### Numerical analysis of stochastic Poisson systems

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### Outline

- I. Motivation
- II. Background material on SDEs
- III. Drift-preserving schemes for problems with additive noise
- IV. Splitting schemes for stochastic Poisson systems

I. Motivation

# **MOTIVATION**

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Consider deterministic Hamiltonian problems of the form (Hamilton 1834):

$$\dot{p}_k = -\frac{\partial H}{\partial q_k}(p,q), \quad \dot{q}_k = \frac{\partial H}{\partial p_k}(p,q), \text{ for } k = 1, \dots, d.$$

This system of differential equations describes the motion of a mechanical system with coordinates  $q_k$  and momenta  $p_k$ . Here  $p = p(t) = (p_1, ..., p_d)^T$ .

Examples: Molecular dynamics, motion of planets, mechanical systems, etc.

Remark: The Hamiltonian

$$H(p,q) = \frac{1}{2}p^T p + V(q)$$

is the total energy of the problem (kinetic energy plus potential energy).

Recall: Hamiltonian systems:

$$\dot{p}_k = -\frac{\partial H}{\partial q_k}(p,q), \quad \dot{q}_k = \frac{\partial H}{\partial p_k}(p,q), \text{ for } k = 1, \dots, d$$

with given initial values  $p(t_0) = p_{\text{init}}, q(t_0) = q_{\text{init}}$ .

Property: The total energy H(p, q) is an invariant:

$$\frac{\mathrm{d}}{\mathrm{d}t}H(p(t),q(t)) = \frac{\partial H}{\partial p}(p(t),q(t))\dot{p}(t) + \frac{\partial H}{\partial q}(p(t),q(t))\dot{q}(t) = 0$$
  
$$\Rightarrow H(p(t),q(t)) = \mathrm{Constant} = H(p_{\mathrm{init}},q_{\mathrm{init}})$$

along the exact solution.

Question: Design and analysis of energy-preserving numer. schemes for ODEs?

Answers 1996–: Brugnano, Celledoni, C., Gonzalez, Hairer, Iavernaro, McLachlan, McLaren, Miyatake, Owren, Quispel, Robidoux, Sato, Sun, Trigiante, Wang, Wu, Zhang, etc.

Recall: Hamiltonian system:

$$\dot{p}_k = -\frac{\partial H}{\partial q_k}(p,q), \quad \dot{q}_k = \frac{\partial H}{\partial p_k}(p,q), \quad \text{for} \quad k = 1, \dots, d.$$

with given initial values  $p(t_0) = p_{\text{init}}, q(t_0) = q_{\text{init}}$ .

Property: The flow  $\varphi_t(p_{\text{init}}, q_{\text{init}}) := (p(t, t_0, p_{\text{init}}, q_{\text{init}}), q(t, t_0, p_{\text{init}}, q_{\text{init}}))$  of the above problem is symplectic (*Poincaré* 1899):

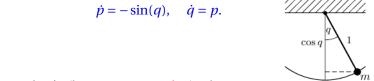
$$\varphi'_t(y)^T J \varphi'_t(y) = J$$
 for all  $y = (p, q)$ ,

where  $J = \begin{pmatrix} 0 & \text{Id} \\ -\text{Id} & 0 \end{pmatrix}$ .

Question: Design and analysis of symplectic numerical schemes?

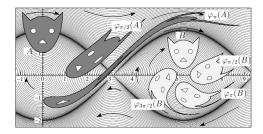
Answers 1956–: Bochev, de Vogelaere, Feng Kang, Hairer, Lasagni, Reich, Ruth, Sanz-Serna, Scovel, Suris, etc.

The mathematical pendulum has  $H(p,q) = \frac{1}{2}p^2 - \cos(q)$  and the Hamiltonian



(IV)

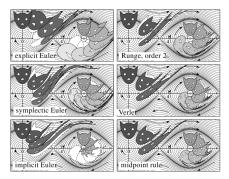
The flow is symplectic (here area preserving), phase space (q, p):



@ Book: Hairer, Wanner, Lubich, Geometric Numerical Integration 2006

# Symplectic integrators

### For ODE $\dot{y}(t) = f(y(t)), y(0) = y_0$ (the mathematical pendulum here):



Euler's scheme:  $y_{n+1} = y_n + hf(y_n) \approx y(t_{n+1})$  is not symplectic. The midpoint rule:  $y_{n+1} = y_n + hf\left(\frac{y_n + y_{n+1}}{2}\right) \approx y(t_{n+1})$  is symplectic. Important for long-term numerical simulations in molecular dynamics or planetary motions (movie click). Keyword: Backward Error Analysis.

@ Book: Hairer, Wanner, Lubich, Geometric Numerical Integration 2006

### Deterministic Poisson systems

Recall: Hamiltonian systems (setting y = (p, q)):  $\dot{y} = J^{-1}\nabla H(y)$ , with the skew-symmetric constant (symplectic) matrix  $J = \begin{pmatrix} 0 & Id \\ -Id & 0 \end{pmatrix}$ .

