

A Hybrid Particle-Continuum Method Coupling a Fluctuating Fluid with Suspended Structures

Aleksandar Donev¹

Courant Institute, *New York University*

&

Alejandro L. Garcia, *San Jose State University*

John B. Bell, *Lawrence Berkeley National Laboratory*

¹This work performed in part under the auspices of the U.S. Department of Energy by Lawrence Livermore National Laboratory under Contract DE-AC52-07NA27344.

AMS von Neumann Symposium
Snowbird, Utah
July 6th, 2011

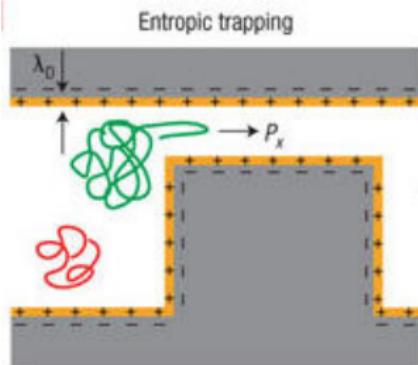
Outline

- 1 Introduction
- 2 Particle Methods
- 3 Fluctuating Hydrodynamics
- 4 Hybrid Particle-Continuum Method
- 5 The Importance of Thermal Fluctuations
 - Brownian Bead
 - Adiabatic Piston
- 6 Fluctuation-Enhanced Diffusion
- 7 Conclusions

Micro- and nano-hydrodynamics

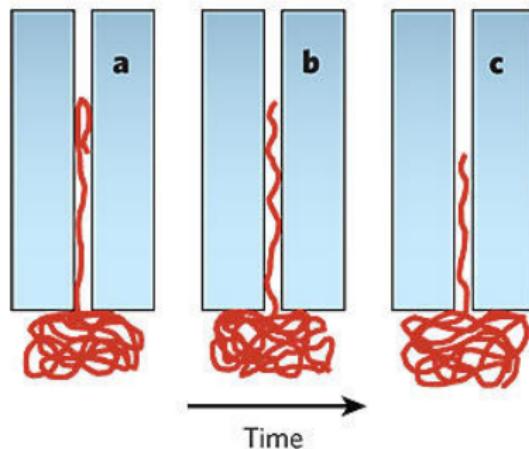
- Flows of fluids (gases and liquids) through micro- (μm) and nano-scale (nm) structures has become technologically important, e.g., **micro-fluidics, microelectromechanical systems (MEMS)**.
- **Biologically-relevant** flows also occur at micro- and nano- scales.
- An important feature of small-scale flows, not discussed here, is **surface/boundary effects** (e.g., slip in the contact line problem).
- Essential distinguishing feature from “ordinary” CFD: **thermal fluctuations!**
- I focus here not on the technical details of hybrid methods, but rather, on using our method to demonstrate the general conclusion that **fluctuations should be taken into account at the continuum level**.

Example: DNA Filtering



Fu et al., *Nature*

Nanotechnology 2 (2007)



H. Craighead, *Nature* 442 (2006)

How to coarse grain the fluid (solvent) and couple it to the suspended microstructure (e.g., polymer chain)?

Levels of Coarse-Graining

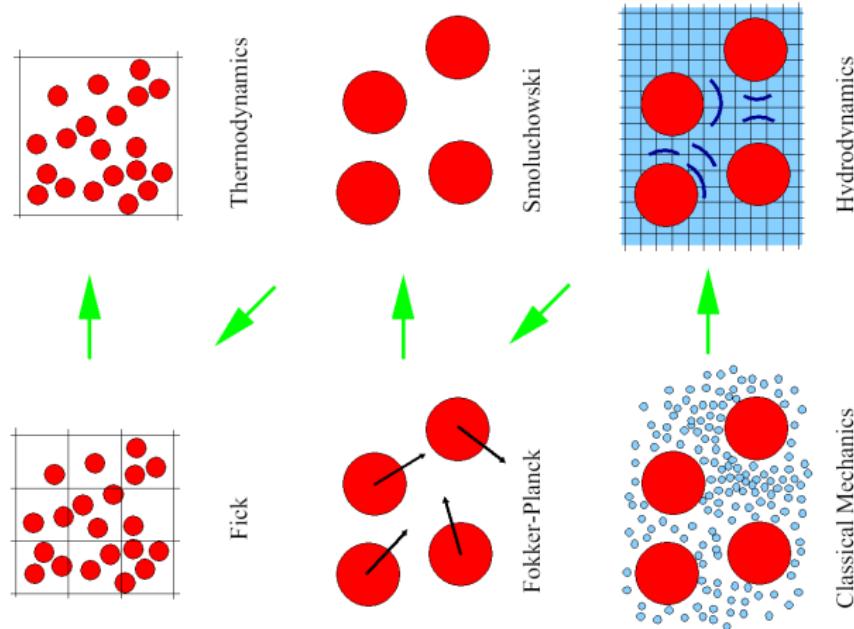


Figure: From Pep Espa ol, "Statistical Mechanics of Coarse-Graining"

