



Diffusion of Monodisperse, Hard-Sphere Colloidal Fluids: Continuous-Time Random-Walks and Beyond

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Background and Introduction

- **Need better prediction and control of**
 - Ion transport in composite electrodes (microstructure scale)
 - Durability of composite electrodes, etc.

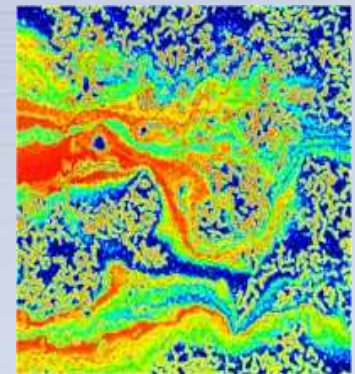
- **Heterogeneous materials**

- Inhomogeneous, “discontinuous”
 - material properties and microstructure
 - multi-phase, multi-material → interfaces
 - “dynamics” (i.e., kinematics of a particle or flux)
 - **Discrete particles** in polymer matrix or **suspending fluid**
 - Effective thermal (electrical, etc.) conductivity of composites

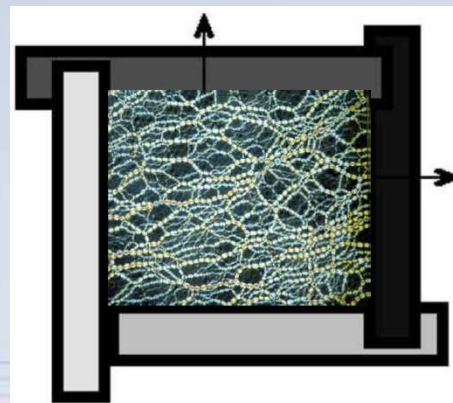
- **Generalized, nonequil. continuum Transport**
 - **Generalized Diffusion**

- “Anomalous”

Multi-fractal

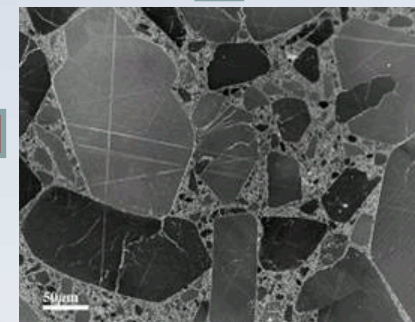


Static μ struct



dynamic

μ struct





Project Goals

- **Develop models that accurately capture events/processes on *broad* range of scales**
 - What are the component/process scale governing equations in complex, far-from-equilibrium systems?
 - Models must be amenable to math analysis
 - Constitutive relations for closure of model equations should be directly accessible experimentally
 - To be predictive
 - Need access to **FULL** range of length and time scales to close gaps between
 - Discrete constituent and component/process scale
 - Simulation/experimental run-time and product/component life span and performance
 - Need workflow path for parameterizing constitutive relations across relevant scales
- **Novel models push the boundary of V&V to become Discovery tools**
 - Advance experimental and math analysis capabilities to elucidate underlying physical phenomena and path to innovation



Brownian Motion and Diffusion Equation: “Micro-Macro” Connection

- Translation of *isolated* small sphere in Newtonian fluid

- Time dependent Stokes drag force on sphere

- Nonlocal: space $\rightarrow$ time

$$\int_{\Omega} \nabla \cdot \boldsymbol{\sigma}_H(t) dV = \int_{\partial\Omega} \boldsymbol{\sigma}_H(t) \cdot \hat{\mathbf{r}} dS \rightarrow \int_{-\infty}^t \boldsymbol{\Gamma}(t-t') \cdot \mathbf{v}_i(t') dt'; \quad \boldsymbol{\Gamma}(t-t') = \gamma(t-t') \mathbf{I}$$

$$m^* \frac{d\mathbf{v}(t)}{dt} = - \int_{-\infty}^t \gamma(t-t') \mathbf{v}(t') dt' + \mathbf{f}(t)$$

- Assume: Gaussian random force, long-time limit (i.e., ignore fluid and particle inertia)

$$\left. \begin{aligned} \gamma(t-t') &= \gamma \delta(t-t') \\ \mathbf{f}(t) &= \mathbf{F}^B(t) \end{aligned} \right\} \begin{aligned} \frac{d\mathbf{r}_i(t)}{dt} &= \frac{1}{\gamma} \mathbf{F}_i^B(t) \\ \langle \mathbf{F}_i^B(t) \rangle &= 0; \langle \mathbf{F}_i^B(t) \mathbf{F}_i^B(t') \rangle = 2k_B T \gamma \delta(t-t') \end{aligned}$$



- Interactionless, Markovian, SDE $\rightarrow$ PDE

$$\frac{\partial \rho(x, t)}{\partial t} = D \nabla^2 \rho(x, t)$$



“Whatever it is I think I see...”

- **Diffusion underlies all irreversible, non-equilibrium processes**
 - Linear, phenomenological Onsager coefficients
 - Fick’s (Second) Law (mass transport)
 - Fourier’s Law (thermal conduction)
 - Newton’s Law (momentum transport)
 - Applicable in the long-time limit (beyond correlation length/time scale)
- **Many systems present multiple, competing length and time scales**
 - Which ones are relevant to the microscopic dynamical processes that underlie macro-transport – what is the “ruler” (i.e., measure) that allows for actual characterization of the system/process
 - How to handle “meso-scale” regimes where correlations are still significant?



Revisiting the Classical Langevin Equation (OU Process)

- “Microscopic” dynamics

$$\frac{d\mathbf{r}_i}{dt} = \mathbf{v}_i$$

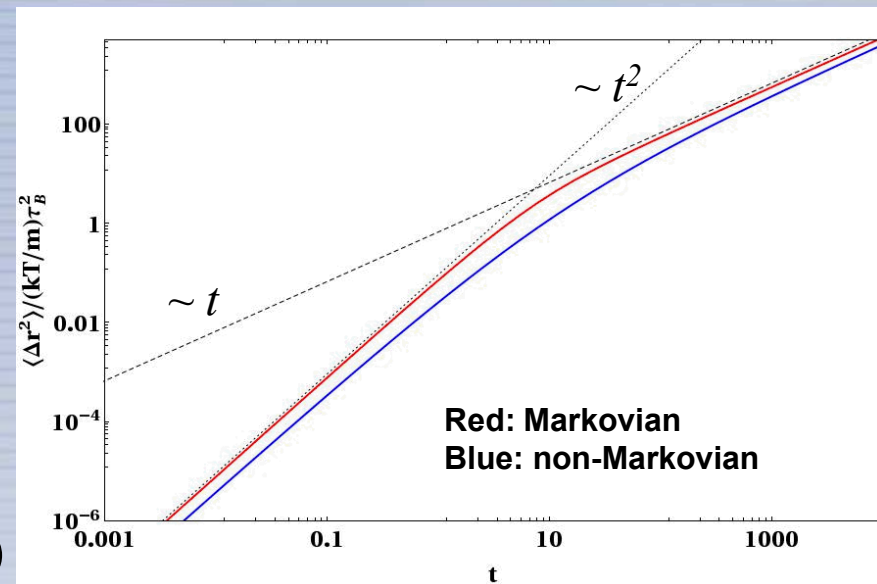
$$\frac{d\mathbf{v}_i(t)}{dt} = -\frac{1}{\tau_B} \mathbf{v}_i(t) + \frac{1}{m} \mathbf{F}_i^B(t)$$

$$\langle \mathbf{F}_i^B(t) \rangle = 0; \langle \mathbf{F}_i^B(t) \mathbf{F}_i^B(t') \rangle = 2k_B T \gamma \delta(t - t')$$

– Solve (formally) and obtain

$$\langle |\Delta \mathbf{r}|^2 \rangle(t) = \frac{6k_B T}{m} \tau_B \left(t - \tau_B (1 - e^{-t/\tau_B}) \right)$$

- Defines/justifies long-time, $t \gg \tau_B$
- $MSD = f(t) \neq Dt^\alpha$; $0 \leq \alpha \leq 2$
 - Temptation: $D = D(t) \rightarrow MSD = D(t)t$
 - » What does this mean **physically**?





Scale-Invariant Diffusion Equation

- **Microscopic dynamics give**

- $MSD = f(t)$

- $MSD = Dt^{\alpha(t)}$; $\alpha(t) = 2H(t)$

- $MSD = D\tau_D(t)$; $\frac{d \ln(\tau_D(t))}{d \ln(t)} = 2H(t)$

$$\langle |\Delta \mathbf{r}|^2 \rangle(t) = \frac{6kT}{m} \tau_B \left(t - \tau_B (1 - e^{-t/\tau_B}) \right) = 6 \left(\frac{kT}{m} \tau_B \right) \tau_D(t)$$

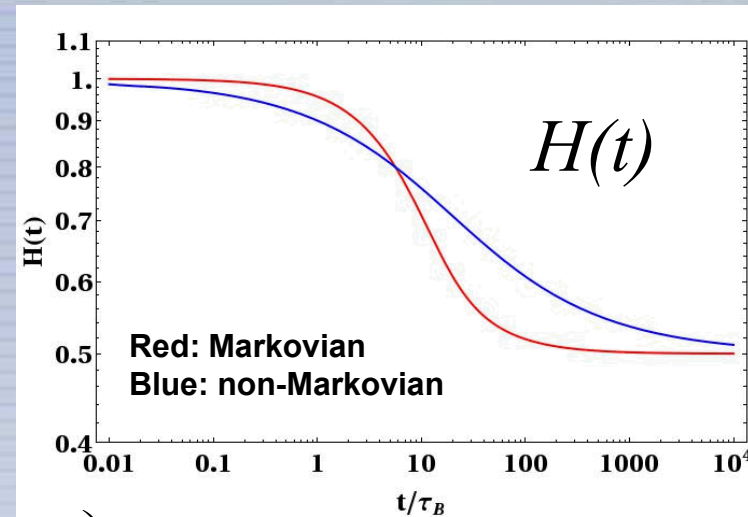
- **A density which gives above as the second moment is**

- Gaussian assumption → distribution is stationary and symmetric; no higher moments matter

$$\rho(\mathbf{x}, \tau_D(t)) = \frac{1}{\sqrt{2\pi D \tau_D(t)}} \exp \left[-\frac{\mathbf{x}^2}{4D \tau_D(t)} \right]$$

- Which is a solution to

$$\frac{\partial \rho(\mathbf{x}, \tau_D)}{\partial \tau_D} = D \nabla^2 \rho(\mathbf{x}, \tau_D), \text{ given } \rho(\mathbf{x}, 0) \sim \delta(\mathbf{x})$$





Generalized Diffusion-Relaxation Equation

- **Recall** $\frac{\partial \rho(\mathbf{x}, \tau_D(t))}{\partial \tau_D} = D \nabla^2 \rho(\mathbf{x}, \tau_D(t))$
 - Note $\partial \tau_D = \frac{d\tau_D(t)}{dt} \partial t$
- **So,** $\frac{\partial \rho(\mathbf{x}, \tau_D(t))}{\partial t} = \frac{d\tau_D(t)}{dt} D \nabla^2 \rho(\mathbf{x}, \tau_D(t))$
 - Here $\frac{\partial \tau_D(t)}{\partial t} = (1 - e^{-t/\tau_B})$, so $\frac{\partial \rho(\mathbf{x}, t)}{\partial t} = D \nabla^2 \rho(\mathbf{x}, t) - e^{-t/\tau_B} D \nabla^2 \rho(\mathbf{x}, t)$
 - Integrating gives,

$$\rho(\mathbf{x}, \tau_D(t)) = \rho(\mathbf{x}, \tau_D(0)) + D \int_0^t \frac{d\tau_D(t-t')}{dt} \nabla^2 \rho(\mathbf{x}, \tau_D(t')) dt'$$



“Fractional” Diffusion?

