



Convergence and approximation of Langevin-like dynamics

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- **Some elements of statistical physics**

- **Equilibrium Langevin dynamics**
 - Convergence results: a review
 - A focus on the approach by Dolbeault, Mouhot and Schmeiser
 - Various extensions/modifications

- **Numerical approximation of Langevin dynamics**
 - Splitting schemes
 - Numerical analysis: error estimates on invariant measures

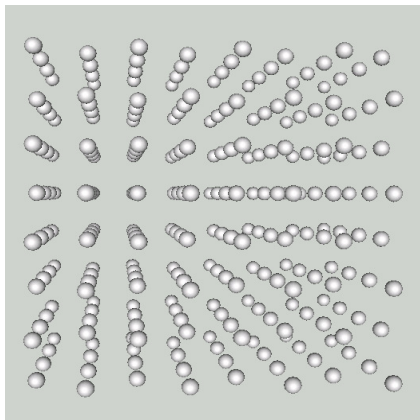
Some elements of statistical physics

General perspective (1)

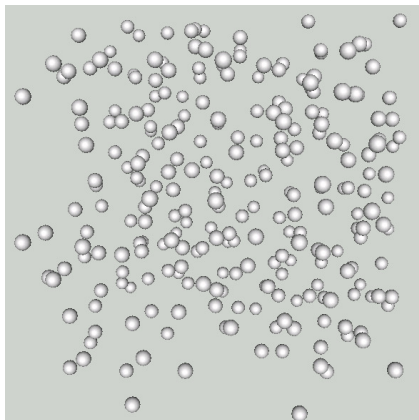
- **Aims** of computational statistical physics:
 - numerical microscope
 - computation of **average properties**, static or dynamic
- Orders of magnitude
 - distances $\sim 1 \text{ \AA} = 10^{-10} \text{ m}$
 - energy per particle $\sim k_B T \sim 4 \times 10^{-21} \text{ J}$ at room temperature
 - atomic masses $\sim 10^{-26} \text{ kg}$
 - **time $\sim 10^{-15} \text{ s}$**
 - number of particles $\sim \mathcal{N}_A = 6.02 \times 10^{23}$
- “Standard” simulations
 - 10^6 particles [“world records”: around 10^9 particles]
 - integration time: (fraction of) ns [“world records”: (fraction of) μs]

General perspective (2)

What is the **melting temperature** of argon?



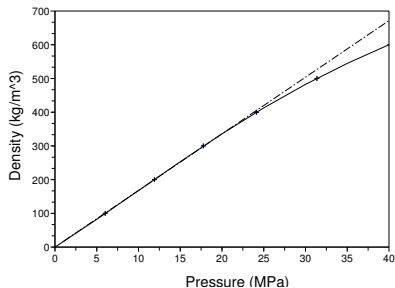
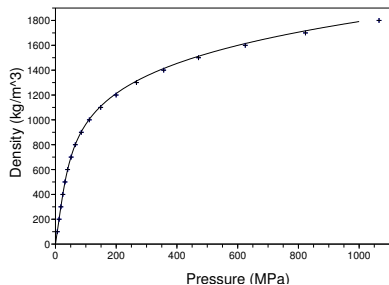
(a) Solid argon (low temperature)



(b) Liquid argon (high temperature)

General perspective (3)

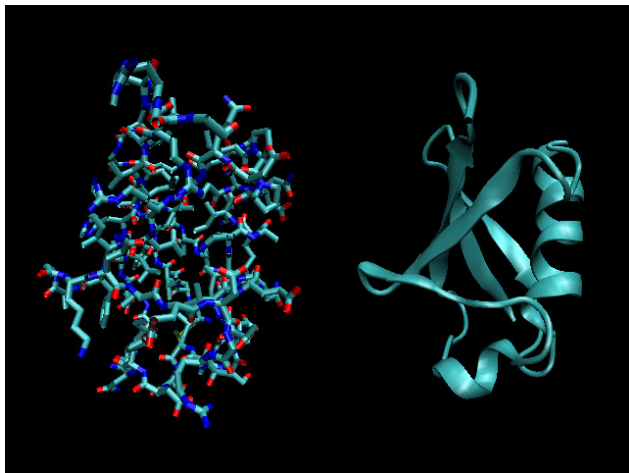
“Given the structure and the laws of interaction of the particles, what are the **macroscopic properties** of the matter composed of these particles?”



