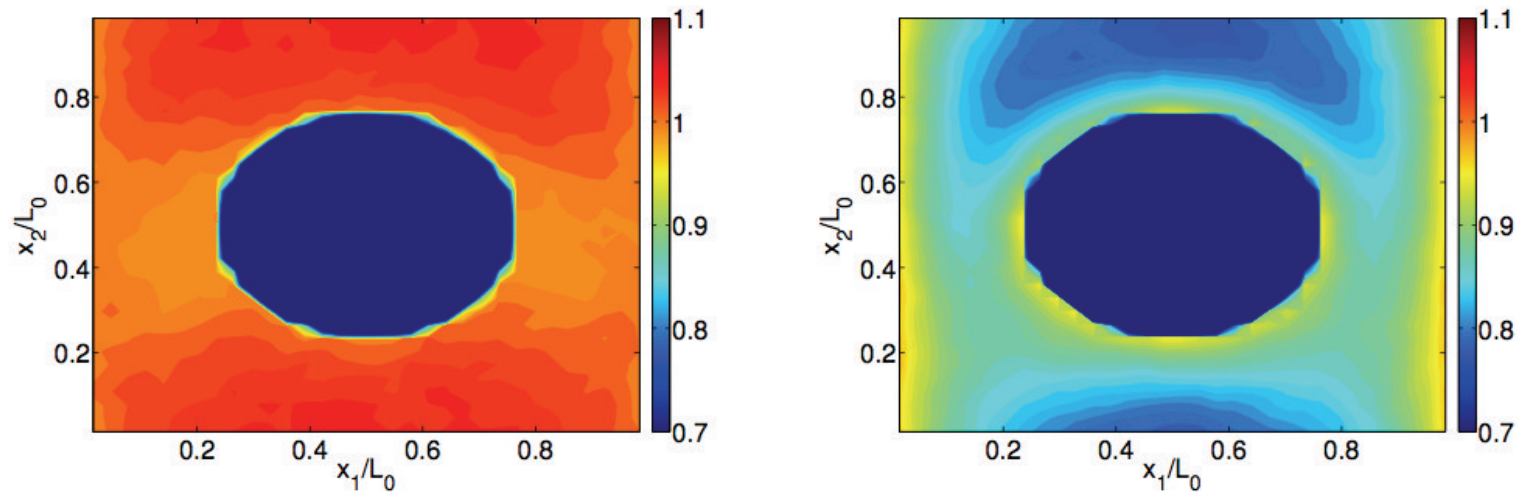


64 × 64, 32 × 32, 16 × 16 and 8 × 8 grids

10 particles / cell averaging factor n_a of 10'000

$$\Delta t = 0.24h / u_{rms}$$

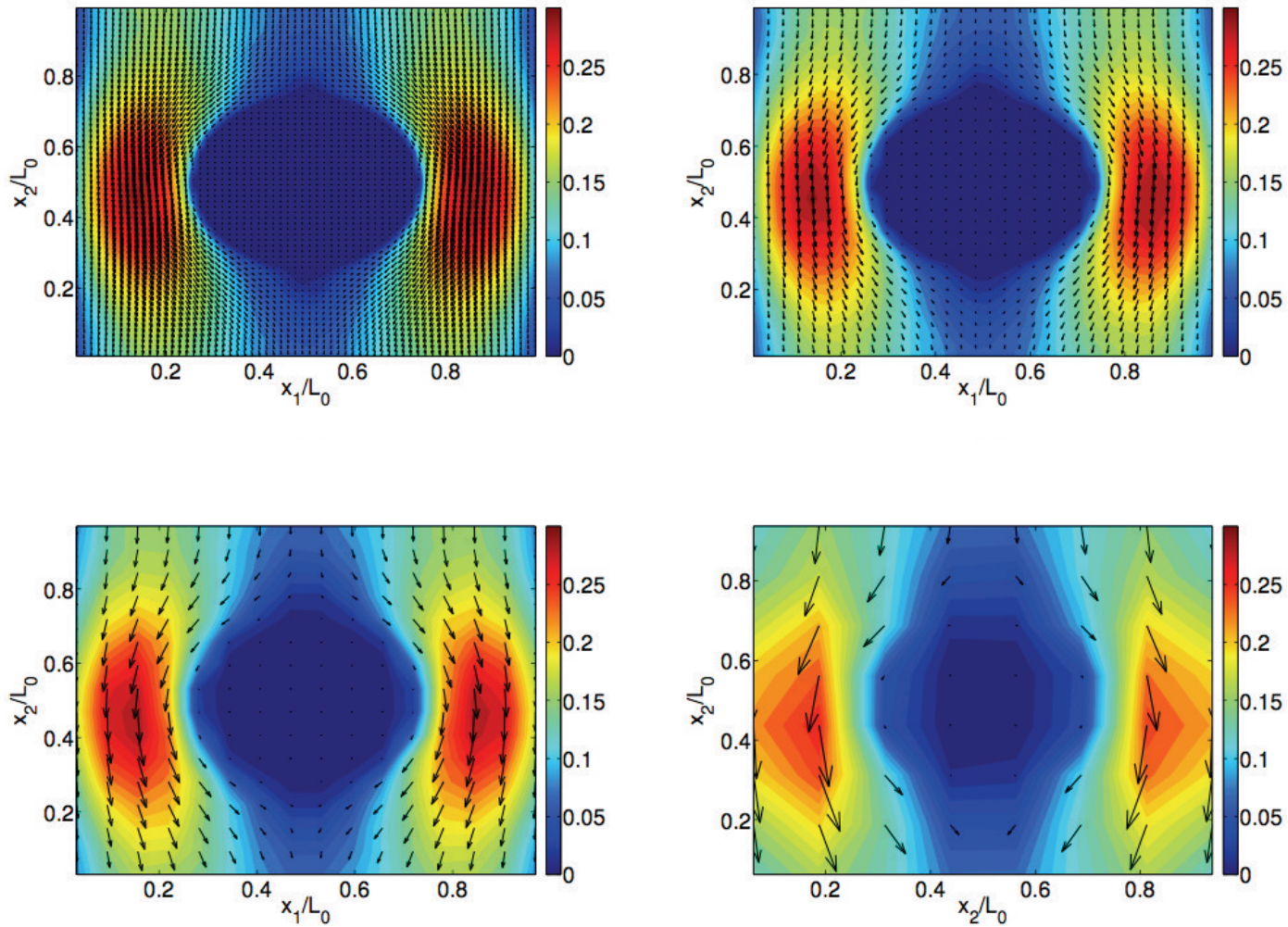
Flow Around a Cylinder



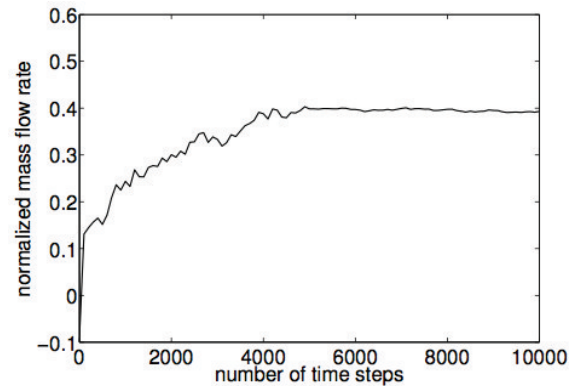
