

Hamiltonian MCMC

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Review: Metropolis-Hastings

- ▶ **Given:** probability density $P(\mathbf{x})$, $\mathbf{x} \in \mathbb{R}^d$
- ▶ **Goal:** generate sample $\mathbf{x} \sim P$



Metropolis-Hastings

- ▶ Initialize $\mathbf{x}^{(0)}$ arbitrarily
- ▶ Given $\mathbf{x}^{(t)} = \mathbf{x}$, propose

$$\mathbf{x}' \sim Q(\cdot | \mathbf{x})$$

- ▶ Accept and set $\mathbf{x}^{(t+1)} = \mathbf{x}'$ with probability $\min(a, 1)$

$$a = \frac{P(\mathbf{x}') \cdot Q(\mathbf{x} | \mathbf{x}')}{P(\mathbf{x}) \cdot Q(\mathbf{x}' | \mathbf{x})}$$

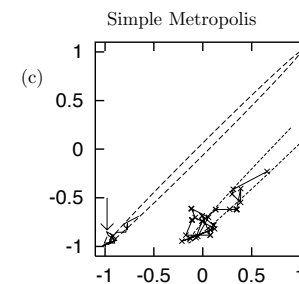
- ▶ Else reject and set $\mathbf{x}^{(t+1)} = \mathbf{x}^{(t)}$

For large enough T , have $\mathbf{x}^{(T)} \sim P$



Problem: random walk

Slow mixing due to “random walk” behavior



Why? Typical proposal is a random *displacement*

- ▶ Spherical Gaussian \rightarrow Brownian motion-like
- ▶ Ignores density surface



Main Idea

- ▶ Idea: use density to guide proposals
- ▶ Select random **velocity** $\mathbf{p}/m \in \mathbb{R}^d$
 - ▶ \mathbf{p} = momentum, m = mass
- ▶ Simulate motion on energy surface

$$\{(\mathbf{x}, E(\mathbf{x})) : \mathbf{x} \in \mathbb{R}^d\} \subseteq \mathbb{R}^{d+1}, \quad E(\mathbf{x}) = -\log P(\mathbf{x})$$

with initial velocity \mathbf{p}/m for some amount of time to get proposal \mathbf{x}' .



Main Idea

Demo: $P(x)$, $E(x)$, motion in 1D



Hamiltonian Mechanics

- ▶ Position $\mathbf{x} \in \mathbb{R}^d$
- ▶ Velocity $\mathbf{p}/m \in \mathbb{R}^d$
- ▶ Potential energy $E(\mathbf{x})$ (= height)

- ▶ Temporal dynamics

$$\begin{aligned} \frac{d\mathbf{x}}{dt} &= \frac{\mathbf{p}}{m} \\ \frac{d\mathbf{p}}{dt} &= -\frac{\partial E(\mathbf{x})}{\partial \mathbf{x}} \end{aligned}$$

Puck of mass m sliding on frictionless surface with velocity \mathbf{p}/m , height at \mathbf{x} equal to $E(\mathbf{x})$ (and thus “incline” $\partial E(\mathbf{x})/\partial \mathbf{x}$).



Generalization: Kinetic Energy

Define $K(\mathbf{p}) = \frac{\mathbf{p}^T \mathbf{p}}{2m}$
kinetic energy

$$\begin{aligned} \frac{d\mathbf{x}}{dt} &= \frac{\mathbf{p}}{m} \\ \frac{d\mathbf{p}}{dt} &= -\frac{\partial E(\mathbf{x})}{\partial \mathbf{x}} \end{aligned} \quad \Longrightarrow \quad \begin{aligned} \frac{d\mathbf{x}}{dt} &= \frac{\partial K(\mathbf{p})}{\partial \mathbf{p}} \\ \frac{d\mathbf{p}}{dt} &= -\frac{\partial E(\mathbf{x})}{\partial \mathbf{x}} \end{aligned}$$



