

The Adaptive Biasing Force algorithm with non-conservative forces

10ème Biennale Française des Mathématiques Appliquées et Industrielles

Tony Lelièvre¹, Lise Maurin², Pierre Monmarché², Jean-Philip Piquemal²

¹ CERMICS, Ecole des Ponts ParisTech

² LJLL & LCT, Sorbonne Université

24th June, 2021



LCT
Laboratoire de Chimie Théorique



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Setting

- N particles, $q = (q_1, \dots, q_N) \in \mathcal{D}$ their positions.
- Potential energy $V \in \mathcal{C}^2(\mathcal{D})$, interaction force $\mathcal{F} = -\nabla V$
- **Boltzmann-Gibbs measure:**

$$\mu(\mathrm{d}q) = Z_\mu^{-1} e^{-\beta V(q)} \mathrm{d}q, \quad Z_\mu = \int_{\mathcal{D}} e^{-\beta V(q)} \mathrm{d}q$$

- **Canonical Mean/ thermodynamic quantity:**

$$\psi \in \mathcal{C}_0^\infty(\mathcal{D}) \text{ observable, and } \mathbb{E}_\mu[\psi] = \int_{\mathcal{D}} \psi \mathrm{d}\mu$$

Goal: being able to sample μ !

Overdamped Langevin dynamics

- **Dynamics:**

$$dQ_t = -\nabla V(Q_t)dt + \sqrt{2\beta^{-1}}dW_t$$

- **Infinitesimal generator:**

$$\mathcal{L}\varphi(X) = \lim_{t \rightarrow +\infty} \frac{\mathbb{E}_\mu[\varphi(X_t)] - \varphi(X)}{t} = -\nabla V \cdot \nabla \varphi(X) + \beta \Delta \varphi(X)$$

- **Fokker-Planck equation:**

$$\partial_t \pi_t = \mathcal{L}^* \pi_t$$

- **Ergodicity:** $(Q_t)_{t \geq 0}$ is ergodic with respect to μ . In other words:

$$\forall \psi \in \mathcal{C}_0^\infty(\mathcal{D}), \quad \lim_{\tau \rightarrow +\infty} \frac{1}{\tau} \int_0^\tau \psi(Q_t) dt = \mathbb{E}_\mu[\psi]$$

- **Problem:** Metastability!

What is metastability?

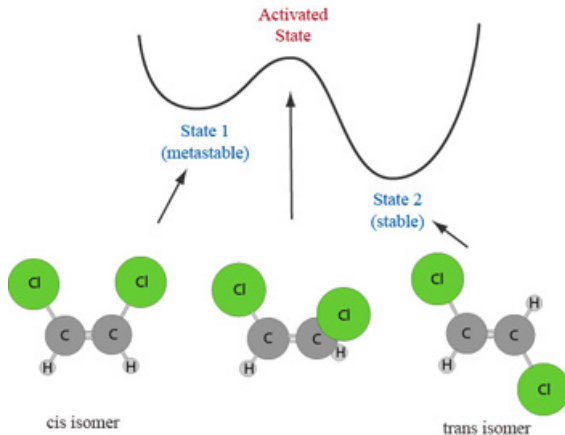


Figure: cis-trans isomerization of 1,2-Dichloroethene
(<http://chemcollective.org/chem/entropy/reactcoord.php>)

Reaction coordinate

- $\xi : \mathcal{D} \rightarrow \mathcal{M}$, with \mathcal{M} manifold of dimension $d' \leq dN$.
- $\Sigma_z = \{q \in \mathcal{D} \mid \xi(q) = z\}$
- **Free energy:**

$$A(z) = -\beta^{-1} \ln(Z_{\Sigma_z}), \quad Z_{\Sigma_z} = \int_{\Sigma_z} e^{-\beta V(q)} \delta_{\xi(q)-z}(\mathrm{d}q)$$

