

Numerical methods

Recall the generalized Markovian description of Langevin dynamics:

$$\dot{x} = -M \cdot \frac{\partial H}{\partial x} + kT \frac{\partial \cdot M}{\partial x} + \sqrt{2kT} \cdot W(t)$$

$$\tilde{M} \tilde{M}^* = (M + M^*)/2$$

where $M(x)$ is the mobility tensor

$$M = \Gamma^{-1} \leftarrow \text{friction tensor}$$

$$D = (kT) M \leftarrow \text{diffusion tensor}$$

Remember that the spurious drift depends on the stochastic integral convention.

As a particular (important) example, consider Brownian dynamics for a collection of colloidal particles suspended in a fluid.

Denote $q = \{q_1, q_2, \dots, q_N\}$ the positions (configuration) of the particles, and $V(q)$ the interaction potential (effective).

Under additivity assumptions, mobility/friction is the Oseen tensor from hydrodynamics:

$$M_{ij} = \frac{1}{6\pi\eta a} \begin{cases} \frac{3}{4} \frac{a}{\Gamma_{ij}} \left(1 + \frac{\vec{r}_{ij} \otimes \vec{r}_{ij}}{\Gamma_{ij}^2}\right), & i \neq j \\ 1 & i = j \end{cases}$$

↑
particle
indexes

Our previous analysis of drift suggests that numerical methods can avoid the evaluation of $\nabla_q \cdot M(q)$ by using a Backward Euler method for the anti-Ito or "isothermal" SODE:

$$\dot{q} = -M \cdot \nabla_q V + \sqrt{2kT} \overset{\longleftarrow}{\square} M \cdot W(t)$$

To avoid an implicit method, we can use an explicit predictor-corrector method

BAD IDEA!

Explicit

"Backward"

Euler

NOT consistent

$$\tilde{q}_t = q_t + \left[M(\nabla_q V)_{t+\Delta t} + \sqrt{2\Delta t(kT)} \tilde{M} W \right]_{t+\Delta t}$$

predictor

$q_{t+\Delta t}$

$$q_{t+\Delta t} = q_t + \Delta t M(\nabla_q V) + \sqrt{2\Delta t(kT)} \tilde{M}(\tilde{q}) \cdot \tilde{W}$$

In the Fixman algorithm one can evaluate the term $M \nabla_q V$ at midpoint to improve the deterministic order of accuracy to second-order, though the formal order of accuracy (weak sense) remains one.

How to construct \tilde{M} from M ?

All that matters at this stage is

$$\text{that } \tilde{M} \tilde{M}^* = M$$

- { ① $\tilde{M} = \tilde{M}^* = M^{1/2}$ (matrix square root)
- ② $\tilde{M} = \text{lower triangular square}$ (Cholesky factorization)
- ③ $\tilde{M} = \text{non-square}$ (physics?)

So far we proposed the algorithm based on "Backward Euler", which we can condense to leading order as:

$$q(t+\Delta t) = q^{n+1} = q^n + M(q^n) \cdot \frac{\partial V}{\partial q^n} \Delta t +$$

$$\sqrt{2\Delta t(h)} \cdot \tilde{M} \left[q^n + \sqrt{2\Delta t(h)} \tilde{M}(q^n) \cdot \tilde{W}^n \right] \cdot \tilde{W}$$

We already checked that this is correct in one dimension, but how about the multivariable case?

Let's check the drift due to the multiplicative noise explicitly:

"Spurious" drift:

$$\Delta q_k = \sqrt{2\Delta t(kT)} \frac{\partial \tilde{M}_{kl}}{\partial q_i} \sqrt{2\Delta t(kT)} (\tilde{M} \tilde{w})_i^l$$

$$\Rightarrow \frac{\langle \Delta q_k \rangle}{2\Delta t(kT)} = \frac{\partial \tilde{M}_{kl}}{\partial q_i} \tilde{M}_{ij} \langle \tilde{w}_j \tilde{w}_l \rangle =$$

$$= \frac{\partial \tilde{M}_{kl}}{\partial q_i} \tilde{M}_{ij} \delta_{jl} = \frac{\partial \tilde{M}_{kj}}{\partial q_i} \tilde{M}_{ij}$$

What we want to get is

$$\frac{1}{2} \left(\frac{\partial M}{\partial q_i} \cdot M \right)_k = \frac{1}{2} \frac{\partial}{\partial q_i} M_{ik}$$

$$\frac{1}{2} \frac{\partial}{\partial q_i} M_{ik} = \frac{1}{2} \frac{\partial}{\partial q_i} (\tilde{M}_{ij} \tilde{M}_{kj}) =$$

$$= \frac{1}{2} \left[\frac{\partial \tilde{M}_{ij}}{\partial q_i} \tilde{M}_{kj} + \tilde{M}_{ij} \frac{\partial \tilde{M}_{kj}}{\partial q_i} \right]$$

↗?