Given a Hamiltonian H and a matrix B(y) (satisfying some properties), the ODE

 $\dot{y} = B(y)\nabla H(y)$ 

is called a Poisson system. The matrix *B* is called the Poisson matrix.

Properties of the exact solution: The Hamiltonian is a conserved quantity. The flow of this ODE is a Poisson map (generalisation of symplecticity). One may have a Casimir function C (first integrals).

Question: Design and analysis of numerical schemes with such properties?

Answers 1988–: Channel, C., Ge, Hairer, Karasözen, McLachlan, Marsden, Reich, Scovel, etc.

### Deterministic free rigid body

**Recall:** Poisson problem:  $\dot{y} = B(y)\nabla H(y)$ .

The equations for a free rigid body reads

 $\dot{y}(t)=B(y(t))\nabla H(y(t)),$ 

where  $y = (y_1, y_2, y_3)^{\top}$  represents the angular momentum in the body frame,  $I = (I_1, I_2, I_3)$  are the principal moments of inertia and  $B(y) = \begin{pmatrix} 0 & -y_3 & y_2 \\ y_3 & 0 & -y_1 \\ -y_2 & y_1 & 0 \end{pmatrix}$ . The Hamiltonian  $H(y) = \frac{1}{2} (y_1^2/I_1 + y_2^2/I_2 + y_3^2/I_3)$  and the Casimir  $C(y) = \frac{1}{2} (y_1^2 + y_2^2 + y_3^2)$  are conserved quantities.

Further examples: Lotka–Volterra equation from population dynamics, discretisations of Euler's equations in fluid dynamics, etc.

Goal of presentation: Analyse (explicit) splitting integrators for random perturbations of Poisson systems.

A map C(y) is a Casimir for the Poisson ODE  $\dot{y} = B(y)\nabla H(y)$  if  $\nabla C(y)B(y) = 0$  for all y. Hence C(y) is also a first integral.

### II. Background material on SDEs



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### Stochastic differential equations (settings)

ODE. Given  $f : \mathbb{R} \to \mathbb{R}$  and an initial value x(0), we look for a solution to

$$\dot{x}(t) := \frac{\mathrm{d}x(t)}{\mathrm{d}t} = f(x(t)) \Longleftrightarrow x(t) - x(0) = \int_0^t f(x(s)) \,\mathrm{d}s.$$

SDE. Given  $f, g: \mathbb{R} \to \mathbb{R}$  and a (non-random) initial value  $X_0$ , a stochastic process  $X_t := X(t) = \{X_t(\omega)\}_{t \in [0,T]} = \{X(t,\omega)\}_{t \in [0,T]}$  is a solution to the SDE

 $dX_t = f(X_t)dt + g(X_t)dW_t$ , with initial value  $X_0$ 

if  $X_t$  solves the integral equation

$$X_t - X_0 = \int_0^t f(X_s) \, \mathrm{d}s + \int_0^t g(X_s) \, \mathrm{d}W_s.$$

Note:  $X_t$  is a stochastic process: i. e. a random variable for each time t (on some probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ ).

Note: Have to define  $W_t$  and the stochastic integral  $\int_0^t g(X_s) dW_s$ .

### **Brownian motion**

Definition. The stochastic process  $W_t$  is a Brownian motion or standard Wiener process over [0, T] if

- $W_0 = 0$  a.s.
- For any  $0 \le s < t \le T$ ,  $W_t W_s \sim N(0, t s)$  (normally dist.).

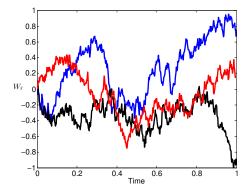


- Independent increments: For  $0 \le s \le t \le u \le v \le T$  the increments  $W_t W_s$  and  $W_v W_u$  are independent.
- $W_t$  has a.s. cont. samples  $\wedge$  nowhere diff.

 $\implies$  At any time t,  $W_t$  is a random variable:  $W_t = W_t - W_0 \sim N(0, t)$  and so  $\mathbb{E}[W_t] = 0$  and  $\mathbb{E}[W_t^2] = t$ .

### **Brownian motion**

Numerical illustration: Discretised Brownian paths over [0, 1].



 $\implies$   $W_t$  is continuous but nowhere differentiable!!

### Stochastic differential equations

Recall SDE: Given  $f, g: \mathbb{R} \to \mathbb{R}$  and a (non-random) initial value  $X_0$ , a stochastic process  $X_t$  is a solution to the SDE

 $\mathrm{d}X_t = f(X_t)\mathrm{d}t + g(X_t)\mathrm{d}W_t$ 

if  $X_t$  solves the integral equation

$$X_t - X_0 = \int_0^t f(X_s) \, \mathrm{d}s + \int_0^t g(X_s) \, \mathrm{d}W_s.$$

Note:  $W_s$  not differentiable (not even finite variation) so that we have to be careful with the definition of the above stochastic integral.