32×32 grid

$$\Delta t = 0.24 h/u_{rms} = 0.15 \tau$$

Flow Around a Cylinder



Performance

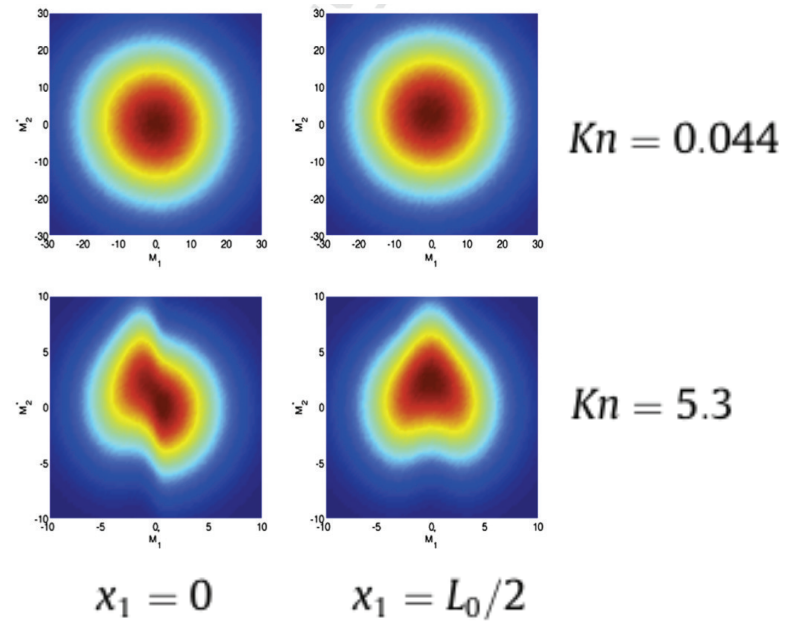
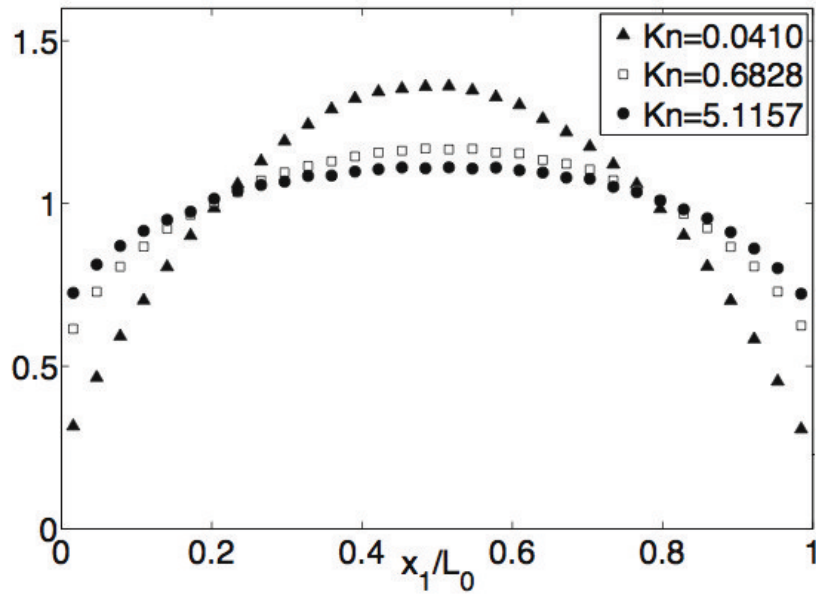


Knudsen paradox (Kn=5.1157):
$$\frac{\partial \mathcal{F}}{\partial t} + V_i \frac{\partial \mathcal{F}}{\partial x_i} + \frac{\partial}{\partial V_i} \left\{ \left[F_i - \frac{1}{\tau} (V_i - U_i) \right] \mathcal{F} \right\} = \frac{\partial^2}{\partial V_i \partial V_i} \left\{ \frac{2e_s}{3\tau} \mathcal{F} \right\}$$

