

# Hybrid Monte Carlo on Hilbert spaces

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## I. A REVIEW OF THE STANDARD HMC ON $\mathbb{R}^N$

- Aim: obtain samples  $q^{(n)}$ ,  $n = 0, 1, \dots$ , from a probability density  $\pi \propto \exp(-V(q))$  in  $\mathbb{R}^N$ .
- Write  $V(q) = \frac{1}{2}\langle q, Lq \rangle + \Phi(q)$ ,  $L$  sym. pos. semdef. (perhaps  $L = 0$ ), so that  $\pi \propto \exp\left(\frac{1}{2}\langle q, Lq \rangle + \Phi(q)\right)$ .
- HMC (Duane *et al* 1987) includes three ingredients:

A Hamiltonian flow in  $\mathbb{R}^{2N}$ .

A numerical integrator for that flow.

An accept/reject rule.

## Hamiltonian flow in $\mathbb{R}^{2N}$

- Introduce Hamiltonian  $H(q, p) = \frac{1}{2}\langle p, M^{-1}p \rangle + \frac{1}{2}\langle q, Lq \rangle + \Phi(q)$ .  $M$  is a user specified sym. pos. def. ('mass') matrix,  $p$  auxiliary variable ('momentum'),  $H$  total energy (kinetic+potential).

- Associated canonical ODEs are

$$\frac{dq}{dt} = \frac{\partial H}{\partial p} = M^{-1}p, \quad \frac{dp}{dt} = -\frac{\partial H}{\partial q} = -Lq + f(q), \quad f = -\nabla\Phi.$$

- For any fixed  $t$ , the  $t$ -flow  $\Xi^t$  of the system

$$\Xi^t(q(0), p(0)) = (q(t), p(t))$$

preserves the volume-element  $dq dp$  and the value of  $H$ .

- Therefore  $\Xi^t$  also preserves the measure in the phase space  $\mathbb{R}^{2N}$  with density  $\Pi(q, p) \propto \exp(-H(q, p))$ , or

$$\Pi(q, p) \propto \exp\left(-\frac{1}{2}\langle p, M^{-1}p \rangle\right) \exp\left(-\frac{1}{2}\langle q, Lq \rangle - \Phi(q)\right)$$

...and by implication the  $q$ -marginal, ie our target  $\pi$ . (The marginal of  $p$  is  $N(0, M)$ .)

- Hence if  $q(0)$  is distributed according to target  $\pi$  and we draw  $p(0) \sim N(0, M)$ , then  $q(t)$  will also be distributed according to  $\pi$ .
- This suggests the following ...

- ‘*Idea for an algorithm*’ ( $T$  and  $M$  have been fixed.)
  - Given  $q^{(n)}$ , draw  $p^{(n)} \sim N(0, M)$ .
  - Find  $(q^*, p^*) = \Xi^T(q^{(n)}, p^{(n)})$ .
  - Set  $q^{(n+1)} = q^*$ , discard  $p^*$ ,  $n \leftarrow n + 1$ .
- The transition  $q^{(n)} \mapsto q^{(n+1)}$  defines a Markov chain that has the target  $\pi$  as invariant distribution.
- The chain is reversible. Transitions are non-local.

## A numerical integrator for the canonical equations

- In practice analytic expression of  $\Xi^t$  not available and one has to resort to numerical approximations.
- Integrator must be **volume preserving** and **reversible**.
- Verlet is method of choice. Convenient to see it here as a splitting algorithm  $\psi_h = \Xi_1^{h/2} \circ \Xi_2^h \circ \Xi_1^{h/2}$ , where  $\Xi_1^t$  and  $\Xi_2^t$  are the flows of the canonical systems with Hamiltonian functions

$$H_1 = \frac{1}{2}\langle q, Lq \rangle + \Phi(q) , \quad H_2 = \frac{1}{2}\langle p, M^{-1}p \rangle, \quad H = H_1 + H_2.$$

## Accept/reject

- Since Verlet does not conserve  $H$  exactly, numerical solution does not preserve  $\Pi$ .
- Exact conservation is enforced through the Metropolis-Hastings rule: if  $(q^{(n)}, p^{(n)})$  is current state of chain and  $(q^*, p^*)$  ('the proposal') is the numerically computed approximation at the end of a time-interval of length  $T$ , then compute

$$a = \min\left(1, \exp\left(H(q^{(n)}, p^{(n)}) - H(q^*, p^*)\right)\right) .$$

- Set  $q^{(n+1)} = q^*$  with probability  $a$  ('acceptance'); otherwise set  $q^{(n+1)} = q^{(n)}$  ('rejection').



