# Underdamped Langevin MCMC

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## **Underdamped Langevin Diffusion Process**

Underdamped Langevin Diffusion Process:

$$dv_t = -\gamma v_t dt - u\nabla(f(x_t))dt + (\sqrt{2\gamma u})dB_t$$
$$dx_t = v_t dt$$

- $(x_t, v_t) \in \mathbb{R}^{2d}$  and f is twice continuously-differentiable
- Under fairly mild conditions, the invariant distribution of the process defined above has invariant distribution proportional to

$$exp(-(f(x) + ||v||_2^2/2u))$$

More specifically, for the marginal distribution of x from the invariant distribution would be proportional to:

$$exp(-f(x))$$



# Assumptions

Lipschitz Gradients

$$\|\nabla f(x) - \nabla f(y)\|_2 \le \|x - y\|_2$$

m-strongly convex

$$f(y) \ge f(x) + \langle \nabla f(x), y - x \rangle + \frac{m}{2} ||x - y||_2^2$$

Notations for f(x):

$$\kappa = L/m$$
$$x^* = \arg\min f(x)$$

#### **Notations**

- Let  $\mu$  and  $\nu$  be two probability measures on  $(\mathbb{R}^d, \mathcal{B}(\mathbb{R}^d))$
- ullet  $\Gamma(\mu, v)$  is the set of all couplings of measure  $\mu$  and v
- Wasserstein Distance

$$W_2(\mu, \upsilon) = (\inf_{\zeta \in \Gamma(\mu, \upsilon)} \int \|x - y\|_2^2 d\zeta(x, y))^{1/2}$$

■  $\Gamma_{opt}(\mu, v)$  is the optimal coupling that achieves the infimum of the Wasserstein distance

#### **Notations**

■ For the Langevin Diffusion Process defined by the SDE before, with initial condition  $(x_0, v_0) \sim p_0$  for some distribution  $p_0$  on  $\mathbb{R}^{2d}$ , let  $p_t$  be the distribution for  $(x_t, v_t)$  and  $\Phi_t$  be the operator that maps  $p_0$  to  $p_t$ 

$$\Phi_t p_0 = p_t$$

# Underdamped Langevin MCMC

- Discretized Underdamped Langevin Process
- Starting from  $(\tilde{x_0}, \tilde{v_0}) \in \mathbb{R}^{2d}$ , the discretized Underdamped Langevin Process will evolve according to the following SDE:

$$d\tilde{v}_t = -\gamma \tilde{v}_t - u \nabla f(\tilde{x}_0) dt + (\sqrt{2\gamma u}) dB_t$$
$$d\tilde{x}_t = \tilde{v}_t dt$$

- Denote  $Z^{\delta}(\tilde{x_0}, \tilde{v_0})$  as the distribution of  $(\tilde{x_\delta}, \tilde{v_\delta})$  evolving according to the discretized process starting from  $(\tilde{x_0}, \tilde{v_0})$
- The discretized version of Underdamped Langevin Process has an explicit form of solution for  $(\tilde{x_t}, \tilde{v_t})$  and easy to implement
- Parallelly, for the discretized SDE with initial condition  $(\tilde{x_0}, \tilde{v_0}) \sim p_0$  for some distribution  $\tilde{p_0}$  on  $\mathbb{R}^{2d}$ , let  $\tilde{p_t}$  be the distribution for  $(\tilde{x_t}, \tilde{v_t})$  and  $\tilde{\Phi_t}$  be the operator that maps  $\tilde{p_0}$  to  $\tilde{p_t}$

$$\tilde{\Phi}_t p_0 = \tilde{p}_t$$



# Underdamped Langevin MCMC

The algorithm for underdamped Langevin MCMC now goes as followed

#### Algorithm 1: Underdamped Langevin MCMC

Initialisation: Choose step-size  $\delta$ , iteration times n and starting point  $(x^0, v^0)$ ;

**for** 
$$0 \le i \le n - 1$$
 **do**

Sample 
$$(x^{i+1}, v^{i+1}) \sim Z^{\delta}(x^i, v^i)$$

end

#### Convergence

For the following analysis on convergence, denote by  $p^*$  the unique invariant distribution of the Underdamped Langevin Process such that  $p^* \propto exp(-(f(x) + \frac{1}{2} ||v||_2^2))$ 

$$x + v$$
), and  $q_t$  be the distribution of  $g(x_t, x_t + v_t)$ 

- Let g(x, v) = (x, x + v), and  $q_t$  be the distribution of  $g(x_t, x_t + v_t)$ ,  $q^*$  be the distribution of g(x, v) when  $(x, v) \sim p^*$
- In the following analysis, we would set u=1/L and  $\gamma=2$

