

Problem:

$$\min_{x \in \mathcal{X}} f(x).$$

$\mathcal{X} \subseteq \text{dom}(f) \subseteq \mathbb{R}^n$. a closed set.

1 A Taxonomy of Solutions to (P)

Will use "solution" and "minimizer" interchangeably.

Definition 1. We say that $x^* \in \mathcal{X} \subseteq \text{dom}(f)$ is

1. a *local minimizer/solution* of (P) if there exists a neighborhood \mathcal{N}_{x^*} of x^* such that for all $x \in \mathcal{N}_{x^*} \cap \mathcal{X}$ we have $f(x) \geq f(x^*)$;
2. a *global minimizer* of (P) if $\forall x \in \mathcal{X}: f(x) \geq f(x^*)$
3. a *strict local minimizer* of (P) if there exists a neighborhood \mathcal{N}_{x^*} of x^* such that for all $x \in \mathcal{N}_{x^*} \cap \mathcal{X}$ and $x \neq x^*$ we have $f(x) > f(x^*)$; (i.e., satisfies part 1 with a strict inequality)
4. an *isolated local minimizer* of (P) if there exists a neighborhood \mathcal{N}_{x^*} such that $\forall x \in \mathcal{N}_{x^*} \cap \mathcal{X}: f(x) \geq f(x^*)$ and \mathcal{N}_{x^*} does not contain any other local minimizer.
5. a *unique minimizer* if it is the only global minimizer.

Ex. prove isolated local min is strict.

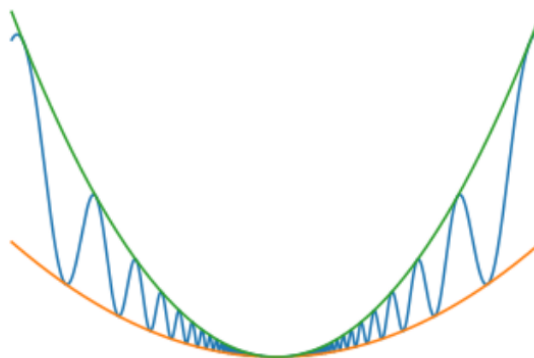
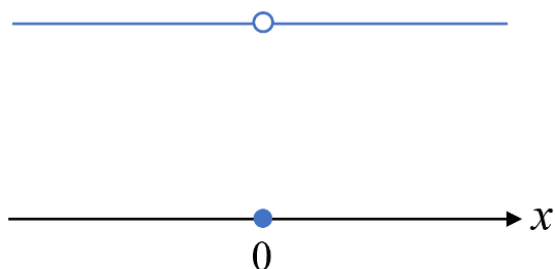
pf: Suppose for contra, strict inequality does not hold. Then $\exists x^{*'} \neq x^*$ s.t. $f(x^{*'}) = f(x^*)$. And $x^{*'} \in \mathcal{N}_{x^*} \cap \mathcal{X}$. This contradicts with x^* is an isolated min.

Strict min ~~is~~ isolated.

Example 3. A strict minimizer that is not isolated:

- (not continuous) $f_1(x) = \begin{cases} 1 & x \neq 0 \\ 0 & x = 0 \end{cases}$ and $x^* = 0$.
- (continuous) $f_2(x) = \begin{cases} x^2 (1 + \sin^2(\frac{1}{x})) & x \neq 0 \\ 0 & x = 0 \end{cases}$ and $x^* = 0$.

Illustration: Left f_1 . Right: f_2 .



Determine a pt is local / global min \Rightarrow Taylor & Thm.

Theorem 1 (Taylor's Theorem; Thm 2.1 in Wright-Recht). Let $f : \mathbb{R}^d \rightarrow \mathbb{R}$ be a continuously differentiable function. Then, for all $x, y \in \text{dom}(f)$ such that $\{(1-\alpha)x + \alpha y : \alpha \in (0,1)\} \subseteq \text{dom}(f)$, we have

1. $f(y) = f(x) + \int_0^1 \langle \nabla f(x + t(y-x)), y-x \rangle dt$
2. $f(y) = f(x) + \langle \nabla f(x + \gamma(y-x)), y-x \rangle$ for some $\gamma \in (0,1)$ (a.k.a. Mean Value Thm).

If f is twice continuously differentiable:

3. $\nabla f(y) = \nabla f(x) + \int_0^1 \nabla^2 f(x + t(y-x)) (y-x) dt$. Here

$$\nabla^2 f(x) = \begin{bmatrix} \dots \\ \frac{\partial^2 f}{\partial x_i \partial x_j}(x) \\ \dots \end{bmatrix} \in \mathbb{R}^{d \times d}$$

denotes the Hessian matrix ("second-order derivative") of f at x .

\downarrow MVT.

4. $\exists \gamma \in (0,1)$:

$$\begin{aligned} f(y) &= f(x) + \langle \nabla f(x), y-x \rangle + \frac{1}{2} \langle \nabla^2 f(x + \gamma(y-x)) (y-x), y-x \rangle \\ &= f(x) + \langle \nabla f(x), y-x \rangle + \frac{1}{2} (y-x)^T \nabla^2 f(x + \gamma(y-x)) (y-x). \end{aligned}$$

Remark 1. A common mistake is to write down the following "Mean-Value Thm" for the gradient:

$$\exists \gamma \in (0,1) : \nabla f(y) = \nabla f(x) + \nabla^2 f(x + \gamma(y-x)) (y-x)? \leftarrow \text{This is wrong!}$$

4. Illustrate:

Start from. 2nd Taylor expansion:

$$f(y) = f(x) + \langle \nabla f(x), y-x \rangle + \int_0^1 (1-t) \langle \nabla^2 f(x + t(y-x)) (y-x), y-x \rangle dt$$

Integral MVT: $\int_a^b f(x) g(x) dx = f(c) \int_a^b g(x) dx$ ($g(x) \geq 0 \Rightarrow$ std MVT)

$$\begin{aligned} \exists \gamma \in (0,1) &= f(x) + \langle \nabla f(x), y-x \rangle + \underbrace{\langle \nabla^2 f(x + \gamma(y-x)) (y-x), y-x \rangle \int_0^1 (1-t) dt}_{\frac{1}{2}} \\ &= f(x) + \langle \nabla f(x), y-x \rangle + \frac{1}{2} (y-x)^T \nabla^2 f(x + \gamma(y-x)) (y-x). \end{aligned}$$

For f continuously differentiable at x , we have

$$f(x+p) = f(x) + \nabla f(x)^T p + o(\|p\|)$$

or equivalently,

$$\lim_{\|P\| \rightarrow 0} \frac{|f(x+p) - f(x) - \nabla f(x)^T P|}{\|P\|} = 0.$$

pf: By part 2 of Taylor Thm above,

$$f(x+p) \stackrel{\exists \lambda \in (0,1)}{=} f(x) + \langle \nabla f(x+\lambda p), P \rangle$$

$$= f(x) + \nabla f(x)^T P + (\nabla f(x+\lambda p) - \nabla f(x))^T P$$

$$\leq f(x) + \nabla f(x)^T P + \|\nabla f(x+\lambda p) - \nabla f(x)\| \|P\|$$

As $\|P\| \rightarrow 0$ $\nabla f(x+\lambda p) \rightarrow \nabla f(x)$. By sandwich we are done. \square .