This talk: Particle/Continuum Hybrid

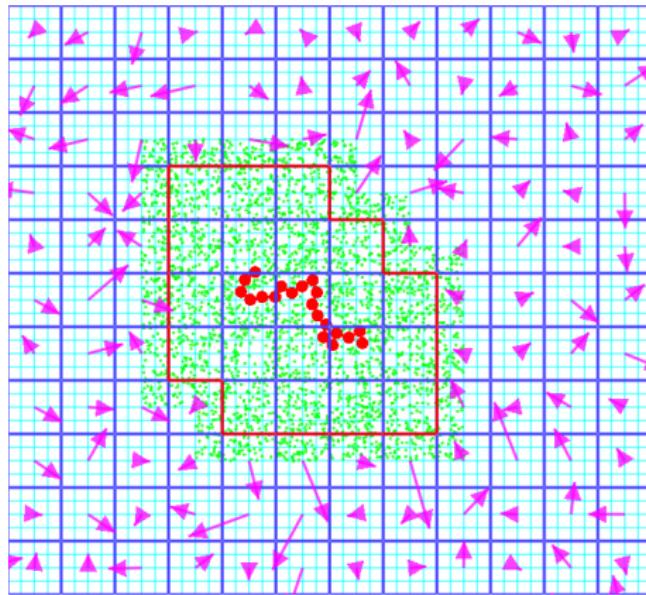


Figure: Hybrid method for a polymer chain.

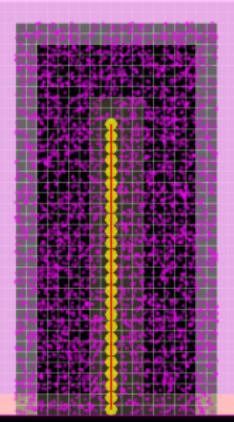
Particle Methods for Complex Fluids

- The most direct and accurate way to simulate the interaction between the **solvent** (fluid) and **solute** (beads, chain) is to use a particle scheme for both: **Molecular Dynamics (MD)**

$$m\ddot{\mathbf{r}}_i = \sum_j \mathbf{f}_{ij}(\mathbf{r}_{ij})$$

- The stiff repulsion among beads demands small time steps, and chain-chain crossings are a problem.
- Most of the computation is “wasted” on the *unimportant solvent particles!*
- Over longer times it is **hydrodynamics** (*local momentum and energy conservation*) and **fluctuations** (Brownian motion) that matter.
- We need to coarse grain the fluid model further: *Replace deterministic interactions with stochastic collisions.*

Direct Simulation Monte Carlo (DSMC)



(MNG)

Tethered polymer chain in shear flow.

- **Stochastic conservative collisions** of randomly chosen nearby solvent particles, as in DSMC (also related to MPCD/SRD and DPD).
- Solute particles still interact with **both** solvent and other solute particles as hard or soft spheres.
- **No fluid structure**: Viscous ideal gas.
- One can introduce biased collision models to give the fluids consistent structure and a **non-ideal equation of state**. [1].

Continuum Models of Fluid Dynamics

- Formally, we consider the continuum field of **conserved quantities**

$$\mathbf{U}(\mathbf{r}, t) = \begin{bmatrix} \rho \\ \mathbf{j} \\ e \end{bmatrix} \cong \widetilde{\mathbf{U}}(\mathbf{r}, t) = \sum_i \begin{bmatrix} m_i \\ m_i \mathbf{v}_i \\ m_i v_i^2 / 2 \end{bmatrix} \delta [\mathbf{r} - \mathbf{r}_i(t)],$$

where the symbol \cong means that $\mathbf{U}(\mathbf{r}, t)$ approximates the true atomistic configuration $\widetilde{\mathbf{U}}(\mathbf{r}, t)$ over **long length and time scales**.

- Formal coarse-graining of the microscopic dynamics has been performed to derive an **approximate closure** for the macroscopic dynamics [2].
- This leads to **SPDEs of Langevin type** formed by postulating a **white-noise random flux** term in the usual Navier-Stokes-Fourier equations with magnitude determined from the **fluctuation-dissipation balance** condition, following Landau and Lifshitz.

Compressible Fluctuating Hydrodynamics

$$D_t \rho = -\rho \nabla \cdot \mathbf{v}$$

$$\rho (D_t \mathbf{v}) = -\nabla P + \nabla \cdot (\eta \bar{\nabla} \mathbf{v} + \boldsymbol{\Sigma})$$

$$\rho c_p (D_t T) = D_t P + \nabla \cdot (\mu \nabla T + \boldsymbol{\Xi}) + (\eta \bar{\nabla} \mathbf{v} + \boldsymbol{\Sigma}) : \nabla \mathbf{v},$$

where the variables are the **density** ρ , **velocity** \mathbf{v} , and **temperature** T fields,

$$D_t \square = \partial_t \square + \mathbf{v} \cdot \nabla (\square)$$

$$\bar{\nabla} \mathbf{v} = (\nabla \mathbf{v} + \nabla \mathbf{v}^T) - 2(\nabla \cdot \mathbf{v}) \mathbf{I}/3$$

and capital Greek letters denote stochastic fluxes:

$$\boldsymbol{\Sigma} = \sqrt{2\eta k_B T} \mathcal{W}.$$

$$\langle \mathcal{W}_{ij}(\mathbf{r}, t) \mathcal{W}_{kl}^*(\mathbf{r}', t') \rangle = (\delta_{ik}\delta_{jl} + \delta_{il}\delta_{jk} - 2\delta_{ij}\delta_{kl}/3) \delta(t - t') \delta(\mathbf{r} - \mathbf{r}').$$

