

Brownian HydroDynamics of Colloidal Suspensions

Aleksandar Donev

Courant Institute, New York University

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Brownian Motion



Experiments: Non-Spherical Designer Colloids

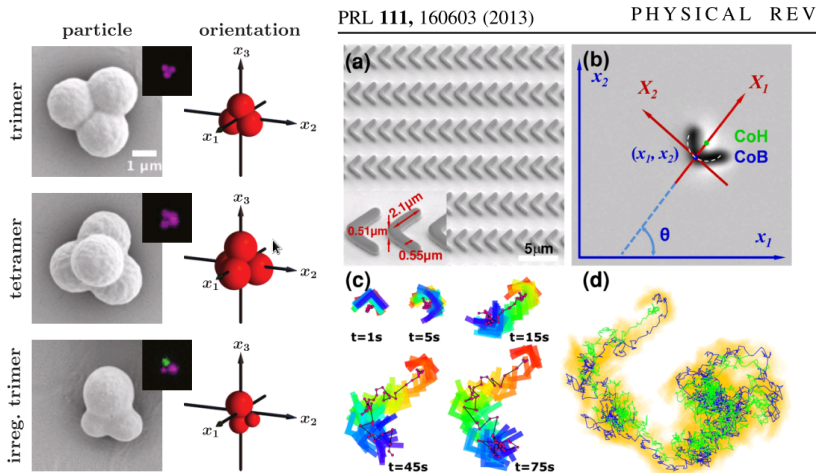


Figure: (Left) Cross-linked spheres; Kraft et al. PRE 2013. (Right) Lithographed boomerangs; Chakrabarty et al. PRL 2013.

Simulations: Dense Boomerang Suspension

PRL **111**, 160603 (2013)

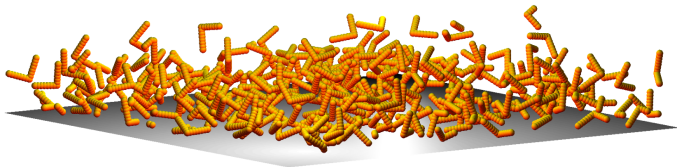
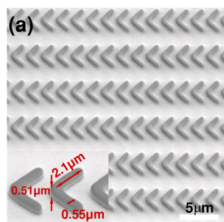
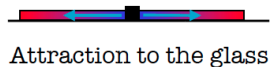
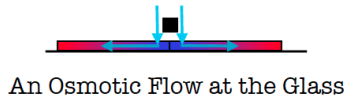
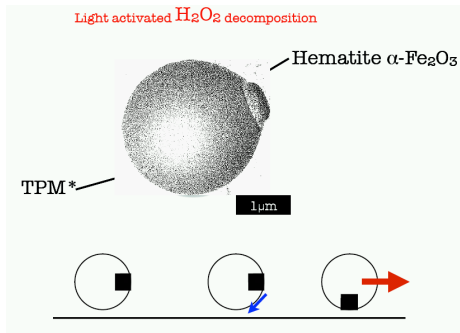


Figure: (Left) Lithographed boomerang colloids. (Right) Brownian dynamics of boomerangs above a bottom wall [1].

Light-Activated Diffusio/Osmophoresis

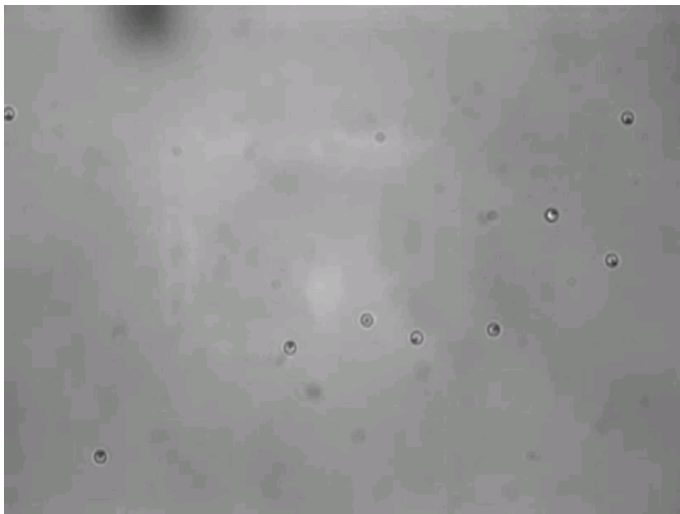


Symmetric Flow but unstable
=> Self-Propulsion



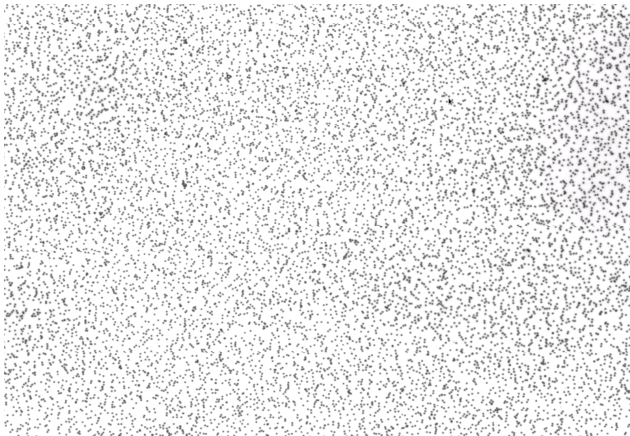
Figure: From Jeremie Palacci and Paul Chaikin (Science 2013)

Light-Activated Colloidal Surfers



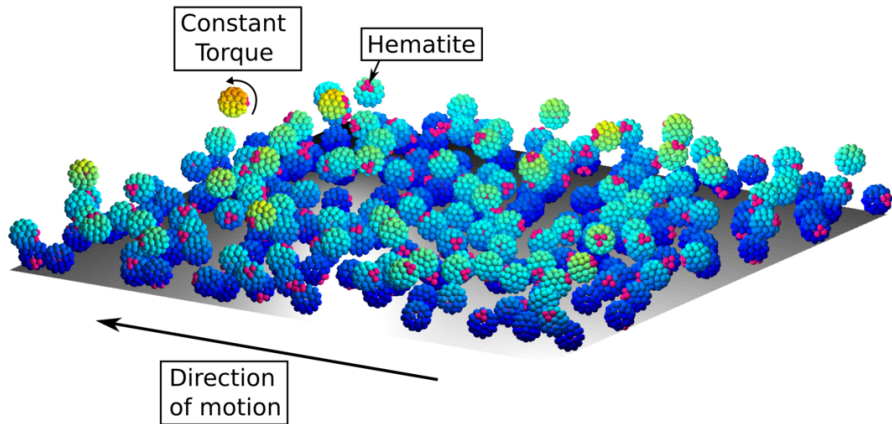
QuickTime

Uniform Suspension of Microrollers: Simulation



Experiments on uniform suspensions by Michelle Driscoll (in progress).