Compare this to previous

$$\tilde{M}_{ij} \frac{\partial \tilde{M}_{kj}}{\partial q_i}$$

Except for a single variable, this does not seem to match.

So our "backward explicit Euler" algorithm is actually not consistent.

Do not use it!

Instead, rewrite the Langevin equation

$$\dot{q} = -M \cdot \frac{\partial V}{\partial q} + \sqrt{2kT} \square M \cdot N \cdot W(t)$$

where

$$NN^* = M^{-1} = \Gamma \leftarrow \text{symmetric}$$

This is equivalent to what we wrote before since all that matters is the noise covariance (think FPE):

$$(MN)(MN)^* = MNN^*M^* = M$$

At first glance this requires both M and M^{-1} , which complicates things..

Fixman proposed the following
mid-point explicit predictor-corrector alg:

$$\tilde{q}^{n+1} = q^n + M(q^n) \cdot \frac{\partial V}{\partial q^n} \Delta t + \sqrt{2 \Delta t(h_T)} M \cdot N \cdot W$$

$$q^{n+1/2} = (q^n + \tilde{q}^{n+1}) / 2 \quad (\text{mid point})$$

$$\left\{ \begin{array}{l} q^{n+1} \\ = q^n + M(q^{n+1/2}) \cdot \frac{\partial V}{\partial q^{n+1/2}} \Delta t + \end{array} \right.$$

$$\sqrt{2 \Delta t(h_T)} \cdot M(q^{n+1/2}) \cdot \underbrace{N \cdot W}_{\text{same as predictor!}}$$

Is this algorithm now consistent?

To leading order, we are doing

$$q^{n+1} = q^n + M(q^n) \cdot \frac{\partial V}{\partial q^n} \Delta t +$$

$$\sqrt{2 \Delta t(h\tau)} \cdot M \left[q^n + \sqrt{\frac{\Delta t(h\tau)}{2}} M(q^n) \cdot N(q^n) \tilde{W} \right] \\ N(q^n) \tilde{W}^n$$

Now the extra drift is

$$\Delta q_k = \sqrt{2 \Delta t(h\tau)} \frac{\partial M_{kl}}{\partial q_i} \cdot \sqrt{\frac{\Delta t(h\tau)}{2}} (M N \tilde{W})_i (N \tilde{W})_l$$

$$= \Delta t(h\tau) \frac{\partial M_{hl}}{\partial q_i} M_{ij} \stackrel{\sim}{N_{jm}} W_m \stackrel{\sim}{N_{ln}} W_n$$

$$\frac{\langle \Delta q_h \rangle}{(kT) \Delta t} = \frac{\partial M_{kl}}{\partial q_i} \underbrace{M_{ij}(N_{jm} N_{lm})}_{(NN^*)_{jl}} =$$

$$= \frac{\partial M_{kl}}{\partial q_i} M_{ij} (M^{-1})_{jl} = \frac{\partial M_{kl}}{\partial q_i} (\delta_{il})$$

$$= \frac{\partial M_{ki}}{\partial q_i} = \frac{\partial M_{ik}}{\partial q_i} = \left(\frac{\partial \cdot M}{\partial q} \right)_k$$

So indeed the extra drift is

$$kT \left(\frac{\partial \cdot M}{\partial q} \right) \text{ as it should be.}$$

But maybe there are other ways also...

