

$$\begin{aligned}
 0 \leq \|x^{k+1} - x^*\|^2 &\leq \|x^k - x^*\|^2 - 2t_k [f(x^k) - f(x^*)] + t_k^2 \|g^k\|^2 \\
 &\leq \|x^0 - x^*\|^2 - 2 \sum_{i=1}^k t_i [f(x^i) - f(x^*)] + L^2 \sum_{i=1}^k t_i^2 \\
 &\leq D^2 - 2 \sum_{i=1}^k t_i [f(x^i) - f(x^*)] + L^2 \sum_{i=1}^k t_i^2
 \end{aligned}$$

$$\Rightarrow 2 \sum_{i=1}^k t_i [f(x_i) - f(x^*)] \leq D^2 + L^2 \sum_{i=1}^k t_i^2$$

$$\Rightarrow \frac{2 \sum_{i=1}^k t_i [f(x_i) - f(x^*)]}{2 \sum_{i=1}^k t_i} \leq \frac{D^2 + L^2 \sum_{i=1}^k t_i^2}{2 \sum_{i=1}^k t_i}$$

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$$f(\bar{x}) - f(x^*) \leq \frac{D^2 + L^2 \sum_{i=1}^k t_i^2}{2 \sum_{i=1}^k t_i} \rightarrow 0 \Leftrightarrow \begin{cases} \sum t_i \rightarrow \infty \\ \sum t_i^2 < \infty \\ t_{i+1} \leq t_i \end{cases}$$

$$\bar{x} = \frac{\sum_{i=1}^k t_i x_i}{\sum_{i=1}^k t_i}$$

$$\operatorname{argmin}_x f(x_k) + g_k \cdot (x - x_k) + \frac{\|x - x_k\|^2}{2t_k}$$

$$= \operatorname{argmin}_x f(x_k) + g_k \cdot (x - x_k) - \frac{x^T x_k}{t_k} + \frac{x^T x}{2t_k}$$

$$= \operatorname{argmin}_x t_k [f(x_k) + g_k \cdot (x - x_k)] - x^T (x_{k-1} - t_{k-1} g_{k-1}) + \frac{x^T x}{2}$$

$$= \operatorname{argmin}_x \frac{\sum_{i=1}^k t_i [f(x_i) + g_i \cdot (x - x_i)]}{\sum_{i=1}^k t_i} + \frac{\|x\|^2}{2 \sum_{i=1}^k t_i}$$

$$\begin{aligned} \text{prox}_t(x) &= \arg\min_z \frac{1}{2t} \|x-z\|^2 + h(z) \\ &= \arg\min_z \frac{x^T x}{2t} - \frac{x^T z}{t} + \frac{z^T z}{2t} + h(z) \end{aligned}$$

$$\frac{1}{t} = \frac{1}{k}, \quad \frac{x}{t} = \hat{g}_k \Rightarrow x = \frac{k \hat{g}_k}{1/k}$$

$$v^* \min f(x)$$

$$f(x) = \max_g g^T x - f^*(g) \quad [\text{By convexity}]$$

$$v^* = \min_{\|x\| \leq D} f(x) \quad [\text{Assume } D \text{ large enough}]$$

$$= \min_{\|x\| \leq D} \max_g g^T x - f^*(g)$$

$$= \max_g \min_{\|x\| \leq D} g^T x - f^*(g) \quad [\text{Strong Duality}]$$

$$= \max_g -f^*(g) - D \max_{\|z\| \leq 1} -z^T y$$

$$= \max_g -f^*(g) - D \|g\|_*$$

$$0 = v^* - v^* = \min_x f(x) - \max_g -f^*(g) - D \|g\|_*$$

$$= \min_{x, g} f(x) + f^*(g) + D \|g\|_*$$