## Choice of mass matrix (Crucial for efficiency.)

- From canonical equations:

$$\frac{d^2q}{dt^2} = -M^{-1}Lq + M^{-1}f(q).$$

- Consider case  $f \equiv 0$  and  $L$  is pos. def. and has some very large eigenvalues ( $\pi$  is normal with some very small variances),

choice  $M = I$  implies inefficiency (Verlet step  $h$  adjusted to smallest variance of  $\pi$ /largest eigenvalue of  $L$ ).

choice  $M = L$  (large mass in directions of large force)  
then  $d^2q/dt^2 = -q$ . Both  $q$  and 'velocity'  $v = dq/dt$  are slowly varying;  $v$  is of the size of  $q$  ( $p = Mv = Lv$  is large).

- This suggests to use  $M = L$  and to write dynamics for Verlet integration in  $q, v$  variables:

$$\frac{dq}{dt} = v, \quad \frac{dv}{dt} = -q + L^{-1}f(q).$$

- $v \sim N(0, L^{-1})$  and initial value should be drawn accordingly.
- Also in accept/reject compute  $H$  in terms of  $v$ :

$$H = \frac{1}{2}\langle v, Lv \rangle + \frac{1}{2}\langle q, Lq \rangle + \Phi(q).$$

## II. THE PROBLEM

- $\pi_0$  a non-degenerate (non-Dirac) centred Gaussian measure with covariance operator  $\mathcal{C}$  in (infinite-dimensional, separable) Hilbert space  $\mathcal{H}$ .

[ $\mathcal{C}$  is a positive, self-adjoint, nuclear (ie its eigenvalues are summable), its eigenfunctions span  $\mathcal{H}$ . (eg  $\mathcal{C}$  inverse Laplacian in  $L^2 + \text{bd}$ ; corresponding to Brownian motion and Brownian bridge.)]

- **Aim** is to sample from a probability measure  $\pi$  defined by its density with respect to  $\pi_0$ :

$$\frac{d\pi}{d\pi_0}(q) \propto \exp(-\Phi(q)).$$

( $\Phi$  small relative to  $\mathcal{C}^{-1}$ .)

- Sources: conditioned diffusion; Bayesian approach to inverse problems . . .

- May also consider case where  $\mathcal{H}$  is replaced by  $\mathbb{R}^N$  with very large  $N$  and  $\mathcal{C}$  by a sym. pos. def. matrix with some of its eigenvalues close to 0. (High-dimensional Gaussian with some very small variances.)
- Such **finite-dimensional version** is required for implementation.
- Standard HMC:
  - not applicable in Hilbert space setting. (No standard density of target  $\pi$ .)
  - applicable in finite-dimensional setting ...
  - ... with performance that rapidly deteriorates as  $N \uparrow \infty$ .

- Situation similar to that for explicit time integrators for PDE  $du/dt = Au$ , with  $A$  an unbounded operator in  $\mathcal{H}$ : they only make sense if equation has been first discretized in space and their performance degrades as the discretization is refined.
- Here we wish to construct an algorithm that may be expressed on the Hilbert space setting (and is therefore likely to perform uniformly well as  $N \uparrow \infty$  in finite-dimensional approximations).