Uniform Suspension of Microrollers: Simulation



Simulations by **Brennan Sprinkle** [1] of a uniform suspension of microrollers at packing fraction $\phi = 0.4$ (MP4).

Bent Active Nanorods

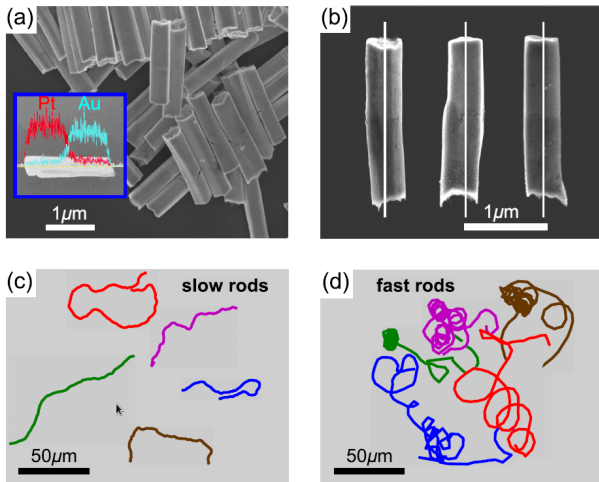
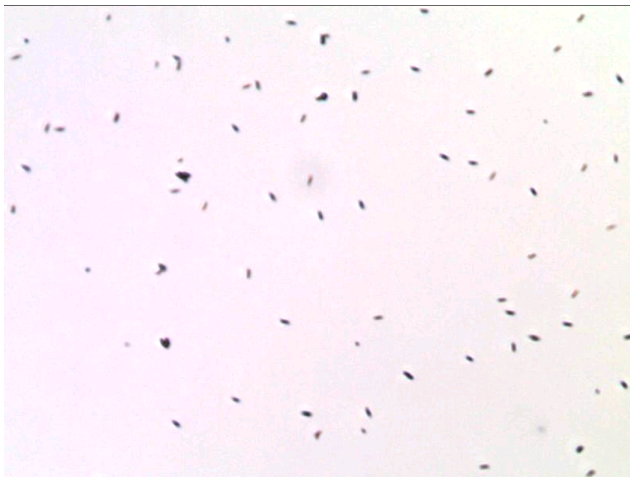


Figure: From the Courant Applied Math Lab of Michael Shelley

Thermal Fluctuation Flips



QuickTime

Brownian Motion

- Consider a **single spherical particle** of radius a with position $\mathbf{q}(t)$ diffusing **in an unbounded domain** with the fluid at rest at infinity.
- If there are no forces applied on the particle, the displacement of the particle in each direction over a time interval Δt has a normal (Gaussian) probability distribution with mean zero and variance (**mean square displacement**)

$$\langle (q_\alpha(t + \Delta t) - q_\alpha(t))^2 \rangle = 2D\Delta t, \quad \alpha = x, y, z$$

where D is the **diffusion coefficient** in units of m^2/s .

- Therefore, we can write the recurrence relationship

$$q_\alpha(t + \Delta t) = q_\alpha(t) + \mathcal{N}(0, 2D\Delta t),$$

where $\mathcal{N}(m, \sigma^2)$ denotes a Gaussian random variable (pseudo-random number on a computer) with mean m and variance σ^2 (standard deviation σ).

The simplest SDE

- If we take the **time step size** $\Delta t \rightarrow 0$ the trajectory $\mathbf{q}(t)$ converges to a continuous-time stochastic process with (almost surely) continuous trajectories that we call **Brownian Motion**.
- In this limit we formally write this as a **stochastic differential equation** (SDE)

$$\frac{d}{dt}q_\alpha(t) = \sqrt{2D} \mathcal{W}_\alpha(t) \quad (\text{physics notation})$$

$$dq_\alpha(t) = \sqrt{2D} d\mathcal{B}_\alpha(t) \quad (\text{math notation}),$$

where $\mathcal{W}(t) \equiv d\mathcal{B}(t)/dt$ is a **white noise process**, and $\mathcal{B}(t)$ is the **standard Wiener process** or standard Brownian motion.

- I will employ heavily vector/matrix (physics) notation,

$$\frac{d\mathbf{q}(t)}{dt} = \sqrt{2D} \mathcal{W}(t).$$

Classical Fluid Dynamics

- Now imagine that the sphere was large (macroscopic) so that Brownian motion did not play a role (more on this soon).
- If we apply a force \mathbf{F} on the sphere (e.g. gravity), small enough so that **Reynolds number** $\text{Re} \ll 1$, hydrodynamics (steady Stokes equations) says that the velocity of the particle (e.g., sedimenting sphere) is

$$\mathbf{u} = \frac{d\mathbf{q}}{dt} = \frac{1}{6\pi\eta a} \mathbf{F} = \mu \mathbf{F},$$

where μ is the **mobility** of the sphere.

- We also have the following fundamental **Einstein relationship** between diffusion and mobility:

$$D = (k_B T) \mu.$$

- Understanding where this comes from requires a whole class on **nonequilibrium statistical mechanics**.

Single Colloidal Sphere

- Assuming we can combine these gives the Stokes-Einstein relation (approximate but nearly exact)

$$D \approx \frac{k_B T}{6\pi\eta a} \Rightarrow \frac{d\mathbf{q}(t)}{dt} = \mu \mathbf{F} + \sqrt{2k_B T \mu} \mathcal{W}(t).$$

- When the particle is **confined near walls** (no-slip boundaries) the diffusion coefficient depends on how far the particle is from the wall. Different directions are also different – mobility is in general a 3×3 matrix, i.e., a **mobility tensor**.
- The more general SDE for Brownian motion is then

$$\frac{d\mathbf{q}(t)}{dt} = \left(\boldsymbol{\mu}(\mathbf{q}) \mathbf{F} + (k_B T) \frac{\partial}{\partial \mathbf{q}} \cdot \boldsymbol{\mu}(\mathbf{q}) \right) + \sqrt{2k_B T \boldsymbol{\mu}(\mathbf{q})} \mathcal{W}(t).$$

Note that $\frac{dx(t)}{dt} = a(x, t) + \sqrt{b(x)} \mathcal{W}(t)$

is notation for the limit as $\Delta t \rightarrow 0$ of

$$x(t + \Delta t) = x(t) + a(x(t), t) \Delta t + \mathcal{N}(0, b(x(t)) \Delta t).$$

When is Brownian motion “important”

- The **diffusion time** is the time it takes a particle to diffuse one radius,

$$\tau = \frac{a^2}{D} = \frac{6\pi\eta a^3}{k_B T} \approx \frac{(a/(1\mu\text{m}))^3}{0.2} \text{s}.$$

- If $a \sim 1\text{mm}$ then $\tau \sim 10^6\text{s}$ which is quite long: We don't see sand particles diffusing.
- But if $a = 1\mu\text{m}$, a typical colloidal particle made in the lab, then $\tau \approx 5\text{s}$ which is observable by microscopes.
- Now what if there was also convective/advective flow carrying the particles with speed v ? We define the dimensionless **Péclet number**

$$\text{Pe} = \frac{va}{D}.$$

- If $\text{Pe} \lesssim 1$, then diffusion is “important” and must be included.
- But importantly, we see that deterministic and random motions are intimately linked and given by the same **hydrodynamic mobility**.

