6. 
$$f(\alpha \boldsymbol{x} + (1-\alpha)\boldsymbol{y}) + \frac{\alpha(1-\alpha)}{2L} \|f'(\boldsymbol{x}) - f'(\boldsymbol{y})\|_2^2 \le \alpha f(\boldsymbol{x}) + (1-\alpha)f(\boldsymbol{y}).$$
  
7. 
$$0 \le \alpha f(\boldsymbol{x}) + (1-\alpha)f(\boldsymbol{y}) - f(\alpha \boldsymbol{x} + (1-\alpha)\boldsymbol{y}) \le \alpha(1-\alpha)\frac{L}{2}\|\boldsymbol{x} - \boldsymbol{y}\|_2^2.$$

Proof:

 $1 \Rightarrow 2$  It follows from the definition of convex function and Lemma 3.4.

2 $\Rightarrow$ 3 Fix  $\boldsymbol{x} \in \mathbb{R}^n$ , and consider the function  $\phi(\boldsymbol{y}) = f(\boldsymbol{y}) - \langle f'(\boldsymbol{x}), \boldsymbol{y} \rangle$ . Clearly  $\phi(\boldsymbol{y})$  satisfies 2. Also,  $\boldsymbol{y}^* = \boldsymbol{x}$  is a minimal solution. Therefore from 2,

$$\begin{split} \phi(\boldsymbol{x}) &= \phi(\boldsymbol{y}^*) \leq \phi\left(\boldsymbol{y} - \frac{1}{L}\phi'(\boldsymbol{y})\right) \leq \phi(\boldsymbol{y}) + \frac{L}{2} \left\|\frac{1}{L}\phi'(\boldsymbol{y})\right\|_2^2 + \langle\phi'(\boldsymbol{y}), -\frac{1}{L}\phi'(\boldsymbol{y})\rangle \\ &= \phi(\boldsymbol{y}) + \frac{1}{2L} \|\phi'(\boldsymbol{y})\|_2^2 - \frac{1}{L} \|\phi'(\boldsymbol{y})\|_2^2 = \phi(\boldsymbol{y}) - \frac{1}{2L} \|\phi'(\boldsymbol{y})\|_2^2. \end{split}$$

Since  $\phi'(\boldsymbol{y}) = f'(\boldsymbol{y}) - f'(\boldsymbol{x})$ , finally we have

$$f(oldsymbol{x}) - \langle f'(oldsymbol{x}), oldsymbol{x} 
angle \leq f(oldsymbol{y}) - \langle f'(oldsymbol{x}), oldsymbol{y} 
angle - rac{1}{2L} \|f'(oldsymbol{y}) - f'(oldsymbol{x})\|_2^2.$$

 $3 \Rightarrow 4$  Adding two copies of 3 with  $\boldsymbol{x}$  and  $\boldsymbol{y}$  interchanged, we obtain 4.

4 $\Rightarrow$ 1 Applying the Cauchy-Schwarz inequality to 4, we obtain  $||f'(\boldsymbol{x}) - f'(\boldsymbol{y})||_2 \leq L||\boldsymbol{x} - \boldsymbol{y}||_2$ . Also from Theorem 6.10,  $f(\boldsymbol{x})$  is convex.

2 $\Rightarrow$ 5 Adding two copies of 2 with  $\boldsymbol{x}$  and  $\boldsymbol{y}$  interchanged, we obtain 5. 5 $\Rightarrow$ 2

$$egin{aligned} f(oldsymbol{y}) - f(oldsymbol{x}) - \langle f'(oldsymbol{x}), oldsymbol{y} - oldsymbol{x} 
angle ) &= \int_0^1 \langle f'(oldsymbol{x} + au(oldsymbol{y} - oldsymbol{x})) - f'(oldsymbol{x}), oldsymbol{y} - oldsymbol{x} 
angle d au \ &\leq \int_0^1 au L \|oldsymbol{y} - oldsymbol{x}\|_2^2 d au = rac{L}{2} \|oldsymbol{y} - oldsymbol{x}\|_2^2. \end{aligned}$$

The non-negativity follows from Theorem 6.10.