Landau-Lifshitz Navier-Stokes (LLNS) Equations

- The **non-linear LLNS equations are ill-behaved stochastic PDEs**, and we do not really know how to interpret the nonlinearities precisely.
- Finite-volume discretizations naturally impose a grid-scale **regularization** (smoothing) of the stochastic forcing.
- A **renormalization** of the transport coefficients is also necessary [3].
- We have algorithms and codes to solve the compressible equations (collocated and staggered grid), and recently also the incompressible ones (staggered grid) [4, 5].
- Solving the LLNS equations numerically requires paying attention to **discrete fluctuation-dissipation balance**, in addition to the usual deterministic difficulties [4].

Finite-Volume Schemes

$$c_t = -\mathbf{v} \cdot \nabla c + \chi \nabla^2 c + \nabla \cdot (\sqrt{2\chi} \mathbf{W}) = \nabla \cdot [-c\mathbf{v} + \chi \nabla c + \sqrt{2\chi} \mathbf{W}]$$

- Generic **finite-volume spatial discretization**

$$\mathbf{c}_t = \mathbf{D} [(-\mathbf{V}\mathbf{c} + \mathbf{G}\mathbf{c}) + \sqrt{2\chi / (\Delta t \Delta V)} \mathbf{W}],$$

where \mathbf{D} : faces \rightarrow cells is a **conservative** discrete divergence,
 \mathbf{G} : cells \rightarrow faces is a discrete gradient.

- Here \mathbf{W} is a collection of random normal numbers representing the (face-centered) stochastic fluxes.
- The **divergence** and **gradient** should be **duals**, $\mathbf{D}^* = -\mathbf{G}$.
- Advection should be **skew-adjoint** (non-dissipative) if $\nabla \cdot \mathbf{v} = 0$,

$$(\mathbf{D}\mathbf{v})^* = -(\mathbf{D}\mathbf{v}) \text{ if } (\mathbf{D}\mathbf{v})\mathbf{1} = \mathbf{0}.$$

Weak Accuracy

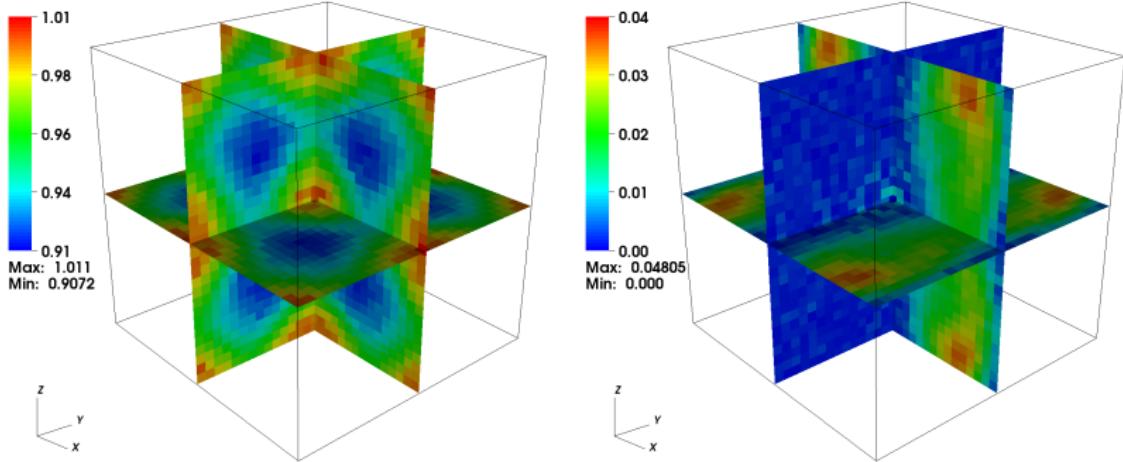


Figure: Equilibrium discrete spectra (static structure factors) $S_{\rho,\rho}(\mathbf{k}) \sim \langle \hat{\rho} \hat{\rho}^* \rangle$ (should be unity for all discrete wavenumbers) and $S_{\rho,v}(\mathbf{k}) \sim \langle \hat{\rho} \hat{v}_x^* \rangle$ (should be zero) for our RK3 collocated scheme.

Fluid-Structure Coupling using Particles

- Split the domain into a **particle** and a **continuum (hydro) subdomains**, with timesteps $\Delta t_H = K \Delta t_P$.
- Hydro solver is a simple explicit **(fluctuating) compressible LLNS** code and is *not aware* of particle patch.
- The method is based on Adaptive Mesh and Algorithm Refinement (AMAR) methodology for conservation laws and ensures **strict conservation** of mass, momentum, *and* energy.