- **Local mean force:** $F (\rightsquigarrow (\nabla \xi)^\top \nabla \xi, V, \beta)$
- **Free energy derivative:** $\nabla A(z) = \mathbb{E}[F(Q) \mid \xi(Q) = z]$
- **Property:** If $Q \sim \mu$ then $\xi(Q) \sim \mu_A$:

$$\mu_A = \xi \star \mu(\mathrm{d}z) = \frac{e^{-\beta A(z)} \mathrm{d}z}{\int_{\mathcal{M}} e^{-\beta A(z)} \mathrm{d}z}. \quad (1)$$

From now on

- Position $q = (x, y) \in \mathbb{T}^n$
- Stochastic process $Q_t = (X_t, Y_t) \quad \forall t \geq 0$
- with $x \in \mathbb{T}^m$, $y \in \mathbb{T}^{n-m}$ and $\xi(x, y) = x$
- $\nabla A(z) = \mathbb{E}[\nabla_x V(x, y) \mid \xi(x, y) = x = z]$

\rightsquigarrow Convergence results are expressed using the relative entropy of two measures:

$$\mathcal{H}(\mu|\nu) = \begin{cases} \int \ln(\frac{d\mu}{d\nu}) d\mu & \text{if } \mu \ll \nu \\ +\infty & \text{else.} \end{cases}$$

The idea

- ξ is a **generalized coordinate**
- (q, ξ) extended coordinates
- Methods using ξ : **generalized ensemble methods**

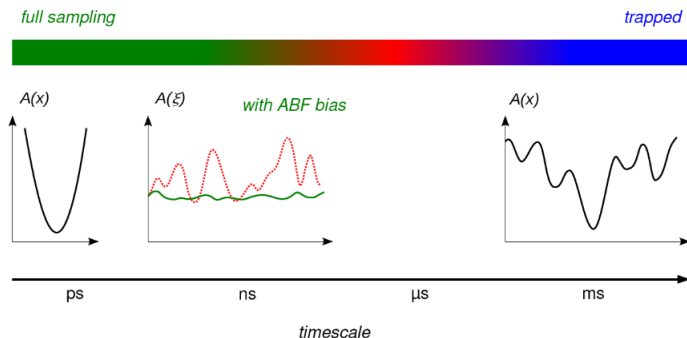


Figure: Concept of ABF (Jérôme Hénin, CECAM workshop, Paris, 2021)

First attempt

- Biased Overdamped Langevin:

$$\begin{cases} dX_t = (-\nabla_x V + \nabla A \circ \xi)(X_t, Y_t) dt + \sqrt{2\beta^{-1}} dW_t^1 \\ dY_t = -\nabla_y V(X_t, Y_t) dt + \sqrt{2\beta^{-1}} dW_t^2 \end{cases}$$

- Fokker-Planck equation:

$$\partial_t \pi_t = \beta^{-1} \Delta \pi_t - \nabla \cdot ((-\nabla V + \nabla A \circ \xi) \pi_t)$$

- Stationary measure: $\pi_\infty \propto e^{-\beta(V - A \circ \xi)}$
- $\xi \star \pi_\infty$ is the uniform law!

ABF

- ABF:

$$\begin{cases} dX_t = (-\nabla_x V + G_t \circ \xi)(X_t, Y_t) dt + \sqrt{2\beta^{-1}} dW_t^1 \\ dY_t = -\nabla_y V(X_t, Y_t) dt + \sqrt{2\beta^{-1}} dW_t^2 \\ G_t(z) = \mathbb{E}[\nabla_x V(X_t, Y_t) | \xi(X_t, Y_t) = z] \end{cases}$$

- Fokker-Planck equation:

$$\partial_t \pi_t = \beta^{-1} \Delta \pi_t - \nabla \cdot ((-\nabla V + G_t \circ \xi) \pi_t)$$

- Known results: T.Lelièvre, F.Otto, M.Rousset, G.Stoltz – 2007

- (i) $\lim_{t \rightarrow +\infty} G_t = \nabla A$
- (ii) $\lim_{t \rightarrow +\infty} \pi_t = \pi_\infty$
- (iii) **Flat histogram property:** the law $\pi_t^\xi = \xi \star \pi_t$ of $(\xi(Q_t))_{t \geq 0}$ is s.t.
 $\lim_{t \rightarrow +\infty} \pi_t^\xi = \lambda$: no more energy barriers.