Two Spheres Far Away

- Now what if there were two spheres and we applied a force on one of them? The force would create a fluid flow velocity $\mathbf{v}(\mathbf{r})$ and the other particle would move also; this is called **hydrodynamic interaction** although this is a misnomer.

Recall $Re \ll 1$ and we assume steady Stokes, so not $\mathbf{v}(\mathbf{r}, t)$.

- Since the steady Stokes equations are linear, we have that

$$\mathbf{u}_2 \approx \mathbf{v}(\mathbf{q}_2) = \boldsymbol{\mu}_{12}(\mathbf{q}_1, \mathbf{q}_2) \mathbf{F}_1,$$

where $\boldsymbol{\mu}_{12}$ is the 3×3 **pair mobility tensor**.

- The Einstein relationship tells us that the Brownian motions of the two spheres would become correlated. So one can call this **hydrodynamic correlations**.
- If the spheres were far apart at a distance $r \gg a$, then they would look like “point particles.”

Point singularity approximation

- A force applied at a point \mathbf{q} is called a Stokeslet. The flow it creates is the solution to the steady Stokes equation

$$\nabla \pi(\mathbf{r}) = \eta \nabla^2 \mathbf{v}(\mathbf{r}) + \mathbf{F} \delta(\mathbf{r} - \mathbf{q})$$

$$\nabla \cdot \mathbf{v} = 0 \quad + \text{boundary conditions,}$$

which is also called the Green's function for Stokes flow,

$$\mathbf{v}(\mathbf{r}) = \mathbb{G}(\mathbf{r}, \mathbf{q}).$$

- For a three dimensional unbounded domain, the Green's function is the so-called **Oseen tensor**, with $\mathbf{r} = \mathbf{r}' - \mathbf{r}''$:

$$\mathbb{G}(\mathbf{r}', \mathbf{r}'') \equiv \mathbb{O}(\mathbf{r}' - \mathbf{r}'') = \frac{1}{8\pi\eta r} \left(\mathbf{I} + \frac{\mathbf{r} \otimes \mathbf{r}}{r^2} \right). \quad (1)$$

- So for two spheres far away with applied forces we can add all the pieces together because of the linearity of Stokes equations, and ignore for now Brownian motion,

$$\mathbf{u}_{1/2} = \frac{d\mathbf{q}_{1/2}}{dt} \approx \frac{1}{6\pi\eta a} \mathbf{F}_{1/2} + \mathbb{O}(\mathbf{q}_1 - \mathbf{q}_2) \mathbf{F}_{2/1}.$$

Mobility matrix

- For a collection of spheres we can very generally write in matrix notation

$$d\mathbf{Q}/dt = \mathcal{M}(\mathbf{Q})\mathbf{F},$$

where $\mathbf{Q}(t) = \{\mathbf{q}_1(t), \dots, \mathbf{q}_N(t)\}$ collects the particle positions and \mathbf{F} collects the forces applied on the particles.

- The **mobility matrix** $\mathcal{M}(\mathbf{Q})$ is symmetric and has all positive eigenvalues (is positive definite) – it encodes the hydrodynamic interactions/correlations.
- If there are applied torques on the particles, this will induce translational motion, especially near boundaries, and we have

$$d\mathbf{Q}/dt = \mathcal{M}\mathbf{F} + \mathcal{M}_c\mathbf{T},$$

- But we can only analytically compute $\mathcal{M}(\mathbf{Q})$ or $\mathcal{M}_c(\mathbf{Q})$ approximately for low-density or **dilute suspensions**.

Blob-blob pairwise mobility

- Now, if the particles are not exactly points, there will be corrections to this. The next order of approximation gives that the pairwise mobility is the so-called **Rotne-Prager mobility** \mathcal{R} ,

$$\mu_{12} \approx \mathcal{R}(\mathbf{q}_1, \mathbf{q}_2) \text{ where}$$

$$\mathcal{R}(\mathbf{r}', \mathbf{r}'') = \eta^{-1} \left(\mathbf{I} + \frac{a^2}{6} \nabla_{\mathbf{r}'}^2 \right) \left(\mathbf{I} + \frac{a^2}{6} \nabla_{\mathbf{r}''}^2 \right) \mathbb{G}(\mathbf{r}', \mathbf{r}'') \Big|_{\substack{\mathbf{r}'=\mathbf{r}_j \\ \mathbf{r}''=\mathbf{r}_i}}.$$

- For an unbounded domain we call $\mathcal{R}(\mathbf{r}', \mathbf{r}'') = \mathcal{R}(\mathbf{r}' - \mathbf{r}'') = \mathcal{R}(\mathbf{r})$ the **Rotne-Prager-Yamakawa (RPY) tensor**, but this can be generalized to confined domains [2].
- When the two spheres overlap we need to define the RPY tensor differently, but this can be done in a way such that [2]

$$\mathcal{R}(\mathbf{r}, \mathbf{r}) = \frac{1}{6\pi\eta a} \mathbf{I}.$$

Blob-bob mobility matrix

- The 3×3 block \mathbf{M}_{ij} of mobility matrix $\mathcal{M}(\mathbf{Q})$ maps a force on particle j to a velocity of particle i .
- For dilute suspensions we might at first assume that each pair of particles is not affected by the other particles, and just add over all pairs by linearity, giving the **pairwise approximation**:

$$\mathbf{M}_{ij}(\mathbf{Q}) \equiv \mathbf{M}_{ij}(\mathbf{q}_i, \mathbf{q}_j) = \mathcal{R}(\mathbf{q}_i, \mathbf{q}_j) \quad \text{for all } (i, j).$$

- We will call spherically-symmetric particles that interact/correlate through the RPY mobility “**blobs**”.
- Even if the suspension is not dilute we may approximate the particles as blobs without violating basic physics laws!