- Start with “diffusion-relaxation” equation

$$\rho(\mathbf{x}, \tau_D(t)) = \rho(\mathbf{x}, \tau_D(0)) + D \int_0^t (1 - e^{-(t-t')/\tau_B}) \nabla^2 \rho(\mathbf{x}, \tau_D(t')) dt'$$

- Take **time** derivative

$$\frac{\partial \rho(\mathbf{x}, \tau_D(t))}{\partial t} = \frac{\partial \rho(\mathbf{x}, \tau_D(0))}{\partial t} - D \frac{\partial \tau_D(0)}{\partial t} \nabla^2 \rho(\mathbf{x}, t) + D \int_0^t \frac{\partial^2 \tau_D(t-t')}{\partial t^2} \nabla^2 \rho(\mathbf{x}, \tau_D(t')) dt'$$

- set $\frac{\partial \rho(\mathbf{x}, \tau_D(0))}{\partial t} = 0$ to get a “diffusion equation with memory”

$$\frac{\partial \rho(\mathbf{x}, \tau_D(t))}{\partial t} = \frac{kT}{m} \int_0^t e^{-(t-t')/\tau_B} \nabla^2 \rho(\mathbf{x}, \tau_D(t')) dt'$$

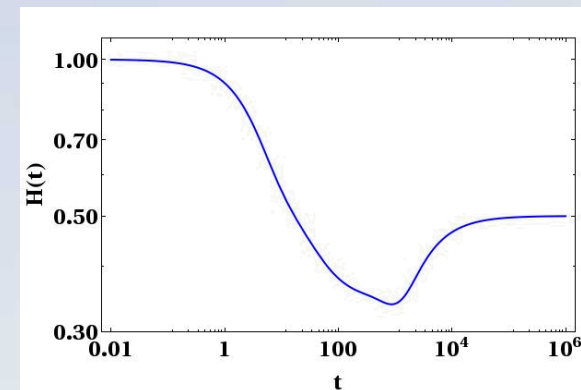
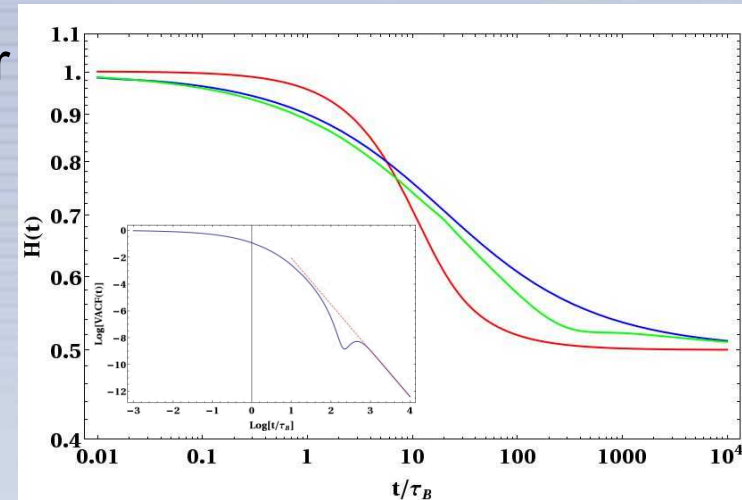
- Which is equivalent to

$$\tau_B \frac{\partial^2 \rho(\mathbf{x}, \tau_D(t))}{\partial t^2} + \frac{\partial \rho(\mathbf{x}, \tau_D(t))}{\partial t} = D \nabla^2 \rho(\mathbf{x}, \tau_D(t))$$



Various Examples of $\tau_D(t)$

- Have analytical forms for dilute particles → non-interacting random walkers
 - Classical Damped Harmonic Oscillator
 - Under, over, critically damped
 - **Classical GLE**
 - Incompressible Newtonian Liquid
 - zero Re
 - Compressible Newtonian Fluid
 - Zero Re
 - High Re, Zero and moderate Ma
 - Incompressible/compressible Non-Newtonian
 - Zero Re
 - **Nearly Classical Ideal Liquid**
- Can obtain from simulation
 - Interacting particles (non-zero concentration)





The “Generalized Langevin Equation”

- **Transient Brownian Dynamics (Interactionless)**
 - **NonMarkovian Langevin Equation**

$$m^* \frac{d\mathbf{v}_i(t)}{dt} = -6\pi\mu a \int_0^t \underbrace{\left(\frac{(t-t')^{-3/2}}{2\Gamma(1/2)} + \delta(t-t') \right)}_{\xi(t-t')} \mathbf{v}_i(t') dt' + \mathbf{F}_i^R(t)$$

$$\langle \mathbf{F}_i^R(t) \rangle = 0$$

$$\xi(t-t')$$

$$\langle \mathbf{F}_i^R(t) \mathbf{F}_i^R(t') \rangle = 2k_B T \xi(t-t')$$

- **If noise is modeled by Gaussian, stationary, self-similar processes, we have fBm**
- **Note Basset-Boussinesq**

$$m^* \frac{d\mathbf{v}_i(t)}{dt} = 6\pi\mu a \int_0^t \left(\frac{(t-t')^{-1/2}}{\sqrt{\pi}} \right) \frac{d\mathbf{v}_i(t')}{dt'} dt' - 6\pi\mu a \mathbf{v}(t)$$

$$m^* = \left(m_p + \frac{1}{2} m_f \right)$$



Non-Markovian Example

- **Note, impulsive response of sphere in incompressible, Newtonian fluid**
 - sphere initially at rest, fluid at rest at infinity – i.e., dilute suspension

$$\mathbf{v}(t) = \frac{1}{2(q_1 - q_2)} \left[q_1 E_{1/2,1}(-t / \tau_1) - q_2 E_{1/2,1}(-t / \tau_2) \right]$$

$$q_{1,2} = \frac{1}{2Za} \left(1 \pm \sqrt{1 - 4Z} \right), \quad Z = \frac{2m_p + m_f}{9m_f}, \quad \tau_{1,2} = (q_{1,2})^2 \nu$$

$$E_{\alpha,\beta}(y) = \sum_{k=0}^{\infty} \frac{y^k}{\Gamma(\alpha k + \beta)}$$

- **We can use this to derive MSD and**

$$\tau_D(t) = \frac{\tau_B}{2(q_1 - q_2)} \left\{ q_1 E_{1/2,3}[-(t / \tau_1)^2] - q_2 E_{1/2,3}[-(t / \tau_2)^2] \right\}$$

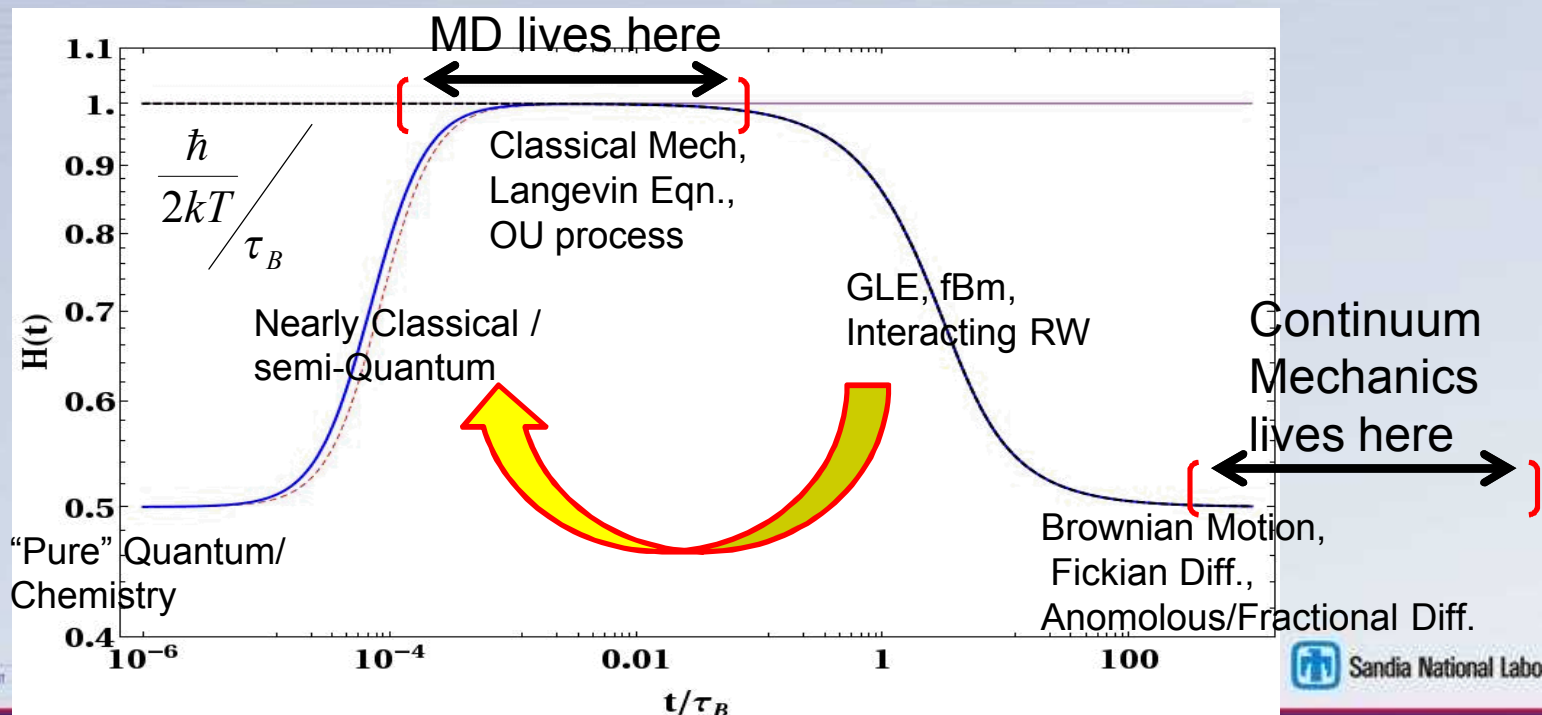


The Semi-Classical Limit and Defining Mesoscales

- **Nearly classical fluid (QM – Langevin equation)**
 - following De Bar 1963 (see also Vineyard 1958, van Hove 1954)

$$\tau_D^{Cl}(t) = \tau_B \left[\frac{t}{\tau_B} - (1 - e^{-t/\tau_B}) \right] \quad \tau_D(t) = \tau_D^{Cl}(t) - i \frac{\hbar}{2kT} \left(\frac{d\tau_D^{Cl}(t)}{dt} \right) = t - \tau_B (1 - e^{-t/\tau_B}) - i \frac{\hbar}{2kT} (1 - e^{-t/\tau_B})$$

- Consider the “average trajectory” of a particle, $\rho_i(x,t)$





Space/Time structure of Probability Densities for Particle Trajectories

- **Single particle density (kinematics) of hard-sphere colloid**

$$\rho_i(\mathbf{r}, t) = \left\langle m \delta(\mathbf{r} - \mathbf{R}_i(t)); f^N(\mathbf{R}^N, \mathbf{P}^N; t) \right\rangle = \lim_{t \rightarrow \infty} \frac{1}{t} \int_{t_0}^t m \delta(\mathbf{r} - \mathbf{R}_i(t')) dt' \approx \frac{1}{M_t} \sum_{k=0}^{M_t} m \delta(\mathbf{r} - \mathbf{R}_i(t_0 + k\Delta t))$$