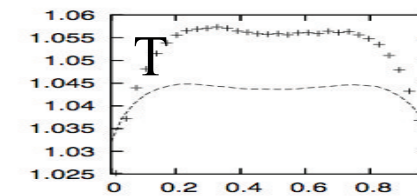
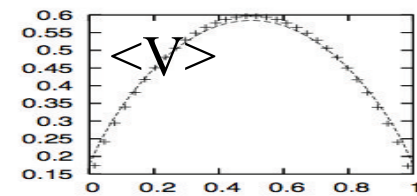
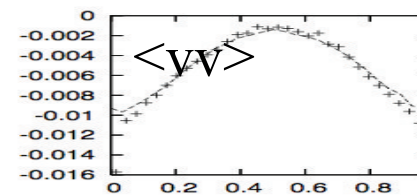
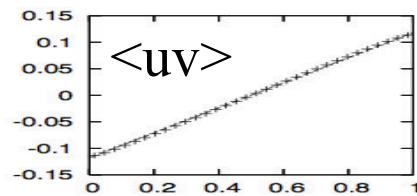
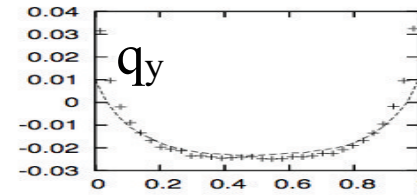
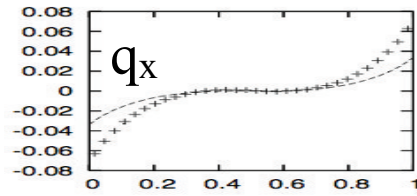
number of grid cells	\hat{J}/\hat{F}
64	1.315
32	1.317
16	1.327
8	1.341
4	1.383

number of grid cells	Δt	\hat{J}/\hat{F}
64×64	0.075τ	0.368653
32×32	0.15τ	0.378591
16×16	0.3τ	0.385789
8×8	0.6τ	0.387631

