

Convergence of Langevin-Simulated Annealing algorithms with multiplicative noise in L^1 -Wasserstein distance

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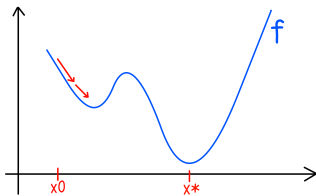
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The Langevin Equation in \mathbb{R}^d

$$dX_t = -\nabla V(X_t)dt + \sigma dW_t,$$

- $V : \mathbb{R}^d \rightarrow \mathbb{R}^+$: coercive function to be minimized
- $\sigma > 0$: noise parameter
- Invariant Gibbs measure : $\nu_\sigma(dx) \propto \exp(-2V(x)/\sigma^2) dx$
- Solve optimization problem with gradient descent type algorithm : $\min_{x \in \mathbb{R}^d} V(x)$.
- Exogenous noise σ added to escape local minima ('traps') and explore the state space (SGLD algorithms)
- For small σ , ν_σ is concentrated around $\operatorname{argmin}(V)$.



The Langevin-Simulated Annealing Equation

$$dX_t = -\nabla V(X_t)dt + a(t)\sigma dW_t,$$

- $V : \mathbb{R}^d \rightarrow \mathbb{R}^+$: coercive function to be minimized
 - $\sigma > 0$: noise parameter
 - $a : \mathbb{R}^+ \rightarrow \mathbb{R}^+$ non-increasing with $a(t) \xrightarrow{t \rightarrow \infty} 0$.
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- The 'instantaneous' invariant measure $\nu_{a(t)\sigma}(dx) \propto \exp(-2V(x)/(a^2(t)\sigma^2))$ converges itself to $\operatorname{argmin}(V)$
 - Schedule $a(t) = A \log^{-1/2}(t)$ then $X_t \xrightarrow{t \rightarrow \infty} \operatorname{argmin}(V)$ [Chiang-Hwang 1987], [Miclo 1992]
 - ([Gelfand-Mitter 1991] proves the convergence of the algorithm
- $$\bar{X}_{n+1} = \bar{X}_n - \gamma_{n+1}(\nabla V(\bar{X}_n) + \zeta_{n+1}) + a(\gamma_1 + \dots + \gamma_n)\sigma\sqrt{\gamma_{n+1}}\mathcal{N}(0, I_d),$$
- (γ_n) : decreasing step sequence,
 (ζ_n) : noise of SGD with $\mathbb{E}[\zeta_n] = 0$.

- Noise $\sigma > 0 \implies$ isotropic, homogeneous noise \implies not adapted to V
- Instead : $\sigma(X_t)$ depends on the position
- In ML literature, a good choice is $\sigma(x)\sigma(x)^\top \simeq (\nabla^2 V(x))^{-1}$.

$$dY_t = -(\sigma\sigma^\top \nabla V)(Y_t)dt + a(t)\sigma(Y_t)dW_t + \left(a^2(t) \left[\sum_{j=1}^d \partial_i(\sigma\sigma^\top)(Y_t)_{ij} \right]_{1 \leq i \leq d} \right) dt$$

$$a(t) = \frac{A}{\sqrt{\log(t)}},$$

- Correction term so that $\nu_{a(t)} \propto \exp(-2V(x)/a^2(t))$ is still the "instantaneous" invariant measure

- Prove the convergence in \mathcal{W}_1 of Y_t and \tilde{Y}_t to ν^* (supported by $\operatorname{argmin}(V)$)

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$$\mathcal{W}_1(Y_t, \nu^*) \leq \mathcal{W}_1(Y_t, \nu_{a(t)}) + \mathcal{W}_1(\nu_{a(t)}, \nu^*)$$

The convergence is limited by the slowness of $a(t)$ as

$\mathcal{W}_1(\nu_{a(t)}, \nu^*) \asymp a(t) \asymp \log^{-1/2}(t)$. In fact we also prove

$$\mathcal{W}_1(Y_t^{x_0}, \nu_{a(t)}) \leq C_\alpha \max(1 + |x_0|, V(X_0)) t^{-\alpha}$$

$$\mathcal{W}_1(\tilde{Y}_t^{x_0}, \nu_{a(t)}) \leq C_\alpha \max(1 + |x_0|, V^2(X_0)) t^{-\alpha}$$

for every $\alpha < 1$.

- **Assumptions :**

- 1 V is strongly convex outside some compact set
- 2 σ is bounded and elliptic : $\sigma \sigma^\top \geq \sigma_0 I_d$, $\sigma_0 > 0$.
- 3 ∇V is Lipschitz
- 4 Decreasing steps (γ_n) for the Euler scheme, with $\sum_n \gamma_n = \infty$, $\sum_n \gamma_n^2 < \infty$,
 $\Gamma_n := \gamma_1 + \dots + \gamma_n$.