Equation of state (pressure/density diagram) for Argon at $T = 300$ K

General perspective (4)

What is the **structure** of the protein? What are its **typical conformations**, and what are the **transition pathways** from one conformation to another?



Microscopic description of physical systems: unknowns

- **Microstate** of a classical system of N particles:

$$(q, p) = (q_1, \dots, q_N, p_1, \dots, p_N) \in \mathcal{E}$$

Positions q (configuration), **momenta** p (to be thought of as $M\dot{q}$)

- Here, periodic boundary conditions: $\mathcal{E} = \mathcal{D} \times \mathbb{R}^{3N}$ with $\mathcal{D} = (L\mathbb{T})^{3N}$
- More complicated situations can be considered: molecular **constraints** defining submanifolds of the phase space
- **Hamiltonian** $H(q, p) = E_{\text{kin}}(p) + V(q)$, where the kinetic energy is

$$E_{\text{kin}}(p) = \frac{1}{2} p^T M^{-1} p, \quad M = \begin{pmatrix} m_1 \text{Id}_3 & & 0 \\ & \ddots & \\ 0 & & m_N \text{Id}_3 \end{pmatrix}.$$

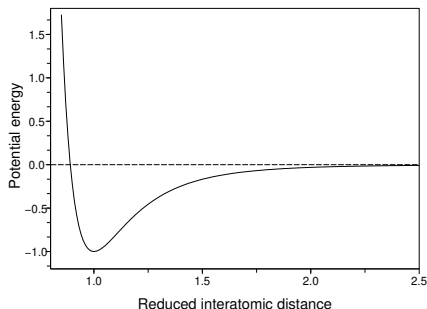
Microscopic description: interaction laws

- All the physics is contained in V
 - ideally derived from **quantum mechanical** computations
 - in practice, **empirical** potentials for large scale calculations
- An example: **Lennard-Jones** pair interactions to describe noble gases

$$V(q_1, \dots, q_N) = \sum_{1 \leq i < j \leq N} v(|q_j - q_i|)$$

$$v(r) = 4\varepsilon \left[\left(\frac{\sigma}{r}\right)^{12} - \left(\frac{\sigma}{r}\right)^6 \right]$$

$$\text{Argon: } \begin{cases} \sigma = 3.405 \times 10^{-10} \text{ m} \\ \varepsilon/k_B = 119.8 \text{ K} \end{cases}$$



Average properties

- **Macrostate** of the system described by a **probability measure**

Equilibrium thermodynamic properties (pressure,...)

$$\langle \varphi \rangle_\mu = \mathbb{E}_\mu(\varphi) = \int_{\mathcal{E}} \varphi(q, p) \mu(dq dp)$$

- Examples of **observables**:

- Pressure $\varphi(q, p) = \frac{1}{3|\mathcal{D}|} \sum_{i=1}^N \left(\frac{p_i^2}{m_i} - q_i \cdot \nabla_{q_i} V(q) \right)$

- Kinetic temperature $\varphi(q, p) = \frac{1}{3Nk_B} \sum_{i=1}^N \frac{p_i^2}{m_i}$

- **Canonical** ensemble = measure on (q, p) (average energy fixed)

$$\mu_{\text{NVT}}(dq dp) = Z_{\text{NVT}}^{-1} e^{-\beta H(q,p)} dq dp, \quad \beta = \frac{1}{k_B T}$$

Aims of computational statistical physics

- **“Numerical microscope”**

- gaining some **insight** into physical mechanisms at the atomic scale
- From the press release for the Nobel prize in Chemistry 2013 (Karplus/Levitt/Warshel)

Today the computer is just as important a tool for chemists as the test tube. Simulations are so realistic that they predict the outcome of traditional experiments.

