

Suppose that π is our target of interest on \mathbb{R}^p

Define the map $T: \mathbb{R}^p \rightarrow \mathbb{R}^p$ s.t. $T^{-1} = T$

Ex: $T(x_1, \dots, x_p) = (x_2, x_1, x_3, \dots, x_p)$

Algo: Given $X_n = x$ do

(1) Compute $y = T(x)$

(2) Set

$$X_{n+1} = \begin{cases} y & \text{w.p. } \alpha(x, y) \\ x & \text{w.p. } 1 - \alpha(x, y) \end{cases}$$

$$\alpha(x, y) = \min \left[1, \frac{\pi(y)}{\pi(x) |\det(\nabla T(y))|} \right]$$

Proposition: If $X_0 \sim \pi$ then $X_1 \sim \pi$.

Pf: We look to show that the chain is π reversible

$$T_1 \equiv \int_A \pi(x) P(x, B) dx = \int_B \pi(x) P(x, A) dx \equiv T_2$$

for all $A, B \in \mathcal{F}$. when $p(x, A) = P(X_1 \in A | X_0 = x)$

[Why is this sufficient?

w.t.s. $P(X_1 \in A) = \int_A \pi(x) dx$ if $X_0 \sim \pi$

But also notice that

$$P(X_1 \in A) = \int P(X_1 \in A | X_0 = x) \pi(dx)$$

So if we can show

$$\int_A \pi(x) dx = \int \pi(x) P(X_1 \in A | X_0 = x) dx$$

then choose A to finish.



Now notice

$$P(X, A) = \alpha(x, T(x)) \mathbb{1}_A(T(x)) + (1 - \alpha(x, y)) \mathbb{1}_A(x)$$

$$T_1 = \int_A \pi(x) \alpha(x, T(x)) \mathbb{1}_B(T(x)) dx + \int \pi(x) (1 - \alpha(x, y)) \mathbb{1}_B(x) dx$$

$$T_2 = \int_B \pi(x) \alpha(x, T(x)) \mathbb{1}_A(T(x)) dx + \int \pi(x) (1 - \alpha(x, y)) \mathbb{1}_A(x) dx$$

With a multivariate change of variables we see

$$T(x) = y \implies x = T^{-1}(y) = T(y) \implies dx = |\det(\nabla T(y))| dy$$

which yields

$$\int_A \pi(x) \alpha(x, T(x)) \mathbb{1}_B(T(x)) dx$$

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$$= \int_{\mathbb{R}^p} \pi(T(y)) \alpha(T(y), y) | \det \nabla T(y) | 1_A(T(y)) 1_B(y) dy$$

therefore $T_1 = T_2$ iff

$$\pi(x) \alpha(x, T(x)) = \pi(T(x)) \alpha(T(x), y) | \det \nabla T(x) |, \quad \forall x \in \mathbb{R}^p.$$

But as

$$\begin{aligned} \pi(x) \alpha(x, T(x)) &= \min \left[\pi(x), \frac{\pi(T(x))}{| \det (\nabla T(T(x))) |} \right] \\ &= \min \left[\pi(x), \pi(T(x)) | \det (\nabla T(x)) | \right] \end{aligned}$$

and $\pi(T(x)) | \det (\nabla T(x)) | \alpha(T(x), x)$

$$= \pi(T(x)) | \det (\nabla T(x)) | \min \left(1, \frac{\pi(x)}{\pi(T(x)) | \det (\nabla T(x)) |} \right)$$

Therefore — is satisfied by our choice of $\alpha(x, y)$.

Parallel Tempering

Suppose we can write

$$\pi(x_1, \dots, x_K) = \prod_{i=1}^K \pi_i(x_i), \quad \pi_i(x) \propto e^{-U(x)/t_i}$$

for $1 = t_1 < t_2 < \dots < t_K$

Suppose $T(x_1, \dots, x_K) = (x_1, x_2, \dots, x_K)$

$$\Rightarrow \nabla T(x_1, \dots, x_K) = \begin{pmatrix} \square & & \\ & \square & \\ & & \square & \\ & & & \square & \\ & & & & \ddots \end{pmatrix}$$

$$d_t + \nabla T(x_1, \dots, x_K) = \pm 1$$

$$\begin{aligned} \alpha(x, T(x)) &= \min \left\{ 1, \frac{\pi(T(x))}{\pi(x)} \cdot \frac{1}{|d_t + \nabla T(x)|} \right\} \\ &= \min \left\{ 1, \frac{\bar{\pi}(x_2, x_1, \dots, x_K)}{\bar{\pi}(x_1, x_2, \dots, x_K)} \right\} \\ &= \min \left\{ 1, \frac{\pi_1(x_1) \pi_1(x_2)}{\pi_1(x_2) \pi_2(x_2)} \right\} \end{aligned}$$

