

4.  $\{\alpha_k\}_{k=-1}^\infty$  is an arbitrary sequence such that  $\alpha_{-1} = 0$ ,  $\alpha_k \in (0, 1]$  ( $k = 0, 1, \dots$ ), and  $\sum_{k=0}^\infty \alpha_k = \infty$ .

Then the pair of sequences  $\left\{ \prod_{i=-1}^{k-1} (1 - \alpha_i) \right\}_{k=0}^\infty$  and  $\{\phi_k(\mathbf{x})\}_{k=0}^\infty$  recursively defined as

$$\phi_{k+1}(\mathbf{x}) = (1 - \alpha_k)\phi_k(\mathbf{x}) + \alpha_k \left[ f(\mathbf{y}_k) + \langle f'(\mathbf{y}_k), \mathbf{x} - \mathbf{y}_k \rangle + \frac{\mu}{2} \|\mathbf{x} - \mathbf{y}_k\|_2^2 \right]$$

is an estimate sequence.

*Proof:*

Let us prove by induction on  $k$ . For  $k = 0$ ,  $\phi_0(\mathbf{x}) = (1 - (1 - \alpha_{-1}))f(\mathbf{x}) + (1 - \alpha_{-1})\phi_0(\mathbf{x})$  since  $\alpha_{-1} = 0$ . Suppose that the induction hypothesis is valid for any index equal or smaller than  $k$ . Since  $f \in \mathcal{S}_\mu^1(\mathbb{R}^n)$ ,

$$\begin{aligned} \phi_{k+1}(\mathbf{x}) &= (1 - \alpha_k)\phi_k(\mathbf{x}) + \alpha_k \left[ f(\mathbf{y}_k) + \langle f'(\mathbf{y}_k), \mathbf{x} - \mathbf{y}_k \rangle + \frac{\mu}{2} \|\mathbf{x} - \mathbf{y}_k\|_2^2 \right] \\ &\leq (1 - \alpha_k)\phi_k(\mathbf{x}) + \alpha_k f(\mathbf{x}) \\ &= \left( 1 - (1 - \alpha_k) \prod_{i=-1}^{k-1} (1 - \alpha_i) \right) f(\mathbf{x}) + (1 - \alpha_k) \left( \phi_k(\mathbf{x}) - \left( 1 - \prod_{i=-1}^{k-1} (1 - \alpha_i) \right) f(\mathbf{x}) \right) \\ &\leq \left( 1 - (1 - \alpha_k) \prod_{i=-1}^{k-1} (1 - \alpha_i) \right) f(\mathbf{x}) + (1 - \alpha_k) \prod_{i=-1}^{k-1} (1 - \alpha_i) \phi_0(\mathbf{x}) \\ &= \left( 1 - \prod_{i=-1}^k (1 - \alpha_i) \right) f(\mathbf{x}) + \prod_{i=-1}^k (1 - \alpha_i) \phi_0(\mathbf{x}). \end{aligned}$$

The remaining part is left for exercise. ■

**Lemma 9.4** Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  be an arbitrary continuously differentiable function. Also let  $\phi_0^* \in \mathbb{R}$ ,  $\mu \geq 0$ ,  $\gamma_0 \geq 0$ ,  $\mathbf{v}_0 \in \mathbb{R}^n$ ,  $\{\mathbf{y}_k\}_{k=0}^\infty$ , and  $\{\alpha_k\}_{k=0}^\infty$  given arbitrarily sequences such that  $\alpha_{-1} = 0$ ,  $\alpha_k \in (0, 1]$  ( $k = 0, 1, \dots$ ). In the special case of  $\mu = 0$ , we further assume that  $\gamma_0 > 0$  and  $\alpha_k < 1$  ( $k = 0, 1, \dots$ ). Let  $\phi_0(\mathbf{x}) = \phi_0^* + \frac{\gamma_0}{2} \|\mathbf{x} - \mathbf{v}_0\|_2^2$ . If we define recursively  $\phi_{k+1}(\mathbf{x})$  such as the previous lemma:

$$\phi_{k+1}(\mathbf{x}) = (1 - \alpha_k)\phi_k(\mathbf{x}) + \alpha_k \left[ f(\mathbf{y}_k) + \langle f'(\mathbf{y}_k), \mathbf{x} - \mathbf{y}_k \rangle + \frac{\mu}{2} \|\mathbf{x} - \mathbf{y}_k\|_2^2 \right],$$

then  $\phi_{k+1}(\mathbf{x})$  preserve the canonical form

$$\phi_{k+1}(\mathbf{x}) = \phi_{k+1}^* + \frac{\gamma_{k+1}}{2} \|\mathbf{x} - \mathbf{v}_{k+1}\|_2^2 \quad (12)$$

