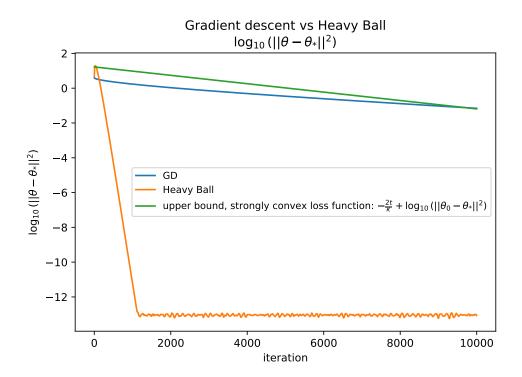
FTML Exercices 5

Pour le 4 mai 2023



1 HEAVY-BALL

In this exercise we study the values of the γ and β parameters in a Heavy-Ball algorithm, which is an acceleration of gradient descent, in the case where f is a least squares loss function.

$$\theta_{t+1} = \theta_t - \gamma \nabla_{\theta} f(\theta_t) + \beta(\theta_t - \theta_{t-1}) \tag{1}$$

The update $\theta_{t+1} - \theta_t$ is then a combination of the gradient $\nabla_{\theta_t} f$ and of the previous update $\theta_t - \theta_{t-1}$. The goal of this method it might balance the effet of oscillations in the gradient. We will use these parameters :

$$\gamma = \frac{4}{(\sqrt{L} + \sqrt{\mu})^2} \tag{2}$$

and

$$\beta = \left(\frac{\sqrt{L} - \sqrt{\mu}}{\sqrt{L} + \sqrt{\mu}}\right)^2 \tag{3}$$

Let λ be an eigenvalue of H and u_{λ} a eigenvector for thie eigenvalue. We are interested in the evolution of $\langle \theta_t - \eta^*, u_{\lambda} \rangle$.

We note

$$a_{t} = \langle \theta_{t} - \eta^{*}, u_{\lambda} \rangle \tag{4}$$

Exercice 1: Show that

$$a_{t+1} = (1 - \gamma \lambda + \beta) a_t - \beta a_{t-1}$$
(5)

Exercice 2: Compute the constant-recursive sequence a_t , and show that there exists a constant C_{λ} that depends on the initial conditions, such that

$$\forall t, a_t \leqslant (\sqrt{\beta})^t C_{\lambda} \tag{6}$$

https://en.wikipedia.org/wiki/Constant-recursive_sequence

If u_i is a basis of orthonormal vectors with eigenvalues λ_i , we have that

$$\begin{split} \|\theta_{t} - \eta^{*}\|^{2} &= \sum_{i=1}^{d} (\langle \theta_{t} - \eta^{*}, u_{i} \rangle)^{2} \\ &\leq \sum_{i=1}^{d} (\sqrt{\beta})^{2t} C_{\lambda_{i}} \\ &= (\sqrt{\beta})^{2t} D \end{split} \tag{7}$$

with

$$D = \sum_{i=1}^{d} C_{\lambda_i}$$
 (8)

We can now remark that

$$\sqrt{\beta} = \frac{\sqrt{L} - \sqrt{\mu}}{\sqrt{L} + \sqrt{\mu}}$$

$$= \frac{1 - \sqrt{\frac{\mu}{L}}}{1 + \sqrt{\frac{\mu}{L}}}$$

$$\leqslant 1 - \sqrt{\frac{\mu}{L}}$$

$$= 1 - \frac{1}{\sqrt{\kappa}}$$
(9)

Exercice 3: Conclude