Hamiltonian Monte Carlo

$$\pi(x) \propto \exp(-U(x)) \quad U: \mathbb{R}^d \mapsto \mathbb{R} \quad U \in C^2$$

Introduce a new variable $p \in \mathbb{R}$ momentum

which gives

$$\bar{\pi}(p, x) \propto \exp \left\{ -\frac{1}{2} p^T C^{-1} p - U(x) \right\}$$

where C is a p.s.d. matrix

$$\pi(x) = \int \pi(p, x) dp \quad \text{and we try to instead}$$

sample from $\overline{\pi}$.

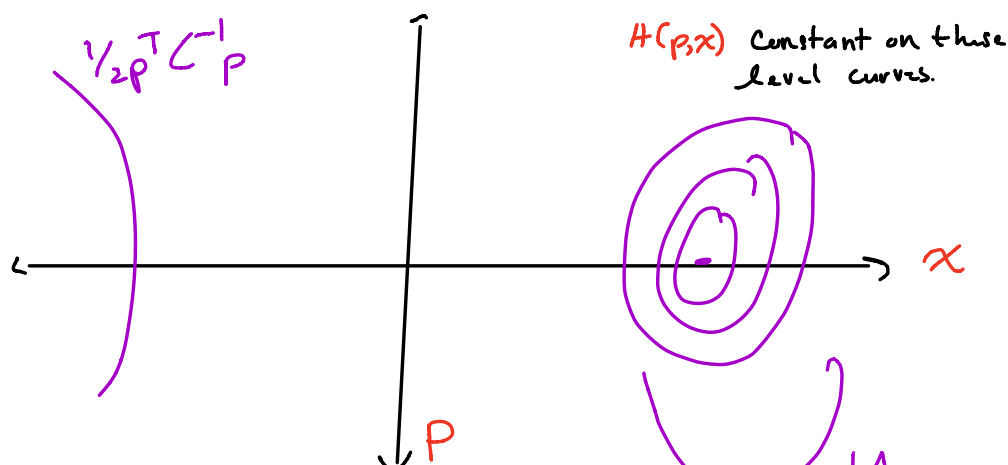
Define the hamiltonian $H(p, x) = \frac{1}{2} p^T C^{-1} p + U(x)$

Want to find the curve in which $H(p, x)$ is constant.

Consider the O.D.E. with initial conditions (p_0, x_0)

$$\begin{cases} \frac{dp_t}{dt} = -\frac{\partial H}{\partial x}(p_t, x_t) = -U'(x) \\ \frac{dx_t}{dt} = \frac{\partial H}{\partial p}(p_t, x_t) = C^{-1} p_t \end{cases}$$

Note: $\frac{d}{dt} H(p_t, x_t) = \frac{\partial}{\partial p} H(p_t, x_t) \frac{dp_t}{dt} + \frac{\partial}{\partial x} H(p_t, x_t) \frac{dx_t}{dt} = 0$



Set $y = \begin{pmatrix} p \\ x \end{pmatrix}$ then the Hamiltonian flow can be written as

$$\frac{dy_t}{dt} = \underbrace{\begin{bmatrix} 0 & -I \\ I & 0 \end{bmatrix}}_{J^T} \underbrace{\begin{pmatrix} \partial_p H(p_t, x_t) \\ \partial_x H(p_t, x_t) \end{pmatrix}}_{\nabla H(y_t)}$$

$$\frac{dy_t}{dt} = J^T \nabla H(y_t)$$

Set $H_t: (p_0, x_0) \mapsto (p_t, x_t)$

H_t satisfies

$$\nabla H_t(y) J \nabla H_t^T(y) = J \quad \forall y$$

$$\Rightarrow \det \nabla H_t(y) = \pm 1$$

With this we can set up a similar alg. as before

Alg 0: Given $x_k = x$:

1. Draw $\bar{p} \sim N(0, C)$

2. Propose $(\bar{p}, y) = H_t(\bar{p}, x)$

$$\text{Set } x_{n+1} = \begin{cases} y & \text{w.p. } \alpha \\ x & \text{w.p. } 1-\alpha \end{cases}$$

$$\alpha = \min \left\{ 1, e^{-H(\bar{p}, y) + H(\bar{p}, x)} \right\}$$