Recall that for additive noise (no x dependence in stochastic term) or for Langevin dynamics there is no problem with extra drift terms, and the "mobility" matrix M is rather simple:

$$H = \frac{m u^2}{2} + V(q) = \frac{P^2}{2m} + V(q)$$

$$\Rightarrow \frac{\partial H}{\partial z} = \begin{bmatrix} \partial V / \partial q \\ \frac{P}{m} = u \end{bmatrix}$$

where $z = (q, P)$

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$$\dot{z} = \begin{bmatrix} \dot{q} \\ \dot{p} \end{bmatrix} = \begin{array}{c} \downarrow \text{reversible} \\ \text{Hamiltonian} \end{array} \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \begin{array}{c} \downarrow \frac{\partial H}{\partial z} \\ \downarrow \frac{\partial V}{\partial q} \end{array} \begin{bmatrix} \dot{V} \\ \dot{u} \end{bmatrix}$$

\uparrow
skew-adjoint Louisville operator

$$- \begin{bmatrix} 0 & 0 \\ 0 & f \end{bmatrix} \begin{bmatrix} \dot{V} \\ \dot{u} \end{bmatrix} \leftarrow \frac{\partial H}{\partial z}$$

"irreversible" \uparrow self-adjoint friction operator

$$+ \begin{bmatrix} 0 \\ \sqrt{2kT} g \end{bmatrix} \xrightarrow{\quad} W(t)$$

fluctuations \tilde{M} , with $\tilde{M}\tilde{M}^* = M$

Recall the FPE for Langevin dynamics:

$$\frac{\partial P}{\partial t} = - \chi P = - (\chi_1 + \chi_2 + \chi_3 + \chi_4) P$$

$$\left\{ \begin{array}{l} \chi_1 = \frac{P}{m} \cdot \frac{\partial}{\partial q} \quad (\text{advection}) \\ \quad \quad \quad \text{or streaming} \\ \chi_2 = - \frac{\partial V}{\partial q} \cdot \frac{\partial}{\partial p} \quad (\text{constant force acceleration}) \\ \chi_3 = - \frac{f}{m} \frac{\partial}{\partial p} P \quad (\text{constant friction}) \\ \chi_4 = - \gamma k T \frac{\partial^2}{\partial p^2} \quad (\text{noise}) \quad \text{In general} \\ \quad \quad \quad f = f(q) \end{array} \right.$$

Recall that for Hamiltonian dynamics (MD) we desired symplectic integrators and constructed them using operator splitting and the Trotter factorization

$$\begin{aligned}
 P(\Delta t) &= e^{-\mathcal{X}\Delta t} P(0) && \text{Strang splitting} \\
 &\approx e^{-\mathcal{X}_A \Delta t/2} e^{-\mathcal{X}_B \Delta t} e^{-\mathcal{X}_A \Delta t/2} P(0)
 \end{aligned}$$

↓
 most expensive
 in middle?

where

$$\mathcal{X} = \mathcal{X}_A + \mathcal{X}_B$$

For this to work as written the two pieces \mathcal{L}_x and \mathcal{L}_p should ideally be exactly solvable so that

$e^{-\mathcal{L}_x \Delta t}$ and $e^{-\mathcal{L}_p \Delta t}$ are exact.

This suggests the splitting

$$\left\{ \begin{array}{l} \mathcal{L}_x = \mathcal{L}_1 \quad (\text{streaming advection}) \\ \mathcal{L}_p = \mathcal{L}_2 + \mathcal{L}_3 + \mathcal{L}_4 \quad (\text{Ornstein-Uhlenbeck momentum update}) \end{array} \right.$$

We already showed how to do this.

Explicitly :

$$\left\{ \begin{array}{l} q^{n+1} = q^n + \frac{P_m^n}{m} \Delta t \quad (\text{this is } \mathcal{Z} \Delta t) \\ P^{n+1} = P^n e^{-\lambda \Delta t} + \frac{m}{\gamma} F(q^n) (1 - e^{-\lambda \Delta t}) \\ \qquad \qquad \qquad + \sqrt{m k T (1 - e^{-2\lambda \Delta t})} \cdot \overset{\text{random}}{\mathcal{N}(0,1)} \end{array} \right.$$

where $\lambda = \gamma/m$

normal variate

This is the so-called quasi-symplectic
Stochastic Verlet algorithm.

Another option is to do the splitting

$$\left\{ \begin{array}{l} \mathcal{L}_2 = \mathcal{L}_1 + \mathcal{L}_2 \quad (\text{Hamiltonian dynamics}) \\ \quad \nearrow \text{use any } \underline{\text{second-order}} \text{ symplectic integrator} \\ \mathcal{L}_B = \mathcal{L}_3 + \mathcal{L}_4 \quad (\text{irreversible dynamics}) \\ \quad \nearrow \text{use previous formula with } F \equiv 0 \end{array} \right.$$

Both splittings give quasi-symplectic integrators (Jacobian not unity but constant), but it seems there is no common agreement on which one is "best".