Generalization: The Hamiltonian

Define $H(\mathbf{x}, \mathbf{p}) = E(\mathbf{x}) + K(\mathbf{p})$

Hamiltonian or total energy

$$\begin{aligned} \frac{d\mathbf{x}}{dt} &= \frac{\partial K(\mathbf{p})}{\partial \mathbf{p}} \\ \frac{d\mathbf{p}}{dt} &= -\frac{\partial E(\mathbf{x})}{\partial \mathbf{x}} \end{aligned} \Rightarrow \begin{aligned} \frac{d\mathbf{x}}{dt} &= \frac{\partial H(\mathbf{x}, \mathbf{p})}{\partial \mathbf{p}} \\ \frac{d\mathbf{p}}{dt} &= -\frac{\partial H(\mathbf{x}, \mathbf{p})}{\partial \mathbf{x}} \end{aligned}$$

Navigation icons

Simulating Hamiltonian Mechanics

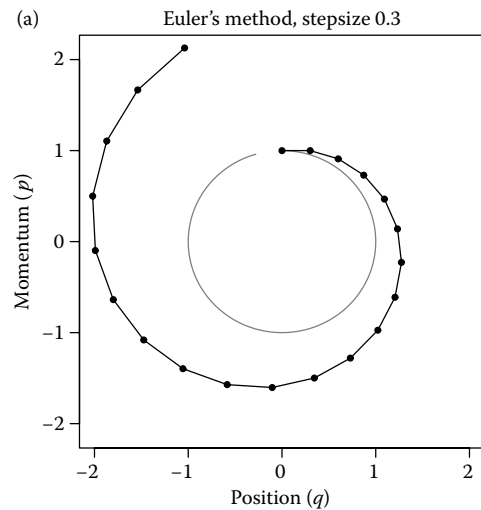
Euler's method

$$\begin{aligned} \mathbf{x}(t + \varepsilon) &= \mathbf{x}(t) + \varepsilon \frac{\mathbf{p}(t)}{m} \\ \mathbf{p}(t + \varepsilon) &= \mathbf{p}(t) - \varepsilon \frac{\partial E(\mathbf{x}(t))}{\partial \mathbf{x}} \end{aligned}$$

Problem: numerically unstable

Navigation icons

Euler's Method



Navigation icons

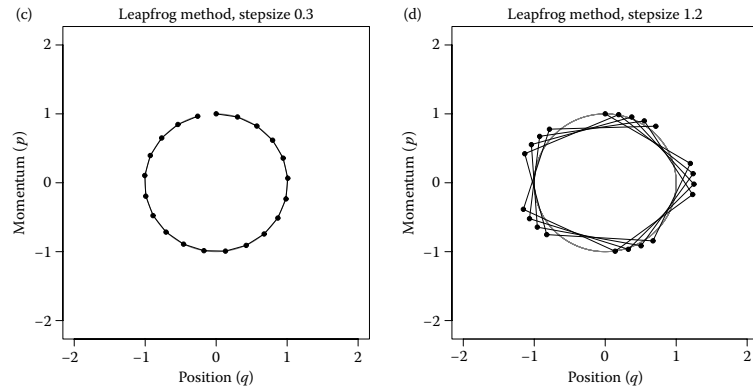
Leapfrog Method

More accurate and stable method

$$\begin{aligned} \mathbf{p}(t + \varepsilon/2) &= \mathbf{p}(t) - (\varepsilon/2) \frac{\partial E(\mathbf{x}(t))}{\partial \mathbf{x}} \\ \mathbf{x}(t + \varepsilon) &= \mathbf{x}(t) + \varepsilon \frac{\mathbf{p}(t + \varepsilon/2)}{m} \\ \mathbf{p}(t + \varepsilon) &= \mathbf{p}(t + \varepsilon/2) - (\varepsilon/2) \frac{\partial E(\mathbf{x}(t + \varepsilon))}{\partial \mathbf{x}} \end{aligned}$$

Navigation icons

Leapfrog Method



Hamiltonian MCMC

Random velocity/momentum instead of random displacement

- ▶ Start at \mathbf{x}
- ▶ Choose random momentum $\mathbf{p} \sim \exp(-\mathbf{p}^T \mathbf{p}/2m)$
- ▶ Simulate Hamiltonian mechanics for s time units
→ end at \mathbf{x}'
- ▶ Propose \mathbf{x}'

Problem: how to compute $Q(\mathbf{x}' | \mathbf{x})$ for acceptance probability?