Projected Adaptive Biasing Force method

- In simulations, the converged bias $G = \lim_{t \rightarrow +\infty} G_t$ is *a priori* not a gradient, and yet we are looking for ∇A !
- PABF:

$$\begin{cases} dX_t = (-\nabla_x V + \nabla H_t \circ \xi)(X_t, Y_t) dt + \sqrt{2\beta^{-1}} dW_t^1 \\ dY_t = -\nabla_y V(X_t, Y_t) dt + \sqrt{2\beta^{-1}} dW_t^2 \\ \nabla H_t(z) = P_{L^2(\lambda)}(G_t) = P_{L^2(\lambda)}(\mathbb{E}[\nabla_x V(X_t, Y_t) | \xi(X_t, Y_t) = z]) \end{cases}$$

- Fokker-Planck equation:

$$\partial_t \pi_t = \beta^{-1} \Delta \pi_t - \nabla \cdot ((-\nabla V + \nabla H_t \circ \xi) \pi_t)$$

- ▶ (H. Alrachid, T. Lelièvre, 2015): technical trick regarding the projection.

Problem

- **Question:** How about nonconservative forces (i.e for a generic interacting force \mathcal{F})?
- **Motivation:**
 - Implementing $-\nabla V$ (*ab initio* MD approximations).
A priori the force $\mathcal{F} = -\nabla V + \varepsilon g$ is not conservative!
 - Is there a stationary measure to the Fokker-Planck equation?
 - Provided it exists, do we still have long-time convergence?
 - Can one obtain an estimation of the error made on the system's free energy?

Final model

Dynamics:

$$\begin{cases} dX_t = \mathcal{F}_1(X_t, Y_t)dt + B_t(X_t)dt + \sqrt{2\beta^{-1}}dW_t^1 \\ dY_t = \mathcal{F}_2(X_t, Y_t)dt + \sqrt{2\beta^{-1}}dW_t^2 \end{cases} \quad (2)$$

where the bias B_t is either:

- $G_t(\cdot) = \mathbb{E}[-\mathcal{F}_1(X_t, Y_t) | \xi(X_t, Y_t) = \cdot]$ (ABF)
- $\nabla H_t = P_{L^2(\lambda)}(G_t)$ (Projected ABF)

where $P_{L^2(\nu)}(G)$ is the Helmholtz projection in $L^2(\nu)$ of the vector field G .

Fokker-Planck equation:

$$\partial_t \pi_t = \beta^{-1} \Delta \pi_t - \nabla \cdot ((\mathcal{F} + B_t \circ \xi) \pi_t) \quad (3)$$

Flat histogram property

Proposition

For both the ABF and PABF algorithm, under several assumptions, π_t^ξ converges towards the Lebesgue measure as $t \rightarrow \infty$. More precisely, for all $t \geq 0$

$$\mathcal{H}(\pi_t^\xi | \lambda) \leq e^{-8\beta^{-1}\pi^2 t} \mathcal{H}(\pi_0^\xi | \lambda).$$

Furthermore, there exists $C > 0$ such that for all initial distribution $\pi_0^\xi \in L^2(\mathbb{T}^m)$, for all $t \geq 1$:

$$\|\pi_t^\xi - 1\|_\infty \leq C e^{-4\beta^{-1}\pi^2 t} \|\pi_0^\xi - 1\|_2.$$

↪ Flat histogram property verified in all cases.

- Entropic: PDE satisfied by π_t^ξ + deriving the entropy + Gronwall.
- L- ∞ : Nash inequality + PDE satisfied by π_t^ξ .

