

1. Pedro Talk

The standard gradient descent dynamics is given by

$$dx(t) = -\nabla U(x(t))dt + \sigma dW(t), \quad x(0) = x_0.$$

In this note we will try to study the modified dynamics

$$(1) \quad \begin{aligned} dX_\varepsilon(t) &= -e^{-\gamma(U(X_\varepsilon(t)) - \min_{s \leq t} U(X_\varepsilon(s)))} \nabla U(X_\varepsilon(t))dt + \varepsilon dW(t), \\ X_\varepsilon(0) &= x_0 \end{aligned}$$

2. Idea of Analysis

In the analysis of (1), let us assume that x_0 is a local minimum and a stationary point; that is, $\nabla U(x_0) = 0$. Define the deterministic flow induced by the vector field ∇U starting at an arbitrary point $x \in \mathbb{R}^d$:

$$(2) \quad \frac{d}{dt} S^t x = -\nabla U(S^t x), \quad S^0 x = x.$$

Then, consider the proxy process to (1) given by

$$(3) \quad d\hat{X}_\varepsilon(t) = -e^{-\gamma(U(\hat{X}_\varepsilon(t)) - \min_{s < t - \tau_i} U(S^s \hat{X}_\varepsilon(\tau_i)))} \nabla U(\hat{X}_\varepsilon(t))dt + \varepsilon dW(t),$$

on the interval $\tau_i < t < \tau_{i+1}$, where the times τ_i are the innovation times defined via

$$\tau_i = \inf \left\{ t \geq \tau_{i-1} : U(\hat{X}_\varepsilon(t)) < U(\hat{X}_\varepsilon(\tau_{i-1})) \right\},$$

with $\tau_0 = 0$ and the same initial condition as X_ε , $\hat{X}_\varepsilon(0) = x_0$. Let us now make some observations.

The time τ_1 is bounded below by the exit time of \hat{X}_ε from the basin of attraction, $\mathcal{B}(x_0)$, of x_0 ; that is, $\tau_1 > \sigma(x_0) = \inf \left\{ t : \hat{X}_\varepsilon(t) \in \partial \mathcal{B}(x_0) \right\}$, since for every $y \in \mathcal{B}(x_0)$, $\nabla U(y) \neq 0$. Now, from FW theory, it is well known that $\sigma(x_0)$ is exponentially distributed with mean $\varepsilon^{-2} (\min_{y \in \partial \mathcal{B}(x_0)} V(y) - V(x_0))$, where V is the quasi-potential of (3). Since, \hat{X}_ε does not find a new minimum of the function U before time $\sigma(x_0)$, and since $\min_{t > 0} U(S^t x_0) = U(x_0)$ we observe that the drift in equation (3) is given by

$$b(x) = -\gamma^{-1} \nabla \left(1 - e^{-\gamma(U(y) - U(x_0))} \right),$$

so the quasi-potential is given by $V(y) = 2 - 2e^{-\gamma(U(y) - U(x_0))}$. As a consequence, the exit time $\sigma(x_0)$ is exponentially distributed with mean λ given by

$$\log \lambda = \frac{2}{\gamma \varepsilon^2} \left(1 - e^{-\gamma(\min_{y \in \partial \mathcal{B}(x_0)} U(y) - U(x_0))} \right).$$

So that if this expression is of constant order, then for some constant $c > 0$ and with $\Delta_{x_0} = \min_{y \in \partial \mathcal{B}(x_0)} U(y) - U(x_0)$ it follows that, for γ and ε small enough,

$$\begin{aligned} \gamma \Delta_{x_0} &= \log(1 - \gamma \varepsilon^2 c) \\ &= \log \left(1 + \sum_{i \geq 1} \gamma^i \varepsilon^{2i} c^i \right) \\ &\approx \gamma \varepsilon^2 c + \gamma^2 \varepsilon^4 c^2. \end{aligned}$$

This quadratic equation has a unique strictly positive solution given by $\varepsilon^4 c^2 \gamma = \Delta_{x_0} - \varepsilon^2 c$, which leads to the following:

CLAIM 1. *By choosing $\gamma = \mathcal{O}(\varepsilon^{-4})$, as $\varepsilon \rightarrow 0$, the exit from the basin of attraction $\mathcal{B}(x_0)$ happens in almost constant time.*

3. Some Potential Discretizations

Let us consider SDE (1) up to time τ_1 , so that the law of X_ε is the same as the law of

$$dY_\varepsilon(t) = -e^{-\gamma(U(Y_\varepsilon(t)) - U(x_0))} \nabla U(Y_\varepsilon(t)) dt + \varepsilon dW(t)$$

Then observe that, if we create a uniform equidistant discretization, $\mathcal{P} = \{0 = t_0 < t_1 < \dots\}$ of $[0, \infty)$ with $t_1 = \eta$, then it follows that

$$\begin{aligned} Y_\varepsilon(t_{i+1}) &= Y_\varepsilon(t_i) - \int_{t_i}^{t_{i+1}} e^{-\gamma(U(Y_\varepsilon(s)) - U(x_0))} \nabla U(Y_\varepsilon(s)) ds + \varepsilon (W(t_{i+1}) - W(t_i)) \\ &\approx Y_\varepsilon(t_i) - \eta \left(e^{-\gamma(U(Y_\varepsilon(t_i)) - U(x_0))} \nabla U(Y_\varepsilon(t_i)) + \frac{\varepsilon}{\sqrt{\eta}} \xi_i \right). \end{aligned}$$

By comparing the last expression with the characterization of the algorithm, it follows that $\varepsilon = \sqrt{\eta}$, so that $\gamma \approx \eta^{-2}$ giving rise to the claim.

CLAIM 2. *Let $y_0 = x_0$, and consider the algorithm,*

$$y_{i+1} = y_i - \eta \left(e^{-\eta^{-2}(U_i - U_i^*)} \nabla U_i + \xi_i \right), \quad i \in \mathbb{N},$$

where we have used the short hand notation $U_i = U(y_i)$ and $U_i^ = \min_{j \leq i} U_j$. Then, with high probability, after $\mathcal{O}(\gamma^{-1})$ time steps, y_i escapes the set $\mathcal{B}(x_0)$.*

Now, as explained below, exit from the basin of attraction occurs in time