

Using exotic aromatic forests to construct order two scheme for the invariant measure sampling of Langevin dynamics with variable diffusion

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Abstract

Exotic aromatic forests, an extension of aromatic forests into the stochastic context, play a crucial role in generating order conditions for invariant measure sampling and in studying the algebraic properties of stochastic integrators. This work demonstrates practical benefits through a new method, a generalization of the Leimkuhler-Matthews method, which achieves order two for overdamped Langevin dynamics with variable diffusion.

1. Stochastic differential equations

Let ϕ denote a test function $\mathbb{R}^d \rightarrow \mathbb{R}$. Consider a stochastic differential equation with multiplicative noise with smooth vector field $F: \mathbb{R}^d \rightarrow \mathbb{R}^d$ and smooth diffusion $D: \mathbb{R}^d \rightarrow \mathbb{R}^{d \times d}$:

$$dX = F(X)dt + \sigma D(X)dW, \quad X(t) \in \mathbb{R}^d,$$

where $W(t) \in \mathbb{R}^d$ is a standard Wiener process. The weak Taylor expansion of the solution $X(t)$ is given by

$$\mathbb{E}[\phi(X(h))] = \phi(X_0) + h\mathcal{L}\phi(X_0) + \dots + \frac{h^k}{k!}\mathcal{L}^{\circ k}\phi(X_0) + \dots,$$

with generator, using Hessian matrix $\nabla^2\phi$, given by

$$\mathcal{L}\phi = \phi'F + \frac{\sigma^2}{2} \sum_{a=1}^d \phi''(D_a, D_a) = F \cdot \nabla\phi + \frac{\sigma^2}{2} \text{Tr}((\nabla^2\phi)DD^T).$$

1.1 Weak order of an integrator

An integrator $X_1 = \Phi_h(X_0)$ with the weak Taylor expansion

$$\mathbb{E}[\phi(X_1)] = \phi(X_0) + h\mathcal{A}_1\phi(X_0) + \dots + h^k\mathcal{A}_k\phi(X_0) + \dots, \quad (1.1)$$

has weak order p if $\mathcal{A}_k = \frac{1}{k!}\mathcal{L}^{\circ k}$ for $k = 1, \dots, p$. [9]

1.2 Order w.r.t. the invariant measure sampling

For an ergodic model (e.g. overdamped Langevin dynamics where $F = -\nabla V$ and mild assumptions) with invariant measure μ , the solution $X(t)$ satisfies

$$\lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T \phi(X(t))dt = \int_{\mathbb{R}^d} \phi(x)d\mu(x), \quad \text{a.s.}$$

An ergodic integrator $X_n \mapsto X_{n+1}$ has order q with respect to invariant measure sampling if

$$\left| \lim_{N \rightarrow \infty} \frac{1}{N+1} \sum_{k=0}^N \phi(X_k) - \int_{\mathbb{R}^d} \phi(x)d\mu(x) \right| \leq Ch^q, \quad (1.2)$$

Given the differential operators \mathcal{A}_k from the weak Taylor expansion (1.1) of $\mathbb{E}[\phi(X_1)]$, the condition (1.2) is satisfied if,

$$\int_{\mathbb{R}^d} \mathcal{A}_k\phi(x)d\mu(x) = 0, \quad k = 1, \dots, q. \quad (1.3)$$

If the integrator has weak order p , then $q \geq p$. [1, 9]

1.3 Taylor expansions are tedious to manipulate!

Third term of the weak Taylor expansion of $X(h)$:

$$\begin{aligned} \mathcal{L}^{\circ 2} = & F^i F^j \partial_{ij} + F^i \partial_i F^j \partial_j + \sigma^2 F^i D_a^j D_a^k \partial_{ijk} + \sigma^2 F^i \partial_i D_a^j D_a^k \partial_{ijk} \\ & + \sigma^2 D_a^j D_a^k \partial_k F^i \partial_{ij} + \frac{1}{2} \sigma^2 D_a^j D_a^k \partial_j^2 F^i \partial_i \\ & + \frac{1}{4} \sigma^4 D_{a_1}^i D_{a_1}^j D_{a_2}^k D_{a_2}^l \partial_{ijkl} + \sigma^4 D_{a_1}^i D_{a_1}^j D_{a_2}^k \partial_j \partial_l \partial_{ikl} \\ & + \frac{1}{2} \sigma^4 D_{a_1}^i \partial_i D_{a_2}^k D_{a_1}^j \partial_j D_{a_2}^l \partial_{kl} + \frac{1}{2} \sigma^4 D_{a_2}^k D_{a_1}^j D_{a_1}^i \partial_i \partial_j D_{a_2}^l \partial_{kl}. \end{aligned}$$