Existence result

Theorem

For both the ABF and PABF algorithms, under several assumptions, there exists a couple of stationary measure and bias $(\pi_\infty^{\mathcal{F}}, B_\infty^{\mathcal{F}})$ to (3), such that $\pi_\infty^{\mathcal{F}} \in \mathcal{C}^0(\mathbb{T}^n)$ is strictly positive. As a consequence,

- (i) $\pi_\infty^{\mathcal{F}}$ satisfies a log-Sobolev inequality for some constant $R > 0$,
- (ii) the conditional density $y \mapsto \pi_{\infty, x}^{\mathcal{F}}(y) := \pi_\infty^{\mathcal{F}}(x, y) / \pi_\infty^{\mathcal{F}, \xi}(x)$ satisfies a log-Sobolev inequality for some constant ρ , for all $x \in \mathbb{T}^m$.

\rightsquigarrow **Note:** $\pi_\infty^{\mathcal{F}}$ different for ABF and PABF.

- Invariant probability measures of homogeneous diffusions.
- Fixed point problem.

Long-time convergence: conservative case

Theorem

Let us consider the ABF and PABF algorithms under several assumptions. Let us suppose moreover that $\mathcal{F} = -\nabla V$, with $V \in \mathcal{C}^2(\mathbb{T}^n)$. Then, there exists $K > 0$ such that, for all $\varepsilon > 0$ and for all $t \geq 0$:

$$\mathcal{H}(\pi_t | \pi_\infty^{\mathcal{F}}) \leq K \left(1 + \frac{1}{\varepsilon^2}\right) e^{-(\Lambda - \varepsilon)t},$$

with $\pi_\infty^{\mathcal{F}} = \mu_A$ being given by (1), $\Lambda = (8\pi^2 \wedge 2\rho) \beta^{-1}$ (ABF) and $\Lambda = (4\pi^2 \wedge 2\rho) \beta^{-1}$ (PABF). Furthermore, $(\pi_\infty^{-\nabla V}, B_\infty^{-\nabla V}) = (\mu_A, \nabla A)$ is the unique fixed point of (3).

↪ PABF case: theoretical gap filled.

- Classical proof skeleton + property of the Helmholtz decomposition.

Long-time convergence: non-conservative case

Theorem

Let us consider the ABF and PABF algorithms under several assumptions (among which \mathcal{F}_1 is M -Lipschitz). Let $\pi_{\infty}^{\mathcal{F},\rho}$ be a stationary measure for (3) and a constant such as in Theorem 1. Suppose moreover that $M\beta < 2\rho$. Then there exists $K \geq 0$ such that, for all $t \geq 0$:

$$\mathcal{H}(\pi_t | \pi_{\infty}^{\mathcal{F}}) \leq K e^{-\Lambda t}$$

with $\Lambda = 2R(1 - \frac{M\beta}{2\rho})\beta^{-1}$.

As a consequence, the dynamics (3) admits a unique stationary measure.

↪ It is ok to use ABF and PABF with non-conservative forces.

- Classical proof skeleton + Helmholtz decomposition.

Error bounds

Proposition

For the PABF algorithm, for all $V \in \mathcal{C}^2(\mathbb{T}^n)$ and $p \geq 1$, there exists $K_V > 0$ and $K_p > 0$ such that the following holds. For all $\mathcal{F} \in \mathcal{C}^1(\mathbb{T}^n)$ satisfying $\|\mathcal{F} + \nabla V\|_\infty \leq 1$, for all stationary measure $\pi_\infty^\mathcal{F}$ of (3), considering the corresponding bias $\nabla H_\infty^\mathcal{F}$, one has

$$\|\nabla A - \nabla H_\infty^\mathcal{F}\|_{L^p(\mathbb{T}^m)} \leq K_V K_p \|\mathcal{F} + \nabla V\|_\infty,$$

and, for all $\psi \in L^\infty(\mathbb{T}^n)$, considering

$$\hat{I}_\psi := \frac{\int_{\mathbb{T}^n} \psi(x, y) e^{-\beta H_\infty^\mathcal{F}(x)} \pi_\infty^\mathcal{F}(x, y) dx dy}{\int_{\mathbb{T}^n} e^{-\beta H_\infty^\mathcal{F}(x)} \pi_\infty^\mathcal{F}(x, y) dx dy},$$

one has

$$\left| \int_{\mathbb{T}^n} \psi d\mu - \hat{I}_\psi \right| \leq K_V \|\psi\|_\infty \|\mathcal{F} + \nabla V\|_\infty.$$

