

## Hamiltonian Monte Carlo (HMC)

Just MH, we only need to be able to evaluate a function

$$\log P(x) = - \overset{\substack{\uparrow \\ \text{potential energy}}}{E(x)} + \text{const} \quad x \in \mathbb{R}^d$$

we will also need to evaluate

$$\nabla E(x) \in \mathbb{R}^d$$

(modern languages, we can often get derivative info at the same time as evaluating function for negligible extra cost.)

~~For~~  $x$  called positions, we introduce momentum variable  $p \in \mathbb{R}^d$

$$(x, p) \in \mathbb{R}^{2d}$$

for the notation  $p$ , introduce a new distribution

$$Q(p) = N(0, \bar{M})$$

this should be: - easy to sample from

- have symmetry  $Q(p) = Q(-p)$

$$\log Q(p) = -K(p) + \text{const}$$

↑

kinetic energy

$$(x, p) \in \mathbb{R}^{2d}$$

new distribution  $\log R(x, p) = -f(x, p) + \text{const}$

where  $\mathcal{H}(x, p) = F(x) + K(p)$

"Hamiltonian"

target dist  $P(x) = \int dp R(x, p)$

is recovered by  
marginalising  $R$  w.r.t.  $p$ .

similarly  $Q(p) = \int dx R(x, p)$

also target if  $x, p \sim R$   
 then  $x | p \sim P$   
 $x | p \sim Q$

we will try to sample  $x, p \sim R$  if we throw out  $p$  then  
 we are left with  $x \sim P$ , as required.

## Equations of Hamiltonian dynamics

$$\frac{dx^k}{dt} = \frac{\partial H}{\partial p^k} \quad \frac{dp^k}{dt} = - \frac{\partial H}{\partial x^k}$$

$t$  is a new ~~parameter~~ made up time coordinate.

choose a fixed interval  $S$

~~$\mathcal{D} : (x, p)$~~

$x(t), p(t)$

evolve E.O.M. forwards in time  $S$ , we get (deterministically)

$x(t+S), p(t+S)$

• Reversible

$$x, p \xrightarrow{S} x', p'$$

$$x', p' \xrightarrow{S} x, p$$

• Volume - preserving

(Liouville's theorem)

Take points in same  $\Delta x, \Delta p$  of phase space, evolve forwards in time  $S$ ,

$\Delta x', \Delta p'$ .

$$\Delta x \Delta p = \Delta x' \Delta p'$$

• Casimir's  $H(x, p) = E(x) + K(p)$

used Hamilton's EOM.

$$\frac{d}{dt} H = \frac{\partial H}{\partial t} + \frac{\partial H}{\partial x} \frac{dx}{dt} + \frac{\partial H}{\partial p} \frac{dp}{dt} = \frac{\partial H}{\partial x} \frac{dx}{dt} + \frac{\partial H}{\partial p} \frac{dp}{dt} = 0$$

It will be necessary to numerically integrate EOM.

$x(t), p(t)$  inputs

$x(t+\Delta t), p(t+\Delta t)$  outputs

Leap frog tries a step forward in  $t$ , approximates Hamiltonian dynamics.

Recall  $\log Q(p) = -K(p) + c$

$$K(p) = \frac{1}{2} \underline{p}^T \underline{M} \cdot \underline{p}$$

$M$  the mass matrix.

(symmetric positive definite)

$$\frac{\partial K}{\partial p_k} = \underline{M} \cdot \underline{p}$$

Leap frog is Reversible + V.I. pres. exactly for any  $\Delta t$ .

but  $f$  is only approximately conserved

$$f(x(t+\Delta t), p(t+\Delta t)) \\ = f(x(t), p(t)) + O(\Delta t^2).$$

2 ways to think about HMC.

1: as a version of MH on space  $(x, p)$

with a proposal  $Q_{MH}(\cdot | y | x) = Q(\hat{p}) \mathcal{J}^{(2d)}(\hat{z} - \text{Hamilton}(x, \hat{p}))$

Let  $\hat{z} = (x, p)$

2: as a version of Gated Gibbs sampler. (sweep)

we 2 blocks  ~~$(x, p)$~~   $(p, x)$

$p_{i+1} \sim R(p | x_i)$  sweep step 1

$p_{i+1} \sim Q(p)$

$x_{i+1} \sim R(x | p_{i+1})$  sweep step 2

$x_{i+1} \sim P(x)$

accept/reject  
according

MH rule

accept with  
 $\min \alpha \approx 1$ .