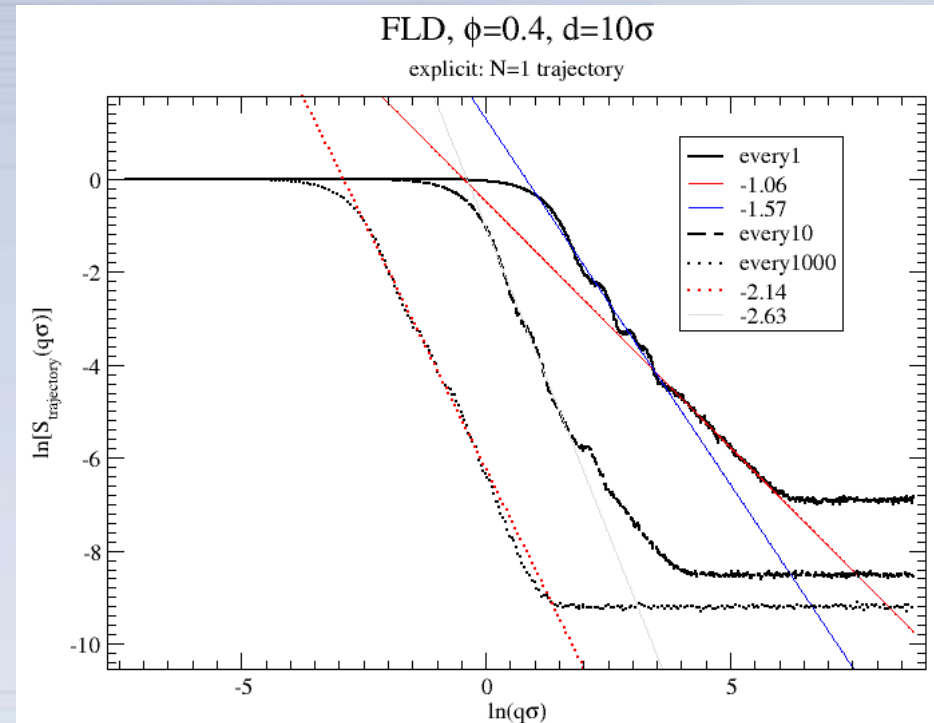
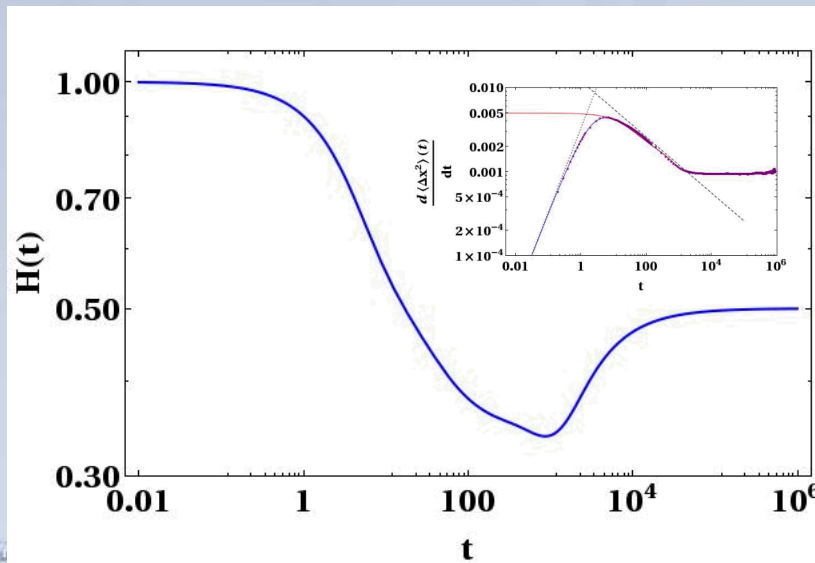
- “Light scattering” of single particle trajectory, $\mathbf{R}_i(t_k)$

- Self-affine record $\rightarrow$ self-similar fractal “trail”

- Multiple length and time scales

- Multifractal: $d_{f,D}(t) \sim 1/H(t)$

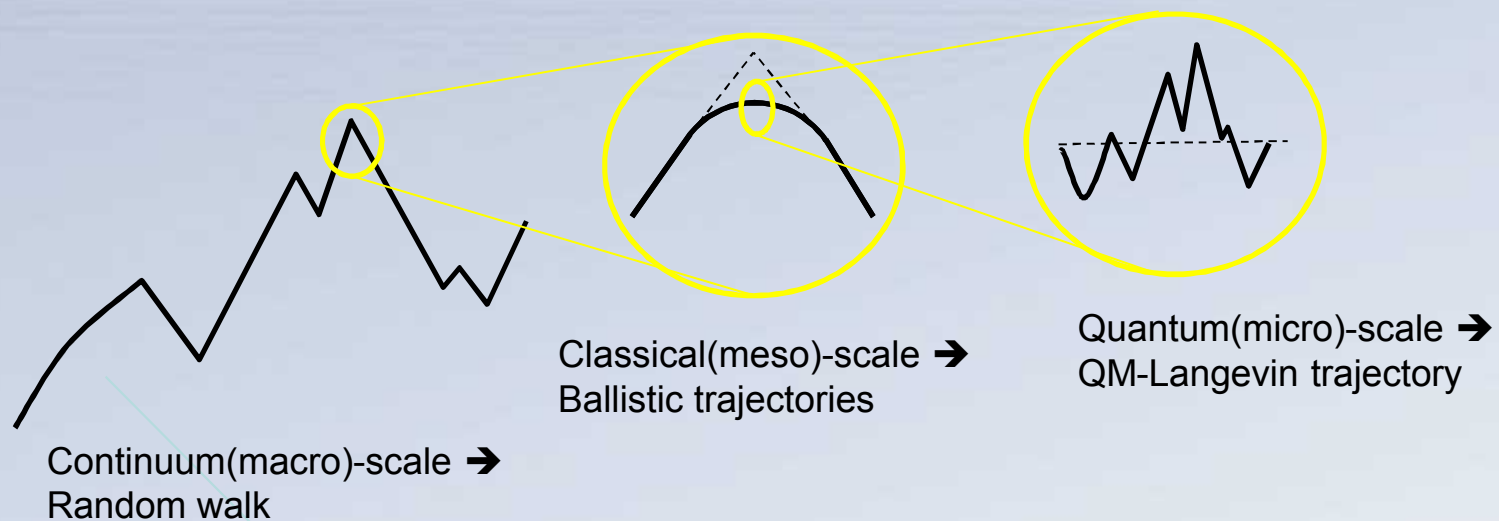
- Change “ruler”: $d_{f,D}(\tau_D) = 2$





Schematic of Scale Dependent Processes

- **Complex heterogeneity implies Multi-fractal functions**
 - Generalized Trajectories (mass, momentum, energy), $r(t)$
 - Non-differentiable
 - Hierarchy of scaling exponents
 - Fundamentally, no derivatives (e.g., particle velocity) can be defined in the classical delta-epsilon, limit sense





Scale Consistent “Diffusion” (at Very Small Scale)

- For “Quantum-Mechanical” Markovian Langevin Eqn. (OU process)

$$\tau_D^{Cl}(t) = \tau_B \left[\frac{t}{\tau_B} - (1 - e^{-t/\tau_B}) \right] \quad \tau_D(t) = \tau_D^{Cl}(t) - i \frac{\hbar}{2kT} \left(\frac{d\tau_D^{Cl}(t)}{dt} \right) = t - \tau_B (1 - e^{-t/\tau_B}) - i \frac{\hbar}{2kT} (1 - e^{-t/\tau_B})$$

$$\frac{\partial \rho(\mathbf{x}, \tau_D(t))}{\partial \tau_D} = D \nabla^2 \rho(\mathbf{x}, \tau_D(t))$$

$$\frac{\partial \rho(\mathbf{x}, \tau_D(t))}{\partial t} = \frac{d\tau_D(t)}{dt} \left(\frac{kT}{m} \tau_B \right) \nabla^2 \rho(\mathbf{x}, \tau_D(t))$$

$$\frac{\partial \rho(\mathbf{x}, \tau_D(t))}{\partial t} = \left(\frac{kT}{m} \tau_B \right) (1 - e^{-t/\tau_B}) \nabla^2 \rho(\mathbf{x}, \tau_D(t)) - i \frac{\hbar}{2m} e^{-t/\tau_B} \nabla^2 \rho(\mathbf{x}, \tau_D(t))$$

- Take limit of $t \rightarrow 0$; keeping leading order term

$$\frac{\partial \rho(\mathbf{x}, \tau_D(t))}{\partial t} \approx -i \frac{\hbar}{2m} \nabla^2 \rho(\mathbf{x}, \tau_D(t)) \quad \lim_{t \rightarrow 0} \tau_D(t) = \frac{t^2}{2\tau_B} - i \frac{\hbar}{2kT} \left(\frac{t}{\tau_B} \right)$$



What is the “density” at small t?

- Recall

$$\frac{\partial \rho(\mathbf{x}, \tau_D(t))}{\partial \tau_D} = \frac{kT}{m} \tau_B \nabla^2 \rho(\mathbf{x}, \tau_D(t))$$

$$\rho(\mathbf{x}, \tau_D(t)) = \frac{1}{\sqrt{2\pi D \tau_D(t)}} \exp\left[-\frac{\mathbf{x}^2}{4D \tau_D(t)}\right]$$

$$\lim_{t \rightarrow 0} \tau_D(t) \approx \frac{t^2}{2\tau_B} - i \frac{\hbar}{2kT} \left(\frac{t}{\tau_B}\right) \approx -i \frac{\hbar}{2m} \left(\frac{t}{D}\right)$$

$$\lim_{t \rightarrow 0} \rho(\mathbf{x}, \tau_D(t)) \approx \frac{1}{\sqrt{-\pi i \frac{\hbar}{m} t}} \exp\left[-\frac{\mathbf{x}^2}{-2i \frac{\hbar}{m} t}\right]$$

$$\lim_{t \rightarrow 0} \rho(\mathbf{x}, \tau_D(t)) = \psi(\mathbf{x}, -t) = \psi^*(\mathbf{x}, t) \rightarrow \text{Gaussian wave function}$$

$$\frac{\partial \psi^*(\mathbf{x}, t)}{\partial t} = -i \frac{\hbar}{2m} \nabla^2 \psi^*(\mathbf{x}, t)$$



Routes to the scale-consistent or scale-invariant diffusion equation



Diffusion Equation for Wiener process (i.e., overdamped Langevin dynamics)

- **Consider phase-space density** $f^N(\mathbf{r}^N; t)$
 - Velocity not defined for these dynamics

$$d\mathbf{r}_i = \frac{1}{\gamma} \mathbf{F}_i^B dt$$

$$\langle \mathbf{F}_i^B(t) \rangle = 0; \langle \mathbf{F}_i^B(t) \mathbf{F}_j^B(t') \rangle = 2k_B T \gamma \delta(t - t') \delta_{ij}$$

- Conservation of phase space volume implies Ito's Formula

$$\frac{df^N}{dt} = \left[\frac{\partial f^N}{\partial t} + \sum_{i=1}^N \frac{kT}{\gamma} \frac{\partial^2}{\partial \mathbf{r}_i^2} f^N \right] = 0$$

- Now, $\rho_k(\mathbf{x}, t) = \langle \delta(\mathbf{x} - \mathbf{r}_k(t)); f^N(\mathbf{r}^N; t) \rangle$

- and

$$\frac{\partial \rho_k(\mathbf{x}, t)}{\partial t} = \frac{kT}{\gamma} \left(\frac{\partial^2 \rho_k}{\partial \mathbf{x}^2} \right)$$



Kramers-Klein Equation for OU process (i.e., Langevin dynamics)

- Consider Liouville phase-space density $f^N(\mathbf{r}^N, \mathbf{p}^N; \tau_d(\Delta t))$

– Conservation of phase space volume implies

$$\frac{d}{d\tau_D} f^N(\mathbf{r}^N, \mathbf{p}^N; \tau_d(t)) = \frac{df^N}{dt} \frac{dt}{d\tau_D} = 0$$