for

$$\begin{aligned} \gamma_{k+1} &= (1 - \alpha_k)\gamma_k + \alpha_k\mu, \\ \mathbf{v}_{k+1} &= \frac{1}{\gamma_{k+1}} [(1 - \alpha_k)\gamma_k \mathbf{v}_k + \alpha_k\mu \mathbf{y}_k - \alpha_k f'(\mathbf{y}_k)], \\ \phi_{k+1}^* &= (1 - \alpha_k)\phi_k^* + \alpha_k f(\mathbf{y}_k) - \frac{\alpha_k^2}{2\gamma_{k+1}} \|f'(\mathbf{y}_k)\|_2^2 \\ &\quad + \frac{\alpha_k(1 - \alpha_k)\gamma_k}{\gamma_{k+1}} \left( \frac{\mu}{2} \|\mathbf{y}_k - \mathbf{v}_k\|_2^2 + \langle f'(\mathbf{y}_k), \mathbf{v}_k - \mathbf{y}_k \rangle \right). \end{aligned}$$

*Proof:*

We will use again the induction hypothesis in  $k$ . Note that  $\phi_0''(\mathbf{x}) = \gamma_0 \mathbf{I}$ . Now, for any  $k \geq 0$ ,

$$\phi_{k+1}''(\mathbf{x}) = (1 - \alpha_k)\phi_k''(\mathbf{x}) + \alpha_k\mu\mathbf{I} = ((1 - \alpha_k)\gamma_k + \alpha_k\mu)\mathbf{I} = \gamma_{k+1}\mathbf{I}.$$

Therefore,  $\phi_{k+1}(\mathbf{x})$  is a quadratic function of the form (12). Also,  $\gamma_{k+1} > 0$  since  $\mu > 0$  and  $\alpha_k > 0$  ( $k = 0, 1, \dots$ ); or if  $\mu = 0$ , we assumed that  $\gamma_0 > 0$  and  $\alpha_k \in (0, 1)$  ( $k = 0, 1, \dots$ ).

From the first-order optimality condition

$$\begin{aligned}\phi_{k+1}'(\mathbf{x}) &= (1 - \alpha_k)\phi_k'(\mathbf{x}) + \alpha_k f'(\mathbf{y}_k) + \alpha_k\mu(\mathbf{x} - \mathbf{y}_k) \\ &= (1 - \alpha_k)\gamma_k(\mathbf{x} - \mathbf{v}_k) + \alpha_k f'(\mathbf{y}_k) + \alpha_k\mu(\mathbf{x} - \mathbf{y}_k) = 0.\end{aligned}$$

Thus,

$$\mathbf{x} = \mathbf{v}_{k+1} = \frac{1}{\gamma_{k+1}} [(1 - \alpha_k)\gamma_k\mathbf{v}_k + \alpha_k\mu\mathbf{y}_k - \alpha_k f'(\mathbf{y}_k)]$$

is the minimal optimal solution of  $\phi_{k+1}(\mathbf{x})$ .

Finally, from what we proved so far and from the definition

$$\begin{aligned}\phi_{k+1}(\mathbf{y}_k) &= \phi_{k+1}^* + \frac{\gamma_{k+1}}{2}\|\mathbf{y}_k - \mathbf{v}_{k+1}\|_2^2 \\ &= (1 - \alpha_k)\phi_k(\mathbf{y}_k) + \alpha_k f(\mathbf{y}_k) \\ &= (1 - \alpha_k)\left(\phi_k^* + \frac{\gamma_k}{2}\|\mathbf{y}_k - \mathbf{v}_k\|_2^2\right) + \alpha_k f(\mathbf{y}_k).\end{aligned}\tag{13}$$

Now,

$$\mathbf{v}_{k+1} - \mathbf{y}_k = \frac{1}{\gamma_{k+1}} [(1 - \alpha_k)\gamma_k(\mathbf{v}_k - \mathbf{y}_k) - \alpha_k f'(\mathbf{y}_k)].$$

Therefore,

$$\begin{aligned}\frac{\gamma_{k+1}}{2}\|\mathbf{v}_{k+1} - \mathbf{y}_k\|_2^2 &= \frac{1}{2\gamma_{k+1}} [(1 - \alpha_k)^2\gamma_k^2\|\mathbf{v}_k - \mathbf{y}_k\|_2^2 + \alpha_k^2\|f'(\mathbf{y}_k)\|_2^2 \\ &\quad - 2\alpha_k(1 - \alpha_k)\gamma_k\langle f'(\mathbf{y}_k), \mathbf{v}_k - \mathbf{y}_k \rangle].\end{aligned}\tag{14}$$

