

Convergence of the kinetic annealing for general potentials

Joint work with Pierre Monmarché

Lucas Journal

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$$U : \mathbb{R}^d \longrightarrow \mathbb{R}_+.$$

Goal :

$$\min_{\mathbb{R}^d} U.$$

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Design stochastic process (X_t) whose law is close to :

$$\pi_{\beta_t}(\mathrm{d}x) \propto e^{-\beta_t U(x)} \mathrm{d}x$$

where $\beta_t \xrightarrow[t \rightarrow \infty]{} \infty$.

$$\pi_{\beta_t}(U(x) > \delta) \xrightarrow[t \rightarrow \infty]{} 0.$$

Examples of process used in simulated annealing

- Overdamped Langevin Process (OLP) :

$$dX_t = -\nabla U(X_t)dt + \sqrt{2\beta_t^{-1}}dB_t.$$

- Kinetic Langevin Process (KLP) :

$$\begin{cases} dX_t = Y_t dt \\ dY_t = -\nabla U(X_t)dt - \gamma_t Y_t dt + \sqrt{2\gamma_t\beta_t^{-1}}dB_t. \end{cases}$$

- Local equilibrium of KLP :

$$\mu_{\beta_t}(dx dy) \propto e^{-\beta_t H(x,y)} dx dy$$

where $H(x, y) = U(x) + |y|^2/2$.

Cooling schedule and energy barrier

Cooling schedule :

$$\beta_t = \frac{\ln(e^{c\beta_0} + t)}{c}.$$

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Largest energy barrier :

$$c^* = \sup_{x_1, x_2} E(x_1, x_2),$$

where

$$E(x_1, x_2) = \inf_{\xi} \left\{ \max_{0 \leq t \leq 1} U(\xi(t)) - U(x_1) - U(x_2) \right\}.$$

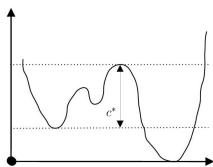


Figure – Example of potential

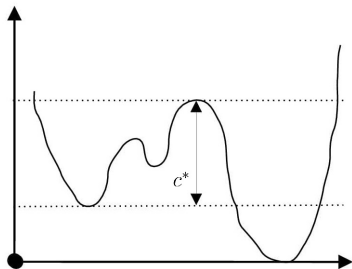
Theorem (Holley, Kusuoka, Stroock 89)

M smooth compact manifold and $U : M \rightarrow \mathbb{R}_+$.

(X_t) OLP with cooling schedule (β_t) .

Then if $c > c^*$, $U(X_t) \rightarrow \min U$ in probability.

If there exists $p \in M$, bottom of a well of height greater than c , then $\mathbb{P}(\inf U(X_t) > U(p)) > 0$.



- $U : \mathbb{R}^d \rightarrow \mathbb{R}_+, U(\infty) = \infty, |\nabla U|(\infty) = \infty,$
- $\inf_{\mathbb{R}^d} |\nabla U|^2 - \Delta U > -\infty.$

(X_t) Overdamped Langevin Process.

Theorem (Chiang, Hwang, Sheu 87)

If $c > 3/2c^$, $U(X_t) \rightarrow \min U$ in probability.*

Theorem (Royer 89, Miclo 92)

If $c > c^$, $U(X_t) \rightarrow \min U$ in probability.*

(X_t) Overdamped Langevin Process.

Theorem (Zitt 08)

$U : \mathbb{R}^d \rightarrow \mathbb{R}_+$, $U(x) \geq \ln(|x|)^m - C$, $\|\nabla U\|_\infty < \infty$,
 $\Delta U \leq 0$ outside a compact.

Then if $c > c^*$, $U(X_t) \rightarrow \min U$ in probability.

Theorem (Fournier, Tardif 21)

$U : \mathbb{R}^d \rightarrow \mathbb{R}_+$, $U(\infty) = \infty$, $\int_{\mathbb{R}^d} e^{-\alpha_0 U} < \infty$

Then if $c > c^*$, $U(X_t) \rightarrow \min U$ in probability.

