

Proof.

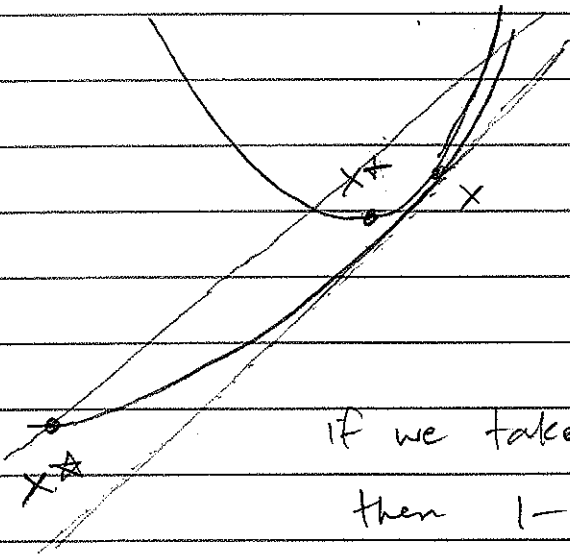
∇f Lipschitz w/ constant L

$$\Leftrightarrow \nabla^2 f(x) \leq LI$$

\Leftrightarrow

$$f(y) \leq f(x) + \nabla f(x)^T (y-x) + \frac{L}{2} \|y-x\|_2^2$$

for all y, x .



$$y = x^+ = x - t \nabla f(x)$$

$$\begin{aligned} f(x^+) &\leq f(x) - \nabla f(x)^T t \nabla f(x) \\ &\quad + \frac{L}{2} t^2 \|\nabla f(x)\|_2^2 \\ &= f(x) - \left(1 - \frac{Lt}{2}\right) \cdot t \|\nabla f(x)\|_2^2 \end{aligned}$$

if we take $t \leq \frac{1}{L}$

then $1 - \frac{Lt}{2} \geq \frac{1}{2}$, so $f(x^+) \leq f(x) - \frac{1}{2} t \|\nabla f(x)\|_2^2$

$$f(x^*) \geq f(x) + \nabla f(x)^T (x^* - x)$$

$$f(x) \leq f(x^*) + \nabla f(x)^T (x - x^*) \leftarrow \text{plug in}$$

$$f(x^+) \leq f(x^*) + \nabla f(x)^T (x - x^*) - \frac{1}{2} \|\nabla f(x)\|_2^2$$

$$f(x^+) - f(x^*) \leq \frac{1}{2t} \left(\|x - x^*\|_2^2 - \underbrace{\|x - t \nabla f(x) - x^*\|_2^2}_{x^+} \right)$$

$$\leq \frac{1}{2t} \left(\|x - x^*\|_2^2 - \|x^+ - x^*\|_2^2 \right)$$

sum this up over $i=1, \dots, k$

$$\begin{aligned} \sum_{i=1}^k (f(x^{(i)}) - f(x^*)) &\leq \frac{1}{2t} \|x^{(0)} - x^*\|_2^2 - \frac{1}{2t} \cancel{\|x^{(k)} - x^*\|_2^2} \\ &\leq \frac{1}{2t} \|x^{(0)} - x^*\|_2^2 \end{aligned}$$