Substituting (14) into (13), we obtain the expression for  $\phi_{k+1}^*$ . ■

**Theorem 9.5** Let  $L \geq \mu \geq 0$ . Consider  $f \in \mathcal{S}_{\mu, L}^{1,1}(\mathbb{R}^n)$ , possible with  $\mu = 0$  (which means that  $f \in \mathcal{F}_L^{1,1}(\mathbb{R}^n)$ ). For given  $\mathbf{x}_0, \mathbf{v}_0 \in \mathbb{R}^n$ , let us choose  $\phi_0^* = f(\mathbf{x}_0)$ . Consider also  $\gamma_0 > 0$  such that  $L \geq \gamma_0 \geq \mu \geq 0$ . Define the sequences  $\{\alpha_k\}_{k=-1}^\infty$ ,  $\{\gamma_k\}_{k=0}^\infty$ ,  $\{\mathbf{y}_k\}_{k=0}^\infty$ ,  $\{\mathbf{x}_k\}_{k=0}^\infty$ ,  $\{\mathbf{v}_k\}_{k=0}^\infty$ ,  $\{\phi_k^*\}_{k=0}^\infty$ , and  $\{\phi_k(\mathbf{x})\}_{k=0}^\infty$  as follows:

$$\begin{aligned}\alpha_{-1} &= 0, \\ \alpha_k \in (0, 1] \quad &\text{root of} \quad L\alpha_k^2 = (1 - \alpha_k)\gamma_k + \alpha_k\mu := \gamma_{k+1}, \\ \mathbf{y}_k &= \frac{\alpha_k\gamma_k\mathbf{v}_k + \gamma_{k+1}\mathbf{x}_k}{\gamma_k + \alpha_k\mu}, \\ \mathbf{x}_k \quad &\text{is such that} \quad f(\mathbf{x}_{k+1}) \leq f(\mathbf{y}_k) - \frac{1}{2L}\|f'(\mathbf{y}_k)\|_2^2, \\ \mathbf{v}_{k+1} &= \frac{1}{\gamma_{k+1}} [(1 - \alpha_k)\gamma_k\mathbf{v}_k + \alpha_k\mu\mathbf{y}_k - \alpha_k f'(\mathbf{y}_k)], \\ \phi_{k+1}^* &= (1 - \alpha_k)\phi_k^* + \alpha_k f(\mathbf{y}_k) - \frac{\alpha_k^2}{2\gamma_{k+1}}\|f'(\mathbf{y}_k)\|_2^2 \\ &\quad + \frac{\alpha_k(1 - \alpha_k)\gamma_k}{\gamma_{k+1}} \left( \frac{\mu}{2}\|\mathbf{y}_k - \mathbf{v}_k\|_2^2 + \langle f'(\mathbf{y}_k), \mathbf{v}_k - \mathbf{y}_k \rangle \right), \\ \phi_{k+1}(\mathbf{x}) &= \phi_{k+1}^* + \frac{\gamma_{k+1}}{2}\|\mathbf{x} - \mathbf{v}_{k+1}\|_2^2.\end{aligned}$$

Then, we satisfy all the conditions of Lemma 9.2 for the  $\lambda_k = \prod_{i=-1}^{k-1} (1 - \alpha_i)$ .

*Proof:*

In fact, due to Lemmas 9.3 and 9.4, it just remains to show that  $\alpha_k \in (0, 1]$  for  $(k = 0, 1, \dots)$  such that  $\sum_{k=0}^{\infty} \alpha_k = \infty$ . In the special case of  $\mu = 0$ ,  $\alpha_k < 1$  ( $k = 0, 1, \dots$ ). And also that  $f(\mathbf{x}_k) \leq \phi_k^*$ .

Let us show both using induction hypothesis.

Consider the quadratic equation in  $\alpha_0$ ,  $q_0(\alpha_0) := L\alpha_0^2 + (\gamma_0 - \mu)\alpha_0 - \gamma_0 = 0$ . Notice that its discriminant  $\Delta := (\gamma_0 - \mu)^2 + 4\gamma_0 L$  is always positive by the hypothesis. Also,  $q_0(0) = -\gamma_0 < 0$ , but due to the hypothesis again. Therefore, this equation always has a root  $\alpha_0 > 0$ . Since  $q_0(1) = L - \mu \geq 0$ ,  $\alpha_0 \leq 1$ , and we have  $\alpha_0 \in (0, 1]$ . If  $\mu = 0$ , and  $\alpha_0 = 1$ , we will have  $L = 0$  which implies  $\gamma_0 = 0$  which contradicts our hypothesis. Then  $\alpha_0 < 1$ . In addition,  $\gamma_1 := (1 - \alpha_0)\gamma_0 + \alpha_0\mu > 0$  and  $\gamma_0 + \alpha_0\mu > 0$ . The same arguments are valid for any  $k$ . Therefore,  $\alpha_k \in (0, 1]$ , and  $\alpha_k < 1$  ( $k = 0, 1, \dots$ ) if  $\mu = 0$ .

