

Temporal integrators for Langevin equations arising in Fluctuating Hydrodynamics

Steven Delong, CIMS
&

Eric Vanden-Eijnden and Aleksandar Donev, CIMS

Courant Institute, New York University

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Light-Activated Diffusion/Osmophoresis

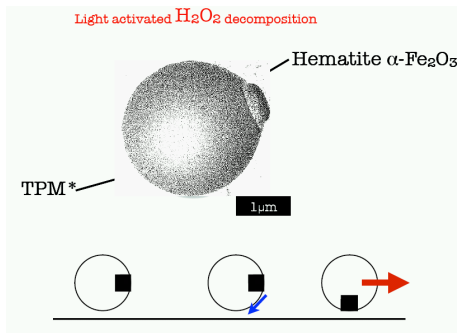


Figure: From Jeremie Palacci, Paul Chaikin lab (NYU Physics) [1]

Non-Spherical Colloids near Boundaries

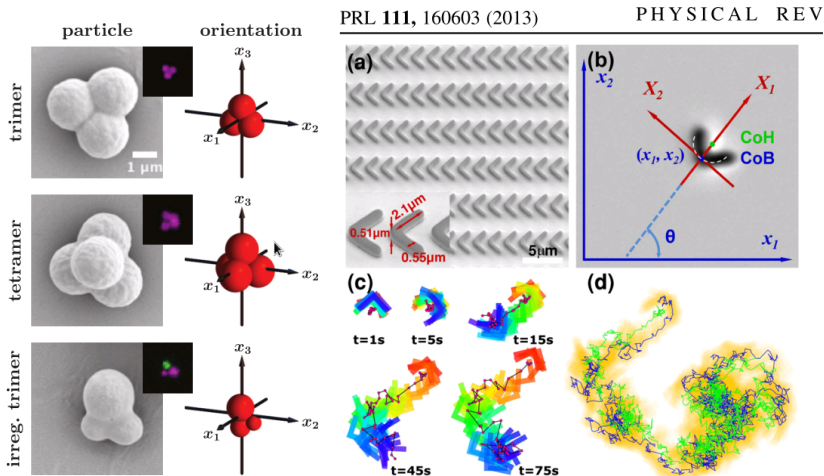


Figure: (Left) Cross-linked spheres; Kraft et al. [2]. (Right) Lithographed boomerangs; Chakrabarty et al. [3].

Immersed Spherical Particles

- For simplicity, we first consider the dynamics of several immersed spherical particles with positions given by $\mathbf{x} = \{\mathbf{x}_i\}$.
- For now we will ignore the orientation of the particles.
- We assume that inertia evolves much faster than particle position, and consider the **overdamped** equations for particle motion.

Brownian Dynamics

- We write the overdamped Langevin equation for the Brownian dynamics of the particles.

$$\partial_t \mathbf{x} = -\mathbf{M} \partial_{\mathbf{x}} U(\mathbf{x}) + \sqrt{2k_B T} \mathbf{M}_{\frac{1}{2}} \boldsymbol{\mathcal{W}} + k_B T \partial_{\mathbf{x}} \cdot (\mathbf{M})$$

- \mathbf{M} is the mobility operator, $\mathbf{M}_{\frac{1}{2}} \mathbf{M}_{\frac{1}{2}}^* = \mathbf{M}$, U is the potential energy, $k_B T$ is the temperature, and $\boldsymbol{\mathcal{W}}$ is a vector of independent white noise.
- The thermal drift term $\partial_{\mathbf{x}} \cdot (\mathbf{M})$ can be written as $\sum_{j=1}^N \partial_{x_j} M_{ij}(\mathbf{x})$.
- These equations of motion are time reversible w.r.t the Gibbs Boltzmann distribution

$$\rho_{eq}(\mathbf{x}) = Z^{-1} \exp(-U(\mathbf{x})/k_B T)$$

Difficulties in Integration

- The mobility \mathbf{M} includes all particle interaction and boundary effects. Constructing this operator is non trivial (Aleksandar Donev will talk more on this).
- Calculating just the application of \mathbf{M} and $\mathbf{M}_{\frac{1}{2}}$ is an expensive process.
- More complicated objects such as $\partial_{\mathbf{x}} \cdot \mathbf{M}$ and \mathbf{M}^{-1} may not be directly computable.
- This talk focuses on the thermal drift. Our goal is to develop schemes that use only application of \mathbf{M} and $\mathbf{M}_{\frac{1}{2}}$.