Knudsen Paradox

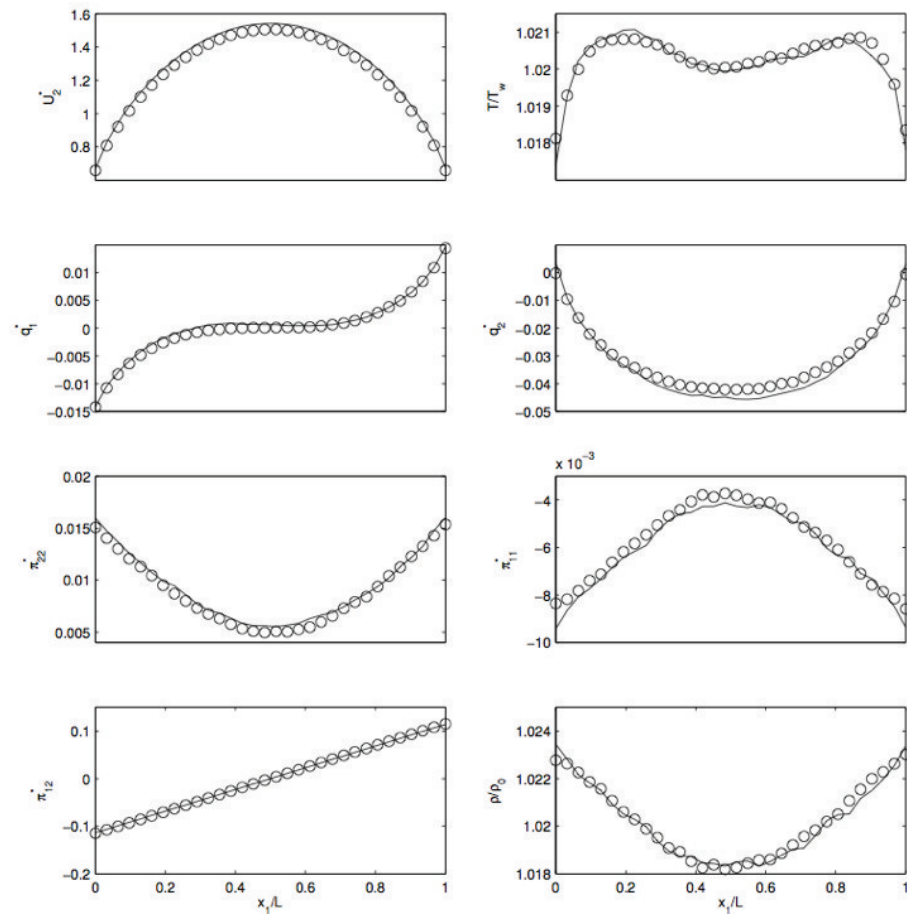


Knudsen Paradox



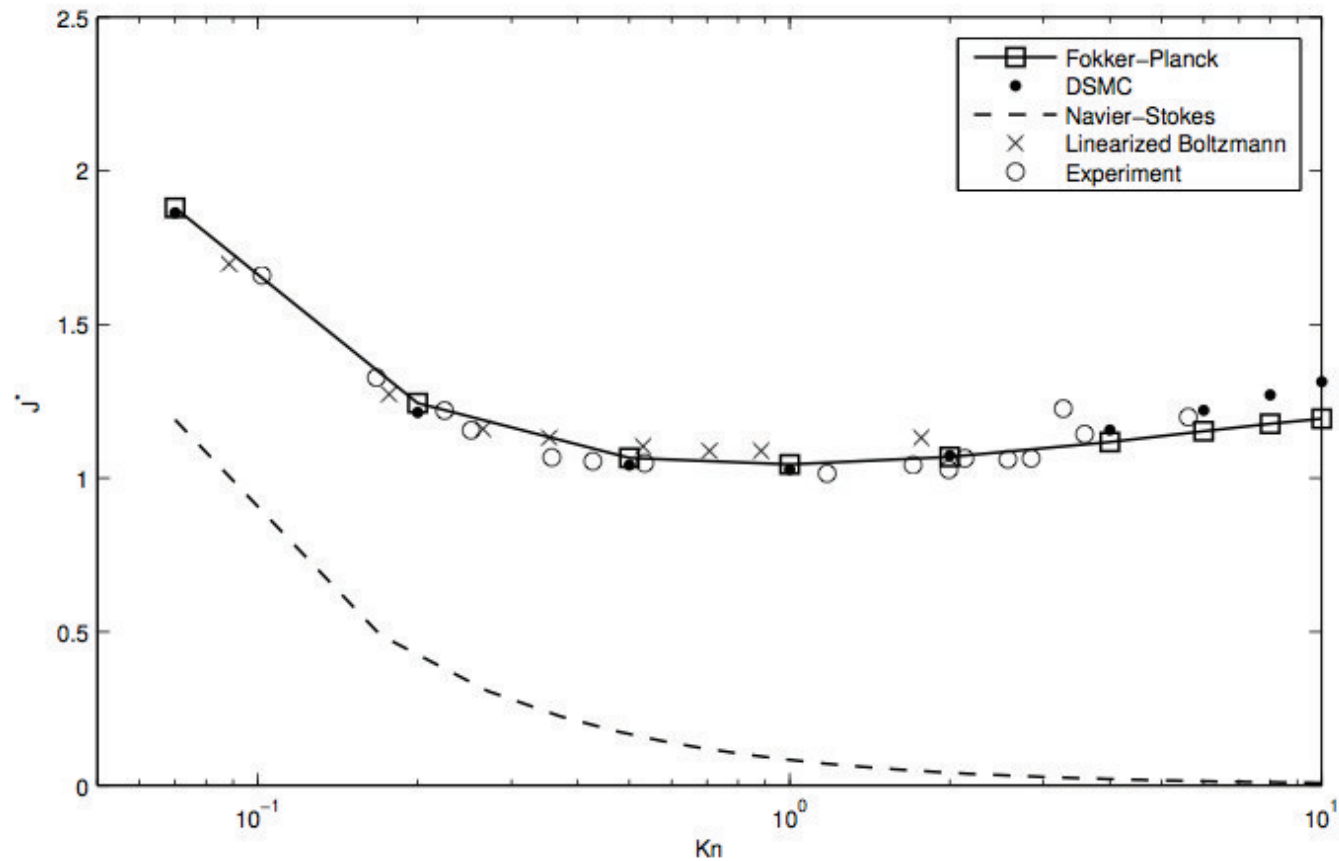
$Kn=0.072$

Results with a Cubic Model