## Convergence

#### Theorem

Let  $(x_0, v_0)$  and  $(y_0, w_0)$  be two arbitrary points in  $\mathbb{R}^{2d}$ . Let  $p_0$  and  $p_0'$  be two dirac measures concentrated on  $(x_0, v_0)$  and  $(y_0, w_0)$ . If we set u = 1/L and  $\gamma = 2$ , then for every t > 0, there exists a  $\zeta_t(x_0, v_0, y_0, w_0) \in \Gamma(\Phi_t p_0, \Phi_t p_0')$  such that:

$$\mathbb{E}_{(x_t, v_t, y_t, w_t) \sim \zeta_t(x_0, v_0, y_0, w_0)} [\|x_t - y_t\|_2^2 + \|(x_t + v_t) - (y_t + w_t)\|_2^2] \le e^{-t/\kappa} [\|x_0 - y_0\|_2^2 + \|(x_0 + v_0) - (y_0 + w_0)\|_2^2]$$

■ From the theorem above, let  $(x_0, v_0) \sim p_0$  and  $g(x_0, v_0) \sim q_0$ , we could easily get:

$$W_2(\Phi_t q_0, q^*) \le e^{-t/2\kappa} W_2(q_0, q^*)$$
  
 $W_2(\Phi_t p_0, p^*) \le 4e^{-t/2\kappa} W_2(p_0, p^*)$ 



## Convergence<sup>1</sup>

- The previous theorem only guarantees the convergence of the continuous process. However, for the discretized version, more assumptions are needed.
- Let  $\delta$  represents a single step of the Underdamped Langevin MCMC algorithm,  $p_t$  be the probability distribution of  $(x_t, v_t)$  from the continuous process and  $\tilde{p}_t$  be the distribution from the discretized version. Recall the definitions of  $\tilde{\Phi}_t$  and  $\Phi_t$  aforementioned.
- **Assumptions:** For the continuous time process, there exists  $\mathcal{E}_{\kappa}$  such that:

$$\forall t \in [0, \delta]$$
  $\mathbb{E}_{p_t}[\|v\|_2^2] \leq \mathcal{E}_{\kappa}$ 

lacksquare  $\mathcal{E}_k$  could be explicitly bounded by function of paramters  $m, \ L$  and d

## Convergence

The following theorem bounds the distance between the continuous process and the discretized process with one step.

#### Theorem

Let  $\tilde{\Phi}_t$  and  $\Phi_t$  be the probability transfer operator as aforementioned. Let  $p_0$  be any arbitrary distributions and the step-size  $\delta < 1$ . If we choose u = 1/L and  $\gamma = 2$ , the Wasserstein distance between the continuous process and the discretized process is upper bounded by:

$$W_2(\tilde{\Phi}_t p_0, \Phi_t p_0) \leq \delta^2 \sqrt{\frac{2\mathcal{E}_{\kappa}}{5}}$$

### Convergence<sup>1</sup>

#### **Theorem**

Let  $p^{(n)}$  be the distribution of the Underdamped Langevin MCMC algorithm after n steps starting from initial distribution  $p^{(0)} = \mathbf{1}_{\{x=x^{(0)},v=0\}}$ , and the initial distribution satisfies  $\|x^{(0)}-x^*\| \leq D^2$ . If we set the step-size to be:

$$\delta = \frac{\epsilon}{104\kappa} \sqrt{\frac{1}{d/m + D^2}}$$

and run the algorithm for:

$$n \geq \big(\frac{52\kappa^2}{\epsilon}\big) \cdot \big(\sqrt{\frac{d}{m} + D^2}\big) \cdot log\big(\frac{24(\frac{d}{m} + R^2)}{\epsilon}\big)$$

we shall have the guarantee that:

$$W_2(p^{(n)}, p^*) \le \epsilon$$



# Comparisons to traditional MCMC

- Strong assumptions for the invariant distribution proportional to exp(-f(x))
- Under mild conditions, Metropolis-Hasting MCMC would converge in terms of total variation distance, the Underdamped Langevin is shown converge only in Wasserstein Distance.

### References I