### Stochastic integrals

For a deterministic function  $h: \mathbb{R} \to \mathbb{R}$  and a partition  $t_n = n\delta t$  with  $\delta t = T/N$ , one defines:

Deterministic Riemann integrals

$$\int_{0}^{T} h(t) dt = \lim_{\delta t \to 0} \sum_{n=0}^{N-1} h(t_n)(t_{n+1} - t_n)$$
$$= \lim_{\delta t \to 0} \sum_{n=0}^{N-1} h\left(\frac{t_n + t_{n+1}}{2}\right)(t_{n+1} - t_n)$$

Stochastic Itô integrals for stochastic process h(t) (left endpoints)

$$\int_0^T h(t) \, \mathrm{d}W_t \stackrel{L^2}{=} \lim_{\delta t \to 0} \sum_{n=0}^{N-1} h(t_n) \underbrace{(W_{t_{n+1}} - W_{t_n})}_{\sim N(0, t_{n+1} - t_n)}.$$

Stochastic Stratonovich integrals for stochastic process h(t) (midpoint)

$$\int_0^T h(t) \circ dW_t \stackrel{L^2}{=} \lim_{\delta t \to 0} \sum_{n=0}^{N-1} h\left(\frac{t_n + t_{n+1}}{2}\right) \underbrace{(W_{t_{n+1}} - W_{t_n})}_{\sim N(0, t_{n+1} - t_n)}.$$

### III. Drift-preserving schemes for problems with additive noise



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### Stochastic Poisson problem

For an integer m > 0 and a nice potential  $V : \mathbb{R}^m \to \mathbb{R}$ , consider the separable Hamiltonian

$$H(p,q) = \frac{1}{2} \sum_{j=1}^{m} p_j^2 + V(q).$$

Problem: Set X(t) = (p(t), q(t)) and consider Poisson system with additive noise:

$$\mathrm{d}X(t) = B(X(t))\nabla H(X(t))\,\mathrm{d}t + \begin{pmatrix} \Sigma \\ 0 \end{pmatrix} \mathrm{d}W(t).$$

Here,  $B(X) \in \mathbb{R}^{2m \times 2m}$  is a smooth skew-symmetric matrix,  $\Sigma \in \mathbb{R}^{m \times d}$  and  $W(t) \in \mathbb{R}^{d}$ .

### Examples:

Generalisation of stoch. Hamilton systems taking

 $B(X) = J^{-1} = \begin{pmatrix} 0 & -Id_m \\ Id_m & 0 \end{pmatrix}$  constant matrix. Obs: odd dimension also ok! Stochastic free rigid body (*B*(*X*) not constant), Lotka–Volterra systems, etc.

### Stochastic Poisson problem

(II)

Recall: Poisson system with additive noise:

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\mathrm{d}X(t) = B(X(t))\nabla H(X(t))\,\mathrm{d}t + \begin{pmatrix} \Sigma \\ 0 \end{pmatrix} \mathrm{d}W(t).
```

Proposition (C., Vilmart 20<sup>\*</sup>,21): Trace formula for the energy: Along the exact solution to the above SDE, one has

$$\mathbb{E}[H(X(t))] = \mathbb{E}[H(X_0)] + \frac{1}{2}\operatorname{Tr}(\Sigma^{\top}\Sigma)t \quad \text{for all time} \quad t > 0.$$

The proof is done using Ito's formula.

Question: What about numerical discretisation?

Drift-preserving scheme for stochastic Poisson problem (I)

Recall: Poisson system with additive noise:

$$\mathrm{d}X(t) = B(X(t))\nabla H(X(t))\,\mathrm{d}t + \begin{pmatrix} \Sigma \\ 0 \end{pmatrix} \mathrm{d}W(t).$$

Based on a splitting idea, we propose a new drift-preserving scheme for stochastic Poisson problem:

$$Y_{1} := X_{n} + {\binom{\Sigma}{0}} \left( W(t_{n} + \frac{h}{2}) - W(t_{n}) \right),$$
  

$$Y_{2} := Y_{1} + hB \left( \frac{Y_{1} + Y_{2}}{2} \right) \int_{0}^{1} \nabla H(Y_{1} + \theta(Y_{2} - Y_{1})) d\theta,$$
  

$$X_{n+1} = Y_{2} + {\binom{\Sigma}{0}} \left( W(t_{n+1}) - W(t_{n} + \frac{h}{2}) \right),$$

where h > 0 is the stepsize of the numerical scheme and  $t_n = nh$ .

### Drift-preserving scheme for stochastic Poisson problem (II)

Recall: The exact solution to the Poisson system with additive noise

$$\mathrm{d}X(t) = B(X(t))\nabla H(X(t))\,\mathrm{d}t + \begin{pmatrix} \Sigma \\ 0 \end{pmatrix} \mathrm{d}W(t).$$

has the trace formula for the energy

$$\mathbb{E}[H(X(t))] = \mathbb{E}[H(X_0)] + \frac{1}{2}\operatorname{Tr}(\Sigma^{\top}\Sigma)t \quad \text{for all time} \quad t > 0.$$

#### Our splitting scheme satisfies:

Theorem (C., Vilmart 20<sup>\*</sup>, 21): Numerical trace formula for the energy

$$\mathbb{E}[H(X_n)] = \mathbb{E}[H(X_0)] + \frac{1}{2} \operatorname{Tr}(\Sigma^{\top} \Sigma) t_n \quad \text{for all discrete times} \quad t_n = nh,$$

where  $n \in \mathbb{N}$ .

# Drift-preserving scheme for stochastic Poisson problem (III)

To show: Drift-preserving scheme:  $\mathbb{E}[H(X_n)] = \mathbb{E}[H(X_0)] + \frac{1}{2} \operatorname{Tr}(\Sigma^{\top} \Sigma) t_n$ . The first step of the drift-preserving scheme can be rewritten as

$$Y_1 = X_n + \int_{t_n}^{t_n + \frac{h}{2}} {\binom{\Sigma}{0}} dW(s)$$

and an application of Itô's formula gives

$$\mathbb{E}[H(Y_1)] = \mathbb{E}[H(X_n)] + \frac{h}{4}\operatorname{Tr}(\Sigma^{\top}\Sigma).$$

Second step of the scheme is a deterministic energy-preserving scheme:

 $\mathbb{E}[H(Y_2)] = \mathbb{E}[H(Y_1)].$ 

The last step of the numerical integrator gives

$$\mathbb{E}[H(X_{n+1})] = \mathbb{E}[H(Y_2)] + \frac{h}{4}\operatorname{Tr}(\Sigma^{\top}\Sigma) = \mathbb{E}[H(Y_1)] + \frac{h}{4}\operatorname{Tr}(\Sigma^{\top}\Sigma)$$
$$= \mathbb{E}[H(X_n)] + \frac{h}{2}\operatorname{Tr}(\Sigma^{\top}\Sigma).$$

A recursion now completes the proof.

## Drift-preservation of Casimirs

If the original ODE has a quadratic Casimir  $C(X) = \frac{1}{2}X^{\top}AX$ , with a symmetric constant matrix  $A = \begin{pmatrix} a & b \\ b^{\top} & c \end{pmatrix}$  with  $a, b, c \in \mathbb{R}^{m \times m}$ , then

Theorem (C., Vilmart 20<sup>\*</sup>,21): Trace formula for the Casimir (exact solution)

$$\mathbb{E}[C(X(t))] = \mathbb{E}[C(X_0)] + \frac{1}{2}\operatorname{Tr}(\Sigma^{\top} a\Sigma) t \text{ for all time } t > 0.$$

Numerical trace formula for the Casimir (numerical solution)