Brownian Hydrodynamics with blobs

- Represent each spherical particle by a **single blob**, and solve the Ito equations of **Brownian HydroDynamics** for the (correlated) positions of the N spherical microrollers $\mathbf{Q}(t) = \{\mathbf{q}_1(t), \dots, \mathbf{q}_N(t)\}$,

$$d\mathbf{Q}/dt = \mathcal{M}\mathbf{F} + \mathcal{M}_c\mathbf{T} + k_B T (\partial_{\mathbf{Q}} \cdot \mathcal{M}) + (2k_B T \mathcal{M})^{\frac{1}{2}} \mathcal{W}(t).$$
- Computational issues (not discussed here heavily but very important to my research group):
 - How to compute **deterministic velocities** $\mathcal{M}\mathbf{F}$ (matrix-vector product) efficiently?
 - How to generate **Brownian increments** $\mathcal{N}(\mathbf{0}, 2k_B T \Delta t \mathcal{M})$ or, equivalently, **Brownian velocities** $\mathcal{N}(\mathbf{0}, (2k_B T / \Delta t) \mathcal{M})$ efficiently?
 - How to generate **stochastic drift** $k_B T (\partial_{\mathbf{Q}} \cdot \mathcal{M})$ efficiently by only multiplying vectors by \mathcal{M} , without derivatives.

Generating Brownian increments

- We need a fast way to compute the **Brownian velocities**

$$\mathbf{U}_b = \sqrt{\frac{2k_B T}{\Delta t}} \mathcal{M}^{\frac{1}{2}} \mathbf{W} = \mathcal{N}(\mathbf{0}, 2k_B T / \Delta t \mathcal{M})$$

where \mathbf{W} is a vector of Gaussian random variables.

- The product $\mathcal{M}^{\frac{1}{2}} \mathbf{W}$ can be computed iteratively by **repeated multiplication** of a vector by \mathcal{M} using (preconditioned) Krylov subspace **Lanczos methods**.
- When particles are sedimented close to a bottom wall, pairwise hydrodynamic interactions decay rapidly like $1/r^3$, which appears to be enough to make the Krylov method converge in a **small constant number of iterations**, without any preconditioning.

Stochastic drift term

$$\frac{d\mathbf{Q}(t)}{dt} = \mathcal{M}\mathbf{F} + (2k_B T \mathcal{M})^{\frac{1}{2}} \mathbf{W}(t) + (k_B T) \partial_{\mathbf{Q}} \cdot \mathcal{M}$$

- Key idea to get $(\partial_{\mathbf{Q}} \cdot \mathcal{M})_i = \partial \mathcal{M}_{ij} / \partial Q_j$ is to use **random finite differences (RFD)** [3]: If $\langle \Delta \mathbf{P} \Delta \mathbf{Q}^T = \mathbf{I} \rangle$,

$$\lim_{\delta \rightarrow 0} \frac{1}{\delta} \langle \left\{ \mathcal{M} \left(\mathbf{Q} + \frac{\delta}{2} \Delta \mathbf{Q} \right) - \mathcal{M} \left(\mathbf{Q} - \frac{\delta}{2} \Delta \mathbf{Q} \right) \right\} \Delta \mathbf{P} \rangle =$$

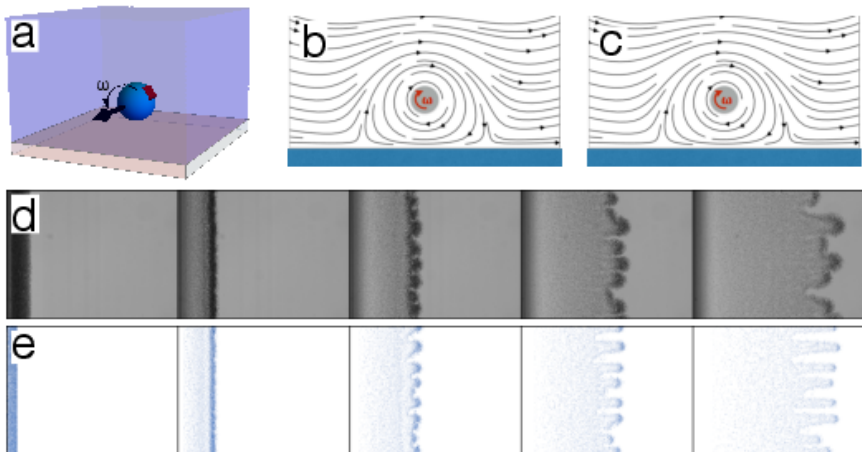
$$\{ \partial_{\mathbf{Q}} \mathcal{M}(\mathbf{Q}) \} : \langle \Delta \mathbf{P} \Delta \mathbf{Q}^T \rangle = k_B T \partial_{\mathbf{Q}} \cdot \mathcal{M}(\mathbf{Q}).$$

- This leads to a **stochastic Adams-Bashforth** temporal integrator [3],

$$\frac{\mathbf{Q}^{n+1} - \mathbf{Q}^n}{\Delta t} = \left(\frac{3}{2} \mathcal{M}^n \mathbf{F}^n - \frac{1}{2} \mathcal{M}^{n-1} \mathbf{F}^{n-1} \right) + \sqrt{\frac{2k_B T}{\Delta t}} (\mathcal{M}^n)^{\frac{1}{2}} \mathbf{W}^n$$

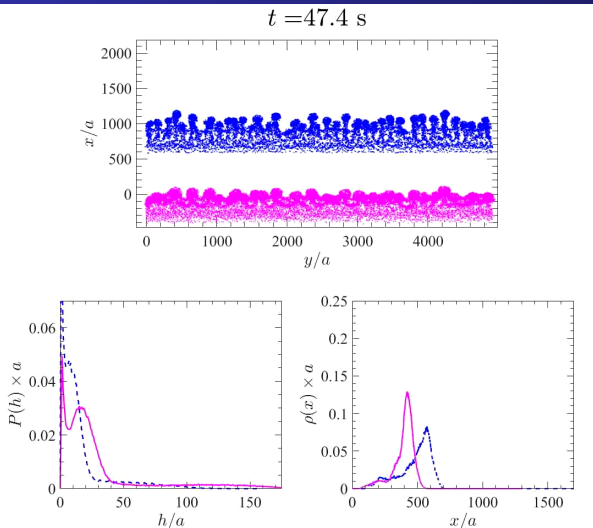
$$+ \frac{k_B T}{\delta} \left(\mathcal{M} \left(\mathbf{Q} + \frac{\delta}{2} \widetilde{\mathbf{W}}^n \right) - \mathcal{M} \left(\mathbf{Q} - \frac{\delta}{2} \widetilde{\mathbf{W}}^n \right) \right) \widetilde{\mathbf{W}}^n.$$

