

Suppose $f(x)$ is c -strongly convex, L -smooth

& x^* is in a set B , e.g., the set of vectors with at most S -nonzeros.

Consider Averaged Projected Gradient Descent:

$$y^{(k)} = \pi_B \left(x^{(k)} - \frac{1}{\mu L} \nabla f(x^{(k)}) \right)$$

$$(\mu \in (0, 1))$$

$$x^{(k+1)} = (1-\mu)x^{(k)} + \mu y^{(k)}$$

Remark: Note that

$$y^{(k)} = \arg \min_{y \in B} \|y - (x^{(k)} - \frac{1}{\mu L} \nabla f(x^{(k)}))\|^2$$

$$= \arg \min_{y \in B} \langle \nabla f(x^{(k)}), y - x^{(k)} \rangle + \frac{\mu L}{2} \|y - x^{(k)}\|^2 \quad (1)$$

Thm: If $\mu \in (0, \frac{1}{K})$ where $K = \frac{L}{c}$, then

$$f(x^{(k)}) - f(x^*) \leq (1-\mu)^k (f(x^{(0)}) - f(x^*))$$

Proof:

L -Smooth.

$$f(x^{(k+1)}) = f(x^{(k)} + \mu(y^{(k)} - x^{(k)})) \stackrel{\downarrow}{\leq} f(x^{(k)}) + \mu \langle \nabla f(x^{(k)}), y^{(k)} - x^{(k)} \rangle$$

$$+ \frac{\mu^2 L}{2} \|y^{(k)} - x^{(k)}\|^2$$

$$\text{optimality of } y^{(k)} \rightarrow \leq f(x^{(k)}) + \mu \langle \nabla f(x^{(k)}), x^* - x^{(k)} \rangle$$

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$$+ \frac{\mu^2 L}{2} \|x^* - x^{(k)}\|^2$$

$$\text{Strong convexity} \rightarrow \leq f(x^{(k)}) + \mu \left(f(x^*) - f(x^{(k)}) - \frac{c}{2} \|x^* - x^{(k)}\|^2 \right)$$

$$\begin{aligned} f(x^*) &\geq f(x^{(k)}) + \langle \nabla f(x^{(k)}), x^* - x^{(k)} \rangle \\ &\quad + \frac{c}{2} \|x^{(k)} - x^*\|^2 \\ &= (1-\mu)(f(x^{(k)}) - f(x^*)) + \underbrace{\left(-\frac{c\mu}{2} + \frac{\mu^2 L}{2} \right)}_{\mu \in \frac{1}{K}} \|x^* - x^{(k)}\|^2 \\ &\quad + f(x^*) \end{aligned}$$

$$\leq (1-\mu)(f(x^{(k)}) - f(x^*)) + f(x^*)$$

$\mu \in \frac{1}{K}$

makes this ≤ 0

$$\begin{aligned}
 \Leftrightarrow f(x^{(k+1)}) - f(x^*) &\leq (1-\mu) (f(x^{(k)}) - f(x^*)) \\
 &\leq \dots \leq (1-\mu)^{(k+1)} (f(x^{(0)}) - f(x^*)) \quad \square
 \end{aligned}$$