Fixman's Method

- A standard way to handle the thermal drift is with Fixman's midpoint method

$$\begin{aligned} \mathbf{x}^{n+\frac{1}{2}} &= \mathbf{x}^n - \frac{\Delta t}{2} \mathbf{M}^n \partial_{\mathbf{x}} U(\mathbf{x}^n) + \sqrt{\frac{k_B T \Delta t}{2}} \mathbf{M}_{\frac{1}{2}}^n \mathbf{W}^n \\ \mathbf{x}^{n+1} &= \mathbf{x}^n - \Delta t \mathbf{M}^{n+\frac{1}{2}} \partial_{\mathbf{x}} U(\mathbf{x}^{n+\frac{1}{2}}) \\ &\quad + \sqrt{2k_B T \Delta t} \mathbf{M}^{n+\frac{1}{2}} \left(\mathbf{M}_{\frac{1}{2}}^n \right)^{-1} \mathbf{W}^n, \end{aligned}$$

where \mathbf{W}^n is a vector of independent $\mathcal{N}(0, 1)$ variables.

- While this achieves the correct drift, it requires knowledge of $\left(\mathbf{M}_{\frac{1}{2}}^n \right)^{-1} = \mathbf{M}^{-1} \mathbf{M}_{\frac{1}{2}}$, which we recall is not easy to obtain.

Random Finite Difference

- We can also generate the drift with $(M(x^n + \delta \widetilde{W}) - M^n) \widetilde{W} / \delta$, where \widetilde{W} is a vector of independent $\mathcal{N}(0, 1)$ variables.
- This leads to the Random Finite Difference (RFD) scheme

$$\begin{aligned} x^{n+1} = & x^n - \Delta t M^n \partial_x U + \sqrt{2k_B T \Delta t} M_{\frac{1}{2}}^n W^n \\ & + \frac{k_B T}{\delta} (M(x^n + \delta \widetilde{W}) - M^n) \widetilde{W}. \end{aligned}$$

- This scheme makes an error of order δ . Fixman takes $\delta \sim \Delta t$, but this is not necessary.
- We need to apply M^n , $M_{\frac{1}{2}}^n$, and $M(x^n + \delta \widetilde{W})$, but *not* $(M_{\frac{1}{2}}^n)^{-1}$.

Bodies with rotation

- We can extend our work to simulate bodies with **rotational DOFs** by formulating the appropriate Langevin equation and using a RFD approach to for temporal integration.
- For simplicity, first we consider a single body with only rotational degrees of freedom.
- Orientation is an element of $SO(3)$ so we need to parameterize it: we use **normalized quaternion** (point on the unit 4-sphere)

$$\boldsymbol{\theta} \in \mathbb{R}^4, \quad \|\boldsymbol{\theta}\|_2 = \boldsymbol{\theta} \cdot \boldsymbol{\theta} = 1.$$

- This offers several advantages over several other common approaches, such as rotation angles, rotation matrices, and Euler angles[4].

Quaternions

- Successive rotations can be accumulated by **quaternion multiplication**.
- In three dimensions, there exists a 4×3 matrix $\Psi(\theta)$ such that, given a conservative potential $U(\theta)$,

$$\dot{\theta} = \Psi\omega, \quad \tau = \Psi^T \partial_{\theta} U(\theta).$$

Here τ is the torque applied to the body, and ω is the angular velocity.

- One can also rotate a body by an oriented angle ϕ , denoted as

$$\theta^{n+1} = \text{Rotate}(\theta^n, \phi).$$

Equations for Rotation

- We assume now that we know the mobility tensor $M_{\omega\tau}$,

$$\omega = M_{\omega\tau}\tau.$$

- Given $M_{\omega\tau}$ and a potential $U(\theta)$, the Langevin Equation for orientation is

$$\begin{aligned} \partial_t \theta = & - \left(\Psi M_{\omega\tau} \Psi^T \right) \partial_\theta U + \sqrt{2k_B T} \Psi M_{\omega\tau}^{\frac{1}{2}} \mathcal{W} \\ & + k_B T \partial_\theta \cdot \left(\Psi M_{\omega\tau} \Psi^T \right). \end{aligned}$$

- This equation preserves the unit norm constraint and is time reversible w.r.t. the Gibbs Boltzmann distribution

$$P_{\text{eq}}(\theta) = Z^{-1} \exp(-U(\theta)/k_B T) \delta(\theta^T \theta - 1).$$