(X_t, Y_t) Kinetic Langevin Process. $H(x, y) = U(x) + |y|^2/2$.

Theorem (Monmarché 18)

$U : \mathbb{R}^d \rightarrow \mathbb{R}_+$, $x \cdot \nabla U(x) \geq r|x|^2 - M$, $\|\nabla^2 U\|_\infty < \infty$.

Then if $c > c^$, $U(X_t) \rightarrow \min U$ in probability.*

Theorem (J, Monmarché 21)

$U : \mathbb{R}^d \rightarrow \mathbb{R}_+$, $U(\infty) = \infty$, $\int_{\mathbb{R}^d} e^{-\alpha_0 U} < \infty$.

Then if $c > c^$, $H(X_t, Y_t) \rightarrow \min U$ in probability.*

If there exists p , bottom of a well of height greater than c^ , then*

$\mathbb{P}(\inf U(X_t) > U(p)) > 0$.

A basic example

$$dX_t = -\nabla U(X_t)dt + \sqrt{2\beta^{-1}}dB_t,$$

U quadratic at infinity.

- π_β stationary measure.
- $f_t = \mathcal{L}aw(X_t)$, $h_t = \frac{df_t}{d\pi_\beta}$.

$$\partial_t h_t = \beta^{-1} \Delta h_t - \nabla U \cdot \nabla h_t =: L^* h_t.$$

Definition (Carré du champ)

$$\Gamma f = \frac{1}{2} (L^* f^2 - 2fL^* f).$$

$$\Gamma f = \sqrt{2\beta^{-1}} |\nabla f|^2.$$

$$H_t = \int_{\mathbb{R}^d} (h_t - 1)^2 d\pi_\beta \geq \|f_t - \pi_\beta\|_{TV}^2.$$

$$H'_t = - \int_{\mathbb{R}^d} \Gamma(h_t) d\pi_\beta.$$

A more basic example

Poincaré inequality

For all $h : \mathbb{R}^d \rightarrow \mathbb{R}$, $\int_{\mathbb{R}^d} h d\pi_\beta = 1$:

$$\int_{\mathbb{R}^d} (h - 1)^2 d\pi_\beta \leq \lambda_\beta \int_{\mathbb{R}^d} \Gamma(h) d\pi_\beta.$$

λ_β satisfies :

$$\lim_{\beta \rightarrow \infty} \beta^{-1} \ln(\lambda_\beta) = c^*.$$

$$H_t = \int_{\mathbb{R}^d} (h_t - 1)^2 d\pi_\beta.$$

$$H'_t = - \int_{\mathbb{R}^d} \Gamma(h_t) d\pi_\beta \leq -\lambda_\beta^{-1} H_t.$$

$$H_t \leq e^{-\lambda_\beta^{-1} t} H_0.$$

- $\beta = \beta_t$.

$$H'_t = - \int_{\mathbb{R}^d} \Gamma(h_t) d\pi_{\beta_t} + \beta'_t A_t.$$

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- U is such that π_β does not satisfy Poincaré inequality.

- $\beta = \beta_t$.

$$H'_t = - \int_{\mathbb{R}^d} \Gamma(h_t) d\pi_{\beta_t} + \beta'_t A_t.$$

- U is such that π_β does not satisfy Poincaré inequality.
- For kinetic Langevin process :

$$\Gamma(f) = |\nabla_y f|^2,$$

hence no Poincaré inequality of the form

$$\int_{\mathbb{R}^d} (h - 1)^2 d\mu_\beta \leq \lambda_\beta \int_{\mathbb{R}^d} \Gamma(h) d\mu_\beta.$$

Generator of the process :

$$L_t = y \cdot \nabla_x - \nabla U \cdot \nabla_y - \gamma_t y \cdot \nabla_y + \gamma_t \beta_t^{-1} \Delta_y$$

Let f_t the law of the process and $h_t = \frac{f_t}{\mu_{\beta_t}}$.