- ([Pages-Panloup 2020] proves the convergence of the Euler scheme of a general SDE $dX_t = b(X_t)dt + \sigma(X_t)dW_t$ to the invariant measure for \mathcal{W}_1 (and d_{TV})
- Recall :

$$\mathcal{W}_1(\pi_1, \pi_2) = \sup \left\{ \int_{\mathbb{R}^d} f(x)(\pi_1 - \pi_2)(dx) : f : \mathbb{R}^d \rightarrow \mathbb{R}, [f]_{\text{Lip}} = 1 \right\}.$$

- *Domino strategy* : $(P, \bar{P} : \text{ kernels of } X, \bar{X})$

$$\begin{aligned} |\mathbb{E}f(\bar{X}_{\Gamma_n}^x) - \mathbb{E}f(X_{\Gamma_n}^x)| &= |\bar{P}_{\gamma_1} \circ \dots \circ \bar{P}_{\gamma_n} f(x) - P_{\Gamma_n} f(x)| \\ &= \left| \sum_{k=1}^n \bar{P}_{\gamma_1} \circ \dots \circ \bar{P}_{\gamma_{k-1}} \circ (\bar{P}_{\gamma_k} - P_{\gamma_k}) \circ P_{\Gamma_n - \Gamma_k} f(x) \right| \\ &\leq \sum_{k=1}^n |\bar{P}_{\gamma_1} \circ \dots \circ \bar{P}_{\gamma_{k-1}} \circ (\bar{P}_{\gamma_k} - P_{\gamma_k}) \circ P_{\Gamma_n - \Gamma_k} f(x)|, \end{aligned}$$

- 1 For large $k \implies$ Error in small time \implies use bounds for $\|X_t^x - \bar{X}_t^x\|_p$
- 2 For small $k \implies$ Ergodicity contraction properties using the convexity of V outside a compact set and the ellipticity of σ [Wang 2020] :

$$\begin{aligned} \mathcal{W}_1(X_t^x, X_t^y) &\leq Ce^{-\rho t} |x - y| \\ \implies \mathcal{W}_1(X_t^x, \nu) &\leq Ce^{-\rho t} (1 + |x|). \end{aligned}$$

- Problems before applying the domino strategy : non-homogeneous Markov chain + the ellipticity parameter fades away in $a(t)$.

⇒ What is the dependency of the constants C and ρ in the ellipticity?

Consider $dX_t = b(X_t)dt + a\sigma(X_t)dW_t$, $a > 0$ with invariant measure ν_a .

$$\mathcal{W}_1(X_t^x, X_t^y) \leq Ce^{C_1/a^2} |x - y| e^{-\rho_a t}, \quad \rho_a := e^{-C_2/a^2}$$

$$\mathcal{W}_1(X_t^x, \nu_a) \leq Ce^{C_1/a^2} e^{-\rho_a t} \mathbb{E}|\nu_a - x|.$$

We first consider the plateau SDE :

$$dX_t = -\sigma\sigma^\top \nabla V(X_t)dt + a_{n+1}\sigma(X_t)dW_t + a_{n+1}^2 \Upsilon(X_t)dt, \quad t \in [T_n, T_{n+1}),$$

$$a_n = A \log^{-1/2}(T_n)$$

We apply the contraction property on every plateau :

$$\mathcal{W}_1(X_{T_{n+1}}, \nu_{a_{n+1}} | X_{T_n}) \leq C e^{C_1/a_{n+1}^2} e^{-\rho_{a_{n+1}}(T_{n+1}-T_n)} \mathbb{E} [|\nu_{a_{n+1}} - X_{T_n}| | X_{T_n}]$$

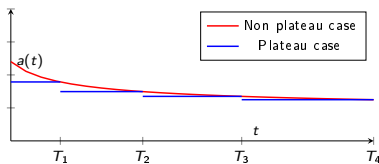
We integrate over the law of X_{T_n} , giving

$$\begin{aligned} \mathcal{W}_1([X_{T_{n+1}}^{x_0}], \nu_{a_{n+1}}) &\leq C e^{C_1/a_{n+1}^2} e^{-\rho_{a_{n+1}}(T_{n+1}-T_n)} \mathcal{W}_1([X_{T_n}^{x_0}], \nu_{a_{n+1}}) \\ &\leq C e^{C_1/a_{n+1}^2} e^{-\rho_{a_{n+1}}(T_{n+1}-T_n)} \left(\mathcal{W}_1([X_{T_n}^{x_0}], \nu_{a_n}) + \mathcal{W}_1(\nu_{a_n}, \nu_{a_{n+1}}) \right). \end{aligned}$$