From Ito's Lemma

$$\frac{df^N}{dt} = \frac{d\tau_D}{dt} \left[\frac{\partial f^N}{\partial t} \frac{\partial t}{\partial \tau_D} + \sum_{i=1}^N \frac{\partial}{\partial \chi_i} \cdot \left(f^N \frac{\partial \chi_i}{\partial t} \frac{\partial t}{\partial \tau_D} \right) + \sum_{i=1}^N \gamma kT \frac{\partial^2}{\partial \mathbf{p}_i^2} f^N \right] = 0$$

- Where, $\chi_i(t) = \begin{Bmatrix} \mathbf{r}_i(t) \\ \mathbf{p}_i(t) \end{Bmatrix}$

- and $\rho_i(\mathbf{x}, \mathbf{v}, \tau_D(t)) = \left\langle m \delta(\mathbf{x} - \mathbf{r}_i(t)) \delta(\mathbf{v} - \frac{\mathbf{p}_i(t)}{m}); f^N(\mathbf{r}^N, \mathbf{p}^N; \tau_D(t)) \right\rangle$

$$\frac{\partial \rho_i(\mathbf{x}, \mathbf{v}, \tau_D(t))}{\partial t} = - \frac{\partial}{\partial \mathbf{x}} \cdot (\rho_i \mathbf{v}) + \frac{\partial}{\partial \mathbf{v}} \cdot \left(\frac{\gamma}{m} \rho_i \mathbf{v} \right) + \frac{\gamma kT}{m^2} \frac{d\tau_D}{dt} \left(\frac{\partial^2 \rho_i}{\partial \mathbf{v}^2} \right)$$



Non-dimensionalized Kramer's Equation

- Choose the following non-dimensionalization

$$\hat{\mathbf{x}} = \mathbf{x} / \lambda ; \hat{\mathbf{v}} = \mathbf{v} / \sqrt{kT/m} ; \tau = t / \tau_B$$

– Where $\lambda = \tau_B \sqrt{kT/m}$

– Now, write

$$\frac{\partial \rho_i(\hat{\mathbf{x}}, \hat{\mathbf{v}}, \tau_D(\tau))}{\partial \tau} = - \frac{\partial}{\partial \hat{\mathbf{x}}} \cdot (\rho_i \hat{\mathbf{v}}) + \frac{\partial}{\partial \hat{\mathbf{v}}} \cdot (\rho_i \hat{\mathbf{v}}) + \frac{d\tau_D(\tau)}{d\tau} \left(\frac{\partial^2 \rho_i}{\partial \hat{\mathbf{v}}^2} \right)$$



Now reduce

- **Note, from microscopic equations of motion**

$$\frac{\partial \langle \hat{\mathbf{r}}_i(t) \rangle_t}{\partial t} = - \frac{\partial \langle \hat{\mathbf{p}}_i(t) / m \rangle_t}{\partial t} \Rightarrow \partial \hat{\mathbf{x}} = -\partial \hat{\mathbf{v}}$$

$$\frac{\langle \hat{\mathbf{r}}_i(t) \rangle_t - \hat{\mathbf{r}}_i(0)}{\langle \hat{\mathbf{p}}_i(t) / m \rangle_t - \hat{\mathbf{p}}_i(0) / m} = -1 = \tan \theta; \theta = -\frac{\pi}{4}$$

$$\frac{\partial \rho_i(\hat{\mathbf{x}}, \hat{\mathbf{v}}, \tau_D(\tau))}{\partial \tau} = -2 \frac{\partial}{\partial \hat{\mathbf{x}}} \cdot (\rho_i \hat{\mathbf{v}}) + \frac{d\tau_D(\tau)}{d\tau} \left(\frac{\partial^2 \rho_i}{\partial \hat{\mathbf{x}}^2} \right)$$

- **Now, Integrate over $\mathbf{v}$ ($\langle \mathbf{v} \rangle = 0$ at equilibrium) to get**

$$\frac{\partial \rho_i(\hat{\mathbf{x}}, \tau_D(\tau))}{\partial \tau} = \frac{d\tau_D(\tau)}{d\tau} \left(\frac{\partial^2 \rho_i}{\partial \hat{\mathbf{x}}^2} \right)$$



Path Integral – Preliminaries

- Choose the non-dimensionalization

$$\hat{\mathbf{x}} = \mathbf{x} / \lambda; \quad \hat{\mathbf{v}} = \mathbf{v} / \sqrt{kT/m}; \quad \tau = t / \tau_B$$

– Where $\lambda = \tau_B \sqrt{kT/m}$

- Consider Transition Probability for Markov Process

$$\rho(\hat{\mathbf{q}}) = \frac{1}{(\sqrt{2\pi})^3} \exp\left[-\frac{\hat{\mathbf{q}}^2}{2}\right] \quad \hat{\mathbf{q}} = \frac{\hat{\mathbf{x}}}{\sqrt{\hat{\tau}_D}}; \quad \hat{\tau}_D = \tau_D / \tau_B$$

- Gives recursion relation

$$P_n(\hat{\mathbf{q}}) = \frac{1}{(\sqrt{2\pi})^{3n}} \int d^3 \hat{\mathbf{q}}_1 d^3 \hat{\mathbf{q}}_2 \dots d^3 \hat{\mathbf{q}}_{n-1} e^{-S(\hat{\mathbf{q}}_0, \hat{\mathbf{q}}_2, \dots, \hat{\mathbf{q}}_n)}$$

$$S(\hat{\mathbf{q}}_0, \hat{\mathbf{q}}_2, \dots, \hat{\mathbf{q}}_n) = \frac{1}{2} \sum_{l=1}^n (\hat{\mathbf{q}}_l - \hat{\mathbf{q}}_{l-1})^2$$

$$\hat{\mathbf{q}}_0 = 0; \quad \hat{\mathbf{q}}_n = \hat{\mathbf{q}}$$



Path Integral (1) a la Standard Approach

- Introduce macroscopic time variables

$$\varepsilon = \frac{\hat{\tau}_D}{n} \quad \hat{\tau}_l = l\varepsilon, \quad \hat{\tau}_n = n\varepsilon = \hat{\tau}_D$$

- Define piecewise continuous path

$$\hat{\mathbf{x}}(\hat{\tau}'_D) = \sqrt{\varepsilon} \left[\hat{\mathbf{q}}_{l-1} + \frac{\hat{\tau}'_D - \hat{\tau}_{l-1}}{\hat{\tau}_l - \hat{\tau}_{l-1}} (\hat{\mathbf{q}}_l - \hat{\mathbf{q}}_{l-1}) \right] \quad \hat{\tau}_{l-1} < \hat{\tau}'_D < \hat{\tau}_l$$

– Note scaling/conversion pre-factor on RHS?

- **Now** $\frac{d\hat{\mathbf{x}}}{d\hat{\tau}'_D} = \frac{1}{\sqrt{\varepsilon}} (\hat{\mathbf{q}}_l - \hat{\mathbf{q}}_{l-1}) = \frac{\hat{\mathbf{x}}_l - \hat{\mathbf{x}}_{l-1}}{\varepsilon} \quad \hat{\mathbf{x}}(0) = 0; \quad \hat{\mathbf{x}}(\hat{\tau}_D) = \sqrt{\varepsilon} \hat{\mathbf{q}} = \hat{\mathbf{x}}$

$$S(\hat{\mathbf{q}}_0, \hat{\mathbf{q}}_2, \dots, \hat{\mathbf{q}}_n) \approx S(\hat{\mathbf{x}}(\hat{\tau}_D)) = \frac{1}{2} \int_0^{\hat{\tau}_D} \left(\frac{d\hat{\mathbf{x}}(\hat{\tau}'_D)}{d\hat{\tau}'_D} \right)^2 d\hat{\tau}'_D$$

– Which leads to

$$P_n(\hat{\mathbf{q}}) = \frac{1}{(\sqrt{2\pi})^3} \int \prod_{l=1}^{n-1} \frac{d^3 \hat{\mathbf{x}}(\hat{\tau}_l)}{(\sqrt{2\pi} \sqrt{\varepsilon})^3} \exp[-S(\hat{\mathbf{x}})]$$



Continuum Limit (1)

- **Now let** $n \rightarrow \infty \Rightarrow \varepsilon \rightarrow 0$ **for fixed** t

$$P_n(\hat{\mathbf{q}}) \rightarrow \Pi(\hat{\mathbf{x}}, \hat{\tau}_D) = \int [d\hat{\mathbf{x}}(\hat{\tau}_D)] \exp[-S(\hat{\mathbf{x}}(\hat{\tau}_D))]$$

$$\Pi(\mathbf{x}, \tau_D) = N(\tau_D) e^{-\mathbf{x}^2 / 4 \frac{kT}{m} \tau_B \tau_D} = \rho(\mathbf{x}, \tau_D)$$

- Notice this limit is perfectly justified even for finite correlation time
- Also, recall

$$\rho_i(\mathbf{r}, t) = \langle m \delta(\mathbf{r} - \mathbf{R}_i(t)); f^N(\mathbf{R}^N, \mathbf{P}^N; t) \rangle = \frac{1}{t} \int_0^t m \delta(\mathbf{r} - \mathbf{R}_i(t')) dt' \approx \frac{1}{M_t} \sum_{k=0}^{M_t} m \delta(\mathbf{r} - \mathbf{R}_i(t_k))$$

$$\rightarrow \Pi(\mathbf{r}, t) = \int [d\mathbf{r}(t)] \exp[-S(\mathbf{r}(t))]$$

$$\Rightarrow \frac{1}{2kT\tau_B} \int_0^t S(t') dt' = \frac{m}{2kT\tau_B} \int_0^t \left(\frac{d\mathbf{R}_i(t')}{dt'} \right)^2 dt' \approx \frac{m}{2kT\tau_B} \sum_{k=0}^{M_t} \left(\frac{\mathbf{R}_i(t_k) - \mathbf{R}_i(t_{k-1})}{\Delta t} \right)^2 \Delta t$$



Diffusion in Inhomogeneous Fluids: Dynamic, Classical DFT

- Consider continuity equation of concentration

$$\frac{d}{dt}P(\mathbf{r}^N, t) = \frac{\partial P(\mathbf{r}^N, t)}{\partial t} + \sum_{i=1}^N \nabla_i \cdot [\mathbf{v}_i P(\mathbf{r}^N, t)] = 0$$

- Recall for steady drag $\mathbf{v}_i(t) = v_{0,i} e^{-t/\tau_B} + \frac{1}{m} \int_0^t e^{-(t-t')/\tau_B} \mathbf{F}_i(t') dt'$

- Assume force *independent of time* and initial velocity is zero

$$\mathbf{F}_i = -\nabla_i V(\mathbf{r}^N) = -kT \nabla_i \ln[P(\mathbf{r}^N)] - \nabla_i V^{ex}(\mathbf{r}^N)$$

$$\mathbf{v}_i(t) = \frac{\mathbf{F}_i}{m} \int_0^t e^{-(t-t')/\tau_B} dt' = \frac{\tau_B}{m} (1 - e^{-t/\tau_B}) \mathbf{F}_i$$

- Which gives

$$\frac{\partial P(\mathbf{r}^N, t)}{\partial t} = \frac{1}{\gamma} (1 - e^{-t/\tau_B}) \sum_{i=1}^N \nabla_i \cdot [P(\mathbf{r}^N, t) \nabla_i V(\mathbf{r}^N)]$$

- Integrating over $N-1$ particles yields

$$\frac{\partial \rho(\mathbf{r}, t)}{\partial t} = \frac{1}{\gamma} (1 - e^{-t/\tau_B}) \nabla \cdot [\rho(\mathbf{r}, t) \nabla \mu(\mathbf{r})]$$

$$\nabla \mu(\mathbf{r}) \equiv \nabla \left[\frac{\delta F[\rho^{eq}(\mathbf{r})]}{\delta \rho^{eq}(\mathbf{r})} \right]$$

$$\nabla \mu(\mathbf{r}) = kT \nabla \ln[\rho(\mathbf{r})] + \nabla \mu^{ex}(\mathbf{r})$$