MNG

Continuum-Particle Coupling

- Each macro (hydro) cell is either **particle or continuum**. There is also a **reservoir region** surrounding the particle subdomain.
- The coupling is roughly of the **state-flux** form:
 - The continuum solver provides *state boundary conditions* for the particle subdomain via reservoir particles.
 - The particle subdomain provides *flux boundary conditions* for the continuum subdomain.
- The fluctuating hydro solver is **oblivious** to the particle region: Any conservative explicit finite-volume scheme can trivially be substituted.
- The coupling is greatly simplified because the ideal **particle fluid has no internal structure**.

"A hybrid particle-continuum method for hydrodynamics of complex fluids", A. Donev and J. B. Bell and A. L. Garcia and B. J. Alder, **SIAM J. Multiscale Modeling and Simulation 8(3):871-911, 2010**

Our Hybrid Algorithm

- ① The hydro solution \mathbf{u}_H is computed everywhere, including the **particle patch**, giving an estimated total flux Φ_H .
- ② **Reservoir particles** are *inserted* at the boundary of the particle patch based on *Chapman-Enskog distribution* from kinetic theory, accounting for *both* collisional and kinetic viscosities.
- ③ Reservoir particles are *propagated* by Δt and *collisions* are processed, giving the total particle flux Φ_p .
- ④ The hydro solution is overwritten in the particle patch based on the particle state \mathbf{u}_p .
- ⑤ The hydro solution is corrected based on the more accurate flux,
$$\mathbf{u}_H \leftarrow \mathbf{u}_H - \Phi_H + \Phi_p.$$

Other Hybrid Algorithms

- For molecular dynamics (non-ideal particle fluids) the insertion of reservoir particles is greatly complicated by the need to account for the **internal structure** of the fluid and requires an **overlap region**.
- A hybrid method based on a flux-flux coupling between molecular dynamics and isothermal compressible fluctuating hydrodynamics has been developed by Coveney, De Fabritiis, Delgado-Buscalioni and co-workers [6].
- Some comparisons between different forms of coupling (state-state, state-flux, flux-state, flux-flux) has been performed by Ren [7].
- Reaching relevant time scales ultimately requires a **stochastic immersed structure** approach coupling immersed structures directly to a fluctuating solver (work in progress).

Brownian Bead

- Thermal fluctuations *push a sphere of size a and density ρ' suspended in a stationary fluid with density ρ and viscosity η (Brownian walker)* with initial velocity $V_{th} \approx \sqrt{kT/M}$, $M \approx \rho' a^3$.
- The classical picture of Brownian motion indicates three **widely-separated timescales**:
 - **Sound waves** are generated from the sudden compression of the fluid and they take away a fraction of the kinetic energy during a **sonic time** $t_{sonic} \approx a/c$, where c is the (adiabatic) sound speed.
 - **Viscous dissipation** then takes over and slows the particle non-exponentially over a **viscous time** $t_{visc} \approx \rho a^2/\eta$, where η is the shear viscosity.
 - **Thermal fluctuations** get similarly dissipated, but their constant presence pushes the particle diffusively over a **diffusion time** $t_{diff} \approx a^2/D$, where $D \sim kT/(a\eta)$.