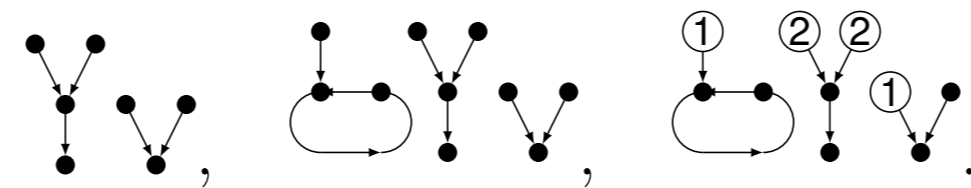
and using exotic aromatic forests:

$$\begin{aligned} \mathcal{L}^{\circ 2} = & \mathcal{F}(\bullet \bullet + \bullet \bullet + \bullet \bullet + \bullet \bullet + \bullet \bullet + \frac{1}{2} \bullet \bullet) \\ & + \frac{1}{4} \bullet \bullet \bullet \bullet + \frac{1}{2} \bullet \bullet \bullet \bullet + \frac{1}{2} \bullet \bullet \bullet \bullet. \end{aligned}$$

Example taken from [5].

2. Exotic aromatic forests

An exotic aromatic forest is a forest with edges oriented from top to bottom. This forest can contain cycles with edges oriented counterclockwise, and some of its vertices may be paired. For example:



In these forests, vertices represent vector fields, and edges represent directional derivatives. Cycles allow us to represent divergences, while paired vertices correspond to Laplacians. [2, 6]

2.1 Using forests to check weak order

We write $\mathcal{A}_k = \mathcal{F}(\sum_{\pi \in F_k} \frac{a(\pi)}{\sigma(\pi)} \pi)$ and obtain the following order conditions for weak order p :

$$a(\pi) = \frac{\alpha(\pi)}{|\pi|!}, \quad \text{for all } \pi \in EAF, |\pi| \leq p,$$

where $a(\pi): EAF \rightarrow \mathbb{R}$ is a functional coming from the integrator, $\sigma(\pi)$ is the symmetry of π , $|\pi|$ is the number of vertices, and $\alpha(\pi)$ is a number of ordered labelings.

2.2 Using forests to check inv. measure sampling order

Theorem 1 ([2, 5]). We can use integration by parts denoted by \sim to modify \mathcal{A}_k without changing the value of the integral in (1.3). The order conditions become

$$(a \circ A)(\tau) = 0, \quad \text{for all } \tau \in EAT, |\tau| \leq q,$$

where A is an adjoint operation of the integration by parts.

For example, we obtain among the order two conditions:

$$\begin{aligned} (a \circ A)(\bullet \bullet) &= a(\bullet \bullet) - 2a(\bullet \bullet) + a(\bullet \bullet) - \frac{1}{2}a(\bullet \bullet \bullet \bullet) = 0, \\ (a \circ A)(\bullet \bullet \bullet) &= a(\bullet \bullet \bullet) - 2a(\bullet \bullet \bullet) + a(\bullet \bullet \bullet) - \frac{1}{2}a(\bullet \bullet \bullet \bullet) = 0, \end{aligned}$$

3. New order two scheme w.r.t. the inv. measure

Consider the Langevin dynamics, $V: \mathbb{R}^d \rightarrow \mathbb{R}$, $D \in \mathbb{R}^{d \times d}$,

$$dX = -D^2 \nabla V(X)dt + \sigma DdW, \quad X(t) \in \mathbb{R}^d,$$

and Leimkuhler-Matthews scheme [7], $\xi_n \sim \mathcal{N}(0, I_d)$,

$$X_{n+1} = X_n - hD^2 \nabla V(X_n) + \sqrt{h}\sigma D \frac{\xi_n + \xi_{n+1}}{2},$$

which can be rewritten in Markovian form as

$$\begin{aligned} X_{n+1} &= X_n - hD^2 \nabla V(\bar{X}_n) + \sqrt{h}\sigma D \xi_n, \\ \bar{X}_n &= X_n + \frac{1}{2}\sqrt{h}\sigma D \xi_n. \end{aligned}$$

Then, $\bar{X}_n \rightarrow \bar{X}_{n+1}$ is second-order w.r.t the invariant measure sampling and has only one evaluation of ∇V per step.