- **Computation of average properties:** **high dimensional integrals**

→ ergodic averages

- **Computation of dynamical quantities**

- reactive paths, transition kinetics
- transport coefficients (**nonequilibrium** steady state simulations)

Standard Langevin dynamics

Langevin dynamics (1)

- Positions $q \in \mathcal{D} = (L\mathbb{T})^d$ or \mathbb{R}^d and momenta $p \in \mathbb{R}^d$
→ phase-space $\mathcal{E} = \mathcal{D} \times \mathbb{R}^d$
- **Hamiltonian** $H(q, p) = V(q) + \frac{1}{2} p^T M^{-1} p$

Stochastic perturbation of the Hamiltonian dynamics **friction** $\gamma > 0$

$$\begin{cases} dq_t = M^{-1} p_t dt \\ dp_t = -\nabla V(q_t) dt - \gamma M^{-1} p_t dt + \sqrt{\frac{2\gamma}{\beta}} dW_t \end{cases}$$

- Almost-sure convergence¹ of **ergodic averages** $\hat{\varphi}_t = \frac{1}{t} \int_0^t \varphi(q_s, p_s) ds$

¹Kliemann, *Ann. Probab.* **15**(2), 690-707 (1987)

Langevin dynamics (2)

- Evolution semigroup $(e^{t\mathcal{L}}\varphi)(q, p) = \mathbb{E} \left[\varphi(q_t, p_t) \mid (q_0, p_0) = (q, p) \right]$
- Generator of the dynamics \mathcal{L}

$$\frac{d}{dt} \left(\mathbb{E} \left[\varphi(q_t, p_t) \mid (q_0, p_0) = (q, p) \right] \right) = \mathbb{E} \left[(\mathcal{L}\varphi)(q_t, p_t) \mid (q_0, p_0) = (q, p) \right]$$

Generator of the Langevin dynamics $\mathcal{L} = \mathcal{L}_{\text{ham}} + \gamma\mathcal{L}_{\text{FD}}$

$$\mathcal{L}_{\text{ham}} = p^T M^{-1} \nabla_q - \nabla V^T \nabla_p, \quad \mathcal{L}_{\text{FD}} = -p^T M^{-1} \nabla_p + \frac{1}{\beta} \Delta_p$$

- Existence and uniqueness of the invariant measure characterized by

$$\forall \varphi \in C_0^\infty(\mathcal{E}), \quad \int_{\mathcal{E}} \mathcal{L}\varphi d\mu = 0$$

- Here, **canonical measure**

$$\mu(dq dp) = Z^{-1} e^{-\beta H(q,p)} dq dp = \nu(dq) \kappa(dp)$$

Fokker–Planck equations

- Evolution of the law $\psi(t, q, p)$ of the process at time $t \geq 0$

$$\frac{d}{dt} \left(\int_{\mathcal{E}} \varphi \psi(t) \right) = \int_{\mathcal{E}} (\mathcal{L}\varphi) \psi(t)$$

- Fokker–Planck equation (with \mathcal{L}^\dagger adjoint of \mathcal{L} on $L^2(\mathcal{E})$)

$$\partial_t \psi = \mathcal{L}^\dagger \psi$$

- It is convenient to work in $L^2(\mu)$ with $f(t) = \psi(t)/\mu$
 - denote the adjoint of \mathcal{L} on $L^2(\mu)$ by \mathcal{L}^*

$$\mathcal{L}^* = -\mathcal{L}_{\text{ham}} + \gamma \mathcal{L}_{\text{FD}}$$

- Fokker–Planck equation $\partial_t f = \mathcal{L}^* f$
- Convergence results for $e^{t\mathcal{L}}$ on $L^2(\mu)$ are very similar to the ones for $e^{t\mathcal{L}^*}$

Hamiltonian and overdamped limits

- As $\gamma \rightarrow 0$, the **Hamiltonian** dynamics is recovered
- **Overdamped** limit $\gamma \rightarrow +\infty$ (or masses going to 0)

$$\begin{aligned}q_{\gamma t} - q_0 &= -\frac{1}{\gamma} \int_0^{\gamma t} \nabla V(q_s) ds + \sqrt{\frac{2}{\gamma\beta}} W_{\gamma t} - \frac{1}{\gamma} (p_{\gamma t} - p_0) \\ &= -\int_0^t \nabla V(q_{\gamma s}) ds + \sqrt{2\beta^{-1}} B_t - \frac{1}{\gamma} (p_{\gamma t} - p_0)\end{aligned}$$

which converges to the solution of $dQ_t = -\nabla V(Q_t) dt + \sqrt{2\beta^{-1}} dB_t$