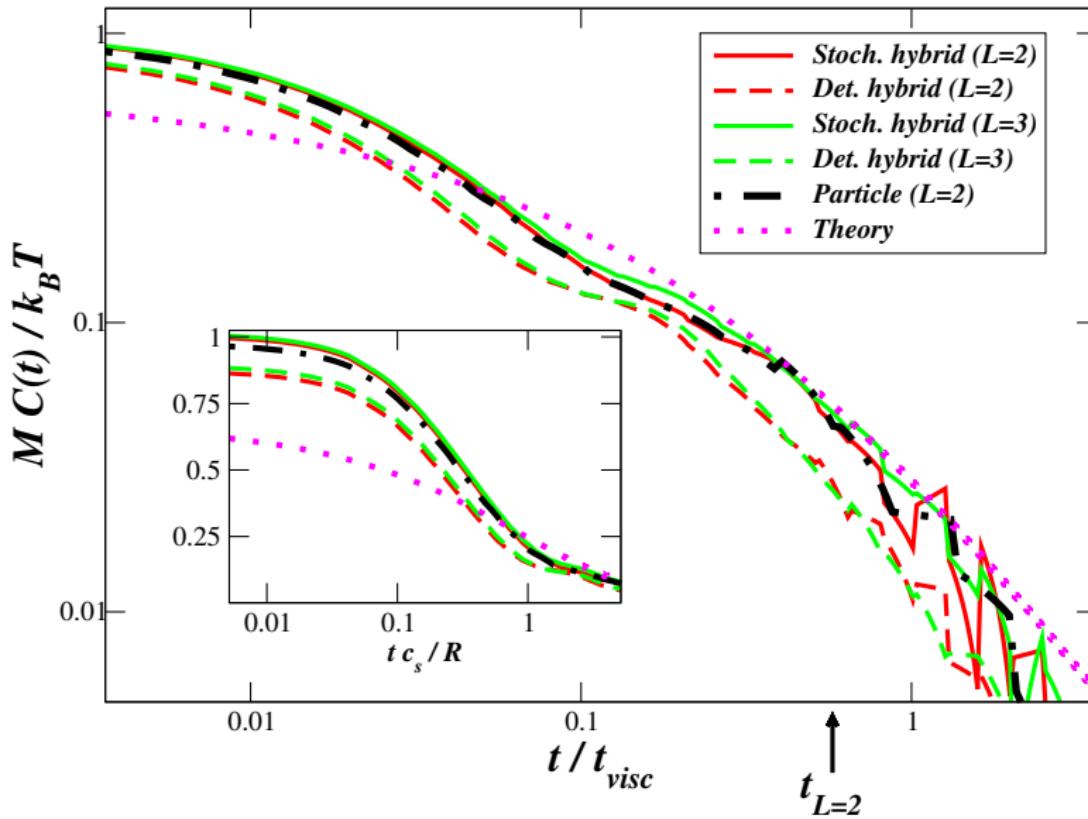
Velocity Autocorrelation Function

- We investigate the **velocity autocorrelation function** (VACF) for a Brownian bead

$$C(t) = 2d^{-1} \langle \mathbf{v}(t_0) \cdot \mathbf{v}(t_0 + t) \rangle$$

- From equipartition theorem $C(0) = k_B T/M$.
- For a **neutrally-boyant** particle, $\rho' = \rho$, incompressible hydrodynamic theory gives $C(0) = 2k_B T/3M$ because one third of the kinetic energy decays at the sound time scale.
- Hydrodynamic persistence (conservation) gives a **long-time power-law tail** $C(t) \sim (k_B T/M)(t/t_{visc})^{-3/2}$ that can be quantified using fluctuating hydrodynamics.
- The diffusion coefficient is the **integral of the VACF** and is strongly-affected by the tail.

VACF



The adiabatic piston problem

MNG

Relaxation Toward Equilibrium

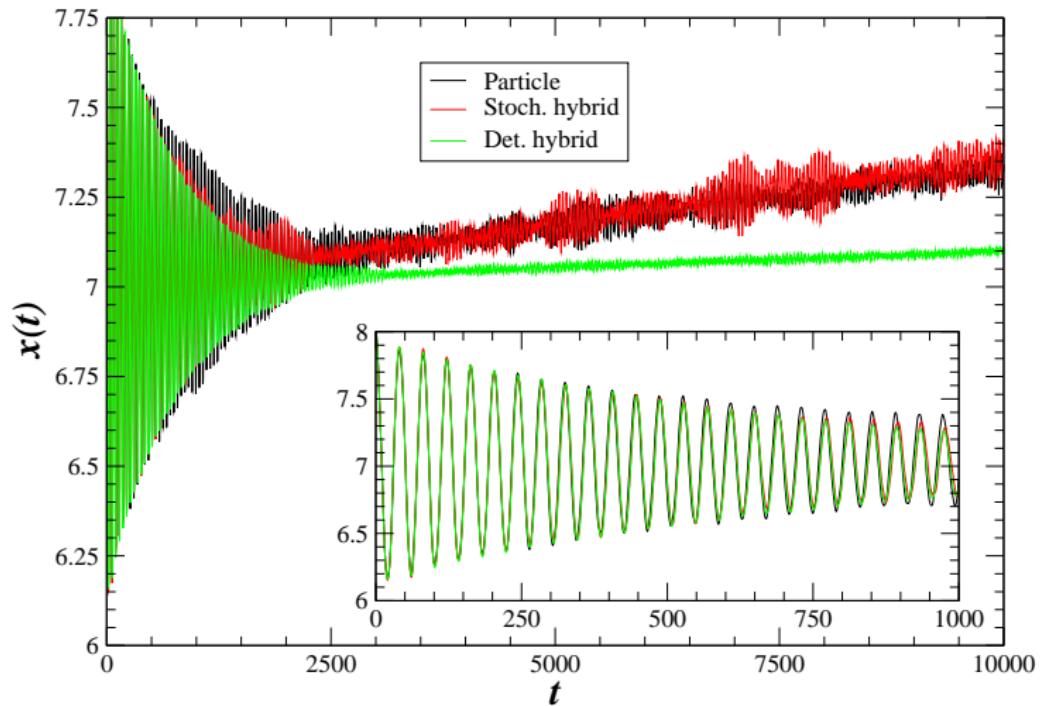


Figure: Massive rigid piston ($M/m = 4000$) not in mechanical equilibrium: **The deterministic hybrid gives the wrong answer!**