 $\mathbb{E}[C(X_n)] = \mathbb{E}[C(X_0)] + \frac{1}{2} \operatorname{Tr}(\Sigma^{\top} a \Sigma) t_n \quad \text{for all discrete times} \quad t_n = nh,$ where  $n \in \mathbb{N}$ .

A map C(X) is a Casimir for the Poisson ODE  $\dot{X} = B(X)\nabla H(X)$  if  $\nabla C(X)B(X) = 0$  for all X. Hence C(X) is also a first integral.

Rates of convergence of the drift-preserving scheme

The proposed drift-preserving scheme has the following rates of convergence under the standard setting.

Theorem (C., Vilmart 20<sup>\*</sup>,21): Mean-square order of convergence 1:

$$\left(\mathbb{E}[\|X(t_n) - X_n\|^2]\right)^{1/2} \le Ch.$$

Weak convergence of order 2:

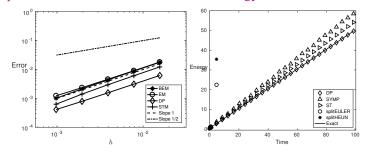
 $|\mathbb{E}[\Phi(X(t_n))] - \mathbb{E}[\Phi(X_n)]| \le Ch^2,$ 

for all test functions  $\Phi \in C_p^6(\mathbb{R}^{2m},\mathbb{R})$ , the space of  $C^6$  functions with all derivatives up to order 6 with at most polynomial growth.

### Linear stochastic oscillator

Problem:  $dX(t) = B(X(t))\nabla H(X(t)) dt + \Sigma dW(t)$ , where X = (p, q),  $H(p, q) = \frac{1}{2}p^2 + \frac{1}{2}q^2$  and with  $\Sigma = 1$  and W scalars. For this problem, the drift-preserving scheme is an explicit time integrator!

Mean-square error and trace formula for the energy:



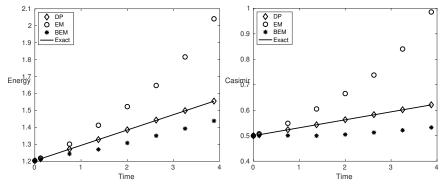
Drift-preserving scheme (DP), the splitting methods with the symplectic Euler method (SYMP), the Störmer–Verlet method (ST), the explicit Euler method (splitEULER), or the Heun method (splitHEUN).

Parameters: (p(0), q(0)) = (0, 1), time interval [0, 100] with 2<sup>7</sup> step sizes,  $M_s = 10^6$  samples.

### Stochastic rigid body

Problem:  $dX(t) = B(X(t))\nabla H(X(t)) dt + \Sigma dW(t)$ , where Hamiltonian  $H(X) = \frac{1}{2} (X_1^2/I_1 + X_2^2/I_2 + X_3^2/I_3)$ , and quadratic Casimir  $C(X) = \frac{1}{2} (X_1^2 + X_2^2 + X_3^2)$ , with  $\Sigma$  and W scalars (acting on first component). Here,  $X = (X_1, X_2, X_3)^{\top}$  and moments of inertia  $I = (I_1, I_2, I_3)$ .

Trace formula for the energy and the Casimir:



Parameters: X(0) = (0.8, 0.6, 0) and I = (0.345, 0.653, 1), stepsizes  $h = 4/2^5$ , time interval [0,4],  $M_s = 2 \cdot 10^6$  samples.

## IV. Splitting schemes for stochastic Poisson systems



Thanks to www.images.google.com

### Stochastic Lie–Poisson problems

Consider stochastic Poisson systems of the form

$$\begin{cases} dy(t) = B(y(t))\nabla H(y(t)) dt + \sum_{k=1}^{m} B(y(t))\nabla \widehat{H}_{k}(y(t)) \circ dW_{k}(t), \\ y(0) = y_{0}, \end{cases}$$

with Hamiltonian functions  $H, \hat{H}_1, \ldots, \hat{H}_m \colon \mathbb{R}^d \to \mathbb{R}$ , with structure matrix  $B \colon \mathbb{R}^d \to \mathbb{R}^{d \times d}$ , and with independent standard real-valued Wiener processes  $W_1, \ldots, W_m$ .

Skew-symmetry: for every  $y \in \mathbb{R}^d$  and for all  $i, j \in \{1, ..., d\}$ , one has

 $B_{ij}(y) = -B_{ji}(y).$ 

■ Jacobi identity: for every  $y \in \mathbb{R}^d$  and for all  $i, j, k \in \{1, ..., d\}$ , one has

$$\sum_{\ell=1}^{d} \left( \frac{\partial B_{ij}(y)}{\partial y_{\ell}} B_{\ell k}(y) + \frac{\partial B_{jk}(y)}{\partial y_{\ell}} B_{\ell i}(y) + \frac{\partial B_{ki}(y)}{\partial y_{\ell}} B_{\ell j}(y) \right) = 0.$$

■ Lie–Poisson systems: *B* depends linearly on *y*.

### Main results

Recall: SDE  $dy(t) = B(y(t))\nabla H(y(t)) dt + \sum_{k=1}^{m} B(y(t))\nabla \widehat{H}_k(y(t)) \circ dW_k(t).$ 

Under technical assumptions, we:

- Prove that the flow of this SDE is a Poisson map: One has a.s., for all y, that  $\varphi'_t(y)B(y)\varphi'_t(y)^{\top} = B(\varphi_t(y))$ .