Finally,  $L\alpha_k^2 = (1 - \alpha_k)\gamma_k + \alpha_k\mu \geq (1 - \alpha_k)\mu + \alpha_k\mu = \mu$ . And we have  $\alpha_k \geq \sqrt{\frac{\mu}{L}}$ , and therefore,  $\sum_{k=0}^{\infty} \alpha_k = \infty$ , if  $\mu > 0$ . For the case  $\mu = 0$ , the argument is the same as the proof of Theorem 9.6.

For  $k = 0$ ,  $f(\mathbf{x}_0) \leq \phi_0^*$ . Suppose that the induction hypothesis is valid for any index equal or smaller than  $k$ . Due to the previous lemma,

$$\begin{aligned} \phi_{k+1}^* &= (1 - \alpha_k)\phi_k^* + \alpha_k f(\mathbf{y}_k) - \frac{\alpha_k^2}{2\gamma_{k+1}} \|f'(\mathbf{y}_k)\|_2^2 \\ &\quad + \frac{\alpha_k(1 - \alpha_k)\gamma_k}{\gamma_{k+1}} \left( \frac{\mu}{2} \|\mathbf{y}_k - \mathbf{v}_k\|_2^2 + \langle f'(\mathbf{y}_k), \mathbf{v}_k - \mathbf{y}_k \rangle \right) \\ &\geq (1 - \alpha_k)f(\mathbf{x}_k) + \alpha_k f(\mathbf{y}_k) - \frac{\alpha_k^2}{2\gamma_{k+1}} \|f'(\mathbf{y}_k)\|_2^2 \\ &\quad + \frac{\alpha_k(1 - \alpha_k)\gamma_k}{\gamma_{k+1}} \left( \frac{\mu}{2} \|\mathbf{y}_k - \mathbf{v}_k\|_2^2 + \langle f'(\mathbf{y}_k), \mathbf{v}_k - \mathbf{y}_k \rangle \right). \end{aligned}$$

Now, since  $f(\mathbf{x})$  is convex,  $f(\mathbf{x}_k) \geq f(\mathbf{y}_k) + \langle f'(\mathbf{y}_k), \mathbf{x}_k - \mathbf{y}_k \rangle$ , and we have:

$$\phi_{k+1}^* \geq f(\mathbf{y}_k) - \frac{\alpha_k^2}{2\gamma_{k+1}} \|f'(\mathbf{y}_k)\|_2^2 + (1 - \alpha_k) \langle f'(\mathbf{y}_k), \frac{\alpha_k \gamma_k}{\gamma_{k+1}} (\mathbf{v}_k - \mathbf{y}_k) + \mathbf{x}_k - \mathbf{y}_k \rangle + \frac{\alpha_k(1 - \alpha_k)\gamma_k \mu}{2\gamma_{k+1}} \|\mathbf{y}_k - \mathbf{v}_k\|_2^2.$$

Recall that since  $f'$  is  $L$ -Lipschitz continuous, if we apply Lemma 3.4 to  $\mathbf{y}_k$  and  $\mathbf{x}_{k+1} = \mathbf{y}_k - \frac{1}{L} f'(\mathbf{y}_k)$ , we obtain

$$f(\mathbf{y}_k) - \frac{1}{2L} \|f'(\mathbf{y}_k)\|_2^2 \geq f(\mathbf{x}_{k+1}).$$

Therefore, if we impose

$$\frac{\alpha_k \gamma_k}{\gamma_{k+1}} (\mathbf{v}_k - \mathbf{y}_k) + \mathbf{x}_k - \mathbf{y}_k = \mathbf{0}$$

it justifies our choice for  $\mathbf{y}_k$ . And putting

$$\frac{\alpha_k^2}{2\gamma_{k+1}} = \frac{1}{2L}$$

it justifies our choice for  $\alpha_k$ . Since  $\frac{\alpha_k(1 - \alpha_k)\gamma_k \mu}{\gamma_{k+1}} \geq 0$ , we finally obtain  $\phi_{k+1}^* \geq f(\mathbf{x}_{k+1})$  as wished. ■

The above theorem suggests an algorithm to minimize  $f \in \mathcal{S}_{\mu, L}^{1,1}(\mathbb{R}^n)$ .

Notice that in the following “optimal” gradient method, the estimated sequence is not necessary anymore.