

Assume $f \in C^1$, f convex, ∂f convex.

$$L\text{-smooth} \left(\|\nabla f(x) - \nabla f(y)\|_* \leq L \|x - y\| \right)$$

$$\langle \nabla f(x) - \nabla f(y), x - y \rangle \leq L \|x - y\|^2$$

$$f(y) \leq f(x) + \langle \nabla f(x), y - x \rangle + \frac{L}{2} \|y - x\|^2$$

$$\langle \nabla f(x) - \nabla f(y), x - y \rangle \geq \frac{1}{L} \|\nabla f(x) - \nabla f(y)\|_*^2$$

Generalization:

$$\nabla f: \mathbb{R}^n \mapsto \mathbb{R}^n$$

$$\partial f: \mathbb{R}^n \mapsto 2^{\mathbb{R}^n} \text{ set-value mapping}$$

(monotone operator)

$$\frac{1}{2L} \|\nabla f(x)\|_*^2 \leq f(x) - f(x^*) \leq \frac{L}{2} \|x - x^*\|^2$$

$$\frac{L}{2} \|x\|_2^2 - f(x) \text{ is convex.}$$

$$f: m\text{-strongly convex} \iff h(x) = f(x) - \frac{m}{2} \|x\|^2 \text{ convex} \iff g(t) = f(x + t(y-x))$$

$$\downarrow$$

x^* exists and unique

Strong monotonicity

$$\langle \nabla f(x) - \nabla f(y), x - y \rangle \geq m \|x - y\|^2$$

$$\downarrow -\frac{m}{2} t^2 \|y - x\|^2 \text{ convex.}$$

$$g'(t) = g'(0) + \int_0^t g''(s) ds$$

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

$$\downarrow S_\alpha = \{x | f(x) \leq \alpha\} \text{ bounded}$$

If f is L -smooth.

$h: L-m$ smooth

$$\omega\text{-coercivity: } \langle \nabla f(x) - \nabla f(y), x - y \rangle \geq \frac{mL}{m+L} \|x - y\|^2 + \frac{1}{m+L} \|\nabla f(x) - \nabla f(y)\|^2$$

Gradient descent:

$$X_{k+1} = X_k - \alpha \nabla f(X_k)$$

$$= \arg \min_X \left\{ f(X_k) + \langle \nabla f(X_k), X - X_k \rangle + \frac{1}{2\alpha} \|X - X_k\|^2 \right\}$$

$$\nabla f(X_k) + \frac{1}{\alpha} (X_{k+1} - X_k) = 0 \iff X_{k+1} = X_k + \alpha \nabla f(X_k)$$

$$\text{If } \frac{1}{\alpha} \geq L$$