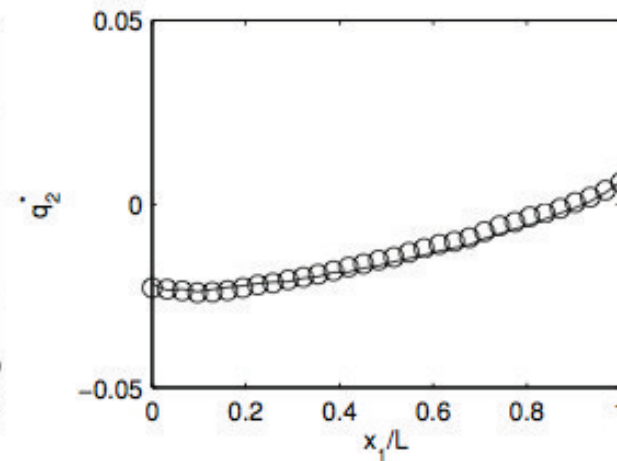
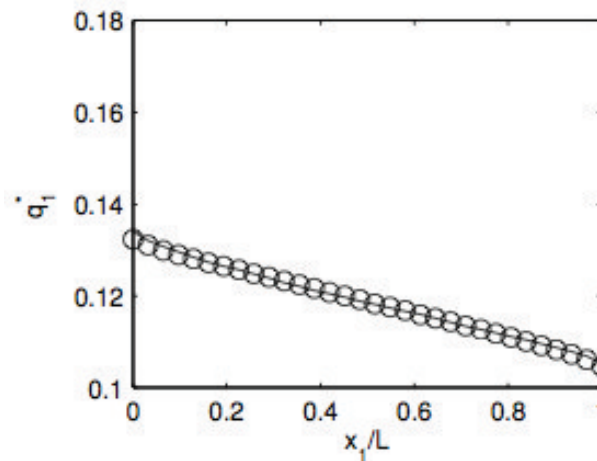
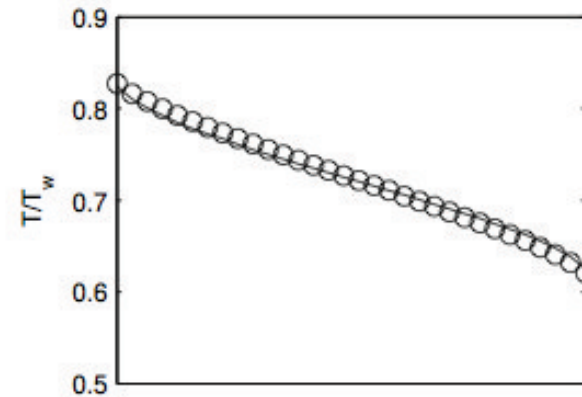
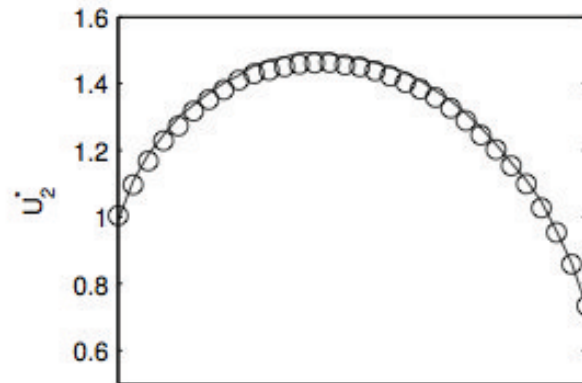
Kn=0.2