- In both cases, **slow convergence to equilibrium**
 - it takes time to change energy levels in the Hamiltonian limit²
 - for fixed masses, time has to be rescaled by a factor γ

²Hairer and Pavliotis, *J. Stat. Phys.*, **131**(1), 175-202 (2008)

Ergodicity results (1)

- Almost-sure convergence³ of **ergodic averages** $\widehat{\varphi}_t = \frac{1}{t} \int_0^t \varphi(q_s, p_s) ds$
- **Asymptotic variance** of ergodic averages

$$\sigma_\varphi^2 = \lim_{t \rightarrow +\infty} t \mathbb{E} [\widehat{\varphi}_t^2] = 2 \int_{\mathcal{E}} (-\mathcal{L}^{-1} \Pi_0 \varphi) \Pi_0 \varphi d\mu$$

where $\Pi_0 \varphi = \varphi - \mathbb{E}_\mu(\varphi)$

- A central limit theorem holds⁴ when the equation has a solution in $L^2(\mu)$

Poisson equation in $L^2(\mu)$

$$-\mathcal{L}\Phi = \Pi_0 \varphi$$

- Well-posedness of such equations?

³Kliemann, *Ann. Probab.* **15**(2), 690-707 (1987)

⁴Bhattacharya, *Z. Wahrsch. Verw. Gebiete* **60**, 185-201 (1982)

Ergodicity results (2)

- **Invertibility** of \mathcal{L} on subsets of $L_0^2(\mu) = \left\{ \varphi \in L^2(\mu) \mid \int_{\mathcal{E}} \varphi d\mu = 0 \right\}$?

$$-\mathcal{L}^{-1} = \int_0^{+\infty} e^{t\mathcal{L}} dt$$

- Prove **exponential convergence** of the semigroup $e^{t\mathcal{L}}$
 - various Banach spaces $E \cap L_0^2(\mu)$
 - **Lyapunov** techniques⁵ $L_W^\infty(\mathcal{E}) = \left\{ \varphi \text{ measurable, } \left\| \frac{\varphi}{W} \right\|_{L^\infty} < +\infty \right\}$
 - standard **hypocoercive**⁶ setup $H^1(\mu)$
 - $E = L^2(\mu)$ after hypoelliptic regularization⁷ from $H^1(\mu)$
 - **coupling** arguments⁸

⁵L. Wu, *Stoch. Proc. Appl.* (2001); Mattingly, Stuart and Higham, *Stoch. Proc. Appl.* (2002); L. Rey-Bellet, *Lect. Notes Math.* (2006); Hairer and Mattingly, *Progr. Probab.* (2011)

⁶Villani (2009) and before Talay (2002), Eckmann/Hairer (2003), Hérau/Nier (2004)

⁷F. Hérau, *J. Funct. Anal.* **244**(1), 95-118 (2007)

⁸A. Eberle, A. Guillin and R. Zimmer, *arXiv preprint* **1703.01617** (2017)

Direct $L^2(\mu)$ approach: lack of coercivity

- The generator, considered on $L^2(\mu)$, is the sum of...
 - a **degenerate** symmetric part $\mathcal{L}_{\text{FD}} = -p^T M^{-1} \nabla_p + \frac{1}{\beta} \Delta_p$
 - an **antisymmetric** part $\mathcal{L}_{\text{ham}} = p^T M^{-1} \nabla_q - \nabla V^T \nabla_p$
- Standard strategy for coercive generators: consider φ with average 0 with respect to μ and compute

$$\begin{aligned} \frac{d}{dt} \left(\|e^{t\mathcal{L}} \varphi\|_{L^2(\mu)}^2 \right) &= \langle e^{t\mathcal{L}} \varphi, \mathcal{L} e^{t\mathcal{L}} \varphi \rangle_{L^2(\mu)} = \langle e^{t\mathcal{L}} \varphi, \mathcal{L}_{\text{FD}} e^{t\mathcal{L}} \varphi \rangle_{L^2(\mu)} \\ &= -\frac{1}{\beta} \|\nabla_p e^{t\mathcal{L}} \varphi\|_{L^2(\mu)}^2 \leq 0, \end{aligned}$$

but no control of $\|\phi\|_{L^2(\mu)}$ by $\|\nabla_p \phi\|_{L^2(\mu)}$ for a Gronwall estimate...