 $3 \Rightarrow 6$  Denote  $\boldsymbol{x}_{\alpha} = \alpha \boldsymbol{x} + (1 - \alpha) \boldsymbol{y}$ . From 3,

$$f(\boldsymbol{x}) \geq f(\boldsymbol{x}_{\alpha}) + \langle f'(\boldsymbol{x}_{\alpha}), (1-\alpha)(\boldsymbol{x}-\boldsymbol{y}) \rangle + \frac{1}{2L} \|f'(\boldsymbol{x}) - f'(\boldsymbol{x}_{\alpha})\|_{2}^{2}$$
  
$$f(\boldsymbol{y}) \geq f(\boldsymbol{x}_{\alpha}) + \langle f'(\boldsymbol{x}_{\alpha}), \alpha(\boldsymbol{y}-\boldsymbol{x}) \rangle + \frac{1}{2L} \|f'(\boldsymbol{y}) - f'(\boldsymbol{x}_{\alpha})\|_{2}^{2}.$$

Multiplying the first inequality by  $\alpha$ , the second by  $1 - \alpha$ , and summing up, we have

$$\alpha f(\boldsymbol{x}) + (1-\alpha)f(\boldsymbol{y}) \ge f(\boldsymbol{x}_{\alpha}) + \frac{\alpha}{2L} \|f'(\boldsymbol{x}) - f'(\boldsymbol{x}_{\alpha})\|_{2}^{2} + \frac{1-\alpha}{2L} \|f'(\boldsymbol{y}) - f'(\boldsymbol{x}_{\alpha})\|_{2}^{2}$$

Finally, using the inequality

$$\alpha \| \boldsymbol{b} - \boldsymbol{d} \|_{2}^{2} + (1 - \alpha) \| \boldsymbol{c} - \boldsymbol{d} \|_{2}^{2} \ge \alpha (1 - \alpha) \| \boldsymbol{b} - \boldsymbol{c} \|_{2}^{2}$$

we have the result.

$$\begin{pmatrix} -\alpha(1-\alpha) \| \boldsymbol{b} - \boldsymbol{c} \|_{2}^{2} \ge -\alpha(1-\alpha)(\| \boldsymbol{b} - \boldsymbol{d} \|_{2} + \| \boldsymbol{c} - \boldsymbol{d} \|)_{2}^{2} \\ \text{Therefore} \\ \alpha \| \boldsymbol{b} - \boldsymbol{d} \|_{2}^{2} + (1-\alpha) \| \boldsymbol{c} - \boldsymbol{d} \|_{2}^{2} - \alpha(1-\alpha)(\| \boldsymbol{b} - \boldsymbol{d} \|_{2} + \| \boldsymbol{c} - \boldsymbol{d} \|_{2})^{2} \\ = (\alpha \| \boldsymbol{b} - \boldsymbol{d} \|_{2} - (1-\alpha) \| \boldsymbol{c} - \boldsymbol{d} \|_{2})^{2} \ge 0$$

 $6\Rightarrow3$  Dividing both sides by  $1-\alpha$  and tending  $\alpha$  to 1, we obtain 3.

 $2 \Rightarrow 7$  From 2,

$$\begin{aligned} f(\boldsymbol{x}) &\leq f(\boldsymbol{x}_{\alpha}) + \langle f'(\boldsymbol{x}_{\alpha}), (1-\alpha)(\boldsymbol{x}-\boldsymbol{y}) \rangle + \frac{L}{2}(1-\alpha)^{2} \|\boldsymbol{x}-\boldsymbol{y}\|_{2}^{2} \\ f(\boldsymbol{y}) &\leq f(\boldsymbol{x}_{\alpha}) + \langle f'(\boldsymbol{x}_{\alpha}), \alpha(\boldsymbol{y}-\boldsymbol{x}) \rangle + \frac{L}{2} \alpha^{2} \|\boldsymbol{x}-\boldsymbol{y}\|_{2}^{2} \end{aligned}$$

Multiplying the first inequality by  $\alpha$ , the second by  $1 - \alpha$ , and summing up, we have

$$\alpha f(\boldsymbol{x}) + (1-\alpha)f(\boldsymbol{y}) \le f(\boldsymbol{x}_{\alpha}) + \frac{L}{2} \left( \alpha (1-\alpha)^2 + (1-\alpha)\alpha^2 \right) \|\boldsymbol{x} - \boldsymbol{y}\|_2^2.$$

The non-negativity follows from Theorem 6.10.