### III. NEW ALGORITHM

## Hamiltonian flow in $\mathcal{H} \times \mathcal{H}$

- In finite-dimensional version, reference Gaussian measure  $\pi_0$  has a density  $\propto \exp(\frac{1}{2}\langle q, \mathcal{C}^{-1}q \rangle)$  wrt standard Lebesgue measure. (Note there is no Lebesgue measure on  $\mathcal{H}$ !)
- Hence target  $\pi$  has standard density  $\exp\left(\frac{1}{2}\langle q, \mathcal{C}^{-1}q \rangle + \Phi(q)\right)$ : a format we considered earlier with the notation  $L = \mathcal{C}^{-1}$ .
- Earlier discussion suggests to proceed as follows:
  - Introduce Gaussian measure  $\Pi_0$  on  $\mathcal{H} \times \mathcal{H}$  given by  $\Pi_0(dq, dv) = \pi_0(dq) \otimes \pi_0(dv)$ .
  - Introduce measure  $\Pi$  on  $\mathcal{H} \times \mathcal{H}$  given by  $d\Pi/d\Pi_0 \propto \exp(-\Phi(q))$ . Target  $\pi$  is  $q$ -marginal of  $\Pi$ .



— Consider system:

$$\frac{dq}{dt} = v, \quad \frac{dv}{dt} = -q + \mathcal{C} f(q), \quad f = -D\Phi.$$

• Under natural hypotheses, it may be shown that

— System defines a global flow  $\Xi^t$  on  $\mathcal{H} \times \mathcal{H}$ .

—  $\Xi^t$  preserves the measure  $\Pi$  on  $\mathcal{H} \times \mathcal{H}$ .

—  $(q^{(n+1)}, v^{(n+1)}) = \Xi^T(q^{(n)}, v^{(n)})$ ,  $v^{(n)} \sim \pi_0$  defines

via  $q^{(n)} \mapsto q^{(n+1)}$ , a Markov chain reversible wrt to  $\pi$ .

- System preserves *formally*

$$H(q, v) = \frac{1}{2} \langle v, C^{-1}v \rangle + \frac{1}{2} \langle q, C^{-1}q \rangle + \Phi(q).$$

(which in the finite-dimensional case is the old energy) and hence  $\exp(-H)$ .

- However  $\langle q, C^{-1}q \rangle$  and  $\langle v, C^{-1}v \rangle$  are almost surely infinite in an infinite-dimensional context. (If  $C$  is inverse Laplacian in  $L^2$ , they are squares of  $H^1$ -norms.)

## A numerical integrator

- Use Strang splitting  $\psi_h = \Xi_1^{h/2} \circ \Xi_2^h \circ \Xi_1^{h/2}$ , where

—  $\Xi_1$  is the flow of

$$\frac{dq}{dt} = 0, \quad \frac{dv}{dt} = C f(q).$$

—  $\Xi_2$  is the flow of

$$\frac{dq}{dt} = v, \quad \frac{dv}{dt} = -q.$$

- $\Xi_1, \Xi_2$  available in closed form.

## Accept/Reject rule

- The natural candidate for the acceptance probability is

$$a = \min\left(1, \exp\left(H(q^{(n)}, v^{(n)}) - H(q^*, v^*)\right)\right),$$

where  $H$  is the invariant we discussed above . . .

- . . . but, as we noted,  $H$  is almost surely infinite in  $\mathcal{H}$ .
- Remedy is to work a formula for the increment  $H(q^{(n)}, v^{(n)}) - H(q^*, v^*)$  that does not include the offending almost surely infinite terms.

- The recipe is:

$$\begin{aligned} \Phi(q_I) - \Phi(q_0) + \frac{h^2}{8} \left( |\mathcal{C}^{\frac{1}{2}} f(q_0)|^2 - |\mathcal{C}^{\frac{1}{2}} f(q_I)|^2 \right) \\ + h \sum_{i=1}^{I-1} \langle f(q_i), v_i \rangle + \frac{h}{2} \left( \langle f(q_0), v_0 \rangle + \langle f(q_I), v_I \rangle \right). \end{aligned}$$

This makes sense in  $\mathcal{H}$  and in the finite-dimensional setting coincides with the energy increment.

- This is discrete analogue of physically meaningful expression:

$$\Phi(q(T)) - \Phi(q(0)) + \int_0^T \langle f(q(t)), v(t) \rangle dt.$$

## MAIN RESULT

THEOREM: The algorithm defines a Markov chain which is reversible wrt to  $\pi$ .

The proof uses finite-dimensional approximations based on the eigenspaces of  $\mathcal{C}$ .