# **Provable Convergence and Limitations of Geometric Tempering for Langevin Dynamics**

**ENSAE** 

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## **Background**

**Sampling process:** Annealed Langevin Dynamics

**Prescribed path:** Interpolate target and Gaussian, using the geometric mean.

$$
\mu_t(x) = c_t \pi(x)^{\lambda_t} \times \mathcal{N}(x; 0, I)^{1 - \lambda_t}
$$
, increasing schedule  $\lambda_t \in [0, 1]$ 

Multimodality appears in small (inverse) log-Sobolev constant  $\alpha_t \geq 0$ .



This path is computationally convenient: the scores  $\nabla \log \mu_t(x)$  are known when  $\pi$  is

known up to a normalizing constant. But does it **accelerate convergence?**

$$
dx_t = \nabla \log \mu_t(x_t) dt + \sqrt{2} dw_t
$$

### **Our results**

### **Convergence is guaranteed:** upper bound

**Convergence can be accelerated** for a unimodal and peaky target: optimal schedule

#### **Convergence is provably slow** for a multimodal target**:** lower bound

$$
KL(p_t, \pi) \le \exp(-2\int_0^t \alpha_s ds) \cdot KL(p_0, \mu_0) + (1 - \lambda_t)A + A\int_0^t \lambda_s \exp(-2\int_s^t \alpha_v dv)ds
$$
  
Initial condition *Terminal condition* Speed of traversal  
vs. Multimodality

1 for a peakier target, 
$$
\alpha_{\pi} \ge \alpha_{\nu}/2
$$

$$
\lambda(t) = \begin{cases} 1 & \alpha_{\nu} = 1 + \alpha_{\nu}t \\ \frac{\alpha_{\nu}}{\alpha_{\nu} - \alpha_{\pi}} \frac{1 + \alpha_{\nu}t}{2 + \alpha_{\nu}t} \end{cases}
$$

for a flatter target, 
$$
\alpha_{\pi} < \alpha_{\nu}/2
$$

$$
TV(p_t, \pi) \ge \frac{1}{20} - 16 \cdot t \cdot e^{-m^2/64}
$$

#### Convergence time is exponential in the distance m between target modes.