- **Change of scalar product** in order to use the antisymmetric part

Almost direct $L^2(\mu)$ approach: convergence result

- Assume that the potential V is **smooth** and^{9,10}
 - the marginal measure ν satisfies a **Poincaré** inequality

$$\|\Pi_0 \varphi\|_{L^2(\nu)}^2 \leq \frac{1}{C_\nu} \|\nabla_q \varphi\|_{L^2(\nu)}^2$$

- there exist $c_1 > 0$, $c_2 \in [0, 1)$ and $c_3 > 0$ such that V satisfies

$$\Delta V \leq c_1 + \frac{c_2}{2} |\nabla V|^2, \quad |\nabla^2 V| \leq c_3 (1 + |\nabla V|)$$

There exist $C > 0$ and $\lambda_\gamma > 0$ such that, for any $\varphi \in L^2_0(\mu)$,

$$\forall t \geq 0, \quad \|e^{t\mathcal{L}} \varphi\|_{L^2(\mu)} \leq C e^{-\lambda_\gamma t} \|\varphi\|_{L^2(\mu)}.$$

with convergence rate of order $\min(\gamma, \gamma^{-1})$: there exists $\bar{\lambda} > 0$ such that

$$\lambda_\gamma \geq \bar{\lambda} \min(\gamma, \gamma^{-1}).$$

⁹Dolbeault, Mouhot and Schmeiser, *C. R. Math. Acad. Sci. Paris* (2009)

¹⁰Dolbeault, Mouhot and Schmeiser, *Trans. AMS*, **367**, 3807–3828 (2015)

Sketch of proof

- Modified square norm $\mathcal{H}[\varphi] = \frac{1}{2}\|\varphi\|^2 - \varepsilon \langle A\varphi, \varphi \rangle$ for $\varepsilon \in (-1, 1)$ and
$$A = \left(1 + (\mathcal{L}_{\text{ham}} \Pi_p)^*(\mathcal{L}_{\text{ham}} \Pi_p)\right)^{-1} (\mathcal{L}_{\text{ham}} \Pi_p)^*, \quad \Pi_p \varphi = \int_{\mathbb{R}^D} \varphi d\kappa$$
- $A = \Pi_p A (1 - \Pi_p)$ and $\mathcal{L}_{\text{ham}} A$ are bounded so that $\mathcal{H} \sim \|\cdot\|_{L^2(\mu)}^2$

Coercivity in the scalar product $\langle\langle \cdot, \cdot \rangle\rangle$ induced by \mathcal{H}

$$\mathcal{D}[\varphi] := \langle\langle -\mathcal{L}\varphi, \varphi \rangle\rangle \geq \tilde{\lambda}_\gamma \|\varphi\|^2$$

- Idea: control of $\|(1 - \Pi_p)\varphi\|^2$ by $\langle -\mathcal{L}_{\text{FD}}\varphi, \varphi \rangle$ (Poincaré); for $\|\Pi_p \varphi\|^2$,

$$\|\mathcal{L}_{\text{ham}} \Pi_p \varphi\|^2 \geq \frac{DC_\nu}{\beta m} \|\Pi_p \varphi\|^2, \quad \text{hence } A\mathcal{L}_{\text{ham}} \Pi_p \geq \lambda_{\text{ham}} \Pi_p$$

- Gronwall inequality $\frac{d}{dt} (\mathcal{H} [e^{t\mathcal{L}}\varphi]) = -\mathcal{D} [e^{t\mathcal{L}}\varphi] \leq -\frac{2\tilde{\lambda}_\gamma}{1+\varepsilon} \mathcal{H} [e^{t\mathcal{L}}\varphi]$

Extensions and modifications

- **General kinetic energy** function $U(p)$ in the Langevin dynamics
 - The generator \mathcal{L} **may not be hypoelliptic...** (even jump processes)
 - Convergence using Lyapunov¹¹ or hypocoercive¹² techniques
- **Nonequilibrium** Langevin dynamics
 - **Invariant measure not known...**
 - Perturbative results,¹³ with some uniformity on the range of perturbations¹⁴
 - Temperature accelerated molecular dynamics¹⁵

¹¹S. Redon, G. Stoltz and Z. Trstanova, *J. Stat. Phys.* (2016)