 $7\Rightarrow2$  Dividing both sides by  $1-\alpha$  and tending  $\alpha$  to 1, we obtain 2. The non-negativity follows from Theorem 6.10.

## 6.4 Diffentiable Strongly Convex Functions

**Definition 6.14** A continuously differentiable function  $f(\mathbf{x})$  is called *strongly convex* on  $\mathbb{R}^n$  (notation  $f \in \mathcal{S}^1_{\mu}(\mathbb{R}^n)$ ) if there exists a constant  $\mu > 0$  such that

$$f(\boldsymbol{y}) \geq f(\boldsymbol{x}) + \langle f'(\boldsymbol{x}), \boldsymbol{y} - \boldsymbol{x} \rangle + rac{1}{2} \mu \| \boldsymbol{y} - \boldsymbol{x} \|_2^2, \quad orall \boldsymbol{x}, \boldsymbol{y} \in \mathbb{R}^n.$$

The constant  $\mu$  is called the *convexity parameter* of the function f.

**Example 6.15** The following functions are strongly convex functions:

- 1.  $f(\boldsymbol{x}) = \frac{1}{2} \|\boldsymbol{x}\|_2^2$ .
- 2.  $f(\boldsymbol{x}) = \alpha + \langle \boldsymbol{a}, \boldsymbol{x} \rangle + \frac{1}{2} \langle \boldsymbol{A} \boldsymbol{x}, \boldsymbol{x} \rangle$ , for  $\boldsymbol{A} \succeq \mu \boldsymbol{I}$ .
- 3. A sum of a convex and a strongly convex functions.

**Corollary 6.16** If  $f \in \mathcal{S}^1_{\mu}(\mathbb{R}^n)$  and  $f'(\boldsymbol{x}^*) = 0$ , then

$$f(\boldsymbol{x}) \ge f(\boldsymbol{x}^*) + \frac{1}{2}\mu \|\boldsymbol{x} - \boldsymbol{x}^*\|_2^2, \quad \forall \boldsymbol{x} \in \mathbb{R}^n.$$

Proof:

Left for exercise.

**Theorem 6.17** Let f be a continuously differentiable function. The following conditions are equivalent:

1. 
$$f \in S^1_{\mu}(\mathbb{R}^n)$$
.  
2.  $\mu \| \boldsymbol{x} - \boldsymbol{y} \|_2^2 \leq \langle f'(\boldsymbol{x}) - f'(\boldsymbol{y}), \boldsymbol{x} - \boldsymbol{y} \rangle$ ,  $\forall \boldsymbol{x}, \boldsymbol{y} \in \mathbb{R}^n$ .  
3.  $f(\alpha \boldsymbol{x} + (1 - \alpha)\boldsymbol{y}) + \alpha(1 - \alpha)\frac{\mu}{2} \| \boldsymbol{x} - \boldsymbol{y} \|_2^2 \leq \alpha f(\boldsymbol{x}) + (1 - \alpha)f(\boldsymbol{y})$ ,  $\forall \boldsymbol{x}, \boldsymbol{y} \in \mathbb{R}^n$ ,  $\forall \alpha \in [0, 1]$ .  
*Proof:*  
Left for exercise.

**Theorem 6.18** If  $f \in \mathcal{S}^1_{\mu}(\mathbb{R}^n)$ , we have

1. 
$$f(\boldsymbol{y}) \leq f(\boldsymbol{x}) + \langle f'(\boldsymbol{x}), \boldsymbol{y} - \boldsymbol{x} \rangle + \frac{1}{2\mu} \| f'(\boldsymbol{x}) - f'(\boldsymbol{y}) \|_2^2, \ \forall \boldsymbol{x}, \boldsymbol{y} \in \mathbb{R}^n,$$