Diffusion in moderately concentrated colloidal suspensions in thermal equilibrium



Langevin Equation with Interactions

- **Brownian Dynamics Simulations**

- Markov assumption

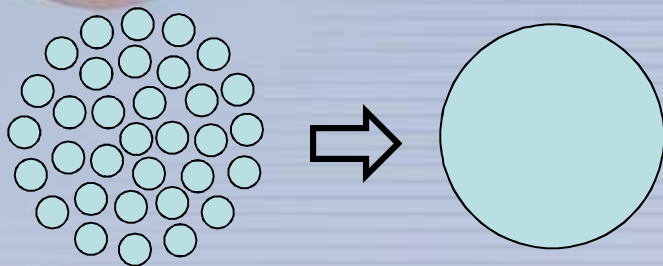
$$m \frac{d\mathbf{v}_i}{dt} = -\mathbf{R} \cdot \mathbf{v}_i + \sum_{j \neq i} \mathbf{F}_{ij}^{colloid}(|\mathbf{r}_i - \mathbf{r}_j|) + \mathbf{F}_i^B(t)$$

$$\langle \mathbf{F}_i^B(t) \rangle = 0; \langle \mathbf{F}_i^B(t) \mathbf{F}_j^B(t') \rangle = 2\mathbf{R}_{FU} k_B T \delta_{ij} \delta(t - t')$$

- $\mathbf{F}^B$ assumed Gaussian distributed and self similar
- Colloid inertia is often neglected (we don't)
- **Can we transform spatial interactions into time (convolution) integral with memory kernel?**
 - This is a constitutive relation we can measure



Simulation Details: Colloid interactions



$$U_A = -\frac{A_{12}}{6} \left[\frac{2a_1a_2}{r_{12}^2 - (a_1 + a_2)^2} + \frac{2a_1a_2}{r_{12}^2 - (a_1 - a_2)^2} + \ln \left(\frac{r_{12}^2 - (a_1 + a_2)^2}{r_{12}^2 - (a_1 - a_2)^2} \right) \right]$$

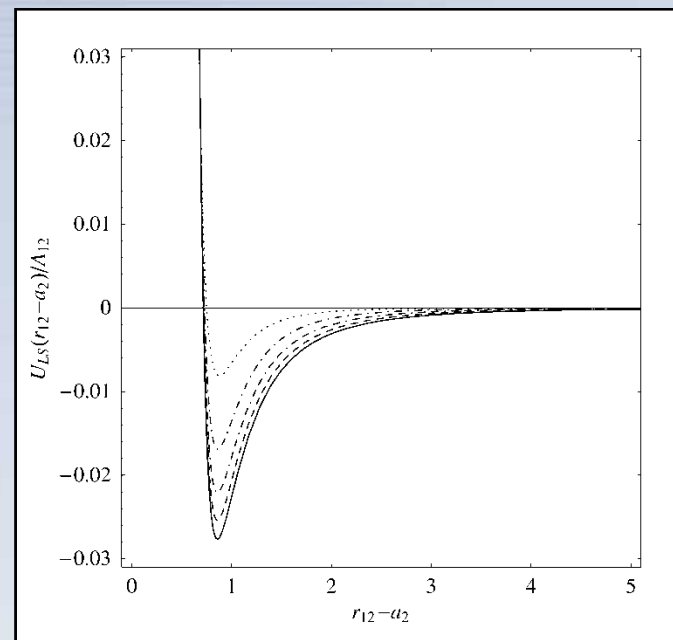
$$U_R = \frac{A_{12}}{37800 r_{12}} \left[\frac{r_{12}^2 - 7r_{12}(a_1 + a_2) + 6(a_1^2 + 7a_1a_2 + a_2^2)}{(r_{12} - a_1 - a_2)^7} + \frac{r_{12}^2 + 7r_{12}(a_1 + a_2) + 6(a_1^2 + 7a_1a_2 + a_2^2)}{(r_{12} + a_1 + a_2)^7} - \frac{r_{12}^2 + 7r_{12}(a_1 - a_2) + 6(a_1^2 - 7a_1a_2 + a_2^2)}{(r_{12} + a_1 - a_2)^7} - \frac{r_{12}^2 - 7r_{12}(a_1 - a_2) + 6(a_1^2 - 7a_1a_2 + a_2^2)}{(r_{12} - a_1 + a_2)^7} \right]$$

$$U = U_A + U_R, \quad r_{12} < r_c$$

- Integrated Lennard-Jones potential¹

- Repulsive soft-sphere

- $a_1 = a_2 = 5\sigma$
- $A_{12} = 1$
- $r_c = r_{\min} = 30^{-1/6}\sigma$



¹R. Everaers and M.R. Ejtehadi, Phys. Rev. E **67**, 41710 (2003)



Simulation Details: Hydrodynamic Interactions

- **Markovian assumption**

- Steady state (quasistatic) incompressible Newtonian flow

- Stokesian Dynamics

- PME

- » $O(N \log N)$

$$R = (I - \mathcal{R})^{-1} R_{1B} + R_{lub}$$

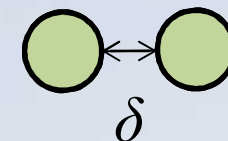
- Fast Lubrication Dynamics

- $O(N)$

$$R = R_0 + R_\delta$$

Isotropic Constant
(mean-field mobility)

$$\mathbf{R}_0 = 3\pi\mu d(1 + 2.16\phi)\mathbf{I}$$



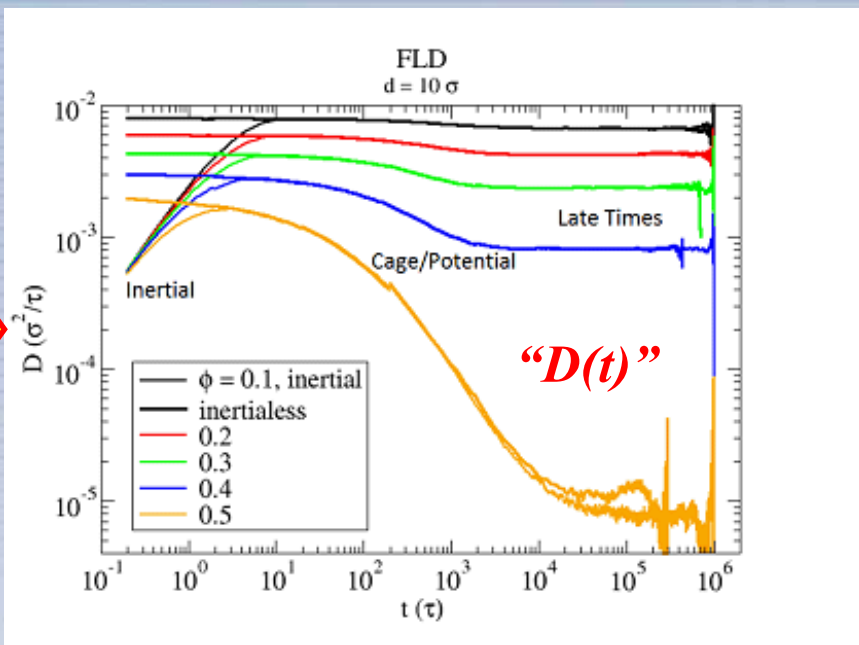
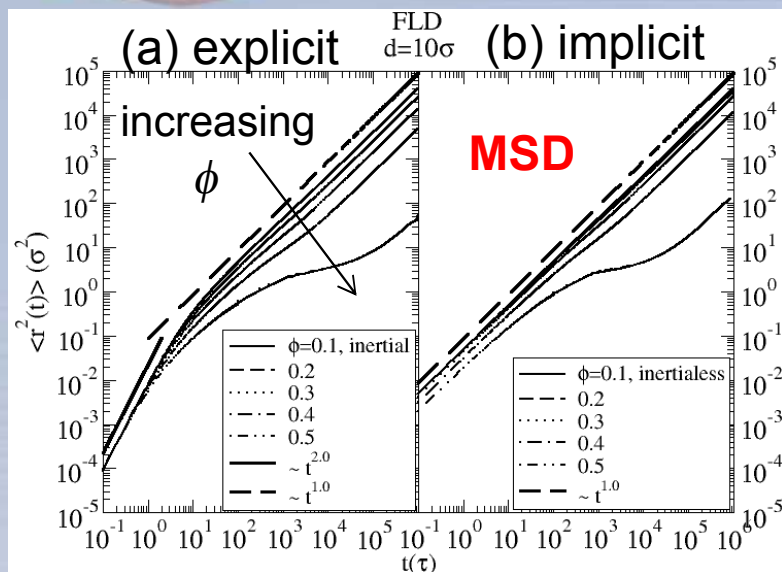
δ^{-1} or $\delta^{-1} + \log(\delta^{-1})$

Kumar and Higdon, Phys Rev E, 82, 051401 (2010)

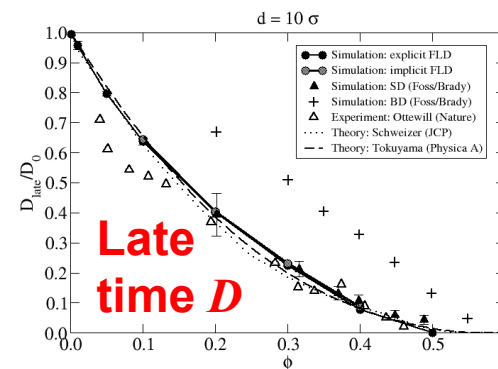
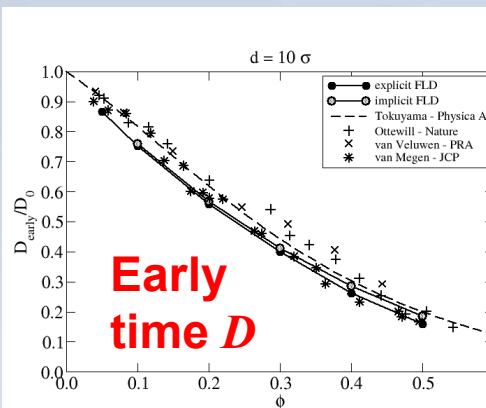
Ball and Melrose, Physica A, 247, 444-472 (1997)