Quick recap

- ▶ ABF and PABF are robust:
 - Stationary measure $\pi_{\infty}^{\mathcal{F}}$.
 - Flat histogram property.
 - Convergence of π_t towards $\pi_{\infty}^{\mathcal{F}}$.
 - Convergence of the bias.
 - Bound on the error made on the system's free energy.
- ▶ Convergence proof for ABF applied to the kinetic Langevin dynamics, where both momenta and positions are considered?

Now, some alchemy!

- **Alchemical reactions:** $\xi \equiv \lambda$ guides the system from an initial state $\lambda = 0$ to a final state $\lambda = 1$.
- **Extended Hamiltonian:**

$$\begin{aligned} H(q, p; \lambda) &= E_k(p, \lambda) + V(q, \lambda) \\ &= E_k(p, \lambda) + V_{elec}(q, \lambda) + V_{vdW}(q, \lambda) \end{aligned}$$

- $H(q, p; 0) = H^{ini}(q, p)$, $H(q, p; 1) = H^{end}(q, p)$.
- In-between states $\lambda \in (0, 1)$ are allowed to not make sense physically!

Applications

Use: estimating free energy differences.

- Hydration FE
- Ligand-binding affinity

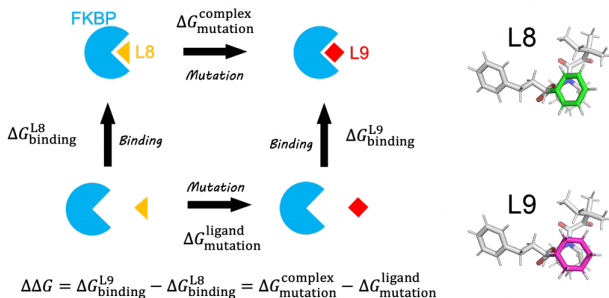


Figure: Thermodynamic cycle of protein-ligand binding and ligand mutation.
 (<https://www.r-ccs.riken.jp/labs/cbrt/tutorials2019/genesis-tutorial-15-1/>)

"Random walk in orthogonal space to achieve efficient free-energy simulation of complex systems"

Lianqing Zheng, Mengen Chen, and Wei Yang, 2008

↪ Choice of a particular reaction coordinate: (λ, F_λ) .

Questions:

- ▶ How about using ABF with (λ, F_λ) ?
- ▶ Is (λ, F_λ) the only good choice? Try with (λ, ξ) ?

Implementation

Treating alchemical reactions in TinkerHP

- Softcores: playing with forces ignition!
- Establish bridge between TinkerHP and Colvars software
- Implement and run λ -dynamics

Toy models: solvation of water in water, solvation of sodium in water

What remains to be done

- Comparing ABF to OSRW in TinkerHP: many ways to do it!

Thank you for your attention!



T. Lelièvre, G. Stoltz

Partial differential equations and stochastic methods in molecular dynamics
In *Acta Numerica*, 2016



H. Alrachid, T. Lelièvre

Long-time convergence of an adaptive biasing force method: variance reduction by Helmholtz projection
In *Journal of computational mathematics*, 2015



T. Lelièvre, F. Otto, M. Rousset, G. Stoltz

Long-time convergence of an adaptive biasing force method
preprint, 2007



X. Kong, C. L. Brooks III

Lambda-dynamics: A new approach to free energy calculations
In *The Journal of Chemical Physics* 105, 2414, 1996



L. Zheng, M. Chen, W. Yang

Random walk in orthogonal space to achieve efficient free-energy simulation of complex systems
In *Proceedings of the National Academy of Sciences of the U.S.A*, 2008