

Second-Order conditions for Optimality

Useful results:

Theorem (Extended Law of the Mean) $D \subseteq \mathbb{R}^n$ open set,
 $f \in C^2(D, \mathbb{R})$ and $x^* \in D$. Then for all $x \in \mathbb{R}^n$

$$f(x) = f(x^*) + (\nabla f)(x^*)^T (x - x^*) + \frac{1}{2} (x - x^*)^T (Hf)(z) (x - x^*)$$

for some point $z \in [x, x^*] := \{tx + (1-t)x^* : t \in [0, 1]\}$

straight line in \mathbb{R}^n
 connecting x and x^* .

Other reformulations of the above theorem: (Taylor)

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

or

$$f(x) = f(x^*) + (\nabla f)(x^*)^T (x - x^*) + \frac{1}{2} (x - x^*)^T (Hf)(x^*) (x - x^*) + o(\|x - x^*\|^2)$$

Recall o -notation:

Let $g: \mathbb{R} \rightarrow \mathbb{R}$, then

" $g(x) = o(x)$ " $\stackrel{\text{def.}}{\iff}$
 (as $x \rightarrow 0$)

$$\lim_{x \rightarrow 0} \left| \frac{g(x)}{x} \right| = 0. \quad \left(\begin{array}{l} \text{asymptotically,} \\ g \text{ goes to zero} \\ \text{faster than } x \end{array} \right)$$

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 different from
 Computer Science
 usage ($N \rightarrow \infty$)

Theorem (2nd-order NECESSARY condition)

Let x^* be a local minimizer of $f \in C^2(\mathbb{R}^n, \mathbb{R})$, then

$$(\nabla f)(x^*) = 0 \quad \text{and} \quad (Hf)(x^*) \succeq 0.$$

Proof: ↗ HW2, Problem 5

Theorem (2nd-order SUFFICIENT conditions)

Let $f \in C^2(\mathbb{R}^n, \mathbb{R})$ and $(\nabla f)(x^*) = 0$. The point $x^* \in \mathbb{R}^n$ is a

1. global minimizer of f if $\forall x \in \mathbb{R}^n \quad (Hf)(x) \succeq 0$.

2. strict global min. of f if $\forall x \in \mathbb{R}^n \quad (Hf)(x) \succ 0$.

3. strict local min. of f if $(Hf)(x^*) \succ 0$.

Proof: 1. Since $(Hf)(x) \succeq 0$ for all $x \in \mathbb{R}^n$, we will also have
 $(Hf)(z) \succeq 0$ in the Extended Law of the Mean:

$$f(x) = f(x^*) + \underbrace{(\nabla f)(x^*)^T (x - x^*)}_{=0} + \underbrace{\frac{1}{2} (x - x^*)^T (Hf)(z) (x - x^*)}_{\geq 0}$$

$$\Rightarrow \forall x \in \mathbb{R}^n \quad f(x) \geq f(x^*).$$

2. Analogous by.

3. Taylor expansion with $(\nabla f)(x^*) = 0$:

$$f(x) = f(x^*) + \frac{1}{2}(x-x^*)^T (Hf)(x^*) (x-x^*) + o(\|x-x^*\|^2).$$

From Problem 6 in Homework 2: $x^T A x \geq \lambda_{\min} \|x\|^2$
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 symmetric and positive semidefinite

$$\Rightarrow f(x) = f(x^*) + \underbrace{\frac{\lambda_{\min}}{2} \|x-x^*\|^2}_{\geq 0 \checkmark} + \underbrace{o(\|x-x^*\|^2)}_{\substack{\text{could be negative} \\ \text{and ruin our plan} \\ \text{of showing } f(x) \geq f(x^*)}}$$

BUT:

$$\frac{o(\|x-x^*\|^2)}{\|x-x^*\|^2} \rightarrow 0 \quad \text{as } x \rightarrow x^* \quad (\|x-x^*\| \rightarrow 0)$$

$$\text{meaning: } \exists \delta > 0 \quad \forall x \in B_\delta(x^*) \quad |o(\|x-x^*\|^2)| \leq \|x-x^*\|^2 \times \underbrace{\frac{\lambda_{\min}}{4}}_{\substack{\text{constant we} \\ \text{introduced} \\ \text{in order to} \\ \text{bound negative} \\ \text{impact of}}}}$$

"worst-case analysis":

$$o(\|x-x^*\|^2) \geq -\frac{\lambda_{\min}}{4} \|x-x^*\|^2$$

$$\Rightarrow f(x) \geq f(x^*) + \underbrace{\frac{\lambda_{\min}}{2} \|x-x^*\|^2 - \frac{\lambda_{\min}}{4} \|x-x^*\|^2}_{= + \frac{\lambda_{\min}}{4} \|x-x^*\|^2 \geq 0 \text{ ("by design")}}$$

$$\Rightarrow f(x) \geq f(x^*) \quad \text{for all } x \in B_\delta(x^*) \quad \square$$

Recap of tests for positive (semi) definiteness:

Theorem: Let $A \in \mathbb{R}^{n \times n}$ be a symmetric matrix.

- $A > 0 \iff x^T A x > 0 \quad \forall x \neq 0 \iff \lambda_i > 0 \quad \forall i=1, \dots, n$
- $A \geq 0 \iff x^T A x \geq 0 \quad \forall x \in \mathbb{R}^n \iff \lambda_i \geq 0 \quad \forall i=1, \dots, n$
- A is indefinite $\iff \exists i, j \quad \lambda_i > 0$ and $\lambda_j < 0$

Next: Sylvester criterion based on "principal minors":

Recall: $\Delta_k \dots$ kth principal minor of a symmetric $A \in \mathbb{R}^{n \times n}$

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix}, \quad \Delta_1 = a_{11}, \quad \Delta_2 = \det \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}, \dots, \Delta_n = \det A$$

Theorem (Sylvester criterion): $A \in \mathbb{R}^{n \times n}$ symmetric, $\Delta_k \dots$ kth principal minor of A . Then

- $A > 0 \iff \Delta_k > 0 \quad \forall k=1, \dots, n$
- $A < 0 \iff (-1)^k \Delta_k > 0 \quad \forall k=1, \dots, n$

Remark: • $\Delta_k \geq 0 \quad \forall k=1, \dots, n$ does not imply $A \geq 0$.

- However, $\Delta_1 > 0, \Delta_2 > 0, \dots, \Delta_{n-1} > 0, \Delta_n = 0$ actually does imply $A \geq 0$.