VACF for Piston

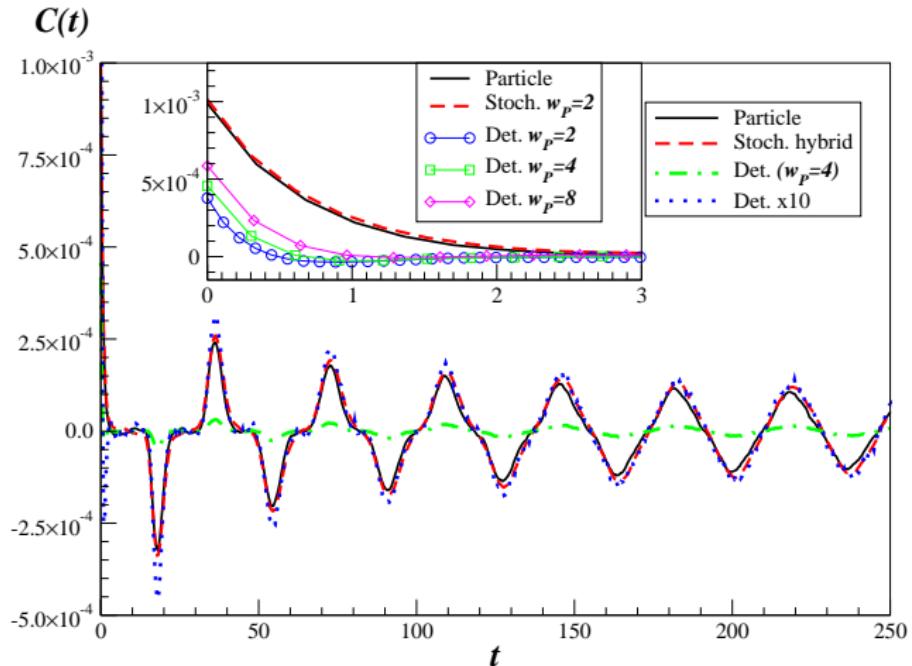


Figure: The VACF for a rigid piston of mass $M/m = 1000$ at thermal equilibrium: **Increasing the width of the particle region does not help: One must include the thermal fluctuations in the continuum solver!**

Nonequilibrium Fluctuations

- When macroscopic gradients are present, steady-state thermal fluctuations become **long-range correlated**.
- Consider a **binary mixture** of fluids and consider **concentration fluctuations** around a steady state $c_0(\mathbf{r})$:

$$c(\mathbf{r}, t) = c_0(\mathbf{r}) + \delta c(\mathbf{r}, t)$$

- The concentration fluctuations are **advected by the random velocities** $\mathbf{v}(\mathbf{r}, t) = \delta \mathbf{v}(\mathbf{r}, t)$, approximately:

$$\partial_t (\delta c) + (\delta \mathbf{v}) \cdot \nabla c_0 = \chi \nabla^2 (\delta c) + \sqrt{2\chi k_B T} (\nabla \cdot \mathcal{W}_c)$$

- The velocity fluctuations drive and amplify the concentration fluctuations leading to so-called **giant fluctuations** [8].

Fractal Fronts in Diffusive Mixing

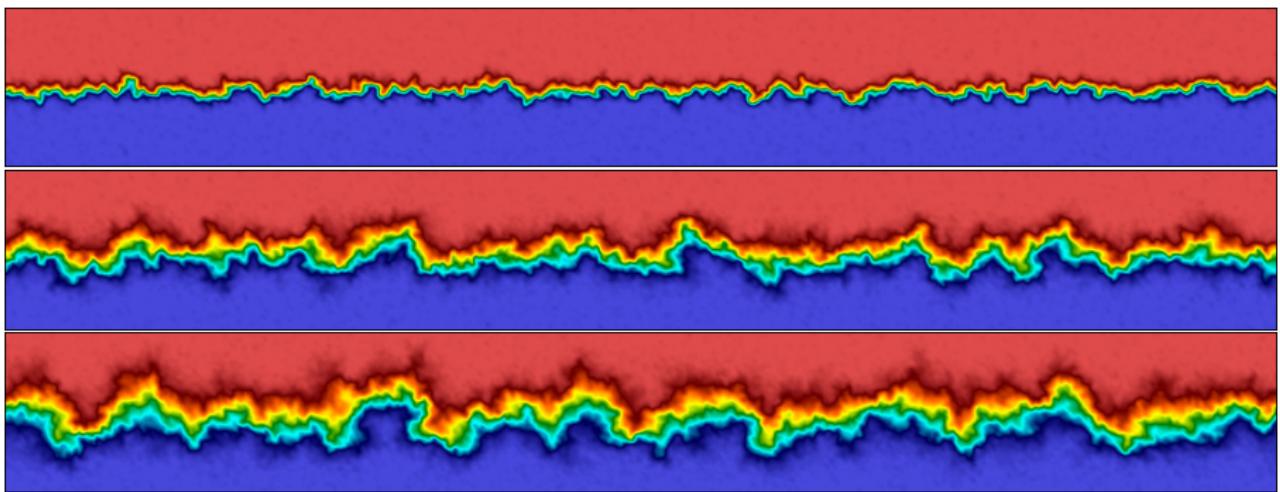


Figure: Snapshots of concentration in a miscible mixture showing the development of a *rough* diffusive interface between two miscible fluids in zero gravity [3, 8, 5].