Microrollers: Fingering Instability



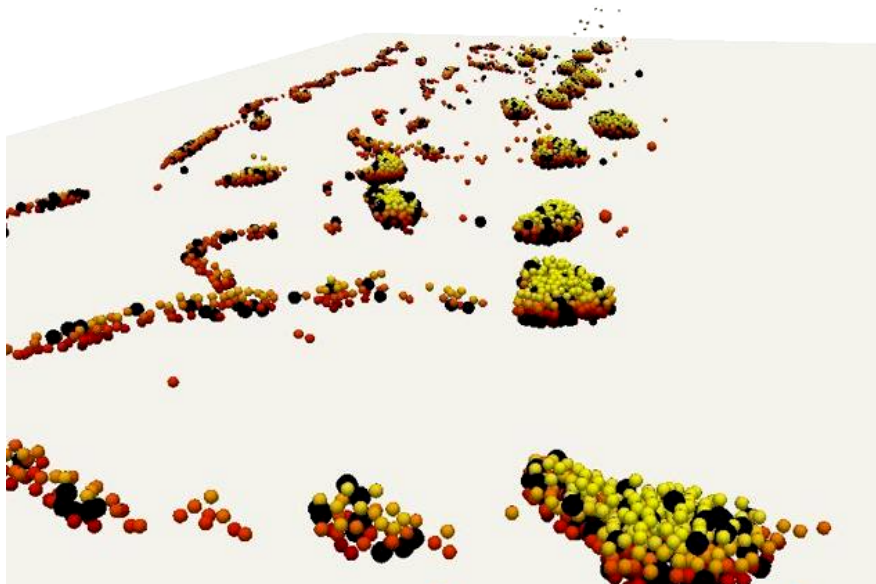
Experiments by Michelle Driscoll (lab of Paul Chaikin, NYU Physics, now at Northwestern), simulations by **Blaise Delmotte** [4, 3].

Role of Brownian Motion



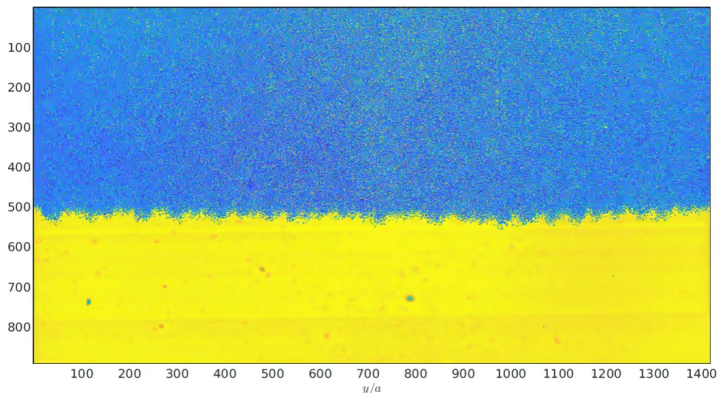
Simulations show that thermal fluctuations are quantitatively important because they set the **gravitational height**. [3].

Critters



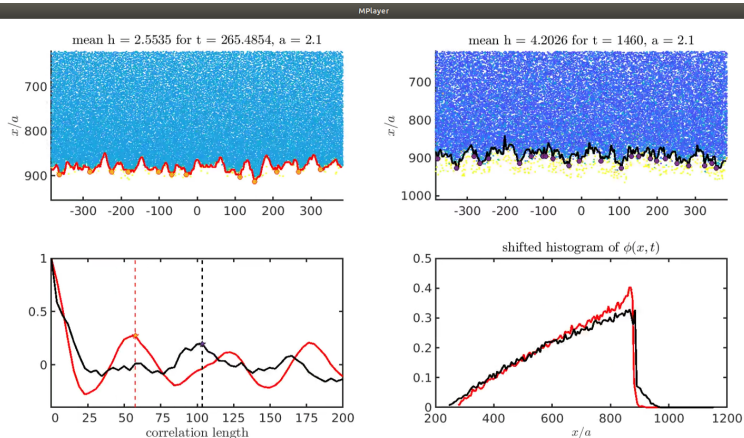
Sedimentation of colloidal monolayer

MPlayer

 $t = 274$ 

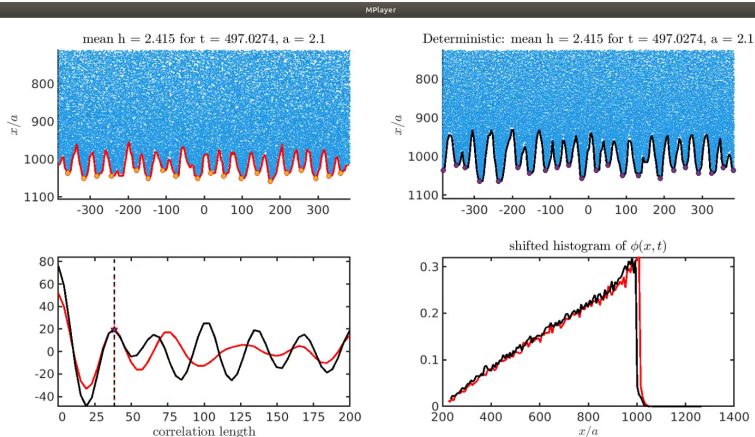
Experiments in lab of Paul Chaikin show that a sedimenting front roughens due to a sort of “instability”.

3D simulations of sedimentation



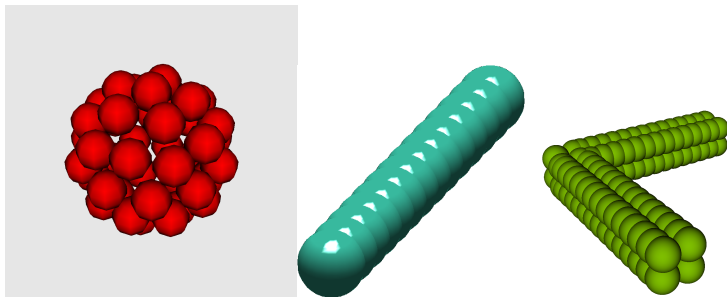
Simulations of **Brennan Sprinkle** show the gravitational height matters, but no precise explanation yet.

2D simulations of sedimentation



Quasi-2D simulations of **Brennan Sprinkle** show that Brownian motion in the plane don't matter that much.

Rigid MultiBlob Models



- The rigid body is discretized through a number of “**beads**” or “**blobs**” with hydrodynamic radius a .
- Standard is **stiff springs** but we want **rigid multiblobs**.
- Equivalent to a (**smartly!**) **regularized first-kind boundary integral formulation** [5].
- **We can efficiently simulate the driven and Brownian motion of the rigid multiblobs.**

Nonspherical Rigid Multiblobs

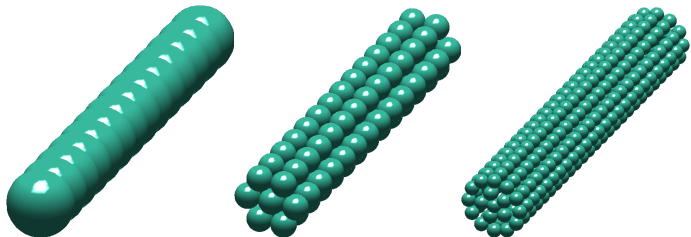


Figure: Rigid multiblob models of a rigid cylinder (rod) going from **minimally resolved** (left) to **well-resolved** (right).