L_t^* the dual of L_t in $L^2(\mu_{\beta_t})$:

$$\partial_t h_t = L_t^* h_t - \beta_t' H h_t.$$

Plan of the proof

1 Almost surely, $\sup_t H(X_t, Y_t) < \infty$.

a Almost surely, $\liminf_{t \rightarrow \infty} H(X_t, Y_t)$.

b For compact set C , there exists $K > 0$ such that

$$\inf_{x \in C} \mathbb{P}(\sup_t H(X_t, Y_t) \leq K) \geq 1/4.$$

2 If X lives in a compact set, there is convergence.

Position in a compact set

Fix $K > 1$, $L_K > 1$, $M_K = (\mathbb{R}/2L_K\mathbb{Z})^d$, such that $\{U \leq K\} \subset M_K$.
 $U^K : M_K \rightarrow \mathbb{R}$, $U^K = U$ on $\{U \leq K\}$

$$\begin{cases} dX_t^K = Y_t^K dt \\ dY_t^K = -\nabla_x U^K(X_t) dt - \gamma_t Y_t^K dt + \sqrt{2\gamma_t \beta_t^{-1}} dB_t. \end{cases} \quad (1)$$

$$\left\{ \sup_{t \geq 0} H(X_t, Y_t) \leq K \right\} = \left\{ \sup_{t \geq 0} H_K(X_t^K, Y_t^K) \leq K \right\},$$

where $H_K(x, y) = U^K(x) + |y|^2/2$.

$$\mu_\beta^K(dx dy) \propto e^{-\beta H_K(x, y)} dx dy.$$

Hypoocoercivity *à la Villani* :

$$\phi_t(h) = |(\nabla_x + \nabla_y)h|^2 + \sigma_t h^2$$

with $\sigma_t = \frac{1}{2} + 2\sqrt{\gamma_t^{-1}\beta_t(1 + \|\nabla U^K\|_\infty + \gamma_t)^2}$, we introduce

$$\tilde{N}(t) = \int_{\mathbb{R}^d \times \mathbb{R}^d} \phi_t \left(h_t^K - 1 \right) d\mu_{\beta_t}^K$$

$$\tilde{I}(t) = \int_{M_K \times \mathbb{R}^d} \left| \nabla h_t^K \right|^2 d\mu_{\beta_t}^K.$$

Differentiating \tilde{N} , one can (formally) check that

$$\tilde{N}'(t) \leq -\frac{1}{2}\tilde{I}(t) + C\beta_t'(1 + \beta_t)\tilde{N}(t)$$

Poincaré inequality :

$$\tilde{N}(t) \leq \lambda_{\beta_t} \tilde{I}(t)$$

With :

$$\frac{1}{\beta} \ln(\lambda_{\beta}) \xrightarrow{\beta \rightarrow \infty} c^*$$

Conclusion :

$$\tilde{N}'(t) \leq \left(-\frac{C'}{(1+t)^{c^*/c}} + \frac{C(1 + \ln(1+t))}{(1+t)} \right) \tilde{N}(t)$$

Proposition

If $c > c^*$, then for all $K > 1$, all C^∞ -probability density f_0 with compact support in $M_K \times \mathbb{R}^d$, and all $\delta > 0$,

$$\mathbb{P}_{f_0} \left(H_K(X_t^K, Y_t^K) > \delta \right) \xrightarrow[t \rightarrow +\infty]{} 0.$$

$$\mathbb{P} \left(H_K(X_t^K, Y_t^K) > \delta \right) \leq \|h_t^K\|_{L^2} (\mu_{\beta_t}(H_K > \delta))^{1/2}$$

Non-convergence for fast cooling schedule

Goal :

$$\mathbb{P} \left(\sup_{t \geq 0} H(X_t, Y_t) \leq H(x_0, y_0) + c + \delta \right) > 0$$

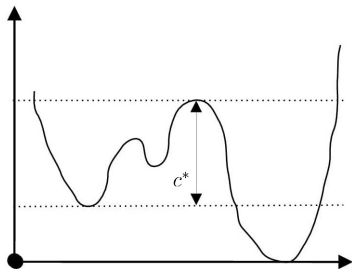


Figure – Example of c^*

THANK YOU FOR YOUR ATTENTION