3.1 New generalization

We consider Langevin equation with variable diffusion matrix $D: \mathbb{R}^d \rightarrow \mathbb{R}^{d \times d}$, with D uniformly s.p.d.

$$dX = -D^2(X)\nabla V(X)dt + \frac{\sigma^2}{2} \text{div}(D^2(X))dt + \sigma D(X)dW,$$

where $D(X)$ is symmetric. The new scheme has the form

$$\begin{aligned} X_{n+1} &= X_n + hF(\bar{X}_n) + \hat{\Phi}_h^D(X_n + \frac{1}{4}hF(\bar{X}_{n-1})), \\ \bar{X}_n &= X_n + \frac{1}{2}\sigma\sqrt{h}D(X_n)\xi_n, \end{aligned} \quad (3.1)$$

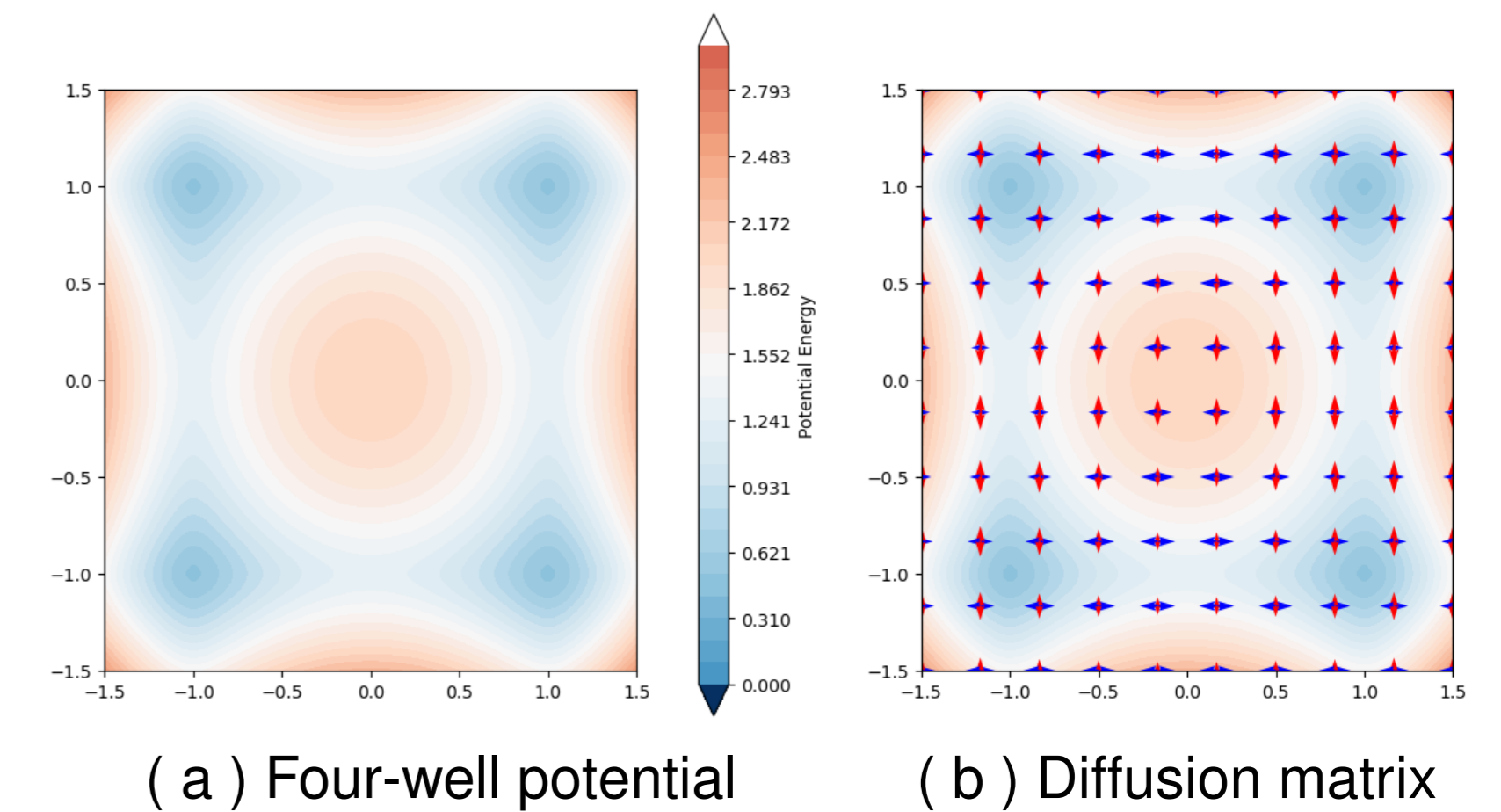
where $I + \hat{\Phi}_h^D$ is a weak order 2 integrator of

$$dX = \sigma D(X)dW.$$

Theorem 2 ([5]). The scheme $\bar{X}_n \rightarrow \bar{X}_{n+1}$ is second-order w.r.t the invariant measure sampling and has only one evaluation of ∇V per step.

3.2 Example model

Diffusion matrix D can be selected to aid in sampling the invariant measure ([8] for details). An example of a potential (a) and eigenvectors of a diffusion matrix (b) are shown.