Simulation Results: MSD and Validation

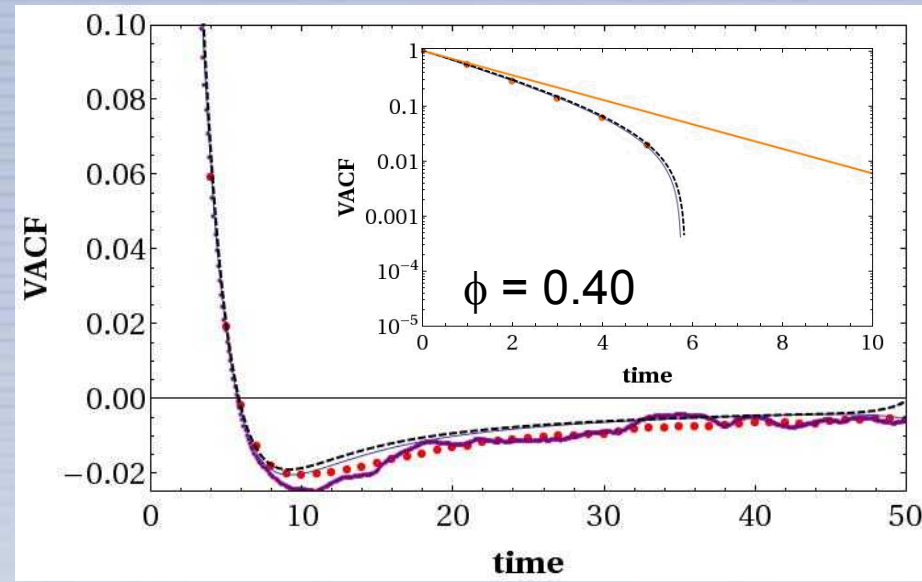
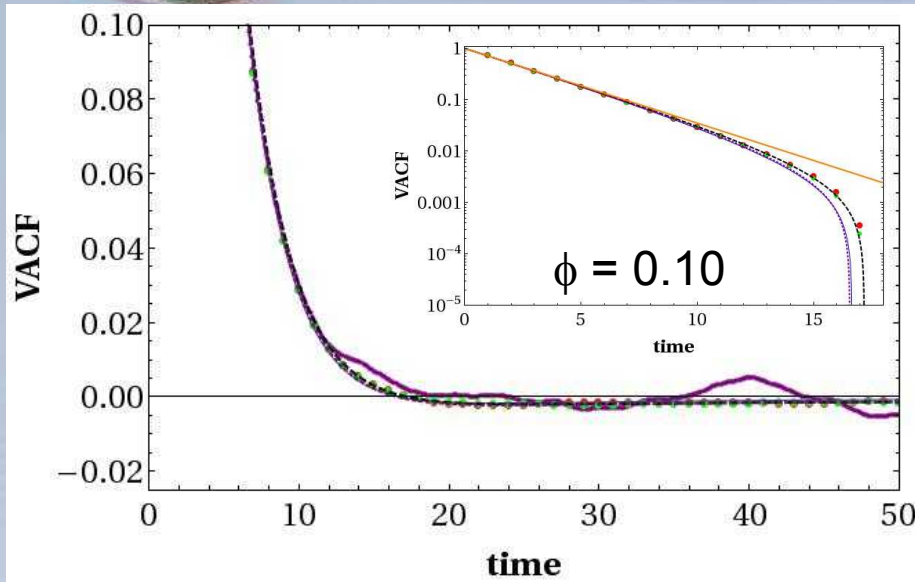


- $MSD = f(t) \neq Dt^\alpha$; $\alpha \neq 1$
 - “Early” and “Late” time D compare well with experiments
 - Engineer’s temptation
 - $D = D(t) \rightarrow MSD = D(t)t$





Simulation Results: VACF, Memory Kernel and “Microscopic Dynamics”



- **Fit VACF with**

$$f(t/\tau_B; \tau_{coll}, \theta) \propto \sum_{n=0}^{\infty} \frac{(-1)^n (t/\tau_{coll})^{2(n+1)-\theta n-2}}{n!} \sum_{k=0}^{\infty} \frac{(k+n)!(-1)^k}{k! \Gamma(1+2n+k-\theta n)} (t/\tau_B)^k$$

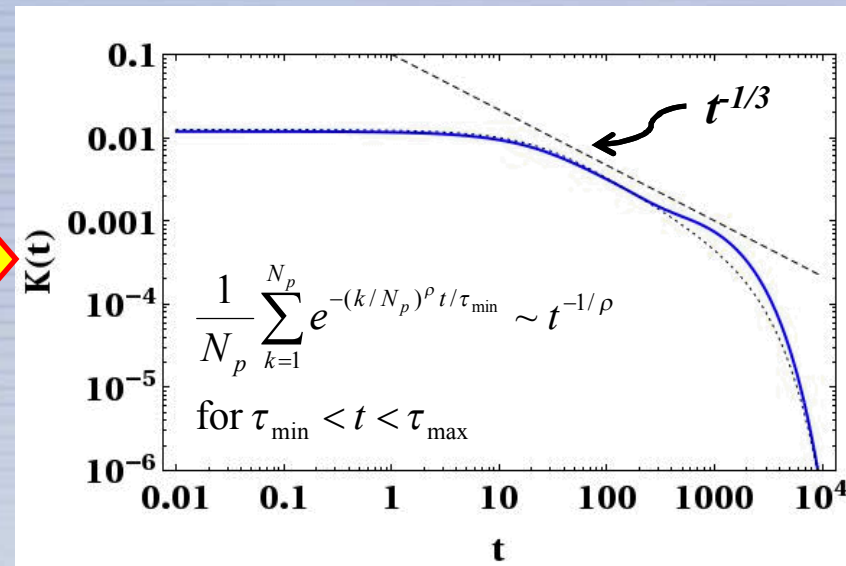
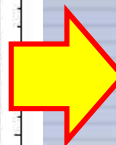
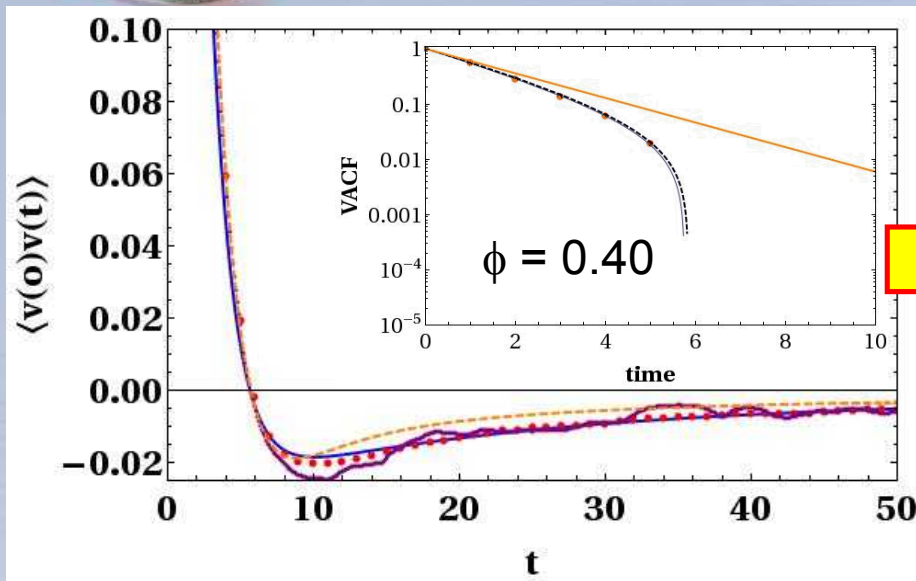
- **Which is a fundamental solution of fBm-type equation**

$$\frac{d\mathbf{v}_i(t)}{dt} = -\frac{1}{\tau_B} \mathbf{v}_i(t) - \frac{1}{\tau_{coll}^{2-\theta}} \int_0^t (t-t')^{-\theta} \mathbf{v}_i(t') dt' + \frac{1}{m_i} (\mathbf{F}_i^B(t) + \mathbf{F}_i^{fBm}(t)); \quad 0 < \theta < 1$$

$$\langle \mathbf{F}_i^{fBm}(t) \rangle = 0; \quad \langle \mathbf{F}_i^{fBm}(t) \mathbf{F}_i^{fBm}(t') \rangle = \frac{2k_B T \gamma}{\tau_{coll}} \left(\frac{t-t'}{\tau_{coll}} \right)^{-\theta}$$



“Microscopic Dynamics”: Alternate Form for VACF and Memory Kernel



- Or, fit VACF with

$$f(t) = -A_N \sum_{i=0}^N c_i \lambda_i e^{-\lambda_i t} \quad A_N = \frac{1}{\sum_{i=0}^N c_i / \lambda_i}$$

- Which leads to

$$m_i \frac{d\mathbf{v}_i(t)}{dt} = - \int_0^t K(t-t') \mathbf{v}_i(t') dt' + \mathbf{F}_i^R(t) \quad K(t) \sim \delta(t) + \frac{1}{N} \sum_{k=1}^N e^{-(k/N)^\rho t / \tau_{\text{coll}}}$$

$$0.00574 e^{-0.0436 t} + 0.00446 e^{-0.0105 t} + 0.00165 e^{-0.000811 t} + 0.664 \delta(t)$$

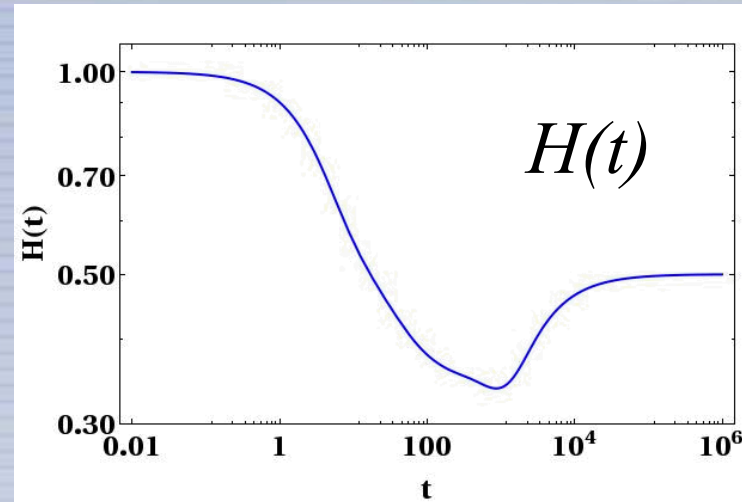


Relaxation-Diffusion Equation for Colloids

- $MSD = f(t) = Dt^{\alpha(t)}; \alpha(t) = 2H(t)$

– Or, $= D\tau_D(t); \frac{d\tau_D(t)}{dt} = 2H(t)$

$$\langle |\Delta \mathbf{r}|^2 \rangle(t) = 6 \left(\frac{kT}{m} \tau_B \right) \tau_D(t)$$



$$\tau_D(t) \rightarrow \frac{t}{\tau_B} + \frac{\tau_{coll}}{N\tau_B} \sum_{k=1}^N \left(\frac{N}{k} \right)^{\beta_{coll}} \left(1 - e^{-(k/N)^{\beta_{coll}} t / \tau_{coll}} \right) - 2(1 - e^{-t/\tau_B}); \tau_B < \tau_{coll}$$



$$\frac{\partial \rho(\mathbf{x}, \tau_D(t))}{\partial \tau_D(t)} = D \nabla^2 \rho(\mathbf{x}, \tau_D(t))$$



$$\rho(\mathbf{x}, t) = \rho(\mathbf{x}, 0) + D \int_0^t \frac{d\tau_D(t-t')}{dt} \nabla^2 \rho(\mathbf{x}, t') dt'$$



Summary: Langevin Equations to Generalized Diffusion Equations

- **Brownian Dynamics Simulations**

- Markovian: Langevin with interactions

$$m \frac{d\mathbf{v}_i}{dt} = -\mathbf{R} \cdot \mathbf{v}_i + \sum_{j \neq i} \mathbf{F}_{ij}^{colloid}(|\mathbf{r}_i - \mathbf{r}_j|) + \mathbf{F}_i^B(t)$$

$$\langle \mathbf{F}_i^B(t) \rangle = 0; \langle \mathbf{F}_i^B(t) \mathbf{F}_j^B(t') \rangle = 2\mathbf{R}_{FU} k_B T \delta_{ij} \delta(t - t')$$

- Non-Markovian: Generalized Langevin

$$m_i \frac{d\mathbf{v}_i(t)}{dt} = -\int_0^t K(t-t') \mathbf{v}_i(t') dt' + \mathbf{F}_i^R(t)$$

$$K(t) \sim \delta(t) + \frac{1}{N} \sum_{k=1}^N e^{-(k/N)^\rho t/\tau_{coll}}$$

$$\langle \mathbf{F}_i^R(t) \rangle = 0; \langle \mathbf{F}_i^R(t) \mathbf{F}_j^R(t') \rangle = 2k_B T \delta_{ij} K(t-t')$$