Auxilliary Variables

Sample both \mathbf{x} and \mathbf{p} from

$$\begin{aligned} P(\mathbf{x}, \mathbf{p}) &= \exp(-H(\mathbf{x}, \mathbf{p})) \\ &= \exp(-E(\mathbf{x})) \exp(-K(\mathbf{p})), \end{aligned}$$

when done, discard \mathbf{p} values

Note: \mathbf{x} and \mathbf{p} are independent

Hamiltonian MCMC

Gibbs step

- ▶ Start at $(\mathbf{x}, \mathbf{p}^-)$
- ▶ Choose random momentum $\mathbf{p} \sim \exp(-\mathbf{p}^T \mathbf{p}/2m)$
- ▶ End at (\mathbf{x}, \mathbf{p})

Metropolis-Hastings step

- ▶ Start at (\mathbf{x}, \mathbf{p})
- ▶ Simulate Hamiltonian mechanics → end at $(\mathbf{x}', \mathbf{p}')$
- ▶ Propose $(\mathbf{x}', -\mathbf{p}')$

Acceptance Probability?

$$\begin{aligned} a &= \frac{P(\mathbf{x}', -\mathbf{p}')}{P(\mathbf{x}, \mathbf{p})} \cdot \frac{Q(\mathbf{x}, \mathbf{p} \mid \mathbf{x}', -\mathbf{p}')}{Q(\mathbf{x}', -\mathbf{p}' \mid \mathbf{x}, \mathbf{p})} \\ &= \frac{P(\mathbf{x}', -\mathbf{p}')}{P(\mathbf{x}, \mathbf{p})} \quad (\text{reversibility, volume preservation}) \\ &= \exp(E(\mathbf{x}) - E(\mathbf{x}') + K(\mathbf{p}) - K(\mathbf{p}')) \quad (K(\mathbf{p}') = K(-\mathbf{p}')) \\ &\approx 1 \quad (\text{conservation of energy}) \end{aligned}$$

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Reversibility

Let $T_{L,\varepsilon} : \mathbb{R}^{2d} \rightarrow \mathbb{R}^{2d}$ be simulation mapping with L steps at time increment ε

$$T_{L,\varepsilon}(\mathbf{x}, \mathbf{p}) = (\mathbf{x}', \mathbf{p}') \implies T_{L,\varepsilon}(\mathbf{x}', -\mathbf{p}') = (\mathbf{x}, -\mathbf{p})$$

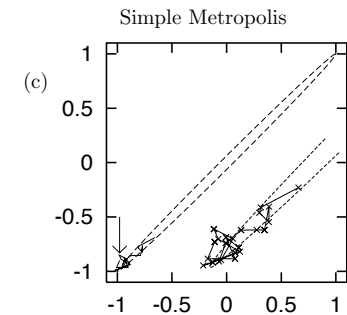
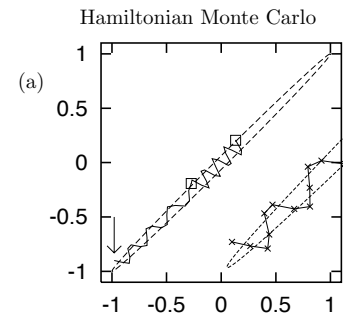
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Demo

Demo with sampling

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Example



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Example

Setup: 100D Gaussian, standard deviations in different dimensions are $0.01, 0.02, \dots, 1.00$