Random Finite Difference for Orientation

- We construct a RFD scheme for timestepping the quaternion orientation

$$\begin{aligned} \boldsymbol{\theta}^* &= \text{Rotate} \left(\boldsymbol{\theta}^n, \delta \widetilde{\boldsymbol{W}} \right) \\ \boldsymbol{\omega}^n &= - \left(\boldsymbol{M} \boldsymbol{\Psi}^T \frac{\partial U}{\partial \boldsymbol{\theta}} \right)^n + \left(\frac{2k_B T}{\Delta t} \boldsymbol{M}^n \right)^{\frac{1}{2}} \boldsymbol{W}^n \\ &\quad + \left(\frac{k_B T}{\delta} \right) (\boldsymbol{M}^* - \boldsymbol{M}^n) \widetilde{\boldsymbol{W}} \\ \boldsymbol{\theta}^{n+1} &= \text{Rotate} \left(\boldsymbol{\theta}^n, \boldsymbol{\omega}^n \Delta t \right). \end{aligned}$$

Algorithm with Translation

- To include translation, we introduce the matrix Ξ , letting $\mathbf{u} = \dot{\mathbf{q}}$ where \mathbf{q} is the location of the body.

$$\Xi = \begin{bmatrix} I & \mathbf{0} \\ \mathbf{0} & \Psi \end{bmatrix}, \quad [\dot{\mathbf{q}}, \dot{\boldsymbol{\theta}}]^T = \Xi [\mathbf{u}, \boldsymbol{\omega}]^T$$

- The equations of motion in this case are

$$\mathbf{v} = \frac{d\mathbf{x}}{dt} = -(\Xi \mathbf{N} \Xi^*) \partial_{\mathbf{x}} U + \sqrt{2k_B T} \Xi \mathbf{N}^{\frac{1}{2}} \boldsymbol{\mathcal{W}} + (k_B T) \partial_{\mathbf{x}} \cdot (\Xi \mathbf{N} \Xi^*),$$

where $\mathbf{x} = (\mathbf{q}, \boldsymbol{\theta})^T$, $\mathbf{v} = (\mathbf{u}, \dot{\boldsymbol{\theta}})^T$, and \mathbf{N} is the grand mobility tensor, such that given force \mathbf{F} on the body, we have.

$$[\mathbf{u}, \boldsymbol{\omega}]^T = \mathbf{N} [\mathbf{F}, \boldsymbol{\tau}]^T.$$

Random Finite Difference with Translation

- We can also write a RFD scheme that works with translation

$$\tilde{\mathbf{v}} = \tilde{\mathbf{W}}$$

$$\tilde{\mathbf{q}} = \mathbf{q}^n + \delta \tilde{\mathbf{u}}$$

$$\tilde{\boldsymbol{\theta}} = \text{Rotate}(\boldsymbol{\theta}^n, \delta \tilde{\boldsymbol{\omega}})$$

$$\mathbf{v}^n = - \left(\mathbf{N} \boldsymbol{\Xi}^T \partial_{\mathbf{x}} U \right)^n + \sqrt{\frac{2k_B T}{\Delta t}} \left(\mathbf{N}^{\frac{1}{2}} \right)^n \mathbf{W}^n + \frac{k_B T}{\delta} \left(\tilde{\mathbf{N}} - \mathbf{N}^n \right) \tilde{\mathbf{W}}$$

$$\mathbf{q}^{n+1} = \mathbf{q}^n + \Delta t \mathbf{u}^n$$

$$\boldsymbol{\theta}^{n+1} = \text{Rotate}(\boldsymbol{\theta}^n, \Delta t \boldsymbol{\omega}^n).$$

Computing Mobility

- There are different ways to obtain M (more in talk by Aleksandar Donev):
 - In unbounded domains we can just use the **Rotne-Prager-Yamakawa tensor** (RPY) (always SPD!).
 - In simple geometries such as a single wall we can use a generalization of RPY;
We use the analytical mobility for a **single wall** (Swan&Brady [5]).
 - In more general cases we can use a fluctuating **FEM/FVM fluid Stokes solver** combined with an immersed-boundary representation of the particles [6].
- In [6] we develop an RFD scheme that requires only one Stokes solve per time step for spherical particles.

Simulation: particles in a channel

- For the single particle case, we can write down the biased distribution that we expect when no drift is present (Euler Maruyama).