Results with a Cubic Model



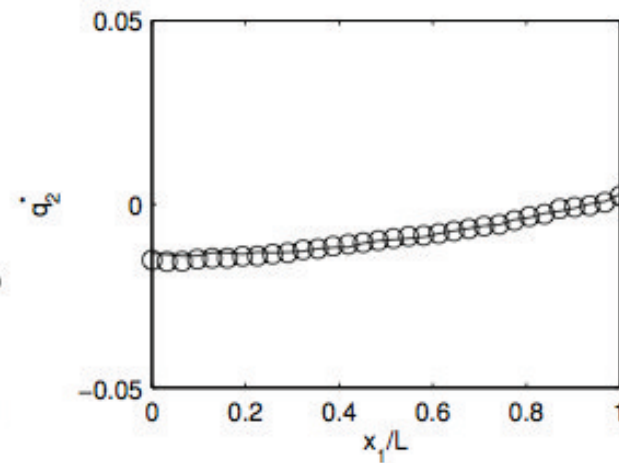
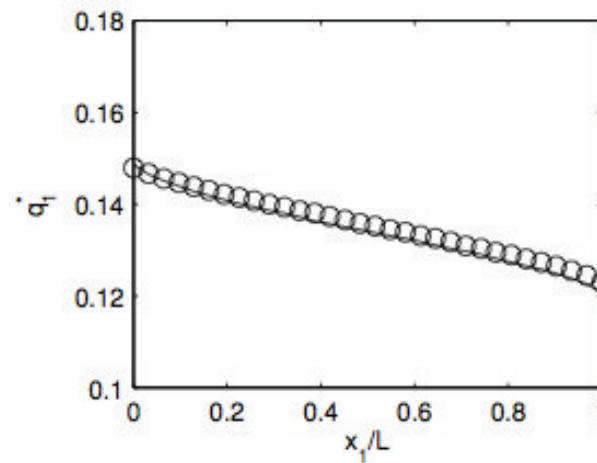
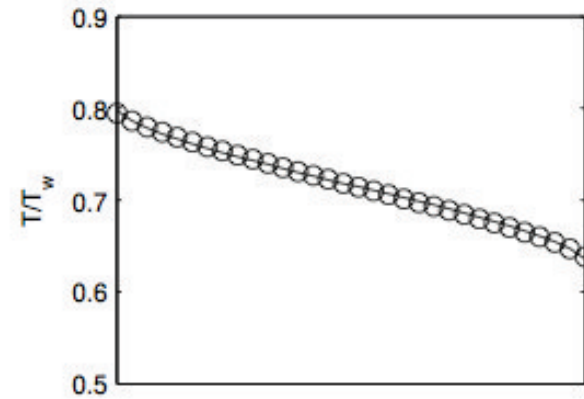
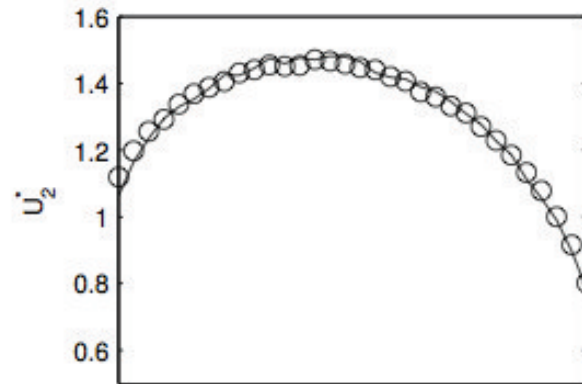
Results with a Cubic Model

Kn=1



Results with a Cubic Model

Kn=2



Multiscale Modeling Framework

Continuum:

$$\begin{aligned} \frac{\partial \rho}{\partial t} + \frac{\partial \rho U_j}{\partial x_j} &= 0, \\ \frac{\partial \rho U_i}{\partial t} + \frac{\partial \rho U_i U_j}{\partial x_j} + \frac{\partial p_{ij}}{\partial x_j} &= \rho F_i, \\ \frac{\partial \rho e_s}{\partial t} + \frac{\partial \rho U_j e_s}{\partial x_j} + \frac{\partial q_j}{\partial x_j} + p_{jk} \frac{\partial U_j}{\partial x_k} &= 0. \end{aligned}$$

Particles:

$$\begin{aligned} M_i^{n+1} &= \dots \\ \Delta X_i^{n+1} &= \dots \end{aligned}$$



Conclusion

- Fokker-Planck collision operator for monatomic gas flow
- Good prediction of Knudsen paradox
- “Exact” particle integration scheme allows for large time steps
- Efficient solution algorithm
- Hybrid framework for multiscale modeling
- Prandtl number problem
- Further validation with Boltzmann and experiments