And we iterate :

$$\begin{aligned} \mathcal{W}_1([X_{T_{n+1}}^{x_0}], \nu_{a_{n+1}}) &\leq \mu_{n+1} \mathcal{W}_1(\nu_{a_n}, \nu_{a_{n+1}}) + \mu_{n+1} \mu_n \mathcal{W}_1(\nu_{a_{n-1}}, \nu_{a_n}) + \dots \\ &\quad + \mu_{n+1} \dots \mu_1 \mathcal{W}_1(\nu_{a_0}, \nu_{a_1}) + \mu_{n+1} \dots \mu_1 \mathcal{W}_1(\delta_{x_0}, \nu_{a_0}), \\ \mu_n &:= C e^{C_1/a_n^2} e^{-\rho_{a_n}(T_n-T_{n-1})}. \end{aligned}$$

$$\mathcal{W}_1(\nu_{a_n}, \nu_{a_{n+1}}) \leq C(a_n - a_{n+1}).$$



$$\mu_n := C e^{C_1/a_n^2} e^{-\rho_{a_n}(T_n - T_{n-1})}, \quad \rho_{a_n} = e^{-C_2/a_n^2}.$$

We now choose

$$T_{n+1} - T_n = Cn^\beta, \beta > 0, \quad a_n = \frac{A}{\sqrt{\log(T_n)}}, \quad A > 0 \text{ large enough}$$

yielding

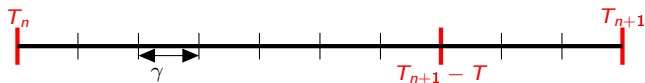
$$\mathcal{W}_1([X_{T_{n+1}}^{x_0}], \nu_{a_{n+1}}) \leq C(1 + |x_0|)\mu_n a_n,$$

where $\mu_n = O(\exp(-Cn^\eta))$. And

$$\mathcal{W}_1([X_{T_{n+1}}^{x_0}], \nu^*) \leq \mathcal{W}_1([X_{T_{n+1}}^{x_0}], \nu_{a_{n+1}}) + \mathcal{W}_1(\nu_{a_{n+1}}, \nu^*) \leq C a_n (1 + |x_0|).$$

Convergence of Y_t with continuously decreasing $(a(t))$

- We apply *domino strategy* to bound $\mathcal{W}_1(X_t, Y_t)$:



- for f Lipschitz-continuous and fixed $T > 0$:

$$\begin{aligned} & \left| \mathbb{E}f(X_{T_{n+1}-T_n}^{X,n}) - \mathbb{E}f(Y_{T_{n+1}-T_n, T_n}^X) \right| \\ & \leq \sum_{k=1}^{\lfloor (T_{n+1}-T_n-T)/\gamma \rfloor} \left| P_{(k-1)\gamma, T_n}^Y \circ (P_{\gamma, T_n+(k-1)\gamma}^Y - P_{\gamma}^{X,n}) \circ P_{T_{n+1}-T_n-k\gamma}^{X,n} f(x) \right| \\ & + \sum_{k=\lfloor (T_{n+1}-T_n-T)/\gamma \rfloor + 1}^{\lfloor (T_{n+1}-T_n)/\gamma \rfloor} \left| P_{(k-1)\gamma, T_n}^Y \circ (P_{\gamma, T_n+(k-1)\gamma}^Y - P_{\gamma}^{X,n}) \circ P_{T_{n+1}-T_n-k\gamma}^{X,n} f(x) \right| \end{aligned}$$

- for $k = 1, \dots, (T_{n+1} - T_n - T)/\gamma$, the kernel $P_{T_{n+1}-T_n-k\gamma}^{X,n}$ has an exponential contraction effect on time $> T$:

$$\begin{aligned} & |(P_{\gamma, T_n+(k-1)\gamma}^Y - P_{\gamma}^{X,n}) \circ P_{T_{n+1}-T_n-k\gamma}^{X,n} f(x)| \\ & = |\mathbb{E}P_{T_{n+1}-T_n-k\gamma}^{X,n} f(X_{\gamma}^{X,n}) - \mathbb{E}P_{T_{n+1}-T_n-k\gamma, n}^X f(Y_{\gamma, T_n+(k-1)\gamma}^X)| \\ & \leq Ce^{C_1 a_{n+1}^{-2}} e^{-\rho_{n+1}(T_{n+1}-T_n-k\gamma)} [f]_{\text{Lip}} \mathbb{E}|X_{\gamma}^{X,n} - Y_{\gamma, T_n+(k-1)\gamma}^X| \\ & \leq Ce^{C_1 a_{n+1}^{-2}} e^{-\rho_{n+1}(T_{n+1}-T_n-k\gamma)} [f]_{\text{Lip}} \sqrt{\gamma} (a_n - a_{n+1}) \end{aligned}$$

- Bounds for the error on time intervals no longer than T :

$$|(P_{\gamma, T_n+(k-1)\gamma}^Y - P_{\gamma}^{X,n}) \circ P_{T_{n+1}-T_n-k\gamma}^{X,n} f(x)| \leq C a_{n+1}^{-2} (a_n - a_{n+1}) [f]_{\text{Lip}} \frac{\gamma}{\sqrt{T_{n+1} - T_n - k\gamma}} V(x)$$

using Taylor formula up to order 4.