(a) Four-well potential (b) Diffusion matrix

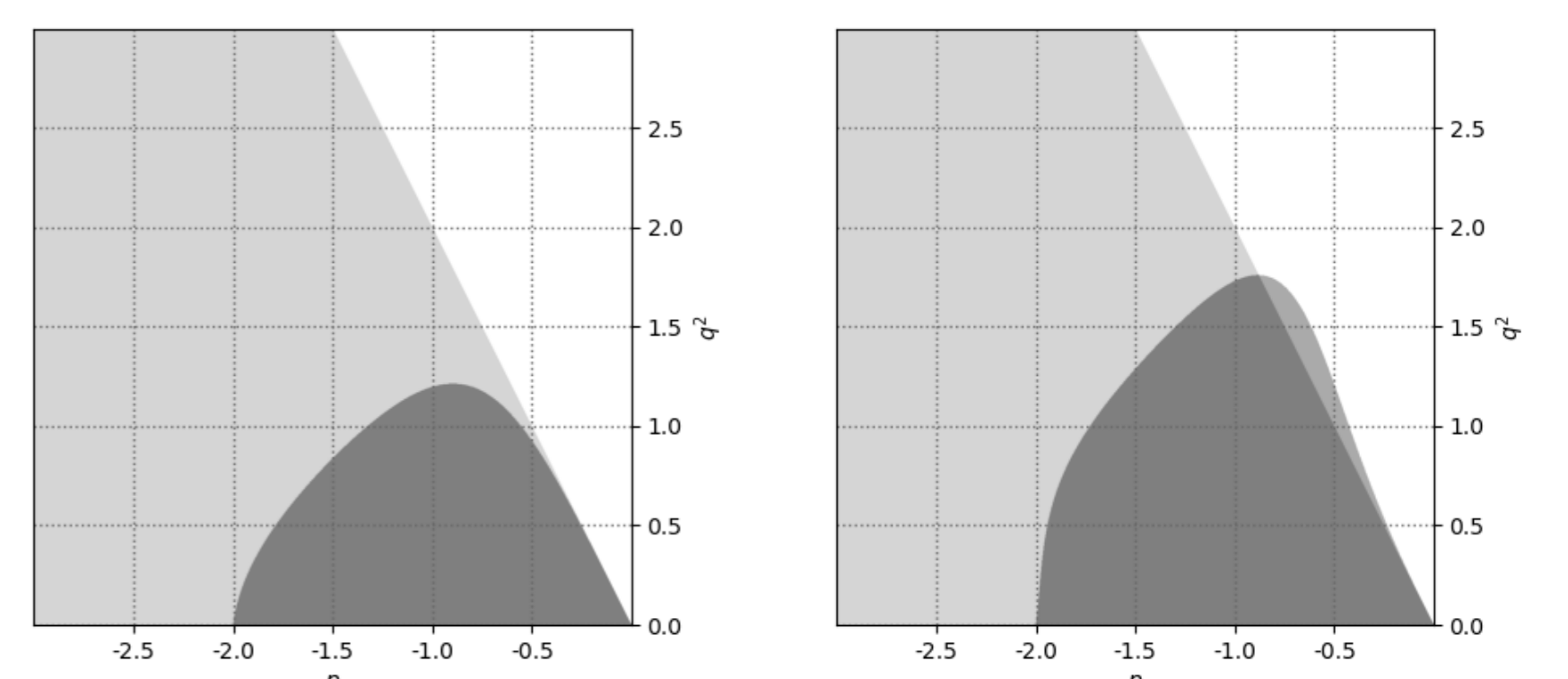
Figure 1: Example model in 2D

3.3 Stability analysis of the new scheme

We consider the following Saito-Mitsui (1996) test model:

$$dX(t) = \lambda X(t)dt + \mu X(t)dW(t), \quad X(0) = 1.$$

We take $\lambda, \mu \in \mathbb{R}$ and let $p = h\lambda$ and $q = \sqrt{h}\mu$ and obtain the following mean-square stability domains (i.e. where $\mathbb{E}(X_n^2)$ is bounded). Light gray denotes the mean-square stability domain of the exact solution:



(a) new method (3.1) (b) stabilized (3.1)

Figure 2: Mean-square stability domains

Figure (b) shows the stability region of the new method with a modified noise integrator $I + \hat{\Phi}_h^D$ for better stabilization.

4. Related ongoing work

In a joint work [4] with Adrien Laurent (INRIA Rennes), we uncover the following algebraic structures of exotic aromatic forests:

- a free tracial pre-Lie-Rinehart algebra,
- a free D-algebra, pre-Hopf algebroid,
- a multi-pre-Lie algebra,
- a comodule-bialgebra structure,

which are essential in the description of the backward error analysis in the context of ergodic stochastic differential equations.

In a joint work [3] with Jean-Luc Falcone (University of Geneva), we develop a Haskell package to automate computations involving exotic aromatic forests. GitLab: <https://gitlab.unige.ch/Eugen.Bronasco/graphalgebra.hs>.

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