Rigid MultiBlobs

- We add **rigidity forces** as Lagrange multipliers $\boldsymbol{\lambda} = \{\boldsymbol{\lambda}_1, \dots, \boldsymbol{\lambda}_n\}$ to constrain a group of blobs forming body p to move rigidly,

$$\sum_j \mathcal{M}_{ij} \boldsymbol{\lambda}_j = \mathbf{u}_p + \boldsymbol{\omega}_p \times (\mathbf{r}_i - \mathbf{q}_p) + \check{u}_p \quad (2)$$

$$\sum_{i \in \mathcal{B}_p} \boldsymbol{\lambda}_i = \mathbf{f}_p$$

$$\sum_{i \in \mathcal{B}_p} (\mathbf{r}_i - \mathbf{q}_p) \times \boldsymbol{\lambda}_i = \boldsymbol{\tau}_p.$$

where \mathbf{u} is the velocity of the tracking point \mathbf{q} , $\boldsymbol{\omega}$ is the angular velocity of the body around \mathbf{q} , \mathbf{f} is the total force applied on the body, $\boldsymbol{\tau}$ is the total torque applied to the body about point \mathbf{q} , and \mathbf{r}_i is the position of blob i .

- This can be a **very large linear system** for suspensions of many bodies discretized with many blobs:
Use **iterative solvers** with a **good preconditioner**.

Suspensions of Rigid Bodies

- In matrix notation we have a **saddle-point** linear system of equations for the rigidity forces λ and unknown motion $\mathbf{U} = (\mathbf{u}, \omega)$,

$$\begin{bmatrix} \mathcal{M} & -\mathcal{K} \\ \mathcal{K}^T & \mathbf{0} \end{bmatrix} \begin{bmatrix} \lambda \\ \mathbf{U} \end{bmatrix} = \begin{bmatrix} \check{\mathbf{u}} \\ \mathbf{F} \end{bmatrix}, \quad (3)$$

where $\mathbf{F} = (\mathbf{f}, \boldsymbol{\tau})$ are the applied forces and torques.

$$\text{Solution } \mathbf{U} = \mathcal{N}\mathbf{F} - (\mathcal{N}\mathcal{K}^T\mathcal{M}^{-1})\check{\mathbf{u}}$$

gives the **multiblob mobility matrix** [sorry for change of notation of letter \mathcal{N}]

$$\mathcal{N} = (\mathcal{K}^T\mathcal{M}^{-1}\mathcal{K})^{-1} \quad (4)$$

- The inverse of the mobility matrix is called the **resistance matrix**, $\mathcal{R} = \mathcal{N}^{-1} = \mathcal{K}^T\mathcal{M}^{-1}\mathcal{K}$.
- The **surface velocity** $\check{\mathbf{u}}$ can be used to model **active slip** or to generate **Brownian velocities** [1].

Lubrication for spherical colloids

- Use **Stokesian Dynamics** approach introduced by Brady to account for the strong lubrication for thin gaps by adding lubrication forces:

$$\begin{pmatrix} \mathcal{M} & -\mathcal{K} \\ \mathcal{K}^T & \mathbf{\Delta}_{MB} \end{pmatrix} \begin{pmatrix} \boldsymbol{\lambda} \\ \mathbf{U} \end{pmatrix} = \begin{pmatrix} -\check{\mathbf{u}} \\ \mathbf{F} \end{pmatrix}, \quad (5)$$

- $\mathbf{\Delta}_{MB}$ is a **lubrication correction to the resistance matrix** formed by adding **pairwise** contributions for each pair of nearby surfaces (either particle-particle or particle-wall) — can be computed semi-analytically or tabulated by using an expensive but accurate reference method (e.g., boundary integral).
- Lubrication-corrected mobility matrix

$$\overline{\mathcal{N}} = [\mathcal{N}^{-1} + \mathbf{\Delta}_{MB}]^{-1} = \mathcal{N} \cdot [\mathbf{I} + \mathbf{\Delta}_{MB} \cdot \mathcal{N}]^{-1}.$$

- One can even use a single blob per sphere (**minimally-resolved**) by adding rotation/torque to the RPY tensor, and setting $\mathcal{K} = \mathbf{I}$.

Generating Brownian Displacements $\sim \mathcal{N}^{\frac{1}{2}} \mathbf{W}$

- Assume that we knew how to efficiently generate Brownian blob velocities $\mathcal{M}^{\frac{1}{2}} \mathbf{W}$.
- Key idea:** Solve the mobility problem with random slip $\check{\mathbf{u}}$,

$$\begin{bmatrix} \mathcal{M} & -\mathcal{K} \\ -\mathcal{K}^T & \mathbf{0} \end{bmatrix} \begin{bmatrix} \lambda \\ \mathbf{U} \end{bmatrix} = - \begin{bmatrix} \check{\mathbf{u}} = (2k_B T)^{1/2} \mathcal{M}^{\frac{1}{2}} \mathbf{W} \\ \mathbf{F} \end{bmatrix}, \quad (6)$$

$$\mathbf{U} = \mathcal{N} \mathbf{F} + (2k_B T)^{\frac{1}{2}} \mathcal{N} \mathcal{K}^T \mathcal{M}^{-1} \mathcal{M}^{\frac{1}{2}} \mathbf{W} = \mathcal{N} \mathbf{F} + (2k_B T)^{\frac{1}{2}} \mathcal{N}^{\frac{1}{2}} \mathbf{W}.$$

which defines a $\mathcal{N}^{\frac{1}{2}} = \mathcal{N} \mathcal{K}^T \mathcal{M}^{-1} \mathcal{M}^{\frac{1}{2}}$:

$$\mathcal{N}^{\frac{1}{2}} \left(\mathcal{N}^{\frac{1}{2}} \right)^\dagger = \mathcal{N} (\mathcal{K}^T \mathcal{M}^{-1} \mathcal{K}) \mathcal{N} = \mathcal{N} \mathcal{N}^{-1} \mathcal{N} = \mathcal{N}.$$

Random Traction Euler-Maruyama

One can use the RFD idea to make more efficient temporal integrators for Brownian rigid multiblobs [1], such as the following **Euler scheme**:

- 1 Solve a mobility problem with a **random force+torque**:

$$\begin{bmatrix} \mathcal{M} & -\mathcal{K} \\ -\mathcal{K}^T & \mathbf{0} \end{bmatrix}^n \begin{bmatrix} \boldsymbol{\lambda}^{RFD} \\ \mathbf{U}^{RFD} \end{bmatrix} = \begin{bmatrix} \mathbf{0} \\ -\widetilde{\mathbf{W}} \end{bmatrix}. \quad (7)$$

- 2 Compute **random finite differences**:

$$\begin{aligned} \mathbf{F}^{RFD} &= \frac{k_B T}{\delta} \left(\mathcal{K}^T \left(\mathbf{Q}^n + \delta \widetilde{\mathbf{W}} \right) - (\mathcal{K}^n)^T \right) \boldsymbol{\lambda}^{RFD} \\ \ddot{\mathbf{u}}^{RFD} &= \frac{k_B T}{\delta} \left(\mathcal{M} \left(\mathbf{Q}^n + \delta \widetilde{\mathbf{W}} \right) - \mathcal{M}^n \right) \boldsymbol{\lambda}^{RFD} + \\ &\quad - \frac{k_B T}{\delta} \left(\mathcal{K} \left(\mathbf{Q}^n + \delta \widetilde{\mathbf{W}} \right) - \mathcal{K}^n \right) \mathbf{U}^{RFD}. \end{aligned}$$

Random Traction EM contd.