Giant Fluctuations in Experiments

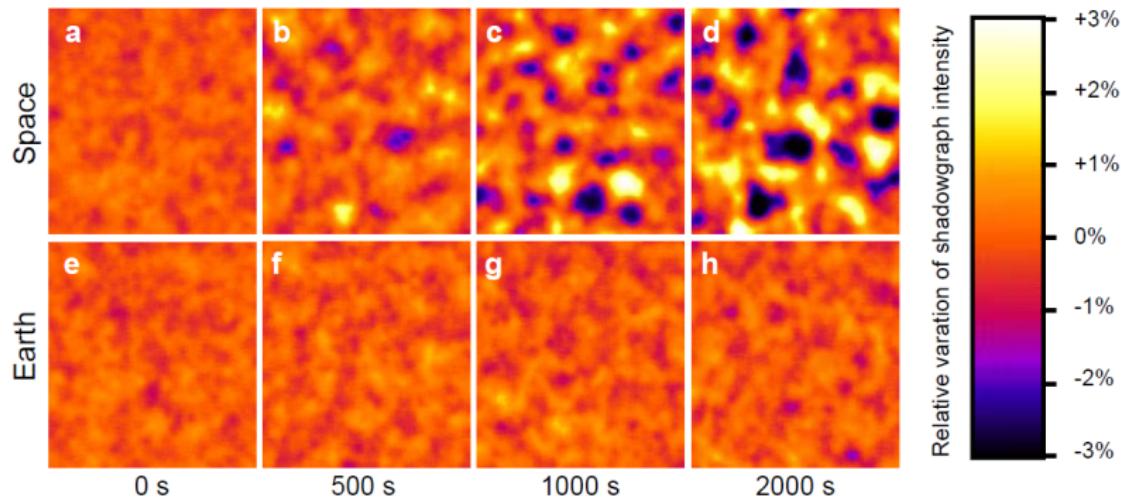


Figure: Experimental results by A. Vailati *et al.* from a microgravity environment [8] showing the enhancement of concentration fluctuations in space (box scale is **macroscopic**: 5mm on the side, 1mm thick).

Fluctuation-Enhanced Diffusion Coefficient

- The **nonlinear** concentration equation includes a contribution to the mass flux due to **advection by the fluctuating velocities**,

$$\partial_t (\delta c) + (\delta \mathbf{v}) \cdot \nabla c_0 = \nabla \cdot [-(\delta c) (\delta \mathbf{v}) + \chi \nabla (\delta c)] + \dots$$

- Simple (quasi-linear) perturbative theory suggests that concentration and velocity fluctuations become correlated and

$$-\langle (\delta c) (\delta \mathbf{v}) \rangle \approx (\Delta \chi) \nabla c_0.$$

- The **fluctuation-renormalized diffusion coefficient** is $\chi + \Delta \chi$ (think of **eddy diffusivity** in turbulent transport).
- Because fluctuations are affected by boundaries, $\Delta \chi$ is **system-size dependent**.

Fluctuation-Enhanced Diffusion Coefficient

- Consider the effective diffusion coefficient in a system of dimensions $L_x \times L_y \times L_z$ with a concentration gradient imposed along the y axis.
- In **two dimensions**, $L_z \ll L_x \ll L_y$, linearized fluctuating hydrodynamics predicts a **logarithmic divergence**

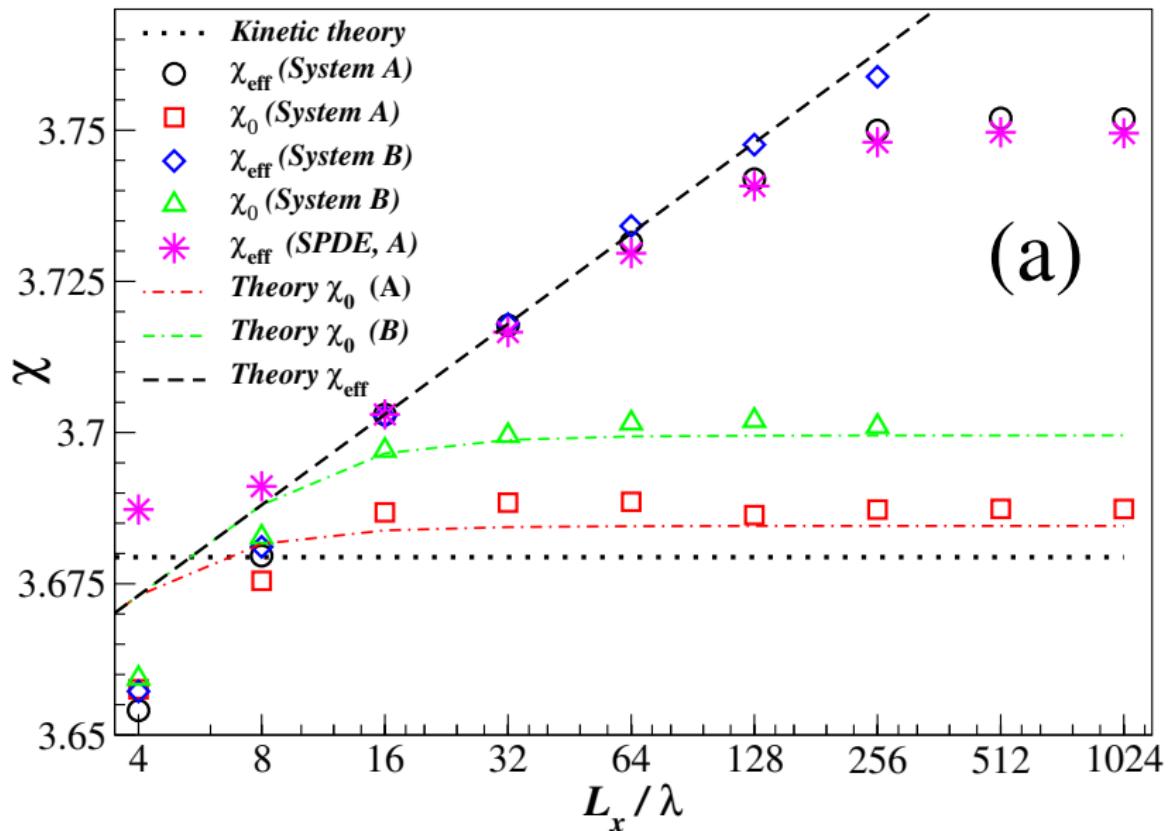
$$\chi_{\text{eff}}^{(2D)} \approx \chi + \frac{k_B T}{4\pi\rho(\chi + \nu)L_z} \ln \frac{L_x}{L_0}$$