- **Can transform spatial interactions into convolution integral + “noise”**

- Then can obtain relaxation-diffusion equation



Micro-rheology to Macro-rheology

- Mason and Weitz**

- microscopic memory function proportional to macroscopic bulk frequency-dependent viscosity (mean-field approximation)

$$\frac{d\mathbf{v}_i(t)}{dt} = \int_0^t K(t-t') \mathbf{v}_i(t') dt' + \mathbf{F}_i^R(t) \quad \tilde{\mu}(s) = \frac{\tilde{K}(s)}{6\pi a} \quad \tilde{\mu}(s) \approx \tilde{G}_r(s)$$

- Relaxation Modulus for Transient BD**

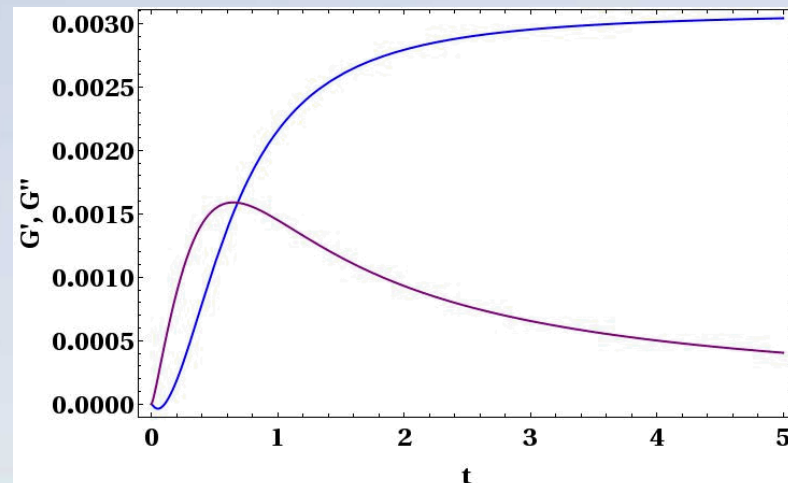
$$\tilde{G}_r(s) \approx \frac{\tilde{K}(s)}{6\pi a} \sim \frac{1}{6\pi Na} \sum_{k=1}^N \frac{\tau_{coll}}{s\tau_{coll} + (k/N)^\rho}$$

$$K(t) \sim \frac{1}{N} \sum_{k=1}^N e^{-(k/N)^\rho t/\tau_{coll}}$$

- Viscoelastic (complex) modulus**

$$\tilde{G}(s) = s\tilde{G}_r(s); \quad s \rightarrow i\omega$$

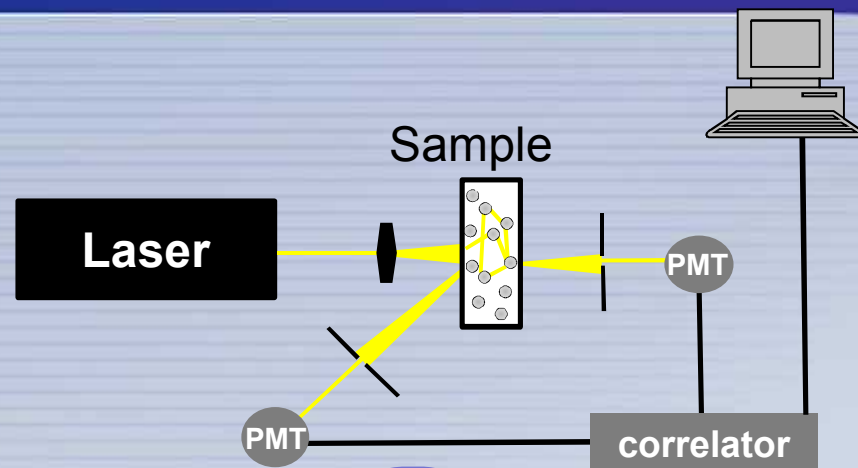
$$G^*(\omega) \sim \frac{1}{6\pi Na} \sum_{k=1}^N \frac{\tau_{coll}}{i\omega\tau_{coll} + (k/N)^\rho}$$



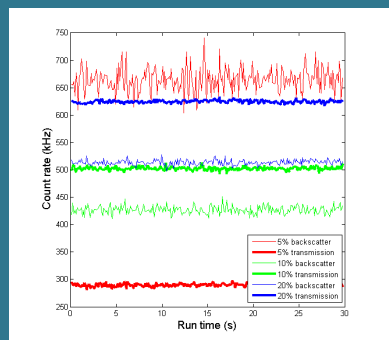


Validation: Diffusing-Wave Spectroscopy

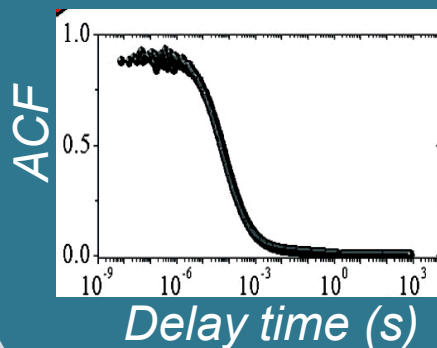
- Diffusivity of turbid suspensions
- Connect stochastic colloidal dynamics to rheology



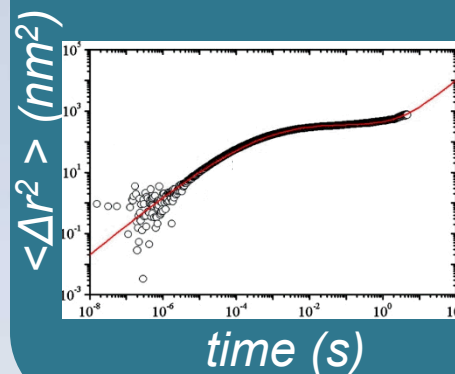
Measure:
intensity fluctuations
vs. time



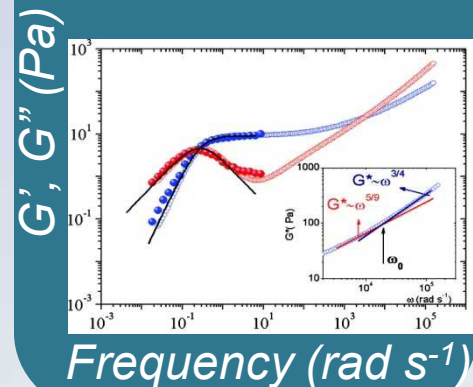
Autocorrelation :
Transmission $\rightarrow f(\tau, I^*)$
Backscatter $\rightarrow f(\tau)$



Mean square displacement



Rheology





Analysis of Macro models: CTRW



Generalized (“Anomalous”) Diffusion

- Some *possible* models – nonlinear FPE, FFPE, etc.
- Consider CTRW with multiple time scales of the form

$$\frac{\partial \rho(x, t)}{\partial t} - \int_0^t g(t-t') \frac{\partial \rho(x, t')}{\partial t'} dt' = D \frac{\partial}{\partial t} \int_0^t g(t-t') \frac{\partial^2 \rho(x, t')}{\partial x^2} dt' \quad ^1$$

- Assume jump pdf can be decoupled into jump length pdf and waiting time pdf

$$\psi(k, s) = \phi(k)g(s)$$

- Assume jump length pdf is Gaussian
 - true for long times in any finite variance jump length distribution

$$\phi(k) \sim 1 - Dk^2 + O(k^4)$$

- Waiting time distribution, $g(t)$

$$\rho(k, s) = \frac{1 - g(s)}{s} \frac{\rho_0(k)}{1 - (1 - Dk^2)g(s)}$$

¹ Fa and Mendes (2010), *J. Stat. Mech.*



Macroscopic, Deterministic Equation for NonMarkovian Stochastic Dynamics

- **CTRW**

- **Solution**

$$\rho(k, s) = \frac{1 - g(s)}{s} \frac{\rho_0(k)}{1 - (1 - Dk^2)g(s)}$$



$$s \langle \Delta \tilde{x}^2(s) \rangle - \langle \Delta x^2(0) \rangle = \frac{2Dg(s)}{1 - g(s)}$$