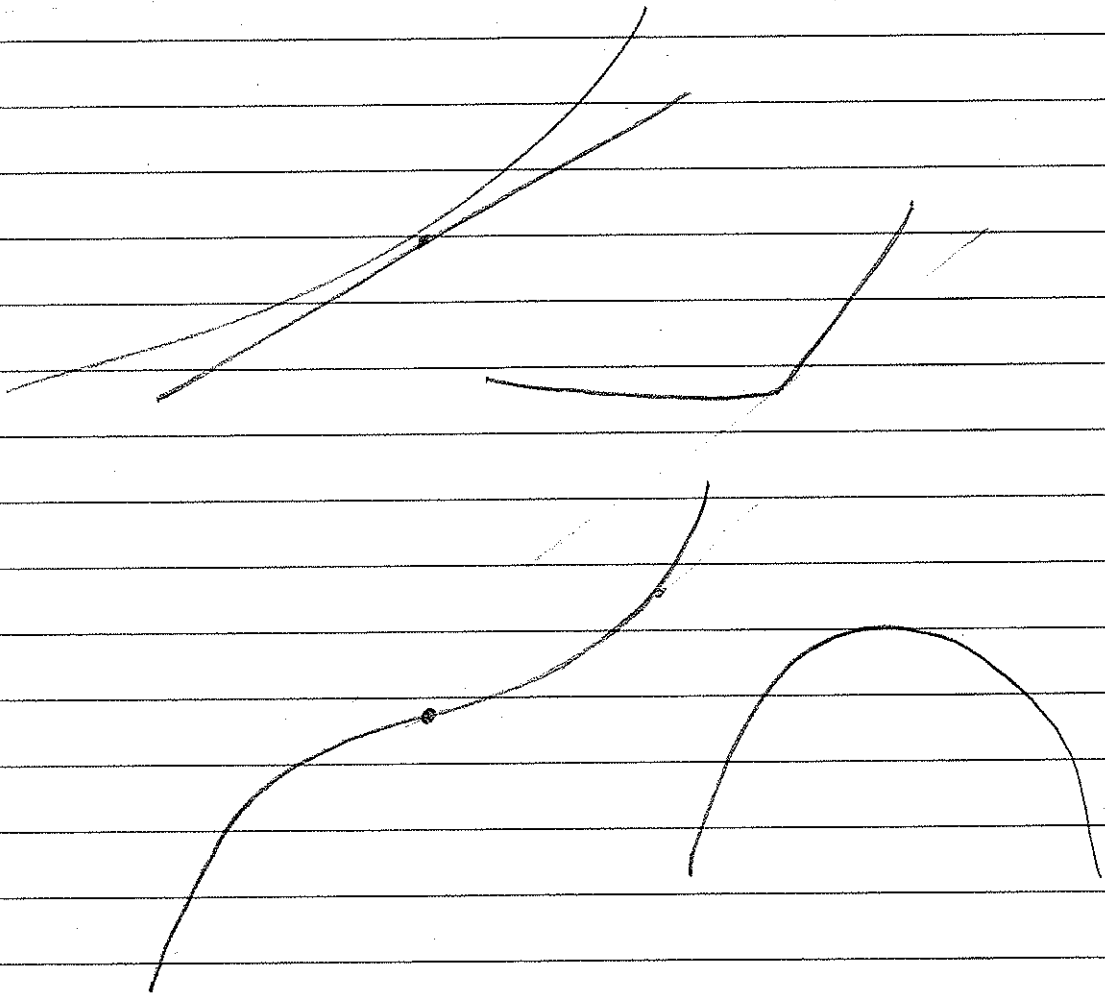
$$f(x^{(k)}) - f(x^*) \leq \frac{1}{k} \sum_{i=1}^k (f(x^{(i)}) - f(x^*))$$

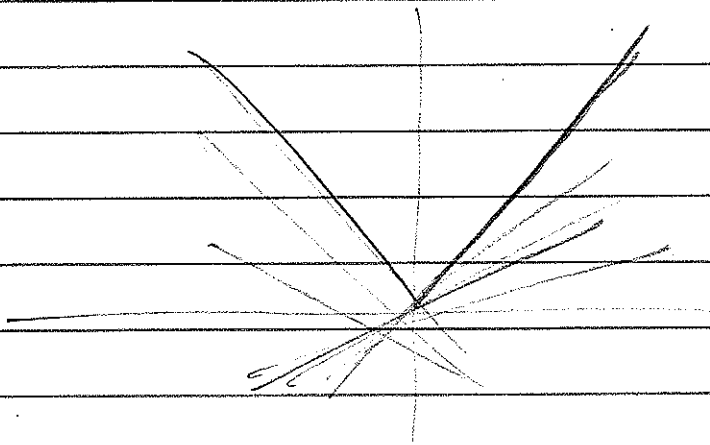
$$\leq \frac{1}{2tk} \|x^{(0)} - x^*\|_2^2 \quad \square$$

$$\varepsilon \text{ st } f(x^{(k)}) - f(x^*) \leq \varepsilon$$

$$e^k = \varepsilon, \quad k = O(\log(1/\varepsilon)) \quad \text{strong convexity}$$

$$\text{vs } O(1/\varepsilon) \quad \text{w/o}$$





$$\nabla f(x) = \frac{x}{\|x\|_2}$$

$$f(y) = \|y\|_2 \geq 0 + g^T y \quad \text{for all } y.$$

g any vector st. $\|g\|_2 \leq 1$

$$\|x\|_1 = \sum_{j=1}^n |x_j|$$

$$g_i = \begin{cases} \text{sign}(x_i) & \text{if } x_i \neq 0 \\ \in [-1, 1] & \text{if } x_i = 0. \end{cases}$$