¹²G. Stoltz and Z. Trstanova, *SIAM MMS* (2018)

¹³E. Bouin, F. Hoffmann, and C. Mouhot, *SIAM J. Math. Anal.* (2017)

¹⁴A. Iacobucci, S. Olla and G. Stoltz, *Ann. Math. Quebec* (2019)

¹⁵G. Stoltz and E. Vanden-Eijnden, *Nonlinearity* (2018)

Direct L^2 hypocoercivity for modified Langevin (2)

- **Spectral discretization** of generator of Langevin dynamics
 - Approximate solutions of Poisson equation for control variates¹⁶
 - Bounds on convergence rates as a function of the basis size¹⁷
- **Adaptive** Langevin dynamics¹⁸ for mini-batching in large scale Bayesian inference
- **Current lines of work:**
 - An even more direct approach **avoiding the change of scalar product?** (with Antoine Levitt, Inria/CERMICS)
 - Quantitative bounds for **atom chains** (Very long term goal...)
 - **Non-perturbative** approach for nonequilibrium dynamics

¹⁶J. Roussel and G. Stoltz, *SIAM MMS* (2019)

¹⁷J. Roussel and G. Stoltz, *M2AN* (2018)

¹⁸Upcoming work with B. Leimkuhler and M. Sachs; currently I. Sekkat

Numerical approximation of Langevin dynamics

Practical computation of average properties

- Numerical scheme = **Markov chain** characterized by evolution operator

$$P_{\Delta t}\varphi(q, p) = \mathbb{E}\left(\varphi(q^{n+1}, p^{n+1}) \mid (q^n, p^n) = (q, p)\right)$$

- Discretization of the Langevin dynamics: **splitting** strategy

$$A = M^{-1}p \cdot \nabla_q, \quad B = -\nabla V(q) \cdot \nabla_p, \quad C = -M^{-1}p \cdot \nabla_p + \frac{1}{\beta}\Delta_p$$

- First order splitting schemes: $P_{\Delta t}^{ZYX} = e^{\Delta t Z} e^{\Delta t Y} e^{\Delta t X} \simeq e^{\Delta t \mathcal{L}}$

- Example: $P_{\Delta t}^{B,A,\gamma C}$ corresponds to (with $\alpha_{\Delta t} = \exp(-\gamma M^{-1}\Delta t)$)

$$\begin{cases} \tilde{p}^{n+1} = p^n - \Delta t \nabla V(q^n), \\ q^{n+1} = q^n + \Delta t M^{-1} \tilde{p}^{n+1}, \\ p^{n+1} = \alpha_{\Delta t} \tilde{p}^{n+1} + \sqrt{\frac{1 - \alpha_{\Delta t}^2}{\beta}} M G^n, \end{cases} \quad (1)$$

where G^n are i.i.d. standard Gaussian random variables

Practical computation of average properties (2)

- **Second order** splitting $P_{\Delta t}^{ZYXYZ} = e^{\Delta t Z/2} e^{\Delta t Y/2} e^{\Delta t X} e^{\Delta t Y/2} e^{\Delta t Z/2}$
- Example: $P_{\Delta t}^{\gamma C, B, A, B, \gamma C}$ (Verlet in the middle)

$$\left\{ \begin{array}{l} \tilde{p}^{n+1/2} = \alpha_{\Delta t/2} p^n + \sqrt{\frac{1 - \alpha_{\Delta t}}{\beta}} M G^n, \\ p^{n+1/2} = \tilde{p}^{n+1/2} - \frac{\Delta t}{2} \nabla V(q^n), \\ q^{n+1} = q^n + \Delta t M^{-1} p^{n+1/2}, \\ \tilde{p}^{n+1} = p^{n+1/2} - \frac{\Delta t}{2} \nabla V(q^{n+1}), \\ p^{n+1} = \alpha_{\Delta t/2} \tilde{p}^{n+1} + \sqrt{\frac{1 - \alpha_{\Delta t}}{\beta}} M G^{n+1/2}, \end{array} \right.$$

- Other category: **Geometric Langevin** algorithms, e.g. $P_{\Delta t}^{\gamma C, A, B, A}$

