

Theorem 1. Let $\mathbf{f}(\mathbf{x})$ be a strongly convex function on a set S with convexity parameter μ , then the term $\|\mathbf{x} - \mathbf{x}^*\|$, which is the distance between any point and optimal point in the set S is upper bounded by $\frac{2\|\nabla \mathbf{f}\|_2}{\mu}$.

Proof. $\mathbf{f}(\mathbf{x})$ is a strongly convex function implies (1) for any $x, y \in S$

$$\mathbf{f}(\mathbf{y}) \geq \mathbf{f}(\mathbf{x}) + \nabla^\top \mathbf{f}(\mathbf{x})(\mathbf{y} - \mathbf{x}) + \frac{\mu \|\mathbf{x} - \mathbf{y}\|^2}{2} \quad (1)$$

The righthand side is a convex quadratic function of y (for a fixed x). Setting the gradient with respect to y equal to zero, we find that $\tilde{\mathbf{y}} = \mathbf{x} - \frac{\nabla \mathbf{f}(\mathbf{x})}{\mu}$ minimizes the RHS of (1). Therefore, we have

$$\begin{aligned} \mathbf{f}(\mathbf{y}) &\geq \mathbf{f}(\mathbf{x}) + \langle \nabla \mathbf{f}(\mathbf{x}), \mathbf{y} - \mathbf{x} \rangle + \frac{\mu}{2} \|\mathbf{y} - \mathbf{x}\|^2 \\ &\geq \mathbf{f}(\mathbf{x}) + \langle \nabla \mathbf{f}(\mathbf{x}), \tilde{\mathbf{y}} - \mathbf{x} \rangle + \frac{\mu}{2} \|\tilde{\mathbf{y}} - \mathbf{x}\|^2 \\ &= \mathbf{f}(\mathbf{x}) - \frac{\|\tilde{\mathbf{y}} - \mathbf{x}\|^2}{2\mu} \end{aligned} \quad (2)$$

Since this holds for any $y \in S$, $\mathbf{f}^* \geq \mathbf{f}(\mathbf{x}) - \frac{1}{2\mu} \|\nabla \mathbf{f}(\mathbf{x})\|^2$

Here $\mathbf{x}^* = \underset{\mathbf{x}}{\operatorname{argmin}} \mathbf{f}(\mathbf{x})$. Similarly, one can derive a bound on $\|\mathbf{x} - \mathbf{x}^*\|$, the distance between \mathbf{x} and any optimal point \mathbf{x}^* in terms of $\|\nabla \mathbf{f}(\mathbf{x})\|$. First apply (1) with $\mathbf{y} = \mathbf{x}^*$ to obtain:

$$\begin{aligned} \mathbf{f}^* = \mathbf{f}(\mathbf{x}^*) &\geq \mathbf{f}(\mathbf{x}) + \langle \nabla \mathbf{f}(\mathbf{x}), \mathbf{x}^* - \mathbf{x} \rangle + \frac{\mu}{2} \|\mathbf{x} - \mathbf{x}^*\|^2 \\ &\geq \mathbf{f}(\mathbf{x}) - \|\nabla \mathbf{f}(\mathbf{x})\| \|\mathbf{x}^* - \mathbf{x}\| + \frac{\mu}{2} \|\mathbf{x} - \mathbf{x}^*\|^2 \end{aligned} \quad (3)$$

Here we use Cauchy-Schwarz inequality in the second part of (3). Since $\mathbf{f}^* \leq \mathbf{f}(\mathbf{x})$, one must have,

$$-\|\nabla \mathbf{f}(\mathbf{x})\| \|\mathbf{x}^* - \mathbf{x}\| + \frac{\mu}{2} \|\mathbf{x} - \mathbf{x}^*\|^2 \leq 0 \quad (4)$$

$$\|\mathbf{x} - \mathbf{x}^*\| \leq \frac{2\|\nabla \mathbf{f}(\mathbf{x})\|}{\mu} \quad \square$$

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