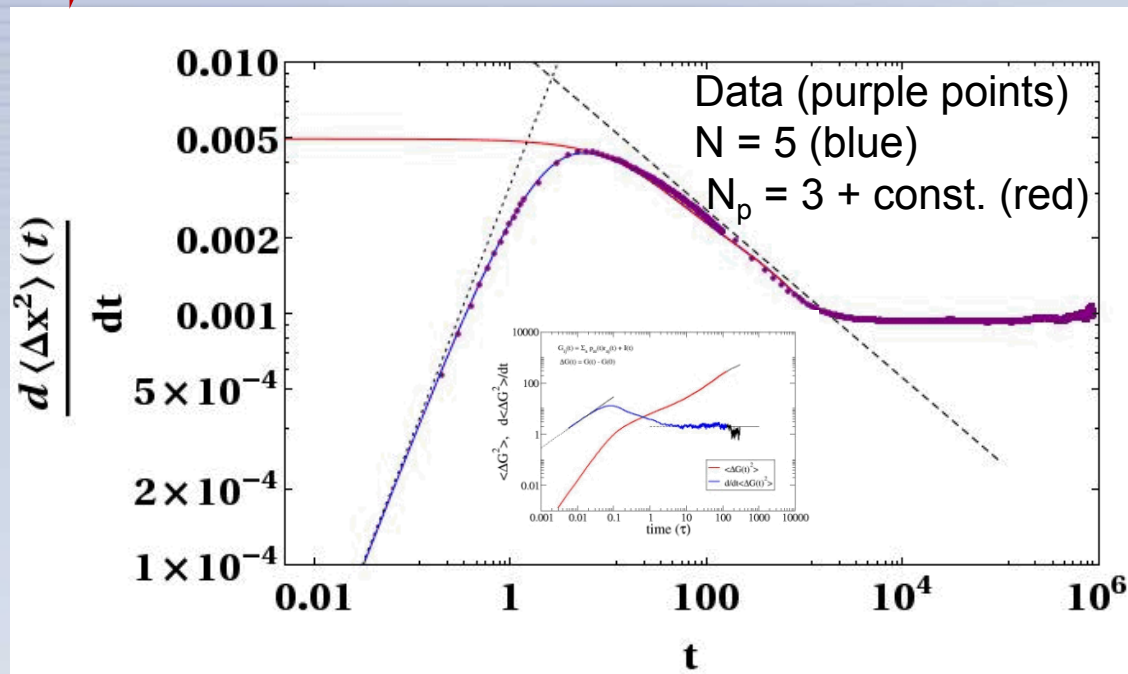
- **Choose $g(t)$**

$$g(t) = A_N \sum_{i=0}^N c_i e^{-\lambda_i t}$$

$$A_N = \frac{1}{\sum_{i=0}^N c_i / \lambda_i}$$

- **Note:**

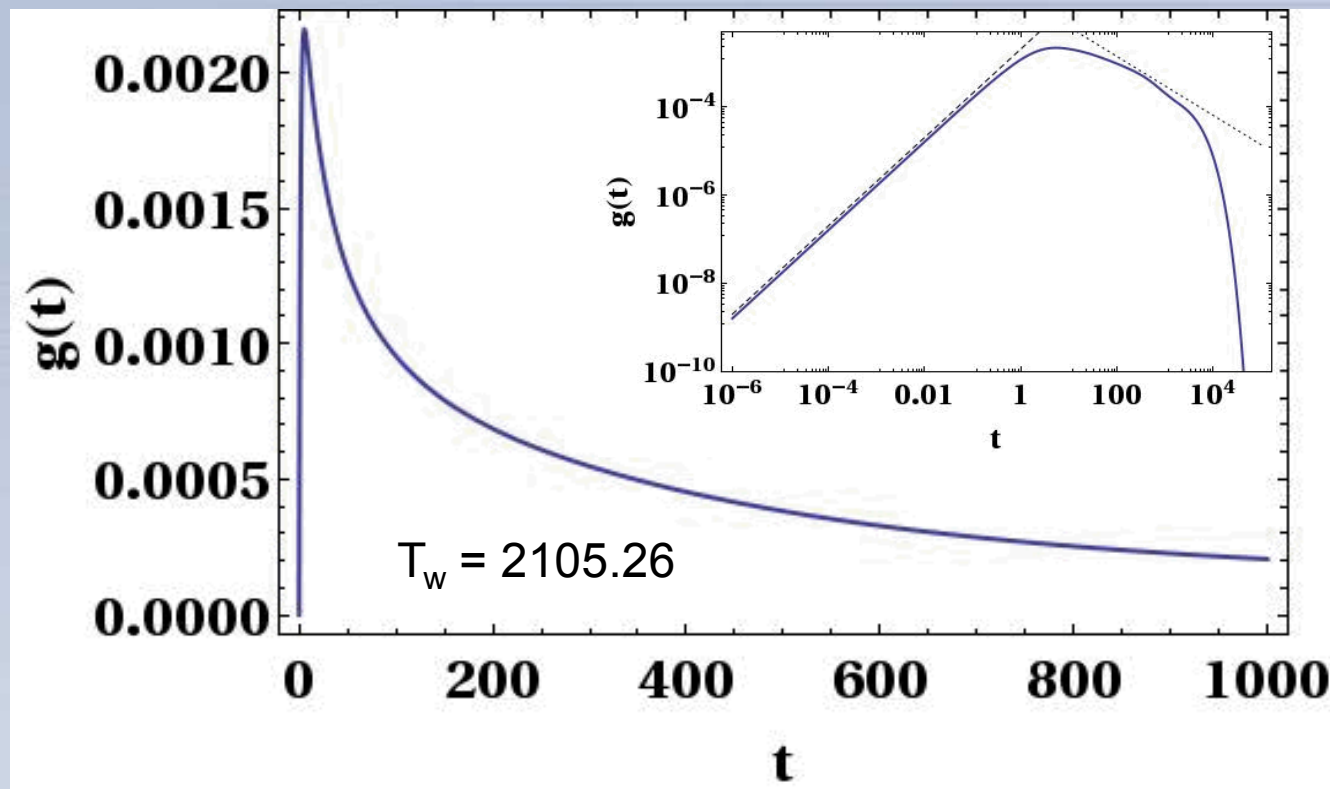
$$\frac{1}{N_p} \sum_{k=1}^{N_p} e^{-(k/N_p)^\rho t / \tau_{\min}} \sim t^{-1/\rho} \quad \text{for } t > \tau_{\min}$$





Waiting Time Distribution

- Note:** $s\langle\Delta\tilde{x}^2(s)\rangle - \langle\Delta x^2(0)\rangle = \frac{2Dg(s)}{1-g(s)}$



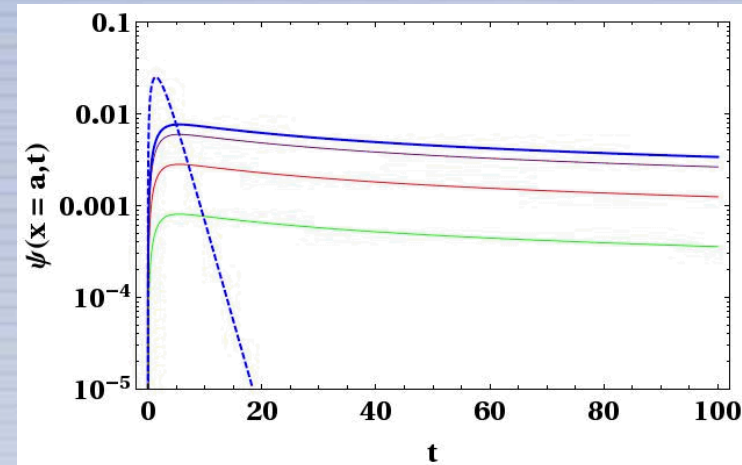
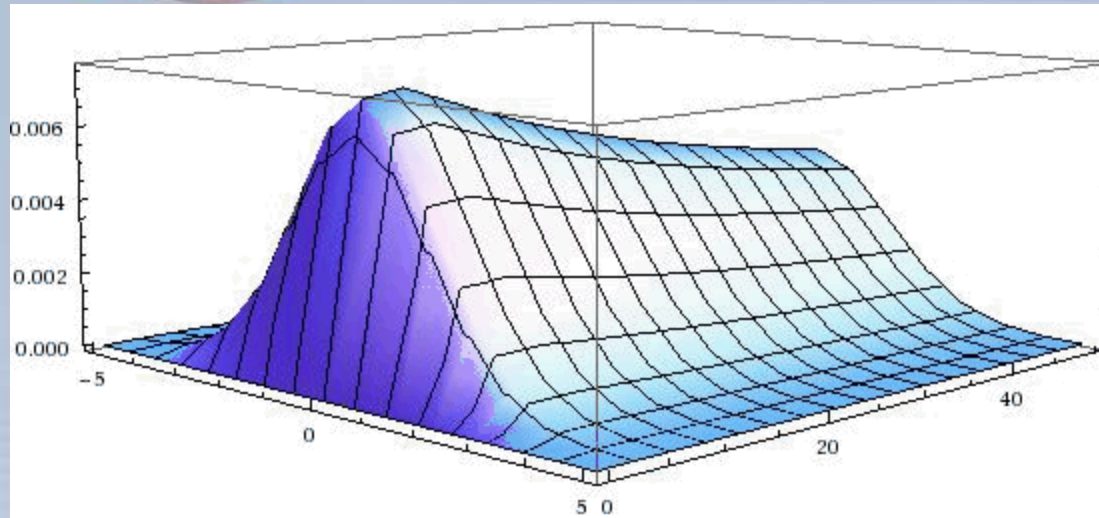
$$g(t) = A_N \sum_{i=0}^N c_i e^{-\lambda_i t}$$

$$A_N = \frac{1}{\sum_{i=0}^N c_i / \lambda_i}$$

$$-0.00249419 e^{-0.664182 t} + 0.000725312 e^{-0.0562511 t} + \\ 0.00074885 e^{-0.0171688 t} + 0.000801025 e^{-0.00280788 t} + 0.000219001 e^{-0.000330834 t}$$



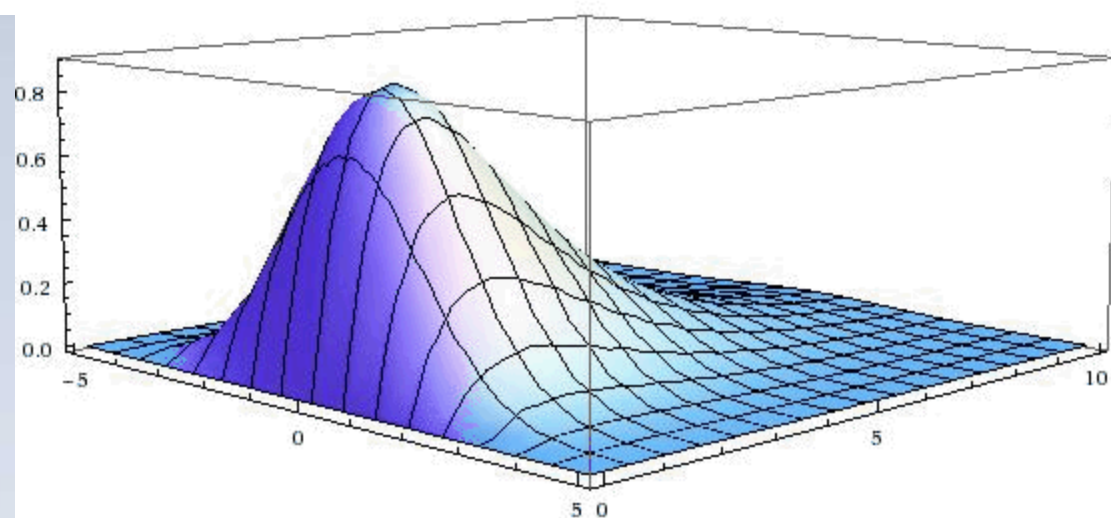
Full Jump Distributions



- **Recall**

$$\psi(k, s) = g(s)\phi(k)$$

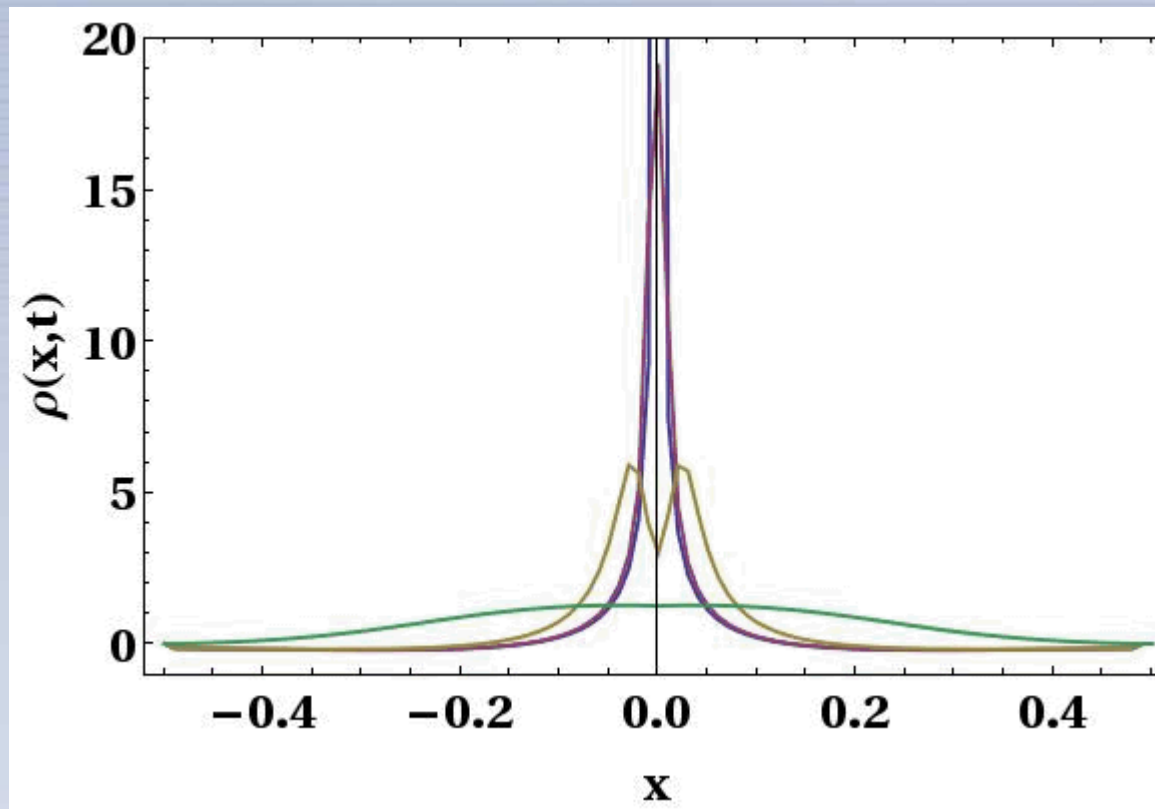
$$\phi(k) = 1 - \sigma^2 k^2$$





Green's Function for CTRW

- Note:** $\rho(k, s) = \frac{1 - g(s)}{s} \frac{W_0(k)}{1 - \psi(k, s)}$





Conclusions

- **Developing Meso-to-Macroscale Modeling framework**
 - Nonequilibrium Statistical Physics/Thermodynamics
- **Solution of Langevin Equation with interactions leads to GLE**
 - Can obtain Memory Kernel of “reasonable” form via VACF
 - Leads to rheology
 - Comparison to Experiments forthcoming
- **Can determine waiting time distribution for CTRW**
 - Assuming Gaussian Jumps
 - Leads to Generalized diffusion equation
 - However, CTRW framework not consistent in dilute limit!?
 - Sum of exponentials may be “degenerate”



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