- We apply on each time interval $[T_n, T_{n+1})$ and obtain the recursive inequality

$$\mathcal{W}_1([X_{T_{n+1}-T_n}^{X,n}, [Y_{T_{n+1}-T_n, T_n}^X]) \leq C e^{C_1 a_{n+1}^{-2}} (a_n - a_{n+1}) \rho_{n+1}^{-1} V(x),$$

$$\begin{aligned} \mathcal{W}_1([X_{T_{n+1}}^{X_0}, [Y_{T_{n+1}}^{Y_0}]) &= \mathcal{W}_1([X_{T_{n+1}-T_n}^{X,n}, [Y_{T_{n+1}-T_n, T_n}^{Y_n}]) \\ &\leq \mathcal{W}_1([X_{T_{n+1}-T_n}^{X,n}, [X_{T_{n+1}-T_n}^{Y_n,n}]) + \mathcal{W}_1([X_{T_{n+1}-T_n}^{Y_n,n}, [Y_{T_{n+1}-T_n, T_n}^{Y_n}]) \\ &\leq \underbrace{C e^{C_1 a_{n+1}^{-2}} e^{-\rho_{n+1}(T_{n+1}-T_n)}}_{\mu_{n+1}} \mathcal{W}_1([X_{T_n}^{X_0}, [Y_{T_n}^{Y_0}]) + \underbrace{C e^{C_1 a_{n+1}^{-2}} (a_n - a_{n+1}) \rho_{n+1}^{-1}}_{\lambda_{n+1}} \mathbb{E} V(Y_{T_n}^{Y_0}), \end{aligned}$$

The convergence is controlled by

$$\lambda_{n+1} := Ce^{C_1 a_{n+1}^{-2}} (a_n - a_{n+1}) \rho_{n+1}^{-1}$$

with

$$a_n \simeq \frac{A}{\sqrt{\log(T_n)}}$$

$$T_{n+1} \simeq Cn^{\beta+1}$$

$$a_n - a_{n+1} \asymp \frac{1}{n \log^{3/2}(n)}$$

$$e^{C_1 a_{n+1}^{-2}} \simeq n^{(\beta+1)C_1/A^2}$$

$$\rho_n^{-1} = e^{C_2 a_{n+1}^{-2}} \simeq n^{(\beta+1)C_2/A^2}$$

\implies Choosing $A > 0$ large enough yields the convergence to 0 of $\mathcal{W}_1([X_{T_{n+1}}^{x_0}], [Y_{T_{n+1}}^{x_0}])$ at rate $n^{-(1-(\beta+1)(C_1+C_2)/A^2)}$. Then :

$$\mathcal{W}_1([Y_{T_{n+1}}^{x_0}], \nu_{a_{n+1}}) \leq \mathcal{W}_1([Y_{T_{n+1}}^{x_0}], [X_{T_{n+1}}^{x_0}]) + \mathcal{W}_1([X_{T_{n+1}}^{x_0}], \nu_{a_{n+1}})$$

$$\mathcal{W}_1([Y_{T_{n+1}}^{x_0}], \nu^*) \leq \mathcal{W}_1([Y_{T_{n+1}}^{x_0}], [X_{T_{n+1}}^{x_0}]) + \mathcal{W}_1([X_{T_{n+1}}^{x_0}], \nu^*)$$

$$\bar{Y}_{\Gamma_{n+1}}^{\times 0} = \bar{Y}_{\Gamma_n} + \gamma_{n+1} \left(b_{a(\Gamma_n)}(\bar{Y}_{\Gamma_n}^{\times 0}) + \zeta_{n+1}(\bar{Y}_{\Gamma_n}^{\times 0}) \right) + a(\Gamma_n) \sigma(\bar{Y}_{\Gamma_n}^{\times 0}) (W_{\Gamma_{n+1}} - W_{\Gamma_n})$$

$$\gamma_{n+1} \text{ decreasing to } 0, \quad \sum_n \gamma_n = \infty, \quad \sum_n \gamma_n^2 < \infty,$$

$$\Gamma_n = \gamma_1 + \dots + \gamma_n.$$

We adopt the same strategy of proof to bound $\mathcal{W}_1(X, \bar{Y})$.

Thank you for your attention !