$$\partial f(x)$$

$$g_1, g_2 \in \partial f(x)$$

$$\alpha g_1 + (1-\alpha) g_2 = g$$

$$\alpha \cdot f(y) \geq \left(f(x) + g_1^T (y-x) \right) \alpha$$

$$(1-\alpha) f(y) \geq \left(f(x) + g_2^T (y-x) \right) (1-\alpha)$$

$$f(y) \geq f(x) + g^T (y-x)$$

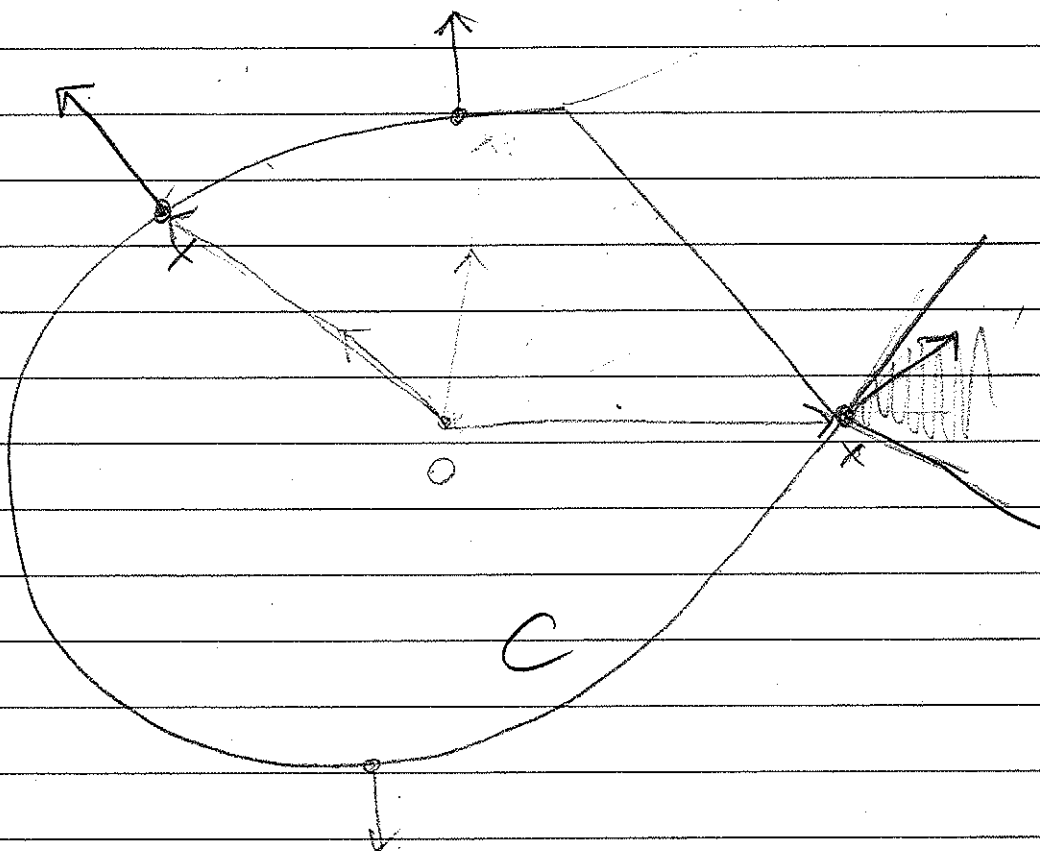
$C \subseteq \mathbb{R}^n$ convex

$$I_C(x) = \begin{cases} 0 & x \in C \\ \infty & x \notin C \end{cases}$$

if $x \in C$

$$\begin{aligned} \partial I_C(x) &= N_C(x) \\ &= \{g : g^T x \geq g^T y \text{ if } y \in C\} \end{aligned}$$

$$I_C(y) \geq I_C(x) + g^T(y-x)$$



$$0 \in \partial f(x^*) \iff x^* \text{ minimizes } f$$

$$\min_{x \in C} f(x) = \min_x (f(x) + I_C(x))$$

$$0 \in \partial f(x^*) + \partial I_C(x^*) \iff$$