- In **three dimensions**, $L_x = L_z = L \ll L_y$, χ_{eff} converges as $L \rightarrow \infty$ to the **macroscopic diffusion coefficient**,

$$\chi_{\text{eff}}^{(3D)} \approx \chi + \frac{\alpha k_B T}{\rho(\chi + \nu)} \left(\frac{1}{L_0} - \frac{1}{L} \right)$$

- We have verified these predictions using particle (DSMC) simulations at hydrodynamic scales [3].

Particle Simulations



Microscopic, Mesoscopic and Macroscopic Fluid Dynamics

- Instead of an ill-defined “molecular” or “bare” diffusivity, one should define a **locally renormalized diffusion coefficient** χ_0 that depends on the length-scale of observation.
- This coefficient accounts for the arbitrary division between continuum and particle levels inherent to fluctuating hydrodynamics.
- A deterministic continuum limit does not exist in two dimensions, and is not applicable to small-scale finite systems in three dimensions.
- **Fluctuating hydrodynamics** is applicable at a broad range of scales if the transport coefficient are renormalized based on the cutoff scale for the random forcing terms.

Conclusions

- **Coarse-grained particle methods** can be used to accelerate hydrodynamic calculations at small scales.
- **Hybrid particle continuum methods** closely reproduce purely particle simulations at a fraction of the cost.
- It is **necessary to include fluctuations** in the continuum solver in hybrid methods.
- **Thermal fluctuations affect the macroscopic transport in fluids.**

Future Directions

- Improve and implement stochastic **particle methods** (parallelize, add chemistry, analyze theoretically).
- **Direct fluid-structure coupling** between fluctuating hydrodynamics and microstructure.
- Develop numerical schemes for **Low-Mach Number** fluctuating hydrodynamics.
- Ultimately we require an **Adaptive Mesh and Algorithm Refinement** (AMAR) framework that couples a particle model (**micro**), with compressible fluctuating Navier-Stokes (**meso**), and incompressible or low Mach solver (**macro**).

References

-  A. Donev, A. L. Garcia, and B. J. Alder.
Stochastic Hard-Sphere Dynamics for Hydrodynamics of Non-Ideal Fluids.
Phys. Rev. Lett., 101:075902, 2008.
-  P. Español.
Stochastic differential equations for non-linear hydrodynamics.
Physica A, 248(1-2):77–96, 1998.
-  A. Donev, A. L. Garcia, Anton de la Fuente, and J. B. Bell.
Diffusive Transport Enhanced by Thermal Velocity Fluctuations.
Phys. Rev. Lett., 106(20):204501, 2011.
-  A. Donev, E. Vanden-Eijnden, A. L. Garcia, and J. B. Bell.
On the Accuracy of Explicit Finite-Volume Schemes for Fluctuating Hydrodynamics.
CAMCOS, 5(2):149–197, 2010.
-  F. Balboa, J. Bell, R. Delgado-Buscallioni, A. Donev, T. Fai, A. Garcia, B. Griffith, and C. Peskin.
Staggered Schemes for Incompressible Fluctuating Hydrodynamics.
Submitted, 2011.
-  G. De Fabritiis, M. Serrano, R. Delgado-Buscallioni, and P. V. Coveney.
Fluctuating hydrodynamic modeling of fluids at the nanoscale.
Phys. Rev. E, 75(2):026307, 2007.
-  W. Ren.
Analytical and numerical study of coupled atomistic-continuum methods for fluids.
J. Comp. Phys., 227(2):1353–1371, 2007.
-  A. Vailati, R. Cerbino, S. Mazzoni, C. J. Takacs, D. S. Cannell, and M. Giglio.
Fractal fronts of diffusion in microgravity.
Nature Communications, 2:290, 2011.