- 1 Compute **correlated random slip**:

$$\check{\mathbf{u}}^n = \left(\frac{2k_B T}{\Delta t} \right)^{1/2} (\mathcal{M}^n)^{\frac{1}{2}} \mathbf{W}^n$$

- 2 Solve the saddle-point system:

$$\begin{bmatrix} \mathcal{M} & -\mathcal{K} \\ -\mathcal{K}^T & \mathbf{0} \end{bmatrix}^n \begin{bmatrix} \boldsymbol{\lambda}^n \\ \mathbf{U}^n \end{bmatrix} = - \begin{bmatrix} \check{\mathbf{u}}^n + \check{\mathbf{u}}^{RFD} \\ \mathbf{F}^n - \mathbf{F}^{RFD} \end{bmatrix}. \quad (8)$$

- 3 Move the particles (rotate for orientation)

$$\mathbf{Q}^{n+1} = \mathbf{Q}^n + \Delta t \mathbf{U}^n.$$

Fluctuating Hydrodynamics

We consider a rigid body Ω immersed in a fluctuating fluid. In the fluid domain, we have the **fluctuating Stokes equation**

$$\begin{aligned}\rho \partial_t \mathbf{v} + \nabla \pi &= \eta \nabla^2 \mathbf{v} + (2k_B T \eta)^{\frac{1}{2}} \nabla \cdot \mathcal{Z} \\ \nabla \cdot \mathbf{v} &= 0,\end{aligned}$$

with **no-slip BCs** on the bottom wall, and the fluid stress tensor

$$\boldsymbol{\sigma} = -\pi \mathbf{I} + \eta (\nabla \mathbf{v} + \nabla^T \mathbf{v}) + (2k_B T \eta)^{\frac{1}{2}} \mathcal{Z} \quad (9)$$

consists of the usual **viscous stress** as well as a **stochastic stress** modeled by a symmetric **white-noise** tensor $\mathcal{Z}(\mathbf{r}, t)$, i.e., a Gaussian random field with mean zero and covariance

$$\langle \mathcal{Z}_{ij}(\mathbf{r}, t) \mathcal{Z}_{kl}(\mathbf{r}', t') \rangle = (\delta_{ik} \delta_{jl} + \delta_{il} \delta_{jk}) \delta(t - t') \delta(\mathbf{r} - \mathbf{r}').$$

Fluid-Body Coupling

At the fluid-body interface the **no-slip boundary condition** is assumed to apply,

$$\mathbf{v}(\mathbf{q}) = \mathbf{u} + \boldsymbol{\omega} \times \mathbf{q} - \check{\mathbf{u}}(\mathbf{q}) \text{ for all } \mathbf{q} \in \partial\Omega, \quad (10)$$

with the **inertial body dynamics**

$$m \frac{d\mathbf{u}}{dt} = \mathbf{F} - \int_{\partial\Omega} \boldsymbol{\lambda}(\mathbf{q}) d\mathbf{q}, \quad (11)$$

$$\mathbf{I} \frac{d\boldsymbol{\omega}}{dt} = \boldsymbol{\tau} - \int_{\partial\Omega} [\mathbf{q} \times \boldsymbol{\lambda}(\mathbf{q})] d\mathbf{q} \quad (12)$$

where $\boldsymbol{\lambda}(\mathbf{q})$ is the normal component of the stress on the outside of the surface of the body, i.e., the **traction**

$$\boldsymbol{\lambda}(\mathbf{q}) = \boldsymbol{\sigma} \cdot \mathbf{n}(\mathbf{q}).$$

To model activity we can add **active slip** $\check{\mathbf{u}}$ due to active boundary layers, or consider external forces/torques.

Mobility Problem

From linearity, the rigid-body motion is defined by a linear mapping $\mathbf{U} = \mathcal{N}\mathbf{F}$ via the deterministic **mobility problem**:

$$\begin{aligned} \nabla \pi &= \eta \nabla^2 \mathbf{v} \quad \text{and} \quad \nabla \cdot \mathbf{v} = 0 \quad + \text{BCs} \\ \mathbf{v}(\mathbf{q}) &= \mathbf{u} + \boldsymbol{\omega} \times \mathbf{q} - \ddot{\mathbf{u}}(\mathbf{q}) \quad \text{for all } \mathbf{q} \in \partial\Omega, \end{aligned} \quad (13)$$

With **force and torque balance**

$$\int_{\partial\Omega} \boldsymbol{\lambda}(\mathbf{q}) d\mathbf{q} = \mathbf{F} \quad \text{and} \quad \int_{\partial\Omega} [\mathbf{q} \times \boldsymbol{\lambda}(\mathbf{q})] d\mathbf{q} = \boldsymbol{\tau}, \quad (14)$$

where $\boldsymbol{\lambda}(\mathbf{q}) = \boldsymbol{\sigma} \cdot \mathbf{n}(\mathbf{q})$ with

$$\boldsymbol{\sigma} = -\pi \mathbf{I} + \eta (\nabla \mathbf{v} + \nabla^T \mathbf{v}). \quad (15)$$

Overdamped Brownian Dynamics

- Consider a suspension of N_b rigid bodies with **configuration** $\mathbf{Q} = \{\mathbf{q}, \boldsymbol{\theta}\}$ consisting of **positions and orientations** (described using **quaternions**) immersed in a Stokes fluid.
- By eliminating the fluid from the equations in the **overdamped limit** (infinite Schmidt number) we get the equations of **Brownian Dynamics**

$$\frac{d\mathbf{Q}(t)}{dt} = \mathbf{U} = \mathcal{N}\mathbf{F} + (2k_B T \mathcal{N})^{\frac{1}{2}} \boldsymbol{\mathcal{W}}(t) + (k_B T) \partial_{\mathbf{Q}} \cdot \mathcal{N},$$