Error estimates on the computation of average properties

- The ergodicity of numerical schemes can be proved (bounded position domain):

$$\frac{1}{N_{\text{iter}}} \sum_{n=1}^{N_{\text{iter}}} \varphi(q^n, p^n) \xrightarrow{N_{\text{iter}} \rightarrow +\infty} \int \varphi(q, p) d\mu_{\gamma, \Delta t}(q, p)$$

- Statistical errors vs. systematic errors (**bias**)¹⁹

Systematic error estimates: α order of the splitting scheme

$$\int_{\mathcal{E}} \varphi(q, p) \mu_{\gamma, \Delta t}(dq dp) = \int_{\mathcal{E}} \varphi(q, p) \mu(dq dp) + \Delta t^\alpha \int_{\mathcal{E}} \varphi(q, p) f_{\alpha, \gamma}(q, p) \mu(dq dp) + O(\Delta t^{\alpha+1})$$

- Correction function $f_{\alpha, \gamma}$ solution of an appropriate **Poisson equation**

$$\mathcal{L}^* f_{\alpha, \gamma} = g_\gamma$$

where g_γ depends on the numerical scheme (adjoints taken on $L^2(\mu)$)

¹⁹B. Leimkuhler, Ch. Matthews and G. Stoltz, *IMA J. Numer. Anal.* (2016)

Proof for the first-order scheme $P_{\Delta t}^{\gamma C, B, A}$ (1)

- By definition of the invariant measure, $\int_{\mathcal{E}} P_{\Delta t} \phi d\mu_{\gamma, \Delta t} = \int_{\mathcal{E}} \phi d\mu_{\gamma, \Delta t}$, so

$$\int_{\mathcal{E}} \left[\left(\frac{\text{Id}_d - P_{\Delta t}}{\Delta t} \right) \phi \right] d\mu_{\gamma, \Delta t} = 0$$

- In view of the [BCH formula](#) $e^{\Delta t A_3} e^{\Delta t A_2} e^{\Delta t A_1} = e^{\Delta t A}$ with

$$A = A_1 + A_2 + A_3 + \frac{\Delta t}{2} \left([A_3, A_1 + A_2] + [A_2, A_1] \right) + \dots,$$

it holds $P_{\Delta t}^{\gamma C, B, A} = \text{Id}_d + \Delta t \mathcal{L} + \frac{\Delta t^2}{2} (\mathcal{L}^2 + S_1) + \Delta t^3 R_{1, \Delta t}$ with

$$S_1 = [C, A + B] + [B, A], \quad R_{1, \Delta t} = \frac{1}{2} \int_0^1 (1 - \theta)^2 \mathcal{R}_{\theta \Delta t} d\theta,$$

Proof for the first-order scheme $P_{\Delta t}^{\gamma C, B, A}$ (2)

- The **correction function** $f_{1,\gamma}$ is chosen so that

$$\int_{\mathcal{E}} \left[\left(\frac{\text{Id}_d - P_{\Delta t}^{\gamma C, B, A}}{\Delta t} \right) \phi \right] (1 + \Delta t f_{1,\gamma}) d\mu = O(\Delta t^2)$$

This requirement can be rewritten as

$$0 = \int_{\mathcal{E}} \left(\frac{1}{2} S_1 \phi + (\mathcal{L}\phi) f_{1,\gamma} \right) d\mu = \int_{\mathcal{E}} \varphi \left[\frac{1}{2} S_1^* \mathbf{1} + \mathcal{L}^* f_{1,\gamma} \right] d\mu,$$

which suggests to choose $\mathcal{L}^* f_{1,\gamma} = -\frac{1}{2} S_1^* \mathbf{1}$ (well posed equation)

- Replace ϕ by $\left(\frac{\text{Id}_d - P_{\Delta t}^{\gamma C, B, A}}{\Delta t} \right)^{-1} \varphi$? No control on the **derivatives**...
- Rely on the “nice” properties of the continuous dynamics, *i.e.* functional estimates²⁰ on \mathcal{L}^{-1} to use pseudo-inverses

$$Q_{1,\Delta t} = -\mathcal{L}^{-1} + \frac{\Delta t}{2} (\text{Id}_d + \mathcal{L}^{-1} S_1 \mathcal{L}^{-1})$$

²⁰D. Talay, Stoch. Proc. Appl. (2002); M. Kopec, arxiv 1310.2599 (2013)

Some start-up references

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