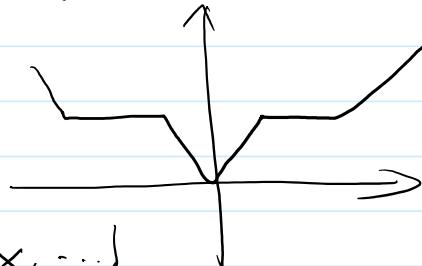


Lecture 15 Non-convex Optimization I Local Analysis

Sunday, October 23, 2016 10:59 PM

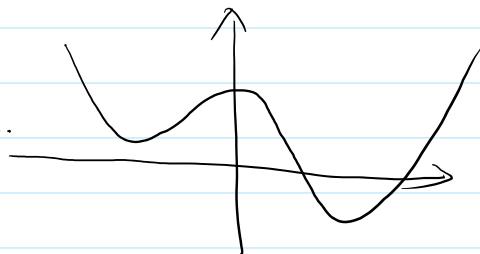
- Non-convex optimization
- what can a non-convex function look like?

- simpler case
still has a unique minimum.



(quasi-convex, pseudo-convex, ...)

- complicated case
multiple local optima..



- optimality conditions
- first order optimality condition

$$\nabla f(x) = 0$$

- such points are called critical points.
- for (strongly) convex function, $\nabla f(x) = 0 \Rightarrow x$ is optimal.

- second order condition

$$\nabla^2 f(x) \succeq 0$$

e.g. $f(x) = x_1^2 + x_2^2$

$$\nabla^2 f(x) = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \succeq 0.$$

- saddle points

$\nabla^2 f(x)$ is not positive semidefinite or negative semidefinite.

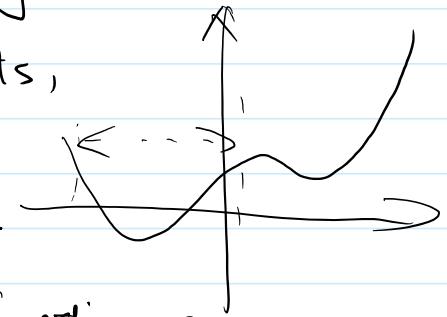
$$\text{e.g. } f(x) = x_1^2 - x_2^2$$

$$\nabla^2 f(x) = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

- Local convergence vs global convergence.

- when multiple local minima exists,

can hope to converge to global minimum with good initialization.



- local structure of a non-convex function can behave like a convex function!

- Approximate Gradient Descent.

- idea: after a good initialization, maybe the function is very similar to convex.

- how to measure "similarity" to convex?

- Consider gradient descent, initial \bar{z}^0 , goal \bar{z}^*

\bar{z}^* is the optimum for convex function $f(\bar{z})$

however, only has non-convex function $g(\bar{z})$

hope: $g(\bar{z})$ close to $f(\bar{z})$

$$\bar{z}^{t+1} = \bar{z}^t - \eta \nabla g(\bar{z}^t)$$

- Def: g is $(\alpha, \beta, \varepsilon)$ -correlated if

$$\langle \nabla g(\bar{z}^t), \bar{z}^t - \bar{z}^* \rangle \geq \alpha \|\bar{z}^t - \bar{z}^*\|^2 + \beta \|\nabla g(\bar{z}^t)\|^2 - \varepsilon$$

Note: if f is μ -strongly convex and L -smooth

$$\langle \nabla f(\bar{z}^t), \bar{z}^t - \bar{z}^* \rangle \geq \frac{\mu L}{\mu+1} \|\bar{z}^t - \bar{z}^*\|^2 + \frac{1}{\mu+1} \|\nabla g(\bar{z}^t)\|^2$$

$$\langle \nabla f(z^+), z^+ - z^* \rangle \geq \frac{\mu L}{\mu + L} \|z^+ - z^*\|^2 + \frac{1}{\mu + L} \|\nabla g(z^+)\|^2$$

$\left(\frac{\mu L}{\mu + L}, \frac{1}{\mu + L}, 0 \right)$ -correlated.

Theorem: $\|z^{t+1} - z^*\|^2 \leq (1 - 2\alpha\eta) \|z^t - z^*\|^2 + 2\eta\varepsilon$,
 (when $\eta \leq 2\beta$), in particular

$$\|z^t - z^*\|^2 \leq (1 - 2\alpha\eta)^t \|z^0 - z^*\|^2 + \frac{\varepsilon}{2}$$

$$\begin{aligned} \text{Proof: } \|z^{t+1} - z^*\|^2 &= \|z^t - z^*\|^2 - 2\eta \langle \nabla g(z^t), z^t - z^* \rangle + \eta^2 \|\nabla g(z^t)\|^2 \\ &= \|z^t - z^*\|^2 - \eta (2 \langle \nabla g(z^t), z^t - z^* \rangle - \eta \|\nabla g(z^t)\|^2) \\ &\leq \|z^t - z^*\|^2 - \eta (2\alpha \|z^t - z^*\|^2 + (2\beta - \eta) \|\nabla g(z^t)\|^2) \geq \varepsilon \\ &\leq (1 - 2\alpha\eta) \|z^t - z^*\|^2 + 2\eta\varepsilon \end{aligned}$$