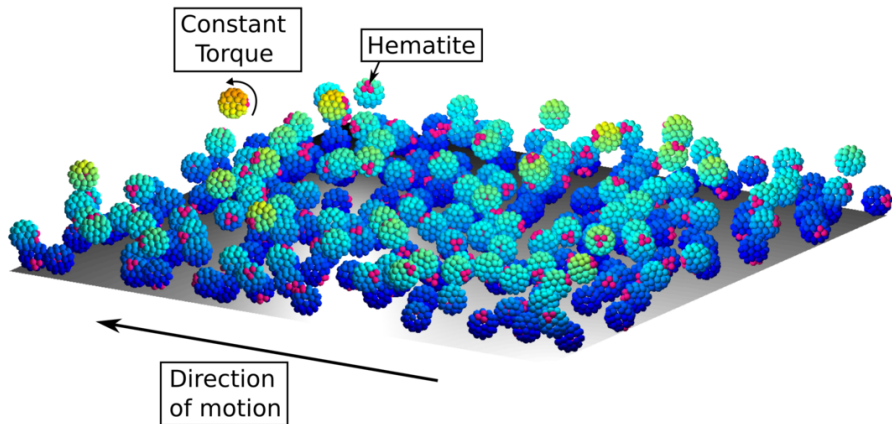
where $\mathcal{N}(\mathbf{Q})$ is the **body mobility matrix**, with “square root” given by **fluctuation-dissipation balance**

$$\mathcal{N}^{\frac{1}{2}} \left(\mathcal{N}^{\frac{1}{2}} \right)^T = \mathcal{N}.$$

$\mathbf{U} = \{\mathbf{u}, \boldsymbol{\omega}\}$ collects the **linear and angular velocities**

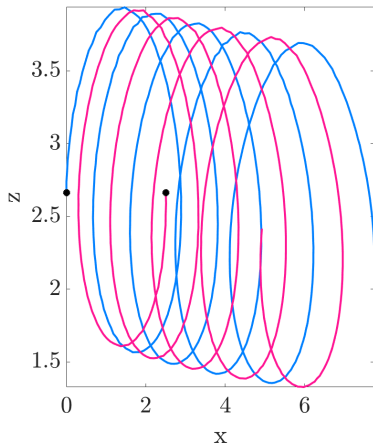
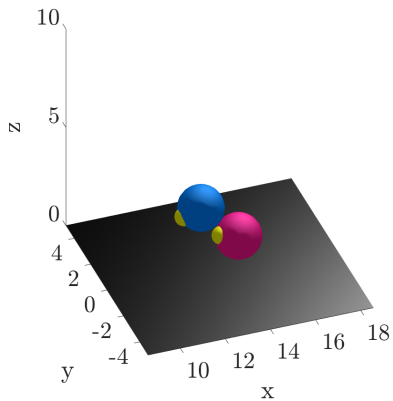
$\mathbf{F}(\mathbf{Q}) = \{\mathbf{f}, \boldsymbol{\tau}\}$ collects the **applied forces and torques**.

Microrollers: Uniform Suspension

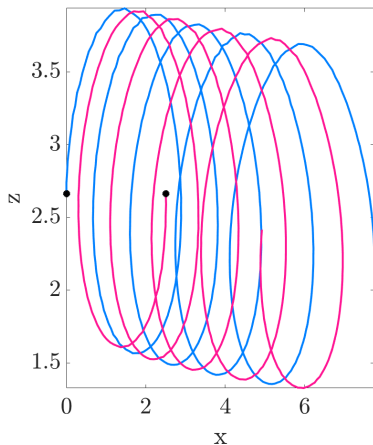
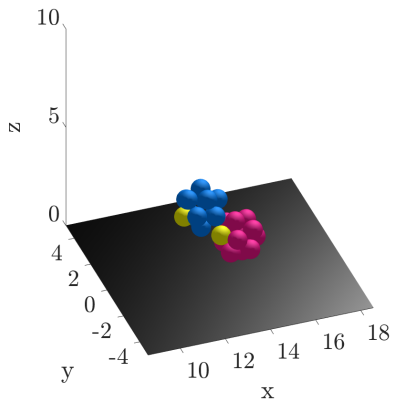


Simulations by **Brennan Sprinkle**+Blaise Delmotte [1] of a uniform suspension of microrollers at packing fraction $\phi = 0.4$ (GIF). Compare to experiments (AVI) by **Michelle Driscoll**.

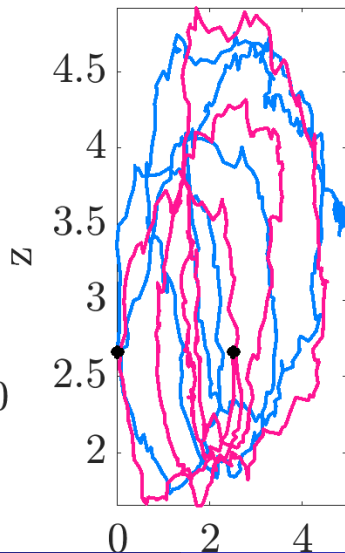
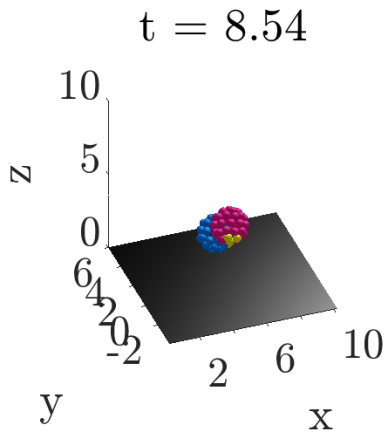
Example 1: Bound Roller Dimer

 $t = 9.8$ 

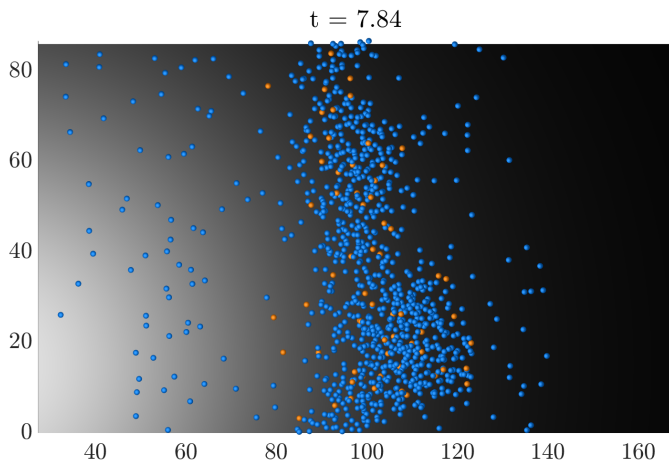
Example 1: Deterministic 12 blobs

 $t = 9.8$ 

Example 1: Bound Roller Dimer



Example 2: Formation of Critters



References



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Software available at <https://github.com/stochasticHydroTools/RigidMultiblobsWall>.



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