- Derive and analyse explicit splitting Poisson integrators for particular stochastic Lie–Poisson systems.
- Prove strong and weak convergence of such integrators, even when the coefficients of the problem are not globally Lipschitz continuous.
- Study asymptotic preserving schemes in the diffusion approximation regime.

Bréhier, C., Jahnke 2021\*, 2023.

### The Poisson map property

**Recall:** SDE  $dy(t) = B(y(t))\nabla H(y(t)) dt + \sum_{k=1}^{m} B(y(t))\nabla \widehat{H}_k(y(t)) \circ dW_k(t).$ 

\*skip\*

**Definition:** Let  $D_y$  denote the Jacobian operator. Let  $U \subset \mathbb{R}^d$  be an open set. A transformation  $\varphi: U \to \mathbb{R}^d$  is called a *Poisson map* for the above SDE, if one has, almost surely, for all  $y \in \mathbb{R}^d$ ,

 $D_y \varphi(y) B(y) D_y \varphi(y)^T = B(\varphi(y)).$ 

Remark: Observe that a composition of Poisson maps is a Poisson map.

Theorem: Introduce the flow  $(t, y) \rightarrow \varphi_t(y)$  of the above SDE with coefficients of class  $\mathscr{C}^3$ . Assume that the flow is globally well defined and of class  $\mathscr{C}^1$  with respect to the variable *y*. Then, for all  $t \ge 0$ ,  $\varphi_t$  is a Poisson map: almost surely, for all  $y \in \mathbb{R}^d$ , one has

 $D_y \varphi_t(y) B(y) D_y \varphi_t(y)^T = B(\varphi_t(y)).$ 

Remark: *Hong, Ruan, Sun, Wang* 21: Proof needs Darboux–Lie theorem and to rewrite SDE. Their Poisson integrators in turn need transformations and are usually implicit.

# Stoch. Poisson integrators based on splitting schemes Recall: SDE $dy(t) = B(y(t))\nabla H(y(t)) dt + \sum_{k=1}^{m} B(y(t))\nabla \hat{H}_k(y(t)) \circ dW_k(t)$ . Assumption: The Hamiltonian *H* can be split as follows: $H = \sum_{k=1}^{p} H_k$ for some $p \ge 1$ .

Let h > 0 be the time step size. A numerical scheme is defined as

$$y^{[n]} = \Phi_h(y^{[n-1]}, \Delta_n W_1, \dots, \Delta_n W_m),$$

with Wiener increments  $\Delta_n W_k = W_k(nh) - W_k((n-1)h)$ , k = 1, ..., m. Provides numerical approximations:  $y^{[n]} \approx y(nh)$ .

Splitting schemes:

 $\Phi_{h}(\cdot) = \Phi_{h}(\cdot, \Delta W_{1}, \dots, \Delta W_{m}) = \exp(hY_{H_{p}}) \circ \exp(hY_{H_{p-1}}) \circ \dots \circ \exp(hY_{H_{1}})$  $\circ \exp(\Delta W_{m}Y_{\widehat{H}_{m}}) \circ \exp(\Delta W_{m-1}Y_{\widehat{H}_{m-1}}) \circ \dots \circ \exp(\Delta W_{1}Y_{\widehat{H}_{1}}),$ 

where  $Y_{H_k} = B\nabla H_k$ , resp.  $Y_{\widehat{H}_k} = B\nabla \widehat{H}_k$ , denote the vector fields of the corresponding differential equations.

### Convergence of the Lie–Poisson splitting schemes

Recall: SDE  $dy(t) = B(y(t))\nabla H(y(t)) dt + \sum_{k=1}^{m} B(y(t))\nabla \widehat{H}_{k}(y(t)) \circ dW_{k}(t)$ . Splitting scheme:  $\Phi_{h}(\cdot) = \exp(hY_{H_{p}}) \circ \ldots \circ \exp(\Delta W_{m}Y_{\widehat{H}_{m}}) \circ \ldots \circ \exp(\Delta W_{1}Y_{\widehat{H}_{1}})$ .

Theorem: Assume SDE admits a Casimir function with compact level sets. Strong convergence. Assume  $B \in \mathscr{C}^2$ ,  $H_1, \ldots, H_p \in \mathscr{C}^2$ , and  $\hat{H}_1, \ldots, \hat{H}_m \in \mathscr{C}^3$ . Then the splitting scheme has strong order of convergence equal to 1/2: for all  $T \in (0, \infty)$  and all  $y_0 \in \mathbb{R}^d$ , there exists a real number  $c(T, y_0) \in (0, \infty)$  such that

$$\sup_{0 \le n \le N} \left( \mathbb{E} \left[ \| y(nh) - y^{[n]} \|^2 \right] \right)^{1/2} \le c(T, y_0) h^{\frac{1}{2}}$$

with time step size h = T/N, and  $y^{[0]} = y_0 = y(0)$ .