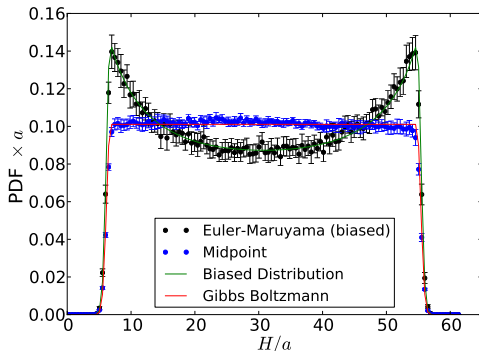


Figure: Distribution of a particle diffusing in a channel.

Importance of stochastic drift

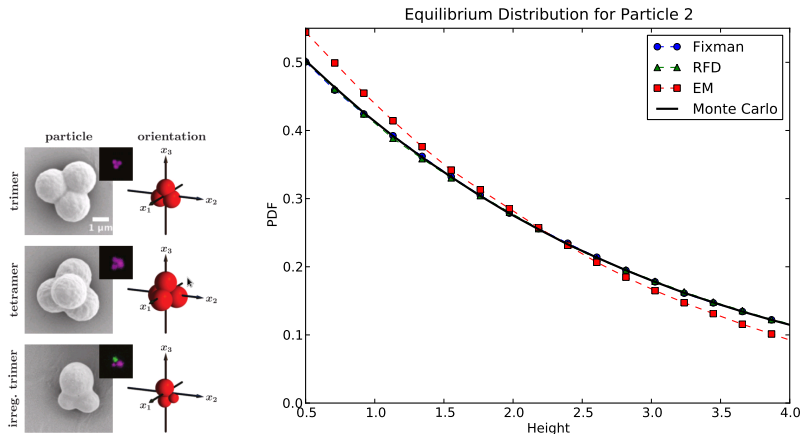


Figure: Height distribution of a vertex of a tetrahedron of spheres near a wall with one vertex tethered (**rotation only**).

Non-uniform Icosahedron

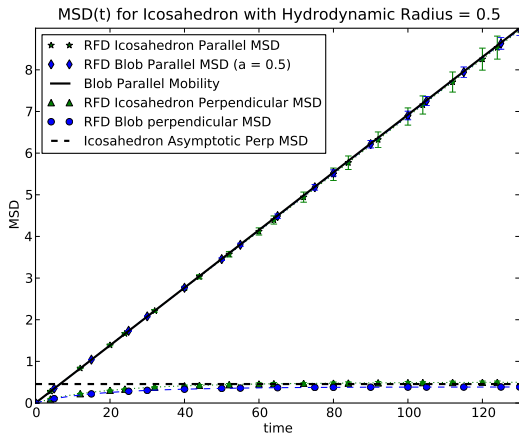
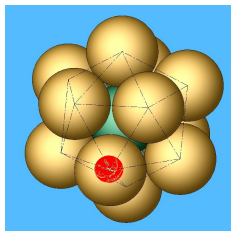


Figure: Translational Mean Square Displacement of a non-uniform Icosahedron.

Non-uniform Icosahedron

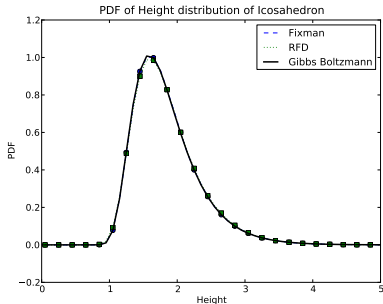
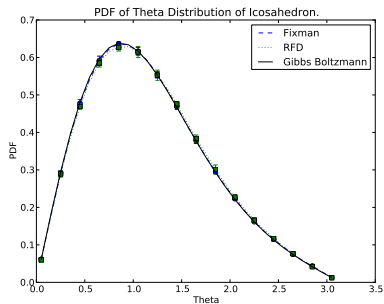


Figure: Height and Orientation distributions for a non-uniform icosahedron.

Conclusion

- We introduced the **Random Finite Difference (RFD)** scheme for timestepping the overdamped Langevin equations.
- This scheme requires only applications of \mathbf{M} and $\mathbf{M}_{\frac{1}{2}}$, and in particular we do **not** need to **invert the mobility**.
- We formulate an overdamped Langevin equation for angular degrees of freedom using **quaternions**.
- We can apply a specialized version of RFD to handle the thermal drift in the equations for orientation.

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