## A continuously differentiable convex function $E: \mathbb{R}^n \to \mathbb{R}$ is L-smooth if and only if $\frac{1}{L}\nabla E$ is firmly nonexpansive

If  $\frac{1}{L}\nabla E$  is firmly nonexpansive, the L-smoothness is a simple conclusion of the Cauchy-Schwarz inequality.

Now let  $E: \mathbb{R}^n \to \mathbb{R}$  be L-smooth. We show the assertion in two steps

- 1. L-smoothness implies that  $\frac{L}{2}||u||^2 E(u)$  is convex.
- 2. If  $\frac{L}{2}||u||^2 E(u)$  is convex, then  $\frac{1}{L}\nabla E$  is firmly nonexpansive.

The first step becomes easy if we assume we know that for a differentiable function E convexity is equivalent to monotononicity of the gradient, i.e.

$$\langle \nabla E(u) - \nabla E(v), u - v \rangle \ge 0.$$

In this case, L-smoothness quickly yields the convexity of  $\frac{L}{2}||u||^2 - E(u)$ .

Let a function  $G(u) = \frac{L}{2}||u||^2 - E(u)$  be convex and differentiable, and fix some point u. Then it holds that all v that

$$\begin{split} 0 \leq & G(v) - G(u) - \langle \nabla G(u), v - u \rangle \\ &= \frac{L}{2} \|v\|^2 - E(v) - \frac{L}{2} \|u\|^2 + E(u) - \langle Lu - \nabla E(u), v - u \rangle \\ &= E(u) - E(v) + \frac{L}{2} \|v\|^2 - \frac{L}{2} \|u\|^2 + L \|u\|^2 - L \langle u, v \rangle - \langle \nabla E(u), v - u \rangle \\ &= E(u) - E(v) + \frac{L}{2} \|v\|^2 - L \langle u, v \rangle + \frac{L}{2} \|u\|^2 - \langle \nabla E(u), v - u \rangle \\ &= E(u) - E(v) + \frac{L}{2} \|v - u\|^2 - \langle \nabla E(u), v - u \rangle \end{split}$$

In other words, we derived the quadratic upper bound

$$\min_{z} E(z) \le E(v) \le E(u) - \langle \nabla E(u), v - u \rangle + \frac{L}{2} ||v - u||^{2}.$$

Minimizing the right hand side yields  $v^* = u - \frac{1}{L}E(u)$  and therefore

$$\min_{z} E(z) \le E(u) - \frac{1}{2L} \|\nabla E(u)\|^{2}. \tag{1}$$

Now consider the function

$$E_v(w) = E(w) - \langle \nabla E(v), w \rangle.$$

It holds that  $E_v$  is L-smooth and convex, since it is just a linear perturbation of E. Furthermore,  $\nabla E_v(v) = 0$ , such that v has to minimize  $E_v$ . According to (1) it holds for any element u that

$$E_v(v) \le E_v(u) - \frac{1}{2L} \|\nabla E_v(u)\|^2$$
  
=  $E_v(u) - \frac{1}{2L} \|\nabla E(u) - \nabla E(v)\|^2$ 

But symmetry also

$$E_u(u) \le E_u(v) - \frac{1}{2L} \|\nabla E_u(v)\|^2$$
  
=  $E_u(v) - \frac{1}{2L} \|\nabla E(v) - \nabla E(u)\|^2$ .

Adding these two inequalities one obtains

$$E_v(v) + E_u(u) \le E_u(v) + E_v(u) - \frac{1}{L} \|\nabla E(u) - \nabla E(v)\|^2$$

and we compute

$$\begin{split} E_v(v) + E_u(u) - E_u(v) - E_v(u) \\ = & E(v) - \langle \nabla E(v), v \rangle + E(u) - \langle \nabla E(u), u \rangle - E(v) + \langle \nabla E(u), v \rangle - E(u) + \langle \nabla E(v), u \rangle \\ = & \langle \nabla E(v), u - v \rangle - \langle \nabla E(u), u - v \rangle \\ = & \langle \nabla E(v) - E(u), u - v \rangle \end{split}$$

which yields the assertion.