Weak convergence. Assume  $B \in \mathscr{C}^5$ ,  $H_1, \ldots, H_p \in \mathscr{C}^5$ , and  $\hat{H}_1, \ldots, \hat{H}_m \in \mathscr{C}^6$ . Then the splitting scheme has weak order of convergence equal to 1: for all  $T \in (0, \infty)$  and all  $y_0 \in \mathbb{R}^d$ , and any test function  $\phi \colon \mathbb{R}^d \to \mathbb{R}$  of class  $\mathscr{C}^4$  with bounded derivatives, there exists a real number  $c(T, y_0, \phi) \in (0, \infty)$  such that

 $\sup_{0 \le n \le N} \left| \mathbb{E} \left[ \phi \left( y(nh) \right) \right] - \mathbb{E} \left[ \phi \left( y^{[n]} \right) \right] \right| \le c(T, y_0, \phi) h.$ 

### Main steps for the proofs

Recall: SDE  $dy(t) = B(y(t))\nabla H(y(t)) dt + \sum_{k=1}^{m} B(y(t))\nabla \hat{H}_{k}(y(t)) \circ dW_{k}(t)$ . Splitting scheme:  $\Phi_{h}(\cdot) = \exp(hY_{H_{p}}) \circ \ldots \circ \exp(\Delta W_{m}Y_{\hat{H}_{m}}) \circ \ldots \circ \exp(\Delta W_{1}Y_{\hat{H}_{1}})$ . Strong order 1/2, weak order 1.

1 Show a.s bounds for the exact and numerical solutions:

 $\sup_{t \in [0,T]} \|y(t)\| \le R(y_0), \quad \sup_{N \ge 1} \sup_{0 \le n \le N} \|y^{[n]}\| \le R(y_0),$ 

\*skip\*

where R(y<sub>0</sub>) = max<sub>y∈ℝ<sup>d</sup>,C(y)=C(y<sub>0</sub>) ||y||, and R(y<sub>0</sub>) < ∞.</li>
Use: Splitting scheme is a Poisson integrator hence preserve Casimir C.
2 Show strong and weak convergence for the auxiliary problem
</sub>

$$dz(t) = \sum_{k=1}^{p} f_k(z(t)) dt + \sum_{k=1}^{m} \widehat{f}_k(z(t)) \circ dW_k(t),$$

with smooth globally Lipschitz continuous functions f<sub>k</sub> and f<sub>k</sub>.
Use: Fundamental theorem by Milstein and the Talay–Tubaro argument.
Conclude to show the convergence results for the above Poisson systems.
Use: Combine above two steps.

## Stochastic Maxwell–Bloch equations

Problem: Let d = 3. The deterministic Maxwell–Bloch equations from laser-matter dynamics read

 $\begin{cases} \dot{y}_1 = y_2 \\ \dot{y}_2 = y_1 y_3 \\ \dot{y}_3 = -y_1 y_2. \end{cases}$ 

This system is a deterministic Lie–Poisson system with Poisson matrix, Hamiltonian and Casimir functions given by

$$B(y) = \begin{pmatrix} 0 & -y_3 & y_2 \\ y_3 & 0 & 0 \\ -y_2 & 0 & 0 \end{pmatrix}, \quad H(y) = \frac{1}{2}y_1^2 + y_3, \quad C(y) = \frac{1}{2}(y_2^2 + y_3^2),$$

respectively, for all  $y = (y_1, y_2, y_3) \in \mathbb{R}^3$ .

Consider the following stochastic version of the Maxwell-Bloch system:

 $dy = B(y) \left( \nabla H(y) dt + \sigma_1 \nabla \hat{H}_1(y) \circ dW_1(t) + \sigma_3 \nabla \hat{H}_3(y) \circ dW_3(t) \right),$ where  $\hat{H}_1(y) = \frac{1}{2}y_1^2$  and  $\hat{H}_3(y) = y_3, \sigma_1, \sigma_3 \ge 0$ , driven by two independent Wiener processes  $W_1$  and  $W_3$ .

### Stochastic Maxwell–Bloch equations

(II)

Recall:  $dy = B(y) (\nabla H(y) dt + \sigma_1 \nabla \hat{H}_1(y) \circ dW_1(t) + \sigma_3 \nabla \hat{H}_3(y) \circ dW_3(t)),$ where  $H(y) = \frac{1}{2}y_1^2 + y_3$  with  $H_1(y) = \hat{H}_1(y) = \frac{1}{2}y_1^2$  and  $H_3(y) = \hat{H}_3(y) = y_3.$ 

The Hamiltonian *H* is split as  $H = H_1 + H_3$ .

The two associated deterministic subsystems can be solved exactly as follows: The deterministic subsystem corresponding with the vector field  $Y_{H_1} = B\nabla H_1$ is given by

$$\dot{y}_1 = 0$$
  
 $\dot{y}_2 = y_3 y_1$   
 $\dot{y}_3 = -y_2 y_1.$ 

Observe that  $y_1$  is constant and thus  $(y_2, y_3)$  is solution to the standard harmonic oscillator: the exact solution of the first subsystem is thus given by

$$\exp(tY_{H_1})y(0) = \begin{pmatrix} 1 & 0 & 0\\ 0 & \cos(y_1(0)t) & \sin(y_1(0)t)\\ 0 & -\sin(y_1(0)t) & \cos(y_1(0)t) \end{pmatrix} y(0)$$

for all  $t \in \mathbb{R}$  and  $y(0) \in \mathbb{R}^3$ . Obs.  $Y_{H_3} = B\nabla H_3$  and stoch. parts ok!

### Stochastic Maxwell–Bloch equations

**Recall:**  $dy = B(y) \left( \nabla H(y) dt + \sigma_1 \nabla \widehat{H}_1(y) \circ dW_1(t) + \sigma_3 \nabla \widehat{H}_3(y) \circ dW_3(t) \right).$ The splitting integrator then reads

 $\Phi_h = \exp(hY_{H_3}) \circ \exp(hY_{H_1}) \circ \exp(\sigma_3 \Delta W_3 Y_{\hat{H}_3}) \circ \exp(\sigma_1 \Delta W_1 Y_{\hat{H}_1}),$ 

where for all  $y \in \mathbb{R}^3$  one has

and

$$\exp(\sigma_{1}\Delta W_{1}Y_{\widehat{H}_{1}})y = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos(y_{1}\sigma_{1}\Delta W_{1}) & \sin(y_{1}\sigma_{1}\Delta W_{1}) \\ 0 & -\sin(y_{1}\sigma_{1}\Delta W_{1}) & \cos(y_{1}\sigma_{1}\Delta W_{1}) \end{pmatrix} y$$
$$\exp(\sigma_{3}\Delta W_{3}Y_{\widehat{H}_{3}})y = \begin{pmatrix} 1 & \sigma_{3}\Delta W_{3} & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} y.$$

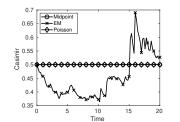
**Remark**: This explicit splitting scheme is a stochastic Poisson integrator: the numerical map is a Poisson map and it preserves all Casimirs of the SDE.

Free rigid body with random inertia tensor

Problem: Let 
$$H(y) = \sum_{k=1}^{3} \frac{y_k^2}{I_k}$$
,  $\hat{H}_k(y) = \frac{y_k^2}{\hat{I}_k}$ , for  $k = 1, 2, 3$ , and consider

$$d\begin{pmatrix} y_1\\y_2\\y_3 \end{pmatrix} = B(y) \left( \nabla H(y) dt + \nabla \widehat{H}_1(y) \circ dW_1(t) + \nabla \widehat{H}_2(y) \circ dW_2(t) \right)$$
$$+ \nabla \widehat{H}_3(y) \circ dW_3(t) \right).$$

Casimir: The above SDE has a conserved quantity, the Casimir:  $C(y) = y_1^2 + y_2^2 + y_3^2.$ 





 $(\mathbf{I})$ 

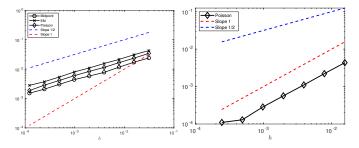
Free rigid body with random inertia tensor

Problem: Let 
$$H(y) = \sum_{n=1}^{3} \frac{y_n^2}{I_n}$$
,  $\widehat{H}_k(y) = \frac{y_k^2}{\widehat{I}_k}$ , for  $k = 1, 2, 3$ , and consider

$$d\begin{pmatrix} y_1\\y_2\\y_3 \end{pmatrix} = B(y) \left( \nabla H(y) \, \mathrm{d}t + \nabla \widehat{H}_1(y) \circ \mathrm{d}W_1(t) + \nabla \widehat{H}_2(y) \circ \mathrm{d}W_2(t) \right)$$
$$+ \nabla \widehat{H}_3(y) \circ \mathrm{d}W_3(t) \right).$$

(II)

### Strong and weak convergence:

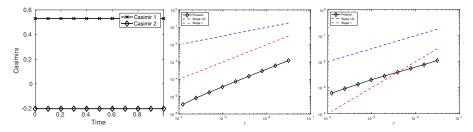


### Stoch. Poisson integrators for the stochastic sine–Euler syst.

The sine–Euler equations consist of a finite-dimensional truncation of the two-dimensional Euler equations in fluid dynamics (Zeitlin 1991).

We consider random perturbations of such systems (horrible equations).

2 Casimirs (quadratic and cubic), ms order 1 (one noise), ms order 1/2 (3 noises):



## Thanks for your attention!!



David Cohen, Gilles Vilmart: *Drift-preserving numerical integrators for stochastic Poisson systems*, 2020\*, Int. J. Comput. Math, 2021 Charles-Edouard Bréhier, David Cohen, Tobias Jahnke: *Splitting integrators for stochastic Lie–Poisson systems*, 2021\*, to appear Math. Comp 2023



Thanks to www.images.google.com and Konstantinos Dareiotis

### Strang version

SDE

$$dy(t) = B(y(t))\nabla H(y(t)) dt + \sum_{k=1}^{m} B(y(t))\nabla \widehat{H}_{k}(y(t)) \circ dW(t)$$

with 
$$H(y) = \sum_{k=1}^{p} H_k$$
.  
Strang version

 $\Phi_{h}(\cdot) = \exp(hY_{H_{p}}) \circ \dots \circ \exp(hY_{H_{1}})$  $\circ \exp(\Delta W/2Y_{\widehat{H}_{1}}) \circ \dots \exp(\Delta WY_{\widehat{H}_{w}}) \circ \dots \circ \exp(\Delta W/2Y_{\widehat{H}_{1}})$ 

Order: 1/2 in mean-square sense and 2 in the weak sense.